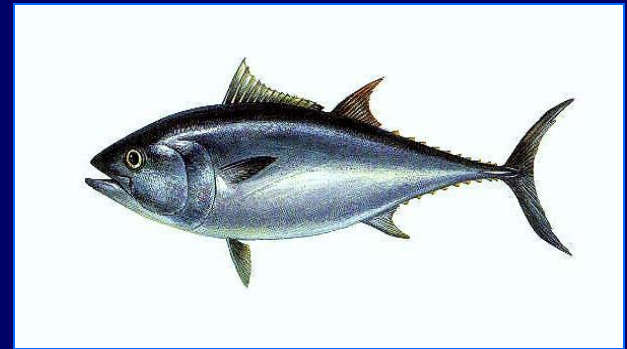
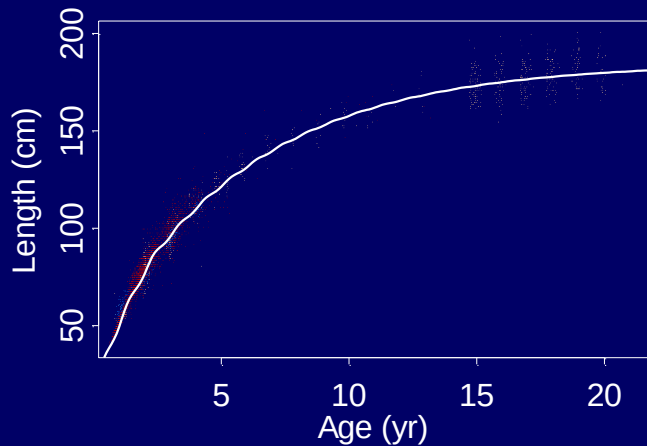


A comprehensive growth model for tuna in the Indian Ocean

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Acknowledgments

- **Model development:** Tom Polacheck, Geoff Laslett
- **Otolith data:** Fany Sardenne, Gael LeCroizier, and other readers
- **Tagging data:** Julien Million

Why do we need growth models?

- fundamental to stock assessments for estimating age composition of catch
- changes in growth have important implications about stock status (i.e. density dependence)

Sources of growth information

Three common sources:

- Change in length and time at liberty data from tag-recapture experiments
- Direct age and length data obtained from hard-part analyses (e.g. otoliths)
- Length-frequency data from commercial catches - *not used here*

A general growth curve

For a fish of length l and age a :

$$l(a) = L_{\infty} f(a - a_0; \theta)$$

L_{∞} is the asymptotic length

f is a monotonic increasing function with parameter set $\{\theta, a_0\}$

$f = 0$ when $a = a_0$ (i.e. a_0 is the theoretical age of length 0)

$$\lim_{a \rightarrow \infty} f = 1$$

Specific growth functions

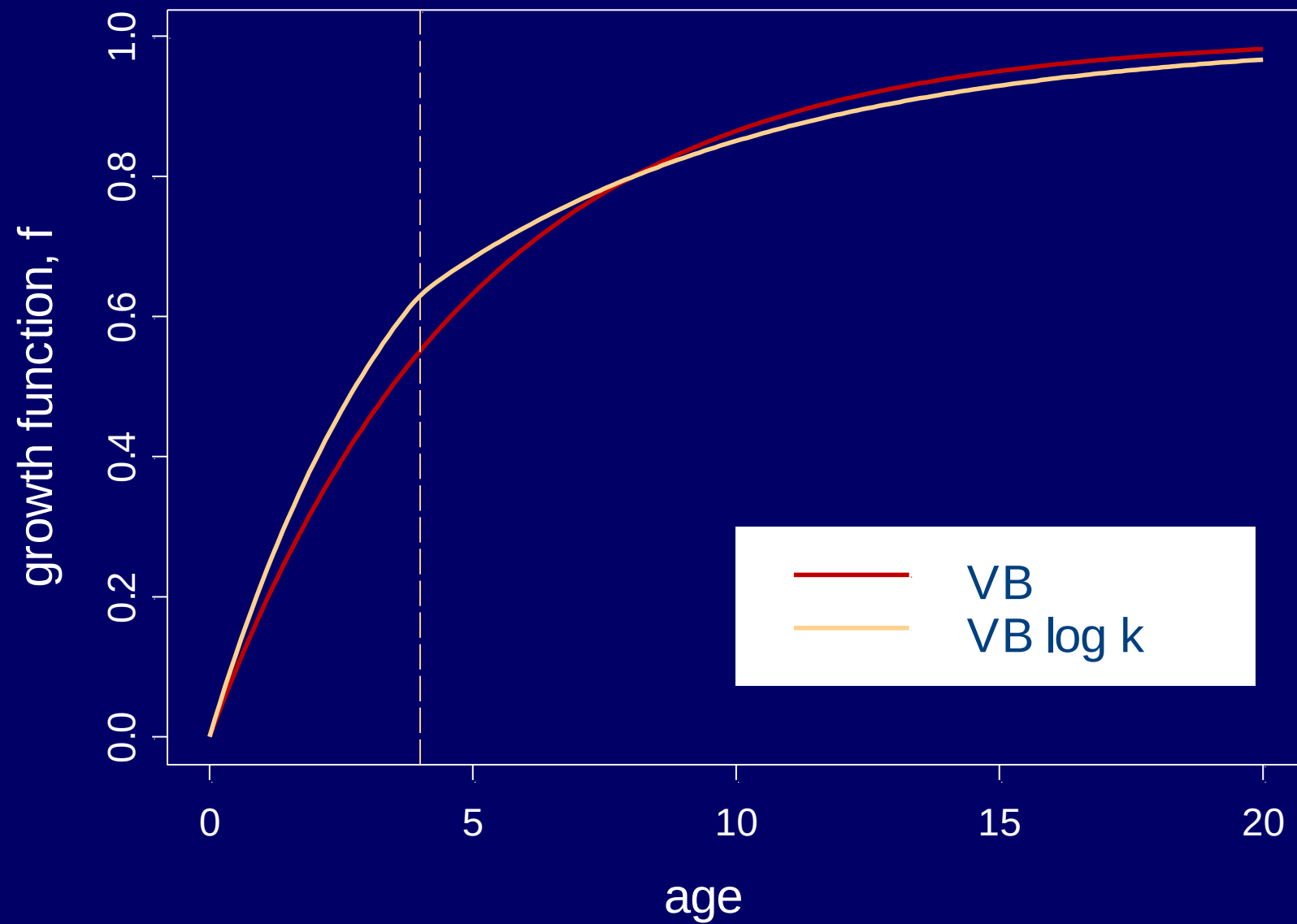
1) VB (von Bertalanffy) function:

