

# SOUTHERN CALIFORNIA REGIONAL RESOURCE KIT

## METRIC DICTIONARY

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Revised: 1 December 2023, version 2.1

Revised: 1 February 2023, version 2.0

28 December 2022, version 1

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## INTRODUCTION

### WHAT IS THE REGIONAL RESOURCE KIT EFFORT?

Reducing the risk of large, high intensity fire (and other mega-disturbances) through forest treatments has become a management imperative in California. A [Strategy for Shared Stewardship](#) (2018) and the USFS [Wildfire Crisis Implementation Plan](#) (2022) reinforce specific goals for pace and scale of strategic forest treatments over the next decade. Concurrently, the State of California has issued a new [Wildfire and Forest Resilience Action Plan \(2022\)](#), designed to strategically accelerate efforts to restore the health and resilience of California forests through a joint State of California - Forest Service framework to improve and enhance forest stewardship in California. The social incentives and the scientific knowledge to pursue meaningful restoration of forested landscapes in California are firmly established.

High quality geospatial data are an essential ingredient to address restoration/conservation of the broad suite of core socio-ecological values across landscapes, and to drive analytic tools for planning management investments. To support these initiatives an interagency team of scientists from the Forest Service/Pacific Southwest Research Station, California Natural Resources Agency/CALFIRE, and the University of California at Berkeley and University of California at Irvine collaborated on development of a comprehensive set of mapped data layers needed to accomplish large-scale landscape planning and restoration. Landscape level assessment using high quality data developed from ecological modeling techniques, informative analytical approaches and the resulting credible scientific outputs will be fundamental to inform and support large landscape restoration planning and execution.

The data layers included in this kit are meant to assist land managers in assessing their current landscape and plan for treatments to enhance resilience to human and natural disturbances. Thus each layer represents what the interagency team believes are the most relevant and reliable geospatial data available at this time. Each layer has been examined by the team and is supported by published data and/or was developed using standard methods. The methods for developing each layer are documented in the metric dictionary; however, the accuracy of each layer has not been quantified. It is anticipated that all data layers will be updated and refined as methods and source data evolve and improve.

### WHAT THIS DOCUMENT IS AND ITS INTENDED PURPOSE

#### ORGANIZATIONAL STRUCTURE

This document has been organized to reflect the “Framework for Resilience” as set forth by the Tahoe Central Sierra Initiative (Manley et al. 2020, 2022). The framework is comprised of ten “**Pillars**” which support the full array of landscape management objectives that are inherently interdependent. Each pillar represents the desired long-term, landscape-scale outcome to restoring resilience. They include ecological values, such as biodiversity, as well as societal benefits to communities, such as water security. Within each pillar are “**Elements**” which represent the primary processes and core functions of that pillar, such as focal species, water quality, or economic health. Finally, within each element are the individual “**Metrics**” which describe the characteristics of elements in quantitative or qualitative terms. Metrics are used to assess, plan for, measure, and monitor progress toward desired outcomes and greater resilience.

The framework pillars are:

- Fire Dynamics
- Forest and Shrubland Resilience
- Biodiversity Conservation
- Wetland Integrity
- Water Security
- Carbon Sequestration
- Air Quality
- Economic Diversity
- Fire Adapted Communities
- Social & Cultural Well-Being

It is important to understand that while pillars and elements are consistent across the Sierra Nevada, the metrics used by a group may vary from region to region based on ecological and social differences (for example forest types or economy), available data, and the user preferences. It is equally important to recognize that due to the interdependent nature of the framework, some metrics overlap into multiple elements/pillars however have only been addressed a single time within this document.

## METRICS

The metrics are organized under the 10 pillars. There are 108 metrics within the Southern California Regional Resource Kit. The Metrics describe the characteristics of the elements (key characteristics) of each pillar in quantitative or, in a few cases, qualitative terms. Metrics are used to assess current conditions, plan treatments, and monitor for, measure, and monitor progress toward desired outcomes and greater resilience. Metrics are selected to be informative, meaningful, and actionable to meet the needs of management.

The metrics included within this Southern California Regional Resource Kit are divided into three "tiers." Among all these metrics, some are created and relevant statewide. Additional metrics are more suited to issues/conditions within a given region. The "Tiers" for metrics included in each RRK:

- **Tier 1** – metrics that are relevant to two or more Regions and a single, consistent data layer is available and provided; can be clipped to the boundary of the region so values within that region are the only ones included for calculations or regional statistics. Example: [Annual Burn Probability](#)
- **Tier 2** – metrics relevant to a single region or relevant to multiple Regions but data layers differ among Regions because of varied data availability (sources) across Regions. Example: [California gnatcatcher habitat suitability](#)

- **Tier 3** - metrics are those that would be of interest to some land managers for specific applications but not included as a core metric in an RRK. Example: [Distribution of the Quino checkerspot butterfly](#)

RRK will contain all Tier 1 and Tier 2 data together to comprise the kit. Tier 3 data will be pointed to for reference and use, as needed.

Some data layers within this kit contain null values. We point this out here so users of the data will be aware and take whatever measures appropriate as they use and analyze the data. For some raster datasets in the RRK, areas have been masked (blanked) out and have a cell value of NoData (also referred to as null, NaN or missing). We, as producers and users of the data, cannot ignore NoData or fill them with zeros, since zero is often a valid value for some datasets. Removing NoData cells is not an option, a raster is a continuous grid. For users of the data performing further analyses and combining or "stacking" rasters, these NoData cells will mask out all values in that location in the output. To avoid this issue, the user must create values for the cells before combining them (i.e. 999 or any numeric value that is not real and clearly out of the range of the other values). Reasons for masking (blanking) out cells in RRK data include:

- Cells are located in water bodies (e.g., lakes, reservoirs, or large rivers)
- Cells are located in urban areas
- Cells are located in areas used for irrigated agriculture
- Cells contain no information relevant to the dataset (e.g. for a streams data layer, areas outside of streams have NoData)
- Area (cells) subject to fire or other disturbance but the post disturbance condition or value is unknown.

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#### INTENDED PURPOSE:

Through this "metric dictionary," each metric has been defined to help end-users of the data (and for use with any decision support tools) to understand:

- What tier the metric is in (1, 2, or 3)
- Data vintage
- The definition meant by a given metric
- The expected use(s) of the metric
- The resolution of the developed data
- The data sources used to derive the metric
- The method of metric derivation
- The root file names

References have been included to help the reader understand potential methods for deriving metrics. It is our hope this information will help people make better use of all the assembled information and how it can best be used with various decision support tools. This dictionary will be updated periodically, as necessary.

Note that all metric data layers have been masked (i.e. blocked out) for open water (lakes, reservoirs) and a selected few have been masked for the urban and agricultural landscape (see the list of operational layers at the end of this document). This is done to avoid confusion with vegetation values coming from urban areas (e.g. city parks) or agricultural areas (e.g. irrigated farm land).

## AIR QUALITY

The goal of healthier forests is aligned with the goal of having healthier air (Cisneros et al., 2014, Long et al., 2018). Forests with sustainable fuel loads create less emissions overall, and support less rapid fire growth, which reduces emissions per day and decreases the chances that smoke from a wildland fire event will create long duration, intense smoke episodes like those we've seen at regional scales during the past decade. Key to supporting the proactive management of smoke and minimization of impacts is a granular understanding at the project scale of where the fuels are, and what potential emissions might occur under wildfire and/or Rx fire scenarios. Those emissions (e.g., from maps like those produced by F3 below) combined with estimates of daily spread can be used to inform operational or scenario-based dispersion modeling (and would be compatible with California's PFIRS smoke management system), which in turn would help fire and air managers better understand where smoke is likely to go, and help inform the public where and when it's likely to occur at potentially unhealthy concentrations.

Tradeoffs between wildfire and Rx fire smoke production (daily, or in total) could be quantified on a first order basis by summing daily or total emissions from high severity vs moderate severity over the area of the respective fire spread polygons. Note that Rx fire smoke impacts are not only different due to per acre differences in emissions, but because the per day emissions can also differ quite substantially. Those emissions numbers could also inform dispersion modeling scenarios showing the relative differences in smoke impacts between wildfire and prescribed scenarios, or even between different wildfire management scenarios.

**DESIRED OUTCOME:** Emissions from fires are limited to primarily low- and moderate-severity fires in wildland ecosystems. Forests improve air quality by capturing pollutants.

## PARTICULATE MATTER

Particle pollution represents a main component of wildfire smoke and the principal public health threat. Fine particles (also known as PM<sub>2.5</sub>) are particles generally 2.5 µm in diameter or smaller and represent a main pollutant emitted from wildfire smoke. Fine particles from wildfire smoke are of greatest health concern.

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### POTENTIAL TOTAL SMOKE PRODUCTION INDEX

#### Tier: 1

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** This metric is an index of the potential smoke production (represented by particulate matter that is 2.5 microns or less in diameter, or PM<sub>2.5</sub>) that could be emitted for a given 30-meter pixel under fire weather conditions that produce high severity fire effects. By showing spatial variation in potential smoke emissions under standardized fuel moisture conditions, this index is intended to help identify potential emissions hotspots within a region if a high severity wildfire occurs in the future. It may be useful for regional scale planning and/or prioritization.

However, the actual moistures and fire weather conditions under which these fuels may convert to smoke will vary; therefore, the map does not represent actual smoke production (PM<sub>2.5</sub> emissions) during an actual fire event. For data users interested in near-term smoke forecasts that reflect the environmental drivers of emissions, project-specific modeling tools are recommended. For example, the BlueSky Playground (<https://tools.airfire.org/playground>) can tailor model inputs based on the fuel and moisture conditions observed or planned for in the project area of interest.

Potential smoke emissions do not consider the probability of a fire or the transport of smoke to more distant locations; they only reflect what would happen locally if a pixel were to burn.

**Data Resolution:** 30m Raster

**Data Units:** 0 - 1, a unitless number serving as an index; on a per 30-m pixel basis

**Creation Method:** Potential TOTAL smoke production index is the smoke production expected for a given pixel under severe fire weather conditions. It is based on model outputs from the First Order Fire Effects Model (FOFEM) developed by the U.S. Forest Service (Spatial FOFEM: <https://www.firelab.org/project/fofem-fire-effects-model>). Key drivers (and model inputs) for this mapped variation are (1) fuel loads spatially extracted from the Landfire FCCS modeled fuelbeds map (LANDFIRE 2022 Update (LF 2.3.0), [https://www.landfire.gov/lf\\_230.php](https://www.landfire.gov/lf_230.php)), and (2) fuel moistures, which are assigned to approximate the extremely dry conditions under which high severity fire generally occurs. The data are dimensionless and linearly normalized from 0 to 1 based on the statewide maximum value, with 1 being the maximum PM<sub>2.5</sub> emissions per 30-m pixel for the given region. Fuels are taken from LANDFIRE LF2022\_FCCS\_220. Spatial FOFEM was run as implemented in FlamMap 6.2 (<https://www.firelab.org/project/flammap>).

This index is a unitless number (ranging between 0 and 1) on a per 30-meter pixel basis, which is calculated using the following equation:

Potential Total Smoke Production Index =  $S_i / (\text{maximum } S_i \text{ statewide})$

where

$S_i$  = high severity PM<sub>2.5</sub> emissions value for pixel i

Calculated with SpatialFOFEM (First Order Fire Effects Model), embedded in FlamMap 6.2. Fuels are LCP and FCCS 2022 from LANDFIRE (LCP\_LF2022\_FBFM40\_220\_CONUS and LF2022\_FCCS\_220\_CONUS). FOFEM Parameters used for this application are:

Seasonality - (Summer)

Canopy consumption – 39%

Duff moisture – 20%

1 hour fuel moisture – 4%

10-hour fuel moisture – 6%

100-hour fuel moisture – 8%

1000-hour fuel moisture – 8%

**Data Source:** LANDFIRE FCCS ([LANDFIRE Program: Data Products – Fuel – Fuel Characteristic Classification System Fuelbeds](#)) 2022

Rocky Mountain Research Station <https://www.firelab.org/project/fofem-fire-effects-model>

**File Name:** PotentialTotalSmoke\_202209.tif

**Tier: 1**

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** This is an index of how much *less* smoke (as defined by PM<sub>2.5</sub> emissions) would be produced from a given pixel by burning under moderate fire weather conditions rather than the extreme conditions that lead to high-severity smoke production. This serves as a proxy for efforts to minimize smoke emissions by allowing a given area to burn under more desirable conditions (e.g., prescribed burning conditions) vs. how it would burn under extreme conditions. Since identical fuelbeds are used as inputs in the high-severity and low-severity model runs, the index does *not* represent the effects of fuel treatments on subsequent wildfire. Rather, this metric represents the maximum potential difference between emissions under high vs. moderate fire weather conditions. Summing these reductions over large areas would be unrealistic because wildland fire burns with a mix of intensities and severities over landscapes, and does not burn everywhere in California, every year.

Wildland fire is often self-limiting in extent. In other words, wildfires may stop spreading when they reach the boundary of a recent burn. Since prescribed fire and managed wildfire can be selected to burn under moderate fire weather conditions, proactive fire use can shift high-severity-type fire emissions to low-severity-type fire emissions. This metric provides a rough index of the potential fire emissions benefits if a fire is allowed to burn under moderate weather conditions rather than in a wildfire under extreme weather. By showing the spatial variation in this potential benefit, this index is intended to help identify where fire management may have the greatest emissions benefit. It may be useful for regional scale planning and/or prioritization.

It is important to note that not all managed fire will produce an emissions benefit, because wildfire may not have otherwise burned in that location within the lifespan of the managed fire's effects, and the managed fire's footprint may not prevent a subsequent wildfire from burning in the same location. Furthermore, actual weather conditions vary from those used in model inputs. Therefore, the map does not represent actual avoided smoke production (PM<sub>2.5</sub> emissions) during an actual fire event that may occur in the future. For data users interested in near-term smoke forecasts that reflect the environmental drivers of emissions, project-specific modeling tools are recommended. For example, the BlueSky Playground (<https://tools.airfire.org/playground>) can tailor model inputs based on the fuel and moisture conditions observed or planned for in the project area of interest.

Potential smoke emissions do not consider the probability of a fire or the transport of smoke to more distant locations; they only reflect what would happen locally if a pixel were to burn.

**Data Resolution:** 30m Raster

**Data Units:** 0 - 1, a unitless number serving as an index; on a per 30-m pixel basis

**Creation Method:** This index is a unitless number (ranging between 0 and 1) on a per 30-meter pixel basis, which is calculated using the following equation:

$$\text{Potential Avoided Smoke Production Index} = ( D_i \text{ for a given pixel } ) / ( \text{the maximum } D_i \text{ statewide} )$$

where

$D_i$  = the difference in modeled PM<sub>2.5</sub> emissions between high severity and low severity scenarios for pixel  $i$  = (high severity PM<sub>2.5</sub> emissions scenario for pixel  $i$ ) – (low severity PM<sub>2.5</sub> emissions scenario for pixel  $i$ )

“High severity PM<sub>2.5</sub> emissions” were calculated as described for the “POTENTIAL TOTAL SMOKE PRODUCTION INDEX” metric.

Calculated with SpatialFOFEM (First Order Fire Effects Model), embedded in FlamMap 6.2. Fuels are LCP and FCCS 2022 from LANDFIRE (LCP\_LF2022\_FBFM40\_220\_CONUS and LF2022\_FCCS\_220\_CONUS). Conditions for Low severity PM 2.5 emissions were calculated for the following settings (FOFEM parameters used for this application):

Seasonality: Spring

Canopy consumption - 5%

Duff moisture - 75%

1 hour fuel moisture - 14%

10-hour fuel moisture - 16%

100-hour fuel moisture - 18%

1000-hour fuel moisture - 25%

**Data Source:** LANDFIRE FCCS ([LANDFIRE Program: Data Products - Fuel - Fuel Characteristic Classification System Fuelbeds](#)) 2022

Rocky Mountain Research Station <https://www.firelab.org/project/fofem-fire-effects-model>

**File Name:** PotentialAvoidedSmoke\_202209.tif

## BIODIVERSITY CONSERVATION

The southern California landscape provides habitat for over 300 species of native vertebrates and thousands of invertebrate species and plants. Management activities over the last century have impacted most species to varying degrees and some have declined significantly in recent decades. Protecting and enhancing native biodiversity has become a management imperative under both federal and state laws and policy. Native plants and animals provide a wide array of benefits to forests and other habitats in southern California; they help forests recover after a fire, control flooding and soil erosion, cycle nutrients, and are valued by people recreating in forests. Greater species diversity promotes adaptability and helps ecosystems withstand and recover from disturbance, including those caused by climate change. The Biodiversity Conservation pillar focuses on species diversity, critical habitat for focal species and non-native species distribution.

Habitat data to model the likelihood of species presence or absence was derived from the [FVEG WHR data layer](#).

**DESIRED OUTCOME:** The network of native species and ecological communities is sufficiently abundant and distributed across the landscape to support and sustain their full suite of ecological and cultural roles.

## SPECIES DIVERSITY

Species diversity is a function of both the number of different species in the community and their relative abundances. Larger numbers of species and more even abundances of species lead to higher species diversity.

Species diversity can be calculated in a variety of ways to represent the type and magnitude of differences among species, their number, and their abundance.

---

#### WILDLIFE SPECIES RICHNESS

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** Native species richness is estimated based on high suitability reproductive habitat for a given species. Reproductive habitat is used to represent suitability because it is critical for species persistence and for most native species it has the most limited requirements. If a habitat is identified as high for a given species, it is considered suitable (1), and habitat identified as moderate, low or not suitable, it is considered unsuitable (0). Species richness values are used as a relative measure of biodiversity value; as such, areas with lower species richness based on these criteria may still have high biodiversity value, but not as high as areas with higher richness values. The number of native species per spatial unit (30m pixel) presented as simply the total number; this can be useful for assessing change in number/composition over space. These values are specific to the southern California species and footprint for this kit.

**Data Resolution:** 30m Raster

**Data Units:** Number of species

**Creation Method:** Generated using the California Wildlife Habitat Relationships model developed and managed by the California Department of Fish and Wildlife. CWHR habitat values are based on the FVEG vegetation data that has been updated. Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed highly suitable for the reproduction of that species in the California Wildlife Habitat Relationship database.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** wildlife\_species\_richness.tif

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#### THREATENED/ENDANGERED VERTEBRATE SPECIES RICHNESS

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** Native species richness is estimated based on high suitability reproductive habitat for a given species. Reproductive habitat is used to represent suitability because it is critical for species persistence and for most native species it has the most limited requirements. If a habitat is identified as high for a given species, it is considered suitable (1), and habitat identified as moderate, low or not suitable, it is considered unsuitable (0). Species richness values are used as a relative measure of biodiversity value; as such, areas with lower species richness based on these criteria may still have high biodiversity value, but not as high as areas with higher richness values. The total number of federally threatened/endangered native species per spatial unit (30m pixel) can be useful for assessing change in number/composition over space. These values are specific to the southern California species and footprint for this kit.

**Data Resolution:** 30m Raster

**Data Units:** Number of species

**Creation Method:** Generated using the California Wildlife Habitat Relationships model developed and managed by the California Department of Fish and Wildlife. CWHR habitat values are based on the FVEG vegetation data that has been updated. Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed highly suitable for the reproduction of that species in the California Wildlife Habitat Relationship database.

Only species classified in the CWHR database as federally endangered, federally threatened, California endangered, or California threatened have been included in the species richness count for this layer.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** t\_e\_species\_richness.tif

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FOREST RAPTORS SPECIES RICHNESS

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** Native species richness is estimated based on high suitability reproductive habitat for a given species. Reproductive habitat is used to represent suitability because it is critical for species persistence and for most native species it has the most limited requirements. If a habitat is identified as high for a given species, it is considered suitable (1), and habitat identified as moderate, low or not suitable, it is considered unsuitable (0). Species richness values are used as a relative measure of biodiversity value; as such, areas with lower species richness based on these criteria may still have high biodiversity value, but not as high as areas with higher richness values. The total number of federally threatened/endangered native species per spatial unit (30m pixel) can be useful for assessing change in number/composition over space.

**Data Resolution:** 30m Raster

**Data Units:** Number of species

**Creation Method:** Generated using the California Wildlife Habitat Relationships model developed and managed by the California Department of Fish and Wildlife. CWHR habitat values are based on the FVEG vegetation data that has been updated. Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed highly suitable for the reproduction of that species in the California Wildlife Habitat Relationship database.

Only raptor species that are associated with forest habitats have been included in the species richness count for this layer. The raptors included in this layer are Bald Eagle, California Spotted Owl, Cooper'S Hawk, Great-Horned Owl, Merlin, Northern Goshawk, Northern Spotted Owl, Osprey, Peregrine Falcon, Red-Shouldered Hawk, RedTailed Hawk, Screech Owl and Sharp-Shinned Hawk.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** forest\_species\_richness\_202304.tif

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OPEN HABITAT RAPTORS SPECIES RICHNESS

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** Native species richness is estimated based on high suitability reproductive habitat for a given species. Reproductive habitat is used to represent suitability because it is critical for species persistence and for most native species it has the most limited requirements. If a habitat is identified as high for a given species, it is considered suitable (1), and habitat identified as moderate, low or not suitable, it is considered unsuitable (0). Species richness values are used as a relative measure of biodiversity value; as such, areas with lower species richness based on these criteria may still have high biodiversity value, but not as high as areas with higher richness values. The total number of federally threatened/endangered native species per spatial unit (30m pixel) can be useful for assessing change in number/composition over space.

**Data Resolution:** 30m Raster

**Data Units:** Number of species

**Creation Method:** Generated using the California Wildlife Habitat Relationships model developed and managed by the California Department of Fish and Wildlife. CWHR habitat values are based on the FVEG vegetation data that has been updated. Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed highly suitable for the reproduction of that species in the California Wildlife Habitat Relationship database.

Only raptor species that are associated with open habitats have been included in the species richness count for this layer. The raptors included in this layer are American Kestrel, Barn Owl, Burrowing Owl, Ferruginous Hawk, Golden Eagle, Long-Eared Owl, Northern Harrier, Prairie Falcon, Rough-Legged Hawk, Short-Eared Owl, Swainson'S Hawk, Turkey Vulture, White-Tailed Kite, California Condor and Great Grey Owl.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** open\_species\_richness\_202304.tif

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HUMMINGBIRDS SPECIES RICHNESS

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** Native species richness is estimated based on high suitability reproductive habitat for a given species. Reproductive habitat is used to represent suitability because it is critical for species persistence and for most native species it has the most limited requirements. If a habitat is identified as high for a given species, it is considered suitable (1), and habitat identified as moderate, low or not suitable, it is considered unsuitable (0). Species richness values are used as a relative measure of biodiversity value; as such, areas with lower species richness based on these criteria may still have high biodiversity value, but not as high as areas with

higher richness values. The total number of federally threatened/endangered native species per spatial unit (30m pixel) can be useful for assessing change in number/composition over space.

**Data Resolution:** 30m Raster

**Data Units:** Number of species

**Creation Method:** Generated using the California Wildlife Habitat Relationships model developed and managed by the California Department of Fish and Wildlife. CWHR habitat values are based on the FVEG vegetation data that has been updated. Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed highly suitable for the reproduction of that species in the California Wildlife Habitat Relationship database.

Only hummingbird species have been included in the species richness count for this layer.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** hummingbirds\_species\_richness\_202304.tif

#### FOCAL SPECIES

For specified species listed below within the Focal Species element section of the Biodiversity Conservation pillar, the species should be considered as *Species of Interest*. It is important for the readers to understand, the listed species are not exhaustive, may be an Endangered Species Act (ESA) species, or considered Sensitive Species as they pertain to forest planning. These species are identified based on their sensitivity to impacts from restoration thinning, prescribed fire, and wildfire. The two wildlife species are California spotted owl and fisher. Black oak is an important species for wildlife as well as for tribes.

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#### CALIFORNIA SPOTTED OWL HABITAT SUITABILITY

**Tier:** 2

**Data Vintage:** 2023

**Metric Definition and Relevance:** California spotted owl is continuously distributed on the western slope of the Sierra and inhabits elevations ranging from 1,000 to over 7,000 feet, it is a Region 5 Forest Service “Sensitive Species” and a “Management Indicator Species” (representing late seral closed canopy coniferous forest). In November, 2019, the USFWS issued a 12-month finding on a petition to list the California spotted owl under the Endangered Species Act and determined listing to be not warranted at this time (USDI Fish and Wildlife Service 2019). Although the species is declining throughout much of its range and faces continued threats due to wildfire, habitat loss, and competition from barred owls, the USFWS determined that existing regulatory mechanisms are sufficient (USDI Fish and Wildlife Service 2019). This species is also recognized as a California “Species of Special Concern and a Species of Greatest Conservation Need.”

A conservation assessment for California spotted owl was conducted in 2017 (Gutiérrez, Manley, and Stine 2017). This was followed by the development of a conservation strategy to guide habitat management on National Forest System Lands (USDA Forest Service 2019). The conservation strategy for the California spotted owl throughout its range, including southern California, aims to balance the need to conserve essential habitat elements around sites

occupied by California spotted owls, while simultaneously restoring resilient forest conditions at the landscape scale (USDA Forest Service 2019).

The USDA Forest Service designates a 300-acre protected activity center (PAC) around each known nesting area or activity center. PACs are a USFS land allocation designed to protect and maintain high-quality California spotted owl nesting and roosting habitat around active sites. Territorial owls typically defend a geographic area consistently used for nesting, roosting, and foraging, containing essential habitat for survival and reproduction. The USDA Forest Service calls for an area of 1,000 acres in the central Sierra Nevada around core use areas, including the associated protected activity center, with a minimum of 400 acres of suitable habitat.

**Data Resolution:** 30m Raster

**Data Units:** Binary, 0 (Low Suitability), 1 (High Suitability)

**Creation Method:** CWHR classifications are based on a combination of FVEG canopy cover, FVEG size class and vegetation data. The vegetation data includes a variety of tree, shrub, grassland, and water dominated habitats. Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed suitable for the reproduction of that species in the California Wildlife Habitat Relationship database. Habitat that meets the following criteria is considered suitable:

- CWHR size and density of 4D, 4M within CWHR vegetation types of DFR, MHC, MHW, MRI, PPN, RFR, SMC, WFR
- CWHR size and density of 5D, 5M, 6 within CWHR vegetation types of DFR, EPN, JPN, LPN, MHC, MHW, MRI, PPN, RFR, SMC, WFR

CWHR high suitability values have been used to create separate data layers which identify suitable nesting and suitable foraging habitat. These data have been combined to create the identified “suitable habitat” layers.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

Conservation Strategy for the California Spotted Owl in the Sierra Nevada, US Forest Service, 2019

**File Name:** CSO\_suitable\_habitat.tif

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COASTAL CALIFORNIA GNATCATCHER HABITAT SUITABILITY

**Tier:** 2

**Data Vintage:** 2015

**Metric Definition and Relevance:** This habitat model was developed to delineate a sampling frame for regional monitoring of coastal California gnatcatchers (*Polioptila californica californica*) to determine: 1) percent area occupied (PAO) in high and very high suitability habitat across conserved lands and participating military lands in the U.S. range in southern California; 2) changes in PAO over time; and 3) extinction and colonization rates. One specific purpose of the model is to identify areas recovering from disturbance, such as wildfire, that may not currently support coastal sage scrub vegetation used by coastal California gnatcatchers but are otherwise highly suitable. This will enable monitoring gnatcatcher occupancy associated with habitat changes over time.

**Data Resolution:** 150m Raster

**Data Units:** Habitat Similarity Index (HSI) prediction indicating habitat suitability ranging from zero, the least suitable, to one, the most suitable

- Very High Suitability = HSI values of 0.75 to 1.0
- High Suitability = HSI values of 0.5 to 0.74
- Low to Moderate Suitability = HSI values of 0 to 0.49

**Creation Method:** The authors used the “Partitioned Mahalanobis D2” modeling technique to construct alternative models with different combinations of environmental variables. Variables were calculated at each point in the center of a 150 m x 150 m cell in a grid of points across the southern California landscape. Variables reflect various aspects of topography, climate, land use (percent vegetation and urbanization at 150 m and 1 km scales), Normalized Difference Vegetation Index, and modeled California sagebrush (*Artemisia californica*) habitat suitability. Due to spatial unevenness in gnatcatcher location data, southern California was divided into five sampling regions and randomly subsampled 50 locations from each region. We repeated this process 1,000 times using a total of 1,063 spatially precise and non-redundant gnatcatcher locations as a calibration dataset. For every model-partition, we calculated Habitat Similarity Index (HSI) predictions for presence and pseudo absence points ranging from Very High (0.75 - 1.00); High (0.50 - 0.74); Low (0.25 - 0.49); and Very Low (0 - 0.24). Suitable habitat is identified as grid cells with HSI greater than or equal to 0.5. We calculated Area Under the Curve (AUC) values from a Receiver Operating Curve (ROC) to determine how well models distinguish between presence and pseudo-absence points. We selected a best performing calibration model and partition based upon median HSI calibration and validation values and AUC results. This model includes the following variables: average minimum January and maximum July temperatures, annual precipitation, elevation, northness, eastness, slope, topographic heterogeneity (30 m x 30 m neighborhood), percent of urban, coastal sage scrub and chaparral land cover at 150 m scale, and predicted California sagebrush habitat suitability. We mapped HSI predictions for each cell in the 150 m-scale grid across the study area.

Credits: USGS; Kristine Preston; Coastal California Gnatcatcher Habitat Suitability Model for Southern California (2015) - ScienceBase-Catalog

**Data Source:** USGS

Vandergast, A.G., Kus, B.E., Preston, K.L. and Barr, K.R., 2019. Distinguishing recent dispersal from historical genetic connectivity in the coastal California gnatcatcher. *Scientific reports*, 9(1), pp.1-12.

**File Name:** Gnatcatcher\_Habitat.tif

MOUNTAIN LION SUITABLE HABITAT

**Tier:** 2

**Data vintage:** 2023

**Metric Definition and Relevance:** This layer shows highly suitable habitats for the reproduction and feeding of Mountain lion (*Puma concolor*).

**Data Resolution:** 30m Raster

**Data Units:** Binary, 0 (not suitable) and 1 (suitable)

**Creation Method:** CWHR classifications are based on a combination of FVEG canopy cover, FVEG size class and vegetation data. The vegetation data includes a variety of tree, shrub, grassland, and water dominated habitats.

Species are considered present, and habitats considered suitable for each 30m cell for which the canopy cover-size-vegetation combination have been deemed highly suitable for the reproduction or feeding of that species in the California Wildlife Habitat Relationship database.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** Mountain\_lion\_suitable\_habitat.tif

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MOUNTAIN YELLOW-LEGGED FROG POTENTIAL HABITAT

**Tier:** 2

**Data Vintage:** 2018

**Metric Definition and Relevance:** This dataset represents a species habitat distribution map for Southern Mountain Yellow-legged Frog (*Rana muscosa*) within the conterminous United States (CONUS) based on 2001 ground conditions.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents potential habitat

**Creation Method:** This habitat map was created by applying a deductive habitat model to remotely-sensed data layers within the species' known range.

This Gap Analysis Project (GAP) habitat map is a prediction of the spatial distribution of suitable environmental and land cover conditions within the United States for the species. Mapped areas represent places where the environment is suitable for the species to occur (i.e. suitable to support one or more life history requirements for breeding, resting, or foraging), while areas not included in the map are those predicted to be unsuitable for the species. While the actual distributions of many species are likely to be habitat limited, suitable habitat will not always be occupied because of population dynamics and species interactions. Furthermore, these maps correspond to midscale characterizations of landscapes, but individual animals may deem areas to be unsuitable because of presence or absence of fine-scale features and characteristics that are not represented in our models (e.g. snags, vernal pools, shrubby undergrowth). These maps are intended to be used at a 1:100,000 or smaller map scale.

This habitat map is created by applying a deductive habitat model to remotely-sensed data layers within the species' range. The deductive habitat models are built by compiling information on the species' habitat associations and entering it into a relational database. Information is compiled from the best available characterizations of the species' habitat, which included species accounts in books and databases, primary peer-reviewed literature. The literature references for each species are included in the "Species Habitat Model Report" and "Machine Readable Habitat Database Parameters" files attached to each habitat map item in the ScienceBase repository. The compiled habitat information is used by a biologist to determine which of the ecological systems and land use classes represented in the National Gap Analysis Project's (GAP) Land Cover Map Ver. 1.0 the species is associated with.

The maps are generated using a python script that queries the model parameters in the database; reclassifies the GAP Land Cover Ver 1.0 and ancillary data layers within the species' range; and combines the reclassified layers to produce the final 30m resolution habitat map. Map output is, therefore, not only a reflection of the ecological systems that are selected in the habitat model, but also any other constraints in the model that are represented by the ancillary data layers.

Source: - Gap Analysis Project (GAP), 2018, Southern Mountain Yellow-legged Frog (*Rana muscosa*)  
aSMFRx\_CONUS\_2001v1 Habitat Map: U.S. Geological Survey data release, <https://doi.org/10.5066/F70Z71KZ>.

