# Effects of fire, fragmentation, and climate change on demographics of Ceanothus verrucosus

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#### Plant Conservation Challenges

 In the US – 22% of endangered plants occur in 8% of landscape where 50% of human population resides.

(Schwartz et al. 2002)





### Plant Conservation Challenges

- Biogeographic patterns of plant endangerment differ from vertebrates.
- Highest plant diversity associated with low productivity.



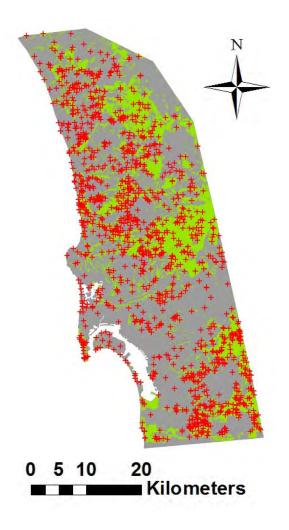


### Plant Conservation Challenges

 Rare plants more likely to be embedded in human dominated landscapes

Within this area

91 taxa 1,581 populations (CNDDB 2013) Rare Plants in Coastal San Diego County



# Link Between Rare Plants and Populated Landscapes

 Human development of landscapes follows predictable patterns (Huston 2005)



# Conservation Focus on Functional Landscapes

 Conservation planning based on functional ecosystems emphasized.

 May result in exclusion of viable plant populations from conservation strategies.



#### **Analysis of CNDDB Data**

- Populations of rare plants are not on average at higher risk in urban areas. (Lawson et al. 2008; Schwartz et al. in press)
- Small populations are not more likely to experience negative growth rates (Lawson et al. 2008; Schwartz et al. in press).
- Conservation efforts are not less likely to be successful in urban environments (Schwartz et al. in press)

### Rare Plant Conservation in Urban Environments

 This is not to say that all species are resistant to the effects of fragmentation and human development.

#### but rather

 There is no generalization that populations of rare species embedded in urbanized landscapes are doomed to extinction.

#### Conservation in Southern California

Biodiversity Hotspot

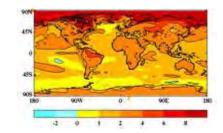
#### **Threats**

 Rapid urbanization (Ewing et al. 2005)

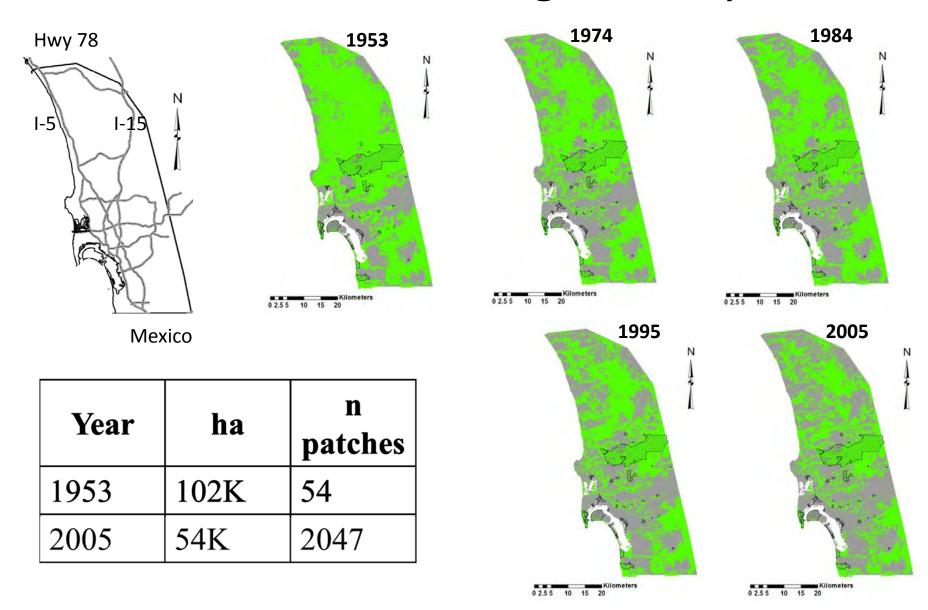


- Significant habitat loss & fragmentation (State of California (2006), Farmland Mapping and Monitoring Data)
- Altered fire regimes
- (Syphard et al. 2007)
- Climate change (Christensen et al. 2007, Cayan et al. 2008)





# Habitat loss and fragmentation western San Diego county



# Existing Threats Under Present Climate

Too frequent





Too infrequent



### Effects of Climate Change

#### Distribution shifts

#### Distribution contractions





(Loarie et al. 2008, Schwartz et al. 2006)

