

Landscape scale risk assessment of cyanobacteria blooms in California lakes

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Unprecedented freshwater HAB events recently in California

- New record high toxin concentrations
 - ▶ Multiples toxins detected simultaneously





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Unprecedented freshwater HAB events recently in California

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 - ▶ Annual dog deaths attributed to cyanotoxins
 - ▶ Wildlife mortality events
- New situations and HAB organisms
 - ▶ Golden algae, *Pyramnesium parvum*
 - ▶ Ubiquitous and year round toxins
 - ▶ Cyanotoxins detected in marine shellfish and marine outflows





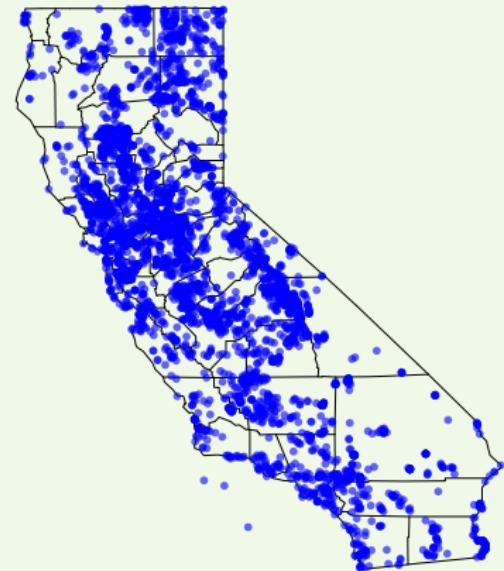
Lack of routine monitoring data for California lakes

Limited *in situ* data for risk assessment, lots of watershed data

NLA07, NLA12: 59 lakes



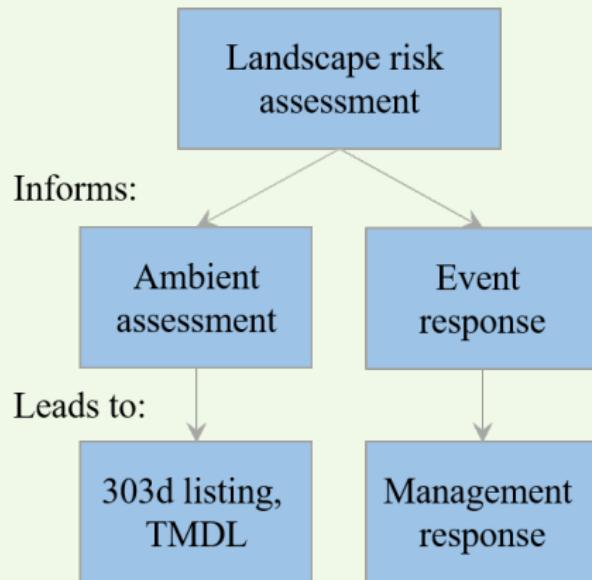
LakeCat: 4924 lakes



[USEPA, 2009, USEPA, 2017, Hill et al., 2018]

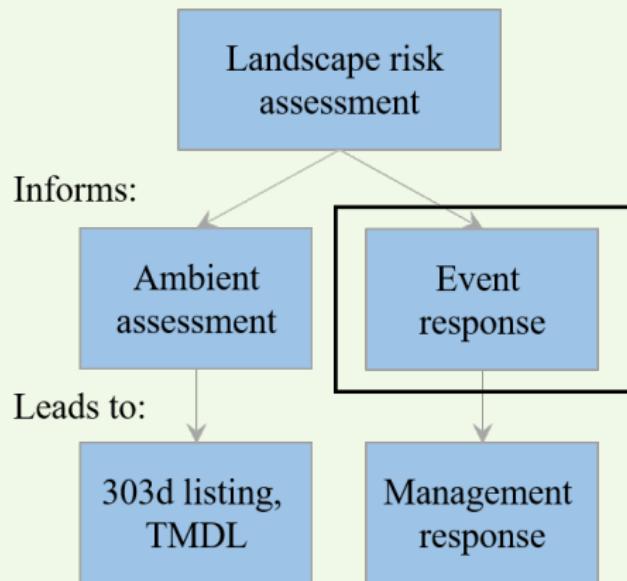


California has a strategy to develop a lake bioassessment program





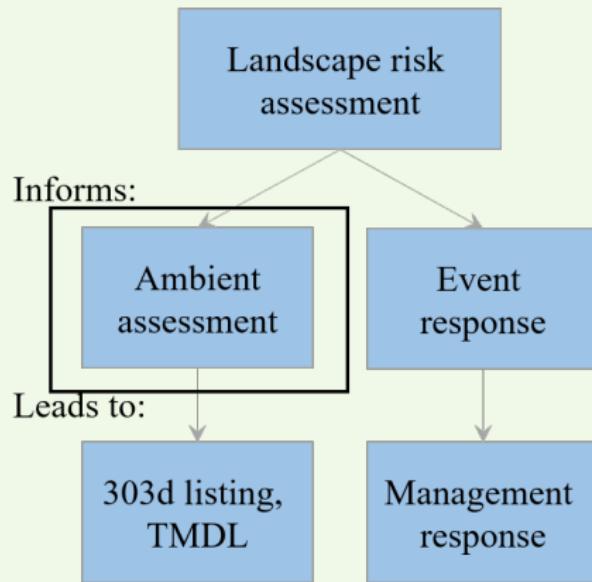
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- Some components developed...



