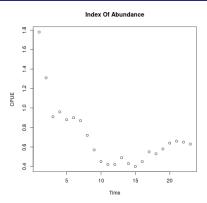
Metamodeling for Bias Estimation of Biological Reference Points Under Two-Parameter SRRs

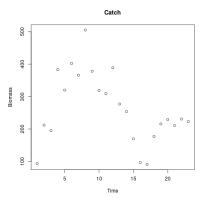


Nick Grunloh

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

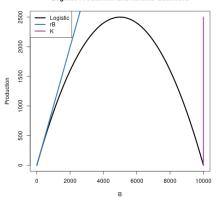
Schaefer Model

Introduction 0000

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

 $\theta = (r, K)$

Logistic Production and Related Quantities





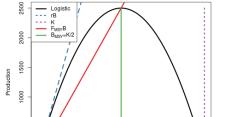
Schaefer Reference Points

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

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

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

Some words on RPs, aybe some citations



4000

6000

В

2000

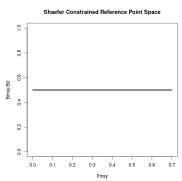
Logistic Production and Related Quantities



8000

10000

Introduction



Introduction 0000

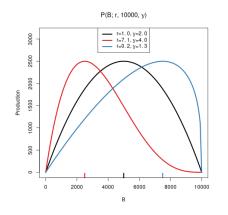
Pella-Tomlinson Production Model

$$I(t) \sim LN(qB(t), \sigma^2)$$

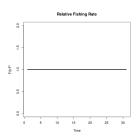
 $\frac{dB(t)}{dt} = P_{\theta}(B(t)) - F(t)B(t)$

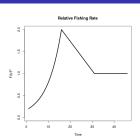
$$P_{\theta}(B) = \frac{rB}{\gamma - 1} \left(1 - \frac{B}{K} \right)^{\gamma - 1}$$
$$\theta = (r, K, \gamma)$$

 $\gamma = 2 \Rightarrow$ Schaefer Model



Catch



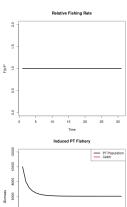


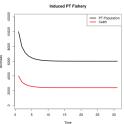
$$C(t) = F^*\left(\frac{F(t)}{F^*}\right)B(t)$$

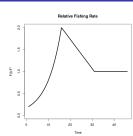


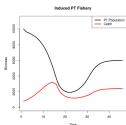
Simulation 00000

Catch





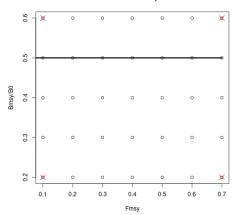






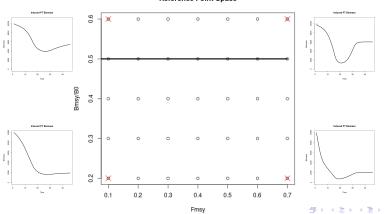
$$\theta = \left[r = F^* \left(\frac{1 - \frac{B^*}{\bar{B}(0)}}{\frac{B^*}{\bar{B}(0)}} \right) \left(1 - \frac{B^*}{\bar{B}(0)} \right)^{\left(\frac{B^*}{\bar{B}(0)} - 1 \right)}, \ K = 10000, \ \gamma = \frac{1}{\frac{B^*}{\bar{B}(0)}} \right]$$

Reference Point Space



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

Reference Point Space



Metamodel

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

$$\hat{\mu} = \beta_0 + \beta' \mathbf{x} + f(\mathbf{x}) + \epsilon$$

$$f(\mathbf{x}) \sim \mathsf{GP}(0, \tau^2 R(\mathbf{x}, \mathbf{x}'))$$

$$\epsilon_i \sim \mathsf{N}(0, \hat{\omega}_i).$$

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

define μ



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

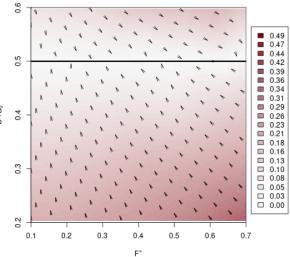
$$\check{B}^* = \frac{\check{K}}{2} \qquad \check{F}^* = \frac{\check{r}}{2}.$$

