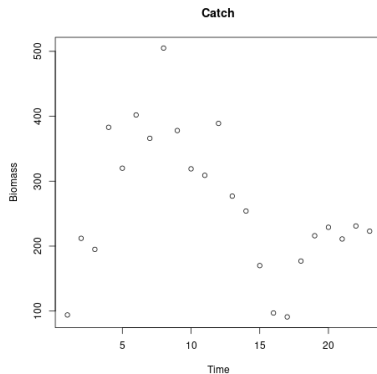
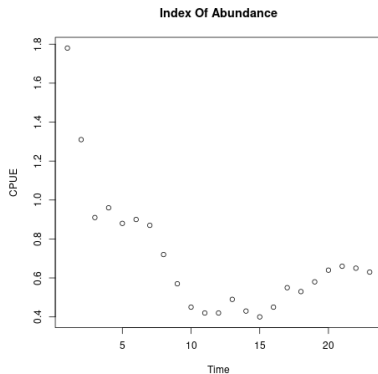


Bias Estimation of Biological Reference Points Under Two-Parameter SRRs

Nick Grunloh

14 March 2022





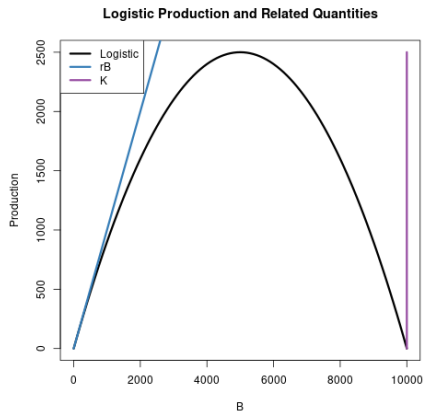
$$I_t = qB_t e^{\epsilon} \quad \epsilon \sim N(0, \sigma^2)$$

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

Schaefer Model

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

$$\theta = (r, K)$$

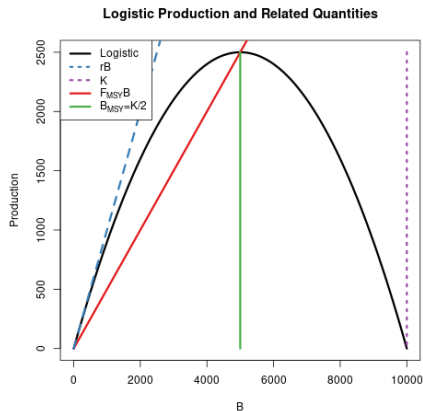


Schaefer Reference Points

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

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

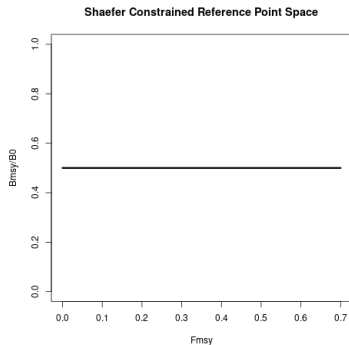
$$MSY = \frac{rK}{4}$$



Something about Model misspecification of RPs

Three parameter SRR

functional form with oprthogonal RPs



Pella-Tomlinson Production Model

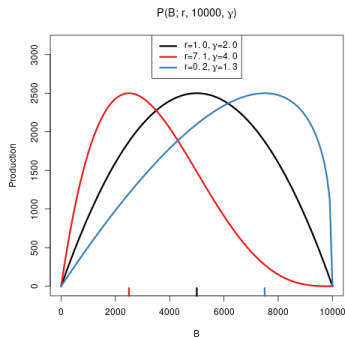
$$I(t) \sim LN(qB(t), \sigma^2)$$

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

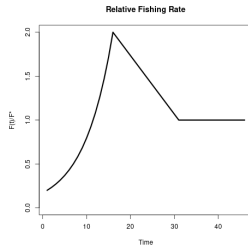
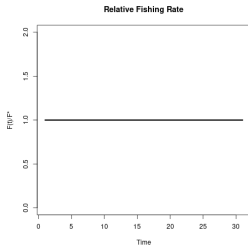
$$P_{\theta}(B) = \frac{rB}{\gamma - 1} \left(1 - \frac{B}{K}\right)^{\gamma-1}$$

$$\theta = (r, K, \gamma)$$

$$\gamma = 2 \Rightarrow \text{Schaefer Model}$$

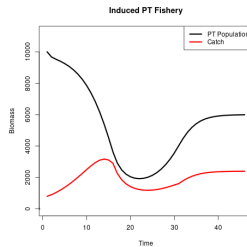
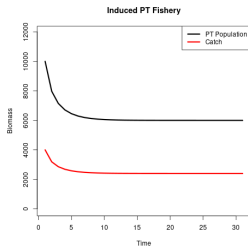
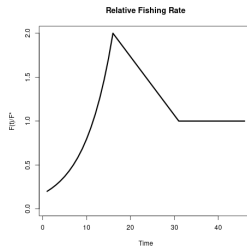
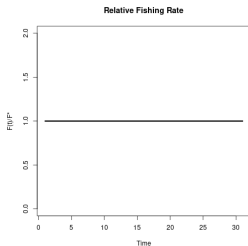