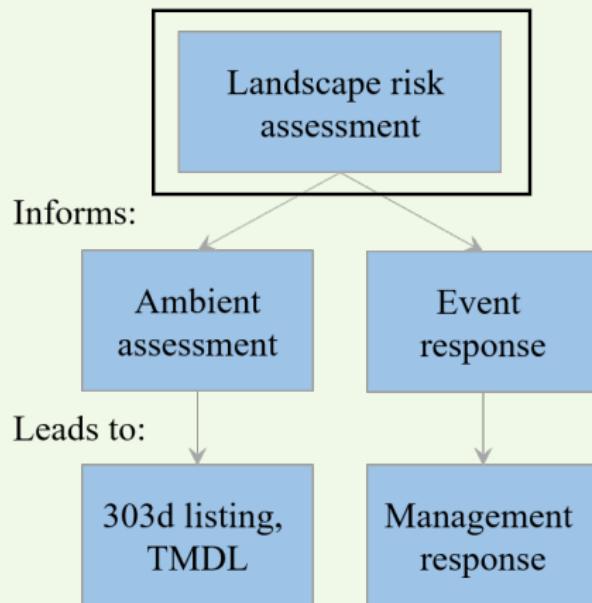
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- Some components developed...
- ...others are not
- Landscape screening can inform the process



Alternative data sources

Remote sensing limited to large lakes, 95 in CA



[https://www.epa.gov/water-research/](https://www.epa.gov/water-research/cyanobacteria-assessment-network-cyan)

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Goal: develop statewide screening tool to evaluate the relative risk of HABs in order to prioritize lakes assessment



Develop a landscape screening tool for microcystin risk

A four-step approach to assess risk from a limited dataset:

1. Develop link between chlorophyll and microcystin from *in situ* data



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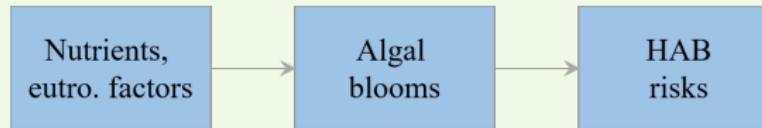


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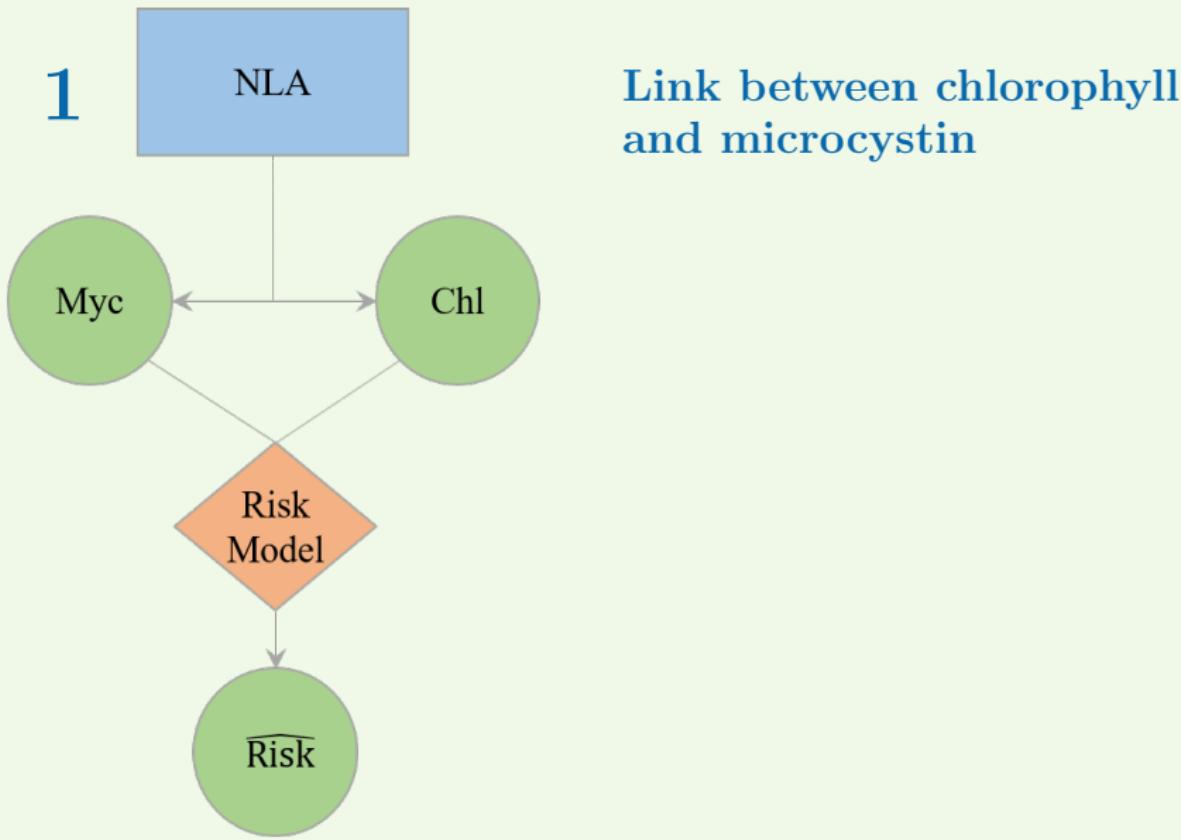
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Why chlorophyll?