$$f(a - a_0; \theta) = 1 - e^{-k(a - a_0)} \quad \text{where } \theta = \{k\}$$

2) VB log k function:

$$f(a - a_0; \theta) = 1 - e^{-k_2(a - a_0)} \left\{ \frac{1 + e^{-\beta(a - a_0 - \alpha)}}{1 + e^{\alpha\beta}} \right\}^{-(k_2 - k_1) / \beta}$$

where $\theta = \{k_1, k_2, \alpha, \beta\}$



Model description

- Maximum likelihood method with separate likelihood for each data source
- Data sets independent so multiply together to obtain overall likelihood to be optimized

Tag-recapture likelihood

- Model the joint density of release and recapture lengths
- Age at release unknown so we model it as a random variable, A :
 - here we assume $A \sim \log N(\mu_A, \sigma_A^2)$
- Allow for individual variability in growth by modelling asymptotic length as a random effect, $L_\infty \sim N(\mu_\infty, \sigma_\infty^2)$

Tag-recapture likelihood

For fish i tagged at time t_{1i} with length l_{1i} and unknown age a_{1i} , and recaptured at time t_{2i} with length l_{2i} :

$$l_{1i} = L_{\infty,i} f(A_i; \theta) + \varepsilon_{1i}$$

$$l_{2i} = L_{\infty,i} f(A_i + t_{2i} - t_{1i}; \theta) + \varepsilon_{2i}$$

where

$$A_i = a_{1i} - a_0 \sim \log N(\mu_A, \sigma_A^2)$$

$$L_{\infty,i} \sim N(\mu_{\infty}, \sigma_{\infty}^2)$$

$$\varepsilon_{1i} \sim N(0, \sigma_s^2)$$

Tag-recapture likelihood

Condition on A , then the joint distribution of l_1 and l_2 is bivariate normal (see Laslett, Eveson & Polacheck 2002 CJFAS 59: 976-986 for exact formulation). Then the unconditional density is:

$$\int h(l_{1i}, l_{2i} | a) p(a) da$$

where

$h(\cdot)$ is the bivariate normal density

$p(\cdot)$ is the density of A (log normal)

Otolith likelihood

For fish i with length l_i and age a_i :

$$l_i = L_{\infty,i} f(a_i - a_0; \theta) + \gamma_i$$

where

$$L_{\infty,i} \sim N(\mu_{\infty}, \sigma_{\infty}^2)$$

$$\gamma_i \sim N(0, \sigma_{\gamma}^2)$$

a_i = age calculated from # of bands in otolith

Otolith (log) likelihood

$$\ln(\lambda_{OTO}) = \frac{1}{2} \sum_i \left[\ln(2\pi V(l_i)) + \frac{(l_i - E(l_i))^2}{V(l_i)} \right]$$

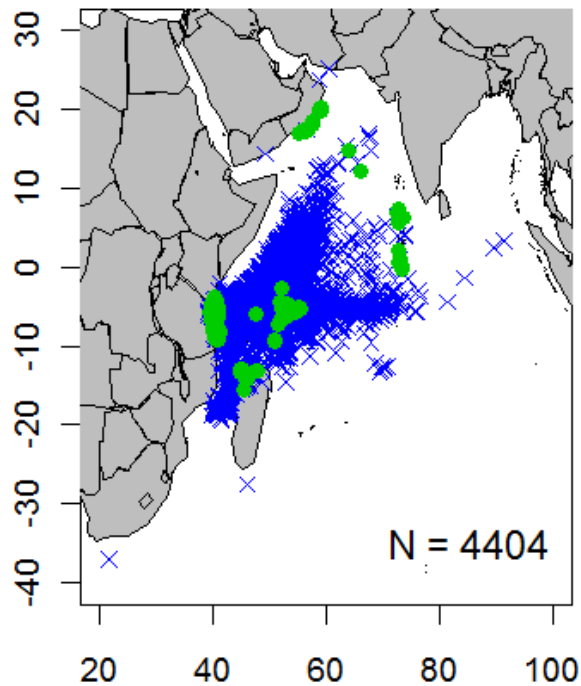
where

$$E(l_i) = \mu_{\infty} f(a_i - a_0; \theta)$$

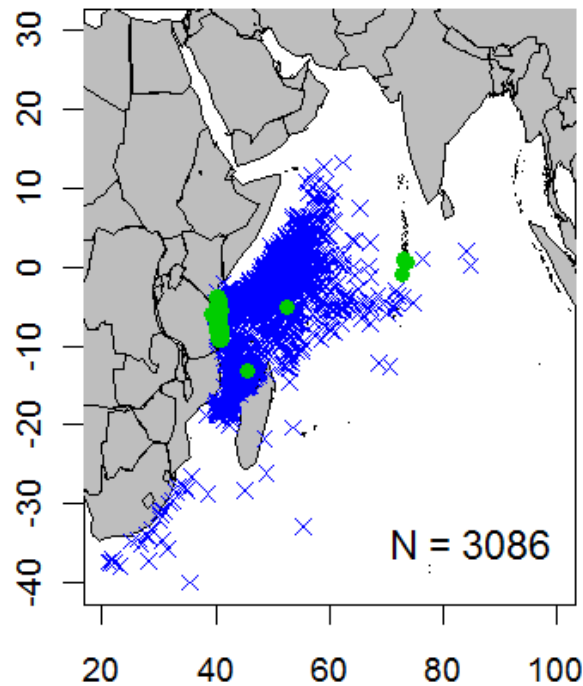
$$V(l_i) = \sigma_{\infty}^2 f(a_i - a_0; \theta)^2 + \sigma_{\gamma}^2$$

