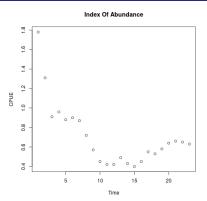
# Bias Estimation of Biological Reference Points Under Two-Parameter SRRs

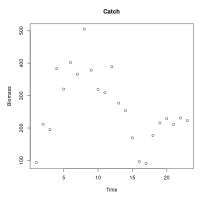


Nick Grunloh

14 March 2022







$$I_t = qB_te^{\epsilon} \quad \epsilon \sim N(0, \sigma^2)$$

$$\frac{dB(t)}{dt} = P(B(t); \theta) - C(t)$$



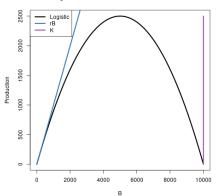
Introduction •000

## Schaefer Model

Introduction 0000

$$P_{\theta}(B) = rB\left(1 - \frac{B}{K}\right)$$
  
 $\theta = (r, K)$ 

#### Logistic Production and Related Quantities





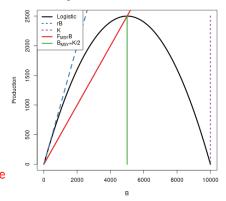
## Schaefer Reference Points

$$F^* = \frac{r}{2}$$

$$\frac{B^*}{B_0} = \frac{1}{2}$$

$$MSY = \frac{rK}{4}$$

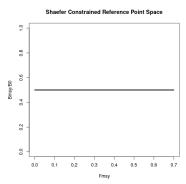
Some words on RPs, aybe some citations



Logistic Production and Related Quantities

Introduction

Something about Model misspecification of RPs Three parameter SRR functional form with oprthogonal RPs



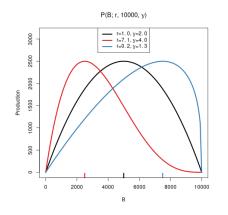
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## Pella-Tomlinson Production Model

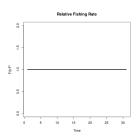
$$I(t) \sim LN(qB(t), \sigma^2)$$
  
 $\frac{dB(t)}{dt} = P_{\theta}(B(t)) - F(t)B(t)$ 

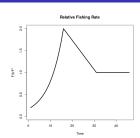
$$P_{\theta}(B) = \frac{rB}{\gamma - 1} \left( 1 - \frac{B}{K} \right)^{\gamma - 1}$$
$$\theta = (r, K, \gamma)$$

 $\gamma = 2 \Rightarrow$  Schaefer Model



#### Catch



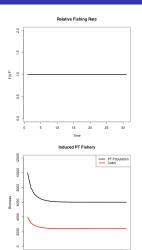


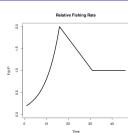
$$C(t) = F^* \left( \frac{F(t)}{F^*} \right) B(t)$$

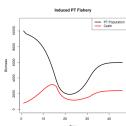


ntroduction Simulation Bias Proposals End
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## Catch





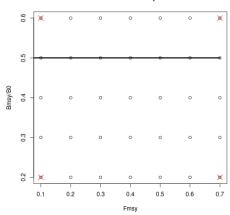




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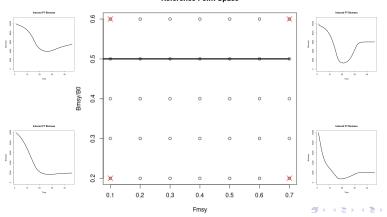
$$\boldsymbol{\theta} = \left[ r = F^* \left( \frac{1 - \frac{B^*}{\bar{B}(0)}}{\frac{B^*}{\bar{B}(0)}} \right) \left( 1 - \frac{B^*}{\bar{B}(0)} \right)^{\left( \frac{B^*}{\bar{B}(0)} - 1 \right)}, \ K = 10000, \ \gamma = \frac{1}{\frac{B^*}{\bar{B}(0)}} \right]$$

#### Reference Point Space



$$\boldsymbol{\theta} = \left[ r = F^* \left( \frac{1 - \frac{B^*}{\overline{B}(0)}}{\frac{B^*}{\overline{B}(0)}} \right) \left( 1 - \frac{B^*}{\overline{B}(0)} \right)^{\left( \frac{\frac{B^*}{\overline{B}(0)} - 1}{\frac{B^*}{\overline{B}(0)}} \right)}, \ K = 10000, \ \gamma = \frac{1}{\frac{B^*}{\overline{B}(0)}} \right]$$

