

# Distributional Patterns of Perennial Pepperweed, *Lepidium latifolium*, in the San Francisco Bay



Melanie Vanderhoof and Chris Rogers,  
Environmental Science Associates

# Research Questions

- Where does *L. latifolium* occur in the San Francisco Bay area?
- Can its distribution pattern be explained and predicted using environmental variables?



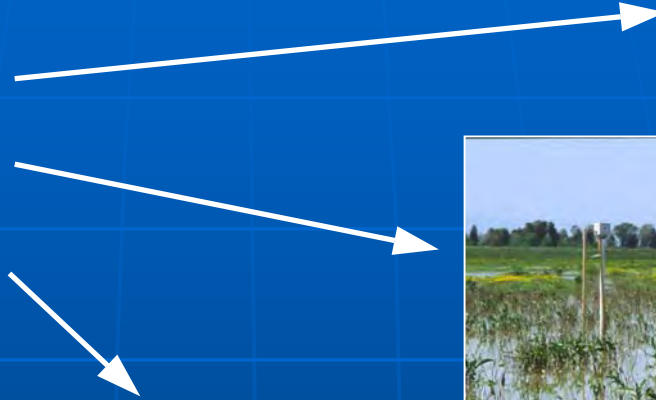
# *Lepidium latifolium*

- A perennial weed
- Member of mustard family.
- Native to Eurasia.
- First recorded in CA in 1936, possibly from contaminated agricultural seed.
- Forms dense colonies or patches that grow 2-3 ft in height.



# Invaded Habitats

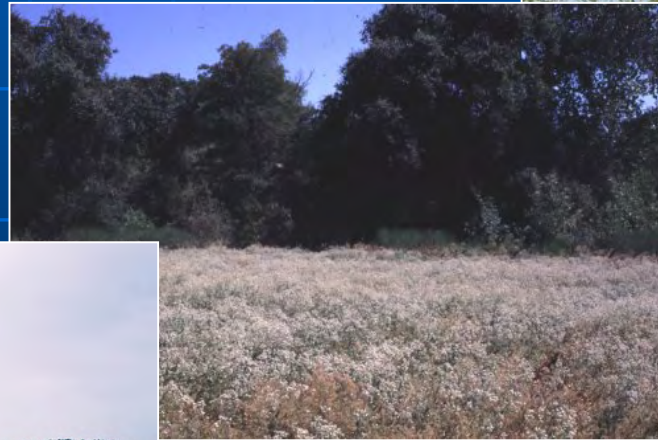
Agricultural areas



Marshes



Vernal Pools



Riparian Zones

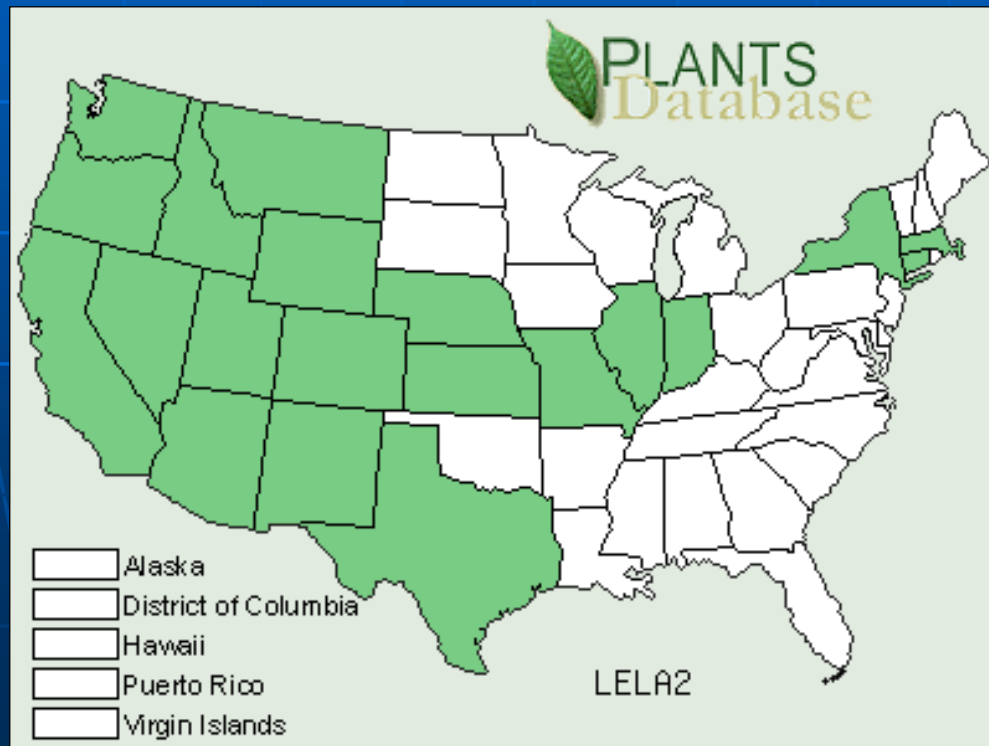
# Objectives

1. Map the distribution of *Lepidium latifolium* along the shoreline of the San Francisco Bay.
2. Develop a predictive model that identifies high risk areas in the larger San Francisco Bay Area based on environmental variables.



# Mapping Methods

- Limited mapping of *L. latifolium* has occurred in the bay area (Grossinger *et al.* 1998; May 1995)