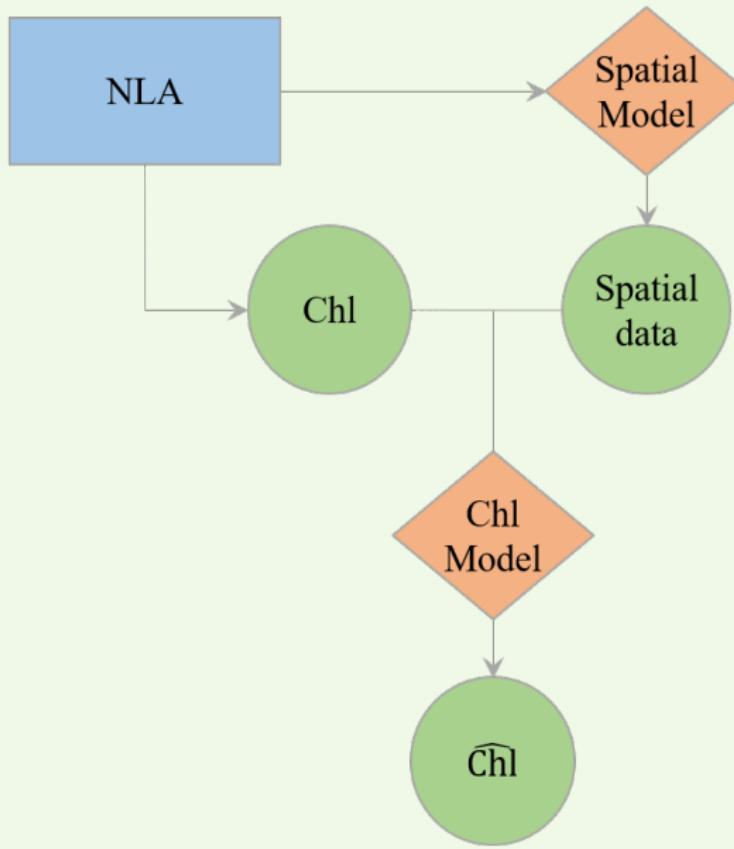


Modelling approach



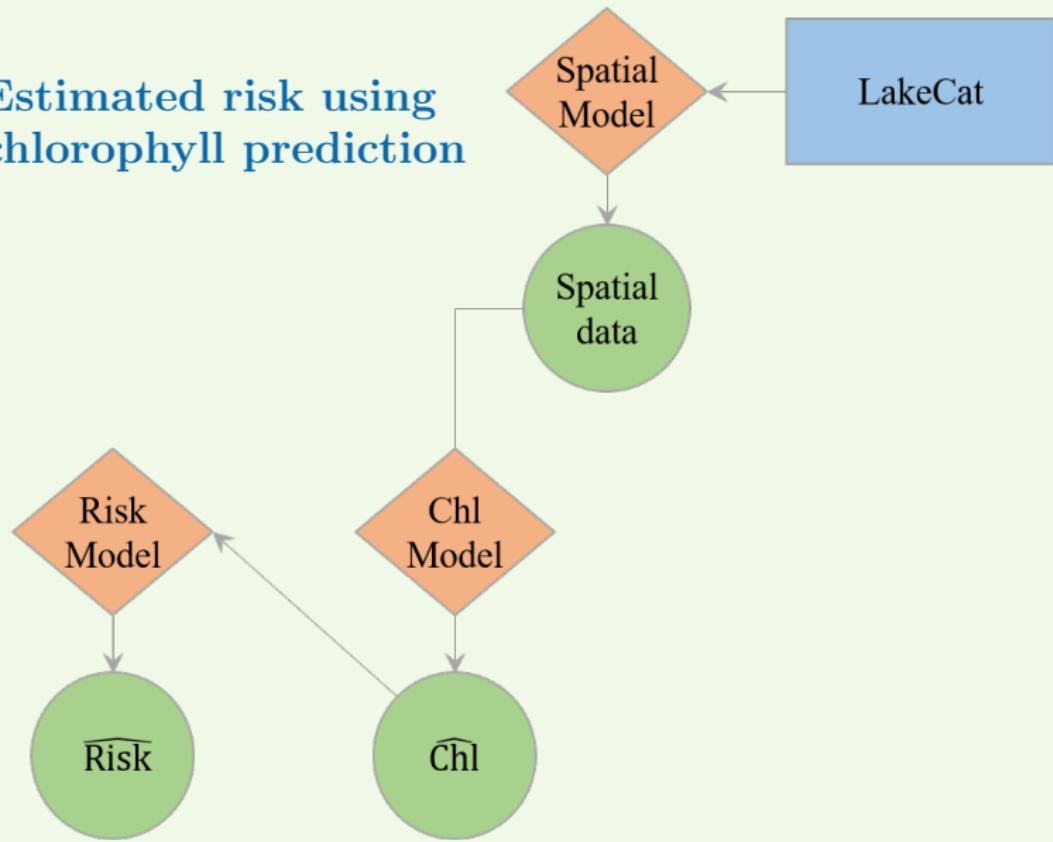
Modelling approach

2



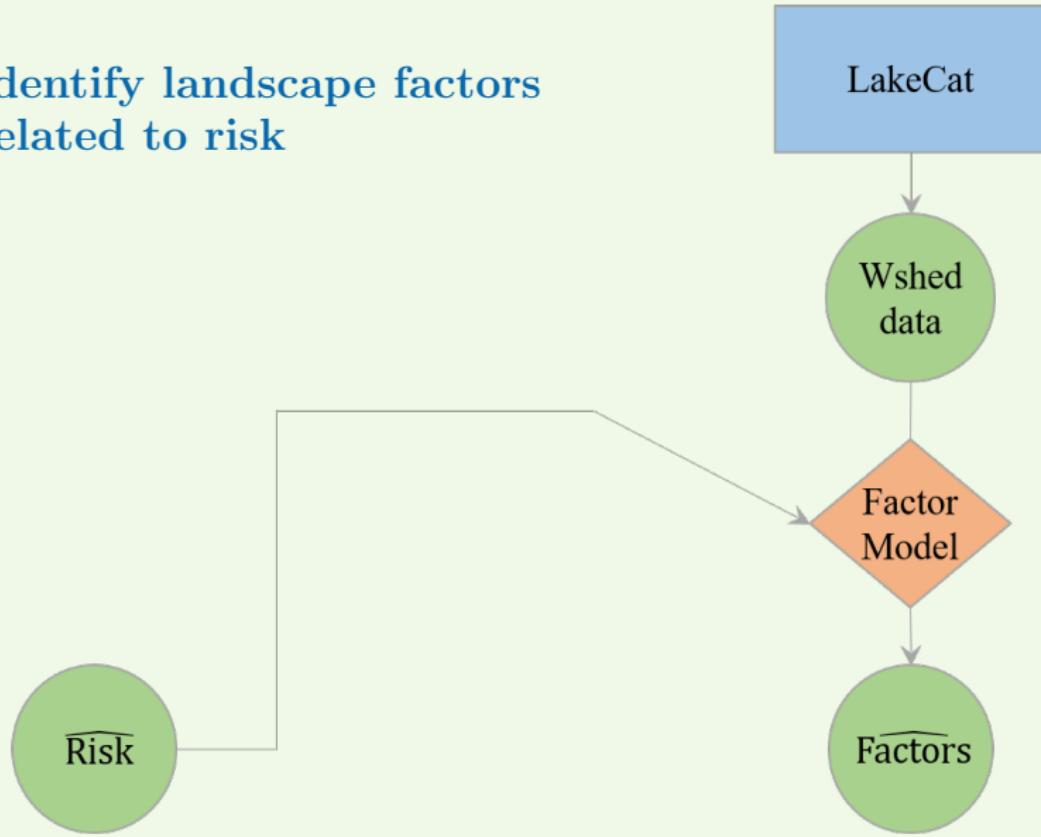
Link between chlorophyll and location