**Data Source:** USGS

**File Name:** mountain\_yellow\_legged\_frog\_potential\_habitat.tif

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#### UNARMORED THREESPINE STICKLEBACK HABITAT

**Tier:** 3

**Data Vintage:** 2021

**Metric Definition and Relevance:** This is a dataset representing the boundaries for the Unarmored threespine stickleback (*Gasterosteus aculeatus williamsoni*) as understood by the United States Fish and Wildlife Service.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents potential habitat

**Creation Method:** Contains the current range polygons as contained in the ECOS database; no generalization was performed and no data was intentionally omitted. Polygons are created by field staff utilizing the ECOS system to select range per office. The office datasets are then dissolved into one species range file and simplified using the Douglas-Peucker algorithm with a 0.0001 tolerance. The polygons were then transformed to a binary 30m raster layer. The polygons were then transformed to a binary 30m raster layer, where 1 represents potential habitat.

**Data Source:** U.S. Fish and Wildlife Service (USFWS), Ecological Services Division

**File Name:** unarmored\_threespine\_stickleback\_current\_range.tif

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#### PENINSULAR BIGHORN SHEEP

**Tier:** 2

**Data Vintage:** 2008

**Metric Definition and Relevance:** These data identify, in general, the areas where the final revised critical habitat for the Peninsular bighorn sheep (*Ovis canadensis nelsoni*) occurs.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents current habitat

**Creation Method:** The authors designate revised critical habitat for the Peninsular bighorn sheep, a distinct population segment (DPS) of desert bighorn sheep (*Ovis canadensis nelsoni*) occupying the Peninsular Ranges of Southern California, under the Endangered Species Act of 1973, as amended (Act). In total, approximately 376,938 acres (ac) (152,542 hectares (ha)) fall within the boundaries of the critical habitat designation. This revised designation of critical habitat for Peninsular bighorn sheep reduces the 2001 designation by approximately 467,959 ac (189,377 ha). The revised critical habitat is located in Riverside, San Diego, and Imperial Counties, California. This revision takes into account comments from the public, including biologists familiar with the Peninsular bighorn sheep, on areas essential to the conservation of the Peninsular bighorn sheep. As a result of these comments, new information received, and a revision of their criteria used to identify critical habitat, they reevaluated the proposed revised critical habitat boundary and determined that some additions and deletions should be made. The critical habitat polygons were then transformed to a binary 30m raster layer, where 1 represents critical habitat.

**Data Source:** U.S. Fish and Wildlife Service, Carlsbad Fish and Wildlife Office.

**File Name:** peninsular\_bighorn\_sheep\_critical\_habitat.tif

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CALIFORNIA RED-LEGGED FROG

**Tier:** 1

**Data Vintage:** 2001

**Metric Definition and Relevance:** This dataset represents a species habitat distribution map for California Red-legged Frog (*Rana draytonii*) within the conterminous United States (CONUS) based on 2001 ground conditions.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents current habitat

**Creation Method:** This Gap Analysis Project (GAP) habitat map is a prediction of the spatial distribution of suitable environmental and land cover conditions within the United States for the species. Mapped areas represent places where the environment is suitable for the species to occur (i.e. suitable to support one or more life history requirements for breeding, resting, or foraging), while areas not included in the map are those predicted to be unsuitable for the species. While the actual distributions of many species are likely to be habitat limited, suitable habitat will not always be occupied because of population dynamics and species interactions. Furthermore, these maps correspond to midscale characterizations of landscapes, but individual animals may deem areas to be unsuitable because of presence or absence of fine-scale features and characteristics that are not represented in our models (e.g. snags, vernal pools, shrubby undergrowth). These maps are intended to be used at a 1:100,000 or smaller map scale.

This habitat map is created using a deductive model to predict areas suitable for occupation within a species range. The deductive habitat models are built by compiling information on the species' habitat associations and entering it into a relational database. Information is compiled from the best available characterizations of the species' habitat, which included species accounts in books and databases, primary peer-reviewed literature. The literature

references for each species are included in the "Species Habitat Model Report" and "Machine Readable Habitat Database Parameters" files attached to each habitat map item in the ScienceBase repository. The compiled habitat information is used by a biologist to determine which of the ecological systems and land use classes represented in the National Gap Analysis Project's (GAP) Land Cover Map Ver. 1.0 the species is associated with.

The maps are generated using a python script that queries the model parameters in the database; reclassifies the GAP Land Cover Ver 1.0 and ancillary data layers within the species' range; and combines the reclassified layers to produce the final 30m resolution habitat map. Map output is, therefore, not only a reflection of the ecological systems that are selected in the habitat model, but also any other constraints in the model that are represented by the ancillary data layers.

Credits: U.S. Geological Survey (USGS) - Gap Analysis Project (GAP), 2018, California Red-legged Frog (*Rana draytonii*) aCRLFx\_CONUS\_2001v1 Habitat Map: U.S. Geological Survey data release, <https://doi.org/10.5066/F7T43RCM>.

**Data Source:** USGS

**File Name:** california\_red\_legged\_frog\_habitat.tif

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JOSHUA TREE

**Tier:** 2

**Data Vintage:** 2003

**Metric Definition and Relevance:** This data set represents the digital range map of Joshua Tree (*Yucca brevifolia*) in western North America.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents current range, 0 not in range

**Creation Method:** Data from multiple sources, including existing digitized maps, tabular data, personal communication, and figures from other publications, were synthesized to create a single digital distribution. Several steps were undertaken in the process of generating the final distribution. Paper map or figure sources were scanned using a flatbed scanner. Scanned images were then georeferenced in ArcMap (ESRI ArcGIS 8.3) using the Georeferencing Toolbar utilities. Once georeferenced, scanned images were then digitized using heads-up digitizing into respective shapefiles using ArcMap (ESRI ArcGIS 8.3). These shapefiles were then used, in conjunction with a digital elevation model (GTOPO30 from <http://edcdaac.usgs.gov/gtopo30/gtopo30.html>), to visually compare, edit, and generate a new distribution. Editing encompassed moving, adding, deleting, and reshaping polygons in the shapefile subjectively based upon visual comparison of all datasets in the GIS. Only polygons of species presence are included in the final version, and all polygons of presence have a CODE equal to 1. Once a draft distribution was completed, the distribution was distributed to regional experts and critically reviewed. Following review, the distribution draft version was iteratively edited and modified to meet the suggestions of the aforementioned reviewers. Subsequent review is possible, and the current version of this data set could be modified again in the near future based on feedback from existing and additional reviewers. The purpose of this digital data set is to

delimit the modern geographic distribution of Joshua Tree (*Yucca brevifolia*) based on existing publicly available data.

The polygons were then transformed to a binary 30m raster layer, where 1 represents Joshua Tree's range.

Credits: Kenneth L. Cole, Keith Pohs, and John A. Cannella. U.S. Geological Survey, Southwest Biological Science Center, Colorado Plateau Field Station

**Data Source:** USGS

**File Name:** joshua\_tree\_range.tif

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HERMES COPPER BUTTERFLY

**Tier:** 2

**Data Vintage:** 2021

**Metric Definition and Relevance:** This is a dataset representing the boundaries for the Hermes copper butterfly (*Lycaena hermes*) as understood by the United States Fish and Wildlife Service. This species is listed as a threatened species by the USFWS.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents current range

**Creation Method:** Contains the current range polygons as contained in the ECOS database; no generalization was performed and no data was intentionally omitted.

**Data Source:** The U.S. Fish and Wildlife Service Region 8

**File Name:** hermes\_copper\_current\_range.tif

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LAGUNA MOUNTAINS SKIPPER

**Tier:** 2

**Data Vintage:** 2006 (critical habitat)

**Metric Definition and Relevance:** The current range represents the boundaries for Laguna Mountains skipper (*Pyrgus ruralis lagunae*) as understood by the United States Fish and Wildlife Service.

The critical data identifies, in general, the areas of Final critical habitat for the species. Critical habitat constitutes areas considered essential for the conservation of a listed species. These areas provide notice to the public and land managers of the importance of the areas to the conservation of this species. Special protections and/or restrictions are possible in areas where Federal funding, permits, licenses, authorizations, or actions occur or are required. This species is listed as an endangered species by the USFWS.

**Data Resolution:** 30m Raster

**Data Units:** integer

**Creation Method:** The current range contains the current range polygons as contained in the ECOS database; no generalization was performed and no data was intentionally omitted.

The critical habitat was initially digitized based on 1:24,000 USGS quadrangle maps, digital orthophoto quarter quadrangle images, species information from various sources, and staff knowledge of the habitat and hydrology in the area, focusing on elevation ranging between 4000 and 6100 feet, where host plant *Horkelia clevelandii* is present, and slope and soils are suitable for species survival.

This layer integrates the critical habitat and the current range of the Laguna Mountains skipper. The critical habitat and current range polygons were transformed to a binary 30m raster layer, where 1 represents the species critical habitat and current range, respectively. In the final layer, overlapping areas of these two layers were assigned a value of '1', areas that are critical habitat but not current range were assigned '2', and areas that are current range but not critical habitat were assigned '3'.

Value	Interpretation
1	critical habitat that is also current range
2	critical habitat that is not current range
3	current range that is not critical habitat

**Data Source:** U. S. Fish and Wildlife Service, Carlsbad Fish and Wildlife Office

**File Name:** laguna\_mountains\_skipper\_current\_range\_and\_critical\_habitat.tif

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QUINO CHECKERSPOT BUTTERFLY

**Tier:** 3

**Data Vintage:** 2009

**Metric Definition and Relevance:** These data identify, in general, the areas of Final critical habitat for *Euphydryas editha quino* (Quino checkerspot butterfly). This species is listed as an endangered species by the USFWS.

Critical habitat constitutes areas considered essential for the conservation of a listed species. These areas provide notice to the public and land managers of the importance of the areas to the conservation of this listed species. Special protections and/or restrictions are possible in areas where Federal funding, permits, licenses, authorizations, or actions occur or are required.

**Data Resolution:** 30m Raster

**Data Units:** Binary layer, 1 represents current habitat

**Creation Method:** The authors considered several quantitative and qualitative criteria in the selection and designation of specific areas or units for Quino checkerspot butterfly critical habitat. These criteria focused on: (1) lands considered to be occupied: lands within recovery unit boundaries and with confirmed recent (since 1986)

Quino checkerspot butterfly locations that are part of identified habitat complexes, (2) lands not known to be occupied that provide landscape connectivity between adjacent occupied habitat complexes, and (3) lands not known to be occupied that contain confirmed historic Quino checkerspot locations and are part of identified habitat complexes, and are contiguous with occupied lands. Critical habitat for the Quino checkerspot butterfly was delineated based on interpretation of multiple data sources available during the preparation of this rule, including confirmed Quino checkerspot butterfly observation and collection records, 2005 one-meter digital aerial photography at 1:24,000 scale (comparable to the scale of a 7.5 minute U.S. Geological Survey topographic quadrangle map), current aerial photography prints, boundaries of approved habitat conservation plans (HCP), and projects authorized for take through section 7 consultations. In defining critical habitat boundaries, the authors made an effort to avoid: (1) developed lands such as towns and agricultural fields, (2) other non-habitat lands that do not contain one or more primary constituent elements of Quino checkerspot butterfly habitat, (3) non-essential habitat lands that may contain Quino checkerspot butterfly habitat features, but are too isolated or fragmented to provide for long-term conservation of populations, and (4) lands covered by an existing, legally operative, incidental take permit under section 10(a)(1)(B) of the Act in which the Quino checkerspot butterfly is a covered species. However, the minimum mapping unit that the authors used to approximate their delineation of critical habitat for the Quino checkerspot butterfly did not allow them to exclude all areas described above. Existing features and structures within the boundaries of the mapped units, such as buildings, paved or improved roads, aqueducts, railroads, airports, other paved areas, lawns, large areas of closed canopy chaparral, portions of agricultural fields, and other urban landscaped areas do not contain the primary constituent elements. Federal actions limited to those areas, therefore, would not trigger a section 7 consultation, unless they affect the species and/or primary constituent elements in adjacent critical habitat. Initial polygons were created using the process described above. Edits based on staff biological input to include essential habitat and eliminate non-essential areas, and to eliminate as much urban and agriculture as visually possible within or adjacent to the target area was completed in this release. Polygon edits were accomplished using heads-up digitizing and ArcInfo coverage feature extraction from various sources. Heads-up digitizing was completed using 1-meter resolution USDA NAIP (National Agriculture Imagery Program) - USDA NAIP 2005) and coverage features were extracted from data sources such as: California Spatial Information Library, CaSIL (roads, county lines, contour lines, and hydrolines), Riverside County GIS (Stevens Kangaroo Rat (SKR) Core Reserve Boundary), San Diego Association of Governments (SANDAG) (Multiple Habitat Planning Areas, major roads, vegetation, ownership, county boundaries), City of Chula Vista (Preserve design), County of San Diego (major amendment areas, pre-approved mitigation areas), U.S. Fish and Wildlife Service (USFWS) (San Diego National Wildlife Refuge, Otay-Sweetwater Unit), California Department of Fish and Game (CDFG) (ownership), and USFWS-Carlsbad Fish and Wildlife Office Proposed Quino Recovery Units Boundaries (See Recovery Data metafile for information on units and creation process).

The critical habitat polygons were then transformed to a binary 30m raster layer, where 1 represents critical habitat.

**Data Source:** U. S. Fish and Wildlife Service, Carlsbad Fish and Wildlife Office

**File Name:** quino\_checkeredspot\_butterfly\_critical\_habitat.tif

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LEAST BELL'S VIREO NESTING HABITAT

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Least Bell's Vireos (LBVI) are a state and federally listed endangered species that nests exclusively in riparian areas in California. Conservation Science and Data Visualization and Aerial Information Systems co-developed a standardized mapping protocol, called VireoVegMap, with the objective of providing a comprehensive map of potential LBVI nesting habitat. Dominant vegetation map units and secondary map attributes were chosen to provide specific information relevant to LBVI ecology, conservation, and management. We then applied this protocol, using 2020 aerial imagery, to create a baseline map of riparian vegetation to support vireo recovery planning and status evaluation.

Mapping of 2020 riparian vegetation establishes a baseline for range-wide LBVI nesting habitat conditions that can be used to:

- Evaluate the range-wide status and distribution of riparian vegetation that LBVI may use for nesting.
- Develop *Arundo donax* (Giant Reed) removal strategies at the scale of HUC8 subbasins that will maintain or expand existing areas of native riparian vegetation nesting habitat. These areas, with follow-up management, will help LBVI reoccupy formerly suitable nesting habitat that has been lost to *Arundo*.
- Understand the extent of the threat to riparian vegetation that is imposed by the emerging stressor of shot-hole borer invasion and develop potential management strategies.
- Evaluate the potential impacts of groundwater management issues on vireo habitat across the species' potential breeding range.
- Evaluate the potential impacts of wastewater management issues on vireo habitat across the species' potential breeding range.
- Develop habitat-based management and restoration strategies.

**Data Resolution:** 30 meters

**Data Units:** Thematic, suitable riparian vegetation types that include at least 15% shrub willow/mulefat cover and/or >15% tree willow cover.

**Creation Method:** This data layer was developed by the Conservation Science Institute. The overall study area consists of twenty-eight USGS Hydrologic Unit Code (HUC) 8 subbasins in Central and Southern Coastal California and Southern California Deserts. This study area is represented by a polygon feature class in the project geodatabase called "Total Study Extent". Within this large study area, a Focused Mapping Area was developed to ensure near-complete coverage of all riparian vegetation within active channels and floodplains of Level 1 stream networks (e.g., streams that terminate in the Pacific Ocean) within each HUC8 subbasin. Similar protocols were developed to cover active channels and floodplains of desert streams that flow eastward from Peninsular Ranges and terminate in playa lakes. Work was performed on the project between 2021 and 2023.

Project data was produced applying heads-up digitizing techniques in an Esri ArcMap environment using 2020 National Agricultural Imagery Program (NAIP) imagery (60-centimeter base; true-color and color infrared). Ancillary imagery and data sources provided context during mapping. Original data are in vector format. Each original map polygon was assessed for 6 attributes:

- Dominant vegetation map unit type (defined as the vegetation type with >50% relative cover, regardless of total percent vegetation).
- Tree Willow Cover (absolute percent cover of all species of willow trees combined). This ensures that all mapped stands include information on this primary tree nesting habitat type for LBVI.

- Shrub Willow/Mulefat Cover (absolute percent cover of all shrub willow species and mulefat, *Baccharis salicifolia*). This ensures that all mapped stands include information on this primary shrub nesting habitat type for LBVI.
- Arundo Cover (absolute percent cover of *Arundo donax*). This ensures that all mapped stands include information on this invasive exotic species, which degrades LBVI habitat and in some cases, replaces it.
- Tamarisk Cover (absolute percent cover of *Tamarix* spp.). This ensures that all mapped stands include information on this invasive exotic species.
- Dieback Index (absolute percent cover of all trees and/or shrubs within a mapped polygon that show signs of dieback). This provides useful information about vegetation conditions and potential drought effects on vireo habitat.

For all but one map unit, the minimum mapping unit (MMU) polygon size was 1 acre. Due to the importance of early detection of *Arundo*, in order to plan for its removal, smaller *Arundo donax* stands were mapped, with polygons as small as a half-acre.

Download link for the original vector data:

<https://drive.google.com/file/d/1mfwlqJq1upEPIQ1f3pVbrmOOyBNb31DV/view>

Download link for the final report describing mapping methods and summarizing results:

<https://nrm.dfg.ca.gov/FileHandler.ashx?DocumentID=215147>

The original vector data were simplified to create a raster layer that represents Least Bell's Vireo nesting habitat for the Southern and Central California Coast Regions by:

- 1 Including only the subset of polygons within these two geographic regions
- 2 Including only the subset of polygons that had >15% shrub willow/mulefat cover and/or >15% tree willow cover (primary characteristics of Least Bell's Vireo nesting habitat).
- 3 Including only the subset of polygons at elevations less than 512 meters (LBVI rarely nest above this)
- 4 Rasterizing this subset of polygons on the VegName field (which represents the dominant classified vegetation type).

Dominant riparian vegetation map units with Least Bell's Vireo Habitat were, from highest to lowest acreage: Willow/Mulefat Shrub, Tree Willow, Cottonwood, Sycamore, Elderberry, Riparian Tamarisk, Riparian Oak, Riparian Alluvial Scrub, Alder, Post-fire disturbance, Post-disturbance (unknown cause), and Exotic Trees.

**Data Source:** Lott, C.A., Reyes, E., A. Glass, and D. Johnson. 2023. A Range Wide Map of Least Bell's Vireo Nesting Vegetation: Mapping Protocol. Conservation Science and Data Visualization; Boise, ID; and Aerial Information Systems, Inc.; Redlands, CA.; 111pp. Available at: <https://nrm.dfg.ca.gov/FileHandler.ashx?DocumentID=215147>

**File Name:** LBVireoNestingHabitat.tif

The ability of communities to adapt to changing ecological, social, and economic conditions. This entails the capability of an ecological system to sustain a community of organisms that retains the pre-settlement species composition, diversity, and functional organization of natural habitats within a region.

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#### HABITAT CONNECTIVITY

#### Tier: 1

**Data Vintage:** last updated 08/21/2019

**Metric Definition and Relevance:** The Terrestrial Connectivity dataset is one of the four key components of the California Department of Fish and Wildlife's (CDFW) Areas of Conservation Emphasis (ACE) suite of terrestrial conservation information. The dataset summarizes the relative ability of a species to move across the landscape between patches of suitable habitat. It shows a compilation of linkages, corridors, and natural landscape blocks identified in statewide and regional connectivity studies. Each hexagon (2.5 mi<sup>2</sup>) is ranked into one of the following categories based on the identification of corridors and linkages in statewide, regional, and species-movement studies:

- **5: Irreplacable and Essential Corridors** – The Nature Conservancy's (TNC) Omniscape model identifies channelized areas and priority species movement corridors. The mapped channelized areas are those areas where surrounding land use and barriers are expected to funnel, or concentrate, animal movement. These areas may represent the last available connection(s) between two areas, making them high priority for conservation.
- **4: Conservation Planning Linkages** – Habitat connectivity linkages are often based on species-specific models and represent the best connections between core natural areas to maintain habitat connectivity. Linkages have more implementation flexibility than irreplaceable and essential corridors; any linkage areas not included in rank 5 are included here.
- **3: Connections with Implementation Flexibility** – Areas identified as having connectivity importance but not identified as channelized areas, species corridors or habitat linkage at this time. Future changes in surrounding land use or regional specific information may alter the connectivity rank. Included in this category are areas mapped in the TNC Omniscape study as 'intensified', core habitat areas, and areas on the periphery of mapped habitat linkages.
- **2: Large Natural Habitat Areas** – Large blocks of natural habitat (> 2000 acres) where connectivity is generally intact. This includes natural landscape blocks from the 2010 CEHC and updated with the 2016 Statewide Intactness dataset. Areas mapped as CEHC NLB and not include in the previous ranks, are included here.
- **1: Limited Connectivity Opportunity** – Areas where land use may limit options for providing connectivity (e.g., agriculture, urban) or no connectivity importance has been identified in models. Includes lakes. Some DOD lands are also in this category because they have been excluded from models due to lack of conservation opportunity, although they may provide important connectivity habitat.

**Data Resolution:** 30m Raster

**Data Units:** Categorical; 5 (listed above)

**Creation Method:** Developed by CDFW, the Terrestrial Connectivity dataset summarizes information on terrestrial connectivity by ACE hexagon (2.5 mi<sup>2</sup>) including the presence of mapped corridors or linkages and the juxtaposition to large, contiguous, natural areas. This dataset was developed to support conservation planning efforts by allowing the user to spatially evaluate the relative contribution of an area to terrestrial connectivity based on the

results of statewide, regional, and other connectivity analyses. This map builds on the 2010 California Essential Habitat Connectivity (CEHC) map, based on guidance given in the 2010 CEHC report. The data are summarized by ACE hexagon.

The ACE Terrestrial Connectivity polygon has been converted to 30m Raster and the connectivity description attribute (HabDesc) is classified into the five connectivity ranks (detailed above).

**Data Source:** California Department of Fish and Wildlife; Terrestrial Connectivity, Areas of Conservation Emphasis (ACE), version 3.1 last updated 08/21/2019

**File Name:** HabitatConnectivity\_2019.tif

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#### PRESENT DAY CONNECTIVITY IN CALIFORNIA (OMNISCAPE)

**Tier:** 1

**Data Vintage:** last updated 01/2023

**Metric Definition and Relevance:** This data represents a wall-to-wall characterization of regional habitat connectivity potential in California for plant and animal species whose movement is inhibited by developed or agricultural land uses.

This model of present-day connectivity assumes there will be more ‘current flow’, representing wildlife movement, coming from and going to areas that are less modified. Wildlife may encounter barriers and land uses that are not conducive to movement en route. They may avoid moving through these areas entirely or these areas will increase their risk of harm. Land use, energy infrastructure, roads, and night lights are some of the factors that affect the ‘resistance’ to movement in this analysis.

Present Day Connectivity is partitioned into 11 classes (and the code used in the data):

- 1) 3 - Land use may restrict movement:
- 2) 4 - Permeable lands that contribute little to regional connectivity
- 3) 19 - Impeded
- 4) 25 - Diffuse - Med
- 5) 29 - Diffuse - High
- 6) 31 - Intensified - Low
- 7) 35 - Intensified - Med
- 8) 39 - Intensified - High
- 9) 41 - Channelized - Low
- 10) 45 - Channelized - Med
- 11) 49 - Channelized - High

Connectivity classes are assembled into categories based on whether an area had more or less flow than would be expected in the absence of barriers. For example, when animal movement is restricted by surrounding land uses, it channelizes into a single movement pathway, or a linkage. These **Intensified** and **Channelized** linkages are areas with more flow and far more flow, respectively, than would be expected in the absence of nearby barriers to movement. **Diffuse** connectivity areas are broadly, permeable areas with as much flow as is expected. Roads and intensive development can cause complete or partial barriers to animal movement, impeding their ability to traverse the landscape. **Impeded** areas are areas where there is less flow than is expected.

The Omniscape output 'current flow' was classified into high, medium and low classes and further categorized by the amount of flow compared to what would be expected in the absence of barriers. The 'Channelized' class has 1.7 times more flow than expected in the absence of barriers and represents the last remaining natural pathway through a modified landscape. The 'Intensified' class has 1.3-1.7 times more flow than expected and represents areas where there are a few remaining natural pathways. The 'Diffuse' class has as much flow as expected and represents lands that have many or unlimited movement options.

**Data Resolution:** 30m Raster

**Data Units:** Categorical; 11 (listed above)

**Creation Method:** The approach uses Omniscape, a modified version of Circuitscape ([www.circuitscape.org/](http://www.circuitscape.org/)) with a moving-window algorithm, to quantify ecological flow (potential connectivity) among all pixels within a 50km radius. Circuitscape treats landscapes as resistive surfaces, where high-quality movement habitat has low resistance and barriers have high resistance. The algorithm incorporates all possible pathways between movement sources and destinations and identifies areas of high flow via low-resistance routes, i.e., routes presenting relatively low movement difficulty because of lower human modification, and thus mortality risk.

**Data Source:** The Nature Conservancy (TNC) Omniscape

[c.k.stanley@tnc.org](mailto:c.k.stanley@tnc.org)

[The Nature Conservancy: A World Where People & Nature Thrive](#)

**File Name:** PresentDayConnectivity\_Omniscape\_202301.tif

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FULL CLIMATE CONNECTIVITY NETWORK

**Tier:** 3

**Data Vintage:** 2021

**Metric Definition and Relevance:** This linkage network is designed to allow for local movements among individual preserves while supporting landscape-scale regional connectivity. Habitat connectivity is the most frequently recommended strategy to support adaptation to climate change, habitat fragmentation, and post-disturbance recolonizations. In southern California, conservation planning efforts have resulted in protected area networks to address widespread habitat fragmentation across the region. These plans are designed to protect biodiversity by establishing networks of core habitats. Connectivity is essential if these networks are to support the long-term goals of protecting biodiversity, particularly as species' ranges are likely to shift in response to climate change.

**Data Resolution:** polygon

**Data Units:** Categorical, 4 - See report for field definitions

**Creation Method:** The Institute of Ecological Monitoring and Management at San Diego State University, funded by California's Wildlife Conservation Board and a State Wildlife Grant from the California Department of Fish and Wildlife, developed a landscape linkage network to support connectivity for preservation of biodiversity across southern California under climate and land-use changes. Linkages depicted in this dataset were developed using an ensemble of approaches. First, species distribution models (SDMs) were developed for five focal species under historic conditions and four future climate scenario projections (warmer, wetter, reduced emissions - CNRM-CM5 RCP 4.5; warmer, wetter, business as usual emissions - CNRM-CM5 RCP 8.5; warmer, drier, reduced emissions - MIROC5 RCP 4.5; warmer, drier, business as usual emissions - MIROC5 RCP 8.5). Suitability surfaces resulting from the SDMs were used to identify core habitat areas to be connected and converted to resistance surfaces for least

cost corridor connectivity modeling under historic conditions and future periods at decadal intervals. Linkage maps for each species were combined into a single, multispecies linkage network based on a prioritization using a fuzzy-logic based modeling toolbox, Environmental Evaluation and Modeling System 2.02 in ArcGIS (Sheehan and Gough, 2016). Prioritization was based on linkage importance under future conditions, feasibility of implementation, connectivity and habitat value, and biological importance derived from metapopulation models. The top linkage segments for each species were selected based on the maximum value for any single species, the average value across all five focal species, and the count of focal species represented. These segments were then supplemented with geodiversity linkages representing valley and narrow valley features on the landscape, which were not already identified in the focal species linkage segments. Linkages in the central portion of the network were expanded based on a multispecies maximum and average from Circuitscape modeling which also relied on the resistance surfaces derived from SDM suitability maps for each species. Finally, the network was manually trimmed in areas of recent development and expanded in several urban areas to match up with urban conservation planning efforts such as the Rim of the Valley Plan, the Emerald Necklace Vision, and the Santa Ana River Parkway. For more information on methods and data products, visit the project page to download the full report at [Climate Resilient Connectivity For The South Coast Ecoregion of California | IEMM \(sdsu.edu\)](https://www.sdsu.edu/~iemm/ClimateResilientConnectivityForTheSouthCoastEcoregionofCalifornia.html)

**Data Source:** San Diego State University CWC Project Team.

Jennings, M.K., E. Conlisk, E. Haueser, and R.L. Lewison. 2019. Climate Resilient Connectivity for the South Coast Ecoregion. Prepared for the California Department of Fish and Wildlife and the California Wildlife Conservation Board.

**File Name:** Full\_climate\_connectivity\_network.shp

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#### FUNCTIONAL GROUP SPECIES RICHNESS

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** Functional groups are sets of species that share life history characteristics that perform particular functions within an ecosystem. The six functional groups are represented and include a range of trophic levels and ecosystem services. A primary consideration in management is to maintain conditions, adapt to changing conditions and transition to alternate but still productive conditions over time. The maintenance of ecosystem services is a primary concern with climate change.

**Data Resolution:** 30m Raster

**Data Units:** Number of species

**Creation Method:** Species list created from CWHR is divided into six functional groups based on literature. The six functional groups include herbivores, predators, insectivores, soil aerators, seed/spore dispersers and cavity nesters/excavators. The diversity of each functional group is first determined by the number of species for which a given location provides high suitability reproductive habitat (as per species richness calculations). Target conditions can be generated based on percentiles of functional group richness across all patches, so that the 90th percentile or higher is considered in target conditions and the 10th percentile or below is considered to be in a fully departed condition.

**Data Source:**

FVEG 2023

California Department of Fish and Wildlife CWHR version 9.0 (CDFW); 2014

**File Name:** cavity\_nesters\_excavators\_species\_richness.tif; herbivores\_species\_richness.tif; insectivores\_species\_richness.tif; predators\_species\_richness.tif; seed\_spore\_dispersers\_species\_richness.tif; soil\_aerators\_species\_richness.tif

## CARBON SEQUESTRATION

Forests play an important role in mitigating climate by sequestering and storing large amounts of carbon. However, forests are at risk of losing carbon because of rates of decay and disturbance, especially with high severity wildfires. Knowing where carbon exists provides a context for where changes in forest conditions will have the greatest impact on carbon storage and sequestration objectives.

**DESIRED OUTCOME:** Carbon sequestration is enhanced in a stable and sustainable manner that yields multiple ecological and social benefits.

Note that all values for carbon have been expressed in Mg C/ha, the international standard for how carbon is measured. If needed, to convert back to the native short tons per acre, divide the Mg/ha by 2.2417023114334.

### CARBON STORAGE

Carbon storage in forest biomass is an essential attribute of stable forest ecosystems and a key link in the global carbon cycle. After carbon dioxide is converted into organic matter by photosynthesis, carbon is stored in forests for a period of time before it is ultimately returned to the atmosphere through respiration and decomposition or disturbance (e.g., fire). A substantial pool of carbon is stored in woody biomass (roots, trunks, branches). Another portion eventually ends up as organic matter in forest floor litter and in soils. Soil carbon does not change very quickly and is difficult to measure directly.

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### TOTAL ABOVEGROUND CARBON

**Tier:** 1

**Data Vintage:** 09/2020

**Metric Definition and Relevance:** Identifying ecosystem carbon is essential to land managers and the Total Aboveground Carbon metric provides an estimate of the amount of existing carbon and its location on California's landscape. The metric also serves to provide context for the other metrics used to quantify carbon sequestration. For example, instability or lack of resilience in forests with low total aboveground carbon would be of less concern than the same degree of instability in a forest that has large total aboveground carbon.

**Data Resolution:** 30m Raster

**Data Units:** Grams dry matter/m<sup>2</sup>

**Creation Method:** The Center for Ecosystem Climate Solutions (CECS) DataEngine model tracks monthly carbon in multiple pools from 1986 to 2021. The carbon components are initialized with eMapR (see [Additional Resources](#)) observations for the early Landsat era; the model then runs freely based on Landsat and other observations. Disturbances and disturbance intensity are tracked annually by Landsat (see other metrics developed by CECS) and used to quantitatively transfer or combust pools. The model allocates and turns over material based on allometry

scaling theory, as adjusted by observational data sets. Aboveground pools (live tree, live shrubs and dead material) are summed for September of 2020.

CECS data that reflect landscape changes resulting from disturbances require 6 to 12 months of Landsat observations **after a given year that included major disturbances (such as a high severity wildfire)** to fully quantify that disturbance. CECS data that reflect disturbance, such as this data layer, are therefore available **through water year 2020 (i.e. through September 2020)**.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** CStocks\_Total\_Above\_202009.tif

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ANNUAL BIOMASS DATA 2001 AND 2021 - ABOVE AND BELOW GROUND, STANDING DEAD, AND LITTER

**Tier:** 2

**Data Vintage:** 2001 and 2021

**Metric Definition and Relevance:** Biomass estimates for shrubland-dominated ecosystems in southern California have, to date, been limited to national or statewide efforts which can underestimate the amount of biomass; are limited to one-time snapshots; or estimate aboveground live biomass only. These data were developed using a consistent, repeatable method to assess four vegetative biomass pools from 2001-2021 for the southern California study area (totaling 6,441,208 ha), defined by the Level IV Ecoregions (Bailey 2016) that intersect with USDA Forest Service lands (Figure 1). Aboveground live biomass estimates were developed first (Schrader-Patton and Underwood 2021), and then belowground, standing dead, and litter biomass pools were calculated using field data in the peer-reviewed literature (Schrader-Patton et al. 2022). Over half (52.3%) of the study area is shrubland, and the method accounts for different amounts of carbon associated with three post-fire shrub regeneration strategies: obligate resprouting, obligate seeding, and facultative seeding. Biomass estimates were also generated for trees and herbs, giving a total of five life form post-fire regeneration strategy types. These data provide an important contribution to the management of shrubland-dominated ecosystems to assess the impacts of wildfire and management activities, such as fuel management and restoration, and for monitoring carbon storage over the long term.