Distribution expansions
(Bradley et al. 2009)

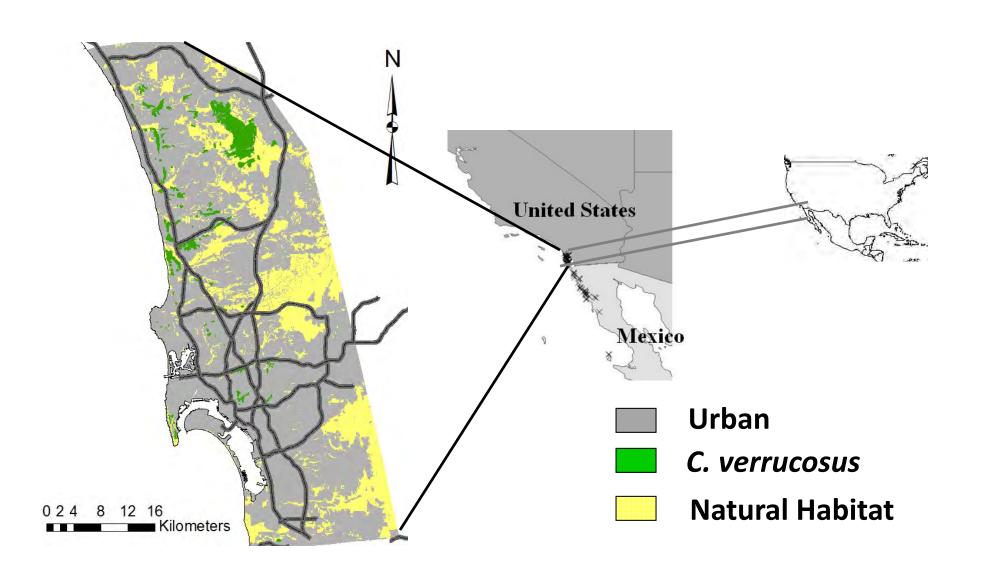


#### Research Questions

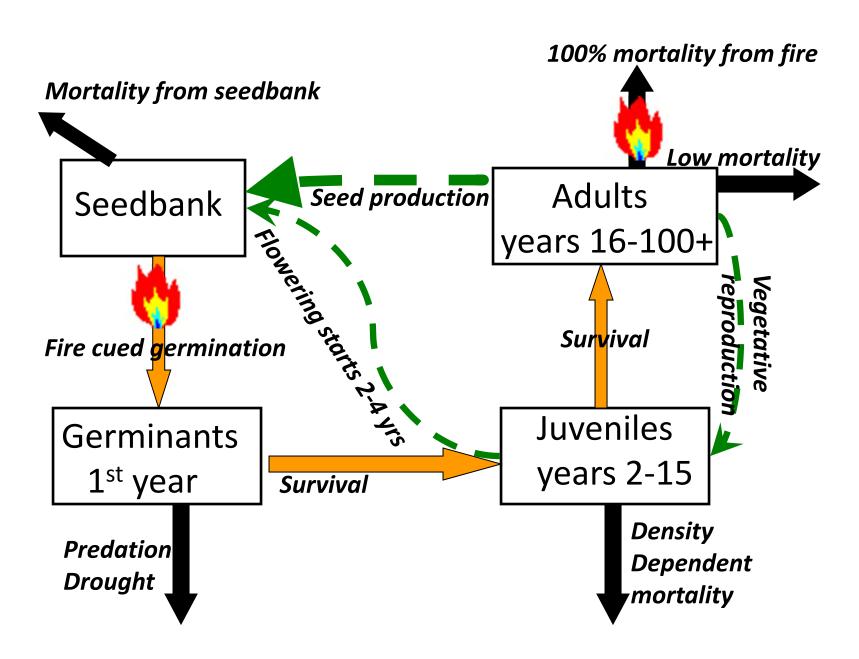
- What are the population-level effects of:
  - -Altered fire regime?
  - -Habitat loss and fragmentation?
- How do climate change projections alter the probability of species persistence?
- Does climate change pose a larger risk to species persistence than existing threats?