IOTTP tag-recapture data

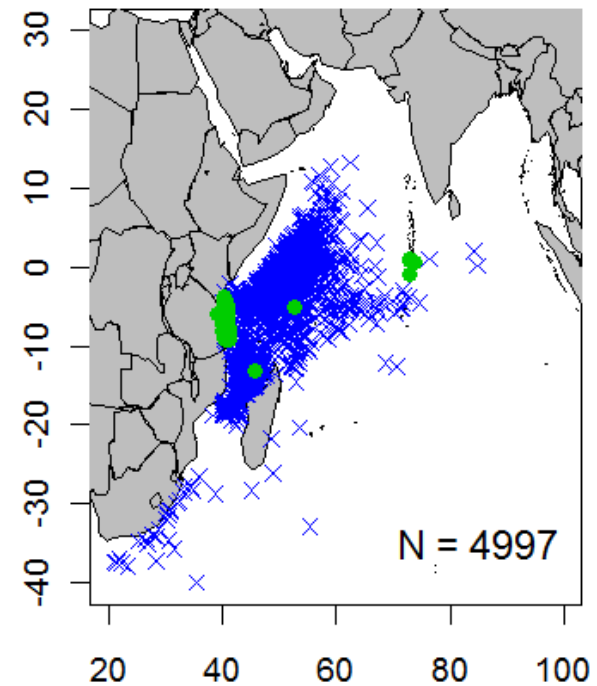
Yellowfin



Bigeye

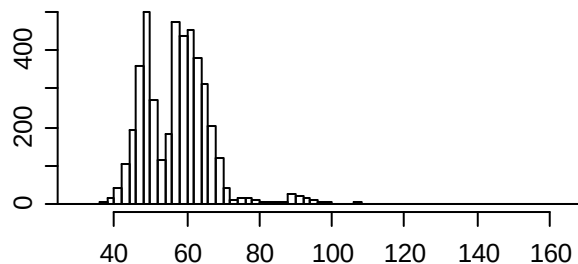


Skipjack

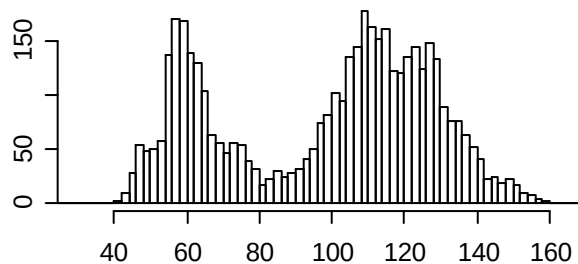


IOTTP tag-recapture data

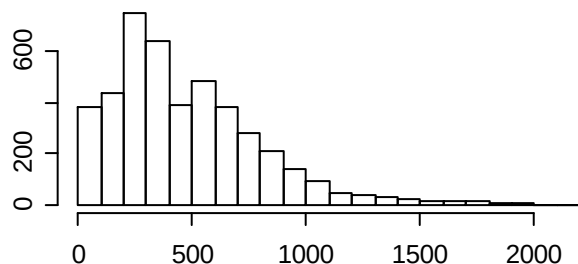
Yellowfin



Release fork length (cm)

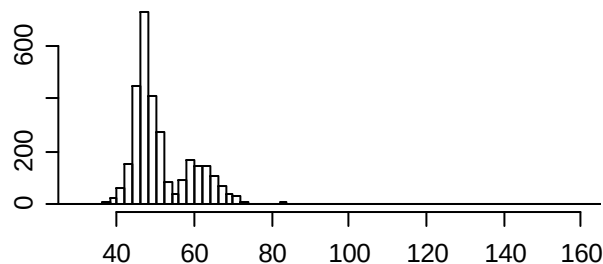


Recapture fork length (cm)

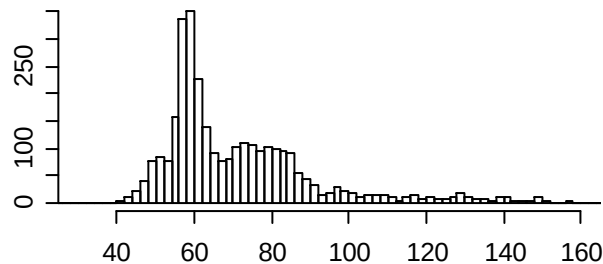


Days at liberty

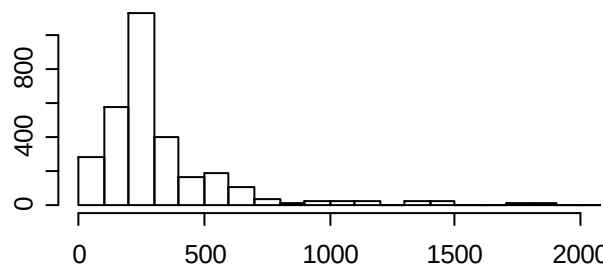
Bigeye



Release fork length (cm)

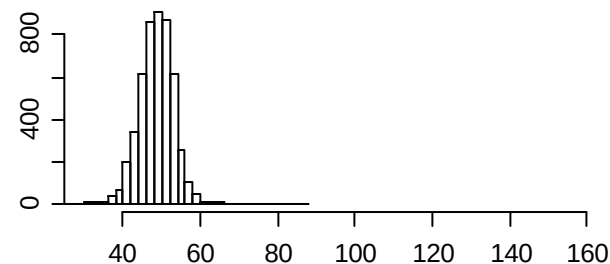


Recapture fork length (cm)

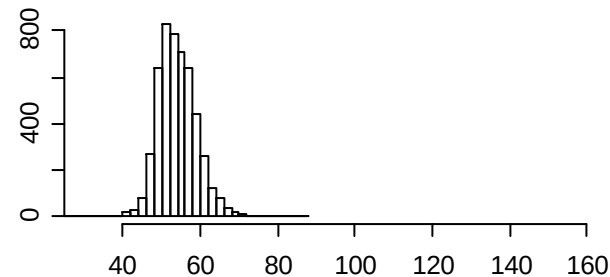


Days at liberty

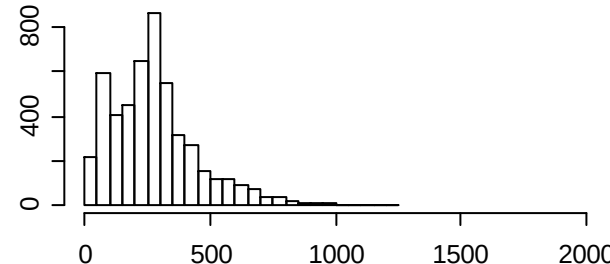
Skipjack



Release fork length (cm)