3 Estimated risk using chlorophyll prediction



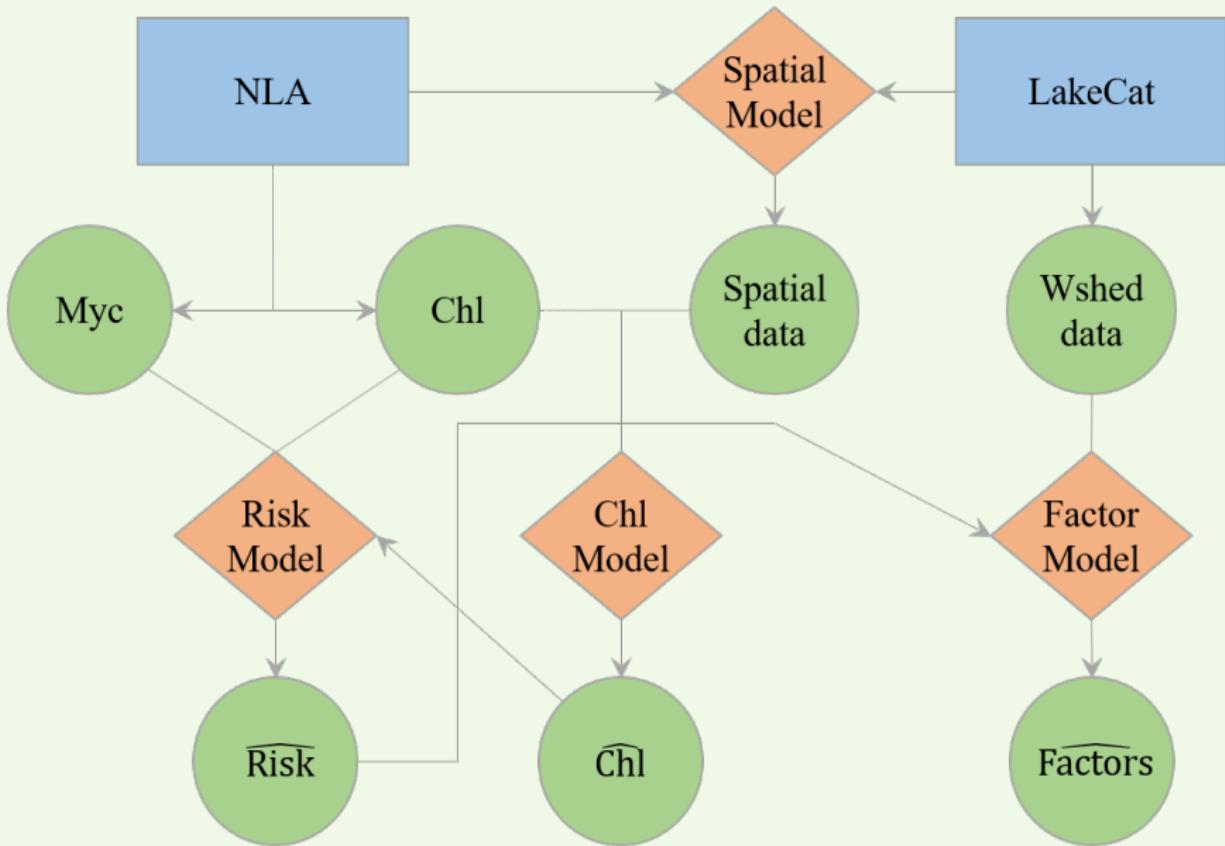
Modelling approach

4 Identify landscape factors related to risk



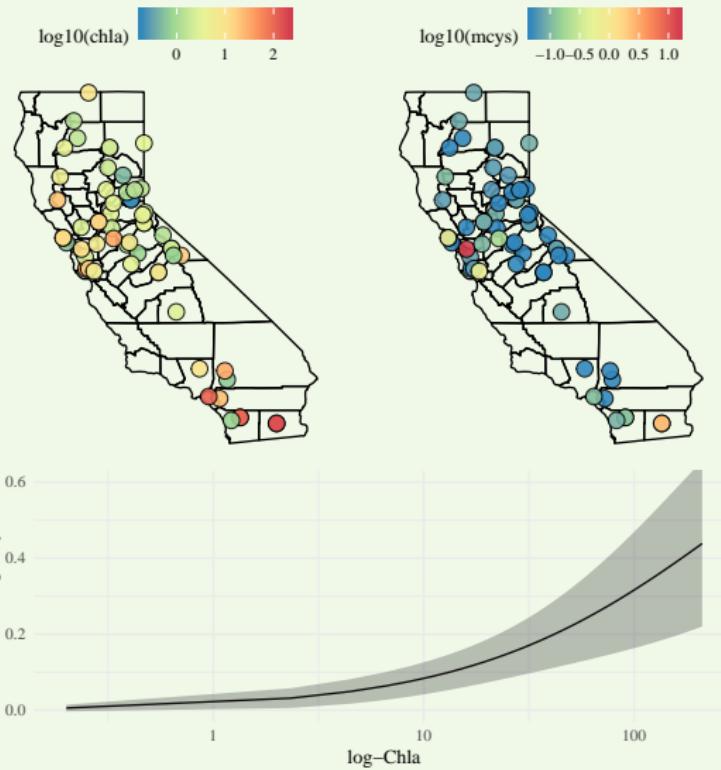