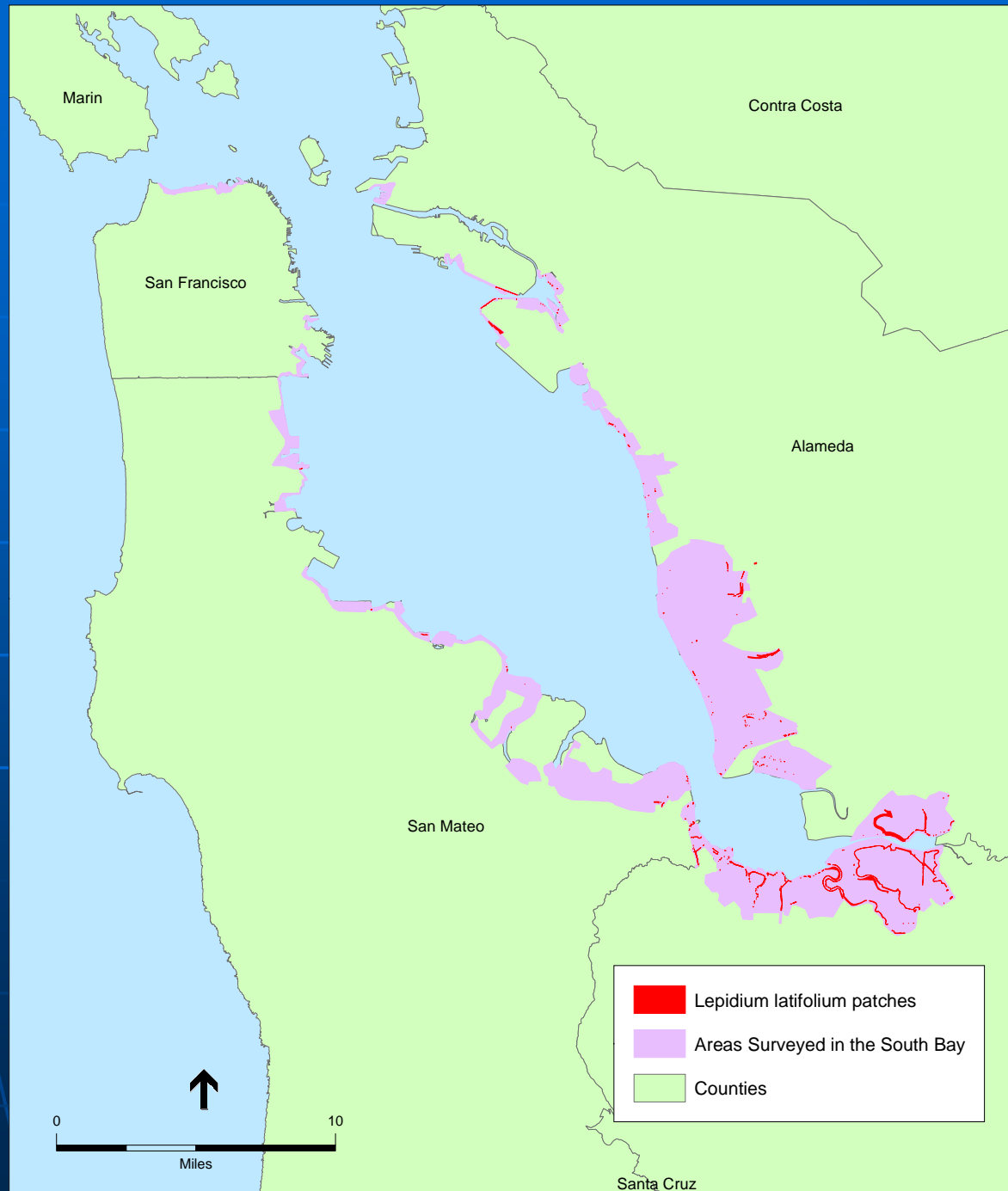
# Mapping the Shoreline

- GPS mapping method
  - the California Weed Mapping Handbook (CDFA)
- All patches larger than 1x1m were recorded.



Photo: ESA

# Patches of *L. latifolium*





# Predictive Modeling

- Predict probabilities of occurrence or spatial distribution of species.
- Gain insight into species/environment relationships.
- Assess risk on local and/or regional scale.

