

Managing Wildfire Risk in an Increasingly Complex Environment

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Agenda

Wildfire Risk Today

AIR Wildfire Model Domain

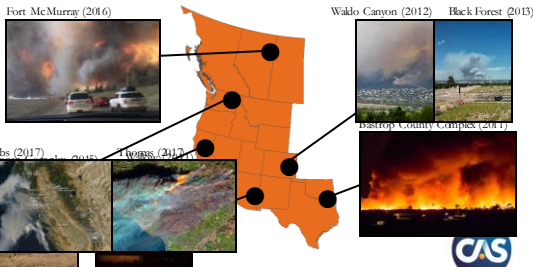
Modeling Ignitions

Spread Model

Modeling Vulnerability

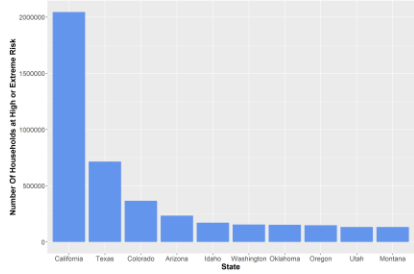


Wildfire Risk is Growing in the Western U.S. and Canada



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Top 10 Wildfire-Prone States (2017)



Number of Households at High or Extreme Risk
State
Number of Households is based on data from the 2010 U.S. Census
Source: "Wildfire-Prone States - Understanding and Reducing the Risk of Wildfire"



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AIR Wildfire Model for the United States

Historical Ignitions ≥ 100 Acres
1992-2015



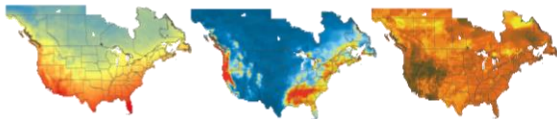
- 13 top loss-causing states
- Contiguous modeling across western United States
- Model based on natural variability in annual fire severity due to variations in weather



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Weather-Dependent Models Enable Correlated Countrywide Modeling of Wildfire Activity

Temperature Precipitation Drought



Wildfire severity is driven by **spatially** and **temporally** varying weather patterns



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Relationships Between Weather and Fire Activity are Complex and Vary Spatially

Fine Fuels

Fire behavior correlated to the drought conditions during the **growing** season

Wet Growing Season = ↑ Fire activity

Dry Growing Season = ↓ Fire activity



Coarse Fuels

Fire behavior correlated to the drought conditions during the **fire** season

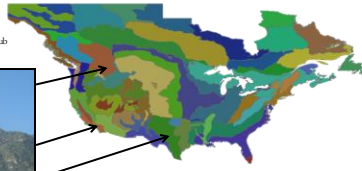
Wet Growing Season = ↓ Fire activity

Dry Growing Season = ↑ Fire activity



Wildfire Activity Modeled at the Ecoprovince Level

California Coastal Range Open Woodland Shrub
 • Coniferous Forest—Meadow Province



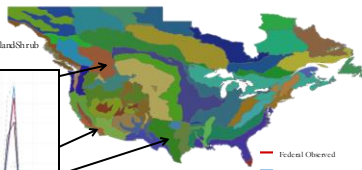
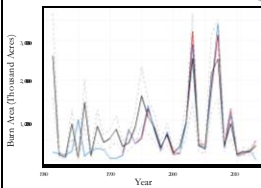
Ecoprovinces identify regions with similar ecosystems, climate, and vegetation—key drivers for fire behavior



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Ecoprovince Approach Captures Unique Relationships Between Weather and Wildfire

California Coastal Range Open Woodland Shrub
 Coniferous Forest Meadow Province



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Understanding the “Perfect Storm” for a WUI Disaster

Buildup of Fuels

- Forest/land management
- Long-term climate patterns

+

Extreme Burning Conditions

- High winds
- Low moisture

+

Residential Areas

- Growing at-risk population

- Weather drives wildfire activity
 - Long-term weather and fuel conditions set the stage for large wildfires
 - Short-term weather conditions alleviate or exacerbate fire conditions
- Wildfires have a large social impact when they meet urban areas