Modelling approach





1) Link between chlorophyll and microcystin

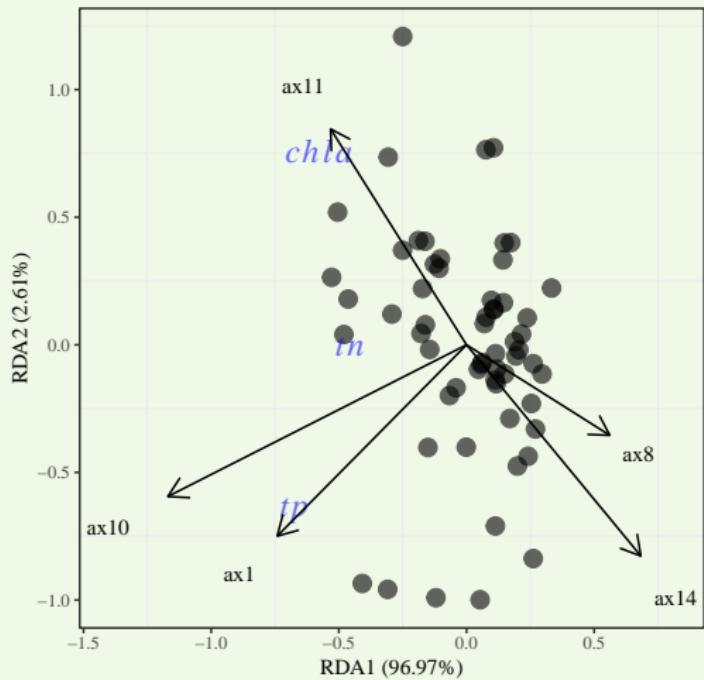
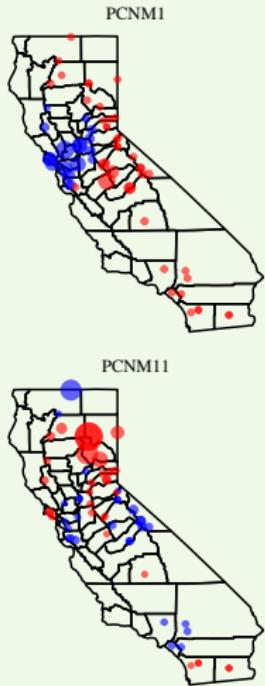


- *In situ* NLA data as probabilistic survey