The biomass data are a key input into the online web mapping tool SoCal EcoServe, developed for US Department Of Agriculture Forest Service resource managers to help evaluate and assess the impacts of wildfire on a suite of ecosystem services including carbon storage. The tool is available at <https://manzanita.forestry.oregonstate.edu/ecoservices/> and described in Underwood et al. (2022).

**Data Resolution:** 30 meters

**Data Units:** kg/m<sup>2</sup>

**Creation Method:** Researchers generated spatial estimates of above ground live biomass (AGLBM) for 2000-2021 for the southern California area, illustrated in the figure below. The study area, totaling 6,441,208 ha, is defined by the 42 Level IV Ecoregions (Bailey 2016) that intersect the four southern US Department of Agriculture (USDA) National Forests in southern California; Angeles, Cleveland, Los Padres, and San Bernardino:



The researchers created biomass raster layers (30m spatial resolution) by modeling a set of covariates (Normalized Difference Vegetation Index [NDVI]), precipitation, solar radiation, actual evapotranspiration, aspect, slope, climatic water deficit, elevation, flow accumulation, landscape facets, hydrological recharge and runoff, and soil type) to predict AGLBM using 766 field plots from the USDA Forest Service Forest Inventory and Analysis (FIA); the Landfire Reference Database (LFRDB) plot data; and other research plots. The dates of field data spanned 2001-2012. The NDVI raster data were derived from Landsat TM/ETM+/OLI multispectral satellite data (onboard Landsat 5, 7, and 8, respectively). NDVI data were composited from all available Landsat images for the months of July and August for each year 2001-2021. Annual precipitation data for each water year (October 1 - September 30) 2001-2021 were downloaded from PRISM (<http://www.prism.oregonstate.edu/>). For each field plot, we extracted the raster values for all covariates; NDVI and precipitation data were matched to the year of plot visit. AGLBM was predicted using the set of 17 covariates in a Random Forest [RF] model in R statistical computing software. To create an AGLBM raster surface for each year 2001-2021, NDVI and precipitation raster data specific to each year were integrated into the RF model (see [Schrader-Patton and Underwood 2021](#) for details).

To estimate other shrubland biomass pools (standing dead, litter, and below ground) a multi-step process was employed:

- 1) First, the study area was segmented by community type using the California Wildlife Habitat Relationships (CWHR) data (Mayer and Laudenslayer 1988). The wildland vegetation of the study area (excluding agricultural, urban, water, and barren classes) contains 45 CWHR classes, 14 of which are  $\geq 0.75\%$  of the wildland vegetation in the study area. CWHR classes were divided into 14 classes; shrubland-dominated versus non-shrubland-dominated types (annual grass, oak, conifer, mixed hardwood).

- 2) For the shrubland types the researchers determined the per pixel proportion of biomass by three plant life forms: tree, shrub, and herb. We further subdivided the shrub life form into three post-fire regeneration strategies: Obligate Resprouters (OR), obligate seeders (OS), and facultative seeders (FS), providing five plant types in total. Rasters depicting the proportion of biomass in each of the five plant types were created by first calculating the proportion of biomass in each type for the plots used in Schrader-Patton and Underwood (2021). The plot data contained individual plant species, crown width and height measurements. Using these measurements, the biomass was estimated for each individual plant within the plot by applying published allometric equations (see

[Schrader-Patton and Underwood 2021](#) for details). The individual plants in the plots were classified into the five plant types and the proportion of biomass in each type were calculated for each plot. A multinomial model was used to relate these proportions to a suite of geophysical and remote sensing variables which, in turn, was applied to raster surfaces of these predictors. The final outputs were raster maps of the proportion of biomass by life form (tree, shrub, herb) and, for shrubs, the proportion of biomass by post-fire regeneration strategy (OR, OS, and FS) (Underwood et al. 2023). We used these raster layers to estimate other vegetative pools of biomass (e.g., below-ground shrub biomass using above- to below- ground ratios) for each post-fire regeneration strategy type (OR, OS, and FS) using information found in the published literature.

3) Third, estimates of standing dead, litter, and below ground biomass pools by either applying fractions of AGLBM gleaned the available published literature or by using biomass estimates in existing spatial datasets. The specific method used was dependent on the percentage of the WHR class in the study area and the vegetation type (shrub or non-shrub)

**File naming:**

WWETAC\_UCD\_above\_ground\_2001\_g\_m2\_v22.tif (Aboveground Live Biomass 2001)

WWETAC\_UCD\_above\_ground\_2021\_g\_m2\_v22 (Aboveground Live Biomass 2021)

WWETAC\_UCD\_below\_ground\_2001\_g\_m2\_v22.tif (Belowground Live Biomass 2001)

WWETAC\_UCD\_below\_ground\_2021\_g\_m2\_v22.tif (Belowground Live Biomass 2021)

WWETAC\_UCD\_litter\_2001\_g\_m2\_v22.tif (Litter Biomass 2001)

WWETAC\_UCD\_litter\_2021\_g\_m2\_v22.tif (Litter Biomass 2021)

WWETAC\_UCD\_standing\_dead\_2001\_g\_m2\_v22.tif (Standing Dead Biomass 2001)

WWETAC\_UCD\_standing\_dead\_2021\_g\_m2\_v22.tif (Standing Dead Biomass 2021)

**Data Source:**

Schrader-Patton, C.C., E.C. Underwood, and Q.M. Sorenson. 2023. Annual biomass spatial data for southern California (2001–2021): Above- and belowground, standing dead, and litter. *Ecology* e4031.

Schrader-Patton, C.C. and E.C. Underwood. 2022. Annual biomass data (2001-2021) for southern California: above- and below-ground, standing dead, and litter. Dryad, Dataset, <https://doi.org/10.5061/dryad.qz612jmjt>

Underwood, E.C., Q.M. Sorenson, C.C. Schrader-Patton, N.A. Molinari and H.D. Safford. 2023. Resprouting, seeding, and facultative seeding shrub species in California’s Mediterranean-type climate region. *Frontiers in Ecology and Evolution* 11:1158265. doi: 10.3389/fevo.2023.1158265

Data available in RRK for 2001 and 2021 (year in file name changes accordingly). The full set of data for intervening years can be downloaded from: <https://doi.org/10.5061/dryad.qz612jmjt> .

**File Name:**

WWETAC\_UCD\_above\_ground\_2021\_g\_m2\_v22.tif; WWETAC\_UCD\_below\_ground\_2021\_g\_m2\_v22.tif;

WWETAC\_UCD\_litter\_2021\_g\_m2\_v22.tif; WWETAC\_UCD\_standing\_dead\_2021\_g\_m2\_v22.tif;

WWETAC\_UCD\_above\_ground\_2001\_g\_m2\_v22.tif; WWETAC\_UCD\_below\_ground\_2001\_g\_m2\_v22.tif;  
WWETAC\_UCD\_litter\_2001\_g\_m2\_v22.tif; WWETAC\_UCD\_standing\_dead\_2001\_g\_m2\_v22.tif

## CARBON STABILITY

Carbon stability is an important feature in carbon sequestration calculations because carbon turnover – high levels of loss, even if followed by high rates of sequestration – are not as ecologically beneficial as high residency rates for carbon and larger pool values, particularly when stored in large live trees which have many other ecological benefits. The carbon in dead biomass is considered a more unstable component of the carbon pool itself, and a potential destabilizing factor for the live carbon pool in fire-adapted forest ecosystems, especially where it exceeds certain thresholds (e.g., over 46 Mg (total biomass)/ha, Stephens et al., 2022).

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### ABOVEGROUND CARBON CARBON TURNOVER TIME

**Tier:** 1

**Data Vintage:** 09/2020

**Metric Definition and Relevance:** The average lifetime of aboveground live and dead carbon in years. Locations where the lifetime or turnover time is longer have more carbon in more stable pools, such as large trees or large coarse woody debris. Locations where the lifetime or turnover time is shorter have more carbon in labile pools, such as live or dead leaves.

**Data Resolution:** 30m Raster

**Data Units:** Years

**Creation Method:** Calculated from the ratio of total aboveground carbon and annual decomposition. Aboveground carbon and annual decomposition are both calculated for 2020 from a Landsat-driven pools and fluxes model, as described for the total aboveground carbon product. Aboveground turnover time does not currently account for carbon losses and removals with combustion or harvest.

CECS data that reflect landscape changes resulting from disturbances require 6 to 12 months of Landsat observations **after a given year that included major disturbances (such as a high severity wildfire)** to fully quantify that disturbance. CECS data that reflect disturbance, such as this data layer, are therefore available **through water year 2020 (i.e. through September 2020)**.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** CStocks\_Turnovertime\_202009.tif

## ECONOMIC DIVERSITY

Economic Diversity increases business opportunities that provide regional economic vitality and additional benefits to rural and vulnerable populations. Ecosystem services and forest products provide a foundation for many local and regional economic activities and employment opportunities. Forest management should support a sustainable natural resource-based economy.

**DESIRED OUTCOME:** Forest management and outdoor activities support a sustainable, natural-resource-based economy, particularly in rural communities.

## WOOD PRODUCT INDUSTRY

### COST OF POTENTIAL TREATMENTS

**Tier:** 2

**Data Vintage:** 2023

**Metric Definition and Relevance:** The principle method for maintaining or restoring resilience to the southern California landscape involves vegetation treatments. There are many variations on treatments involving different kinds of equipment and different activities of managing vegetation. The metric has gathered available information on the costs of the major treatment methods and incorporated this information into a geospatial database.

There are no treatments of vegetation in southern California that generate revenue. All treatments included here are represented simply as costs per acre.

**Field definitions:**

Mastication = CALFIRE estimates for treatments per acre (Brush = \$1,669, Herbaceous = \$1,813, Woodland = \$1,198, Forest = \$1,788)

Masticat\_1 = USFS estimates per acre (low end = \$800), depends on amount of vegetation

Masticat\_2 = USFS estimates per acre (high end = \$1700), depends on amount of vegetation

Thinning\_m = CALFIRE estimates for manual thinning per acre (Brush = \$2,534, Herbaceous = \$1,851, Woodland = \$2,683, Forest = \$1,461)

Thinning\_1 = USFS estimates per acre (low end = \$450), depends on amount of vegetation

Thinning\_2 = USFS estimates per acre (high end = \$950), depends on amount of vegetation

Thinning\_3 = CALFIRE estimates mechanical thinning per acre (Brush = \$2,500, Herbaceous = N/A, Woodland = \$2,807, Forest = \$957)

Thinning\_4 = USFS estimates mechanical thinning per acre (low end = \$945), depends on amount of vegetation

Thinning\_5 = USFS estimates mechanical thinning per acre (high end = \$1,800), depends on amount of vegetation

Piling\_man = CALFIRE estimates manual piling per acre (Brush = \$2,551, Herbaceous = N/A, Woodland = N/A, Forest = \$1,071)

Piling\_m\_1 = USFS estimates manual piling per acre (low end = \$400), depends on amount of vegetation

Piling\_m\_2 = USFS estimates manual piling per acre (high end = \$1,200), depends on amount of vegetation

Piling\_mec = CALFIRE estimates mechanical piling per acre (Brush = \$1,521, Herbaceous = N/A, Woodland = \$251, Forest = \$640)

Piling\_m\_3 = USFS estimates mechanical piling per acre (low end = \$800), depends on amount of vegetation

Piling\_m\_4 = USFS estimates mechanical piling per acre (high end = \$1,200), depends on amount of vegetation

LopScatter = CALFIRE estimates lop and scatter per acre (Brush = \$1,263, Herbaceous = N/A, Woodland = \$1,217, Forest = \$1,616)

LopScatt\_1 = USFS estimates lop and scatter per acre N/A

LopScatt\_2 = USFS estimates lop and scatter per acre N/A

Herbicide\_ = CALFIRE estimates herbicide (post-treatment) per acre (Brush = \$675, Herbaceous = \$396, Woodland = \$667, Forest = \$325)

Herbicide1 = USFS estimates herbicide (post-treatment) per acre (low end = \$250), depends on amount of vegetation

Herbicide\_1 = USFS estimates herbicide (post-treatment) per acre (high end = \$450), depends on amount of vegetation

Pileburn\_C = CALFIRE estimates pile burn per acre (Brush = \$2,303, Herbaceous = \$3,125, Woodland = N/A, Forest = \$810)

Pileburn\_U = USFS estimates lop and scatter per acre N/A

Pileburn\_1 = USFS estimates lop and scatter per acre N/A

**Data Resolution:** 30m Raster

**Data Units:** Dollars per acre

**Creation Method:** Multiple land managers in southern California (Forest Service, CALFIRE) were contacted to obtain current estimates of costs of different treatment methods. We received current estimates from both on treatment costs per acre for a variety of treatment methods. Those cost estimates varied by vegetation type and treatment method. These data were linked to the updated FVEG spatial data and rolled up into a single raster with attributes reflecting these two cost variables. These data are subject to further refinement and changes in costs. Data will continue to be gathered to improve these estimates.

**Data Source:**

CALFIRE

USDA Forest Service

**File Name:** cost\_per\_acre\_vegtype.tif

## FIRE ADAPTED COMMUNITIES

Wildfires are a keystone disturbance process in western US forests. However, the capacity for humans to coexist in the wildland urban interface (WUI) requires different restoration strategies aimed at the protection of life and property. This pillar evaluates the degree to which communities are living safely with fire and are accepting of management and natural ecological dynamics. It also evaluates the capacity for communities to manage desired, beneficial fire and suppress unwanted fire.

The definition of WUI used here, from Carlson et al 2022, adopts the definitions of interface and intermix WUI developed for previous census-based WUI mapping efforts based on U.S. Federal Register definitions (Radeloff et al., 2005; USDA & USDI, 2001). According to the definitions used for the building-based maps and for the census-based maps, WUI is where building density exceeds 6.17 units/km<sup>2</sup> and where land cover is either (1) at least 50% wildland vegetation (intermix) or (2) under 50% wildland vegetation but within 2.4 km (1.5 miles) of a patch of wildland vegetation at least 5 km<sup>2</sup> in area that contains at least 75% vegetation (interface). The distance selected for the interface definition is based on research from the California Fire Alliance suggesting that this is the average distance firebrands can travel from an active wildfire front (Stewart et al., 2007).

**DESIRED OUTCOME:** Communities have adapted to live safely in forested landscapes and understand the significance of fire to maintaining healthy forests. They have sufficient capacity to manage desired fire and suppress unwanted fire.

### HAZARD

The fire hazard element characterizes the risk of high and moderate severity fire and threat to infrastructure. This is typically but not exclusively applied to the Wildland Urban Interface (WUI) defense and threat zones.

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### STRUCTURE EXPOSURE SCORE

#### Tier: 1

**Data vintage:** 08/2023. Includes disturbances through the end of 2022

**Metric Definition and Relevance:** This metric combines two data layers; one is the Wildland Urban Interface (WUI) as defined by Carlson et al. 2022 (see [WUI definition](#) in the operational data section for more information), and a second data layer, Structure Exposure Score (SES), developed by Pyrologix LLC. The WUI includes the intermix and interface zones which collectively identify areas where structures occur and/or where structures are within a 1.5 miles wildland vegetation (see definition above) . The distance selected for the interface definition is based on research from the California Fire Alliance suggesting that this is the average distance firebrands can travel from an active wildfire front.

Structure Exposure Score is an integrated rating of wildfire hazard that includes the likelihood of a wildfire reaching a given location along with the potential intensity and ember load when that occurs. SES varies considerably across the landscape.

Pyrologix uses a standard geometric-interval classification to define the ten classes of SES, where each class break is 1.5 times larger than the previous break. So, homes located within Class X are 1.5 times more exposed than those in Class IX, and so on. This metric represents SES for WUI areas only.

1. 1 (SES I): 0

2. 2 (SES II): 0.01 to 50
3. 3 (SES III): 50 to 75
4. 4 (SES IV): 75 to 113
5. 5 (SES V): 113 to 169
6. 6 (SES VI): 169 to 253
7. 7 (SES VII): 253 to 380
8. 8 (SES VIII): 380 to 570
9. 9 (SES IX): 570 to 854
10. 10 (SES X): 854+

**Data Resolution:** 30m Raster

**Data Units:** Relative index, 10 classes

**Creation Method:**

**WUI:**

The current delineation of the WUI (Carlson et al. 2022) uses a mapping algorithm with definitions of the WUI; two classes of WUI were identified:

- the intermix, where there is at least 50% vegetation cover surrounding buildings
- the interface, where buildings are within 2.4 km (1.5 miles) of a patch of vegetation at least 5 km<sup>2</sup> in size that contains at least 75% vegetation.

Both classes required a minimum building density of 6.17 buildings per km<sup>2</sup> (using a range of circular neighborhood sizes).

This is a proprietary index representing the level of wildfire exposure for a structure (e.g., a home) if one were to exist on a given pixel. It is an integrated measure that includes three components: the likelihood of a wildfire of any intensity occurring in a given year (annual burn probability), potential wildfire intensity for a given pixel, and ember load to that pixel from surrounding vegetation.

SES data was produced by Pyrologix LLC, a wildfire threat assessment research firm, as part of a spatial wildfire hazard assessment across all land ownerships for the state of California. The ongoing work generally follows the framework outlined in Scott and Thompson (2013), with custom methods and significant improvements developed by Pyrologix. The project generally consists of three components: fuelscape calibration and updates, wildfire hazard assessment, and risk assessment. It utilizes a combination of wildfire models and custom tools, including the FSim large wildfire simulator (Finney et al., 2011), and WildEST, a custom modeling tool developed by Pyrologix (Scott, 2020). To date, this work has resulted in a wide variety of spatial data layers related to wildfire hazard and risk, including Structure Exposure Score (SES), representing conditions prior to the 2023 fire season. Work to date has been funded by the USDA Forest Service Region 5, the California Energy Commission, and the USDI Bureau of Land Management with data contributions from CAL FIRE.

For this project, the FSim large-fire simulator is used to quantify annual wildfire likelihood across the analysis area. FSim is a comprehensive fire occurrence, growth, behavior, and suppression simulation system that uses locally relevant fuel, weather, topography, and historical fire occurrence information to make a spatially resolved estimate of the contemporary likelihood and intensity of wildfire across the landscape.

WildEST (Wildfire Exposure Simulation Tool) is used to quantify wildfire intensity and ember loads across the analysis area. WildEST is a deterministic wildfire modeling tool developed by Pyrologix that integrates spatially continuous weather input variables, weighted based on how they will likely be realized on the landscape. This makes the deterministic intensity values developed with WildEST more robust for use in effects analysis than the stochastic intensity values developed with FSim. This is especially true in low wildfire occurrence areas where predicted intensity values from FSim are reliant on a very small sample size of potential weather variables. It also allows for more appropriate weighting of high-spread conditions into fire behavior calculations. WildEST also produces indices of conditional and expected ember production from vegetated areas (pixels) and load to other pixels in the analysis area. Please reference the Pyrologix 2021 project report (Volger et al., 2021) for more information on WildEST analysis.

FSim was run for the CAL 2022 fuelscape at 120m resolution. WildEST was run for the CAL 2022 fuelscape at 30-m resolution. Both models utilized gridded hourly historical California weather data provided by CALFIRE. Results for annual burn probability (FSim), fire intensity (WildEST) and ember load (WildEST) were used to create Structure Exposure Score.

The final step was to overlay the 2022 version of SES with the 2022 footprint of the WUI.

**Data Source:**

Pyrologix, LLC

WUI, Carlson et al, 2022

**File Name:** StructureExposureScore\_WUI\_2023.tif

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DAMAGE POTENTIAL

**Tier:** 1

**Data Vintage:** 08/2023. Includes disturbances through the end of 2022

**Metric Definition and Relevance:** This metric combines two data layers; one is the Wildland Urban Interface (WUI) as defined by Carlson et al. 2022 (see [WUI definition](#) in the operational data section for more information), and a second data layer, Damage Potential (DP), developed by Pyrologix LLC. The WUI includes the intermix and interface zones which collectively identify areas where structures occur. The distance selected for the interface definition is based on research from the California Fire Alliance suggesting that this is the average distance firebrands can travel from an active wildfire front.

The composite Damage Potential (DP) dataset represents a relative measure of wildfire’s potential to damage a home or other structure if one were present at a given pixel, and if a wildfire were to occur (conditional exposure). It is a function of ember load to a given pixel, and fire intensity at that pixel, and considers the generalized consequences to a home from fires of a given intensity (flame length). This index does not incorporate a measure of annual wildfire likelihood.

**Data Resolution:** 30m Raster

**Data Units:** Relative index, low to high

**Creation Method:** This metric represents DP for WUI areas only. DP values were binned based on the following ranges into 6 classes and assigned class names.

- 0 (None): Values = 0

- 1 (Very Low): Values 0.01 to 20
- 2 (Low): Values 20 to 35
- 3 (Moderate): Values 35 to 50
- 4 (High): Values 50 to 80
- 5 (Very High): Values 80+

The current delineation of the WUI (Carlson et al. 2022) uses a mapping algorithm with definitions of the WUI; two classes of WUI were identified:

- the intermix, where there is at least 50% vegetation cover surrounding buildings
- the interface, where buildings are within 2.4 km (1.5 miles) of a patch of vegetation at least 5 km<sup>2</sup> in size that contains at least 75% vegetation.

Both classes required a minimum building density of 6.17 buildings per km<sup>2</sup> (using a range of circular neighborhood sizes).

Damage Potential (DP) data was produced by Pyrologix LLC, a wildfire threat assessment research firm, as part of a spatial wildfire hazard assessment across all land ownerships for the state of California. The ongoing work generally follows the framework outlined in Scott and Thompson (2013), with custom methods and significant improvements developed by Pyrologix. The project generally consists of three components: fuelscape calibration and updates, wildfire hazard assessment, and risk assessment. It utilizes a combination of wildfire models and custom tools, including WildEST (Wildfire Exposure Simulation Tool), a custom modeling tool developed by Pyrologix (Scott, 2020). To date, this work has resulted in a wide variety of spatial data layers related to wildfire hazard and risk, including Damage Potential (DP), representing conditions prior to the 2023 fire season. Work to date has been funded by the USDA Forest Service Region 5, the California Energy Commission, and the USDI Bureau of Land Management with data contributions from CAL FIRE. Please reference the Pyrologix 2021 project report (Volger et al., 2021) for more information about the project or WildEST analysis.

Damage Potential (DP) is a proprietary index developed by Pyrologix LLC representing wildfire's potential to damage a home or other structure if a wildfire were to occur (conditional exposure). It is a function of ember load to a given pixel and fire intensity at that pixel, and it considers the generalized consequences to a home from fires of a given intensity (flame length). DP is calculated based on two other datasets developed by Pyrologix: conditional risk to potential structures (cRPS) and conditional ember load index (CELI).

cRPS represents the potential consequences of fire to a home at a given location if a fire occurs there and if a home were located there. It is a measure that integrates wildfire intensity with generalized consequences to a home on every pixel. Wildfire intensity (flame length) is calculated using Pyrologix' WildEST tool. WildEST is a scripted geospatial process used to perform multiple deterministic simulations under a range of weather types (wind speed, wind direction, fuel moisture content). Rather than weighting results solely according to the temporal relative frequencies of the weather scenarios, the WildEST process integrates results by weighting them according to their weather type probabilities (WTP), which appropriately weights high-spread conditions into the calculations. For fire-effects calculations, WildEST generates flame-length probability rasters that incorporate non-heading spread directions, for which fire intensity is considerably lower than at the head of the fire.

The response function characterizing potential consequences to an exposed structure is applied to fire effects flame lengths from WildEST for all burnable fuel types on the landscape regardless of whether an actual structure is present or not. The response function does not consider building materials of structures and is meant as a measure of the effect of fire intensity on structure exposure. The response function is provided below:

- Flame length probability of 0-2 ft: -25

- Flame length probability of 2-4 ft: -40
- Flame length probability of 4-6 ft: -55
- Flame length probability of 6-8 ft: -70
- Flame length probability of 8-12 ft: -85
- Flame length probability of >12 ft: -100

These results were calculated using 30m fire-effects flame-length probabilities from the WildEST wildfire behavior results and then further smoothed.

cELI is also calculated in WildEST, and represents the relative ember load per pixel, given that a fire occurs, based on surface and canopy fuel characteristics, climate, and topography within the pixel. Units are the relative number of embers. cELI is based on heading-only fire behavior.

Damage Potential is then calculated as the arithmetic mean of cELI and cRPS for each pixel across the landscape as follows:

$$DP = cRPS + cELI/2$$

Although flame length and its potential impact to structures is a function of the fire environment at the subject location only, ember load is a function of ember production and transport in the area surrounding the subject location. A location with light fuel (and therefore low flame length) could still have significant Damage Potential if surrounded by a fire environment that produces copious embers.

**Data Source:**

Pyrologix, LLC

WUI, Carlson et al, 2022

**File Name:** DamagePotential\_WUI\_2023.tif

EMBER LOAD INDEX

**Tier:** 1

**Data Vintage:** 08/2023. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** This ember load dataset represents the ember load index (ELI) per pixel, for a given pixel, based on surface and canopy fuel characteristics, climate, and topography within the pixel. The Ember Load Index (ELI) incorporates burn probability (BP). BP is incorporated into calculations of the ember production before the distribution of embers across the landscape to determine ember load. Given that ELI incorporates burn probability, this index can be used to identify where on the landscape hardening buildings may be needed to resist ignition and the priority for doing so according to the likelihood of the area being visited by fire.

**Data Resolution:** 30m Raster

**Data Units:** Relative number of embers.

**Creation Method:** ELI is not simply the multiplication of ember load (ELI) and burn probability (BP). Rather, BP is incorporated into calculations of the ember production prior to the distribution of embers across the landscape to determine ember load. ELI is based on heading-only fire behavior.

**Data Source:** Pyrologix, LLC

**File Name:** EmberLoadIndex\_2023.tif

**Tier: 1**

**Data Vintage:** 08/2023. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** The ember transport model used in WildEST tracks the travel of embers from each source pixel to downwind receiving pixels. The relative number of embers landing on a given receiving pixel is summed across all potential source pixels. If the receiving pixel has a nonzero WRC Building Cover value (meaning the pixel is within 75 m of a qualifying building), then we separately sum the relative number of embers from the source pixel. The final SELB raster represents the expected annual relative ember production that lands on building cover across all weather types.

**Data Resolution:** 30m Raster

**Data Units:** Relative index

**Creation Method:** The WildEST modeling contains a module for producing indices of conditional and expected ember production and load. The Conditional Ember Production Index (cEPI) is an index of the relative number of embers lofted at a given landscape pixel given the fire environment there, given that a fire occurs. Ember Production Index (EPI) is the expected value of cEPI; it is the expected annual relative number of embers lofted from a given landscape pixel.

The Conditional Ember Load Index (cELI) is a relative index of the relative number of embers that land at a given landscape location, including nonburnable pixels. Finally, Ember Load Index combines the conditional ELI and the likelihood of that ember load occurring. All ember characteristics are based on headfire behavior. These

The ember load indices represent relative ember load at a pixel. Similar to ember production, ember load is also based on surface and canopy fuel characteristics, climate, and topography at the pixel. Ember load incorporates downwind ember travel.

The Ember Load Index (ELI) incorporates burn probability; however, ELI is not simply the multiplication of condition ember load (cELI) and burn probability (BP). Rather, BP is incorporated into calculations of the ember production before the distribution of embers across the landscape to determine ember load. Given that ELI incorporates burn probability, this index can be used to identify where on the landscape hardening buildings may be needed to resist ignition and the priority for doing so according to the likelihood of the area being visited by fire.

**Data Source:** Pyrologix, LLC

**File Name:** SourceEmberLoadToBuildings\_2023.tif

**Tier: 1**

**Data Vintage:** 1992 - 2020

**Metric Definition and Relevance:** The original point layer (WildfireOccurrence\_CA\_1992\_2020.shp ) contains a spatial database of wildfires that occurred in the United States from 1992 to 2020. It is the fifth update of a publication originally generated to support the national Fire Program Analysis (FPA) system. The wildfire records were acquired from the reporting systems of federal, state, and local fire organizations. The following core data elements were required for records to be included in this data publication: discovery date, final fire size, and a

point location at least as precise as a Public Land Survey System (PLSS) section (1-square mile grid). The data were transformed to conform, when possible, to the data standards of the National Wildfire Coordinating Group (NWCG), including an updated wildfire-cause standard (approved August 2020). Basic error-checking was performed and redundant records were identified and removed, to the degree possible. The resulting product, referred to as the Fire Program Analysis fire-occurrence database (FPA FOD), includes 2.3 million geo-referenced wildfire records, representing a total of 180 million acres burned during the 29-year period. Identifiers necessary to link the point-based, final-fire-reporting information to published large-fire-perimeter and operational-situation-reporting datasets are included. Short, Karen C. 2022. Spatial wildfire occurrence data for the United States, 1992-2020 [FPA\_FOD\_20221014]. 6th Edition. Fort Collins, CO: Forest Service Research Data Archive. <https://doi.org/10.2737/RDS-2013-0009.6>

**Data Resolution:** Vector(points) (1 layer) and 30m Raster (3 layers)

**Data Units:** Categorical

**Creation Method:** Rocky Mountain Research Station (U.S. Forest Service) scientist, Karen Short, is the principal creator of this data set. Points were converted to 30m raster cells using the “most frequent” function on the NWCG\_CAUSE\_CLASSIFICATION attribute (Broad classification of the reason the fire occurred) creating three rasters:

- Human caused ignition
- Lightning (natural) caused ignition
- All causes of ignition - Human or Natural and Missing data/not specified/undetermined

“MostFrqCau” indicates the most frequent cause of the fire in that location.. “FireCount” indicates the number of fires that occurred between 1992 and 2020, regardless of cause. It is noted that locations with hundreds of counts may be a result of the method of how ignitions are reported/recorded. Both the accuracy and precision of the location estimates are generally much lower than that implied by the stored coordinate information – which, for example, may have been calculated from a PLSS section centroid. Efforts were made to purge redundant records to the best of the authors’ ability. Despite this, some locations may have multiple records that may reflect redundant records or multiple reports of fires due to the imprecision of the location record, the reporting process of an individual authority, or the possible reality of multiple initiations at a given location.

**Data Source:** Rocky Mountain Research Station, U.S. Forest Service

File Name: WldfireAllCausesCount\_1992\_2020.tif; WldFireOccCause\_Human\_1992\_2020.tif;  
WldFireOccCause\_Natural\_1992\_2020.tif; WildfireOccurrence\_CA\_1992\_2020.shp

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FIRE IGNITION PROBABILITY

**Tier:** 3

**Data Vintage:** 1992 to 2015

**Metric Definition and Relevance:** These rasters depict the predicted human- and lightning-caused ignition probability for the state of California. Ignition is regulated by complex interactions among climate, fuel, topography, and humans. Considerable studies have advanced our knowledge on patterns and drivers of total areas burned and fire frequency, but much is less known about wildfire ignition. To better design effective fire prevention and management strategies, it is critical to understand contemporary ignition patterns and predict the probability of

wildfire ignitions from different sources. UC Davis researchers modeled and analyzed human- and lightning-caused ignition probability across the whole state and sub-ecoregions of California, USA.

Findings reinforce the importance of varying humans vs biophysical controls in different fire regimes, highlighting the need for locally optimized land management to reduce ignition probability. Based on the most complete ignition database available, researchers developed maximum entropy models to predict the spatial distribution of long-term human- and lightning-caused ignition probability at 1 km and investigated how a set of biophysical and anthropogenic variables controlled their spatial variation in California and across its sub-ecoregions. Results showed that the integrated models with both biophysical and anthropogenic drivers predicted well the spatial patterns of both human- and lightning-caused ignitions in statewide and sub-ecoregions of California. Model diagnostics of the relative contribution and marginalized response curves showed that precipitation, slope, human settlement, and road network were the most important variables for shaping human-caused ignition probability, while snow water equivalent, lightning density, and fuel amount were the most important variables controlling the spatial patterns of lightning-caused ignition probability. The relative importance of biophysical and anthropogenic predictors differed across various sub-ecoregions of California.

**Data Resolution:** 1km Raster

**Data Units:** Probability, 0-1

**Creation Method:** Maximum entropy models were developed to estimate wildfire ignition probability and understand the complex impacts of anthropogenic and biophysical drivers, based on a historical ignition database.

UC Davis researchers developed maximum entropy models to estimate wildfire ignition probability and understand the complex impacts of anthropogenic and biophysical drivers, based on a historical ignition database. Researchers used the US Forest Service Fire Program Analysis-Fire Occurrence Database (FPA-FOD), compiled from reporting systems of US federal, state, and local fire agencies (Short [2017](#)). This homogenized and comprehensive dataset includes wildfire ignition records on both public and private lands from 1992 to 2015, and accounted for many small fires that are not included in many other fire datasets.

Researchers used spatial layers of population density, transportation road network, and nighttime lights, to quantify human settlement and accessibility. Researchers assembled statewide geospatial layers to evaluate the biophysical controls from topography, climate, and fuels on spatial variation of wildland ignitions (table [1](#)). The 2010 global 250 m terrain elevation data (GMTED2010) was used to characterize slope and aspect at 1 km spatial resolution. Weather information came from the gridded Daily Surface Weather and Climatological Summaries meteorological data at 1 km (Daymet) (Thornton *et al* [2020](#)), including precipitation (Prcp), minimum and maximum temperature (Tmin and Tmax), incident shortwave radiation (Srad), water vapor pressure (VP), and snow water equivalent (SWE), or the amount of water that would be released from melting snowpack. We derived long-term annual means during 1992–2015 for these meteorological variables at 1 km.