Catch

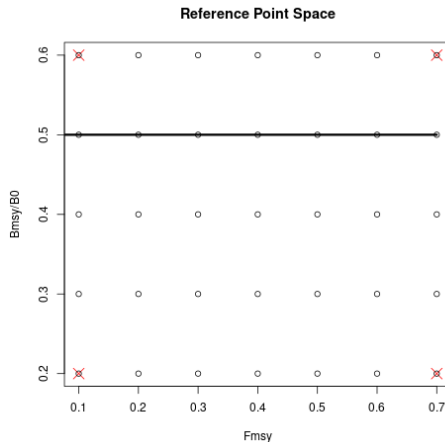


$$C(t) = F^* \left(\frac{F(t)}{F^*} \right) B(t)$$

Catch

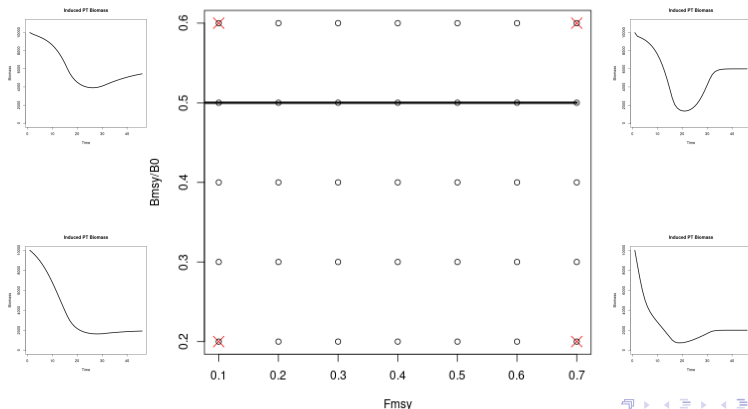


$$\theta = \left[r = F^* \left(\frac{1 - \frac{B^*}{\bar{B}(0)}}{\frac{B^*}{\bar{B}(0)}} \right) \left(1 - \frac{B^*}{\bar{B}(0)} \right)^{\left(\frac{\frac{B^*}{\bar{B}(0)} - 1}{\frac{B^*}{\bar{B}(0)}} \right)}, K = 10000, \gamma = \frac{1}{\frac{B^*}{\bar{B}(0)}} \right]$$



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

Reference Point Space



Metamodel

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

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

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

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

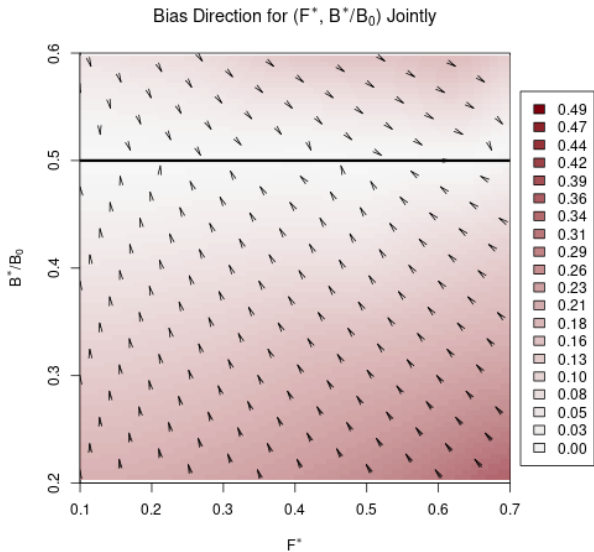
$$R(\mathbf{x}, \mathbf{x}') = \exp \left(\sum_{j=1}^2 \frac{-(x_j - x'_j)^2}{2\ell_j^2} \right)$$

