

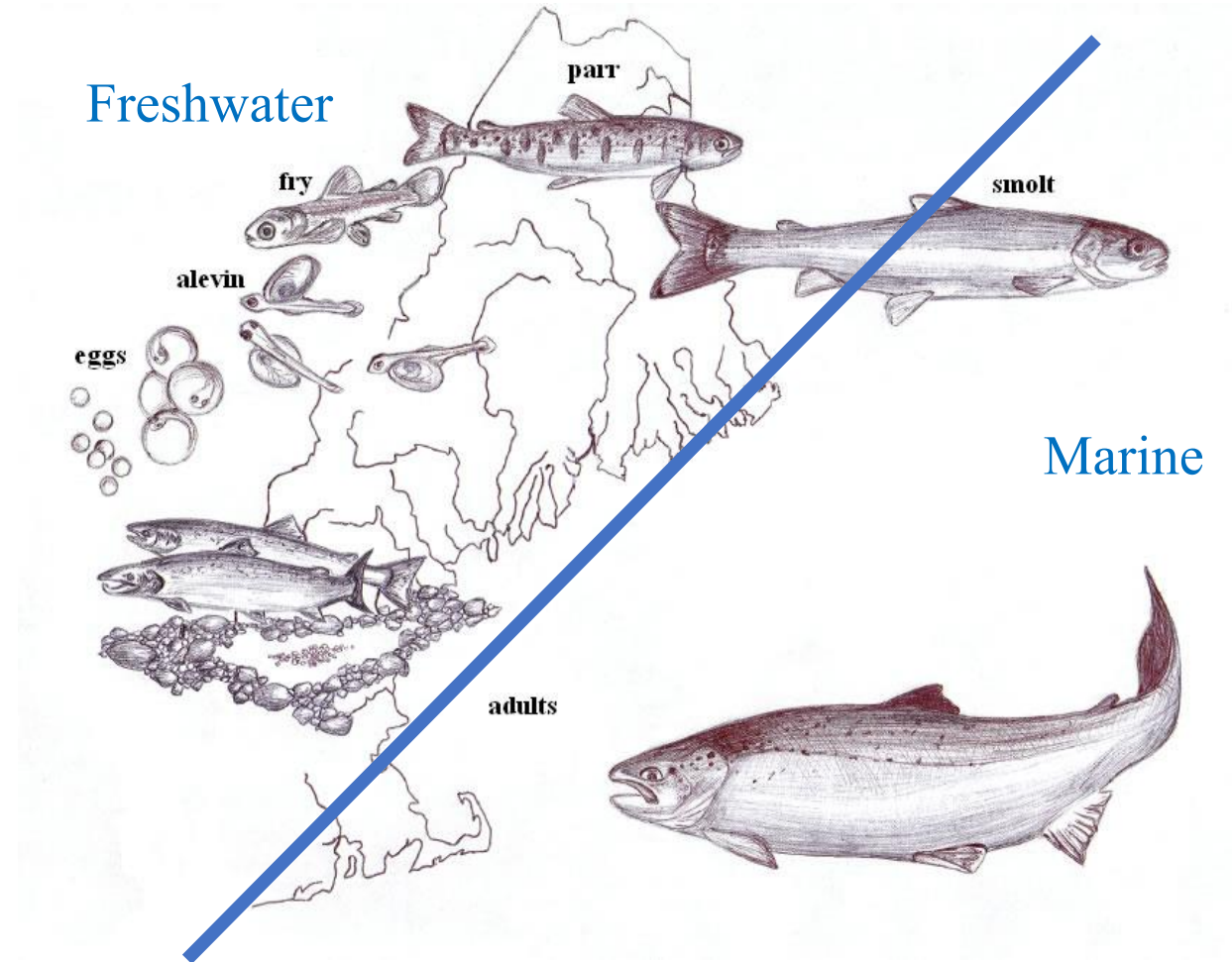
Length of Atlantic salmon smolt and their subsequent marine survival

