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Some Challenges in Classifying Habitats with Model Predictions using Hard Suitability Thresholds

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Talking Points

Habitat Suitability Analyses
 Role of Modeling
 Quantifying Habitat Suitability in Suisun Marsh
 Some Challenges



Habitat Suitability Analyses

Common analyses in restoration and rehabilitation projects

This talk: Habitat suitability for Delta Smelt in Suisun Marsh

□Goal

• to improve the recruitment, growth and survival of Delta Smelt

 by implementing actions designed to increase the quantity and quality of Delta Smelt abiotic habitat and food supply

 Various actions including hydrodynamic modeling to improve spatiotemporal resolution of habitat assessment



Habitat Suitability Analyses

Habitat Suitability Indices for Delta Smelt

Common parameters based on past studies

- Salinity
- Temperature
- Turbidity
- Current Speed



Role of Hydrodynamic Modeling

Metrics are developed from discrete measurementNumerical models fill in data in space and time





Bever et al. (2016)

For our work:

We considered

- Fixed Deterministic Thresholds
 - Salinity < 6 PSU
 - Turbidity > 12 NTU
 - Temperature < 24°C

• Bever et al. (2016) regression equations

- 0.67S + 0.33V, Secchi Depth < 0.5 m
- (0.67S + 0.33V) x 0.42, Secchi Depth > 0.5 m
- RMA (2021) adaptation to include temperature
 - (0.67S + 0.33V) x T



Suisun Marsh General Patterns - Desirable Good, Not Desirable







Quantifying Habitat Suitability SCHISM Model Predictions

-Honker Bay



Observed
SCHISM

NSE = 0.98 Bias = 0.30 RMSE = 0.6

Quantifying Habitat Suitability **SCHISM Model Predictions**

-Grizzly Bay





NSE = 0.97 Bias = 0.36 RMSE = 0.7 25

Challenges

- Time series in Aug 2020
- Just so happens...
- Also, 23.5°C vs 24°C vs 24.5°C ??



Temperature



Cloud Cover



https://weatherspark.com/

Cloud Cover at Antioch





https://weatherspark.com/

How to address these issues?

Depends

 Hindcasts vs Forecasts



How to address these issues?

Hindcast

Correction over space and time

- Simple constant shift in predictions across region using average bias
- Shift in predictions using a correction surface
- Newtonian relaxation of predictions at known locations towards observed values

Identifying cause and refining model inputs

• e.g. reducing incoming radiation in this example

•Using a not-so-hard classification approach

• Fuzzy classification??



How to address these issues?

Constant Shift





How to address these issues?

Identifying cause and refining model inputs – reducing incoming radiation



How to address these issues?

Remedies will work well for historical runs

 What to do for scenarios with no observations/ forecasting cases??



How to address these issues?

• What to do for simulating cases with no observations??

• Impossible to predict conditions accurately

Understanding relative influence of parameters is key

- Could potentially base HSI on past observed values where necessary
- Assumption that temperature and turbidity not significantly affected by ops



How to address these issues?

Forecasting example

Current speed and salinity from SCHISM Model

 Quantiles of past temperature and turbidity interpolated to model grid used

- Suitability assigned based on quantiles,
 - e.g.: Example on next slide



Sample Suitability Index, S_i, Estimation

Based on Bever et al. (2016)

 $S_i = 0.67S + 0.33V$, turbidity > 12 NTU $S_i = (0.67S + 0.33V) \times 0.42$, turbidity < 12 NTU

where *S* is based on the fraction of time salinity < 6 PSU (computed with SCHISM) *V* is based on the maximum current speed (computed with SCHISM)

We weighted equations based on turbidity quantiles

For example, if the 75% quantile was 12 NTU, the suitability index was calculated as:

 $S_i = 0.75 \times [(0.67S + 0.33V) \times 0.42] + 0.25 \times [0.67S + 0.33V]$

• We then adjusted S_i with factors based on temperature quantiles

 $S_{i_{final}} = 1.00 \times S_{i_{final}}$ $S_{i_{final}} = 0.75 \times S_{i_{final}}$

if 75% quantile < 24°C if 50% quantile < 24°C ≤ 75% quantile

Evaluation of approach





Comments

 It can be a challenge to paint an accurate picture of habitat suitability without ACCURATE predictions when hard thresholds are used

 Softer classification/threshold approach is likely to provide more robust estimates of suitable habitat



Questions?



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Issues with hard thresholdsPhysical

Suitable Salinity???

* 5.95 PSU vs 6.0 PSU vs 6.05 PSU ????





$$SI_c = \frac{Normalized \ Percent \ Catch + Normalized \ Total \ Catch}{2}$$





Issues with hard thresholds

- Physical
 - Suitable Salinity???





Evaluation of approach (Bever et al)



Figure 8 Two-dimensional maps of the station index (SI_H) based on (A) the predicted percent of time the depth-averaged salinity was less than 6 psu; (B) the maximum depth-averaged current speed; (C) percent of time the depth-averaged salinity was less than 6 psu and the Secchi depth threshold; and (D) the salinity and velocity metrics with the Secchi depth threshold.



Nobriga et al (2008)



Temperature



Cloud Cover



https://weatherspark.com/