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The WUI is Critical to Understanding Wildfire Risk

Cerro Grande Fire, Los Alamos
Santa Chata Wildfire

- Development intersects with undeveloped lands
- Largest economic losses from wildfire will occur in the WUI
 - One-third of U.S. households are located in the WUI
 - 4,000 acres of wildland are converted to WUI *daily*

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Physically-Based Model Realistically Captures Fire Spread Across a Landscape

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Wind Speed and Direction are the Primary Drivers of Loss

- Wind is the primary propagator of spread direction and wildfire intensity
 - Fires with high wind speeds produce higher losses
- Wind speed, fuel, and topography are considered when determining the final intensity



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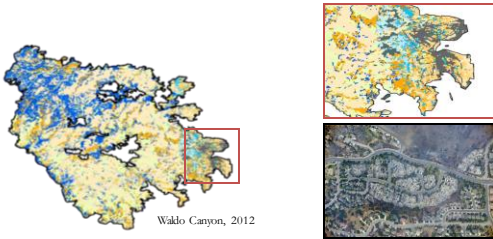
Spread Model Explicitly Accounts for All the Ways Fire Can Spread

- Surface Spread
- Surface-to-Canopy Transition
- Canopy Spread
- Fire Branding/Spotting



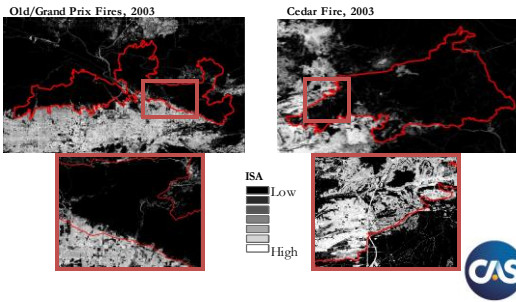
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WUI Penetration is the Biggest Driver of Loss



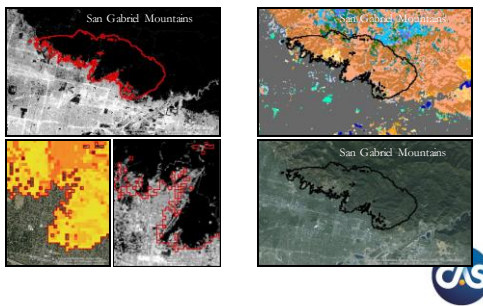
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Impervious Surface Area (ISA) is Used as a Measure of Urbanicity When Modeling WUI Penetration



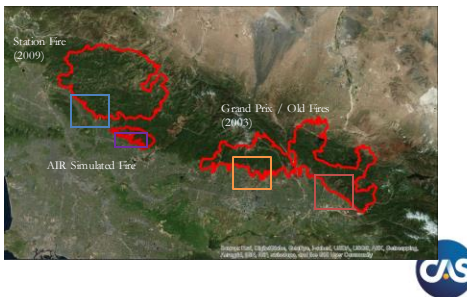
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Model Realistically Captures Fire Spread into Urban Areas

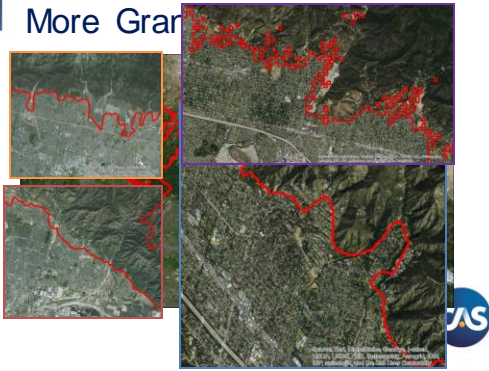


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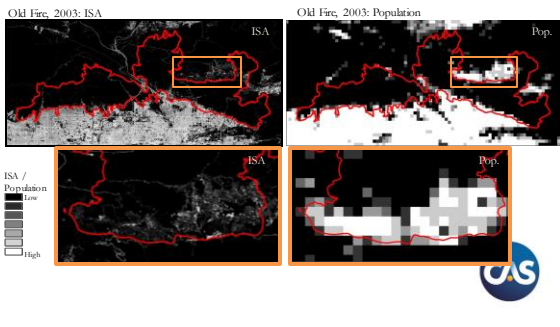
Modeled Fire Perimeters Are More Granular