define μ

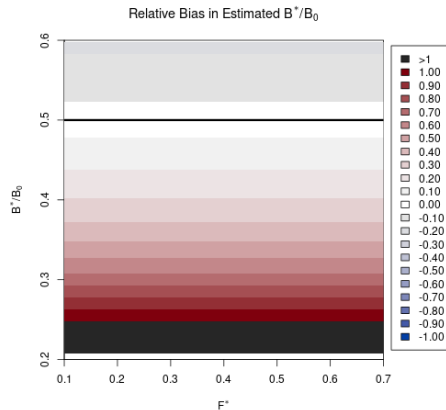
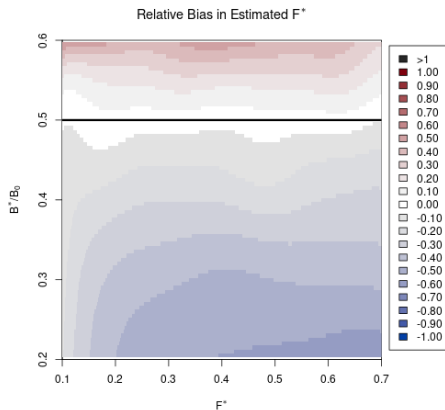
$$\check{\mu}(\check{s}) = \beta_0 + \mathbf{x}(\check{s})\boldsymbol{\beta} + R_\ell(\check{s}, s)R_\ell^{-1}(s, s)\left(\hat{\mu}(s) - (\beta_0 + \mathbf{x}(s)\boldsymbol{\beta})\right)$$

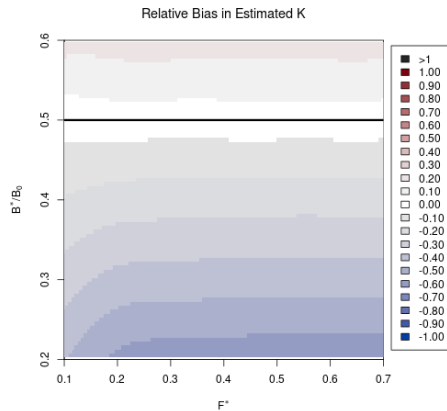
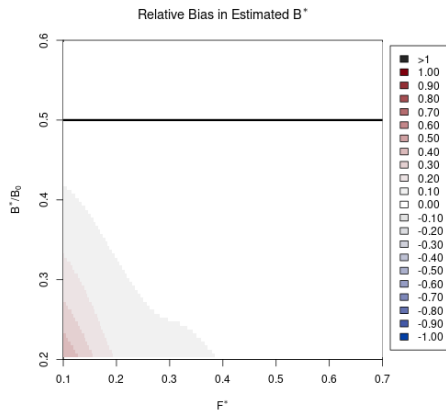
$$\check{B}^* = \frac{\check{K}}{2} \quad \check{F}^* = \frac{\check{r}}{2}.$$

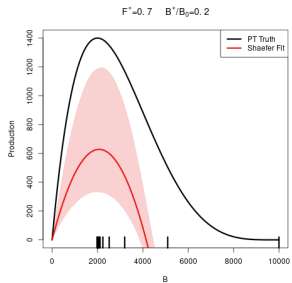
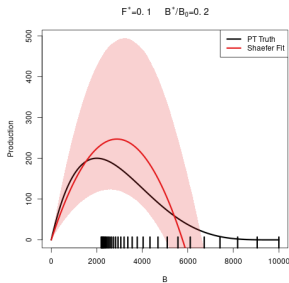
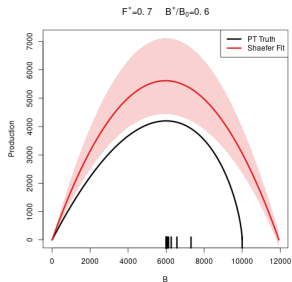
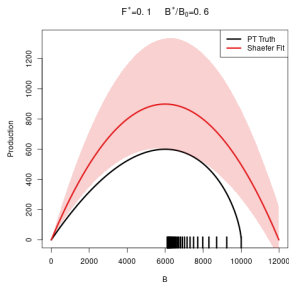
$$\text{Relative Bias} = \frac{\check{R}P - RP}{RP}$$