# Embedded in Urban Landscape



## Life History



#### Population Model

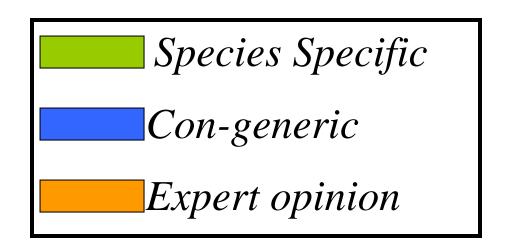
- Spatially explicit (151 populations)
- Age-based matrix model
- Fecundity and survival based on age
- Carrying capacity based on size of plants
- Stochastic
- Linked to fire hazard functions (Moritz 2003)
- Explicit response to fire

#### **Model Parameterization**

- Models can be difficult to parameterize due to sparse data
  - Data Sources
    - •Use of data from con-generics
    - Expert opinion
    - •Collect data where feasible

#### Data Sources for Vital Rates

- Fecundities
  - Seed Prod
  - Veg. repro
- Survival Rates
  - Longevity of seedbank
  - Seed Germination
  - Survival years 1-15
  - Survival from age 16-95
  - Survival from age 96



## Demography of Long-lived Plants

- Most knowledge of plant population dynamics based on studies of short-lived species
- Population trends of long-lived species difficult to detect on time scales convenient for human observation



# Seed Production (seedbank input)



### Seedbank Longevity

- Leveraged seedbank study (Cummins 2003)
  - quantified seedbank under live & dead pairs of CEVE
- Used ring counts to age dead CEVE and establish lower bound on seedbank longevity (Lawson 2011).
  - Longevity lower bound = 44 years





### **Seed Germination**



#### Germinant survival

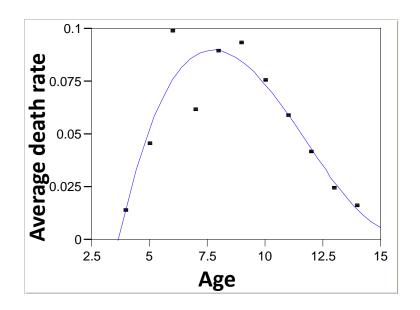


# Survival age 3-5



#### Survival age 6-15

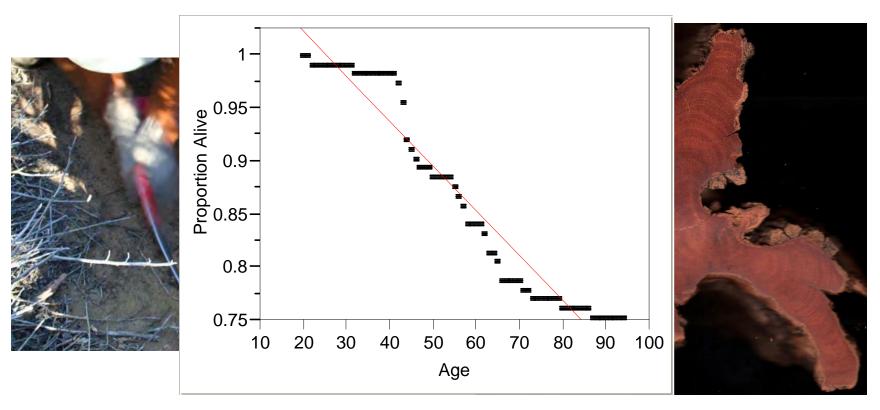
- •Data for *C. megacarpus* from (Schlesinger and Gill 1978)
- •Mean mortality= $0.160-0.0084*age-0.0028(age-9)^2+0.00034(age-9)^3$



- •Mean survival = 1- mean mortality
- •Stdev = 10% of mean survival

#### Survival at 16 to end of life span

 Lawson (2011) ring counts of dead individuals at Pt. Loma (CNM and Navy Lands)



• Survival = 98.8% Stdev 0.646%

### Longevity of adults

• 85-155 yea

nd Zedler 1993)

 Based on r approxima and MCAS

a (Zedler 1995) 11).

tand age is

• Tested 100 sensitivity

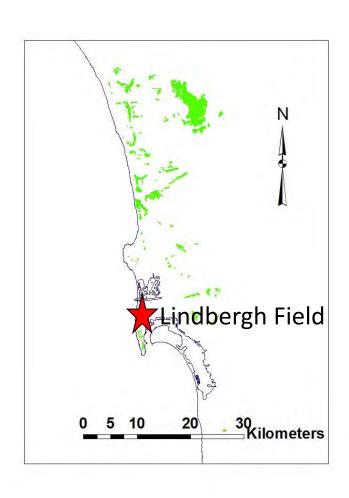
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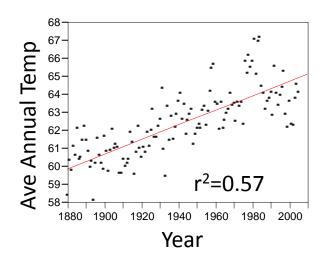