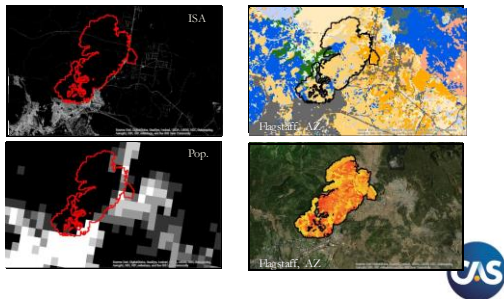
Modeled Fire Perimeters Are More Granular



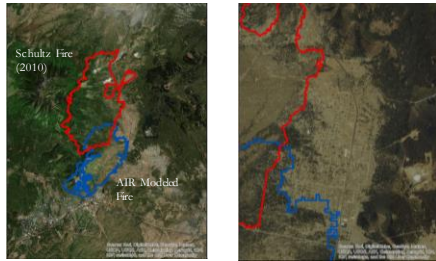
Population Is Used to Prioritize Suppression Resources



Model Accounts for Areas Where Fire Suppression Is Likely to Occur



Comparison with Historical Fires: Schultz Fire, 2010

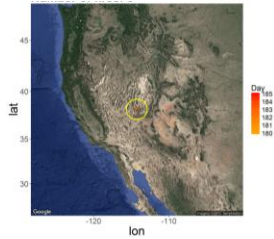


Unique Events are Identified by Clusters

Wildfires typically burn in clusters

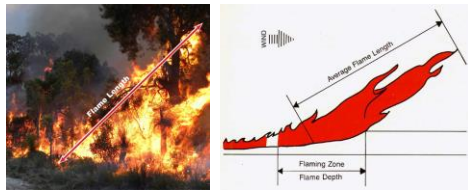
- Northern CA Fires (2017)
- Southern CA Fires (2017)

• Cluster fires by hours clause



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Building Damage Estimated Based on Flame Length



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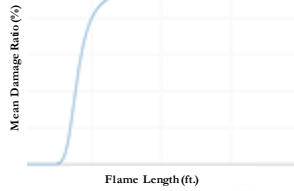
Building Damage Estimated Based on Flame Length

- Flame length is correlated with damage
 - 1-4 ft.: extinguish by hand tool
 - 4-8 ft.: too intense for hand tools
 - 8-11 ft.: serious control problems



Rothermel, 1985, How to predict the spread and intensity of fires and range fires, U.S. DoA

Estimated Damage Ratio Using Flame Length



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Binary Damage Pattern Revealed by Damage Surveys



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Vulnerability Module

- **Primary Risk Characteristics**
 - Construction, Occupancy, Height, Age, Region
- **Coverages A/B/C/D**
- **Secondary Risk Characteristics**
 - Firewise community
 - Defensible space
 - Roof cover
 - Wall siding
 - Glass type
 - Roof apertures



Photos from BEIS Wildfire Vulnerability Studies



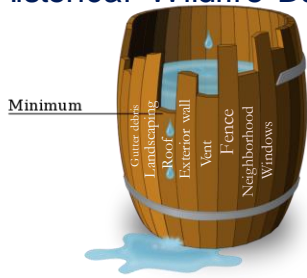
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Lessons Learned from Historical Wildfire Damages



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Lessons Learned from Historical Wildfire Damages



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Relative "Vulnerability" Across Construction/Occupancy/Height



Steel frame

Masonry

Wood frame



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Secondary Risk Characteristics Capture Critical Factors Affecting Structure Vulnerability



Defensible space



Roof type



Siding



Skylight and dormer



Questions and Discussion