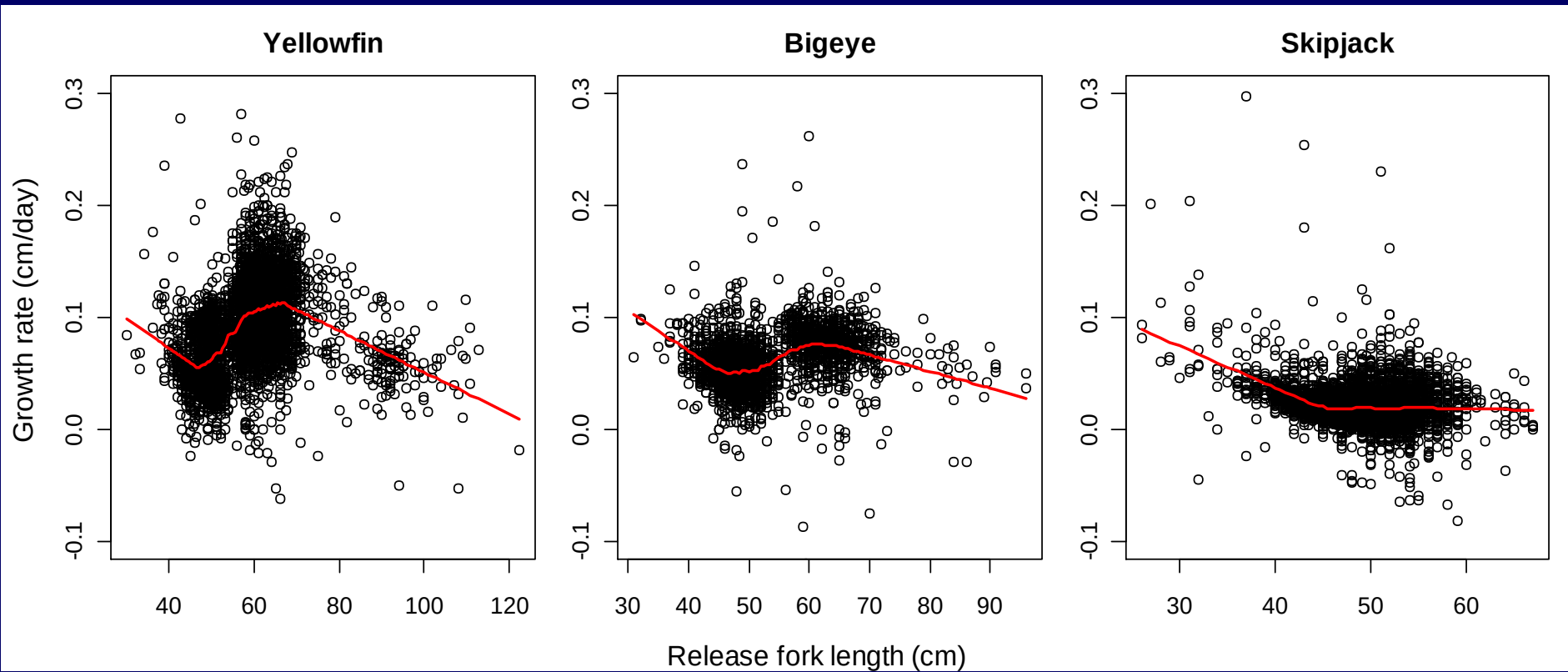


Recapture fork length (cm)