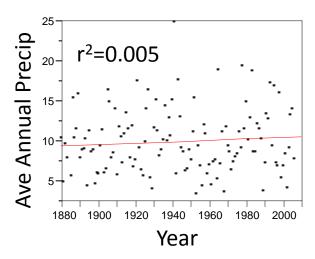
# Vegetative Reproduction



#### Climate







(Western Regional Climate Center 2009 a & b)

#### Climate Models

- NOAA GFDL CM2.1 SRES A2
- Medium High Emissions Scenario
- Predicts hotter and drier climate
  - 36% increase in temperature
- 26% decrease in precipitation
- NCAR PCM1 (DOE) SRES A2
- Medium High Emissions Scenario
- 17% increase in temperature
- 8% increase in precipitation

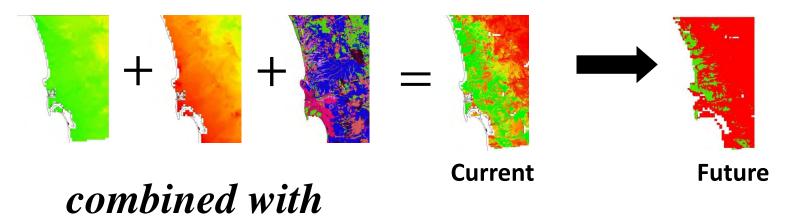
# Applying HSMs to Climate Change Questions

 Common approach - Prediction of range shifts using bioclimatic envelopes.
 (Loarie et. al. 2008; Thomas et. al. 2004).

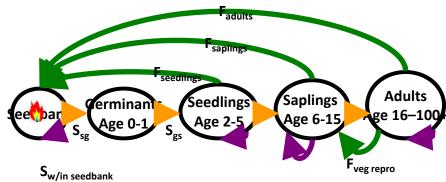
- Limitations
  - Shifts and contractions of suitable climates do not easily translate into extinction risks
  - Ignores demographic processes
- Recent approach Link dynamic bioclimatic envelopes with stochastic demographic models. (Anderson et al. 2009, Keith et al. 2008)

# Methods Effects of Climate Change

Habitat suitability models

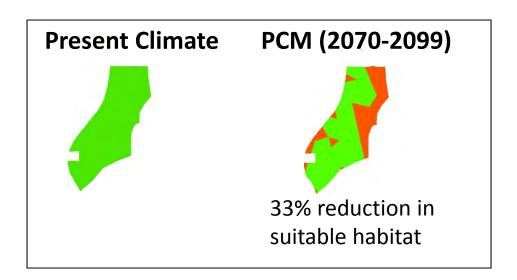


• Spatially explicit population viability models



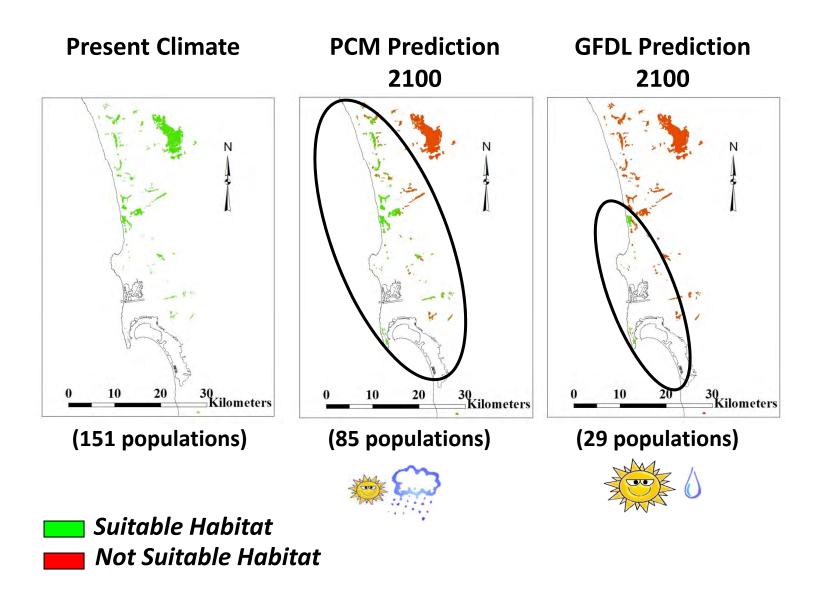
## Linking HSM to Population Model

- Used temporal trend in K.
- •Calculate % habitat loss per patch based on sequential HSM predictions.