Stephen Gregory*, Rasmus Lauridsen, William Beaumont & William Riley



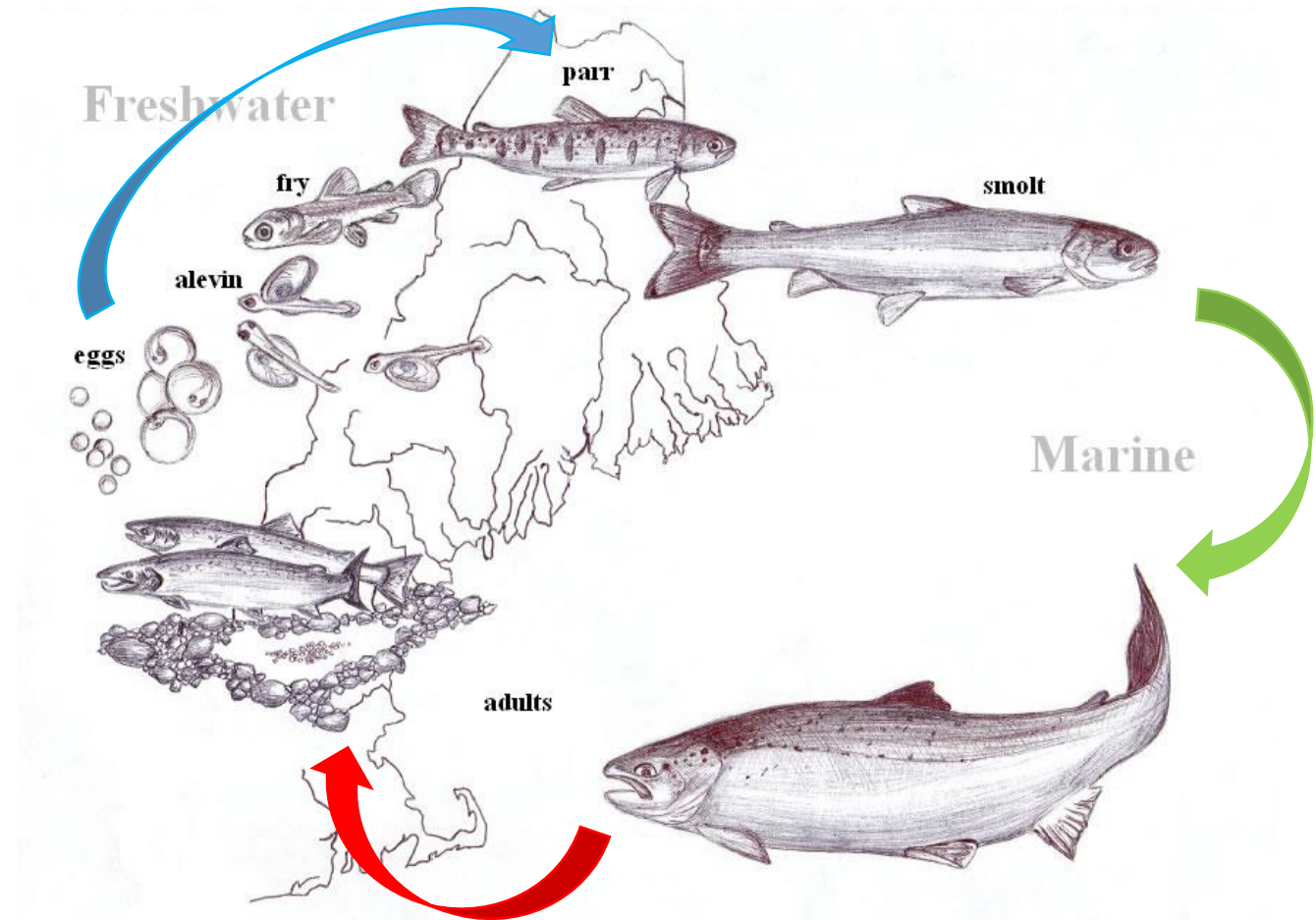
Atlantic salmon life cycle

- Anadromous
 - Freshwater
 - Marine



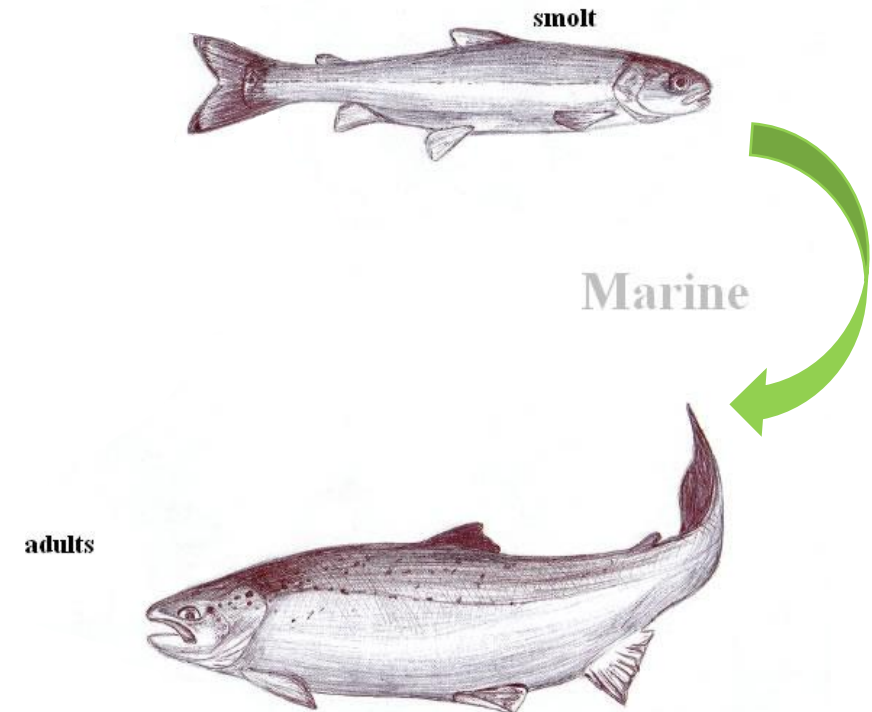
Atlantic salmon life cycle

- Anadromous
 - Freshwater
 - Marine
- Three major transitions
 - Egg > Smolt (freshwater survival)
 - Smolt > Adult (marine survival)
 - Adult > Spawner (fishing mortality)



Atlantic salmon life cycle

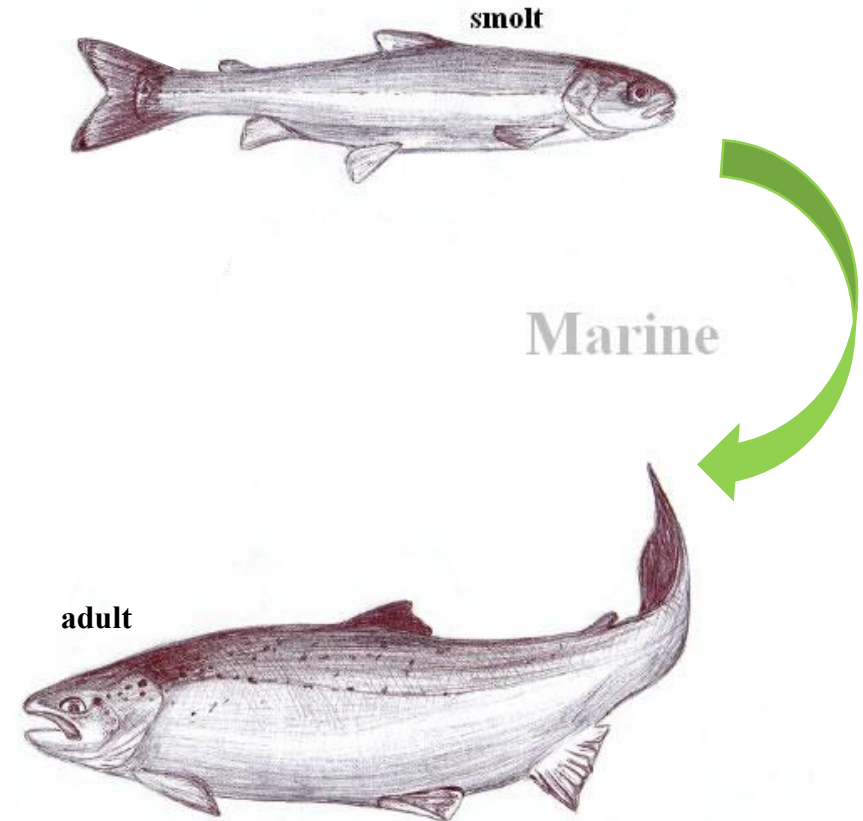
– Smolt > Adult (marine survival)