Researchers modeled the spatial pattern of ignition probability using the maximum entropy statistical method (MaxEnt v3.3.3k) (Phillips *et al* [2004](#), [2006](#), [2021](#)). MaxEnt is a machine-learning technique originally designed to model species distribution from presence-only data using multidimensional environmental inputs (Phillips *et al* [2004](#), [2006](#)). It estimates a target probability distribution by iteratively searching for the probability distribution with maximum entropy (i.e. the one that is most uniform), subject to the environmental variables at each observation (i.e. presence-only point).

The models captured well the spatial patterns of human and lightning started wildfire ignitions in California. The human-caused ignitions dominated the areas closer to populated regions and along the traffic corridors. Model diagnosis showed that precipitation, slope, human settlement, and road network shaped the statewide spatial distribution of human-started ignitions. In contrast, the lightning-caused ignitions were distributed more remotely in Sierra Nevada and North Interior, with snow water equivalent, lightning strike density, and fuel amount as primary drivers. Separate region-specific model results further revealed the difference in the relative importance of the key drivers among different sub-ecoregions.

**Data Source:** Bin Chen and Yufang Jin, University of California Davis, [bch@ucdavis.edu](mailto:bch@ucdavis.edu)

[Spatial patterns and drivers for wildfire ignitions in California - IOPscience](#)

[Short K C 2017 Spatial wildfire occurrence data for the United States, 1992-2015 \[FPA FOD 20170508\]](#)

**File Name:** PredictedHumanIgnitionProb\_1km.tif; PredictedLightningIgnitionProb\_1km.tif

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#### WILDFIRE HAZARD POTENTIAL

**Tier:** 1

**Data Vintage:** 08/2022. Includes disturbances through the end of 2021.

**Metric Definition and Relevance:** Wildfire Hazard Potential (WHP) is an index that quantifies the relative potential for wildfire that may be difficult to control. WHP can be used as a measure to help prioritize where fuel treatments may be needed.

**Data Resolution:** 30m Raster

**Data Units:** Relative index

**Creation Method:** Pyrologix calculated WHP following the methods established by Dillon et al. (2015) and Dillon (2018). The original methods utilize lower-resolution FSim inputs, while our approach uses higher-resolution inputs including 30-m CAL vegetation inputs (derived from LANDFIRE 2016), 30-m CAL fuel model outputs, 30-m CAL burn probability results, and 30-m CAL fire-effects flame-length probabilities from the WildEST wildfire behavior results.

**Data Source:** Pyrologix, LLC

**File Name:** WildfireHazardPotential\_2022.tif

## FIRE DYNAMICS

Fire dynamics reflect fire as an ecological process and the function that it performs. It can be broken into two key elements: functional fire and fire severity. Although fire dynamics pertain to the entire landscape, the ecological role of fire is most relevant to landscapes outside of the wildland urban interface (WUI). Within the WUI, protection of life and property takes priority over the role of fire as a process. As a result, this fire dynamics pillar pertains to areas outside of the WUI while the fire-adapted communities pillar pertains to areas inside the WUI.

**DESIRED OUTCOME:** Fire burns in an ecologically beneficial and socially acceptable way that perpetuates landscape heterogeneity and rarely threatens human safety or infrastructure.

#### FUNCTIONAL FIRE

Increasing the pace and scale of restoration on the landscape will require using a variety of tools to accomplish restoration targets. The use of prescribed fire and managed wildfires, where appropriate, can contribute to the restoration need. This is particularly true where fires burn at low and moderate severity, which we are referring to as “functional fire”. Functional fire is when fire burns in an ecologically beneficial and socially acceptable way, perpetuating landscape heterogeneity and rarely threatening human safety or infrastructure.

### ***Discussion of the Fire Return Interval Departure (FRID) Methods***

**Definition and Relevance:** The fire return interval departure (FRID) analysis quantifies the difference between current and pre-settlement fire frequencies, allowing managers to target areas at high risk of threshold-type responses owing to altered fire regimes and interactions with other factors.

**Creation Method:** The FRID methodology was developed and described by Van de Water and Safford (2011). The feature class is now produced and maintained by the U.S. Forest Service, Region 5, Information Management – Mapping and Remote Sensing (MARS) Team. Contemporary FRIs were calculated using the fire dates and footprints from California Interagency Fire Perimeters database (maintained by the California Department of Forestry and Fire Protection (CalFire-FRAP). The vegetation type stratification (i.e. to calculate the FRI for individual vegetation types) was based on the MARS Existing Vegetation (EVEG) map for California from the year 2011, with the vegetation typing (“CALVEG”) cross-walked (grouped) into 28 pre-settlement fire regime (PFR) types.

For assorted reasons, portions of San Benito and San Luis Obispo Counties never received a full EVEG Baseline Mapping assessment and thus data in the FRID Central California layer has some holes in these areas. In 2009, an EVEG mapping project was started for these areas but never finalized. San Luis Obispo County, the southern part of Santa Clara County, and all of San Benito County were baseline mapped using the Hardwood Dataset as a foundation for regional dominance (vegtype). Additional data sources from the National Land Cover Database, San Luis Obispo County Farm Data, Farmland Mapping & Monitoring Program, Bureau of Reclamation, and National Hydrology Database were then used to overwrite the Hardwood data where it was relevant. Structural attributes for forested conditions came primarily from the Hardwoods Dataset for canopy values while tree size was derived from a classification of Thematic Mapper 30-meter imagery.

Preparation of the Fire Return Interval Departure (FRID) data requires use of up to date statewide vegetation data. For this purpose we have been using EVEG, as described above. This has been adequate for most of California but there are some areas, because of missing data, that required some adjustments.

Although incomplete as an EVEG database, these “best available data” were used by the RRK team to fill holes in FRID for the Central California RRK project. The MARS team completed a crosswalk from Regional Dominance Type 1 (vegtype) to the FRID PFR attribute and calculations for the “gap” areas were run for fire return interval departure. We have used this “patch” to address FRID needs for the near-term. The data for these areas will show vulnerabilities to analysis at larger scales until a time that these areas can be visually edited to match the level of precision seen in the adjoining Los Padres NF.

### **Other gaps (NoData):**

Although areas mapped as grasslands and meadows were included in the GIS layer, FRI and departure statistics were not calculated for these types because reliable information about pre-Euro American settlement fire regimes is lacking. These values (-999) have been converted to NoData in the RRK datasets.

**Data Source:** USDA Forest Service, Region 5, MARS Team

**References:** Information on pre-Euromerican settlement FRIs (fire return intervals) was compiled from an exhaustive review of the fire history literature, expert opinion, and vegetation modeling (Van de Water and Safford 2011; Safford and Van de Water 2014). Contemporary FRIs were calculated using the California Interagency Fire Perimeters database (maintained by the California Department of Forestry and Fire Protection (CAL FIRE-FRAP)). The vegetation type stratification was based on the US Forest Service existing vegetation map (USDA Forest Service, Mapping and Remote Sensing Team) for California from the year 2011, with the vegetation typing (“CALVEG”) grouped into 28 pre-settlement fire regime (PFR) types, as defined by Van de Water and Safford (2011). The 2011 eVeg map is used as the baseline for all subsequent FRID maps to freeze the underlying vegetation template and permit temporal comparisons without introducing vegetation type change as a confounding factor.

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#### CURRENT FIRE RETURN INTERVAL DEPARTURE, SINCE 1908

**Tier:** 3

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** The fire return interval departure (FRID) analysis quantifies the difference between current and pre-settlement fire frequencies, allowing managers to target areas at high risk of threshold-type responses owing to altered fire regimes and interactions with other factors. This is a measure of the extent to which contemporary fires (i.e. since 1908) are burning at frequencies similar to the frequencies that occurred prior to Euro-American settlement.

**Data Resolution:** 30m Raster

**Data Units:** Average Years

**Creation Method:** Current fire return interval 1908 is calculated by dividing the number of years in the fire record by the number of fires occurring between 1908 and the current year in a given polygon plus one.

$$\text{CurrentFRI} = \text{Number of years} / \text{Number of Fires} + 1$$

**Data Source:**

Fire History (2022), CAL FIRE  
Existing Vegetation (CALVEG 2011), USDA Forest Service, Region 5, MARS Team

**File Name:** currentFRI\_2022.tif

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#### CURRENT FIRE RETURN INTERVAL DEPARTURE, SINCE 1970

**Tier:** 2

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** The fire return interval departure (FRID) analysis quantifies the difference between current and pre-settlement fire frequencies, allowing managers to target areas at high risk of threshold-type responses owing to altered fire regimes and interactions with other factors. This is a measure of the extent to which contemporary fires (i.e. since 1970) are burning at frequencies similar to the frequencies that occurred prior to Euro-American settlement, with the mean reference FRI as the basis for comparison. With this metric, mPFRID\_1970, the same formulas are used as with meanPFRID but with 1970 as the baseline rather than 1908. Important note: because 1970 is the baseline for this measure, no fires before 1970 are taken into account and all PFRs start at a PFRID of zero beginning in 1970.

**Data Resolution:** 30m Raster

**Data Units:** Average Years

**Creation Method:** Current fire return interval 1970 is calculated by dividing the number of years in the fire record by the number of fires occurring between 1970 and the current year in a given area plus one.

$$\text{CurrentFRI}_{1970} = \text{Number of years} / \text{Number of Fires} + 1$$

**Data Source:**

Fire History (2022), CAL FIRE

Existing Vegetation (CALVEG 2011), USDA Forest Service, Region 5, MARS Team

**File Name:** currentFRI\_1970\_2022.tif

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MEAN PERCENT FRI DEPARTURE, SINCE 1908

**Tier:** 3

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** This metric, mean percent FRID, is a measure of the extent to which contemporary fires (i.e., since 1908) are burning at frequencies similar to the frequencies that occurred prior to Euro-American settlement, with the mean reference FRI as the basis for comparison. Mean PFRID is a metric of fire return interval departure (FRID) and measures the departure of current FRI from mean reference FRI in percent.

**Data Resolution:** 30m Raster

**Data Units:** Percent

**Creation Method:** The current FRI is calculated by dividing the number of years in the fire record (e.g., 2019-1908=112 years inclusive) by the number of fires occurring between 1908 and the current year in a given polygon plus one (CurrentFRI = Number of years/Number of fires +1). The mean reference FRI is an approximation of how often, on average, a given PFR likely burned in the three or four centuries prior to significant Euro-American settlement. This measure does not return to zero when a fire occurs, unlike FRID values used in some other analyses (e.g., NPS FRID Index). Instead, the following formulas are used to calculate Mean PFRID:

When current FRI is longer than reference FRI (the common condition in most coniferous PFRs) the formula is:

$$[1 - (\text{MeanRefFRI} / \text{CurrentFRI})] * 100$$

When current FRI is shorter than reference FRI (common in some shrub dominated PFRs, and areas in the Wildland Urban Interface) the formula is:

$$- \{ [1 - (\text{CurrentFRI} / \text{MeanRefFRI})] \} * 100$$

For areas dominated by PFRs with a mean reference FRI greater than 112 years, and that have not burned in the period of historical record considered in this analysis (i.e., since 1908), the FRID is assumed to equal zero.

**Data Source:**

Fire History (2022), CAL FIRE

Existing Vegetation (CALVEG 2011), USDA Forest Service, Region 5, MARS Team

**File Name:** meanPFRID\_2022.tif

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## MEAN PERCENT FRI DEPARTURE, SINCE 1970

**Tier:** 2

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** Mean Percent FRID (meanPFRID\_1970) quantifies the extent in percentage to which recent fires (i.e., since 1970) are burning at frequencies similar to those that occurred prior to Euro-American settlement, with the mean reference FRI as the basis for comparison. Mean PFRID measures the departure of current FRI from reference mean FRI in percent

**Data Resolution:** 30m Raster

**Data Units:** Percent

**Creation Method:** The current FRI is calculated by dividing the number of years in the fire record (e.g., 2019-1970=49 years inclusive) by the number of fires occurring between 1970 and the current year in a given polygon plus one ( $\text{CurrentFRI} = \text{Number of years} / \text{Number of fires} + 1$ ). The mean reference FRI is an approximation of how often, on average, a given PFR likely burned in the three or four centuries prior to significant Euro-American settlement. This measure does not return to zero when a fire occurs, unlike FRID values used in some other analyses (e.g., NPS FRID Index).

**Data Source:**

Fire History (2022), CAL FIRE

Existing Vegetation (CALVEG 2011), USDA Forest Service, Region 5, MARS Team

**File Name:** meanPFRID\_1970\_2022.tif

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## MEAN FRID CONDITION CLASS FOR DEPARTURE

**Tier:** 3

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** This metric uses the mean percent FRID to a measure of the extent to which contemporary fires (i.e., since 1908) are burning at frequencies similar to the frequencies that occurred prior to Euro-American settlement, with the mean reference FRI binned into another basis for comparison. Mean PFRID is a metric of fire return interval departure (FRID), and measures the departure of current FRI from reference mean FRI in percent.

**Data Resolution:** 30m Raster

**Data Units:** Integer, -3 to 3

**Creation Method:** This is a condition class categorization of the data in the Mean PFRID field. MeanCC\_FRI categorizes the percent differences calculated in Mean PFRID using the following scale:

- 1: 0 to 33.3% departure
- 2: 33 to 66.7% departure
- 3: >66.7% departure

Negative condition classes (i.e., where fires are burning more often than under pre-Anglo-American settlement conditions) are categorized on the negative of the same scale:

- 1: 0 to -33.3%
- 2: -33 to -66.7%
- 3: <-66.7%

CC1 and CC-1 are mapped in the same class because they are both within 33% of the mean pre-settlement value.

**Data Source:**

Fire History (2022), CAL FIRE

Existing Vegetation (CALVEG 2011), USDA Forest Service, Region 5, MARS Team

**File Name:** meanCC\_FRI\_2022.tif

TIME SINCE LAST FIRE

**Tier:** 2

**Data Vintage:** 2022. Includes disturbances through the end of 2022.

**Metric Definition and Relevance:** Time Since Last Fire (TSLF), from the Fire Return Interval Departure (FRID) map, provides information (in years) to indicate the length of time since an area last burned.

**Data Resolution:** 30m Raster

**Data Units:** Years

**Creation Method:** Time Since Last Fire (TSLF), from the Fire Return Interval Departure (FRID) map, provides information (in years) to indicate the length of time since an area last burned. Specifically, the number of years elapsed between the most recent fire recorded in the fire perimeters database and the version year of the FRID map being used. To illustrate, if the version year of the FRID map is 2019, and the area in question last burned in 1995, TSLF will be 24 (2019 minus 1995).

**Data Source:**

Fire History (2022), CAL FIRE

Existing Vegetation (CALVEG 2011), USDA Forest Service, Region 5, MARS Team

**File Name:** TSLF\_2022.tif

SEVERITY

Uncharacteristic proportions of high severity fire over the area burned, particularly in the last decade, has been a common theme in the megafires that have occurred throughout the Sierra recently. The following metrics characterize, map, and quantify some of the factors that contribute.

PROBABILITY OF HIGH FIRE SEVERITY

**Tier:** 1

**Data Vintage:** 08/2023. Includes disturbances through the end of 2022

**Metric Definition and Relevance:** These metrics depicts the probability of high severity fire as constructed by Pyrologix LLC. This operational-control probability raster indicates the probability that the headfire flame length in each pixel will exceed 8 foot flame lengths, the threshold that defines fires that would exceed manual control.

**Data Resolution:** 30m Raster

**Data Units:** Probability, 0 to 1

**Creation Method:** Probability of High Fire Severity (defined as >8 ft) was produced by Pyrologix LLC, a wildfire threat assessment research firm, as part of a spatial wildfire hazard assessment across all land ownerships for the state of California. The ongoing work generally follows the framework outlined in Scott and Thompson (2013), with custom methods and significant improvements developed by Pyrologix. The project generally consists of three components: fuelscape calibration and updates, wildfire hazard assessment, and risk assessment. To date, this work has resulted in a wide variety of spatial data layers related to wildfire hazard and risk, including operational control probabilities based on conditions prior to the 2023 fire season. Work to date has been funded by the USDA Forest Service Region 5, the California Energy Commission, and the USDI Bureau of Land Management with data contributions from CAL FIRE. Please reference the Pyrologix 2021 project report (Volger et al., 2021) for more information.

Pyrologix uses the Wildfire Exposure Simulation Tool (WildEST) to develop this data layer, a deterministic wildfire modeling tool that integrates variable weather input variables and weights them based on how they will likely be realized on the landscape. WildEST is more robust than the stochastic intensity values developed with FSim. This is especially true in low wildfire occurrence areas where predicted intensity values from FSim are reliant on a very small sample size of potential weather variables.

**Data Source:** Pyrologix, LLC

**File Name:** ProbabilityHighFireSev\_2023.tif

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#### ANNUAL BURN PROBABILITY

**Tier:** 1

**Data Vintage:** 08/2023. Includes disturbances through the end of 2022

**Metric Definition and Relevance:** Annual Burn Probability represents the likelihood of a wildfire of any intensity occurring at a given location (pixel) in a single fire season. In a complete assessment of wildfire hazard, wildfire occurrence and spread are simulated in order to characterize how temporal variability in weather and spatial variability in fuel, topography, and ignition density influence wildfire likelihood across a landscape. In such cases, the hazard assessment includes modeling of burn probability, which quantifies the likelihood that a wildfire will burn a given point (a single grid cell or pixel) during a specified period of time. Burn probability for fire management planning applications in this case is reported on an annual basis - the probability of burning during a single fire season.

**Data Resolution:** 30m Raster

**Data Units:** Probability, 0 to 1

**Creation Method:** Annual Burn Probability was produced by Pyrologix LLC, a wildfire threat assessment research firm, as part of a spatial wildfire hazard assessment across all land ownerships for the state of California. The ongoing work generally follows the framework outlined in Scott and Thompson (2013), with custom methods and significant improvements developed by Pyrologix. The project generally consists of three components: fuelscape calibration and updates, wildfire hazard assessment, and risk assessment. It utilizes a combination of wildfire

models and custom tools, including the FSim large wildfire simulator (Finney et al., 2011). To date, this work has resulted in a wide variety of spatial data layers related to wildfire hazard and risk, including Annual Burn Probability, representing conditions prior to the 2023 fire season. Work to date has been funded by the USDA Forest Service Region 5, the California Energy Commission, and the USDI Bureau of Land Management with data contributions from CAL FIRE.

For this project, the USFS modeling system called FSim is used to quantify annual wildfire likelihood across California. The model is parameterized using spatial datasets of historical weather, fire occurrence, fuels, weather, and topography in order to simulate thousands of fire-years on a landscape. Annual Burn Probability is calculated from these simulations using a Monte Carlo approach to make a spatially resolved estimate of the contemporary annual likelihood of wildfire across the landscape. For more information on FSim or the wildfire hazard modeling being performed by Pyrologix, please see Volger et al., 2021.

**Data Source:** Pyrologix, LLC

**File Name:** AnnualBurnProbability\_2023.tif

## FOREST AND SHRUBLAND RESILIENCE

Forest and shrubland resilience is the ability of forest and shrubland vegetation and structure to remain a forest or shrubland in the face of disturbance (e.g., fire, forest management, climate change, etc.). The Forest and Shrubland Resilience Pillar evaluates forest and shrubland vegetation composition and structure to determine its alignment with desired disturbance dynamics and within tolerances of current and future biophysical conditions when considering changes due to climate change. The last 100 years of forest and shrubland management, combined with changing climates, have resulted in forest and shrubland structure and composition which are not resilient to contemporary disturbances. Forest or shrubland structure and composition are one of the few elements of a wildland that management can modify through treatments to improve conditions.

**DESIRED OUTCOME:** Vegetation composition and structure align with topography, desired disturbance dynamics, and landscape conditions, and are adapted to climate change.

### STRUCTURE

Forest or shrubland structure is the spatial distribution of vegetation (live and dead) both vertically and horizontally on the landscape. Prior to European settlement, forests in the Sierra Nevada were characterized by heterogeneous spatial patterns replete with individual large trees, gaps, and tree clumps of various sizes – patterns that were shaped by recurrent fire and other disturbances. After a century-plus of fire exclusion, timber harvesting, agricultural development, urbanization, and other land-use practices, the predominant trend across Californian landscapes is that they have become less resilient to natural and human-caused disturbances. In many cases some sort of restoration treatment may be necessary to reverse these trends.

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### DENSITY – LARGE TREES

**Tier:** 2

**Data Vintage:** 06/2020

**Metric Definition and Relevance:** Large trees are important to forest managers as they have a greater likelihood of survival from fire, provide sources of seed stock, wildlife habitat, and contribute to other critical processes like

carbon storage and nutrient cycling. Large trees are often the focus of management in order to protect existing ones and to foster future ones. In consultation with National Forests, “large trees” have been designated as greater than 30” dbh.

**Data Resolution:** 30m Raster

**Data Units:** Percent live trees per pixel

**Creation Method:** To determine the cutoff for the definition of large trees in the southern California area (> 30” dbh - need reference ), a statistical relationship between tree dbh and tree height was developed. We used Forest Inventory & Analysis (FIA) 2022 plot data from the region, testing three fits of DBH ~ HGT: Saturating (Michelis Menton), Power, or Linear. Saturating was the best according to Akaike information criterion (AIC). We then extracted heights that predicted DBH cut-offs:

Predicted DBH (inches)	Height (feet)	RMSE (inches)
<=1	2.5	1.5
>1 - 6	14.5	2.4
>6 - 11	27	5.3
>11 - 24	64.5	6.8
>=30	83.5	7.3

Block statistics were run on California Forest Observatory (CFO) canopy height pixels greater than or equal to 83.5’ (25m) with 3x3 window to calculate the sum for input cells within a 30m rectangular neighborhood. This assigned number of pixels per 30m (900m2) cell. Resultant values of 1 through 9 were converted to percent. All background values were calculated to equal 0, meaning 0% large tree existence.

**Data Source:** California Forest Observatory (Salo Sciences), 2020

**File Name:** LargeTreeDensity\_2020.tif

CANOPY VEG COVER

**Tier:** 1

**Data Vintage:** 06/2020

**Metric Definition and Relevance:** This layer represents horizontal cover fraction occupied by tree canopies. Maps community type & fire regime, as well as available habitat for tree-dwelling species.

**Data Resolution:** 30m Raster

**Data Units:** Canopy cover is a 0-100% cover fraction and may be more precisely described as "canopy density." It calculates the proportion of all lidar returns >=5m divided by the total number of returns in that grid cell. This, therefore, does not include all vegetation, but instead describes the density of vegetation in the canopy vertical stratum (veg 5m and taller).

**Creation Method:** Each forest structure metric was derived directly from airborne lidar data, hosted by the USGS 3D Elevation Program. However, these data are only available for a small fraction of California’s 423,970 km<sup>2</sup> area.

To overcome this, we trained deep learning models—a form of pattern recognition—to identify these forest structure patterns in satellite imagery, then mapped each metric statewide.

These algorithms are of the U-net family of neural network architectures that perform pixel-wise regression and classification tasks. The satellite data includes imagery from Sentinel-1 C-band radar sensors and Sentinel-2 multispectral sensors at 10 m spatial resolution, collected in Fall 2019. Future versions will include imagery from PlanetScope multispectral sensors at 3 m resolution.

The 10m raster was resampled to 30m resolution by the RRK team.

Original dataset downloaded from [California Forest Observatory - Organizations - WIFIRE Commons Data Catalog \(sdsc.edu\)](https://forestobservatory.com/about.html#about). For more information, go to <https://forestobservatory.com/about.html#about>

**Data Source:** California Forest Observatory (Salo Sciences), 2020

**File Name:** CFO\_CanopyCover2020Summer.tif

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#### CANOPY VEG HEIGHT

**Tier:** 1

**Data Vintage:** 06/2020

**Metric Definition and Relevance:** This layer represents distance between the ground and the top of the canopy. Canopy height is a proxy for aboveground biomass and the amount of foliage that may be consumed in a canopy fire. Since LANDFIRE doesn't support a NoData value, all NoData pixels in canopy fuel metrics were set to 0 in the Landscape files. (e.g., canopy cover was set to 0 in all NoData locations). Topographic data and surface fuel model remain unaltered.

**Data Resolution:** 30m Raster

**Data Units:** meters, min 0 - max 80; each pixel value represents the average height above ground for vegetation within that pixel

**Creation Method:** Each forest structure metric was derived directly from airborne lidar data, hosted by the USGS 3D Elevation Program. However, these data are only available for a small fraction of California's 423,970 km<sup>2</sup> area. To overcome this, we trained deep learning models—a form of pattern recognition—to identify these forest structure patterns in satellite imagery, then mapped each metric statewide.

These algorithms are of the U-net family of neural network architectures that perform pixel-wise regression and classification tasks. The satellite data includes imagery from Sentinel-1 C-band radar sensors and Sentinel-2 multispectral sensors at 10 m spatial resolution, collected in Fall 2019. Future versions will include imagery from PlanetScope multispectral sensors at 3 m resolution.

The 10m raster was resampled to 30m resolution by the RRK team.

Original dataset downloaded from [California Forest Observatory - Organizations - WIFIRE Commons Data Catalog \(sdsc.edu\)](https://forestobservatory.com/about.html#about). For more information, go to <https://forestobservatory.com/about.html#about>

**Data Source:** California Forest Observatory (Salo Sciences), 2020

**File Name:** CFO\_CanopyHeight2020Summer.tif

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#### CANOPY LAYER COUNT

**Tier:** 1

**Data Vintage:** 06/2020

**Metric Definition and Relevance:**

This layer represents the number of distinct vertical canopy layers of trees. Vertical layer count is a proxy for leaf area index, and maps canopy complexity. Since LANDFIRE doesn't support a NoData value, all NoData pixels in canopy fuel metrics were set to 0 in the Landscape files. (e.g., canopy cover was set to 0 in all NoData locations). Topographic data and surface fuel model remain unaltered.

**Data Resolution:** 30m Raster

**Data Units:** Count

**Creation Method:** Each forest structure metric was derived directly from airborne lidar data, hosted by the USGS 3D Elevation Program. However, these data are only available for a small fraction of California's 423,970 km<sup>2</sup> area. To overcome this, we trained deep learning models—a form of pattern recognition—to identify these forest structure patterns in satellite imagery, then mapped each metric statewide.

These algorithms are of the U-net family of neural network architectures that perform pixel-wise regression and classification tasks. The satellite data includes imagery from Sentinel-1 C-band radar sensors and Sentinel-2 multispectral sensors at 10 m spatial resolution, collected in Fall 2019. Future versions will include imagery from PlanetScope multispectral sensors at 3 m resolution.

The 10m raster was resampled to 30m resolution by the RRK team.

Original dataset downloaded from [California Forest Observatory - Organizations - WIFIRE Commons Data Catalog \(sdsc.edu\)](https://forestobservatory.com/about.html#about). For more information, go to <https://forestobservatory.com/about.html#about>

**Data Source:** California Forest Observatory (Salo Sciences), 2020

**File Name:** CFO\_CanopyLayerCount2020Summer.tif

## COMPOSITION

The composition of a forest is a reference to the biodiversity of the landscape; this includes a diversity of vegetation species, types (e.g., trees, shrubs, forbs, etc.), and distribution. Tree species composition affects many aspects of forest dynamics and function. A diversity of tree and shrub species can confer greater resilience to climate change and beetle outbreaks. The vegetation composition also affects fire dynamics, water reliability, carbon pools and sequestration, and economic diversity pillars. Since European settlement and the adoption of fire suppression and logging, forests of the Sierra Nevada shifted to increased dominance of shade-tolerant and fire-intolerant species like white fir and red fir, incense cedar, Douglas fir, and tanoak. Other species like ponderosa pine, Jeffrey pine, sugar pine, and black oak, which are more shade-intolerant and fire-tolerant, declined in coverage. With increasingly larger and higher-severity fire occurring, forest-cover loss may be significant and shrub cover will increase.

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## SERIAL STAGE

**Tier:** 2

**Data Vintage:** 04/2023

**Metric Definition and Relevance:** The seral stages are categories that represent the developmental progression of forest ecosystems from initial establishment or following a stand replacing event (e.g., high severity fire) to a forest dominated by trees in the upper age classes for a given forest type. Late seral forests are also often characterized by multiple ages of forest trees and dead and dying trees in some form of equilibrium. Seral conditions across landscapes were highly variable prior to major European settlement in the western US. These patterns were highly attuned to dominant disturbance regimes and the multi-scaled variability in environmental conditions across topographic and climatic gradients. These patterns helped to reinforce fire regimes dominated by low- to moderate-severity fire across much of the region and provided for multiple habitat requirements for a wide variety of species.

This metric contains three related data layers. The first is an assignment to each 30 meter pixel of the seral stage it is currently in, either early, mid, or late seral stage. The other two layers represent the proportion of a HUC 12 watershed that is in 1) early seral stage or 2) late seral stage.

**Data Resolution:** 30m Raster, HUC 12 watersheds

**Data Units:** Categorical 1 - 3 (seral stage), continuous variable 0-1 representing percentage of a HUC (early and late seral stage)

**Creation Method:** The FVEG data, used in characterizing vegetation and habitat conditions for a number of metrics in this kit, contain data on tree size ([see FVEG discussion below](#)). Seral stages for forested lands are binned into one of three categories of tree size (Early, Mid, Late) and those are defined by tree diameter, per the CWHR system.

Size Class	Size (inches DBH)	Seral Stage
1 - Seedling	less than 1	Early (1)
2 - Sapling	1 – 6	Early (1)
3 - Pole	6 – 11	Mid (2)
4 - Small	11 – 24	Mid (2)
5 - Medium to Large	24+	Late (3)
6 - <u>Multi-storied</u>	36 – 48	Late (3)

Late Seral conditions have been lumped into a single classification (24” and up).

The first layer provided here assigns a early, mid, or late seral value to each cell based on dominant tree size in the canopy. Null or NoData values in this raster represent areas of non-forest. The second and third data layer provided identify the proportion of the HUC12-scale (typically 10,000-30,000 acres in size) that is either early seral forest or late seral forest, respectively. These patterns can be highly variable at finer-scales so we used a HUC 12 watershed as the unit for expressing relative abundance. For each HUC12, the proportion of the watershed covered by the evaluated seral stage has been calculated.

**Data Source:** FVEG 2023

**File Name:** SeralStage\_EML\_202304.tif; early\_SeralStage\_prop\_202304.tif; late\_SeralStage\_prop\_202304.tif

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#### TREE COVER

**Tier:** 1

**Data Vintage:** 12/2021

**Metric Definition and Relevance:** Total tree cover as measured by the fractional non-overlapping absolute tree cover, viewed vertically. Provides a first order measure of vegetation type when combined with parallel observations of shrub and herbaceous cover. Data from the National Land Cover Database (NLCD) are used for training, and NLCD definitions for cover (for example, the distinction between tree vs shrub) are expected to be similar in the CECS data sets.

**Data Resolution:** 30m Raster

**Data Units:** Fractional non-overlapping absolute cover; continuous variable from 0 to 1.

**Creation Method:** Machine learning (Random Forest) using the National Land Cover Database for training and Landsat observations as predictors. See <https://doi.org/10.1029/2021AV000654> for further information.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** VegCover\_Tree\_2021.tif

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#### SHRUB COVER

**Tier:** 1

**Data Vintage:** 12/2021

**Metric Definition and Relevance:** Total shrub cover as measured by the fractional non-overlapping absolute shrub cover, viewed vertically. Provides a first order measure of vegetation type when combined with parallel observations of tree and herbaceous cover. Data from the National Land Cover Database (NLCD) are used for training, and NLCD definitions for cover (for example, the distinction between tree vs shrub) are expected to be similar in the CECS data sets.

**Data Resolution:** 30m Raster

**Data Units:** Fractional non-overlapping absolute cover; continuous variable from 0 to 1.

**Creation Method:** Machine learning (Random Forest) using the National Land Cover Database for training and Landsat observations as predictors. See <https://doi.org/10.1029/2021AV000654> for further information.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** VegCover\_Shrub\_2021.tif

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#### HERBACEOUS COVER

**Tier:** 1

**Data Vintage:** 12/2021

**Metric Definition and Relevance:** Total herbaceous cover as measured by the fractional non-overlapping absolute herbaceous cover, viewed vertically. Provides a first order measure of vegetation type when combined with parallel observations of tree and herbaceous cover. Data from the National Land Cover Database (NLCD) are used for

training, and NLCD definitions for cover (for example, the distinction between tree vs shrub) are expected to be similar in the CECS data sets.

**Data Resolution:** 30m Raster

**Data Units:** Fractional non-overlapping absolute cover; continuous variable from 0 to 1.

**Creation Method:** Machine learning (Random Forest) using the National Land Cover Database for training and Landsat observations as predictors. See <https://doi.org/10.1029/2021AV000654> for further information.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** VegCover\_Herb\_2021.tif

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#### ABSOLUTE CHANGE IN HERBACEOUS COVER

**Tier:** 3

Data Vintage: 2020

**Metric Definition and Relevance:** A raster dataset representing absolute change of herbaceous fractional (%) land cover in the Southern California Region.