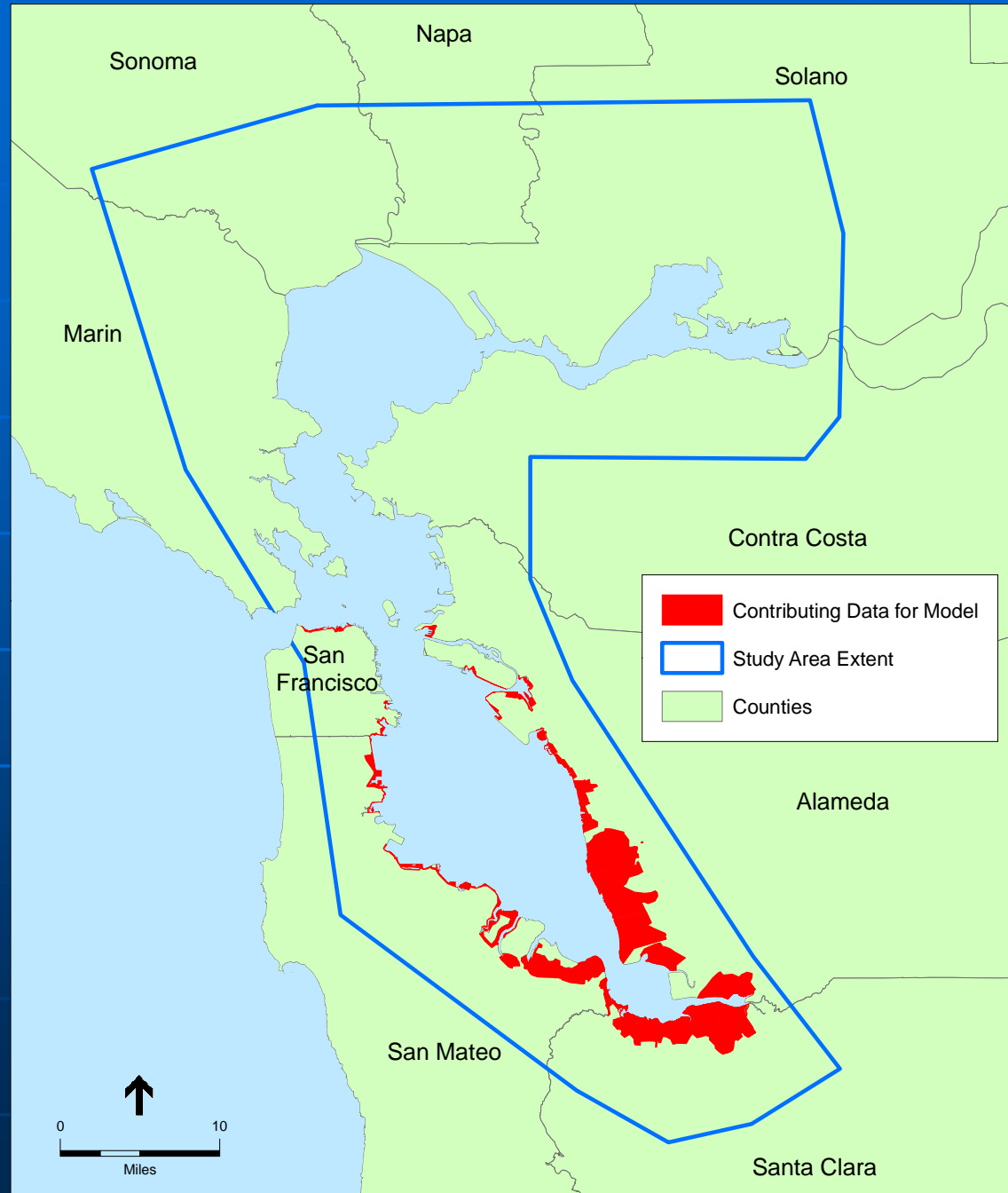


# Predictive Modeling

1. Dependent variable: known presence/absence of species
2. Independent variables: environmental factors
3. Relate variables to distribution using a statistical model (Binomial logistic regression)
4. Transform model into a GIS probability map.

# Study Area

- Randomly selected 500 presence points and 500 absence points within areas surveyed.



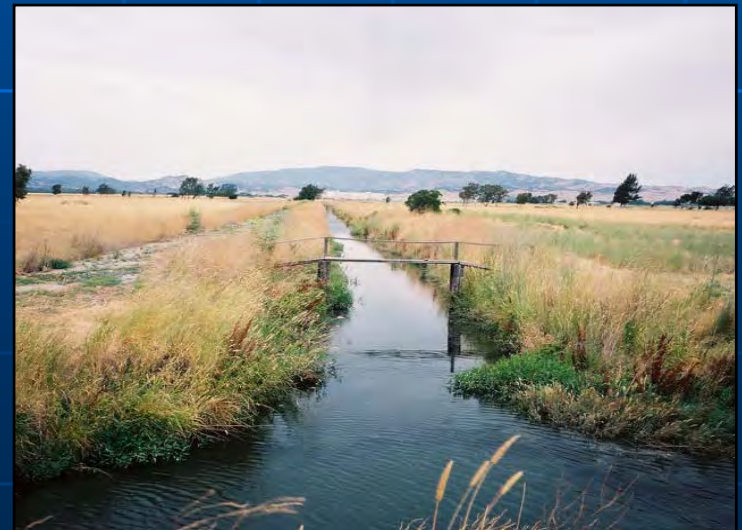
# Transferability

- Making predictions outside of the area in which the model was developed.
- Difficulties...
- Model valid for area surveyed.
- Model can provide guidance and risk assessment for outside areas.



# Environmental Predictor Variables

- Habitat type
- Tidal regime
- Elevation
- Distance to open water
- Distance to road
- Distance to levees
- Distance to agricultural land



# Spatial Extents

## Spatial Extent 1



## Spatial Extent 2



All variables considered

Elevation and Tidal Regime excluded

# Results and Discussion



Photo: ESA, Pittsburg, CA

# Individual Variables: Nagelkerke R<sup>2</sup>

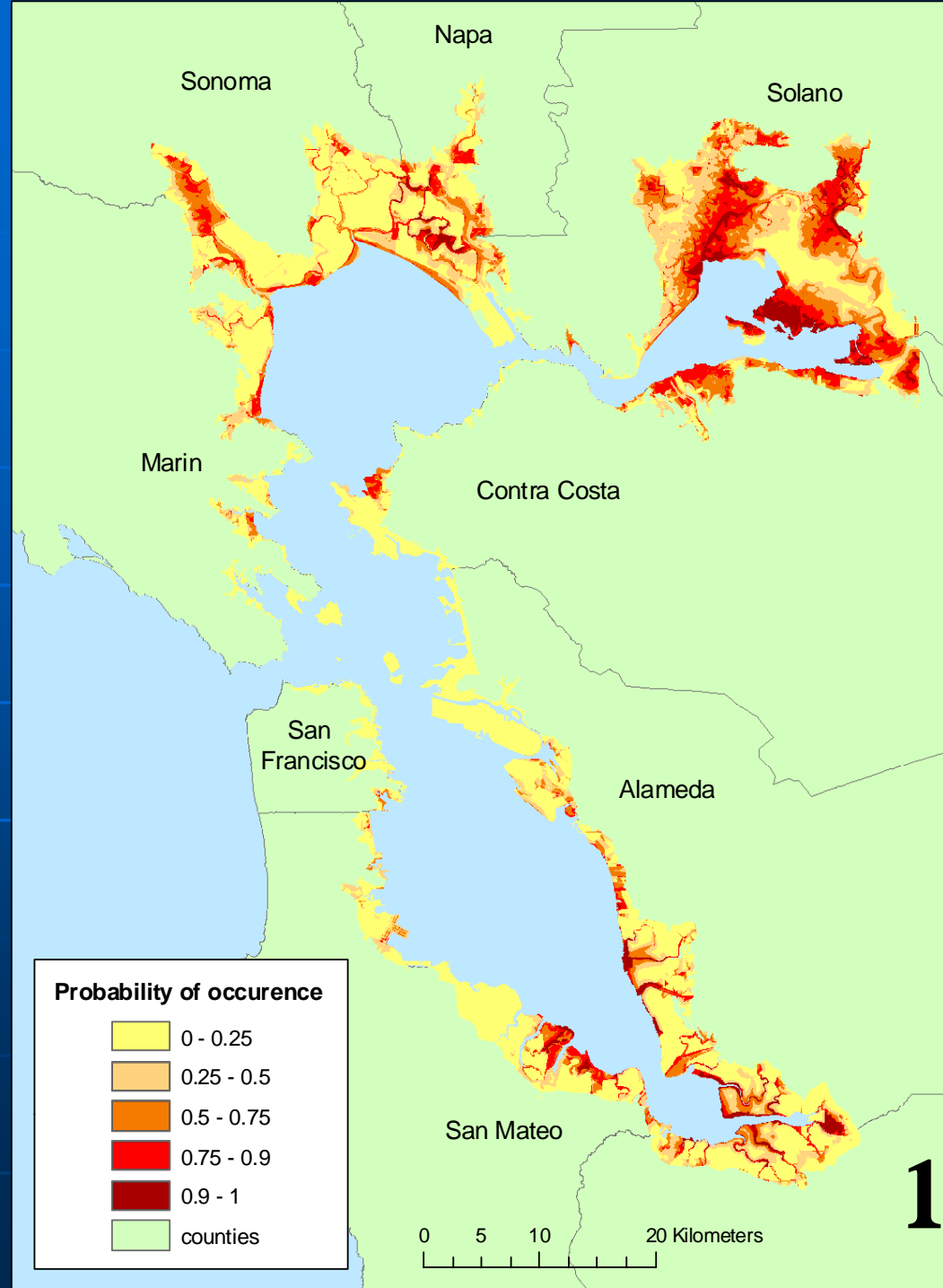
Variable	Nagelkerke R <sup>2</sup>
<b>Distance to <i>L. latifolium</i> patch</b>	<b>0.92</b>
<b>Tidal, tidal</b>	<b>0.44</b>
<b>Cover, estuarine wetland</b>	<b>0.42</b>
<b>Cover, water</b>	<b>0.36</b>
<b>Tidal, diked</b>	<b>0.36</b>
<b>Bay habitat, salt ponds</b>	<b>0.32</b>
<b>Bay habitat, marsh</b>	<b>0.31</b>
<b>Distance from water</b>	<b>0.25</b>
Distance from paved roads	0.094
Bay habitat, developed	0.04
Distance from levee	0.036
Cover, high-intensity development	0.028
Cover, palustrine wetland	0.021
Tidal, muted	0.013
Bay habitat, water	0.012
Tidal, non-tidal	0.008
Bay habitat, mud flat	0.007
Distance from agriculture	0.006
Cover, other estuarine habitat	0.002
Bay habitat, other	0.001
DEM	0.001
Cover, low-intensity development	0.001
Cover, other	<0.001
Cover, bare land	<0.001
Cover, grassland	<0.001