#### Reference Point Space



## Metamodel

$$\mathbf{x} = \left(F^*, \frac{B^*}{\bar{B}(0)}\right)$$

$$\hat{\mu} = \beta_0 + \boldsymbol{\beta'} \boldsymbol{x} + f(\boldsymbol{x}) + \epsilon$$
 $f(\boldsymbol{x}) \sim \mathsf{GP}(0, \tau^2 R(\boldsymbol{x}, \boldsymbol{x'}))$ 
 $\epsilon_i \sim \mathsf{N}(0, \hat{\omega}_i).$ 

$$R(\boldsymbol{x}, \boldsymbol{x'}) = \exp\left(\sum_{j=1}^{2} \frac{-(x_j - x_j')^2}{2\ell_j^2}\right)$$

define  $\mu$ 



$$\check{\mu}(\check{s}) = \beta_0 + \boldsymbol{x}(\check{s})\boldsymbol{\beta} + R_{\ell}(\check{s},s)R_{\ell}^{-1}(s,s)\Big(\hat{\mu}(s) - \big(\beta_0 + \boldsymbol{x}(s)\boldsymbol{\beta}\big)\Big)$$

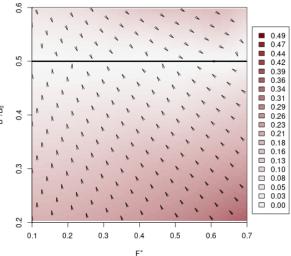
$$\check{B}^* = \frac{\check{K}}{2} \qquad \check{F}^* = \frac{\check{r}}{2}.$$

Relative Bias = 
$$\frac{\mathring{RP} - RP}{RP}$$





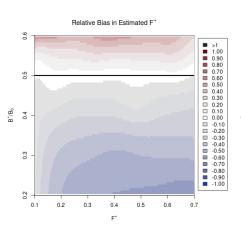
Bias •00000

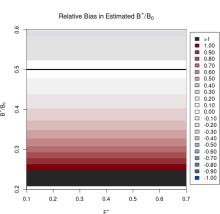




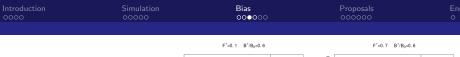
Bias

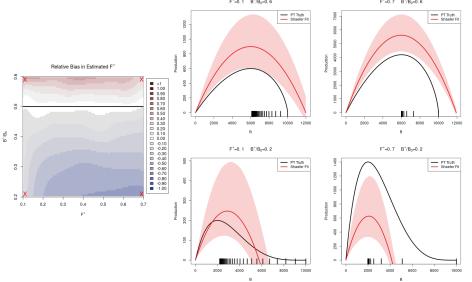
# Components of Bias



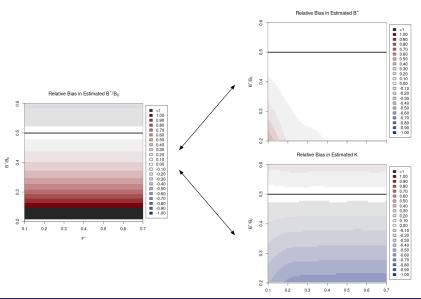




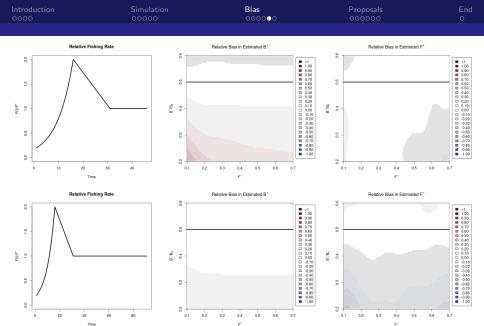




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# Summary

 $F^*$ ,  $B^*$ ,  $B_0$  are not directly observable, but rather modeled quantities

Model Misspecification, Posterior Uncertainty, Bias

RP bias can be very large when production function assumptions are wrong

As model misspecification increases, RP biases tend to increase B\* often appears relatively less sensative to model misspecification than either  $F^*$  or  $B_0$ 

F\* bias is strongly catch dependent

Bias depends on how similar the modeled and true production functions can be at the observed biomasses

A rich simulation-based method for describing global RP bias and a stepping stone for understanding other models