**Data Resolution:** 30m Raster

**Data Units:** Percent

Creation Method: Absolute change was calculated for annual and perennial grasses and forbs and litter, called “herbaceous” vegetation, using five-year-averages for 1986-1990 and 2016-2019 with data from the Rangeland Analysis Platform (RAP).  $\text{Absolute Change} = (2016 - 2020 \text{ Average}) - (1986-1990 \text{ Average})$

In areas where herbaceous cover increased over this period, values will be positive (up to 60%); where herbaceous cover decreased, values will be negative (down to -60%). These data may be used as an indicator of vegetation type conversion resulting from high fire frequency or other disturbances.

These data only represent herbaceous cover; therefore, a second layer of information is required to determine what the cover changed to in areas that experienced a decrease in herbaceous cover over the time period examined.

The Rangeland Analysis Platform (RAP) was created in collaboration with the U.S. Department of Agriculture’s Natural Resources Conservation Service, the Department of Interior’s Bureau of Land Management, and the University of Montana to map western United States rangelands. The data can be used by landowners, managers, conservationists, and scientists to inform land management, planning, decision making, and evaluation of outcomes. For more information about RAP, please refer to the website: <https://rangelands.app/>

RAP fractional (percent) cover maps are produced by combining field plots from the Bureau of Land Management (BLM) Assessment, Inventory, and Monitoring (AIM) datasets and National Resources Conservation Service (NRCS) National Resources Inventory (NRI) with the historical Landsat satellite record. Due to a low amount of field points collected in the CWC focus area, the data products should be used alongside local knowledge and site-specific data to inform actions and decision making. To learn more about how RAP maps and data are intended to be used, refer to this site: <https://support.rangelands.app/article/46-new-to-rap-start-here-landing>

**Data Source:** San Diego State University CWC Project Team

**File Name:** AbsChgeInHerbCover1986\_90To2016\_20.tif

Tier: 2

Data Vintage: 2011

**Metric Definition and Relevance:** This dataset consists of 5 raster files of the proportion of above ground live biomass in these vegetation type categories: (1) Shrub obligate resprouting; (2) Shrub obligate seeding; (3) Shrub facultative seeding; (4) Tree; and (5) Herb. The spatial extent of these data cover 6,441,208 ha and is defined by the 42 Level IV Ecoregions (Bailey 2016) that intersect the four southern US Department of Agriculture (USDA) National Forests in southern California (Angeles, Cleveland, Los Padres, and San Bernardino):



Mediterranean-climate region (MCR) shrublands have evolved a set of regeneration strategies in response to periodic, high-intensity wildfires: obligate seeding (OS), obligate resprouting (OR), and facultative seeding (FS) species. Spatial variation is seen in different regeneration strategies. In California, previous studies have found a higher abundance of OR species in mesic environments and OS species in xeric environments).

Analyzing the spatial rasters depicting OS-, OR- or FS- dominated pixels, researchers found dramatically different spatial patterns between the three shrub regeneration strategies. FS species covered the greatest spatial distribution, accounting for 3,372,125 ha (71%) of shrub dominated pixels in the study area: the FS group covered a range of productivity gradients and vegetation types, although it was notably absent from high elevation areas. In contrast, OS-dominated pixels covered the smallest spatial area (21,899 ha, 5% of shrub dominated pixels) occurring throughout the study area from coastal Big Sur and the Santa Ynez Mountains on the Los Padres National Forest to interior locations including the eastern fringes of the San Jacinto mountains (desert shrub vegetation) in the south. Finally, pixels dominated by OR species covered a similarly small area, 25,075 ha (5% of shrub dominated pixels in the study area), showing aggregations in the San Bernardino and San Gorgonio Mountains on the San Bernardino National Forest; San Gabriel Mountains on the Angeles National Forest, and throughout higher elevations on the Los Padres National Forest. OR-dominated pixels were notably absent in lower elevation areas

with low water availability, as indicated in the relationship with climatic water deficit and solar radiation. The proportion of each post-fire shrub regeneration type is critical for a number of reasons, including assessing the ability of shrublands to recover from multiple, short-interval fires or helping to prioritize areas for post-fire restoration.

**Data Resolution:** 30 meters

**Data Units:** Percentage of life history/post-fire regeneration type per pixel (0 – 100)

**Creation Method:** Researchers developed a multinomial model using temporally dynamic and static variables to predict the distribution of the three shrub post-fire regeneration strategies - obligate seeders (OS), facultative seeders (FS), obligate resprouters (OR), plus trees and herbs, in southern California. Researchers used 222 USDA Forest Service Forest Inventory and Analysis (FIA) plots and 17 predictor variables including: median 2001-2020 NDVI and biomass, vegetation type, precipitation, slope, aspect, and soil characteristics. Each predictor variable was available as a raster dataset to which the model was applied to produce the 5 raster files which estimate the proportion (percentage) of biomass within each of the five groups. Overall, model cross-validation showed the accuracy achieved 50% of predicted value within 8 to 24 percent of the actual value, with prediction accuracy highest for herb biomass and lowest for OR. Of the three shrub regeneration strategies, OS biomass was predicted with the highest accuracy and narrowest environmental range.

The files are named as follows:

SoCal\_ObResprouters\_pp\_v1.tif = percentage of obligate resprouter biomass per pixel.

SoCal\_ObSeeders\_pp\_v1.tif = percentage of obligate seeder biomass per pixel.

SoCal\_FacSeeders\_pp\_v1.tif = percentage of facultative seeder biomass per pixel.

SoCal\_Tree\_pp\_v1.tif = percentage of tree biomass per pixel.

SoCal\_Herb\_pp\_v1.tif = percentage of herb biomass per pixel.

**Data Source:** Underwood, E.C., Q.M. Sorenson, C.C. Schrader-Patton, N.A. Molinari and H.D. Safford. 2023. Resprouting, seeding, and facultative seeding shrub species in California's Mediterranean-type climate region. *Frontiers in Ecology and Evolution* 11:1158265. doi: 10.3389/fevo.2023.1158265

Underwood, E.C.; Q.M. Sorenson, C.C. Schrader-Patton (2023). Obligate resprouting, obligate seeding, and facultative seeding shrub species in California's Mediterranean-type climate region [Dataset]. Dryad. <https://doi.org/10.25338/B8FS9V>

**File Name:** SoCal\_ObResprouters\_pp\_v1.tif; SoCal\_ObSeeders\_pp\_v1.tif; SoCal\_FacSeeders\_pp\_v1.tif; SoCal\_Tree\_pp\_v1.tif; SoCal\_Herb\_pp\_v1.tif

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SHRUBLANDS WITH LOW NATURAL REGENERATION POTENTIAL POST-FIRE IN SOUTHERN CALIFORNIA

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Identifying locations where shrubland vegetation will not recover naturally post-fire is a challenge given the vast areas that are regularly burned in southern California. When shrublands are within the historic fire return interval, e.g., 55 years for low-elevation shrubland, biomass accumulates and shrub cover recovers after 10–14 years. However, in many parts of southern California, the fire return interval has

decreased, often in conjunction with an increase in non-native plant species, drought, and nitrogen deposition. Under these conditions, post-fire biomass recovery can be impeded and, in some cases, may result in type conversion from native shrubland to non-native grassland. Researchers developed a repeatable method to identify areas of low regeneration potential in southern California using fire history data and applying two thresholds guided by the published literature. Low regeneration pixels either had a 'number of fires in the last 40 years' of three or more fires, or the 'time since last fire' was <10 years. Researchers identified pixels that met these criteria as having low natural regeneration potential post-fire and, as a consequence, these areas could represent candidate areas for post-fire restoration in shrublands.

The spatial extent of these data cover 6,441,208 ha and is defined by the 42 Level IV Ecoregions (Bailey 2016) that intersect the four southern US Department of Agriculture (USDA) National Forests in southern California (Angeles, Cleveland, Los Padres, and San Bernardino):



**Data Resolution:** 30 meters

**Data Units:** Binary. 1 = shrub pixel has low regeneration potential post-fire, 0 = shrub pixel has potential to regenerate post-fire (based on decision rules).

**Creation Method:** Researchers obtained the historical wildfire perimeter database from the California Department of Forestry and Fire Protection (FRAP 2021). They then generated from this database a set of binary rasters indicating the occurrence of fire in a pixel for a given year across the southern California study area. The years for this input data stack range from 1967 to 2020. Researchers implemented a script to calculate the low natural regeneration potential rasters. Two binary rasters were created for each year, one with the number of fires that burned in the last 40 years and one with the number of fires in the last 10 years. These inputs were then used to create a raster depicting low shrub regeneration potential for any given year, defined as 3 or more fires in 40 years or 1 or more fires in the last 10 years (see Dryad publication for more details).

**Description of the data and file structure:** The full dataset contains 9 rasters for the years 2008, 2009, 2010, 2015, 2016, 2017, 2018, 2019, and 2020. Using the fire perimeters specific for each year (FRAP 2021) the researchers identified shrub pixels with low regeneration potential using two criteria: either the 'number of fires in the last 40

years' was three or more fires or the 'time since last fire' was <10 years. These two criteria are based on published literature[ECU1] [SC(FU2)] .

Shrubland pixels identified with these fire history characteristics are unlikely to regenerate naturally post-fire. This information provides a key input into the SoCal EcoServe tool designed for US Department of Agriculture Forest Service resource managers and also the Post-fire Restoration Prioritization tool which integrates these data with other spatial data which might influence shrubland recovery post-fire, such as drought and non-native annual grasses.

**Data Source:** Underwood, E.C. and A. D. Hollander. 2023. Areas of low natural regeneration potential post-fire in shrublands of southern California (selected years between 2008 and 2020) [Dataset]. Dryad.

<https://doi.org/10.25338/B8CH2T>

See Underwood and Hollander 2023 for rasters for the other 8 years

**File Name:** prepregen2020.tif

## DISTURBANCE

California forests evolved with a suite of frequent disturbances: wildfires (both from lightning and burning by indigenous people), bark beetle-caused mortality, drought-caused mortality, avalanches, landslides, and windthrow, all of which created forest heterogeneity across the landscape. This heterogeneity included variations in surface and ladder fuels, which moderated fire behavior and spread. The variations in stand density and forest opening also served as critical habitats for wildlife. Forested areas are now more homogeneous due to lack of disturbance. The lack of disturbance is evident in the forest structure.

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## GOLDSPOTTED OAK BORER

**Tier:** 2

**Data Vintage:** 2010 to 2021

**Metric Definition and Relevance:** Tree mortality and other forest damage is detected by annual aerial surveys over forested lands by state and federal agency staff. The primary purpose of the aerial survey is to create sketch maps of areas containing current year conifer and hardwood mortality, defoliation, and other damage. Number of trees and acres with damage are calculated for areas surveyed and reported annually using the methodology described below. Aerial surveys have been recognized for over fifty years as an efficient and economical method of detecting and monitoring forest change events over large forested areas. As with all remotely sensed data, some amount of ground-truthing is required before the data can be considered reliable. The goldspotted oak borer was identified based on field surveys starting in 2006, and coded as goldspotted oak borer during aerial detection surveys beginning in 2010.

**Field definitions:**

DCA\_CODE = damage casual agent code

DCA = Damage casual agent

Damage\_typ = damage type

Mort\_TPA = mortality trees per acre, the modifier used to determine number of trees when multiplied by acres

NUM\_TREES = estimated number of trees

RPT\_YR = report year  
HOST\_ALL\_C = compilation of host and HostGroup names and codes  
HOST\_ALL = compilation of host and HostGroup names

For more information, go to  
<https://www.fs.usda.gov/foresthealth/applied-sciences/mapping-reporting/digital-mobile-sketch-mapping.shtml>

**Data Resolution:** 30 m Raster

**Data Units:** specific to each attribute

**Creation Method:** Recent tree mortality and currently active non-mortality damage is sketch mapped on a mobile device by an aerial observer flying in a small fixed wing aircraft at  $\approx 1000'$  above ground level (AGL) searching for visibly dried and discolored foliage, typically yellow to reddish brown. Generally, two observers are onboard the aircraft surveying on opposite sides with about a two-mile swath width for each surveyor.

Drawn polygons are then attributed with the following: a) damage type, mortality or one of several non-lethal damage types (topkill, defoliation, branch flagging, die back or discoloration) (field name: DAMAGE\_TYP) b) percentage category of forest area affected, c) affected tree species and d) probable damage-causing agent (e.g., goldspotted oak borer). For more information go to

[https://www.fs.usda.gov/detail/r5/forest-grasslandhealth/?cid=fsbdev3\\_046707](https://www.fs.usda.gov/detail/r5/forest-grasslandhealth/?cid=fsbdev3_046707)

[https://www.fs.usda.gov/detail/r5/forest-grasslandhealth/?cid=fsbdev3\\_046696](https://www.fs.usda.gov/detail/r5/forest-grasslandhealth/?cid=fsbdev3_046696)

**Data Source:** R5 Aerial Detection Monitoring

**File Name:** goldspotOakBorer\_mortality.tif

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#### RISK OF TREE DIEOFF DURING DROUGHT

**Tier:** 1

**Data Vintage:** 12/2021

**Metric Definition and Relevance:** A quantitative continuous variable that reflects the risk of tree dieoff during a significant drought period (SPI48 drought = -2).

**Data Resolution:** 30m Raster

**Data Units:** This is a dimensionless index that ranges from 0 to  $\sim 20000$ . Low values indicate minimal or no risk of tree dieoff during drought, either or both because there are few trees in the pixel and/or there is ample local moisture even during periods of extreme precipitation shortfall. High values indicate significant risk of tree dieoff during drought, as a result of both a high density of trees at the site and likelihood of extreme local moisture shortfall.

**Creation Method:** Calculated by combining information on the local moisture balance and tree density. Local moisture balance was calculated as the ratio of Annual Evapotranspiration with the canopy observed in 2021 to Precipitation during a SPI 48 drought = -2 based on local P observations during 1991-2020. This ratio quantifies the local moisture deficit/surplus that would be expected during a 48-month period with precipitation that is 2 standard deviations below the local 30 year Normal. Tree cover was determined from Landsat. See <https://doi.org/10.1038/s41561-019-0388-5> for further information.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** Vulner\_TreeDieoff\_SPI\_2\_2021.tif

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#### MULTI-STRESSOR REFUGIA

**Tier:** 3

**Data Vintage:** 2021

**Metric Definition and Relevance:** This raster dataset represents sites that may provide protection for natural communities from multiple threats including climate, fire, altered river channels, and density of recreational activities.

**Data Resolution:** 270m Raster

**Data Units:** This is a dimensionless index that ranges from 1.91 to 3.68. Low values indicate lower resilience to threats. High values indicate significant protection from threats.

**Creation Method:** Domains of Refugia: To consider how refugial conditions from a range of stressors can inform conservation planning and management, we integrated metrics of refugial capacity across different domains, which we define as social, ecological, or physical drivers, processes, or cycles that influence landscape structure, function, or composition. To persist in the Southern California landscape, species and ecosystems may need refugia from shifting climatic conditions, including extremely hot summers and prolonged droughts, but non-climate stressors can also affect conservation outcomes. In this landscape, changes in fire frequency can be a significant stressor affecting plant community structure and persistence. Anthropogenic features that modify hydrologic flows alter the ability of watersheds to sustain functional habitats. And finally, protected areas are often designed to mitigate the impacts of anthropogenic activities; however, recreational activities may alter the refugial capacity of the protected land, affecting the ability of the landscape to sustain species and their habitats. We combined these individual metrics to assess landscape level refugial capacity.

Sites with high refugial capacity (super-refugia sites) have, on average, 30% fewer extremely warm summers, 20% fewer fire events, 10% less exposure to altered river channels and riparian areas, and 50% fewer recreational trails than the surrounding landscape. Our results suggest that super-refugia sites (~8,200 km<sup>2</sup>) for some natural communities are underrepresented in the existing protected area network, a finding that can inform efforts to expand protected areas.

**Data Source:** San Diego State University CWC Project Team

**File Name:** refugia\_capacity\_4domains\_sum.tif

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#### CUMULATIVE TREE COVER LOSS

**Tier:** 1

**Data Vintage:** 12/2020

**Metric Definition and Relevance:** The cumulative loss of tree cover over a 30-year period (1992-2020). Tree cover loss reflects fires, harvest/management and dieoff. Only disturbances that are sufficient to trigger the Continuous Change Detection and Classification algorithm are included; low-level, diffuse dieoff is likely missed.

**Data Resolution:** 30m Raster

**Data Units:** Cumulative fractional non-overlapping absolute tree cover loss, where tree cover is a continuous variable from 0 to 1. Cumulative loss can exceed 1 in cases with multiple disturbances.

**Creation Method:** Vegetation disturbances were identified over the Landsat TM/ETM+/OLI era using the Continuous Change Detection and Classification algorithm (CCDC). The corresponding annual change in tree cover was determined with machine learning (Random Forest) using the National Land Cover Database for training and Landsat/CCDC observations as predictors; this produced a ~35-year stack of rasters that identified the locations and severity of tree cover loss. This stack was then summed for 1992-2020 to calculate the cumulative tree cover loss over a 30-year period. See <https://doi.org/10.1029/2021AV000654> for further information.

CECS data that reflect landscape changes resulting from disturbances require 6 to 12 months of Landsat observations **after a given year that included major disturbances (such as a high severity wildfire)** to fully quantify that disturbance. CECS data that reflect disturbance, such as this data layer, are therefore available **through water year 2020 (i.e. through September 2020)**.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** DistHist\_Severe\_Tree\_19922020.tif

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#### CUMULATIVE SHRUB COVER LOSS

**Tier:** 1

**Data Vintage:** 12/2020

**Metric Definition and Relevance:** The cumulative loss of shrub cover over a 30-year period (1992-2020). Shrub cover loss reflects fires, harvest/management and dieoff. Only disturbances that are sufficient to trigger the Continuous Change Detection and Classification algorithm are included; low-level, diffuse dieoff is likely missed.

**Data Resolution:** 30m Raster

**Data Units:** Cumulative fractional non-overlapping absolute shrub cover loss, where shrub cover is a continuous variable from 0 to 1. Cumulative loss can exceed 1 in cases with multiple disturbances.

**Creation Method:** Vegetation disturbances were identified over the Landsat TM/ETM+/OLI era using the Continuous Change Detection and Classification algorithm (CCDC). The corresponding annual change in shrub cover was determined with machine learning (Random Forest) using the National Land Cover Database for training and Landsat/CCDC observations as predictors; this produced a ~35-year stack of rasters that identified the locations and severity of shrub cover loss. This stack was then summed for 1992-2021 to calculate the cumulative tree cover loss over a 30-year period. See <https://doi.org/10.1029/2021AV000654> for further information.

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**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** DistHist\_Severe\_Shrub\_19922020.tif

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SHRUB RESILIENCY. NUMBER OF DISTURBANCE EVENTS PER 15 YEAR INTERVAL SINCE 1975

**Tier:** 2

**Data Vintage:** 2019

**Metric Definition and Relevance:** Count of short return interval fires (within 15 years) across southern CA (1950-2019) that can be used to identify sites that have experienced frequent fire (e.g. more than one fire within a 15 year period) that could lead to vegetation-type conversion.

**Data Resolution:** 270m Raster

**Data Units:** A count metric indicating the number of times each site (at the pixel level) met a threshold of a 15-year fire-return interval from 1950 to 2019.

**Creation Method:** Counts were aggregated from the CalFire state fire perimeter dataset using a moving time window of 15 years, which has been identified as a minimum threshold for chaparral or coastal sage scrub recovery after fire. Areas with more frequent fires within a 15-year window could result in decreased chaparral or coastal sage scrub recovery and increased occurrence of invasive species. Counts of 0 indicate a site experienced a single fire but no repeat fires within the 15-yr moving window. Counts of 1 to 9 indicate the number of fires beyond the initial event that recurred at the same pixel within the 15-yr moving window. Data can be used to identify sites that are likely to experience vegetation-type conversion.

Source: Conlisk, E., A.D. Syphard, E. Storey, K. West, C. Ross, M.K. Jennings, D. Stow. 2021. Connecting Wildlands and Communities Research Team; San Diego State University

**Data Source:** San Diego State University CWC Project Team

**File Name:** fireswithin15years\_1950-2019\_2ormorefires.tif

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POTENTIAL CLIMATE REFUGIA -BASELINE (HISTORICAL) CONDITIONS

**Tier:** 3

**Data Vintage:** 1981 - 2010

**Metric Definition and Relevance:** This raster dataset represents habitat types (natural vegetation communities) and their distribution across the array of climate conditions that each separate habitat type is found in under the baseline climate conditions. A 2015 map of the state's natural vegetation compiled from multiple sources was classified to the National Vegetation Classification Standard's mid-level classification, called "Macrogroup". Thirty one natural vegetation macrogroups are identified in the map, covering 99.87% of the state's natural terrestrial vegetation, and occupying 353,271 km<sup>2</sup>.

This serves as the foundation from which habitat types will be exposed to predicted changes in climate. Data are arrayed across 0 to 1 in terms of their exposure to current climate conditions. This data layer provides a baseline of vegetation adapted to "historic" conditions; i.e. climate conditions from the recent past; 1981-2010.

**Data Resolution:** 270m Raster

**Data Units:** 0- 1. Low values indicate higher resilience to threats. High values indicate significant exposure to climate change. -1 represents 'non analog' areas, i.e. locations that are outside the historic climate envelope of a given vegetation type.

**Creation Method:** The vegetation climate exposure analysis takes advantage of the 2015 vegetation map compiled for California by CALFIRE. Each Macrogroup (MG) was analyzed to determine which California habitats and associated dominant plant species make up its definition. California habitats are defined by the California Department of Fish and Wildlife (CDFW) through their California Wildlife Habitat Relationship (WHR) models<sup>9</sup>. WHR types are made up of plant species, such as the dominant trees, shrubs, and smaller plants. CDFW experts determined which WHR types correspond to each individual macrogroup; this cross-walk was used to develop a list of the dominant plant species that comprise each macrogroup.

The climate space occupied by each distinct vegetation macrogroup (largely equivalent to a CWHR habitat type) from the current time period was identified. This was done by using the points for each type and applying a kernel density estimator on a 2-d surface composed of the first two principal components of the climate conditions. The result is a smoothed continuous point density surface, showing the prevalence of each vegetation type across the range of sampled climatic conditions. This surface was partitioned by fitting contour lines so that they enclose a proportion of the original points from the current time period. Contours were calculated at 5% increments. For example the innermost 5% contour line encloses the 5% of pixels for the given vegetation type which are at the core of the climate space for that type, as determined by its density in the climate space. Cells further away from the dense central core, are considered to be more marginal in the vegetation type's distribution. The outer contours are fit to enclose the 95-99% of climatically marginal points, with the last 1% of cells (beyond the 99% contour) being the most marginal. In addition, if a cell lies outside the space defined by the 99% contour of any vegetation type, it is considered to be "non-analog," which means that it experiences climatic conditions outside of the conditions where we have a good sample in the initial time period. Excluded from this assessment are non-vegetated types such as snow, open water, and ice; and non-natural landcover types mapped as vineyards, tilled earth, orchards and Urban.

For more information on methods for the development of these climate refugia data see:

Thorne et al. 2015

Thorne et al. 2016

Thorne et al. 2017

Thorne et al. 2020

**Data Source:** Information Center for the Environment, UC Davis

**File Name:** hst8110.tif

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POTENTIAL CLIMATE REFUGIA - UNDER MODELED CLIMATE CHANGE (MIROC MODEL - HOTTER AND DRIER)

**Tier:** 3

**Data Vintage:** 2016

**Metric Definition and Relevance:** This raster dataset represents habitat types (CWHR habitat classes) and their predicted exposure to climate stress across the array of predicted climate conditions (separate layers for early (2010 - 2039), mid (2040-2069), and late century (2070-2099)) for all habitat types in comparison to the baseline

climate conditions. This serves as the foundation from which habitat types will be exposed to predicted changes in climate. Data are arrayed across 0 to 1 in terms of their exposure to current climate conditions. These three data layers can be used to help land managers allocate limited resources for climate-adaptive field work by providing a view of climate risk that varies across the lands they manage.

The Climate Change Model used in this analysis is the Miroc Earth System Model. This ESM, named “MIROC-ESM”, is based on a global climate model MIROC (Model for Interdisciplinary Research on Climate) which has been cooperatively developed by researchers in Japan and others. This model suggests a hotter and drier future. The emission scenario used is the RCP 8.5, which represents a range of warming statewide from 1.99 to 4.56°C and between a 24.8% decrease in precipitation and a 22.9% increase, respectively.

**Data Resolution:** 270m Raster

**Data Units:** 0- 1. Low values indicate higher resilience to threats. High values indicate significant exposure to climate change. -1 represents ‘non analog’ areas, i.e. locations that are outside the historic climate envelope of a given vegetation type.

**Creation Method:** The vegetation climate exposure analysis takes advantage of the 2015 vegetation map compiled for California, which is described above. The vegetation climate exposure model is implemented in the R programming language, and takes the vegetation and climate raster files as the primary input data. The values of the climate raster files were randomly sampled at 100,000 points on the landscape, which were used to fit a statistical model characterizing the relationship between the variables both in the current time and for the modeled future data.

At each of these 100,000 points, 9 hydro-climatic variables were sampled to characterize the range and variation of conditions in the study region. These variables were: annual mean minimum temperature (Tmin), annual mean maximum temperature (Tmax), annual precipitation (PPT), actual evapotranspiration (AET), potential evapotranspiration (PET), climatic water deficit (CWD), snowpack depth on April 1st, runoff, and recharge. The variation between these variables was modeled using a principal component analysis<sup>21</sup> (PCA) to identify the dominant components of variation. The top-two principal components axes, representing about 79% of the variability across the four climate projections, were extracted as a two-dimensional space, and are portrayed as the axes for the PCA plots shown in each macrogroup chapter below. This was done to simplify the representation of the climate space, while maintaining the most important information on the variables to be associated with the observed vegetation distributions.

The climate space occupied by each distinct macrogroup from the current time period was identified. This was done by using the points for each type and applying a kernel density estimator on a 2-d surface composed of the first two principal components of the climate conditions. The result is a smoothed continuous point density surface, showing the prevalence of each vegetation type across the range of sampled climatic conditions. This surface was partitioned by fitting contour lines so that they enclose a proportion of the original points from the current time period. Contours were calculated at 5% increments. For example the innermost 5% contour line encloses the 5% of pixels for the given vegetation type which are at the core of the climate space for that type, as determined by its density in the climate space. Cells further away from the dense central core, are considered to be more marginal in the vegetation type’s distribution. The outer contours are fit to enclose the 95-99% of climatically marginal points, with the last 1% of cells (beyond the 99% contour) being the most marginal. In addition, if a cell lies outside the space defined by the 99% contour of any vegetation type, it is considered to be “non-analog,” which means that it experiences climatic conditions outside of the conditions where we have a good sample in the initial time period. As a result, the status of that point is uncertain. There are occasionally a few extreme points which appear to be far outside the general distribution for the type. These may be due to misclassified vegetation

types in the source data, microclimatic conditions not captured by the climate data, historic anomalies in long-lived species, etc.

Climate exposure is the level of climate change expected in the areas where each macrogroup is dominating. This report uses the term “vegetation climate exposure analysis” to describe the following analysis which was conducted on each macrogroup. The vegetation climate exposure analysis is calculated using the mapped extent of each macrogroup. Every grid cell of each macrogroup was ranked as to its level of exposure, relative to the entire area of that macrogroup. This was done for the current time, and used to define the common climate found for each macrogroup. Once each type’s “climate envelope” was defined, we then assessed how much every grid cell changed under various future climate projections. This allowed a measure of the vegetation stress, or climate exposure. The area extent of each macrogroup that will be lost from the most commonly occurring climate conditions ( $\leq 80\%$ ) and the area that will fall into current marginal, or stressed, climate conditions ( $> 95\%$ ) or outside the current climate conditions was calculated. This approach is particularly useful for resource managers, who often are constrained to work in specified areas, and need estimates of what areas within their jurisdiction are likely to be highly stressed, and what areas are likely to be less stressed, in effect climate refuge areas.

To consider how refugial conditions from a range of stressors can inform conservation planning and management, the authors integrated metrics of refugial capacity across different domains, which are defined as social, ecological, or physical drivers, processes, or cycles that influence landscape structure, function, or composition. To persist in the California landscape, species and ecosystems may need refugia from shifting climatic conditions, including extremely hot summers and prolonged droughts, but non-climate stressors can also affect conservation outcomes. In this landscape, changes in fire frequency can be a significant stressor affecting plant community structure and persistence. Anthropogenic features that modify hydrologic flows alter the ability of watersheds to sustain functional habitats. And finally, protected areas are often designed to mitigate the impacts of anthropogenic activities; however, recreational activities may alter the refugial capacity of the protected land, affecting the ability of the landscape to sustain species and their habitats. We combined these individual metrics to assess landscape level refugial capacity.

Sites with high refugial capacity (super-refugia sites) have, on average, 30% fewer extremely warm summers, 20% fewer fire events, 10% less exposure to altered river channels and riparian areas, and 50% fewer recreational trails than the surrounding landscape. Our results suggest that super-refugia sites ( $\sim 8,200 \text{ km}^2$ ) for some natural communities are underrepresented in the existing protected area network, a finding that can inform efforts to expand protected areas.

For more information on methods for the development of these climate refugia data see:

Thorne et al. 2015

Thorne et al. 2016

Thorne et al. 2017

Thorne et al. 2020

**Data Source:** Information Center for the Environment, UC Davis

**File Name:** miroc\_85\_1039.tif; miroc\_85\_4069.tif; miroc\_85\_7099.tif

### Tier: 3

#### Data Vintage: 2016

**Metric Definition and Relevance:** This raster dataset represents habitat types (Macro Veg Type, largely equivalent to CWHR habitat classes) and their predicted exposure to climate stress across the array of predicted climate conditions (separate layers for early (2010 - 2039), mid (2040-2069), and late century (2070-2099)) for all habitat types in comparison to the baseline climate conditions. This serves as the foundation from which habitat types will be exposed to predicted changes in climate. Data are arrayed across 0 to 1 in terms of their exposure to current climate conditions. These three data layers can be used to help land managers allocate limited resources for climate-adaptive field work by providing a view of climate risk that varies across the lands they manage.

This analysis uses both the Miroc Earth System Model and the CNRM-CM5. CNRM-CM5 is an Earth system model designed to run climate simulations. It consists of several existing models designed independently and coupled through the [OASIS](#) software. Both were used under the RCP 8.5 emission scenario given that this is more likely under current emission levels.

This data layer is provided as a summary of likely exposure results. **Exposure Scores:**

- 1 = Refugia: CNRM-CM5 only (CNRM exposure values < 80%)
- 2 = Refugia: MIROC-ESM only (MIROC exposure values < 80%)
- 3 = Refugia Consensus (both models agree exposure values < 80%)
- **8 = High Exposure (both models agree exposure values >95%)**
- **9 = Very High Exposure (both models agree exposure values >99%)**

#### Data Resolution: 270m Raster

**Data Units:** 0, 1, 2, 3, 8, 9 Low values indicate higher resilience to threats. High values indicate significant exposure to climate change. -1 represents 'non analog' areas, i.e. locations that are outside the historic climate envelope of a given vegetation type.

**Creation Method:** Each dominant species is scored for its sensitivity to, and ability to adapt (adaptive capacity) to climate change. Sensitivity refers to the degree to which changes in climate are thought to directly impact different species. Adaptive capacity refers to estimates of the degree to which different species can use their life history characteristics to moderate impacts from changing climate. These two sets of scores represent the biological attributes of the dominant species in each macrogroup. We scored each of the dominant species comprising each macrogroup, according to life history characteristics defined in attribute tables of the California Manual of Vegetation, and supplemented by information found in the USDA plants database and the Jepson Interchange, a web portal for California plant taxonomy. The scores were combined to generate a single sensitivity and adaptive capacity (S&A) score.

Climate exposure is the level of climate change expected in the areas where each macrogroup is dominating. This report uses the term "vegetation climate exposure analysis" to describe the following analysis which was conducted on each macrogroup. The vegetation climate exposure analysis is calculated using the mapped extent of each macrogroup. Every grid cell of each macrogroup was ranked as to its level of exposure, relative to the entire area of that macrogroup. This was done for the current time, and used to define the common climate found for each macrogroup. Once each type's "climate envelope" was defined, we then assessed how much every grid cell changed under various future climate projections. This allowed a measure of the vegetation stress, or climate exposure. The area extent of each macrogroup that will be lost from the most commonly occurring climate conditions ( $\leq 80\%$ ) and the area that will fall into current marginal, or stressed, climate conditions ( $>95\%$ ) or outside the current climate conditions was calculated. This approach is particularly useful for resource managers, who

often are constrained to work in specified areas, and need estimates of what areas within their jurisdiction are likely to be highly stressed, and what areas are likely to be less stressed, in effect climate refuge areas.

To consider how refugial conditions from a range of stressors can inform conservation planning and management, the authors integrated metrics of refugial capacity across different domains, which are defined as social, ecological, or physical drivers, processes, or cycles that influence landscape structure, function, or composition. To persist in the California landscape, species and ecosystems may need refugia from shifting climatic conditions, including extremely hot summers and prolonged droughts, but non-climate stressors can also affect conservation outcomes. In this landscape, changes in fire frequency can be a significant stressor affecting plant community structure and persistence. Anthropogenic features that modify hydrologic flows alter the ability of watersheds to sustain functional habitats. And finally, protected areas are often designed to mitigate the impacts of anthropogenic activities; however, recreational activities may alter the refugial capacity of the protected land, affecting the ability of the landscape to sustain species and their habitats. We combined these individual metrics to assess landscape level refugial capacity.