# Spatial Extent 1

## Significant Variables

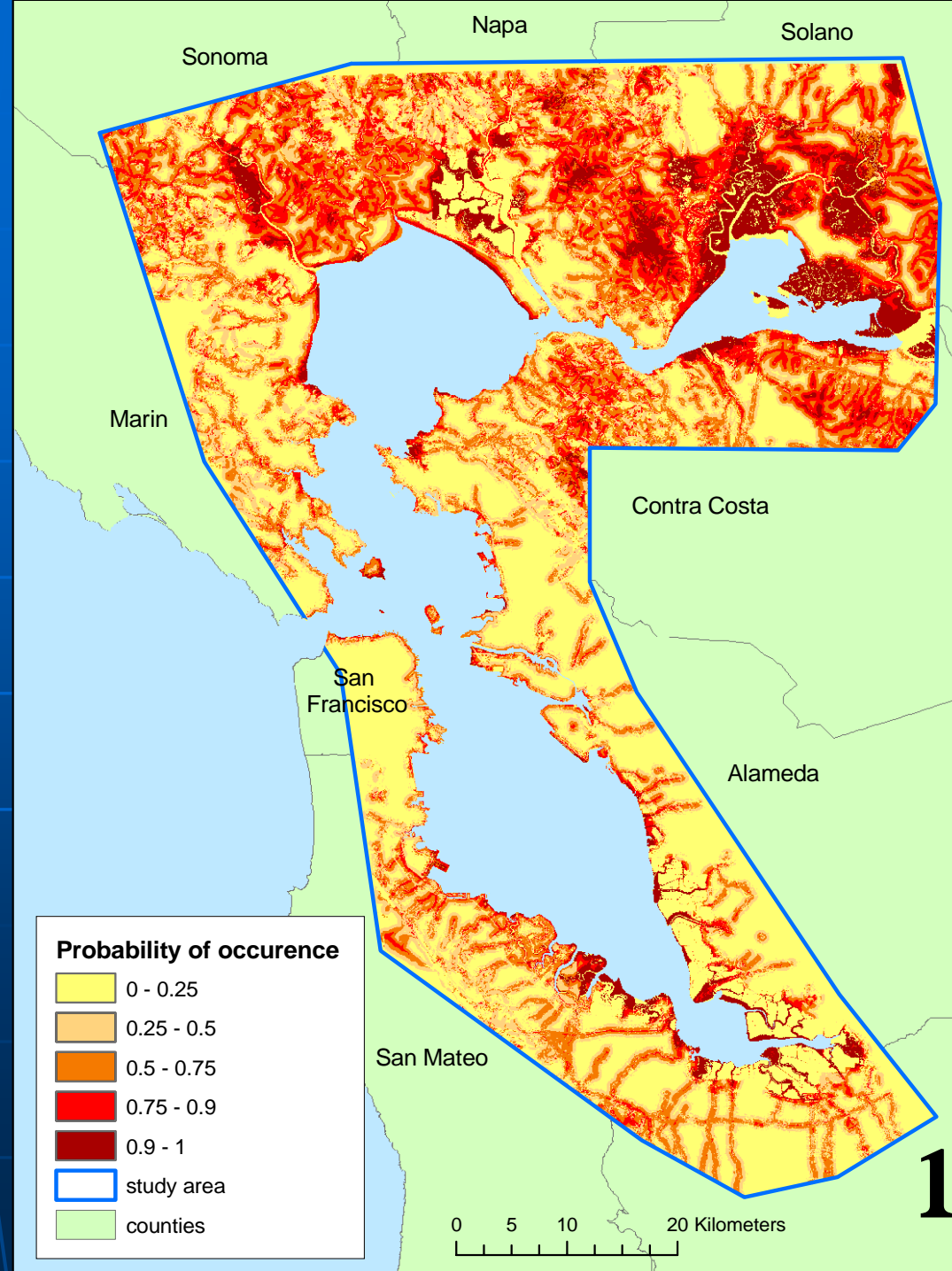
- Found in wetlands
- Outside of diked tidal areas
- Closer to water and levees
- Further from roads



# Spatial Extent 2

## Significant Variables

- Found in wetlands
- Found in grassland, low-intensity devel. and bare ground
- Not in water
- Closer to water
- Further from road and agriculture