Atlantic salmon marine survival

- Risks

- Physiological stress of F/W -> S/W
 - Temperature, salinity, etc.
- Novel, abundant predators
- Distant-water fisheries
 - Highly regulated since 1990



Previous research & hypotheses

- Marine survival might be related to:
 - Smolt length
 - Larger smolt better survive
 - Origin
 - Wild (vs hatchery) smolts better survive
 - Environmental conditions
 - Better growth conveys better survival



Potter, E. C. E., Maoileidigh, N. O. & Chaput, G. (Eds.)
2003. Marine mortality of Atlantic salmon, *Salmo salar* L:
methods and measures. DFO Canadian Science Advisory
Secretariat - Research Document 2003/101

Question & prediction



- Question:

Does marine survival increase with increasing smolt length?

- Prediction:

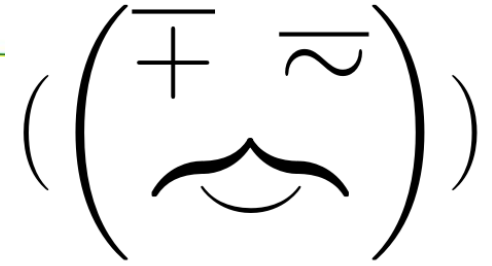
Estimated marine survival will increase with smolt length.

Method


$$\left(\begin{pmatrix} \bar{+} & \bar{2} \\ \underbrace{\hspace{1cm}} \end{pmatrix} \right)$$

- 1) Estimate marine survival from Bayesian State-Space model
- 2) Adapt BSSM to estimate effect of individual smolt length
- 3) Consider alternative hypotheses: model comparison [Todo]

Method

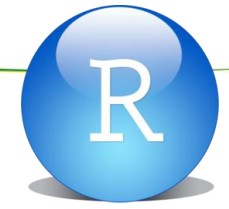


1) Estimate marine survival from Bayesian State-Space model

State matrix of individual i $z_i = [A, ?, B, ?]$
Space matrix of individual i $w_i = [1, 0, 1, 0]$

Individual i observed in state $k = A$ at time $t = 1$, was unobserved at $t = 2$, was observed in $k = B$ at $t = 3$ and was unobserved at $t = 4$.

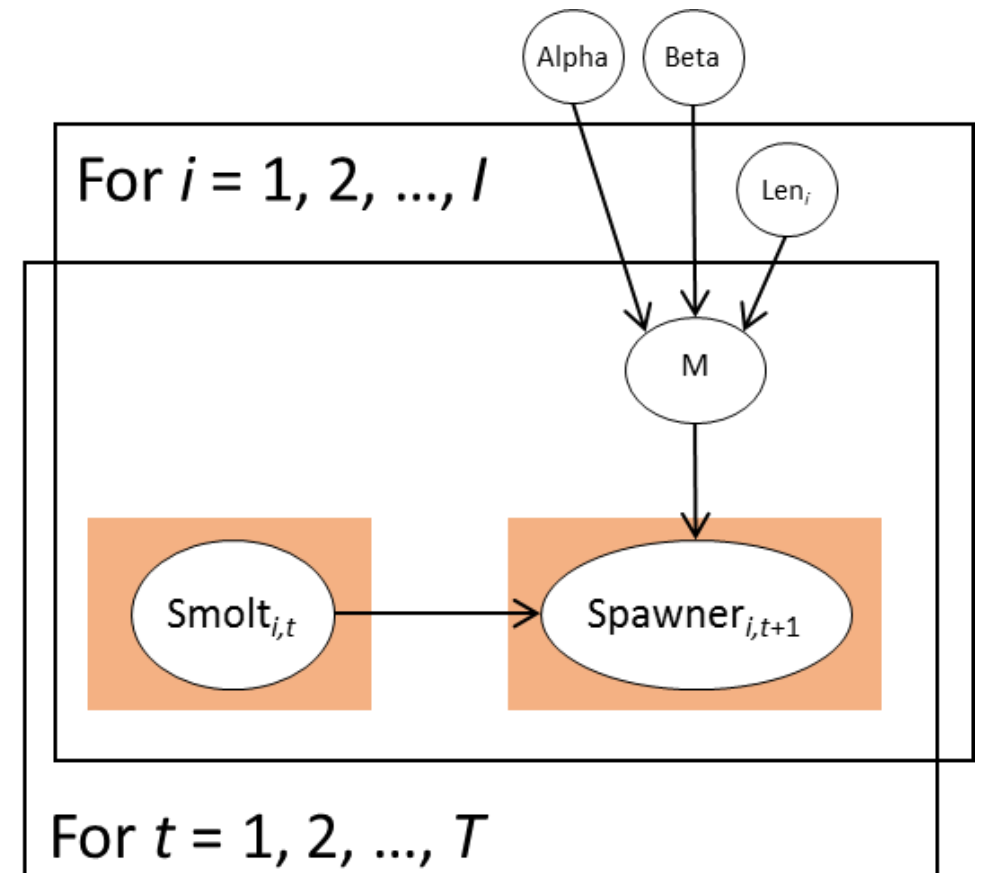
Method



JAGS

1) Estimate marine survival from Bayesian State-Space model

- Individual-based BSSM:
 - admits individual smolt lengths
- Assimilates information to stock level:
 - admits population-level covariates



1) Estimate marine survival from Bayesian State-Space model

- Estimates joint probability of transition (ϕ) and detection (p) probabilities by MCMC

$$\pi(\phi, p | z, w) \propto \pi(z, w | \phi, p) \pi(\phi) \pi(p)$$

where:

$\pi(\phi, p | z, w)$ = posterior parameter probabilities

$\pi(z, w | \phi, p)$ = product of the likelihood of the data given the parameters