- Build a simple model of the likelihood of exceeding some threshold
- Define a criteria threshold, arbitrary at this point



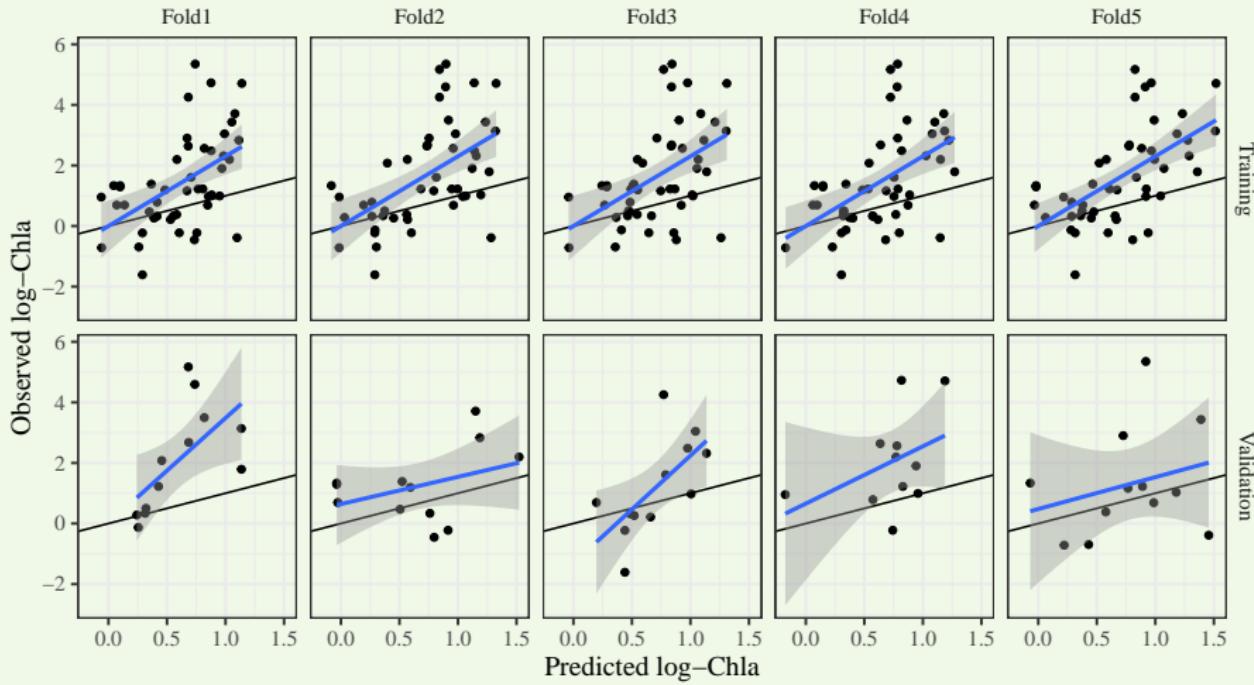
2) Link between chlorophyll and location