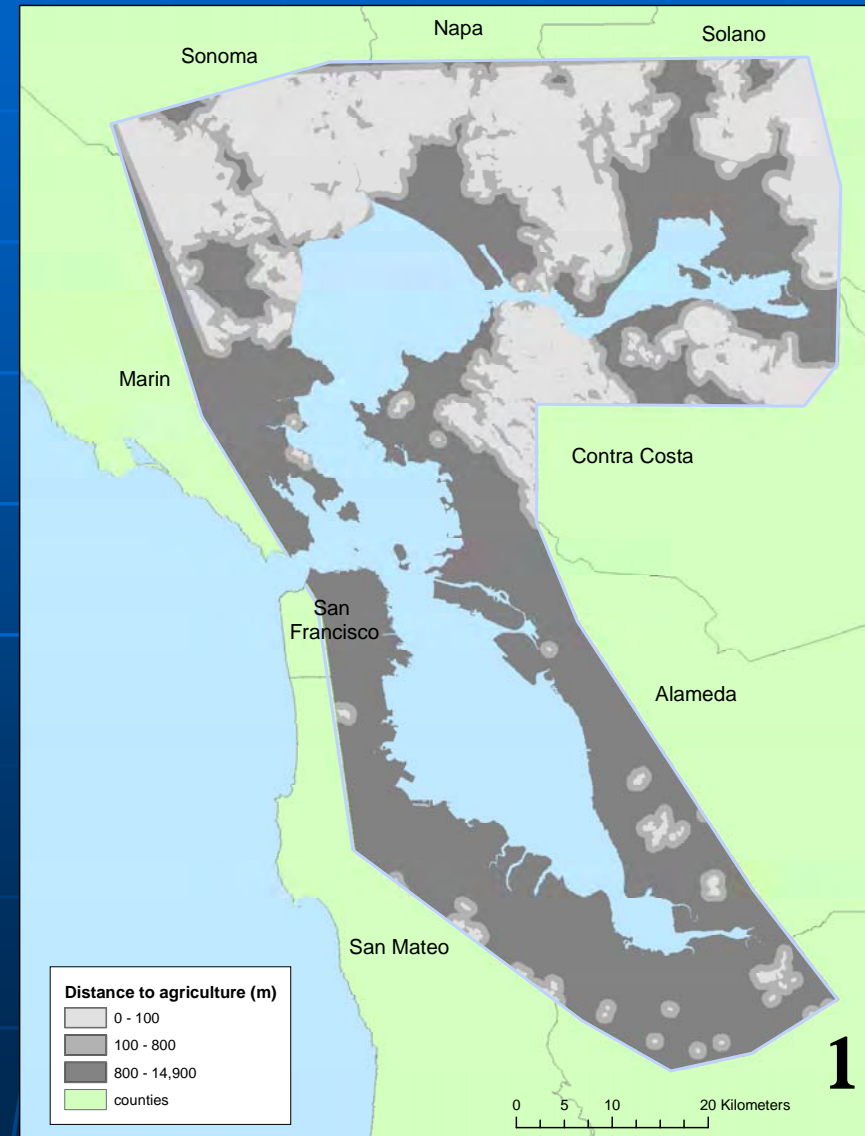
# Significant in Both Models

- Wetlands
- Shorter distance to water
- Longer distance to roads

# Unexpected Relationships

- Distance to Agriculture
- Distance to Roads
- Pattern may differ in San Pablo Bay and Suisun Marsh.

Variable, Distance to Agriculture →



# Internal Accuracy Tests (SPSS)

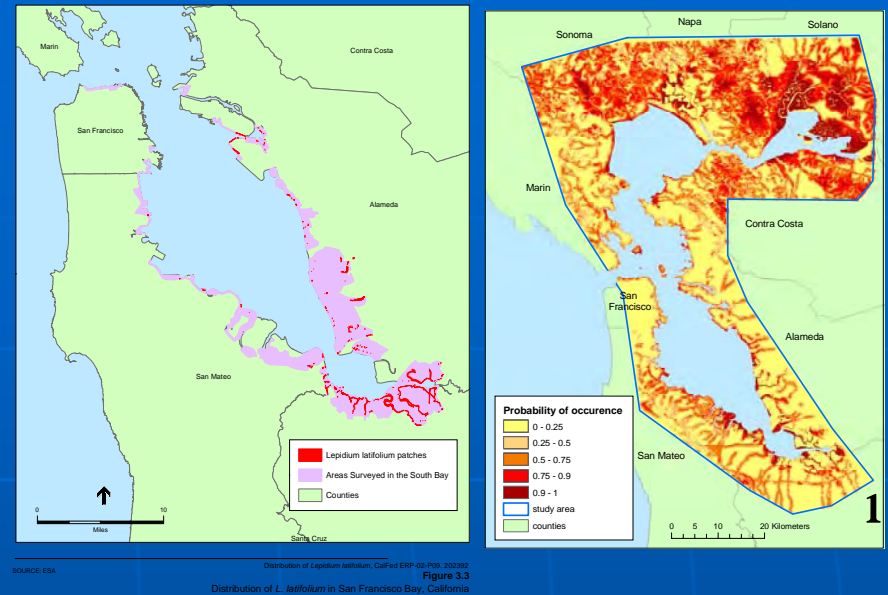
- Nagelkerke  $R^2$ 
  - **Model 1:** 0.542
  - **Model 2:** 0.623
- Classification Table
  - **Model 1:**
    - 80.4% absent correct
    - 82.8% present correct.
  - **Model 2:**
    - 79.3% absent correct
    - 87.4% present correct.



Photo: Melanie Vanderhoof

# External Accuracy

(Random 30% of original data)



1. The percent of cells coded correctly (either present or absent).
2. The percent of absent cells correctly coded.
3. The percent of present cells correctly coded.
4. The number of present cells correctly coded compared to the total number of cells predicted to be present (as a percent). (*# of false positives or over-predicting presence*)

# External Accuracy Findings

## ACCURACY OF BINARY LOGISTIC REGRESSION MODELS AT DIFFERENT CUTPOINTS

Spatial Extent	Cutpoint	1(%)	2(%)	3(%)	4(%)
1	0.5	86.01	85.9	93.28	9.23
1	0.75	92	92.16	91.5	15.22
1	0.9	98	98.74	50.2	37.97
2	.5	65.06	64.62	93.85	3.87
2	0.75	77.68	77.5	89.48	5.69
2	0.9	89.05	89.4	68.12	8.88

1. The percent of cells coded correctly (either present or absent).
2. The percent of absent cells correctly coded.
3. The percent of present cells correctly coded.
4. The number of present cells correctly coded compared to the total number of cells predicted to be present (as a percent).