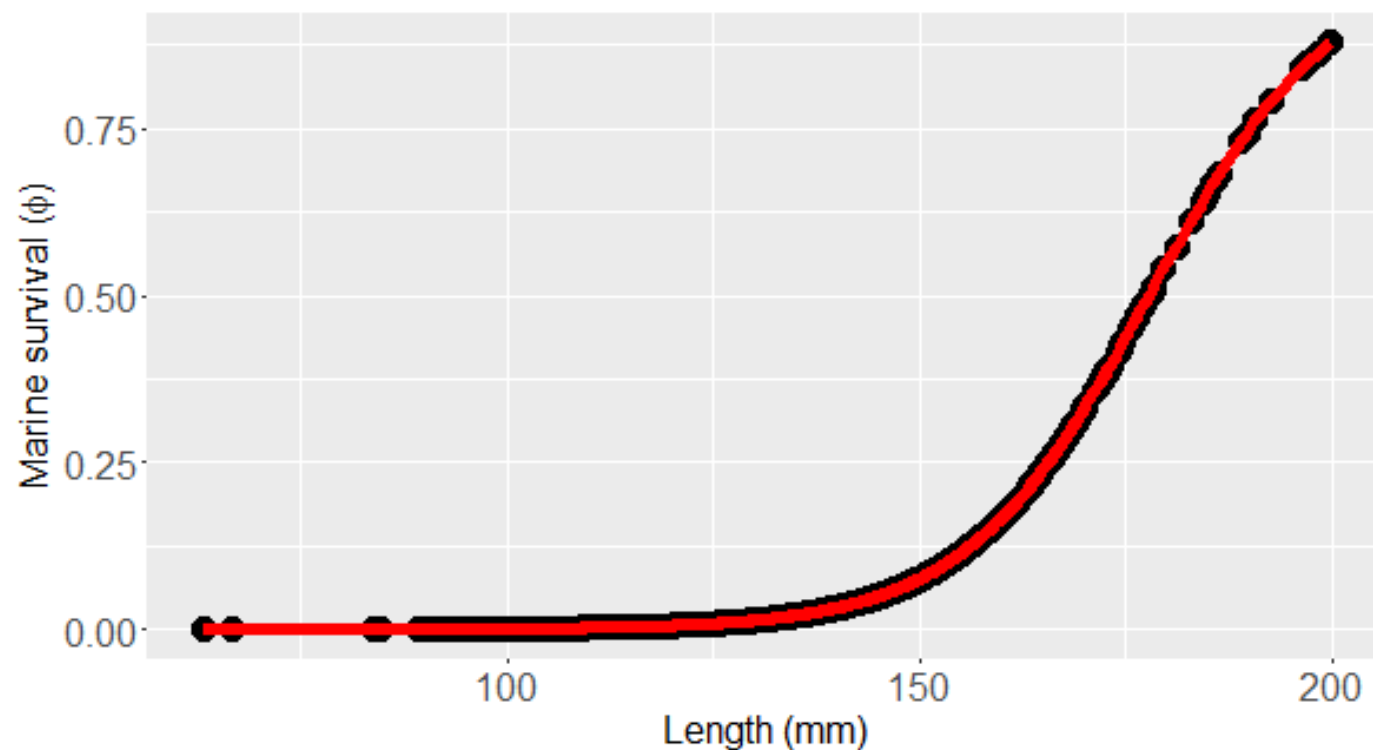
$\pi(\phi)$ = prior probabilities of transition

$\pi(p)$ = prior probabilities of detection

Method

$$\left(\begin{pmatrix} \bar{+} & \bar{2} \\ \underbrace{\hspace{1cm}} \end{pmatrix} \right)$$

2) Adapt BSSM to estimate effect of individual smolt length



$$\phi = \frac{1}{1 + \exp(-lp)}$$

$$lp \sim \alpha + \beta \text{Length}$$

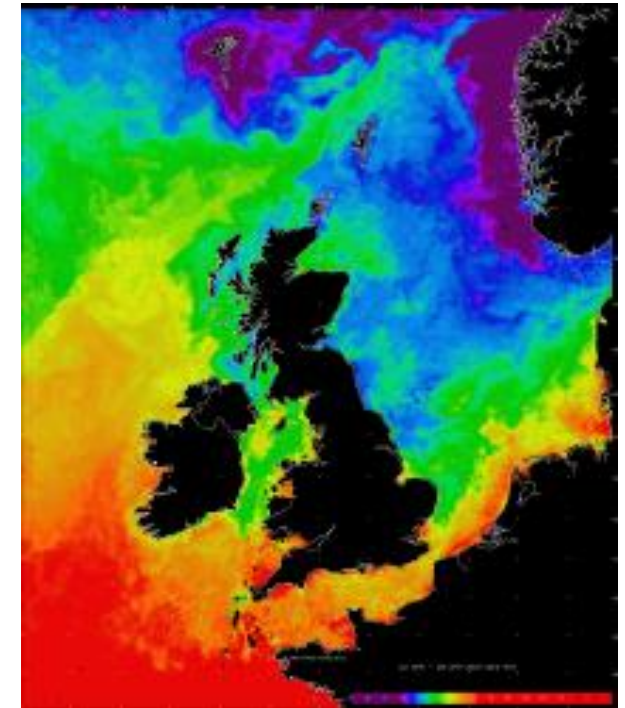
Method

$$\left(\begin{pmatrix} \bar{F} & \bar{I} \\ \underbrace{\hspace{1cm}} \end{pmatrix} \right)$$

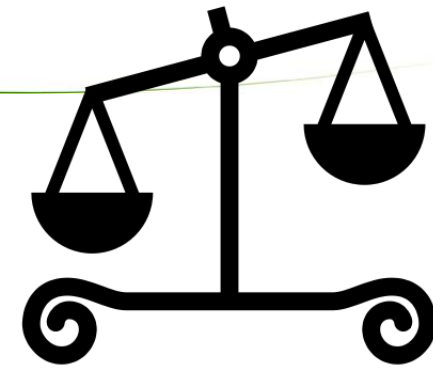
3) Consider alternative hypotheses: model comparison [Todo]

Additional variables to explain marine survival:

- Sea surface temperature ~ growth conditions
- Other...