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For more information on methods for the development of these climate refugia data see:

Thorne et al. 2015

Thorne et al. 2016

Thorne et al. 2017

Thorne et al. 2020

**Data Source:** Information Center for the Environment, UC Davis

**File Name:** combine85\_all7.tif

## SOCIAL AND CULTURAL WELL-BEING

The landscape provides a place for people to connect with nature, recreate, to maintain and improve their overall health, and an opportunity to contribute to environmental stewardship. While the elements of this pillar include public health and engagement, recreation quality, and equitable opportunities producing quantifiable, measurable and actionable metrics remains challenging. These metrics are still under development and insights into these potential metrics are appreciated.

**DESIRED OUTCOME:** The landscape provides a place for people to connect with nature, to recreate, to maintain and improve their overall health, and to contribute to environmental stewardship, and is a critical component of their identity.

EQUITABLE OPPORTUNITY

Equitable opportunity is the fair treatment and meaningful involvement of all people regardless of race, color, national origin or income regarding the development, implementation and enforcement of environmental laws, regulations, policies, and land management.

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#### HOUSING BURDEN PERCENTILE

**Tier:** 1

**Data Vintage:** 10/2021

**Metric Definition and Relevance:** Housing-Burdened Low-Income Households. Percent of households in a census tract that are both low income (making less than 80% of the HUD Area Median Family Income) and severely burdened by housing costs (paying greater than 50% of their income to housing costs). (5-year estimates, 2013-2017).

The cost and availability of housing is an important determinant of well-being. Households with lower incomes may spend a larger proportion of their income on housing. The inability of households to afford necessary non-housing goods after paying for shelter is known as housing-induced poverty. California has very high housing costs relative to much of the country, making it difficult for many to afford adequate housing. Within California, the cost of living varies significantly and is largely dependent on housing cost, availability, and demand.

Areas where low-income households may be stressed by high housing costs can be identified through the Housing and Urban Development (HUD) Comprehensive Housing Affordability Strategy (CHAS) data. We measure households earning less than 80% of HUD Area Median Family Income by county and paying greater than 50% of their income to housing costs. The indicator takes into account the regional cost of living for both homeowners and renters, and factors in the cost of utilities. CHAS data are calculated from US Census Bureau's American Community Survey (ACS).

**Data Resolution:** 30m Raster

**Data Units:** Percent

**Creation Method:** CalEnviroScreen, Version 4.0, is a science-based method for identifying impacted communities by taking into consideration pollution exposure and its effects, as well as health and socioeconomic status, at the census-tract level. CalEnviroScreen 4.0 uses the census tract as the unit of analysis. Census tract boundaries are available from the Census Bureau. CalEnviroScreen uses the Bureau's 2010 boundaries. New boundaries will be drawn by the Census Bureau as part of the 2020 Census but will not be available until 2022. OEHHA will address updates to census tract geography in CalEnviroScreen at that time. There are approximately 8,000 census tracts in California, representing a relatively fine scale of analysis. Census tracts are made up of multiple census blocks, which are the smallest geographic unit for which population data are available. Some census blocks have no people residing in them (unpopulated blocks).

The CalEnviroScreen model is based on the CalEPA working definition in that:

- The model is place-based and provides information for the entire State of California on a geographic basis. The geographic scale selected is intended to be useful for a wide range of decisions.
- The model is made up of multiple components cited in the above definition as contributors to cumulative impacts.
- The model includes two components representing Pollution Burden – Exposures and Environmental Effects

- The model includes two components representing Population Characteristics – Sensitive Populations (e.g., in terms of health status and age) and Socioeconomic Factors.

The American Community Survey (ACS) is an ongoing survey of the US population conducted by the US Census Bureau and has replaced the long form of the decennial census. Unlike the decennial census, which attempts to survey the entire population and collects a limited amount of information, the ACS releases results annually based on a sub-sample of the population and includes more detailed information on socioeconomic factors. Multiple years of data are pooled together to provide more reliable estimates for geographic areas with small population sizes. Each year, the HUD receives custom tabulations of ACS data from the US Census Bureau. These data, known as the "CHAS" data (Comprehensive Housing Affordability Strategy), demonstrate the extent of housing problems and housing needs, particularly for low-income households. The most recent results available at the census tract scale are the 5-year estimates for 2013-2017. The data are available from the HUD user website (see page 174 in the document link below:

<https://oehha.ca.gov/media/downloads/calenviroscreen/report/calenviroscreen40reportf2021.pdf>

**Data Source:** California Environmental Protection Agency, CalEnviroScreen 4.0

**File Name:** HousingBurdenPctl.tif

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#### UNEMPLOYMENT PERCENTILE

**Tier:** 1

**Data Vintage:** 10/2021

**Metric Definition and Relevance:** Percentage of the population over the age of 16 that is unemployed and eligible for the labor force. Excludes retirees, students, homemakers, institutionalized persons except prisoners, those not looking for work, and military personnel on active duty (5-year estimate, 2015-2019).

Because low socioeconomic status often goes hand-in-hand with high unemployment, the rate of unemployment is a factor commonly used in describing disadvantaged communities. On an individual level, unemployment is a source of stress, which is implicated in poor health reported by residents of such communities. Lack of employment and resulting low income often constrain people to live in neighborhoods with higher levels of pollution and environmental degradation.

**Data Resolution:** 30m Raster

**Data Units:** Percent

**Creation Method:** CalEnviroScreen, Version 4.0, is a science-based method for identifying impacted communities by taking into consideration pollution exposure and its effects, as well as health and socioeconomic status, at the census-tract level. CalEnviroScreen 4.0 uses the census tract as the unit of analysis. Census tract boundaries are available from the Census Bureau. CalEnviroScreen uses the Bureau's 2010 boundaries. New boundaries will be drawn by the Census Bureau as part of the 2020 Census but will not be available until 2022. OEHA will address updates to census tract geography in CalEnviroScreen at that time. There are approximately 8,000 census tracts in California, representing a relatively fine scale of analysis. Census tracts are made up of multiple census blocks, which are the smallest geographic unit for which population data are available. Some census blocks have no people residing in them (unpopulated blocks).

The CalEnviroScreen model is based on the CalEPA working definition in that:

- The model is place-based and provides information for the entire State of California on a geographic basis. The geographic scale selected is intended to be useful for a wide range of decisions.
- The model is made up of multiple components cited in the above definition as contributors to cumulative impacts.
- The model includes two components representing Pollution Burden – Exposures and Environmental Effects
- The model includes two components representing Population Characteristics – Sensitive Populations (e.g., in terms of health status and age) and Socioeconomic Factors.

The American Community Survey (ACS) is an ongoing survey of the US population conducted by the US Census Bureau. Unlike the decennial census, which attempts to survey the entire population and collects a limited amount of information, the ACS releases results annually based on a sub-sample of the population and includes more detailed information on socioeconomic factors such as unemployment. Multiple years of data are pooled together to provide more reliable estimates for geographic areas with small population sizes. The most recent results available at the census tract level are the 5-year estimates for 2015-2019. The data are made available using the U.S. Census data download website.

**Data Source:** California Environmental Protection Agency, CalEnviroScreen 4.0

**File Name:** Unemployment\_Pctl.tif

#### POVERTY PERCENTILE

**Tier:** 1

**Data Vintage:** 10/2021

**Metric Definition and Relevance:** Percent of population living below two times the federal poverty level. The U.S. Census Bureau determines the federal poverty level each year. The poverty level is based on the size of the household and the age of family members. If a person or family's total income before taxes is less than the poverty level, the person or family are considered in poverty. Many studies have found that people living in poverty are more likely than others to become ill from pollution.

**Data Resolution:** 30m Raster

**Data Units:** percentile

**Creation Method:** CalEnviroScreen, Version 4.0, is a science-based method for identifying impacted communities by taking into consideration pollution exposure and its effects, as well as health and socioeconomic status, at the census-tract level. CalEnviroScreen 4.0 uses the census tract as the unit of analysis. Census tract boundaries are available from the Census Bureau. CalEnviroScreen uses the Bureau's 2010 boundaries. New boundaries will be drawn by the Census Bureau as part of the 2020 Census but will not be available until after 2022. OEHHA will address updates to census tract geography in CalEnviroScreen at that time. There are approximately 8,000 census tracts in California, representing a relatively fine scale of analysis. Census tracts are made up of multiple census blocks, which are the smallest geographic unit for which population data are available. Some census blocks have no people residing in them (unpopulated blocks).

The poverty percentile is derived from

- The 2015-2019 American Community Survey, a dataset containing the number of individuals below 200 percent of the federal poverty level was downloaded by census tracts for the state of California.

- The number of individuals below 200% of the poverty level was divided by the total population for whom poverty status was determined.
- Unlike the US Census, ACS estimates come from a sample of the population and may be unreliable if they are based on a small sample or population size. The standard error (SE) and relative standard error (RSE) were used to evaluate the reliability of each estimate.
- The SE was calculated for each census tract using the formula for approximating the SE of proportions provided by the ACS (American Community Survey Office, 2013, pg. 13, equation 4). CalEnviroScreen 4.0 189 When this approximation could not be used, the formula for approximating the SE of ratios (equation 3) was used instead.
- The RSE is calculated by dividing a tract's SE by its estimate of the percentage of the population living below twice the federal poverty level, and taking the absolute value of the result.
- Census tract estimates that met either of the following criteria were considered reliable and included in the analysis:
  - RSE less than 50 (meaning the SE was less than half of the estimate) OR
  - SE was less than the mean SE of all California census tract estimates for poverty.
- Census tracts with unreliable estimates received no score for the indicator (null). The indicator was not factored into that tract's overall CalEnviroScreen score.
- Census tracts that met the inclusion criteria were sorted and assigned percentiles based on their position in the distribution.

**Data Source:** California Environmental Protection Agency, CalEnviroScreen 4.0

**File Name:** Poverty\_PctI.tif

#### TRIBAL LAND DESIGNATIONS

**Tier:** 1

**Data Vintage:** 2022

**Metric Definition and Relevance:** Lands under the control of federally recognized Tribes. These boundaries have been clipped to the boundary of California so that only the portions of each Tribal area that fall within California are included. For purposes of this designation, a Tribe may establish that a particular area of land is under its control even if not represented as such on CalEPA's DAC map and therefore should be considered a DAC by requesting a consultation with the CalEPA Deputy Secretary for Environmental Justice, Tribal Affairs and Border Relations at TribalAffairs@calepa.ca.gov

**Data Resolution:** 30m Raster

**Data Units:** Thematic

Definition of field name COMPTYPE (Component type):

- "R" is Federally recognized American Indian Reservations (AIRs)
- "T" is Off-Reservation Trust Lands (ORTL)

**Creation Method:** CalEnviroScreen, Version 4.0, is a science-based method for identifying impacted communities by taking into consideration pollution exposure and its effects, as well as health and socioeconomic status, at the census-tract level. CalEnviroScreen 4.0 uses the census tract as the unit of analysis. Census tract boundaries are available from the Census Bureau. CalEnviroScreen uses the Bureau's 2010 boundaries. New boundaries will be drawn by the Census Bureau as part of the 2020 Census but will not be available until 2022. OEHA will address

updates to census tract geography in CalEnviroScreen at that time. There are approximately 8,000 census tracts in California, representing a relatively fine scale of analysis. Census tracts are made up of multiple census blocks, which are the smallest geographic unit for which population data are available. Some census blocks have no people residing in them (unpopulated blocks).

The CalEnviroScreen model is based on the CalEPA working definition in that:

- The model is place-based and provides information for the entire State of California on a geographic basis. The geographic scale selected is intended to be useful for a wide range of decisions.
- The model is made up of multiple components cited in the above definition as contributors to cumulative impacts.
- The model includes two components representing Pollution Burden – Exposures and Environmental Effects
- The model includes two components representing Population Characteristics – Sensitive Populations (e.g., in terms of health status and age) and Socioeconomic Factors.

The American Community Survey (ACS) is an ongoing survey of the US population conducted by the US Census Bureau and has replaced the long form of the decennial census. Unlike the decennial census, which attempts to survey the entire population and collects a limited amount of information, the ACS releases results annually based on a sub-sample of the population and includes more detailed information on socioeconomic factors. Multiple years of data are pooled together to provide more reliable estimates for geographic areas with small population sizes. Each year, the HUD receives custom tabulations of ACS data from the US Census Bureau. These data, known as the "CHAS" data (Comprehensive Housing Affordability Strategy), demonstrate the extent of housing problems and housing needs, particularly for low-income households. The most recent results available at the census tract scale are the 5-year estimates for 2013-2017. The data are available from the HUD user website (see page 174 in the document link below:

<https://oehha.ca.gov/media/downloads/calenviroscreen/report/calenviroscreen40reportf2021.pdf>

**Data Source:** California Environmental Protection Agency, CalEnviroScreen 4.0

**File Name:** SB535tribalboundaries2022.tif

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#### TRAIL DENSITY

**Tier:** 2

**Data Vintage:** 2023

**Metric Definition and Relevance:** A raster dataset representing density of trails in the Southern California Region.

**Data Resolution:** 30m Raster

**Data Units:** km/km<sup>2</sup> of trails

**Creation Method:** This dataset represents the density of trails across the landscape, measured as km/km<sup>2</sup> of trails. The information was derived from the OpenStreetMap data and includes footways, cycleways, pedestrian paths, service roads, paths, tracks, and unclassified features. Based on local knowledge, inclusion of these different categories was required to capture the majority of popular hiking destinations including authorized and unauthorized routes.

**Data Source:** San Diego State University CWC Project Team

**File Name:** Trail\_Density.tif

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AMERICAN INDIAN OR ALASKA NATIVE RACE ALONE AND MULTIRACE POPULATION CONCENTRATION

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Relative concentration of the Southern California region’s American Indian population. The variable AIAN\_ALN\_AND\_MULTIRACEAIANALN includes BOTH individuals who select American Indian or Alaska Native as their sole racial identity (they *only* identify as American Indian), AND individuals who select American Indian / Alaska Native as one of two or more racial identities (they *partly* identify as American Indian) in response to the Census questionnaire. IMPORTANT: this self reported ancestry and Tribal membership are distinct identities and one does not automatically imply the other. These data should not be interpreted as a distribution of “Tribal people.” Numerous Rancherias in the Southern California region account for the wide distribution of very to extremely high concentrations of American Indians.

“Relative concentration” is a measure that compares the proportion of population within each Census block group data unit that identify as American Indian / Alaska Native alone to the proportion of all people that live within the 13,312 block groups in the Southern California RRK region that identify as American Indian / Alaska native alone. Example: if 5.2% of people in a block group identify as AIANALN, the block group has twice the proportion of AIANALN individuals compared to the Southern California RRK region (2.6%), and more than three times the proportion compared to the entire state of California (1.6%). If the local proportion is twice the regional proportion, then AIANALN individuals are highly concentrated locally.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total
- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total
- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total
- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total

- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total
- Class Code 6: *Very high*. The subject population concentration is very high; the local proportion of the subject population variable roughly 2 to 3 times that of the corresponding proportion in the Southern California region population in total
- Class Code 7: *Extremely high*. The subject population concentration is very extremely high; the local proportion of the subject population variable is at least 3 times that of the corresponding proportion in the Southern California region population in total (the upper limit is determined by natural breaks, if exceptional outliers are present, but is typically over 6 times (600%))
- Class Code 8: *Exceptionally high*. The subject population concentration is so high that it is an exceptional outlier; the local proportion of the subject population variable is typically greater than 6 or 7 times that of the corresponding proportion in the region
- Class Code 99: *Unclassifiable*. The 90% confidence interval for the estimate is wide enough to cause the values to span four or more classes. In these cases, it is impossible to say with any reasonable certainty whether the concentration is "low" or "high."

**Creation Method:** Data reporting units are Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g., Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Note; 1) The pixels attributed with a categorical data unit describing the relative concentration of AIANALN population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with the Southern California RRR boundary. The resulting 13,312 block group features are modified by first erasing from the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to

the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307

Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 Decennial Census Redistricting File (PL 94-171).

Racial identity data are reported in Table P1 of the PL 94-171 release. Population counts were obtained via the Data.Census.Gov web portal and joined to the Census Bureau's TIGER/line feature classes for block groups (see reporting units above).

**File Name:** AmerIndianAndMultiRace\_2020.tif

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#### HISPANIC/LATINO POPULATION CONCENTRATION

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Relative concentration of the Southern California region's Hispanic/Latino population. The variable HISPANIC records all individuals who select Hispanic or Latino in response to the Census questionnaire, regardless of their response to the racial identity question.

"Relative concentration" is a measure that compares the proportion of population within each Census block group data unit that identify as American Indian / Alaska Native alone to the proportion of all people that live within the 13,312 block groups in the Southern California RRK region that identify as American Indian / Alaska native alone. Example: if 5.2% of people in a block group identify as HISPANIC, the block group has twice the proportion of HISPANIC individuals compared to the Southern California RRK region (2.6%), and more than three times the proportion compared to the entire state of California (1.6%). If the local proportion is twice the regional proportion, then HISPANIC individuals are highly concentrated locally.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total

- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total
- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total
- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total
- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total
- Class Code 6: *Very high*. The subject population concentration is very high; the local proportion of the subject population variable roughly 2 to 3 times that of the corresponding proportion in the Southern California region population in total

**Creation Method:** Data reporting units are Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g., Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Note: 1) The pixels attributed with a categorical data unit describing the relative concentration of HISPANIC population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with the Southern California RRK boundary. The resulting 13,312 block group features are modified by first erasing from

the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307

Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

Data were derived from the 2020 Census Total population for the block group from the redistricting file (PL 94-171) of the 2020 Census, released summer 2021. The raw data were obtained directly from the Census Bureau data set table named in "Origin"; all data sets downloaded from census.data.gov and joined to TIGER Census block group features. There are 13,312 Census block groups within or intersecting the Southern California RRK region boundary.

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 Decennial Census Redistricting File (PL 94-171).

Racial identity data are reported in Table P1 of the PL 94-171 release. Population counts were obtained via the Data.Census.Gov web portal and joined to the Census Bureau's TIGER/line feature classes for block groups (see reporting units above).

**File Name:** Hispanic\_2020.tif

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#### BLACK/AFRICAN AMERICAN POPULATION CONCENTRATION

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Relative concentration of the Southern California region's Black/African American population. The variable BLACKALN records all individuals who select black or African American as their SOLE racial identity in response to the Census questionnaire, regardless of their response to the Hispanic ethnicity question. Both Hispanic and non-Hispanic in the Census questionnaire are potentially associated with black race alone.

"Relative concentration" is a measure that compares the proportion of population within each Census block group data unit that identify as Black/African American alone to the proportion of all people that live within the 13,312 block groups in the Southern California RRK region that identify as Black/African American alone. Example: if 5.2% of people in a block group identify as BLACKALN, the block group has twice the proportion of BLACKALN individuals compared to the Southern California RRK region (2.6%), and more than three times the proportion compared to the entire state of California (1.6%). If the local proportion is twice the regional proportion, then BLACKALN individuals are highly concentrated locally.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total
- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total
- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total
- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total
- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total
- Class Code 6: *Very high*. The subject population concentration is very high; the local proportion of the subject population variable roughly 2 to 3 times that of the corresponding proportion in the Southern California region population in total
- Class Code 7: *Extremely high*. The subject population concentration is very extremely high; the local proportion of the subject population variable is at least 3 times that of the corresponding proportion in the Southern California region population in total (the upper limit is determined by natural breaks, if exceptional outliers are present, but is typically over 6 times (600%))
- Class Code 8: *Exceptionally high*. The subject population concentration is so high that it is an exceptional outlier; the local proportion of the subject population variable is typically greater than 6 or 7 times that of the corresponding proportion in the region
- Class Code 99: *Unclassifiable*. The 90% confidence interval for the estimate is wide enough to cause the values to span four or more classes. In these cases, it is impossible to say with any reasonable certainty whether the concentration is "low" or "high."

**Creation Method:** Data reporting units are Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g.,

Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Note: 1) The pixels attributed with a categorical data unit describing the relative concentration of BLACKALN population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with the Southern California RRK boundary. The resulting 13,312 block group features are modified by first erasing from the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307

Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

Data were derived from the 2020 Census Total population for the block group from the redistricting file (PL 94-171) of the 2020 Census, released summer 2021. The raw data were obtained directly from the Census Bureau data set table named in "Origin"; all data sets downloaded from census.data.gov and joined to TIGER Census block group features. There are 13,312 Census block groups within or intersecting the Southern California RRK region boundary.

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 Decennial Census Redistricting File (PL 94-171).

Racial identity data are reported in Table P1 of the PL 94-171 release. Population counts were obtained via the Data.Census.Gov web portal and joined to the Census Bureau's TIGER/line feature classes for block groups (see reporting units above).

**File Name:** Black\_2020.tif

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HISPANIC AND/OR BLACK, INDIGENOUS OR PERSON OF COLOR (HSPBIPOC)

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Relative concentration of the Southern California region's Black/African American population. The variable HSPBIPOC is equivalent to all individuals who select a combination of racial and ethnic identity in response to the Census questionnaire EXCEPT those who select "not Hispanic" for the ethnic identity question, and "white race alone" for the racial identity question. This is the most encompassing possible definition of racial and ethnic identities that may be associated with historic underservice by agencies, or be more likely to express environmental justice concerns (as compared to predominantly non-Hispanic white communities). Until 2021, federal agency guidance for considering environmental justice impacts of proposed actions focused on how the actions affected "racial or ethnic minorities." "Racial minority" is an increasingly meaningless concept in the USA, and particularly so in California, where only about 3/8 of the state's population identifies as non-Hispanic and white race alone - a clear majority of Californians identify as Hispanic and/or not white. Because many federal and state map screening tools continue to rely on "minority population" as an indicator for flagging potentially vulnerable / disadvantaged/ underserved populations, our analysis includes the variable HSPBIPOC which is effectively "all minority" population according to the now outdated federal environmental justice direction. A more meaningful analysis for the potential impact of forest management actions on specific populations considers racial or ethnic populations individually: e.g., all people identifying as Hispanic regardless of race; all people identifying as American Indian, regardless of Hispanic ethnicity; etc.

"Relative concentration" is a measure that compares the proportion of population within each Census block group data unit that identify as HSPBIPOC alone to the proportion of all people that live within the 13,312 block groups in the Southern California RRK region that identify as HSPBIPOC alone. Example: if 5.2% of people in a block group identify as HSPBIPOC, the block group has twice the proportion of HSPBIPOC individuals compared to the Southern California RRK region (2.6%), and more than three times the proportion compared to the entire state of California (1.6%). If the local proportion is twice the regional proportion, then HSPBIPOC individuals are highly concentrated locally.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total
- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total

- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total
- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total
- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total

**Creation Method:** Data reporting units are Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g., Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Note; 1) The pixels attributed with a categorical data unit describing the relative concentration of HSPBIPOC population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with the Southern California RRK boundary. The resulting 13,312 block group features are modified by first erasing from the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307

Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

Data were derived from the 2020 Census Total population for the block group from the redistricting file (PL 94-171) of the 2020 Census, released summer 2021. The raw data were obtained directly from the Census Bureau data set table named in "Origin"; all data sets downloaded from census.data.gov and joined to TIGER Census block group features. There are 13,312 Census block groups within or intersecting the Southern California RRK region boundary.

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 Decennial Census Redistricting File (PL 94-171).

Racial identity data are reported in Table P1 of the PL 94-171 release. Population counts were obtained via the Data.Census.Gov web portal and joined to the Census Bureau's TIGER/line feature classes for block groups (see reporting units above)

**File Name:** HSPBIPOC\_2020.tif

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#### ASIAN POPULATION CONCENTRATION

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** Relative concentration of the Southern California region's Asian American population. The variable ASIANALN records all individuals who select Asian as their SOLE racial identity in response to the Census questionnaire, regardless of their response to the Hispanic ethnicity question. Both Hispanic and non-Hispanic in the Census questionnaire are potentially associated with the Asian race alone.

"Relative concentration" is a measure that compares the proportion of population within each Census block group data unit that identify as ASIANALN alone to the proportion of all people that live within the 13,312 block groups in the Southern California RRK region that identify as ASIANALN alone. Example: if 5.2% of people in a block group identify as HSPBIPOC, the block group has twice the proportion of ASIANALN individuals compared to the Southern California RRK region (2.6%), and more than three times the proportion compared to the entire state of California (1.6%). If the local proportion is twice the regional proportion, then ASIANALN individuals are highly concentrated locally.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total

- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total
- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total
- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total
- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total
- Class Code 6: *Very high*. The subject population concentration is very high; the local proportion of the subject population variable roughly 2 to 3 times that of the corresponding proportion in the Southern California region population in total
- Class Code 7: *Extremely high*. The subject population concentration is very extremely high; the local proportion of the subject population variable is at least 3 times that of the corresponding proportion in the Southern California region population in total (the upper limit is determined by natural breaks, if exceptional outliers are present, but is typically over 6 times (600%))
- Class Code 99: *Unclassifiable*. The 90% confidence interval for the estimate is wide enough to cause the values to span four or more classes. In these cases, it is impossible to say with any reasonable certainty whether the concentration is "low" or "high."

**Creation Method:** Data reporting units are Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g., Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and

Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Note: 1) The pixels attributed with a categorical data unit describing the relative concentration of ASIANALN population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with the Southern California RRK boundary. The resulting 13,312 block group features are modified by first erasing from the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307

Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

Data were derived from the 2020 Census Total population for the block group from the redistricting file (PL 94-171) of the 2020 Census, released summer 2021. The raw data were obtained directly from the Census Bureau data set table named in "Origin"; all data sets downloaded from [census.data.gov](https://www.census.gov) and joined to TIGER Census block group features. There are 13,312 Census block groups within or intersecting the Southern California RRK region boundary.

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 Decennial Census Redistricting File (PL 94-171).

Racial identity data are reported in Table P1 of the PL 94-171 release. Population counts were obtained via the [Data.Census.Gov](https://www.census.gov) web portal and joined to the Census Bureau's TIGER/line feature classes for block groups (see reporting units above)

**File Name:** Asian\_2020.tif

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MULTI-RACE, EXCEPT PART-AMERICAN INDIAN POPULATION CONCENTRATION

**Tier:** 2

**Data Vintage:** 2020

**Metric Definition and Relevance:** The Relative concentration of the Southern California region's population that identifies as "Multiracial", EXCEPT those with part-American Indian identity, in response to the Census

questionnaire. "Relative concentration" is a measure that compares the proportion of population within each Census block group data unit that identifies as Multi-racial to the proportion of all people that live within the 13,312 census block groups in the Southern California RRA region. People with part-American Indian identity are not included here but are included in the American Indian or Alaska Native Race Alone and Multirace Population, described above.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total
- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total
- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total
- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total
- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total
- Class Code 6: *Very high*. The subject population concentration is very high; the local proportion of the subject population variable roughly 2 to 3 times that of the corresponding proportion in the Southern California region population in total
- Class Code 7: *Extremely high*. The subject population concentration is very extremely high; the local proportion of the subject population variable is at least 3 times that of the corresponding proportion in the Southern California region population in total (the upper limit is determined by natural breaks, if exceptional outliers are present, but is typically over 6 times (600%))
- Class Code 99: *Unclassifiable*. The 90% confidence interval for the estimate is wide enough to cause the values to span four or more classes. In these cases, it is impossible to say with any reasonable certainty whether the concentration is "low" or "high."

**Creation Method:** Data reporting units are Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g., Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Note: 1) The pixels attributed with a categorical data unit describing the relative concentration of AIAN\_ALN\_AND\_MULTIRACE\_2020 population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with the Southern California RRK boundary. The resulting 13,312 block group features are modified by first erasing from the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307

Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

Data were derived from the 2020 Census Total population for the block group from the redistricting file (PL 94-171) of the 2020 Census, released summer 2021. The raw data were obtained directly from the Census Bureau data set table named in "Origin"; all data sets downloaded from census.data.gov and joined to TIGER Census block group features. There are 13,312 Census block groups within or intersecting the Southern California RRK region boundary.

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 Decennial Census Redistricting File (PL 94-171).

Racial identity data are reported in Table P1 of the PL 94-171 release. Population counts were obtained via the Data.Census.Gov web portal and joined to the Census Bureau’s TIGER/line feature classes for block groups (see reporting units above)

**File Name:** MultiRaceNotAmerInd\_2020.tif

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LOW INCOME POPULATION CONCENTRATION

**Tier:** 2

**Data Vintage:**2020

**Metric Definition and Relevance:** Relative concentration of the estimated number of people in the Southern California region that live in a household defined as “low income.” There are multiple ways to define low income. These data apply the most common standard: low income population consists of all members of households that collectively have income less than twice the federal poverty threshold that applies to their household type. Household type refers to the household’s resident composition: the number of independent adults plus dependents that can be of any age, from children to elderly. For example, a household with four people – one working adult parent and three dependent children – has a different poverty threshold than a household comprised of four unrelated independent adults.

Due to high estimate uncertainty for many block group estimates of the number of people living in low income households, some records cannot be reliably assigned a class and class code comparable to those assigned to race/ethnicity data from the decennial Census.

“Relative concentration” is a measure that compares the proportion of population within each Census block group data unit to the proportion of all people that live within the 13,312 block groups in the Southern California RRK region. See the “Data Units” description below for how these relative concentrations are broken into categories in this “low income” metric.

**Data Resolution:** 30m Raster

**Data Units:** Categorical

- Class Code 0: *Zero or nearly zero*. The variable is absent (observed value = 0) or is very low; the local proportion of the subject population variable is 10% or less than the same proportion in the Southern California region population in total
- Class Code 1: *Low*. The subject population concentration is low; the local proportion of the subject population variable is between roughly 10% and 50% of the corresponding proportion in the Southern California region population in total
- Class Code 2: *Somewhat low*. The subject population concentration is somewhat low; the local proportion of the subject population variable is between roughly 50% and 85% of the corresponding proportion in the Southern California region population in total

- Class Code 3: *Proportionate*. The subject population concentration is roughly proportionate to the corresponding proportion in the Southern California region population in total - from about 85% to 115% of the regional proportion
- Class Code 4: *Somewhat high*. The subject population concentration is somewhat high; the local proportion of the subject population variable is between roughly 115% and 150% of the corresponding proportion in the Southern California region population in total
- Class Code 5: *High*. The subject population concentration is high; the local proportion of the subject population variable is between roughly 150% and 200% of the corresponding proportion in the Southern California region population in total
- Class Code 6: *Very high*. The subject population concentration is very high; the local proportion of the subject population variable roughly 2 to 3 times that of the corresponding proportion in the Southern California region population in total
- Class Code 7: *Extremely high*. The subject population concentration is very extremely high; the local proportion of the subject population variable is at least 3 times that of the corresponding proportion in the Southern California region population in total (the upper limit is determined by natural breaks if exceptional outliers are present, but is typically over 6 times (600%))
- Class Code 99: *Unclassifiable*. The 90% confidence interval for the estimate is wide enough to cause the values to span four or more classes. In these cases, it is impossible to say with any reasonable certainty whether the concentration is "low" or "high."

**Creation Method:** Data are reported in Census block groups. Standard block groups are clusters of Census blocks within the same census tract that have the same first digit of their 4-character census block number (e.g., Blocks 3001, 3002, 3003 to 3999 in census tract 1210.02 belong to block group 3). Block groups delineated for the 2020 Census generally contain 600 to 3,000 people.

Census blocks are statistical areas bounded on all sides by visible features (e.g., streets, roads, streams, and railroad tracks), and by non-visible boundaries (e.g., city, town, township, county limits, and short line-of-sight extensions of streets and roads). Census blocks in suburban and rural areas may be large, irregular, and bounded by a variety of features (e.g., roads, streams, and/or transmission line rights-of-way). In remote areas, census blocks may encompass hundreds of square miles. Census blocks cover all territory in the United States, Puerto Rico, and the Island areas. Blocks do not cross the boundaries of any entity for which the Census Bureau tabulates data. See note 1.

Data describing concentrations of population characteristics that are potentially related to environmental justice issues were provided to CWI through a collaboration with the USDA Forest Service, Geospatial Technology and Applications Center. The concentration methodology was created by GTAC for social science analysis applications within the Forest Service; it is based on research published in 2018 and 2020 (See Note 2). Data were compiled and prepared for incorporating in the regional resource kits by Mark Adams, Geographer, USFS-GTAC. For more information, contact: [mark.adams1@usda.gov](mailto:mark.adams1@usda.gov).

Notes: The pixels attributed with a categorical data unit describing the relative concentration of LOW\_INCOME population are derived from a vector polygon feature that has been modified as follows: Census block groups from the Census Bureau's TIGER/Line geodatabase features for 2021 are selected based on their spatial intersection with

the Southern California RRK boundary. The resulting 13,312 block group features are modified by first erasing from the feature the area of all constituent Census blocks which have neither housing nor population recorded in the PL-94171 Redistricting dataset for 2020. In a second step, areas of federal and state public lands on which housing by definition is not located are erased from the interim feature. The result is a block group feature that depicts to the maximum practicable extent the areas within the block group where people that are represented by the Census Bureau's Census count could actually be residing. It is this modified block group feature that has been rasterized to match the 30m pixel grid that all biophysical datasets are reported in.