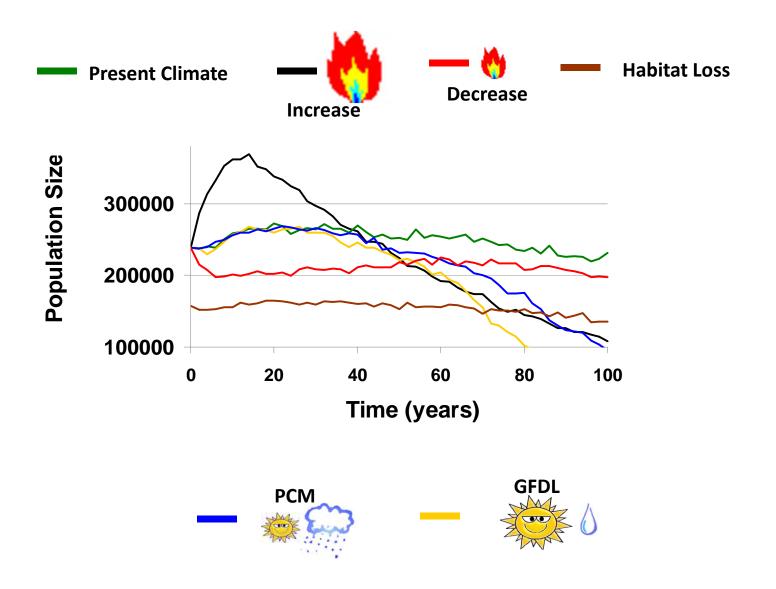


• Reduce K by a constant amount per time step to achieve projected decline.

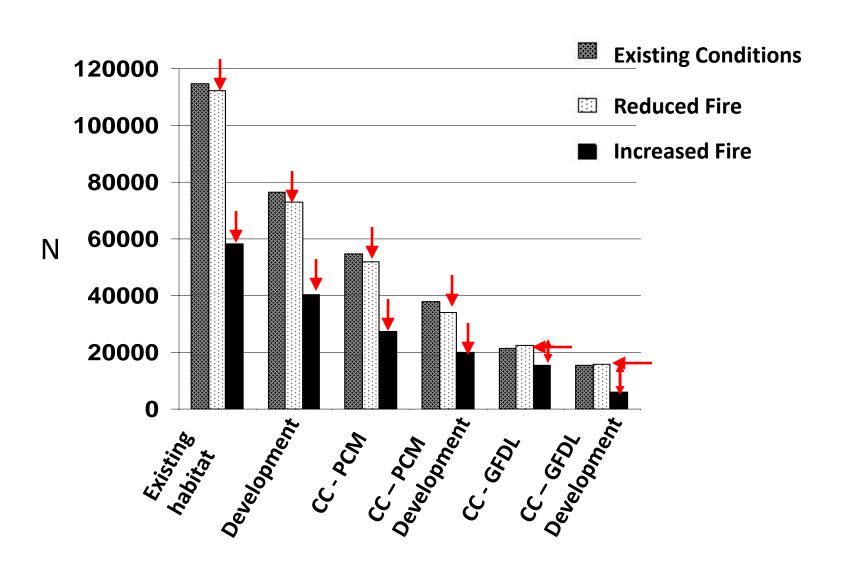
## Results - Habitat Suitability Model



### **Population Trajectories**



### **Expected Minimum Abundance**



#### Conclusions

- More frequent fires are bigger risk than less frequent fires
- Even though fire frequency is increasing, species can be threatened by extended fire intervals
- Development reduces abundance but population trajectory stable
- Climate change poses greatest risk

#### Conclusions

- Range shifts unlikely due to fragmented landscape and poor dispersal
- Interactions among threats may alter relative risks
- Plant conservation must address uncertain future objectives as climate change unfolds
- Impacts to obligate seeders can provide insights to community and ecosystem level effects of climate change

#### Are the model results "true"

- Does climate change really pose the greatest risk?
- Maybe not in the near term.
- There are huge uncertainties.
- Models are useful in adaptive management to:
  - synthesize what we know.
  - prioritize data collection.
  - generate hypotheses.
- Models are not the truth.

#### Conservation Decision Making

- Complex
- High Uncertainty
- Expert opinion only gets us so far
  - Biased risk estimates (Kahneman and Tversky)
  - Over-confidence in opinions (Tetlock)
- Quantitative tools needed for transparency, repeatability and accountability

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