Testing the BSSM

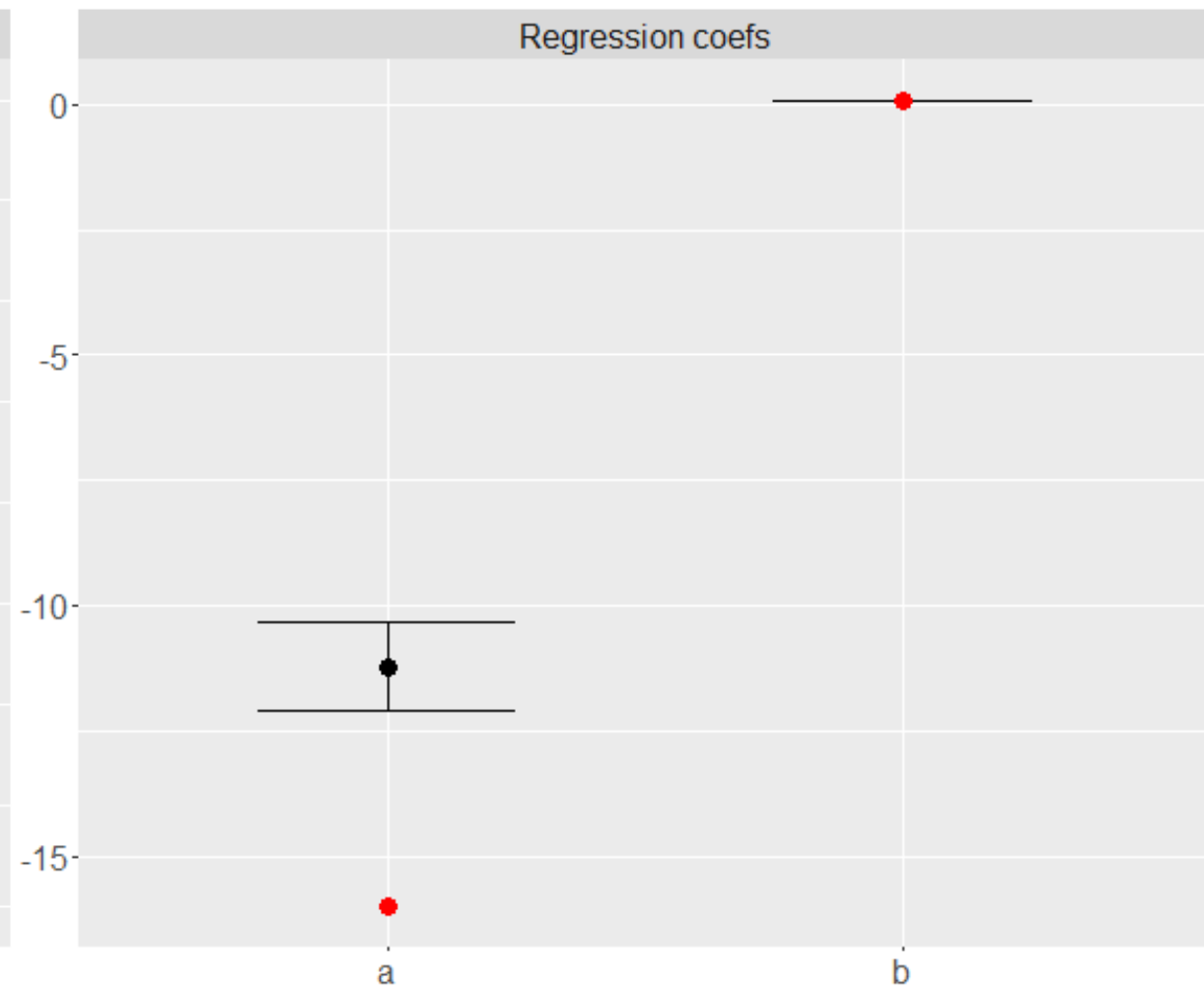
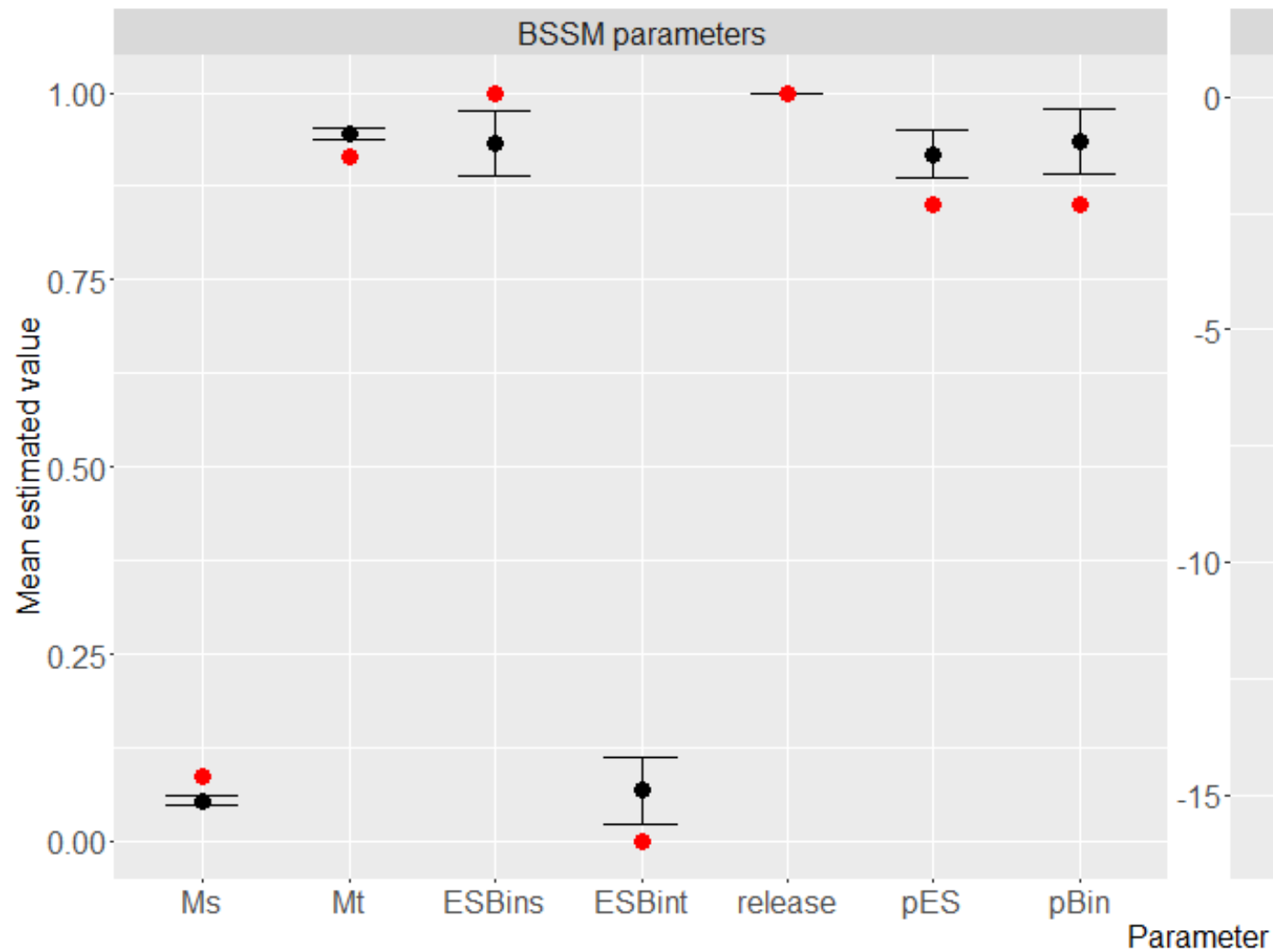