# Models: extent 1 vs. extent 2

- Model 2 tested slightly higher on internal accuracy.
- Model 1 tested better with external accuracy.
- Tidal regime categories.



# Weaknesses of Models

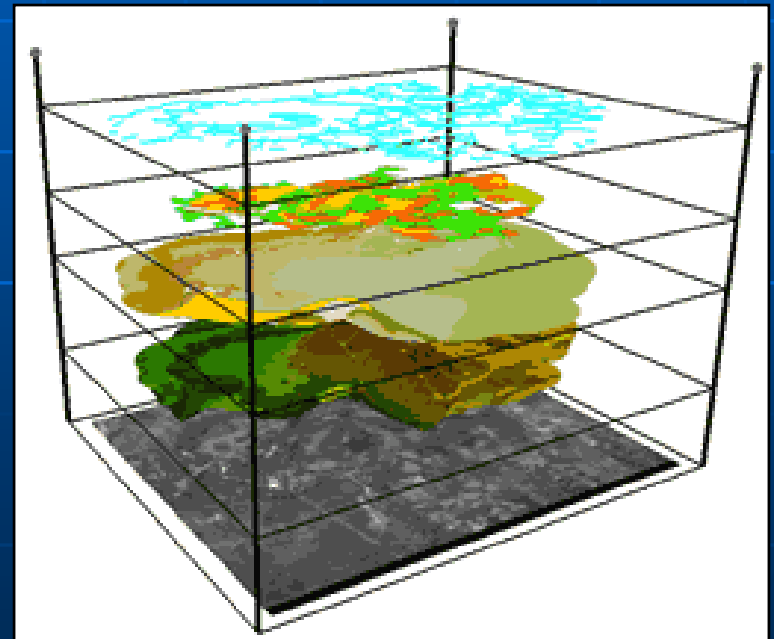
- Resolution not ideal (limits accuracy and precision of relationship)
- Additional variables
- Site specific relationships



Photo: ESA

# Threats to Validity

- Absence data (hasn't spread yet)
- Conversions of patches to 30 m raster grid.
- Errors and inaccuracies within the data layers

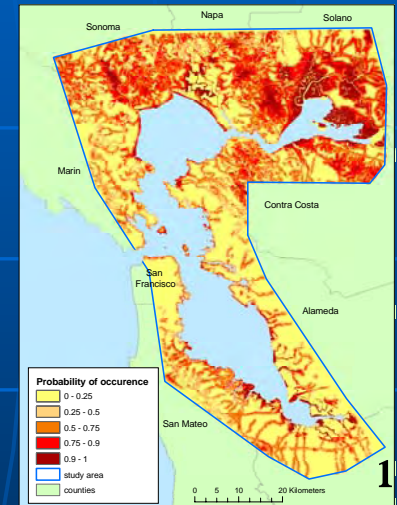
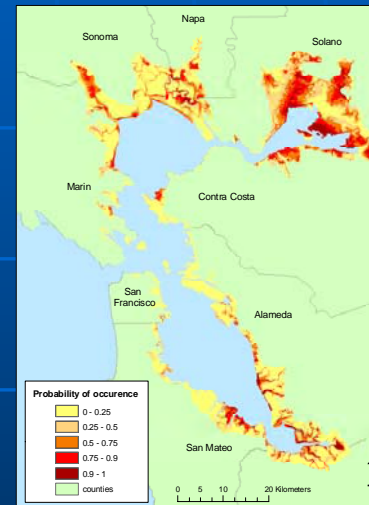


# Risk Assessment



Photo: ESA

- Is largely restricted to tidal marsh and riparian habitat.
- Predicted areas:
  - N. part of Suisun Bay
  - Grizzly Bay
  - Petaluma River
  - Napa-Sonoma marshes
  - Marshes in Don Edwards Wildlife refuge



# Risk and Public Lands

## **Distribution of medium, high and very high risk land:**

- Private land – 85% of medium risk land, 70% of very high risk land.
- CDFG manages the most amount of land at medium, high and very high risk.
- Department of Defense, U.S. Fish and Wildlife and local water districts.

## High Risk

- California State Lands Commission: >85% at high or very high risk
- Conservancy / Land Trusts ~ 60% at high or very high risk
- CDFG: >40% at high or very high risk of invasion

## Low Risk

- NASA 3% at medium, high or very high risk
- National Park Service: <3% at high or very high risk
- Open Space District ~4% at high or very high risk

# Management Recommendations

- Identify high risk lands.
- Control infestations early.
- Monitoring priorities:
  - Conservation/Restoration areas
  - Marsh habitat
  - Border between marshes and grasslands or low-intensity development.
  - Close to water



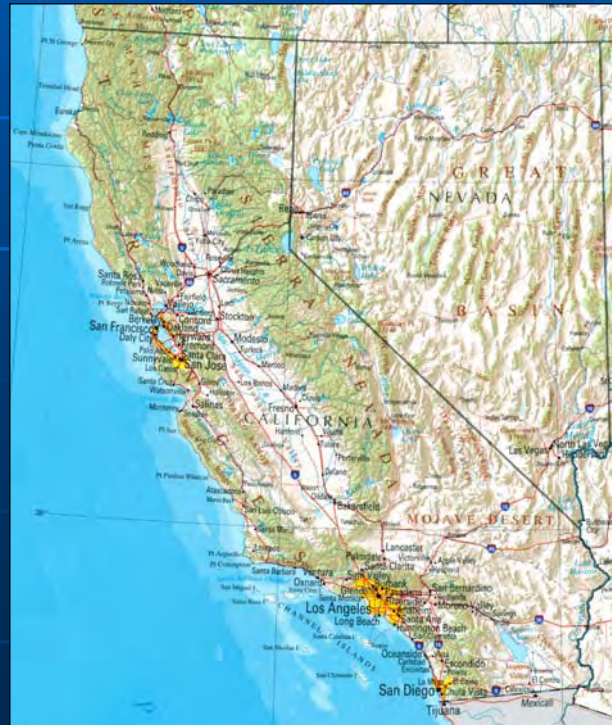
# Conclusions

- Mapping efforts established baseline distribution data.
- Within the S.F. Bay Delta:
  - *L. latifolium* prefers to grow within the tidal zone
  - In brackish and salt marshes
  - Close to water



# Conclusions

- 1<sup>st</sup> modeling attempt for *L. latifolium* in California.
- Significant relationships discerned in model can contribute to knowledge of *L. latifolium*.
- Modeling attempt shows promise for other invasive species.