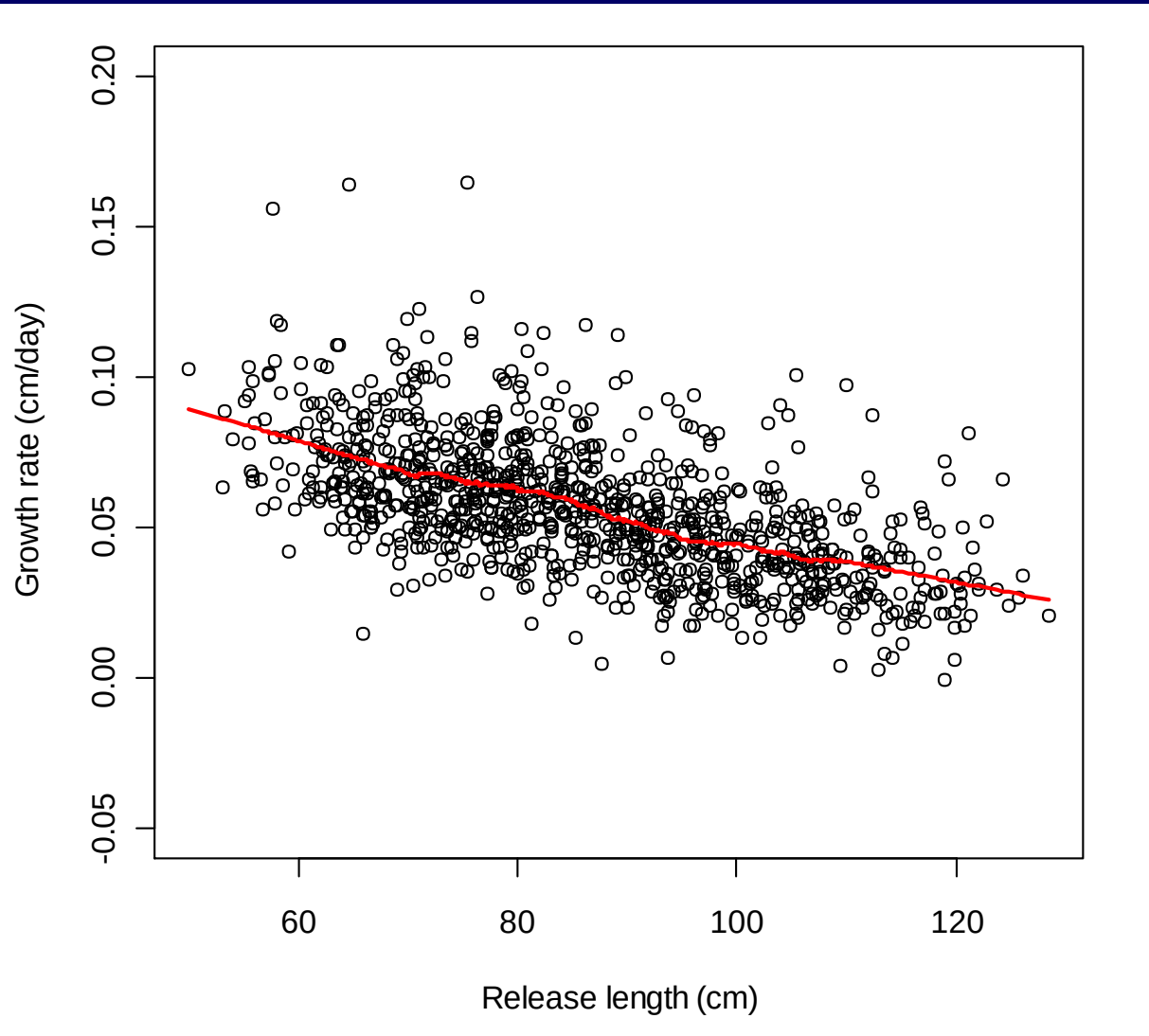


Days at liberty

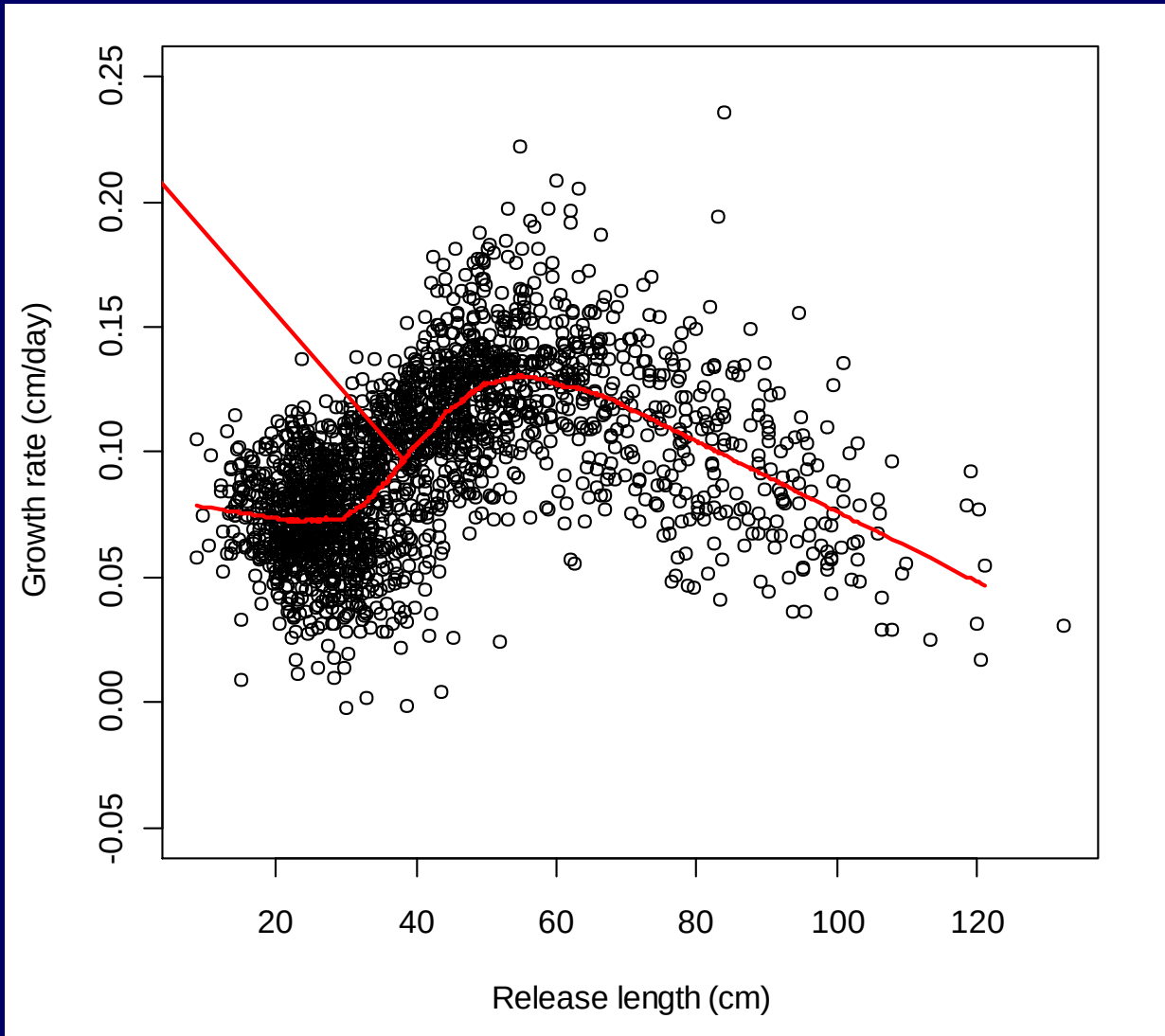
IOTTP tag-recapture data



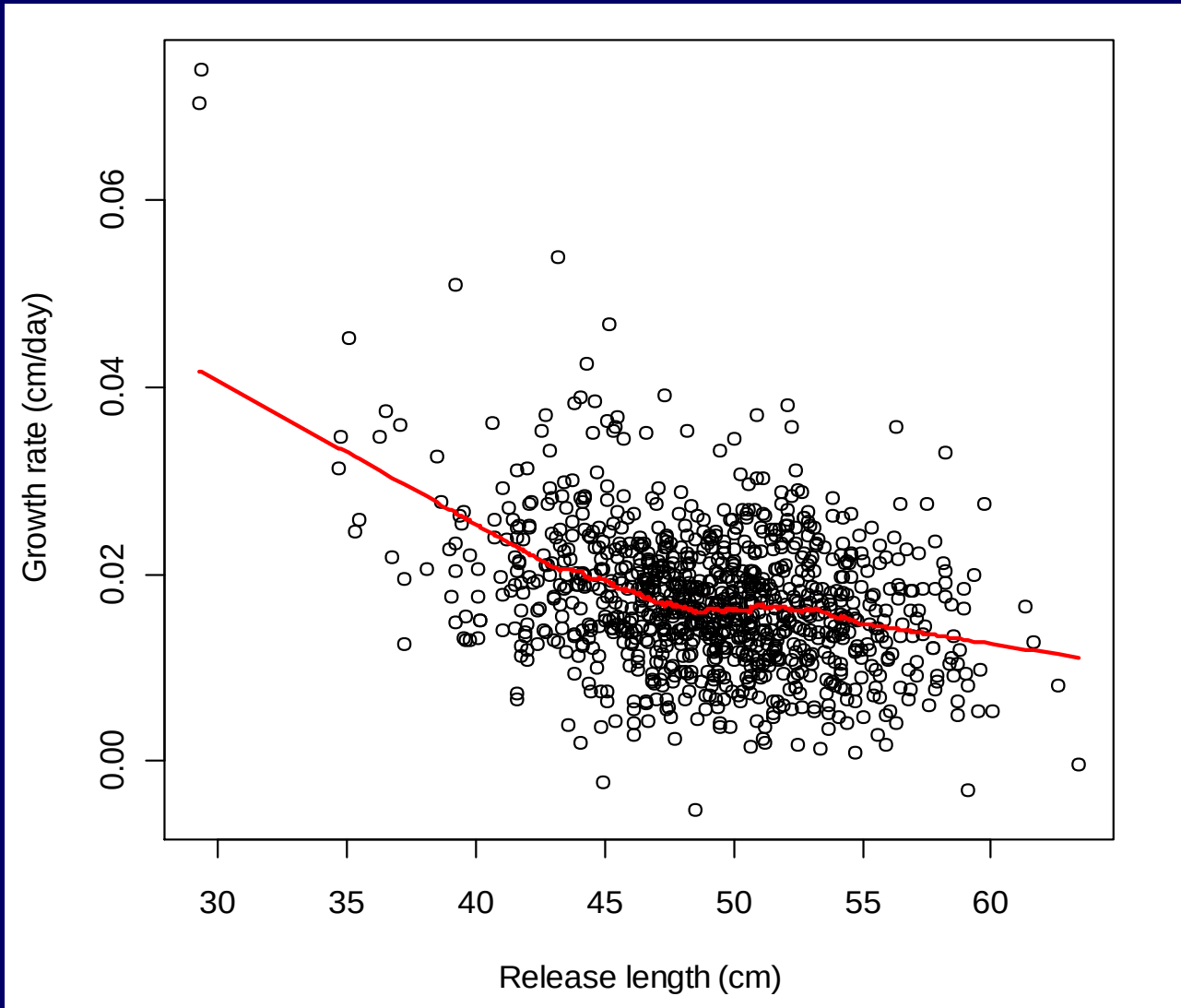
Simulated VB growth