- Generate data & estimate parameters from same model
- Generating parameters:

Parameter	Abbreviation	Value
Mean individual marine survival	Ms	0.09
Individual detection at release	release	1.00
Individual detection probabilities at PIT readers	pES, pBn	0.85, 0.85
Survival between PIT readers	ESBins	1.00
Length-Survival logistic regression coefficients	a, b	-16, 0.09

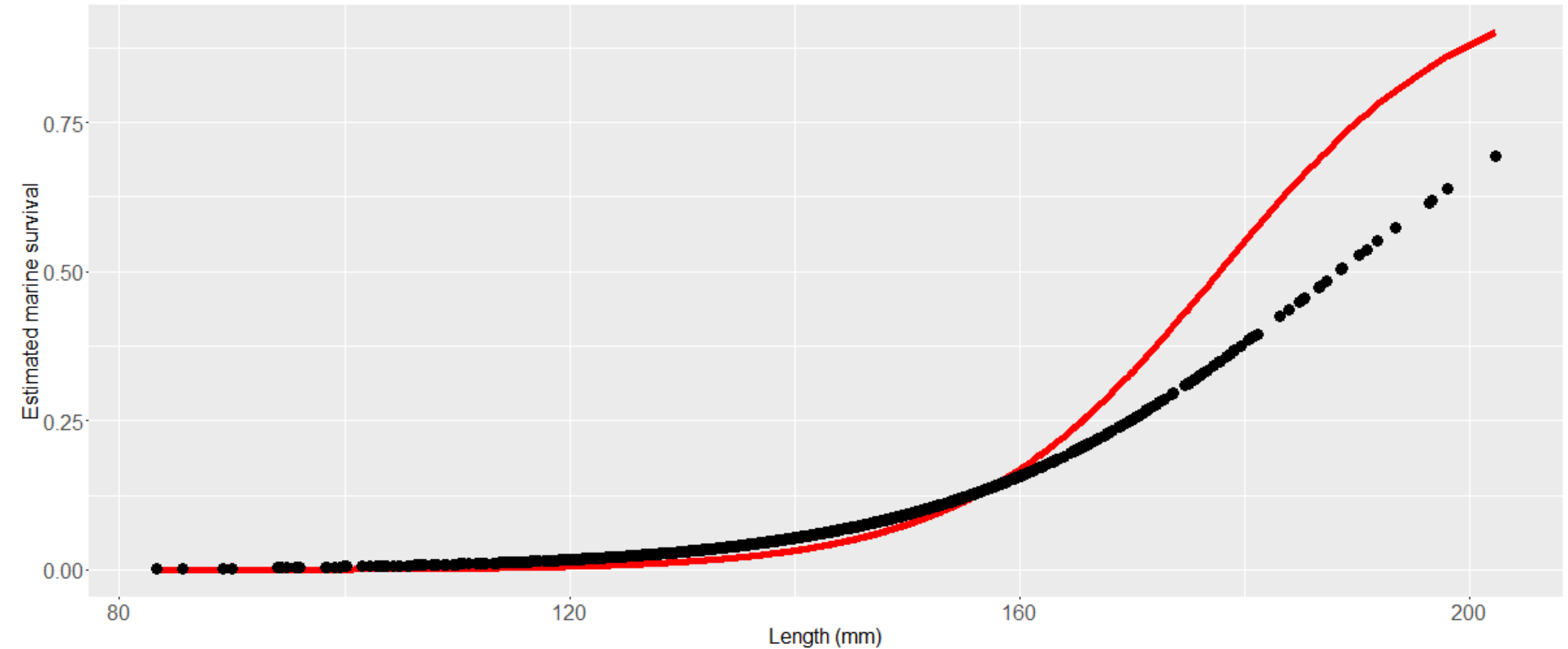
Testing the BSSM

• Actual • Estimated

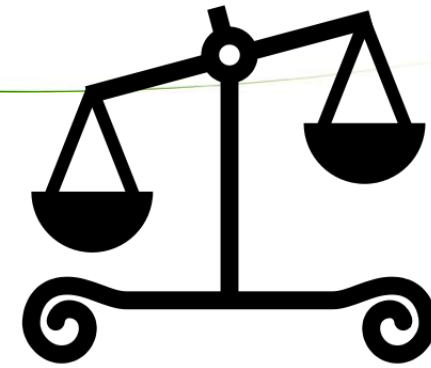


Testing the BSSM

• Actual • Estimated



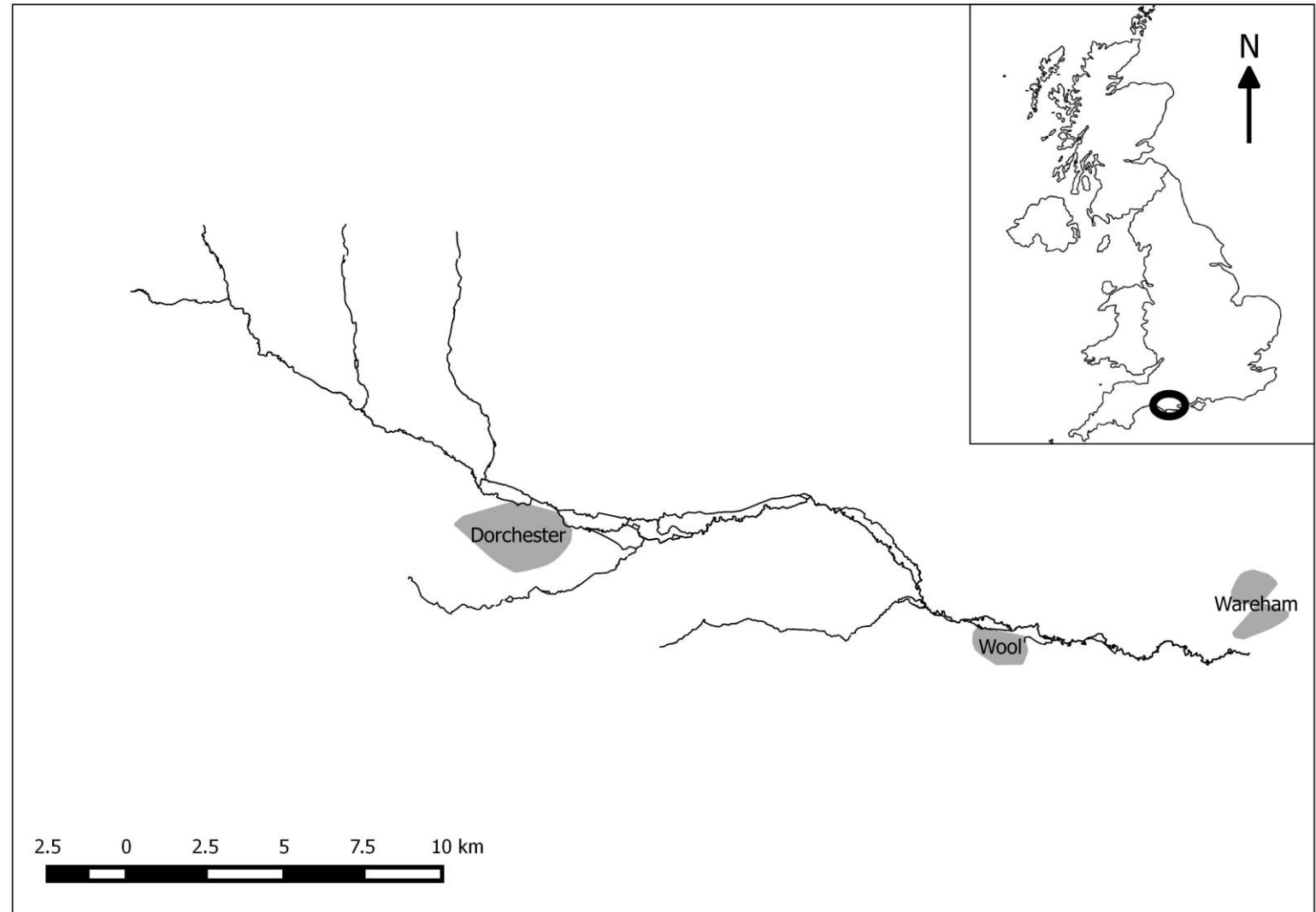
Testing the BSSM