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

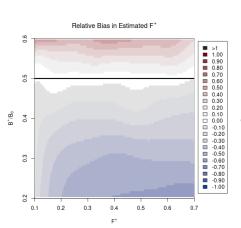


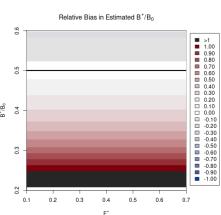


Bias •00000



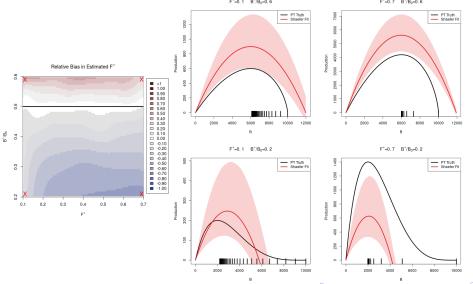
Components of Bias



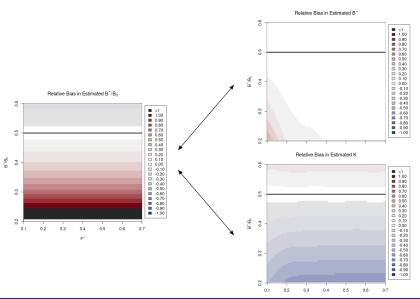




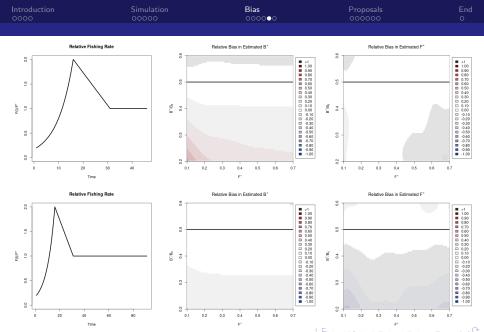




ntroduction Simulation Bias Proposals End
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Summary

 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

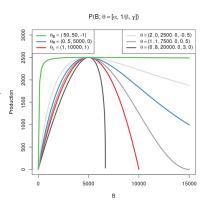
BH and Ricker SRRs
Age-Structured and Delay Difference Models

Productivity Extension

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

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

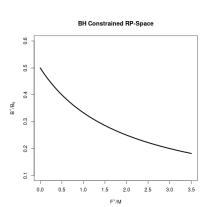
$$\gamma = -1 \Rightarrow$$
 Beverton-Holt $\gamma = 1 \Rightarrow$ Logistic $\gamma \rightarrow 0 \Rightarrow$ Ricker



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

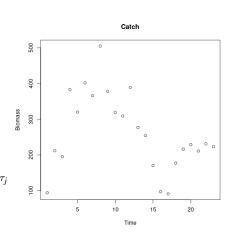


Catch Interpolation

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

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

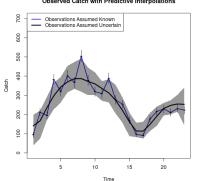
$$x(t) = \beta_0 + \sum_{i=1}^{T-1} \beta_j (t - \tau_j) \mathbb{1}_{t > \tau_j}$$

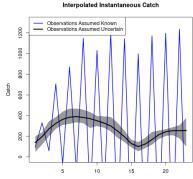




$$y(\tau_i) = \beta_0 + \sum_{j=1}^{i-1} \beta_j \left[\left(\frac{\tau_i^2}{2} - \tau_j \tau_i \right) \mathbb{1}_{\tau_i > \tau_j} - \left(\frac{\tau_{i-1}^2}{2} - \tau_j \tau_{i-1} \right) \mathbb{1}_{\tau_{i-1} > \tau_j} \right] + \epsilon_i$$
$$\beta_j \sim N(0, \phi) \qquad \phi \sim \mathsf{Half-Cauchy}(0,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

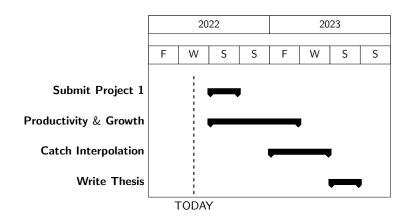
Observed Catch with Predictive Interpolations





Time





Thanks and Acknowldgements NOAA, Sea Grant Ecetra

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

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

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

Bias