Simulated VB log k growth with bi-modal release lengths & $k_1 < k_2$

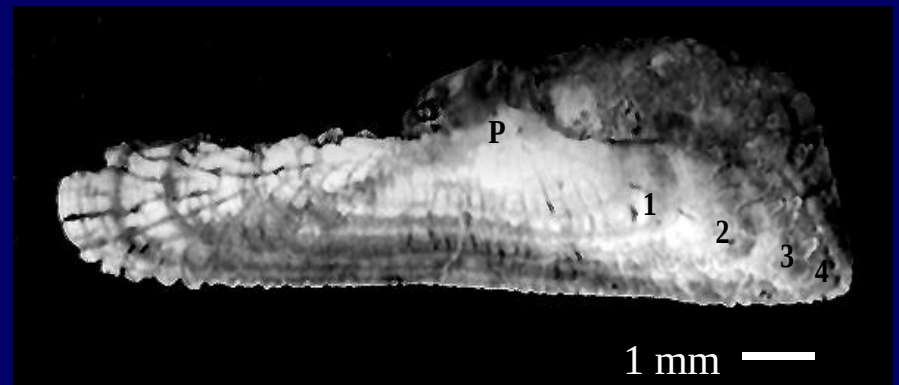


Simulated VB log k growth with uni-modal release lengths & $k_1 > k_2$

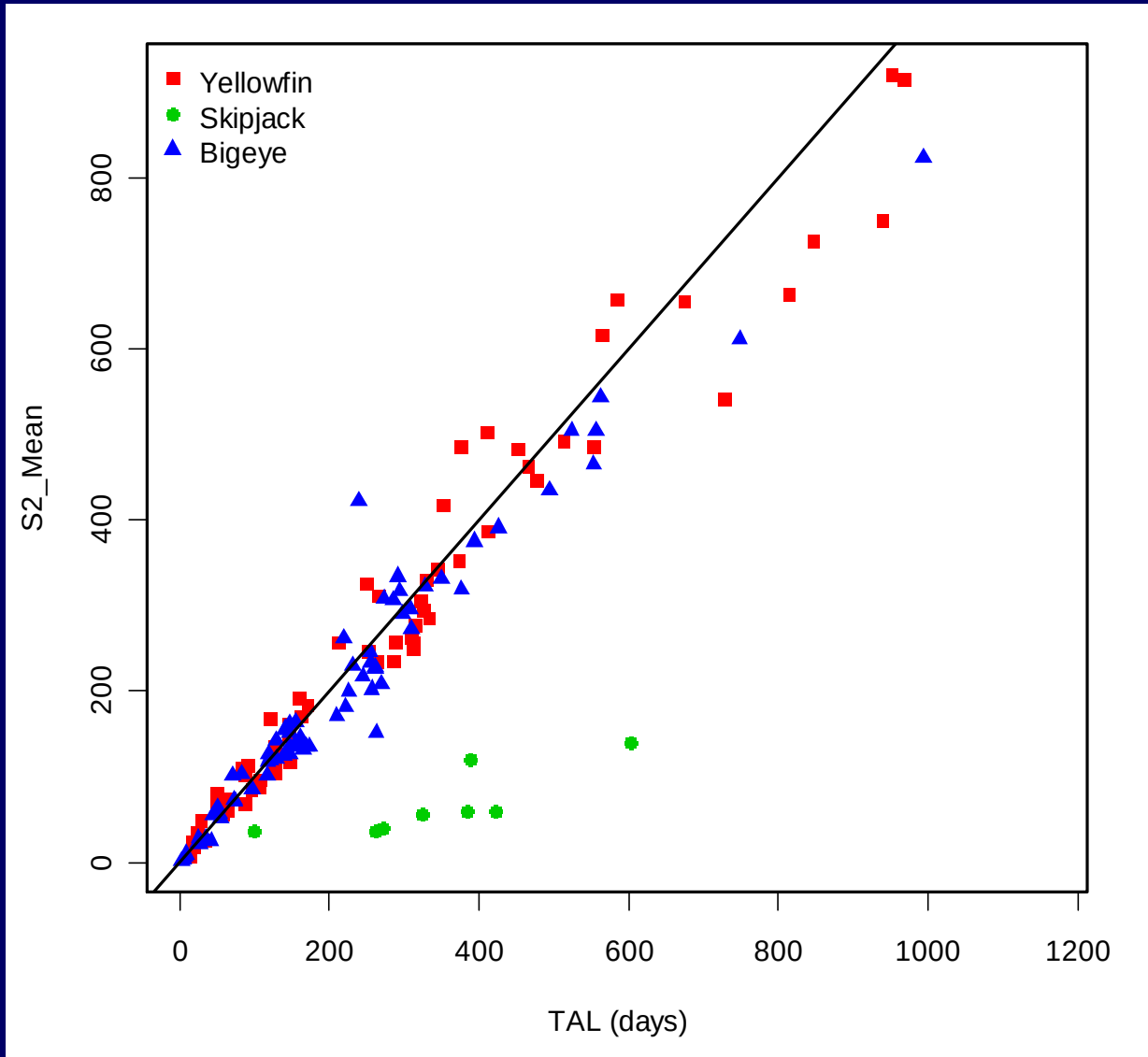


Otolith data

- Verify hypothesis that growth bands are formed daily in otoliths using OTC-marked recaptures:
 - compare counts after OTC mark with number days at liberty
- Correct counts based on relationship between number of counts and days at liberty