- Further tests reveal that reduced pES and pBn introduce bias in parameter and coefficient estimates.
- Marine survival estimates largely unaffected by bias.
- Could be improved by informative priors or model re-specification.
- Further work needed.

Example: *A. salmon* in the Frome, Dorset

- Chalk stream
- Wild salmon
- 98% smolts age 1
- PIT tags



Example: A. salmon in the Frome, Dorset

PIT tagging programme

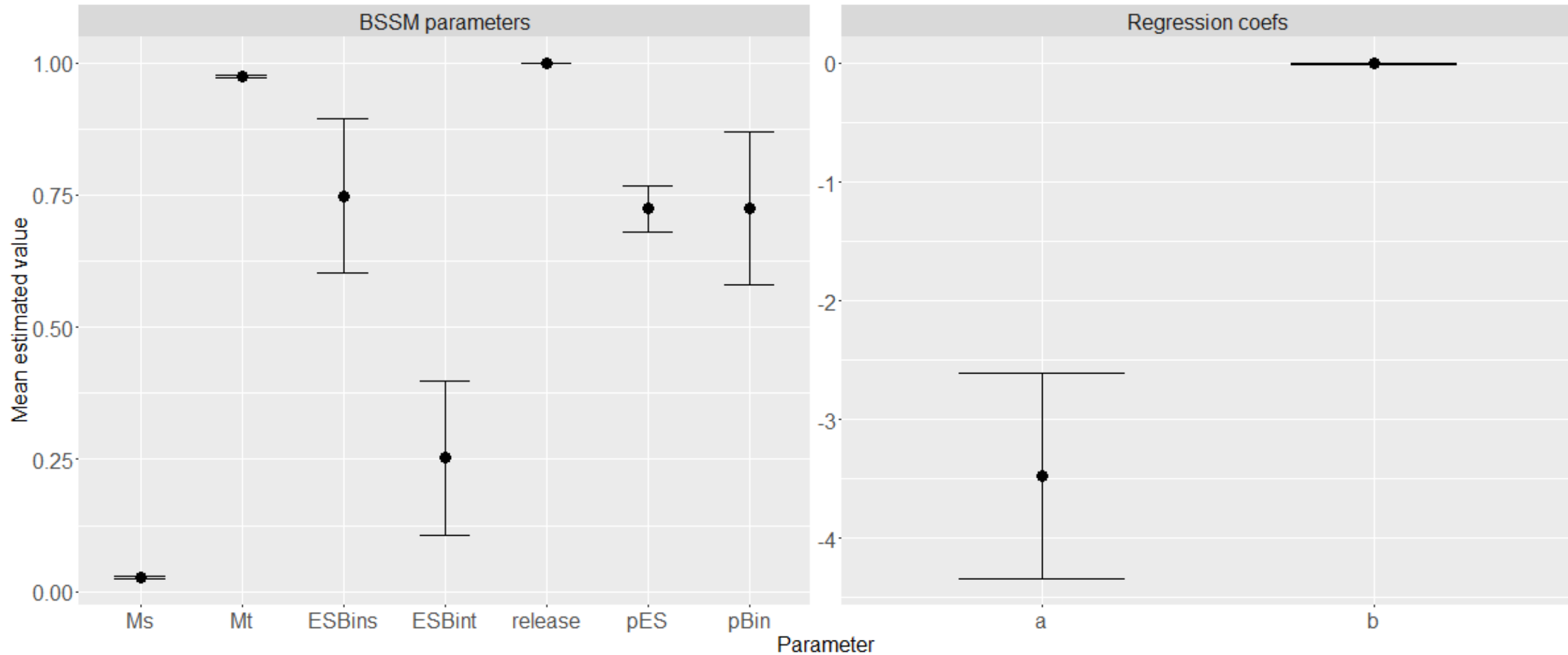
- Annual from 2003 – present
- Cover entire Frome catchment
- Aim to distribute 10,000 PIT tags
- PIT readers throughout catchment