Using a spatial model to predict chlorophyll from lat/lon



2) Link between chlorophyll and location

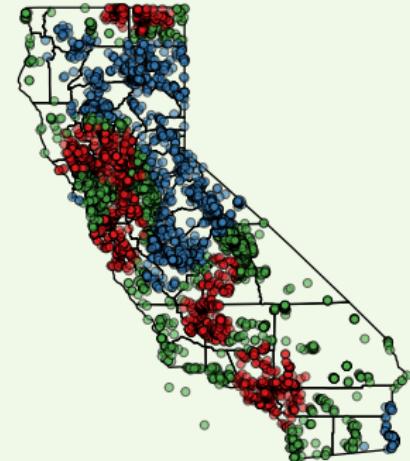
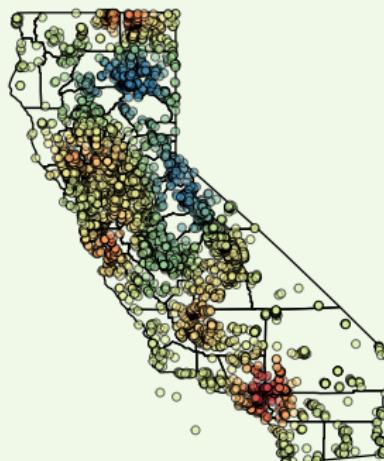
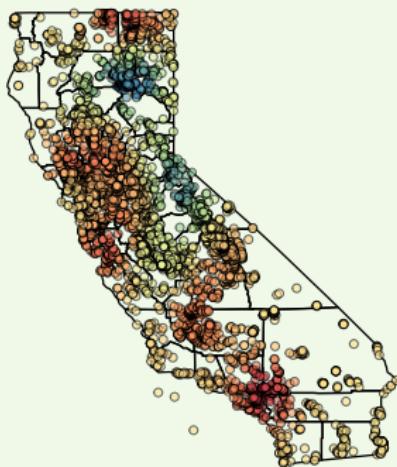
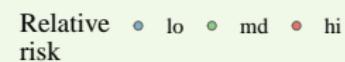
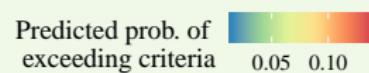
Predicted chlorophyll from location seems okay





3) Estimated risk from chla prediction

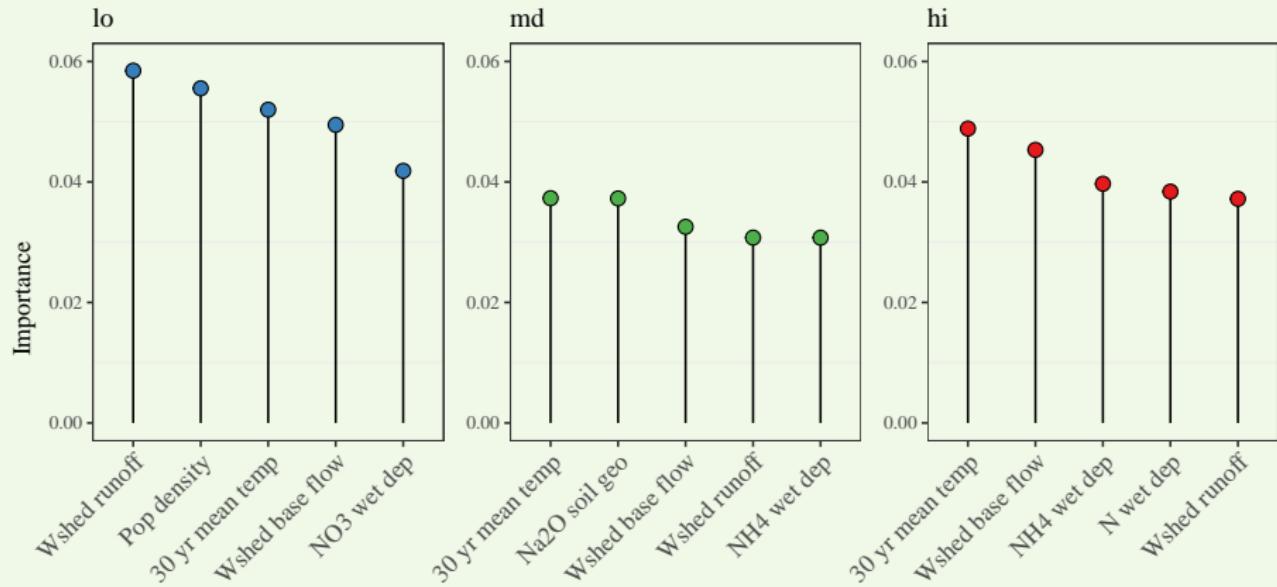
Use predicted chlorophyll to estimate probability of exceeding threshold, categorize relative risk