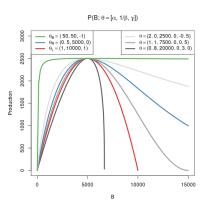
RH and Ricker SRRs Age-Structured and Delay Difference Models

# Productivity Extension

$$\frac{dB}{dt} = P(B; \theta) - (M + F)B$$

$$P(B; [\alpha, \beta, \gamma]) = \alpha B(1 - \beta \gamma B)^{\frac{1}{\gamma}}$$

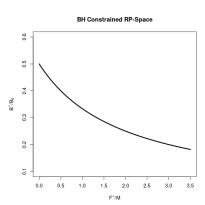
$$\gamma = -1 \Rightarrow$$
 Beverton-Holt  $\gamma = 1 \Rightarrow$  Logistic  $\gamma \to 0 \Rightarrow$  Ricker



# Productivity Extension

$$P_{\mathsf{BH}}(B; [\alpha, \beta, -1]) = \frac{\alpha B}{(1 + \beta B)}$$

$$\frac{B^*}{\bar{B}(0)} = \frac{1}{\frac{F^*}{M} + 2}$$



## Growth Extension

$$\frac{dB}{dt} = \underbrace{w(a_0)R(B;\theta)}_{\text{Recruitment Biomass}} + \underbrace{\kappa \left[w_{\infty}N - B\right]}_{\text{Net Growth}} - \underbrace{(M+F)B}_{\text{Mortality}}$$

$$\frac{dN}{dt} = R(B;\theta) - (M+F)N$$

$$R(B; [\alpha, \beta, \gamma]) = \alpha B(t - a_0) (1 - \beta \gamma B(t - a_0))^{\frac{1}{\gamma}}$$
  
$$w(a) = w_{\infty} (1 - e^{-\kappa a})$$

bullets of primary points of individual growth and maturity

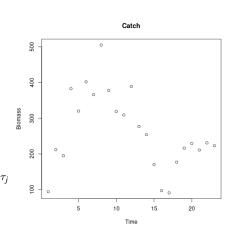


# Catch Interpolation

$$t \in \mathbb{R}^+$$
  $au = \lceil t 
ceil - 1$ 

$$\mathbb{E}[y(t)] = \int_{\tau}^{t} x(t^*) dt^*$$

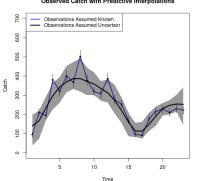
$$x(t) = \beta_0 + \sum_{i=1}^{T-1} \beta_j (t - \tau_j) \mathbb{1}_{t > \tau_j}$$



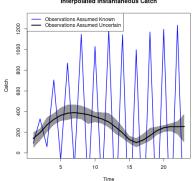


$$y(\tau_i) = \beta_0 + \sum_{j=1}^{i-1} \beta_j \left[ \left( \frac{\tau_i^2}{2} - \tau_j \tau_i \right) \mathbb{1}_{\tau_i > \tau_j} - \left( \frac{\tau_{i-1}^2}{2} - \tau_j \tau_{i-1} \right) \mathbb{1}_{\tau_{i-1} > \tau_j} \right] + \epsilon_i$$
$$\beta_j \sim N(0, \phi) \qquad \phi \sim \mathsf{Half-Cauchy}(0, 1) \qquad \epsilon_i \sim N(0, \sigma_i^2)$$

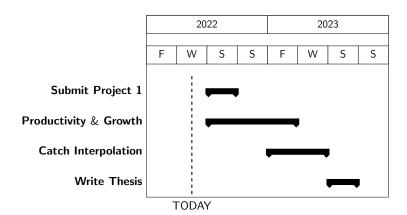
#### Observed Catch with Predictive Interpolations



#### Interpolated Instantaneous Catch







Thanks and Acknowldgements NOAA, Sea Grant Ecetra

$$\frac{B^*}{\bar{B}(0)} = \frac{\left(\frac{\alpha}{M+F^*}\right)^{\frac{1}{\gamma}} - 1}{\left(\frac{\alpha}{M}\right)^{\frac{1}{\gamma}} - 1}$$

$$\alpha = (M+F^*) \left[1 - \frac{1}{\gamma} \left(\frac{F^*}{M+F^*}\right)\right]^{-\gamma}$$

$$\beta = \frac{1}{\gamma \bar{B}(0)} \left(1 - \left(\frac{\alpha}{M}\right)^{\frac{1}{\gamma}}\right)$$