Sample size for analysis

- 8423 PIT tagged smolts
- 191 PIT tagged adult returns



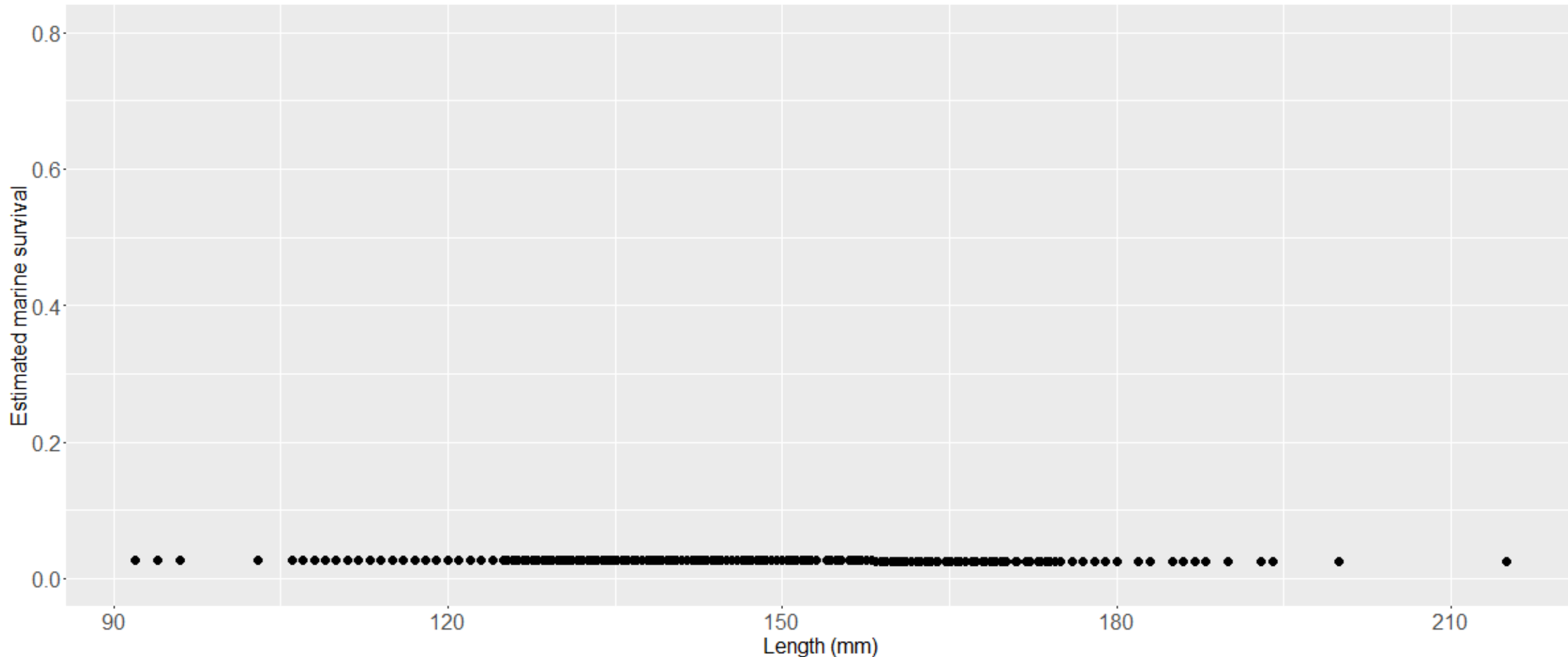
Example: A. salmon in the Frome, Dorset



Example: A. salmon in the Frome, Dorset

- Frome A. salmon marine survival estimates:
 - Chapman model = 1.8 – 2.5%
 - This BSSM = 2.6%
- PIT reader detection probability estimates:
 - Chapman model = 0.85
 - This BSSM = 0.73
- Transition between PIT readers appears low: 0.75

Example: *A. salmon* in the Frome, Dorset



Preliminary conclusions

- BSSM can estimate 1SW marine survival accounting for detection uncertainty.
- Admits individual covariates, e.g., length, environmental conditions, ...
- Individual length appears unrelated to marine survival for Frome Atlantic salmon smolts. But this requires more work.

Next steps



- Improve model to reduce bias from low detection probabilities.
- Alternative equations to include covariates, e.g., linear...
- Extend the model to estimate 1SW and MSW survival separately.
- Compare models including additional explanatory variables.

A background image showing a small boat on a river, surrounded by dense green foliage and trees. The image is slightly faded to allow text to be overlaid.

Thanks for listening

sgregory@gwct.org.uk