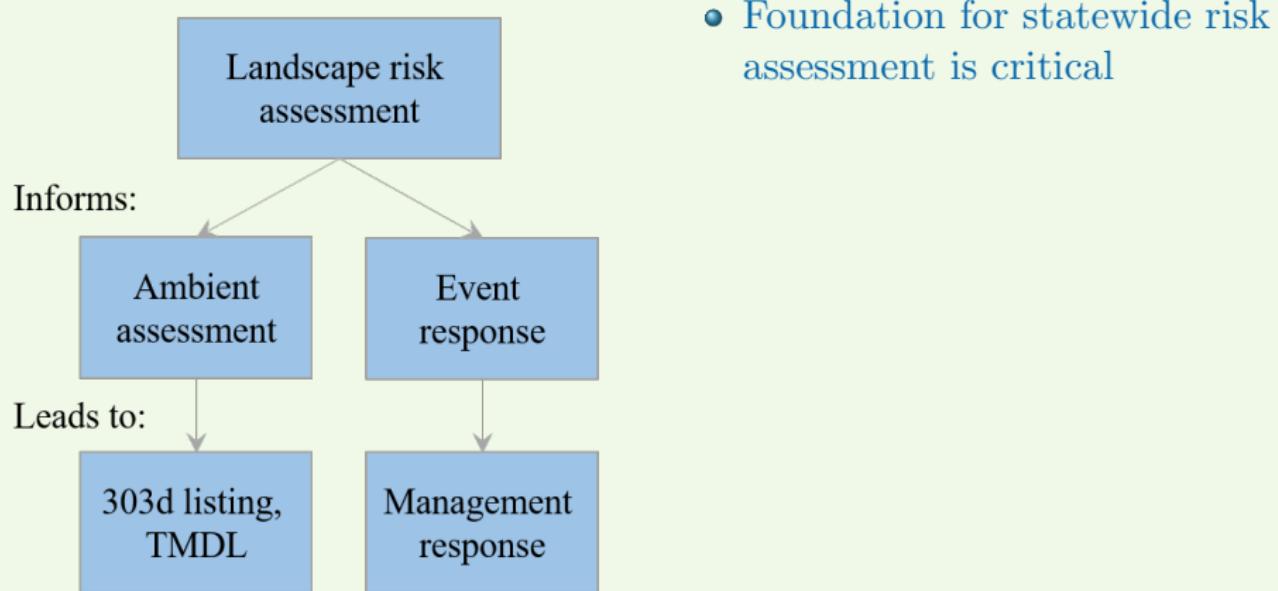
4) Identify landscape factors related to risk

Top five most important watershed factors linked to risk categories



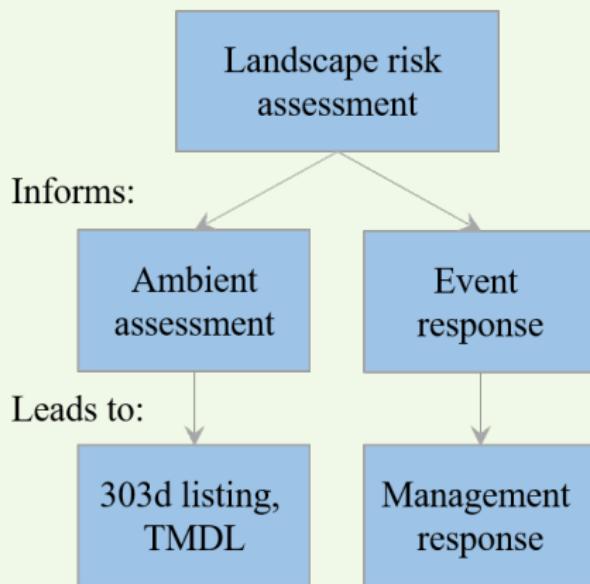


California has a strategy to develop a lake bioassessment program





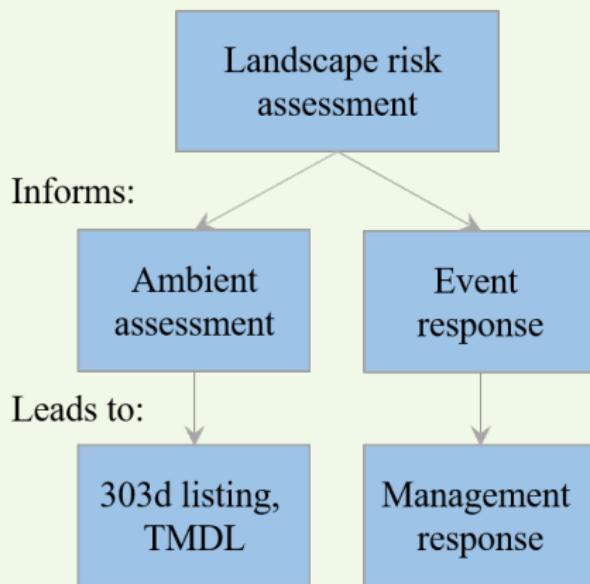
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- Foundation for statewide risk assessment is critical
- Leads to informed decisions for developing bioassessment program



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- Landscape position is a powerful predictor



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- Foundation for statewide risk assessment is critical
- Leads to informed decisions for developing bioassessment program
- Landscape position is a powerful predictor
- A potential for data poor situations

Acknowledgments:

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GitHub (project):
https://github.com/fawda123/cali_lake

GitHub (presentation):
https://github.com/fawda123/SFS_2018

Twitter: @fawda123

References

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