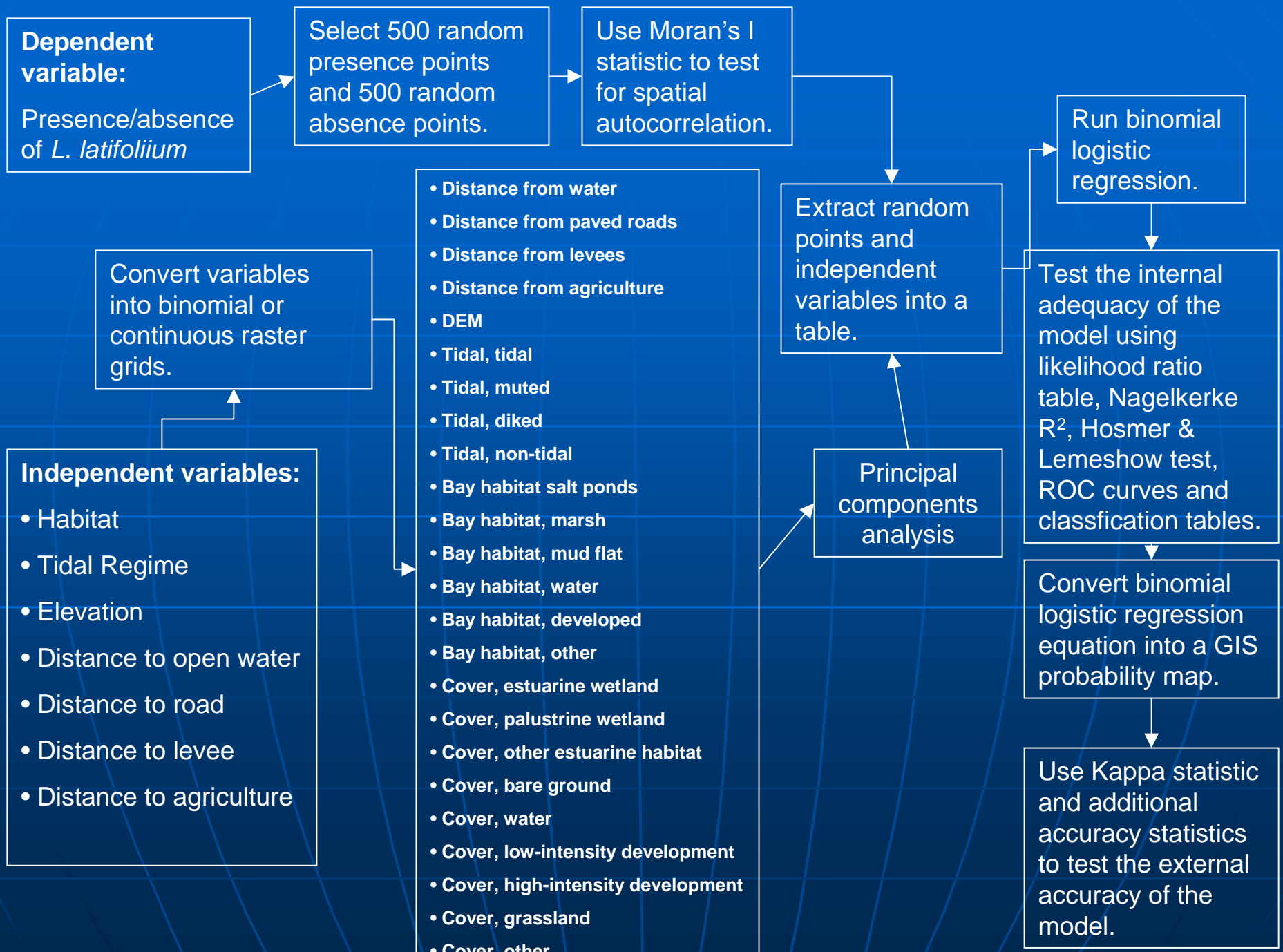
# Acknowledgements

- CALFED
- Graduate school advisors, Barbara Holzman and Ellen Hines (SFSU)
- Martha Lowe, Mark Fogiel, field work
- Jerry Davis, Barry Nickel and Casey Cleve, GIS assistance
- Clyde Morris (USFWS) and John Krause (CDFG), access
- ❖ GIS data will be available through CDFG's BIOS.



# Questions...





**Dependent variable:**

Presence/absence of *L. latifolium*

Select 500 random presence points and 500 random absence points.

Use Moran's I statistic to test for spatial autocorrelation.

Run binomial logistic regression.

Extract random points and independent variables into a table.

Test the internal adequacy of the model using likelihood ratio table, Nagelkerke R<sup>2</sup>, Hosmer & Lemeshow test, ROC curves and classification tables.

Convert binomial logistic regression equation into a GIS probability map.

Use Kappa statistic and additional accuracy statistics to test the external accuracy of the model.

Convert variables into binomial or continuous raster grids.

