# A life cycle model and population viability analysis for wild delta smelt 

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## What affects recruitment and survival in the wild?



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1) Standardize survey data

2) Develop and fit a life-cycle model


## 1) Standardize survey data

Need abundance estimates for each life-stage from many cohorts

Cohorts: 1995-2012


Life-stage
Juveniles
Subadults
Adults

Survey
20mm
FMWT
SMWT (1996-2001)
SKT (2002-2013)

Month
June
Nov (modeled)
Jan/Feb (of year+1)

## Observations: Design-based abundance estimates

Stratified ratio expansions
Total abundance = Sum of subregion abundances
Subregion abundance = Subregion Catch density * subregion volume


## Catch density =

## Sum of adjusted catch

## Sum of effective volume sampled

Adjusted catch=Catch/Prob(catch)
Adjust according size of fish caught

Effective volume adjusts volume by:
Gear Selectivity Curves
i) How much water was sampled in the top 4 m of water

Oblique vs. surface tows
ii) How the density of fish is assumed distributed within this 4 m slice

## Observed subadult abundance is modeled

Under the hood


Currently using a FMWT specific state-space model to further account for relative FMWT bias

Still, 2005 cohort: Subadult $=374,726$, Adult=480,448

## Covariate data

## Recruitment (9)

Food, outflow, X2 location, previous adult size, OMR, water temperature, temperature, predator abundance indices (ISS and striped bass)

## Juvenile survival (8)

Food, outflow, X2 location, predator abundance indices

## Subadult survival (4)

Food, outflow, X2 location, OMR

## All together

9*8*4=288 different unique combinations of covariate triplets

## 2) Develop and fit a life-cycle model

Process (X)


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Survival ~ Logit-normal(mean=f(covariates) ,variance)
Recruitment ~ Log-normal(mean=f(covariates), variance)

## 2) Develop and fit a life-cycle model



Survival ~ Logit-normal(mean=f(covariates) ,variance)
Recruitment ~ Log-normal(mean=f(covariates), variance)

Hierarchical Bayesian state-space model
Allows for:
Covariates to influence recruitment and survival
Serial dependence in predicted abundances

Abundance estimate error

Fit:
Using Bayesian inference (JAGS)
Diagnostics look good

## Result from an "all flow" model

All vital rates depend on mean outflow




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## Evidence across models

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Good: Food, outflow, spawning adult size

No support: A water temp index

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Good: Food, outflow, temperature, age 0 striped bass


Bad: High X2, age 1 striped bass, inland silversides

## Adult survival

Good: Food, high X2

Bad: High outflow and OMR ???


## Good years and bad years

$$
\text { Population growth }=\lambda_{t}=\frac{\mathrm{N}_{\text {Adults }, \mathrm{t}}}{\mathrm{~N}_{\text {Adults },-1}}
$$

Good years: 1995, 1997, 1999, 2010, 2011, maybe 2006
Bad years: 1996, 2002, 2004, 2005, 2007, 2009, 2012
All flow model


## Unpacking good and bad years

 1995 Decent recruitment, high survival 1998 Poor recruitment, decent survival



1996 Great recruitment, poor survival 2005, 2009, 2012- Marginal recruitment and survival


## Growth rate comparison SSM vs. IBM of Rose et al.

Growth rate: - Negative ~1 Around 1 + Positive

|  | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 | 2003 | 2004 | 2005 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| All flow model | + | - | + | + | + | - | $\sim 1$ | - | $\sim 1$ | - | - |
| Rose et al. | + | + | - | + | + | - | - | - | - | - | - |




## 3) Population viability analysis

Simulate future abundances using a fitted model

2 future scenarios:

1) The future is stochastically similar to the past
2) What if spring or summer never experience high flows?

## Percent of 10,000 simulations declining

1) When the future is like the past


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All 288 models


## Percent of 10,000 simulations declining

1) When the future is like the past


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2) When the future is constrained, all flow model

- Low flows
- Medium flows
- High flows

March-June flows


## Percent of 10,000 simulations declining

1) When the future is like the past


All 288 models

2) When the future is constrained, all flow model

- Low flows - Medium flows - High flows

March-June flows


June-Nov flows


## Many elephants

Model is simple, data is noisy, no larva life-stage, drought,...

1) Flow matters
2) Perhaps more so for recruitment than survival
3) Good vital rates for all life-stages needed to realize positive population growth


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