References for the concentration levels analysis:

- Adams, Mark D. O. and S. Charnley. 2020. The Environmental Justice Implications of Managing Hazardous Fuels on Federal Forest Lands, *Annals of the American Association of Geographers*, 110:6, 1907-1935, DOI: 10.1080/24694452.2020.1727307
- Adams, Mark D. O. and S. Charnley. 2018. Environmental justice and U.S. Forest Service hazardous fuels reduction: A spatial method for impact assessment of federal resource management actions. <https://doi.org/10.1016/j.apgeog.2017.12.014>

**Data Source:** U.S. Department of Commerce, Census Bureau, 2020 American Community Survey 5-Year Survey Estimates.

Data estimating household income as a percent of the applicable federal poverty threshold are reported in Table C17002 of the 2016-2020 ACS 5-year data. Estimates of population living in low income households were obtained via the Data.Census.Gov web portal and joined to the Census Bureau's TIGER/line feature classes for block groups (see reporting units below). Table C17002 provides estimates and error margins for total population living in households with income, and population by ratio of income to applicable poverty: 50% of poverty, 50-99%, etc. Additional calculations are performed to generate an estimate for all people in households with income less than 200% of applicable poverty.

FMI: <https://www.census.gov/newsroom/press-releases/2022/acs-5-year-estimates.html>

**File Name:** LowIncome\_2020.tif

## WATER SECURITY

Forests serve as natural water collection, storage, filtration, and delivery systems as water flows from forests into rivers providing critical aquatic and wetland habitat, while also supplying water for drinking and agriculture. From a more mechanistic perspective, the energy and water balance of forest ecosystems are fundamentally linked. Water is essential to photosynthesis and the latent energy exchange of transpiration is a major driver of water loss. In short, the fate of forests directly influences the quantity and quality of California's freshwater supply.

**DESIRED OUTCOME:** Watersheds provide a reliable supply of clean water despite wide swings in annual precipitation, droughts, flooding, and wildfire.

QUANTITY

Understanding the interaction between water supply and ecosystem demand informs both the extent of moisture stress and the amount of water available for storage.

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#### ACTUAL EVAPOTRANSPIRATION TO PRECIPITATION FRACTION DURING DROUGHT

**Tier:** 1

**Data Vintage:** 09/2021

**Metric Definition and Relevance:** Plants respond to conditions in their immediate vicinity. Thus, to understand the vegetative moisture stress during drought, it is important to measure the local moisture balance. The actual evapotranspiration fraction (AETF) provides such a measure. Specifically, it indicates whether a location is expected to experience local drying during a drought, or whether the location receives sufficient precipitation that it will remain moist even during an extended drought. An extended drought is defined by a 48-month period where the Standardized Precipitation Index (SPI, NCAR 2022) is two standard deviations below the long-term mean (SPI-48 = negative 2). Such a drought is expected approximately once every 50 years in the Southern California Region.

The AETF ranges from 0 to > 1; a low value indicates a wetter location during drought and a high value indicates a drier location. Locations <1 would be expected to generate runoff, even during a significant drought (SPI-48 drought = negative 2.0), and would be expected to continue generating runoff. Locations > 1 would be expected to desiccate the soil during drought, with negligible runoff, and increasing vegetation drought stress. AET/P does not account for lateral water inflow, either as runoff or irrigation.

**Data Resolution:** 30m Raster

**Data Units:** Dimensionless fraction (AET in mm/P in mm).

**Creation Method:** Calculated as the ratio of actual evapotranspiration (AET) during 2021 Water Year (WY) and precipitation that would be expected for each pixel under a significant drought (SPI-48 drought = negative 2.0). AET is calculated based on Landsat observations and eddy covariance, along with information on local monthly irradiance that accounts for Top of Atmosphere and topographic effects. The AET calculated for 2021 is then divided by the precipitation that would be expected for each pixel under a significant drought (SPI-48 drought = negative 2.0). This quantity of precipitation is calculated for each pixel based on local, down-scaled PRISM data for 1991-2020. This fraction provides a measure of the local water balance during drought, with the higher values indicating a drier location. See <https://doi.org/10.1029/2012JG002027> and <https://doi.org/10.1073/pnas.1319316111> for further information.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** WaterFlux\_AETFrac\_SPI-2\_2021.tif

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#### PRECIPITATION MINUS ACTUAL EVAPOTRANSPIRATION DURING AVERAGE CONDITIONS

**Tier:** 1

**Data Vintage:** 09/2021

**Metric Definition and Relevance:** Runoff is a measure of the water available for storage. It is determined by both the water supply and the demand of the existing vegetation. Annual mean runoff measures the “average” vegetative demand and thus provides a comparative index on the potential available runoff. Specifically, Annual

Mean Runoff is the expected surplus water that would discharge to surface or groundwater flows during a series of years with average precipitation. Larger values indicate more runoff under mean conditions.

**Data Resolution:** 30m Raster

**Data Units:** mm/y

**Creation Method:** The Center for Ecosystem Climate Solutions at UC Irvine (CECS) is working with the State and Federal governments in developing scientifically rigorous, stakeholder-informed methods that have produced tailored, integrated data for land management decision makers. The CECS DataEngine model tracks monthly water balance from 1986 to 2021. The Annual Mean Runoff layer is calculated using this CECS DataEngine model logic forced with a series of 4 years that each received precipitation according to the timing and magnitude of the 30-year climate Normal Precipitation (SPI = 0 by definition).

The model water inputs are determined from downscaled PRISM gridded datasets (<https://prism.oregonstate.edu/>). In the case of the Annual Mean Runoff, this reflects the monthly 30 year Normal for each pixel calculated for 1991-2020. Actual evapotranspiration (AET) is calculated from Landsat observations and eddy covariance during 2021, along with information on local monthly irradiance that accounts for Top of Atmosphere (TOA) and topographic effects, as well as monthly temperature and drought stress. Precipitation Minus Actual Evapotranspiration is calculated as the difference; it provides an excellent measure of the long-term runoff from upland pixels. Areas with a higher P-ET produce greater runoff, and areas with a low P-ET tend to produce little or no runoff during average or dry years. See <https://doi.org/10.1029/2012JG002027> and <https://doi.org/10.1073/pnas.1319316111> for further information.

**Data Source:** CECS; <https://california-ecosystem-climate.solutions/>

**File Name:** WaterFlux\_Runoff\_SPI0\_2021.tif

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#### DROUGHT SENSITIVITY

**Tier:** 3

**Data Vintage:** 2018

**Metric Definition and Relevance:** This layer represents an estimation of the probability that drought will substantially impact post-fire shrub recovery, potentially leading to vegetation type conversion to invasive grasses and forbs. This type conversion may increase the risk of fire ignition and fire spread.

**Data Resolution:** 30m Raster

**Data Units:** Relative index, low to high

**Creation Method:** These data are based on estimates of change in canopy cover over the period 1984-2018 from satellite data, and includes some areas that have already degraded, but mostly areas predicted to change in future drought episodes.

Specifically, this data set portrays estimates of the probability that absolute shrub cover will decline substantially (up to 30 percent) as a result of failed vegetation recovery consequential of drought. This drought sensitivity index is based on estimates of change in shrub cover between the periods 1984–1989 and 2014–2018, which were derived from June-solstice Landsat Normalized Difference Vegetation Index (NDVI) image time series (source: [espa.cr.usgs.gov](http://espa.cr.usgs.gov)). These 30-m spatial resolution estimates of change in fractional cover were sampled randomly based on 491 points located within chaparral areas that burned only once during the period 1984–2018, and

therefore were not subject to the potential impacts of repeated (short-interval) fires upon seed production or resprouting vigor.

**Credits:** Emanuel Storey, Ph. D; San Diego State University's Connecting Wildlands and Communities Project Team

**Data Source:** San Diego State University CWC Project Team

**File Name:** Drought\_Sensitivity.tif

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#### CHANGE IN AVERAGE CLIMATIC WATER DEFICIT

**Tier:** 3

**Data Vintage:** 2022

**Metric Definition and Relevance:** Change in Average Annual Climatic Water Deficit - Near Future - Drier. This raster dataset represents a projection of the difference in the mean annual climatic water deficit between the baseline period (1950-1980), and a near future period ( 2030-2059) under the MIROC (Drier) scenario of climate change.

**Data Resolution:** 30m Raster

**Data Units:** mm

**Creation Method:** The climate variable in this layer is the change in average annual accumulated climatic water deficit relative to a baseline period. Climatic water deficit (CWD) measures the evaporative demand that exceeds available water, reflecting the impacts of available water or drought stress on plants. Higher CWD indicates a higher level of drought stress.

Baseline annual accumulated CWD was averaged between 1950-1980

Near Future shows the average change in CWD as modeled for 2030-2059

Drier future projections were generated using the MIROC5 climate model under the business as usual RCP 8.5.

**Data Source:** San Diego State University; Connecting Wildlands and Communities Project Team

**File Name:** ChangeInAverageAnnualCWD-NearFutureDrier.tif

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#### GROUNDWATER BASIN BOUNDARIES

**Tier:** 3

**Data Vintage:** 02/2022

**Metric Definition and Relevance:** This dataset shows the boundaries of groundwater basins and subbasins as defined by the California Department of Water Resources as last modified by the Basin Boundary Emergency Regulation adopted on October 21, 2015 and subsequent modifications requested through the Basin Boundary Modification Request Process.

**Data Resolution:** 30m raster

**Data Units:** Binary

**Creation Method:** Groundwater basins are represented as polygon features and designated on the basis of geological and hydrological conditions - usually the occurrence of alluvial or unconsolidated deposits. When practical, large basins are also subdivided by political boundaries, as in the Central Valley. Basins are named and numbered per the convention of the Department of Water Resources.

These boundaries have been converted from a polygon vector to a 30 meter raster by the RRK team and clipped to the Southern California RRK.

**Data Source:** California Department of Water Resources

[https://gis.water.ca.gov/arcgis/rest/services/Geoscientific/i08\\_B118\\_CA\\_GroundwaterBasins/FeatureServer](https://gis.water.ca.gov/arcgis/rest/services/Geoscientific/i08_B118_CA_GroundwaterBasins/FeatureServer)

**File Name:** i08\_B118\_CA\_GroundwaterBasins.tif

## QUALITY

Understanding the interaction between water supply and ecosystem demand informs both the extent of moisture stress and the amount of water available for storage.

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## PERCENT IMPERVIOUS SURFACE

**Tier:** 1

**Data Vintage:** 2019

**Metric Definition and Relevance:** This National Land Cover Database (NLCD) product represents urban impervious surfaces as a percentage of developed surface over every 30-meter pixel of California, extracted from a nationwide layer. The definition of impervious means water does not seep into the ground, it runs off into storm sewers and then into local creeks. Examples of impervious surfaces include highways, streets and pavement, driveways, and house roofs. The relevance of impervious surfaces is the higher the proportion of impervious surfaces the more likely flooding can occur.

**Data Resolution:** 30m Raster

**Data Units:** Percent Imperviousness

**Creation Method:** The NLCD 2019 design aims to provide consistent and robust methodologies for production of a multi-temporal land cover and land cover change database from 2001 to 2019 at 2–3-year intervals. Comprehensive research was conducted and resulted in developed strategies for NLCD 2019: continued integration between impervious surface and all landcover products with impervious surface being directly mapped as developed classes in the landcover, a streamlined compositing process for assembling and preprocessing based on Landsat imagery and geospatial ancillary datasets; a multi-source integrated training data development and decision-tree based land cover classifications; a temporally, spectrally, and spatially integrated land cover change analysis strategy; a hierarchical theme-based post-classification and integration protocol for generating land cover and change products; a continuous fields biophysical parameters modeling method; and an automated scripted operational system for the NLCD 2019 production. For information see [Data | Multi-Resolution Land Characteristics \(MRLC\) Consortium](#)

**Data Source:** National Land Cover Database (NLCD)

**File Name:** nlcd\_2019\_imperviousPercent\_CA.tif

## WETLAND INTEGRITY

Wetlands provide critical habitat, store carbon, enhance water quality, control erosion, filter and retain nutrient pollution, and provide spaces for recreation. They are local and regional centers of biodiversity, and support species found nowhere else across western landscapes. Functional wetland ecosystems will serve increasingly important roles in buffering impacts from extreme climate events, and upland disturbances such as flooding and erosion. Meadow and riparian ecosystems provide ecosystem services and are key linkages between upland and aquatic systems in forested landscapes.

**DESIRED OUTCOME:** Wetland ecosystems are biologically intact, provide multiple ecosystem services, and meadow and riparian ecosystems provide key linkages between upland and aquatic systems in forested landscapes.

#### HYDROLOGIC FUNCTION

Hydrologic systems in the Sierra Nevada function through a complex interaction of topographic patterns, interannual variability of precipitation, and heterogeneous mosaics of vegetation to yield water and maintain valuable wetland habitats. Land management can have profound impacts on the hydrologic function of mountainous landscapes.

*No metrics currently in this element*

#### COMPOSITION

Wetland composition pertains to the array of different wetland types, their relative abundance, the uniqueness of their co-occurrence and composition, and their integrity in a given location and area within and across landscapes. Wetland ecosystems include all lentic (e.g. lakes, ponds, bogs, fens) and lotic (e.g., rivers, streams, springs, seeps) aquatic ecosystems, as well as associated vegetated wetlands such as wet meadows and riparian vegetation.

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#### AQUATIC SPECIES RICHNESS

**Tier:** 1

**Data Vintage:** 02/2018

**Metric Definition and Relevance:** Aquatic native species richness is a measure of species biodiversity, and is one measurement used to describe the distribution of overall species biodiversity in California for the California Department of Fish and Wildlife (CDFW) Areas of Conservation Emphasis Project (ACE). Native species richness represents a count of the total number of native aquatic species potentially present in each watershed based on species range and distribution information. The data can be used to view patterns of species diversity, and to identify areas of highest native richness across the state. The species count consists of four taxonomic groups – fish, aquatic invertebrates, aquatic amphibians, and aquatic reptiles.

**Data Resolution:** 30m Raster

**Data Units:** Count

**Creation Method:** For more information, see the Aquatic Native Species Richness Factsheet (2018) at <https://nrm.dfg.ca.gov/FileHandler.ashx?DocumentID=150852>

The California Department of Fish and Wildlife (CDFW) Areas of Conservation Emphasis (ACE) is a compilation and analysis of the best-available statewide spatial information in California on biodiversity, rarity and endemism, harvested species, significant habitats, connectivity and wildlife movement, climate vulnerability, climate refugia, and other relevant data (e.g., other conservation priorities such as those identified in the State Wildlife Action Plan (SWAP), stressors, land ownership). ACE addresses both terrestrial and aquatic data.

**Data Source:**

Aquatic Native Species Richness Summary, Areas of Conservation Emphasis (ACE), version 3.0, California Department of Fish and Wildlife (CDFW)  
ACE database

**File Name:** aquatic\_species\_richness\_CA\_2018.tif

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## WETLAND DIVERSITY

**Tier:** 1

**Data Vintage:** 06/2018

**Metric Definition and Relevance:** This data set represents the extent, approximate location, and type of wetlands and deepwater habitats in California. These data delineate the aerial extent of wetlands and surface waters as defined by Cowardin et al. (1979).

**Data Resolution:** 30m raster

**Data Units:** Thematic

**Creation Method:** Downloaded from the National Wetlands Inventory (NWI), polygon converted to 30 meter raster. For more information see <https://www.fws.gov/program/national-wetlands-inventory>.

**Definition of values:**

- Lake = Lake or reservoir basin. Lacustrine wetland and deepwater (L).
- Freshwater Emergent Wetland = Herbaceous marsh, fen, swale and wet meadow. Palustrine emergent (PEM).
- Estuarine and Marine Wetland = Vegetated and non-vegetated brackish and saltwater marsh, shrubs, beach, bar, shoal or flat. Estuarine intertidal and Marine intertidal wetland (E2, M2).
- Other = Farmed wetland, saline seep and other miscellaneous wetland. Palustrine wetland (Misc. types, PUS, Pf..)
- Freshwater Pond = Pond. Palustrine unconsolidated bottom, Palustrine aquatic bed (PUB, PAB).
- Estuarine and Marine Deepwater = Open water estuary, bay, sound, open ocean. Estuarine and Marine subtidal water (E1, M1).
- Riverine = River or stream channel. Riverine wetland and deepwater (R).
- Freshwater Forested/Shrub Wetland = Forested swamp or wetland shrub bog or wetland. Palustrine forested and/or Palustrine shrub (PFO, PSS).

**Data Source:** The National Wetlands Inventory, US Fish & Wildlife Service (USFWS)

**File Name:** NWI\_WetlandsType\_2018\_30m.tif

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## RIPARIAN AREAS

**Tier:** 1

**Data vintage:** 04/2019

**Metric Definition and Relevance:** These data depict riparian areas for 50-year flood heights for California in 2019.

**Data Resolution:** 30m Raster

**Data Units:** binary

**Creation Method:** Fifty-year flood heights were estimated using U.S. Geological Survey (USGS) stream gage information. NHDPlus version 2.1 was used as the hydrologic framework to delineate riparian areas. The U.S. Fish and Wildlife Service’s National Wetland Inventory and USGS 10-meter digital elevation models were also used in processing these data. See <https://doi.org/10.2737/RDS-2019-0030>

**Credits:** Sinan Abood, Ph.D. GISP; Research Scientist, Forest Service Washington Office (WO) – Biological & Physical Resources (BPR)

**Data Source:** USDA Forest Service

**File Name:** RiparianAreas\_2019.tif

## OPERATIONAL DATA LAYERS

In addition to the metric data layers assembled for this RRK project, a set of “operational” GIS data layers have been assembled to support use of the metrics. These data provide land use context (e.g. ownership, land use designations, background ecological information (e.g. climate refugia, stream locations, climate classes), infrastructure (roads, operational constraints, powerline corridors), and Forest Service policy information (spotted owl PACs, critical habitat maps for listed species, wilderness/roadless/wild and scenic rivers). These data are provided to assist managers in putting proposed treatments into context for what is feasible and what might constrain project planning.

Data layers provided within this designation of operational data are in their native projection and format with any embedded metadata maintained.

### FIRE

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#### FIRE SEVERITY CLASS (FIRES FROM 2012-2021)

**Definition and Relevance:** This dataset includes the footprint of all fires in California since 2012, designated in three classes of severity; low, moderate, and high.

**Data Vintage:** 2012-2021

**Data Resolution:** Raster, 30 meter pixels

**Data Units:** Categorical (fire severity class 1 (low severity), 2 (moderate severity), or 3 (high severity))

**Creation Method:** The difference-adjusted relativized difference normalized burn ratio (RdNBR) was calculated using methods modified from Parks et al (2018). Fire perimeters were obtained from CAL FIRE’s April 2021 fire perimeter database. A function for estimating basal area loss from RdNBR values was fit to data from Miller et al (2009) using quasibinomial logistic regression and applied to the 2012-2021 fires. Estimated basal area loss was thresholded to represent low (< 25% loss), moderate (25% – 75% loss), and high (> 75% loss) burn severity. For areas where multiple sequential fires burned from 2012-2021 the maximum burn severity is reported. Updated April 2023 to incorporate CAL FIRE’s October-2022 revisions to fire perimeters and to minimize data loss resulting from spatial reprojection.

- 1: Low Severity

- 2: Moderate Severity
- 3: High Severity

**Data Source:** Landsat 8, NASA, Fire History (2022), CAL FIRE Postfire mortality data, Miller et al. 2009; assembled at UC Davis (Joe Stewart)

**File Name:** fire\_severity\_class\_max\_2012to2021.tif

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#### HOUSING UNIT DENSITY

**Data Vintage:** 01/2020

**Definition and Relevance:** HUDen is a raster of housing-unit density measured in housing units per square kilometer. The HUDen raster was generated using population and housing-unit count and data from the U.S. Census Bureau, building footprint data from Microsoft, and land cover data from LANDFIRE.

**Data Resolution:** 30m Raster

**Data Units:** Housing units per square kilometer

**Creation Method:** Generate the HUDen raster from the building points. We first converted the building points to a 30-m raster where the raster value is the sum of the housing-units-per-centroid attribute of all building centroids within each raster grid cell. We then generated a smoothed density raster using a three-step process: 1) calculate a 200-m radius moving-window sum of the 30-m housing-unit count raster; 2) calculate a 200-m radius moving-window sum of habitable land cover (in sq km), where habitable land cover is all land covers except open water and permanent-snow/ice; and 3) divide the smoothed housing-unit count raster by the smoothed habitable land cover raster to generate housing unit density in housing units/sq km. To produce the final integer version of the HUDen raster, we set values less than 0.1 HU/sq km to zero, values between 0.1 and 1.5 to a value of 1 HU/sq km, and rounded all other values to the nearest integer.

**Data Source:** Pyrologix, LLC

**File Name:** HUden\_2020.tif

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#### WILDLAND URBAN INTERFACE

**Data Vintage:**2022

**Definition and Relevance:** The wildland urban interface (WUI) is the area where urban development is in close proximity to wildland vegetation. WUI data for the conterminous U.S. based on 125 million building locations where buildings intermingle with wildland vegetation according to the Federal Register definitions of the WUI. According to the definitions used for our building-based maps and for the census-based maps, WUI is where building density exceeds 6.17 units/km<sup>2</sup> and where land cover is either (1) at least 50% wildland vegetation (intermix) or (2) under 50% wildland vegetation but within 2.4 km (1.5 miles) of a patch of wildland vegetation at least 5 km<sup>2</sup> in area that contains at least 75% vegetation (interface). The distance selected for the interface definition is based on research from the California Fire Alliance suggesting that this is the average distance firebrands can travel from an active wildfire front (Stewart et al., 2007).

**Data Resolution:** 30m Raster

**Data Units:** Categorical

**Creation Method:** Building point locations were obtained from a Microsoft product released in 2018, updated to 2019-2020 for most of California, which classified building footprints based on high-resolution satellite imagery. Maps were also based on wildland vegetation mapped by the 2016 National Land Cover Dataset (Yang et al., 2018). The mapping algorithm utilized definitions of the WUI from the U.S. Federal Register (USDA & USDI, 2001) and Radeloff et al. (2005). Both classes required a minimum building density of 6.17 buildings per km<sup>2</sup>. This map of intermix and interface WUI was generated using a circular neighborhood size based on radius distance of 100m to determine building density and vegetation cover on a pixel-by-pixel basis (Bar Massada et al., 2013). Source: USGS ScienceBase Data Catalog; <https://www.sciencebase.gov/catalog/item/617bfb43d34ea58c3c70038f>

Values in the raster are defined as:

- 0: Not WUI
- 1: Intermix WUI
- 2: Interface WUI

**Data Source:** WUI, Carlson et al, 2022

**File Name:** MSB\_WUI\_CA\_100m.tif

## ADMINISTRATIVE

### URBAN-AGRICULTURE LAND USE

**Definition and Relevance:** This dataset covers the urban and agricultural landscape for all forms of urban and agricultural land use in California. It was created using a combination of best available land cover data from multiple sources (see below). These data are used as a mask for selected metrics in the RRK project where inclusion of urban and agricultural cover potentially creates confusion in calculations of the metric.

**Data Vintage:** 06/2020

**Data Resolution:** Raster, 30m

**Data Units:** Thematic

**Creation Method:**

1. [Farmland Mapping and Monitoring Program \(FMMP\)](#) land-use data from 2018 was converted to 30m raster as the base input, using the values from the Type field of:
  - Farmland of Statewide Importance
  - Unique Farmland
  - Farmland of Local Importance
  - Urban and Built-Up Land
  - Rural Residential Land
  - Confined Animal Agriculture

2. Secondly, to bring more current data in, [LANDFIRE 2020 Existing Vegetation Type \(EVT\)](#) from 2020 was converted to 30m raster, using the values from EVT group name of:
  - Developed-Low Intensity
  - Developed-Medium Intensity
  - Developed-High Intensity
  - Agriculture-Cultivated Crops and Irrigated Agriculture
3. Lastly, [Building Footprints - Bing Maps \(microsoft.com\)](#) polygons were converted to 30m raster and added to the stack to include the most recent urban footprints.

**Data Source:**

Farmland Mapping and Monitoring Program (FMMP)

LANDFIRE: Existing Vegetation Type, U.S. Department of Agriculture and U.S. Department of the Interior

MS Building Footprints

**File Name:** UrbanAgLanduse\_RRK\_2020.tif

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BUILDING STRUCTURE DENSITY

**Definition and Relevance:** A raster dataset containing building footprints of California.

**Data Vintage:** 2012-2020. The vintage of the footprints depends on the vintage of the underlying imagery. Bing Imagery is a composite of multiple sources with different capture dates.

**Data Resolution:** Raster, 10m

**Data Units:** binary

**Creation Method:** Vector spatial data called US Building Footprints contained in a Microsoft dataset (available at <https://github.com/microsoft/USBuildingFootprints>) downloaded, clipped to California and converted to a 10m raster. For more information visit: [Building Footprints - Bing Maps \(microsoft.com\)](#)

**Data Source:** MS Building Footprints

**File Name:** CA\_bldgFootprints\_10m.tif

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HIGH-USE RECREATION AREAS

**Definition and Relevance:** A recreation site is a discrete area on a National Forest that provides recreation opportunities, receives recreational use, and requires a management investment to operate and/or maintain to standard under the direction of an administrative unit in the National Forest System. Recreation sites range in development from relatively undeveloped areas, with little to no improvements (Development Scale 0 and 1), to concentrations of facilities and services evidencing a range of amenities and investment (Development Scale 2 through 5).

Recreation opportunities are point locations of recreational site activities available to visitors and populates the Forest Service websites (<https://www.fs.usda.gov/>), and the interactive visitor map (<https://www.fs.usda.gov/ivm/>).

**Data Vintage:** 01/2023

**Data Resolution:** Points

**Data Units:** Tabular attributes

**Creation Method:** see Metadata

**Data Source:** USFS Enterprise Data Warehouse (EDW)

**File Name:** RECAREAACTIVITIES\_V\_2023.shp

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#### LAND DESIGNATIONS

**Definition and Relevance:** Wilderness, Roadless, Wild and Scenic River

**Data Vintage:** 01/2023

**Data Resolution:** ArcGIS file geodatabase: Vector, polygon

**Data Units:** Tabular attributes

**Creation Method:** Data layers pulled from the Enterprise Data Warehouse for land designations:

- *Wilderness* – area designated as a National Wilderness in the National Wilderness Preservation System
- *Inventoried Roadless Areas* – the 2001 Roadless Rule establishes prohibitions on road construction, road reconstruction, and timber harvesting on inventoried roadless areas on National Forest System lands by the following classifications:
  - 1B = Inventoried Roadless Areas where road construction and reconstruction is prohibited
  - 1B-1 = Inventoried Roadless Areas that are recommended for wilderness designation in the forest plan and where road construction and reconstruction is prohibited
  - 1C = Inventoried Roadless Areas where road construction and reconstruction is not prohibited
- *Wild and Scenic Rivers* – area designated as a National Wild, Scenic, or Recreational River within the National Wild and Scenic River System. The designations and definitions are:
  - Wild (W) – Those rivers or sections of rivers that are free of impoundments and generally inaccessible except by trail, with watersheds or shorelines essentially primitive and waters unpolluted. These represent vestiges of primitive America.
  - Scenic (S) – Those rivers or sections of rivers that are free of impoundments, with shorelines or watersheds still largely primitive and shorelines largely undeveloped, but accessible in places by roads.
  - Recreational (R) – Those rivers or sections of rivers that are readily accessible by road or railroad, that may have some development along their shorelines, and that may have undergone some impoundment or diversion in the past.

**Data Source:** USFS Enterprise Data Warehouse (EDW)

**File Name:** Wilderness\_2023.shp; Roadless\_2001.shp; WildScenicRiver\_2023.shp

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#### WILDERNESS - PROTECTED AREA DATABASE 3.0

**Definition and Relevance:** The PAD-US geodatabase was originally developed to organize and assess the management status (i.e. apply 'GAP Status Code') of elements of biodiversity protection by identifying species and plant communities not adequately represented in existing conservation lands (<https://www.usgs.gov/core-science-systems/science-analytics-and-synthesis/gap> ). In cooperation with the United Nations Environment Programme World Conservation Monitoring Centre (UNEP-WCMC), PAD-US also supports global conservation analyses to inform policy decisions (<https://www.protectedplanet.net/country/USA> , <https://www.protectedplanet.net/en/resources/global-reports> ). The dataset is robust and has been expanded in

recent years with major additions of local parks data to PAD-US 2.1, to support the recreation, natural resource management, wildfire planning, emergency management, transportation, research, and public health communities. New applications of the data are frequently discovered. Multiple attributes and a flexible database structure provide context for data to be used at local, regional, state, national, and international scales. See <https://www.usgs.gov/gapanalysis/PAD-US-resources> for more information.

Includes wilderness on ALL federal lands within California.

**Data Vintage:** 01/2022

**Data Resolution:** File Geodatabase Polygon Feature Class

**Data Units:** Tabular attributes

**Creation Method:** Polygons with [Designation Type] = 'Wilderness' extracted from the PADUS3\_0\_State\_CA\_GDB feature class for use in the RRK project. Polygons projected to California Teale Albers projection and 'exploded' into single part features.

**Data Source:** U.S. Geological Survey (USGS)

**File Name:** PAD3\_0\_Wilderness.gdb/PAD\_Wilderness3\_0\_2022\_CA\_sPart

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#### OWNERSHIP

**Definition and Relevance:** Ownership is a commonly used base layer used in a wide range of business functions and these data are intended to provide a depiction of the land ownership within the RRK project area. Two different data layers are provided, from different sources.

**Data Vintage:** FS\_BasicOwnership: 01/2023, ownership: 05/2022

**Data Resolution:** Vector, polygon

**Data Units:** Tabular attributes

**Creation Method:**

- *FS\_BasicOwnership\_2022.shp* – an area depicted as surface ownership parcels dissolved on the same ownership classification administered by the USDA Forest Service (USFS).
- *ownership22\_1* – California Multi-Source Land Ownership, includes lands owned by each federal agency (including USFS), state agency, local government entities, conservation organizations, and special districts. It does not include lands of private ownership.

**Data Source:** USDA Forest Service, CAL FIRE

**File Name:** FS\_BasicOwnership\_2023.shp; ownership22\_1.shp

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#### ROADS

**Definition and Relevance:** This California statewide dataset was downloaded from [Geofabrik's free download server](#) for California. This server has data extracts from the OpenStreetMap project which are normally updated every day.

**Data Vintage:** 09/2022

**Data Resolution:** Vector, line

**Data Units:** Tabular attributes

**Creation Method:** To simplify the layer, major roads were exported with the following selection of the attribute “fclass”:

- 5111 = motorway
- 5112 = trunk
- 5113 = primary
- 5114 = secondary
- 5121 = unclassified
- 5122 = residential
- 5123 = living street

**Data Source:** [Open Street Map](#) roads based on Tiger Lines (OSM)

**File Name:** OSM\_majorRoads\_CA\_2022.shp

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#### SDG&E TRANSMISSION LINES

**Definition and Relevance:** The purpose of the Geographic Information Systems (GIS) data is for both the electrical corporations and Energy Safety to have a holistic understanding of the electrical corporation infrastructure, annual targets as reported in its WMP, and quarterly planning and progress for each initiative. These data enable Energy Safety to evaluate the electrical corporation’s WMP and compliance.

**Data Vintage:** All data in the Transmission Line feature class represents the as-built electric data model; the data represents the active and active to be removed Transmission Lines as at the time the data is extracted from PRO\_VAQ\_ELEC (approximately 1 month prior to submission date). The data does not contain preliminary segments as they do not physically exist as the time of the data extract.

**Data Resolution:** ArcGIS file geodatabase: Vector, line

**Data Units:** Tabular attributes

See

<https://energysafety.ca.gov/who-we-are/department-organization/electrical-infrastructure-directorate/data-analytics-division/> “Energy Safety Data Guidelines” for further detail on the attributes (page 37 for transmission lines and page 40 for distribution lines).

**Creation Method:** The Transmission Line feature class is derived from data within SDG&E's Electric GIS database (PRO\_VAQ\_ELEC) in addition to leveraging other source systems data, PowerWorkz and SAP. The data is extracted within 5 business days after the quarter closes to ensure the submission contains a constant dataset.

**Data Source:** San Diego Gas & Electric® Company

**File Name:** SDGE\_2023\_Q2NonConfidential.gdb\SDGE\_TransmissionLine\_2023\_Q2

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#### SDG&E PRIMARY DISTRIBUTION LINES

**Definition and Relevance:** The purpose of the Geographic Information Systems (GIS) data is for both the electrical corporations and Energy Safety to have a holistic understanding of the electrical corporation infrastructure, annual targets as reported in its WMP, and quarterly planning and progress for each initiative. These data enable Energy Safety to evaluate the electrical corporation’s WMP and compliance.

**Data Vintage:** All data in the Primary Distribution Line feature class represents the as-built electric data model; the data represents the active and active to be removed Transmission Lines as at the time the data is extracted from PRO\_VAQ\_ELEC (approximately 1 month prior to submission date). The data does not contain preliminary segments as they do not physically exist as the time of the data extract.

**Data Resolution:** ArcGIS file geodatabase: Vector, line

**Data Units:** Tabular attributes

See

<https://energysafety.ca.gov/who-we-are/department-organization/electrical-infrastructure-directorate/data-analytics-division/> “Energy Safety Data Guidelines” for further detail on the attributes (page 37 for transmission lines and page 40 for distribution lines).

**Creation Method:** The Primary Distribution Line feature class is derived from data within SDG&E's Electric GIS database (PRO\_VAQ\_ELEC\_PORTAL) in addition to leveraging other source systems data, PowerWorkz and SAP. The data is extracted within 5 business days after the quarter closes to ensure the submission contains a constant dataset.