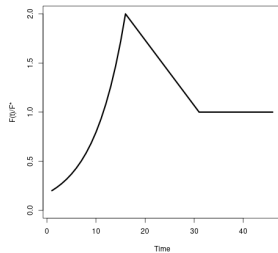
Components of Bias



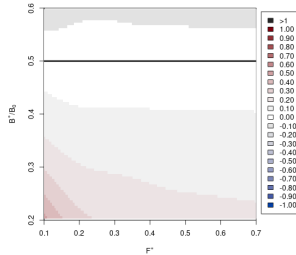
Split $\frac{B^*}{B_0}$ 



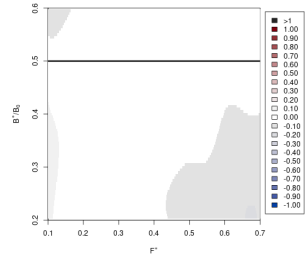
Relative Fishing Rate



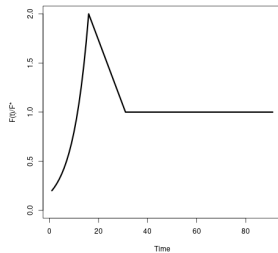
Relative Bias in Estimated B*



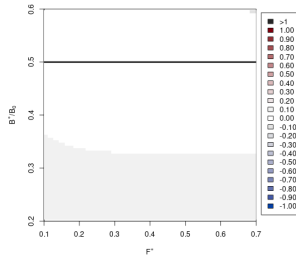
Relative Bias in Estimated F*



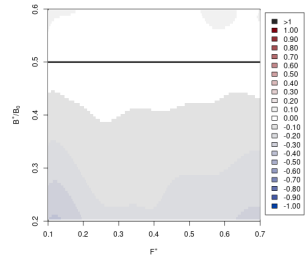
Relative Fishing Rate



Relative Bias in Estimated B*



Relative Bias in Estimated F*



Conclusions

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 sensitive 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

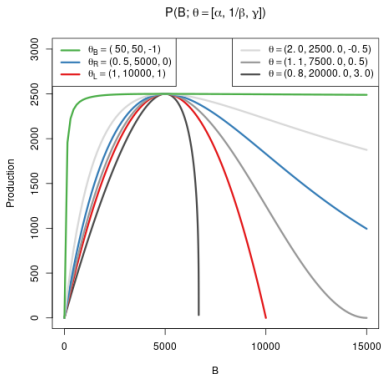
$$\frac{dB}{dt} = P(B; \theta) - (M + F)B$$

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

$\gamma = -1 \Rightarrow$ Beverton-Holt

$\gamma = 1 \Rightarrow$ Logistic

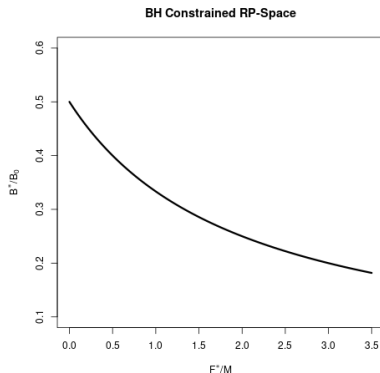
$\gamma \rightarrow 0 \Rightarrow$ Ricker



Productivity Extension

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

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



Growth Extension

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

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

$$R(B; [\alpha, \beta, \gamma]) = \alpha B(t - a_0)(1 - \beta \gamma B(t - a_0))^{\frac{1}{\gamma}}$$

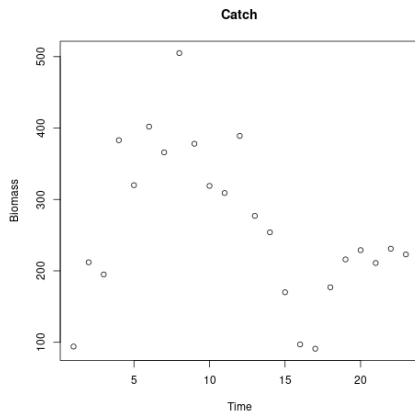
$$w(a) = w_\infty(1 - e^{-\kappa a})$$

Catch Interpolation

$$t \in \mathbb{R}^+ \quad \tau = \lceil t \rceil - 1$$

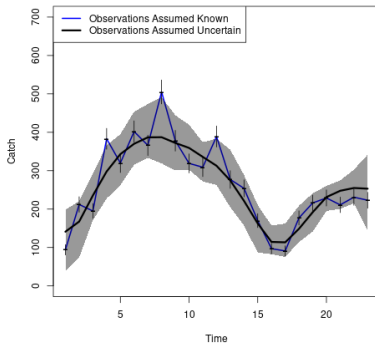
$$\mathbb{E}[y(t)] = \int_{\tau}^t x(t^*) dt^*$$

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

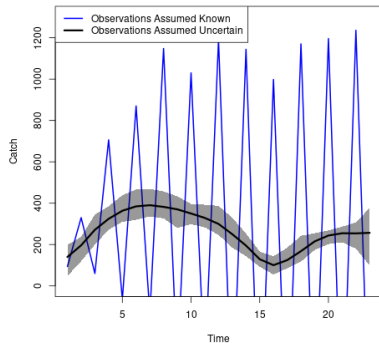


$$y(\tau_i) = \beta_0 + \sum_{j=1}^{i-1} \beta_j \left[\left(\frac{\tau_i^2}{2} - \tau_j \tau_i \right) \mathbb{1}_{\tau_i > \tau_j} - \left(\frac{\tau_{i-1}^2}{2} - \tau_j \tau_{i-1} \right) \mathbb{1}_{\tau_{i-1} > \tau_j} \right] + \epsilon_i$$
$$\beta_j \sim N(0, \phi) \quad \phi \sim \text{Half-Cauchy}(0,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

Observed Catch with Predictive Interpolations



Interpolated Instantaneous Catch



Thanks and Acknowledgements NOAA, Sea Grant Ecetra

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