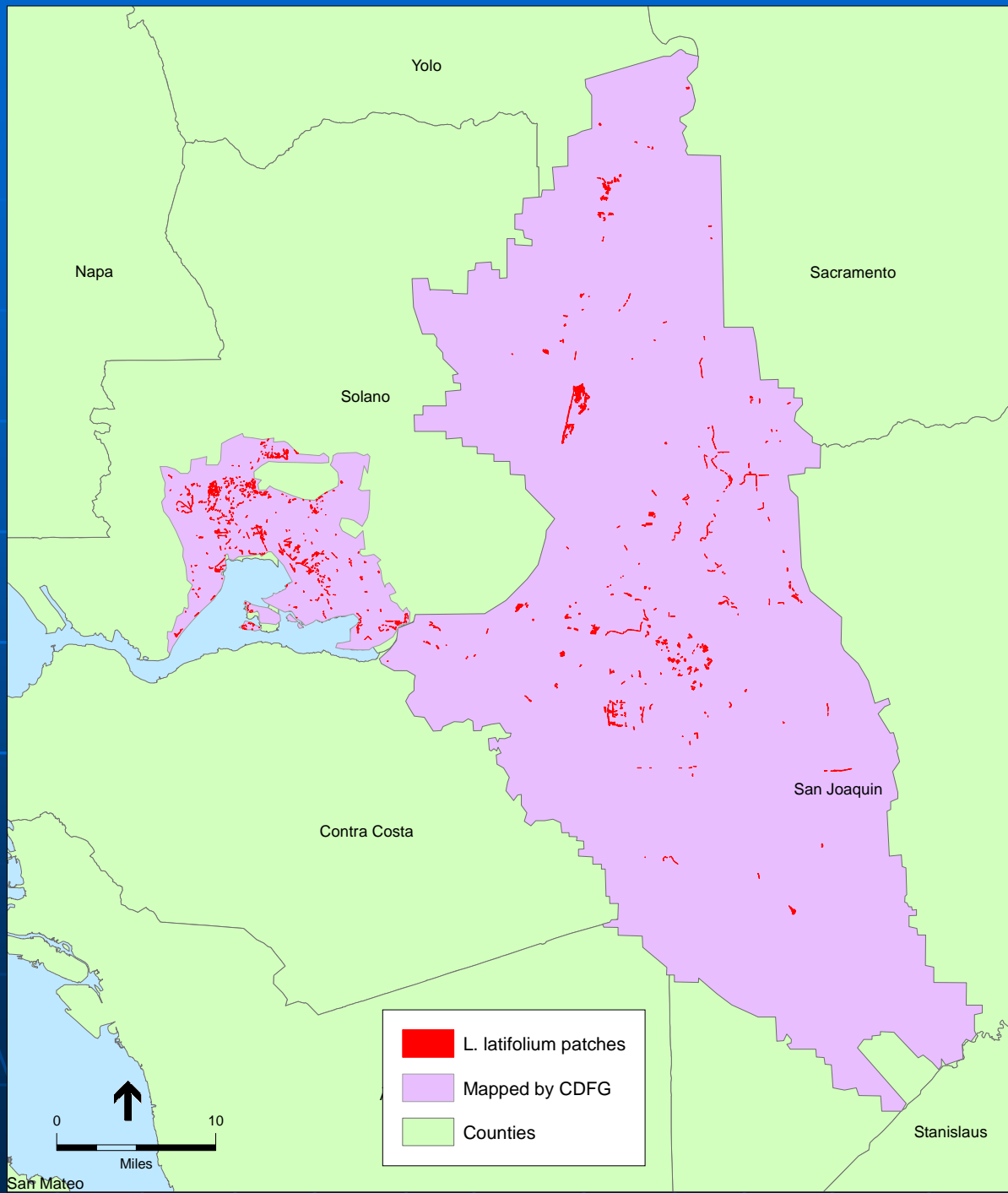
**Independent variables:**

- Habitat
- Tidal Regime
- Elevation
- Distance to open water
- Distance to road
- Distance to levee
- Distance to agriculture

- Distance from water
- Distance from paved roads
- Distance from levees
- Distance from agriculture
- DEM
- Tidal, tidal
- Tidal, muted
- Tidal, diked
- Tidal, non-tidal
- Bay habitat salt ponds
- Bay habitat, marsh
- Bay habitat, mud flat
- Bay habitat, water
- Bay habitat, developed
- Bay habitat, other
- Cover, estuarine wetland
- Cover, palustrine wetland
- Cover, other estuarine habitat
- Cover, bare ground
- Cover, water
- Cover, low-intensity development
- Cover, high-intensity development
- Cover, grassland
- Cover, other

Principal components analysis





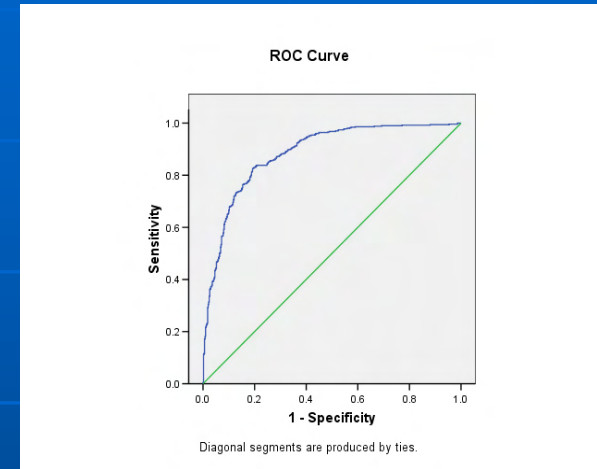
# Internal Accuracy Tests

Spatial Extent	Likelihood ratio table	Nagelkerke R <sup>2</sup>	Hosmer & Lemeshow	% Absent Classified Correctly	% Present Classified Correctly
1	<0.001	0.542	0.059	80.4	82.8
2	<0.001	0.623	0.631	79.3	87.4

# ROC Curves

**Table 12.** The area under the ROC curve and its statistical significance for both *L. latifolium* predictive models. The cutpoint was assumed to be  $P = 0.5$ .

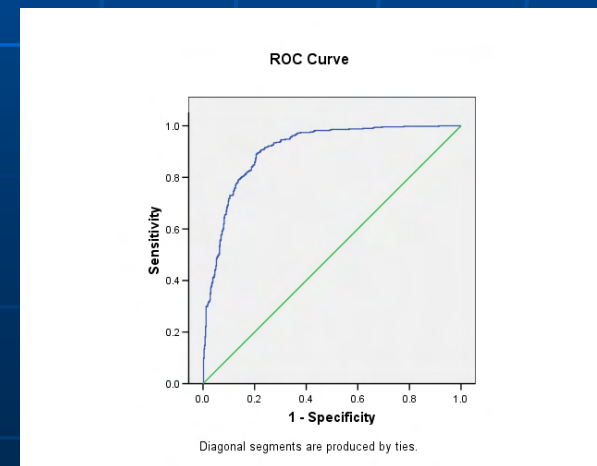
Spatial Extent	ROC Area	ROC Asymp
1	0.884	<0.001
2	0.907	<0.001



Spatial extent 1

**Table 13.** The sensitivity and specificity values on the ROC curve at different cutpoints. The goal is to maximize sensitivity and minimize specificity.

Spatial Extent	Cutpoint	Sensitivity	Specificity
1	0.5	0.826	0.196
1	0.75	0.679	0.104
1	0.9	0.217	0.012
2	0.5	0.874	0.206
2	0.75	0.714	0.1
2	0.9	0.31	0.02



Spatial extent 2