**Data Source:** San Diego Gas & Electric® Company

**File Name:** SDGE\_2023\_Q2NonConfidential.gdb\SDGE\_PrimaryDistributionLine\_2023\_Q2

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#### SCE TRANSMISSION LINES

**Definition and Relevance:** Southern California Edison (SCE) provides data on its electrical facilities through their Distributed Resources Plan External Portal (DRPEP). DRPEP is home to SCE’s Distribution Resource Plan (DRP)-related information supporting SCE’s efforts to support customer use of clean energy technologies, streamline the interconnection process and help California meet its clean energy goals. As described in the Rulemaking 14-08-013 issued in February 2015, the California Public Utilities Commission (CPUC) required SCE and other utilities to publish a DRP by July 1, 2015. In the final guidance for the completion of the DRP, the CPUC ordered each utility to perform what is referred to as an Integration Capacity Analysis (ICA) for each line section or node in the distribution system and to perform the Commission-approved Locational Net Benefit Analysis (LNBA) methodology in their distribution systems.

DRPEP is an interactive web portal providing public access to:

- General locations of SCE distribution circuits, substations, and subtransmissions systems
- Distribution Energy Resources (DER) Integration Capacity Analysis (ICA) results (i.e., hosting capacity) Current, queued, and total distributed generation interconnection amounts
- Downloadable datasets with future API capabilities
- Location Net Benefit Analysis (LNBA) results
- Grid Needs Assessment (GNA) data
- Distribution Deferral Opportunity Report (DDOR) data

- Load Integration Capacity Analysis (ICA) - Substation & Circuit data
- Load Integration Capacity Analysis (ICA) Heat Map
- Historical Public Safety Power Shutoff (PSPS) data
- Distribution Circuits
- Future Transmission projects
- High Fire Risk Areas

For purposes of this Regional Resource Kit we are focusing on the data that indicate the footprint of the electrical lines and their voltage. The provided data include other attributes that may or may not be of interest. The attributes in the attribute table for the transmission lines include the following:

**Transmission table**

Circuit\_ID = Circuit ID

Circuit\_NO = Circuit Number

Circuit\_TY = Circuit Type: Transmission or Subtransmission

Circuit\_VO = Circuit Voltage

**Data Vintage:** All data in the Transmission Line shapefile represents the as-built electric data. The data are updated regularly by SCE. These data are current as of the fall of 2023.

**Data Resolution:** Vector, line

**Data Units:** Tabular attributes

**Creation Method:**

For additional information specific to the SCE data please visit:

<https://drpep.sce.com/drpep/>

and their user guide at:

<https://drpep.sce.com/drpep/drpep-interactive-user-guide/index.html#/>

**Data Source:** Southern California Edison (SCE)

**File Name:** SCE\_ICA\_TransmissionLines.shp

SCE PRIMARY DISTRIBUTION LINES

**Definition and Relevance:** Southern California Edison (SCE) provides data on its electrical facilities through their Distributed Resources Plan External Portal (DRPEP). DRPEP is home to SCE’s Distribution Resource Plan (DRP)-related information supporting SCE’s efforts to support customer use of clean energy technologies, streamline the interconnection process and help California meet its clean energy goals. As described in the Rulemaking 14-08-013 issued in February 2015, the California Public Utilities Commission (CPUC) required SCE and other utilities to publish a DRP by July 1, 2015. In the final guidance for the completion of the DRP, the CPUC ordered each utility to perform what is referred to as an Integration Capacity Analysis (ICA) for each line section or

node in the distribution system and to perform the Commission-approved Locational Net Benefit Analysis (LNBA) methodology in their distribution systems.

DRPEP is an interactive web portal providing public access to:

- General locations of SCE distribution circuits, substations, and subtransmissions systems
- Distribution Energy Resources (DER) Integration Capacity Analysis (ICA) results (i.e., hosting capacity) Current, queued, and total distributed generation interconnection amounts
- Downloadable datasets with future API capabilities
- Location Net Benefit Analysis (LNBA) results
- Grid Needs Assessment (GNA) data
- Distribution Deferral Opportunity Report (DDOR) data
- Load Integration Capacity Analysis (ICA) - Substation & Circuit data
- Load Integration Capacity Analysis (ICA) Heat Map
- Historical Public Safety Power Shutoff (PSPS) data
- Distribution Circuits
- Future Transmission projects
- High Fire Risk Areas

The purpose of the Geographic Information Systems (GIS) data is for both the electrical corporations and Energy Safety to have a holistic understanding of the electrical corporation infrastructure, annual targets as reported in its WMP, and quarterly planning and progress for each initiative. These data enable Energy Safety to evaluate the electrical corporation's WMP and compliance. For purposes of this Regional Resource Kit we are focusing on the data that indicate the footprint of the electrical lines and their voltage. The provided data include other attributes that may or may not be of interest. The attributes in the attribute table for the distribution lines include the following:

#### **Distribution table**

Note = notes (if applicable)

Sub\_name = Substation Name

Circuit\_ty = Circuit Type: A – DISTRICT POLE TOP; D – DISTRIBUTION ; S – SUBTRANSMISSION

Substation = Substation Name with high voltage/low voltage class appended

Sys\_name = System Name only

Circuit\_id = Circuit ID

Circuit\_vo = Circuit Voltage

SHAPE\_Len = Total length (approximate) of the circuit geometry perimeter in terms of meters.

Circuit\_nam = Circuit Name

**Data Vintage:** All data in the Distribution Line shapefile represents the as-built electric data. The data are updated regularly by SCE. These data are current as of the fall of 2023.

**Data Resolution:** Vector, line

**Data Units:** Tabular attributes

**Creation Method:**

For additional information specific to the SCE data please visit:

<https://drpep.sce.com/drpep/>

and their user guide at:

<https://drpep.sce.com/drpep/drpep-interactive-user-guide/index.html#/>

**Data Source:** Southern California Edison (SCE)

**File Name:** SCE\_Distribution\_circuits.shp

TERRESTRIAL

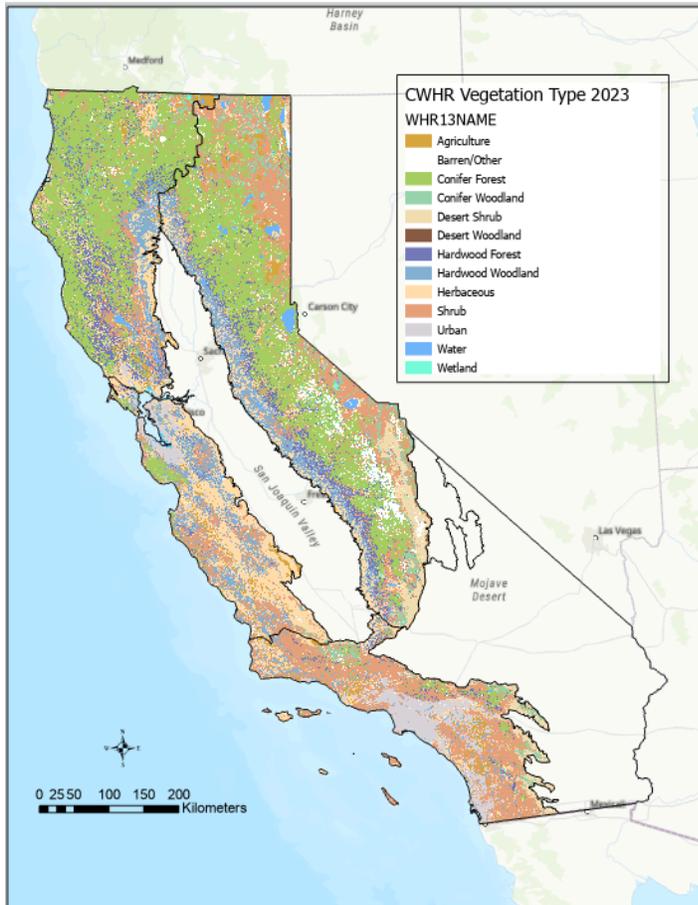
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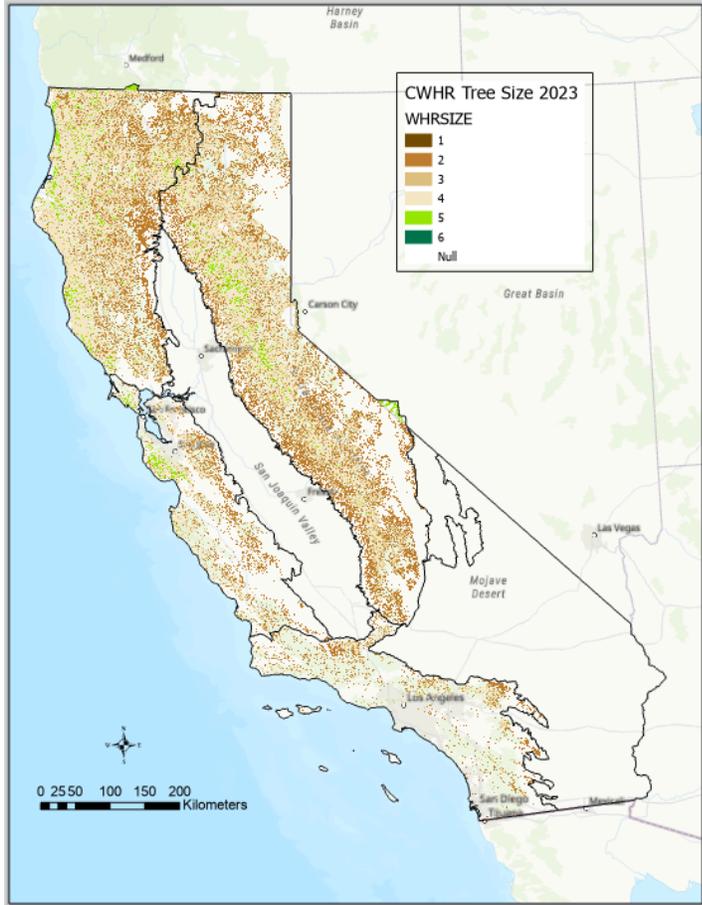
CWHR VEGETATION

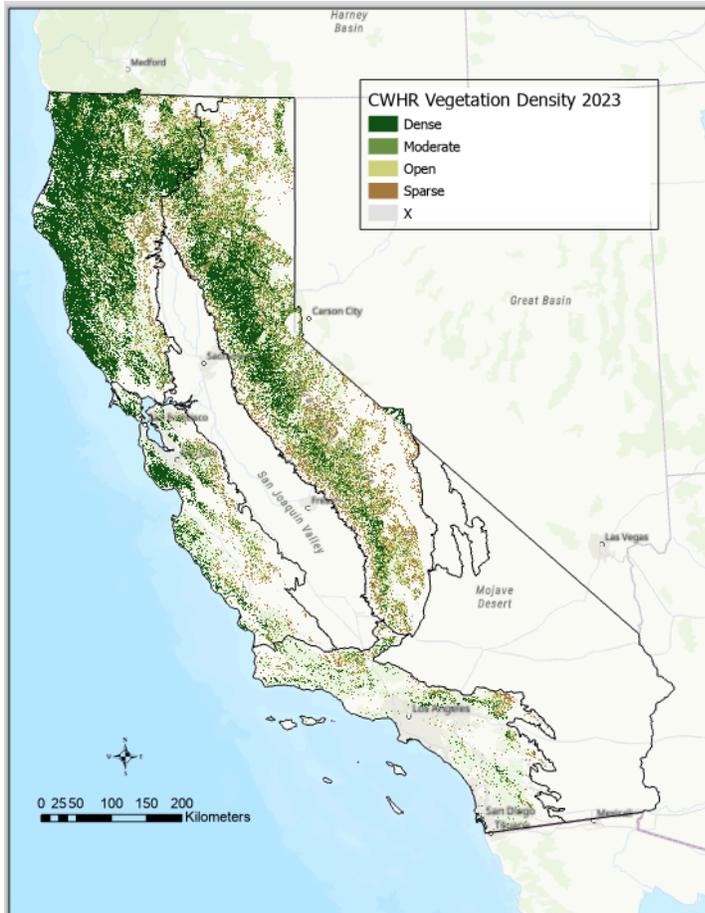
Vegetation maps play a vital role in characterizing conditions for many metrics. We need high resolution details on vegetation composition and structure; a vegetation classification, mapping, and inventory processes that provides vegetation information for a wide variety of metrics. The broader land management community continues to develop such products and make incremental improvements.

The current version displayed here is a new iteration of the Fveg. This is a product of CALFIRE-FRAP, which compiles the "best available" land cover data into a single data layer, to support the various analyses required for the Forest and Rangeland Assessment. Fveg attempts to provide an accurate depiction of the spatial distribution of habitat types within California, required for a variety of legislatively-mandated government functions. This is a product of the California Department of Forestry and Fire Protections CALFIRE Fire and Resource Assessment Program (FRAP), in cooperation with California Department of Fish and Wildlife VegCamp program and extensive use of USDA Forest Service Region 5 Mapping and Remote Sensing (MARS). The data span a period from approximately 1990 to 2014. Typically the most current, detailed and consistent data were collected for various regions of the state. Decision rules were developed that controlled which layers were given priority in areas of overlap. Cross-walks were used to compile the various sources into the common classification scheme, the California Wildlife Habitat Relationships (CWHR) system.

This iteration of Fveg has been updated as explained below. It only covers the four regions of the RRKs that have been mapped and only within the state of California. There are issues with the data in the portions of the map that extend into Nevada. This continues to be a work in progress where we hope to make improvements in the future. There are three separate rasters provided; one for CWHR Vegetation Type, one for CWHR Tree Size Class, and one for CWHR Veg Canopy Cover (Density) Class.







**Definition and Relevance:** Vegetation maps are important for characterizing many important features of a landscape such as wildlife habitat, fuels conditions, forest composition, and carbon. Such data are most useful if they can depict vegetation type, cover, and tree size class. This version was created to capture current conditions as best as possible through a variety of existing and current sources. Cross-walks were used to compile the various sources into the common classification scheme, the California Wildlife Habitat Relationships (CWHR) system. See CWHR for more details on the CWHR system ([California Wildlife Habitat Relationships](#)).

Key field names in this data set (there are others) are defined as follows:

WHRALL - Unique habitat data label. Concatenated from separate habitat attributes WHRtype, WHRsize and WHRdensity.

WHRNUM - Unique number for each Wildlife Habitat Relationship class (WHRtype).

WHRNAME - Unique name for each Wildlife Habitat Relationship class (WHRtype)

WHRTYPE - Unique Wildlife Habitat Relationship (WHR) class code

WHRSIZE - Wildlife Habitat Relationship Size Class (tree types only)

WHRDENSITY - Wildlife Habitat Relationship class (tree types only)

SOURCE\_NAME - General description of where the source data layer used for a given geography

SOURCE\_YEAR - Year of base imagery that source data layer references for a given geography

WHR Codes for Vegetation Types:

**Tree Dominated Habitats**

CWHR Code	Type Description
ASP	Aspen
BOP	Blue Oak-Foothill Pine
BOW	Blue Oak Woodland
COW	Coastal Oak Woodland
CPC	Closed-Cone Pine-Cypress
DFR	Douglas Fir
DRI	Desert Riparian
EPN	Eastside Pine
EUC	Eucalyptus
JPN	Jeffrey Pine
JST	Joshua Tree
JUN	Juniper
KMC	Klamath Mixed Conifer
LPN	Lodgepole Pine
MHC	Montane Hardwood-Conifer
MHW	Montane Hardwood
MRI	Montane Riparian
PJN	Pinyon-Juniper
POS	Palm Oasis

PPN	Ponderosa Pine
RDW	Redwood
RFR	Red fir
SCN	Subalpine Conifer
SMC	Sierran Mixed Conifer
VOW	Valley Oak Woodland
VRI	Valley Foothill Riparian
WFR	White fir

### Shrub Dominated Habitats

CWHR Code	Type Description
ADS	Alpine Dwarf-Shrub
ASC	Alkali Desert Scrub
BBR	Bitterbrush
CRC	Chamise-Redshank Chaparral
CSC	Coastal Scrub
DSC	Desert Scrub
DSS	Desert Succulent Shrub
DSW	Desert Wash
LSG	Low Sage
MCH	Mixed Chaparral
MCP	Montane Chaparral
SGB	Sagebrush

### Herbaceous Dominated Habitats

CWHR Code	Type Description
AGS	Annual Grass
FEW	Fresh Emergent Wetland
PAS	Pasture
PGS	Perennial Grass
SEW	Saline Emergent Wetland
WTM	Wet Meadow

#### Aquatic Habitats

CWHR Code	Type Description
EST	Estuarine
LAC	Lacustrine
MAR	Marine
RIV	Riverine

#### Developed Habitats

CWHR Code	Type Description
CRP	Cropland
DGR	Dryland Grain Crops
DOR	Deciduous Orchard
EOR	Evergreen Orchard
IGR	Irrigated Grain Crops
IRF	Irrigated Row and Field Crops
IRH	Irrigated Hayfield

OVN	Orchard - Vineyard
RIC	Rice
URB	Urban
VIN	Vineyard

**Non-vegetated Habitats**

CWHR Code	Type Description
BAR	Barren

WHR Codes for Tree Size Classes:

CWHR Code	CWHR Size Class	Conifer Crown Diameter	Hardwood Crown Diameter	DBH
1	Seedling tree	n/a	n/a	<1.0"
2	Sapling tree	n/a	<15.0'	1.0" - 5.9"
3	Pole tree	<12.0'	15.0' - 29.9'	6.0" - 10.9"
4	Small tree	12.0' - 23.9'	30.0' - 44.9'	11.0" - 23.9"
5	Medium/large tree	≥24.0'	≥45.0'	≥24.0"
6	Multi-layered tree	A distinct layer of size class 5 trees over a distinct layer of size class 4 and/or 3 trees, and total tree canopy of the layers ≥60% (layers must have ≥10.0% canopy cover and distinctive height separation).		

WHR Codes for Density Classes:

WHR Code	CWHR Closure Class	Vegetation Cover (Canopy Closure)
S	Sparse Cover	10.0 - 24.9%
P	Open Cover	25.0 - 39.9%
M	Moderate Cover	40.0 - 59.9%

D	Dense Cover	$\geq 60\%$
X	Not Determined / Not Applicable	

**Data Vintage:** 04/2023

**Data Resolution:** Raster, 30 meter pixels

**Data Units:** Categorical (see above)

**Creation Method:** Vegetation maps are an important feature of any natural resource management portfolio. Currently the vegetation map for the entire state that is considered the "best available" data is the CALFIRE data known as FVEG (*Vegetation (fveg) - CALFIRE FRAP [ds1327]*). This is an excerpt from the metadata:

“The California Department of Forestry and Fire Protections CALFIRE Fire and Resource Assessment Program (FRAP), in cooperation with California Department of Fish and Wildlife VegCamp program and extensive use of USDA Forest Service Region 5 Remote Sensing Laboratory (RSL) [*now known as Mapping and Remote Sensing Team (MARS)*], has compiled the "best available" land cover data available for California into a single comprehensive statewide data set. The data span a period from approximately 1990 to 2014. Cross-walks were used to compile the various sources into the common classification scheme, the California Wildlife Habitat Relationships (CWHR) system.”

Given the degree of fire in California in the last 30 plus years, especially in areas that experienced high severity fire, our RRK team thought that using the last version of FVEG (from 2015 but source data could be as old as 1987) would have too many glaring errors. Notwithstanding the challenge of creating reliable vegetation maps, we thought it would be possible to make improvements over the most recent map.

There are many avenues for improving vegetation maps. However, we did not have time to build anything from a new starting point, so we constructed a few simple rules for making [updates to the FVEG data layer](#).

There are three separate rasters provided; one for CWHR Vegetation Type, one for CWHR Tree Size Class, and one for CWHR Veg Canopy Cover (Density) Class.

The sources for updated data include:

- Fire severity data (from CALFIRE)
- LANDFIRE 2021 land cover data (wildland fire management programs of the USDA Forest Service and USDI)
- Herbaceous cover (Region 5 MARS Team)
- California Forest Observatory (SALO)

**Updated FVEG; Methods for 2023 statewide updates to FVEG WHRtype, WHRsize, and WHRdensity**

An accurate depiction of the spatial distribution of vegetation/habitat types within California is required for a number of the metrics included in this kit, particularly for some of the fire, forest and rangeland resiliency, and

biodiversity metrics . The California Department of Forestry and Fire Protections CALFIRE Fire and Resource Assessment Program (FRAP), in cooperation with California Department of Fish and Wildlife VegCamp program and extensive use of USDA Forest Service Region 5 Mapping and Remote Sensing unit (MARS) data, has compiled the "best available" land cover data available for California into a single comprehensive statewide data set. The data span a period from approximately 1990 to 2014.

Because the source data are in many cases fairly old and there has been extensive disturbance, particularly from wildfire, over the last 25 years, we made some updates to the 2015 version of FVEG. The methods for making those changes are described here.

*WHRtype update*

FVEG’s WHRtype was updated with the LANDFIRE Existing Vegetation Type (EVT) data product version 2.2.0 (LANDFIRE 2020) and the Rangeland Analysis Platform (RAP) fractional ground cover data product version 3.0 (Jones et al. 2018, Allred et al. 2021). Pixels were considered for update where high severity wildfire occurred after the FVEG mapping date. High severity was defined as wildfire burned areas that experienced  $\geq 75\%$  loss in basal area (Parks et al. 2018, Young-Hart et al. 2022) following the wildfire event. The type of update that occurred in each “high severity” pixel was dependent upon a lifeform conversion comparison (FVEG-to-LANDFIRE EVT), vegetation height (SALO 2020), and percent ground cover by annual and perennial grasses (RAP) (Table 1).

Table 1. FVEG-LANDFIRE update type for high severity pixels. Annual grass (AG) cover and perennial grass (PG) cover data were from the Rangeland Analysis Platform fractional ground cover data product version 3.0. Canopy height (CH) data were from the SALO forest observatory data product.

		FVEG Lifeform		
		<i>Herbaceous</i>	<i>Shrub</i>	<i>Tree</i>
LANDFIRE EVT Lifeform	<i>Herbaceous</i>	<ul style="list-style-type: none"> <li>Where AG cover &gt;50%, update to “Annual Grassland” WHRname (“AGS” WHRtype)</li> </ul>	<ul style="list-style-type: none"> <li>Where CH &lt;1m &amp; AG cover &gt;50%, update to “Annual Grassland” WHRname (“AGS” WHRtype)</li> </ul>	<ul style="list-style-type: none"> <li>Where CH &lt;2m &amp; AG cover &gt;50%, update to “Annual Grassland” WHRname (“AGS” WHRtype)</li> </ul>
		<ul style="list-style-type: none"> <li>Where PG cover &gt;50%, update to “Perennial Grassland” WHRname (“PGS” WHRtype)</li> </ul>	<ul style="list-style-type: none"> <li>Where CH &lt;1m &amp; PG cover &gt;50%, update to “Perennial Grassland” WHRname (“PGS” WHRtype)</li> </ul>	<ul style="list-style-type: none"> <li>Where CH &lt;2m &amp; PG cover &gt;50%, update to “Perennial Grassland” WHRname (“PGS” WHRtype)</li> </ul>
		<ul style="list-style-type: none"> <li>Otherwise, no update to WHRtype</li> </ul>	<ul style="list-style-type: none"> <li>Otherwise, FVEG-LANDFIRE crosswalk update</li> </ul>	<ul style="list-style-type: none"> <li>Otherwise, FVEG-LANDFIRE crosswalk update</li> </ul>
	<i>Shrub</i>	FVEG-LANDFIRE crosswalk update	No update to WHRtype	FVEG-LANDFIRE crosswalk update
	<i>Tree</i>	FVEG-LANDFIRE crosswalk update	<ul style="list-style-type: none"> <li>FVEG-LANDFIRE crosswalk update</li> </ul>	No update to WHRtype

*WHRdensity and WHRsize updates*

Following the WHRtype update, pixels that had lifeform “tree” then had the FVEG attributes “WHRdensity” and “WHRsize” updated using the SALO Forest Observatory canopy height and canopy cover data

products (SALO 2020). SALO data were available for the years 2016-2020, values of canopy height and canopy cover were averaged across years for the update[2] .

To update WHRdensity, SALO canopy cover was converted to WHRdensity canopy closure class per the [Wildlife Habitat Relationships, Standards for Canopy Closure Table 114C](#).

To update WHRsize, we developed allometric equations that predict tree DBH (diameter at breast height, breast height = 4.5 ft) as a function of tree height (HT, ft). We used data from the USDA Forest Inventory and Analysis program (FIA) for California (FIA DataMart 2023; California 2022 database; ver. 9.0.1). For this analysis, we included live trees ≥ 4.5 ft tall with a crown class code of dominant, co-dominant, or open grown (N = 165,224 tree measurements between 1991 and 2019). Trees were grouped by region based on the “fuzzed” location of the plot. Regions were defined by the Regional Resource Kits (2023, 4 regions) and separated into softwoods and hardwoods as defined by FIA (2 categories). For each analysis, three functions were evaluated: linear, saturating, and power:

Linear:  $DBH = a + (b \cdot HT)$ ;

Saturating (Michaelis–Menten):  $DBH = (V_m \cdot HT) / (K + HT)$ ;

Power:  $DBH = aHT^b$ .

For the most informative model (i.e., lowest AIC), we report both the root mean squared error (RMSE) and the pseudo R<sup>2</sup>. In this case, pseudo R<sup>2</sup> was calculated as the coefficient of determination between the observed and predicted DBHs (Table 2). We used the most informative HT-to- DBH function for the region and tree category to convert SALO canopy height data to DBH that was then converted to WHRsize class per the [Wildlife Habitat Relationships, Standards for Tree Size Table 114B](#).

Table 2. Height-to-DBH conversion equations by California region and tree class. DBH is in inches; Height (HT) is in feet. Only included trees with a HT > = 4.5 feet. Only included canopy class = dominant, co-dominant, or open. Equation (EQN) code: MM = Michaelis Menton; POWER = power; Linear = linear.

Region	Tree Class	EQN	a (Vm)	b (K)	RMSE	pseudoR <sup>2</sup>	EQN formula
Sierra Nevada	Softwood	MM	223.39	712.20	6.57	0.69	$DBH = (V_m \cdot HT) / (K + HT)$
Sierra Nevada	Hardwood	Linear	-0.391	0.294	4.69	0.57	$DBH = a + b(HT)$
Southern California	Softwood	MM	108.97	216.30	7.47	0.55	$DBH = (V_m \cdot HT) / (K + HT)$

<b>Southern California</b>	Hardwood	MM	175.17	424.31	5.55	0.52	$DBH = (Vm*HT)/(K+HT)$
<b>Northern California</b>	Softwood	POWER	0.128	1.13	6.51	0.74	$DBH = a*HT^b$
<b>Northern California</b>	Hardwood	Linear	0.135	0.242	5.2	0.49	$DBH = a+b(HT)$
<b>Central California</b>	Softwood	Linear	0.588	0.244	8.25	0.62	$DBH = a+b(HT)$
<b>Central California</b>	Hardwood	MM	68.51	161.40	6.24	0.45	$DBH = (Vm*HT)/(K+HT)$

*Availability of Data and Materials*

Data used for the 2023 FVEG updates can be obtained from the following:

- LANDFIRE – <http://www.landfire.gov/>
- Rangeland Analysis Platform – <https://rangelands.app/products/>
- SALO Forest Canopy – <https://forestobservatory.com/download>
- 10-year summary of basal area lost – <https://data.fs.usda.gov/geodata/rastergateway/acre>
- Perturbed FIA data – <https://experience.arcgis.com/experience/3641cea45d614ab88791aef54f3a1849>

Google Earth Engine Python API script can be obtained from: [https://github.com/kjohnston73/fveg\\_update](https://github.com/kjohnston73/fveg_update)

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**Data Source:** CALFIRE, CDFW, LANDFIRE, California Forest Observatory (SALO), USDA Forest Service

**File Name:** RRK\_Fveg\_WHRtype\_2023Apr\_4regions\_v2.tif; RRK\_Fveg\_WHRsize\_2023Apr\_4regions.tif; RRK\_Fveg\_WHRDensity\_2023Apr\_4regions.tif

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#### STATEWIDE CROP MAPPING - PROVISIONAL

**Definition and Relevance:** Land use data is critically important to the work of the Department of Water Resources (DWR) and other California agencies. Understanding the impacts of land use, crop location, acreage, and management practices on environmental attributes and resource management is an integral step in the ability of Groundwater Sustainability Agencies (GSAs) to produce Groundwater Sustainability Plans (GSPs) and implement projects to attain sustainability.

**Data Vintage:** 2021

**Data Resolution:** Vector, polygons

**Data Units:** thematic - Fields were attributed with DWR crop categories and included citrus/subtropical, deciduous fruits and nuts, field crops, grain and hay, idle, pasture, rice, truck crops, urban, vineyards, and young perennials.

**Creation Method:** Land IQ was contracted by DWR to develop a comprehensive and accurate spatial land use database for the 2021 water year (WY 2021). The primary objective of this effort was to produce a spatial land use

database with accuracies exceeding 95% using remote sensing, statistical, and temporal analysis methods. This project is an extension of the 2014, 2016, 2018, 2019, and 2020 land use mapping, which classified over 14 million acres of land into irrigated agriculture and urban areas. Unlike the 2014 and 2016 datasets, the WY 2018, 2019, 2020, and 2021 datasets include multi-cropping and incorporates DWR ground-truth data from Siskiyou, Modoc, Lassen and Shasta counties. Land IQ integrated crop production knowledge with detailed ground truth information and multiple satellite and aerial image resources to conduct remote sensing land use analysis at the field scale. Individual fields (boundaries of homogeneous crop types representing cropped area, rather than legal parcel boundaries) were classified using a crop category legend and a more specific crop type legend. A supervised classification method using a random forest approach was used to classify delineated fields and was carried out county by county where training samples were available. Random forest approaches are currently some of the highest performing methods for data classification and regression. To determine frequency and seasonality of multiple-cropped fields, peak growth dates were determined for annual crops.

Data Source: Land IQ, [www.LandIQ.com](http://www.LandIQ.com), California Department of Water Resources, Division of Regional Assistance Regional Offices: Northern, North Central, South Central and Southern Regional Offices, and Water Use Efficiency Branch (Sacramento Headquarters).

Statewide Crop Mapping - Datasets - California Natural Resources Agency Open Data

File Name: i15\_Crop\_Mapping\_2021\_Provisional.shp

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## AQUATIC

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### LAKES/RESERVOIRS

**Definition and Relevance:** Water Bodies such as lake and reservoir features are represented in this layer pulled from the National Hydrography Dataset (NHD). These data were used to erase areas of lakes and ponds from every raster metric in the RRK project dataset.

**Data Vintage:** 12/2022

**Data Resolution:** 30m Raster

**Data Units:** Binary, 0/1

**Creation Method:** This dataset is a subset of vector polygon NHD water bodies, encompassing the RRK project boundary and converted to a raster grid at 30m resolution based on existence/non-existence.

**Data Source:** USGS National Hydrography Dataset (NHD);  
<https://www.usgs.gov/national-hydrography/national-hydrography-dataset>

**File Name:** NHD\_lakesReservoirs\_2022\_RRK.tif

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### PERENNIAL, INTERMITTENT AND EPHEMERAL STREAMS

**Definition and Relevance:** USGS National Hydrography Dataset (NHD); Flowline is the fundamental flow network consisting predominantly of stream/river and artificial path vector features. It represents the spatial geometry and carries the attributes

**Data Vintage:** 12/2022

**Data Resolution:** Vector, line

**Data Units:** Tabular attributes

**Creation Method:** Data selected from NHD Flowline feature class to contain only FType code 460, StreamRiver (Perennial, Ephemeral, Intermittent) for the state of California.

**Data Source:** USGS National Hydrography Dataset (NHD);  
<https://www.usgs.gov/national-hydrography/national-hydrography-dataset>

**File Name:** NHD\_Flowline\_2022\_RRK.tif

## DATA DISCLAIMERS

The user is responsible to verify the limitations of the geospatial data and to use the data accordingly. For all data layers, please check each data source and adhere to any limitations established by the data creator. In general, adhere to the terms for use listed at the bottom of this section.

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### DEPARTMENT OF FORESTRY AND FIRE PROTECTION (CALFIRE)

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### ADDITIONAL RESOURCES

California Department of Fish and Wildlife Areas of Conservation Emphasis program:

<https://wildlife.ca.gov/Data/Analysis/Ace>

California Department of Fish and Wildlife. California Interagency Wildlife Task Group. 2014. CWHR version 9.0 personal computer program. Sacramento, CA. <http://wildlife.ca.gov/Data/CWHR>

California Office of Environmental Health Hazard Assessment CalEnviroScreen 4.0 report:

<https://oehha.ca.gov/calenviroscreen/report/calenviroscreen-40>

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[Connecting Wildlands & Communities, Conservation Ecology Lab - San Diego State University.](#) [Connecting Wildlands & Communities | Climate Science Alliance](#)

Monitoring Trends in Burn Severity (MTBS) program: <https://www.mtbs.gov/>

Multi-Resolution Land Characteristics Consortium (MRLC): <https://www.mrlc.gov/>

Oregon State University Environmental Monitoring, Analysis, and Process Recognition (eMapR) Lab:  
<http://emapr.ceoas.oregonstate.edu/>

Rapid Assessment of Vegetation Condition after Wildfire (RAVG): <https://burnseverity.cr.usgs.gov/ravg/>

Spatial Informatics Group: [Home - SIG \(sig-gis.com\)](http://sig-gis.com)

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