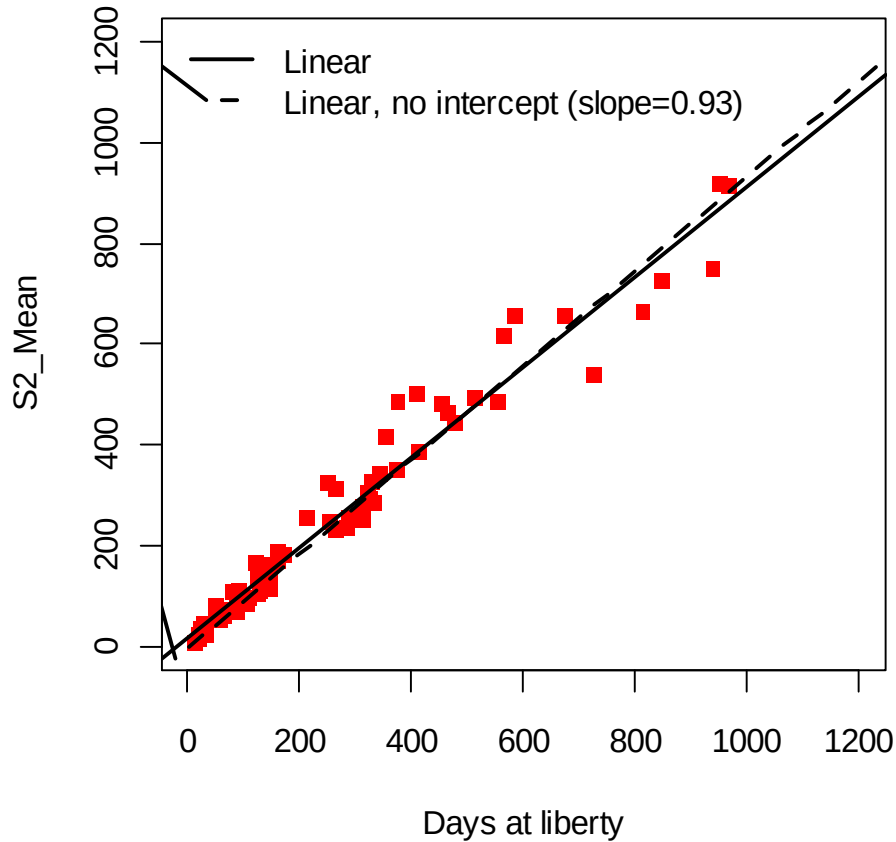


Otolith counts from OTC-marked fish vs. days at liberty

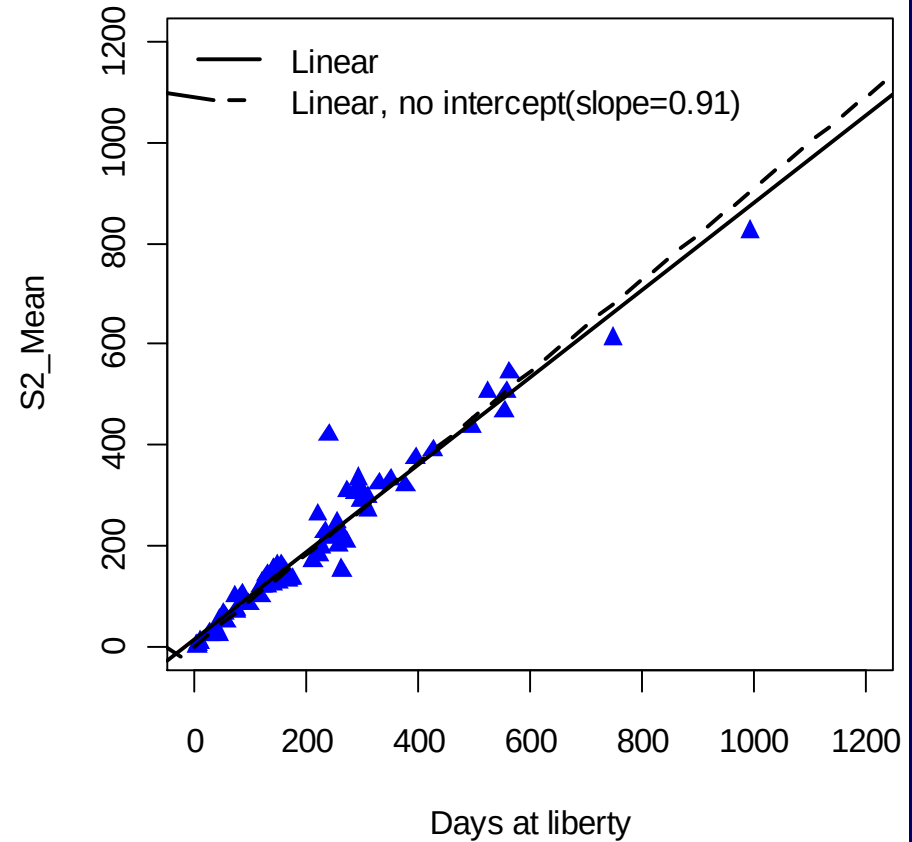


Otolith counts from OTC-marked fish vs. days at liberty

Yellowfin

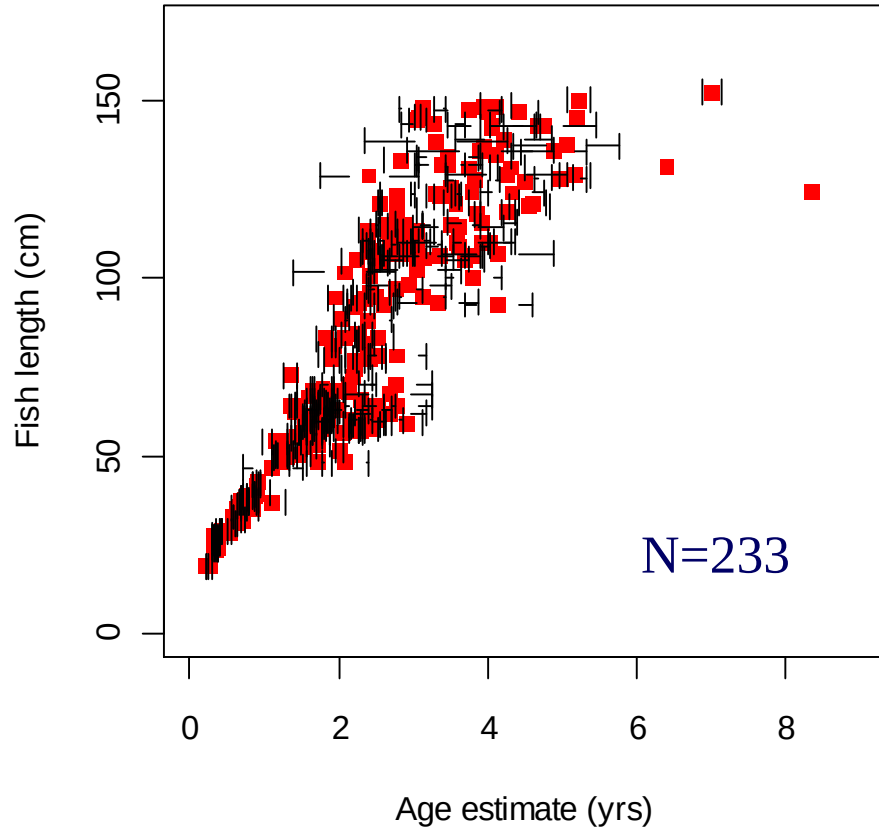


Bigeye

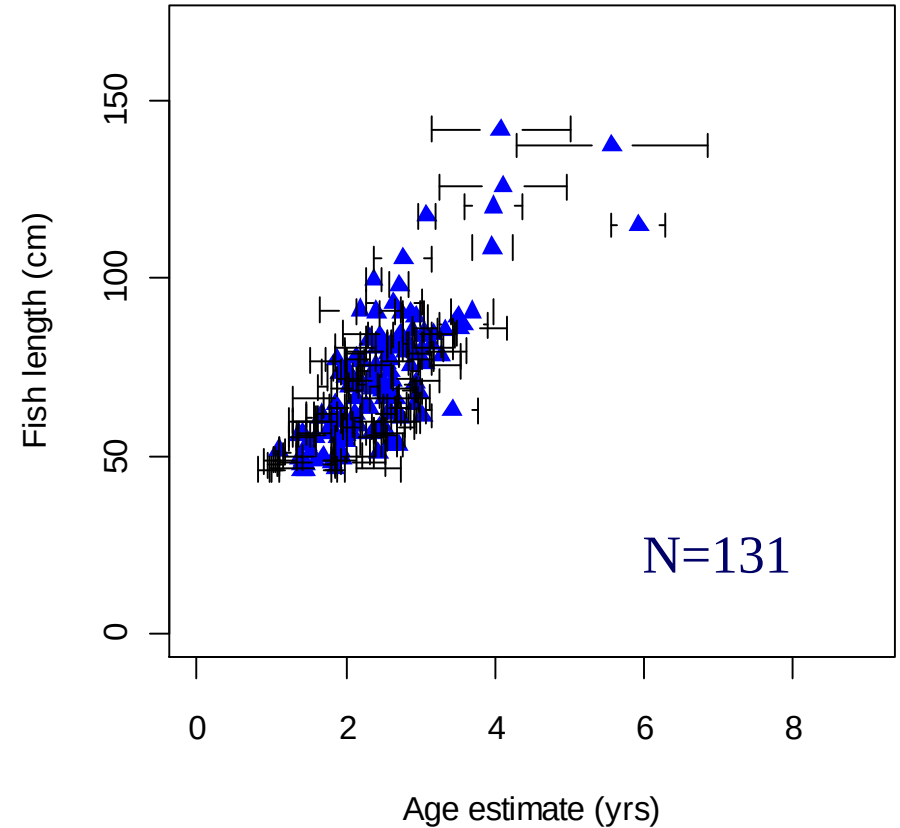


Corrected otolith age and length data

Yellowfin



Bigeye



Yellowfin results

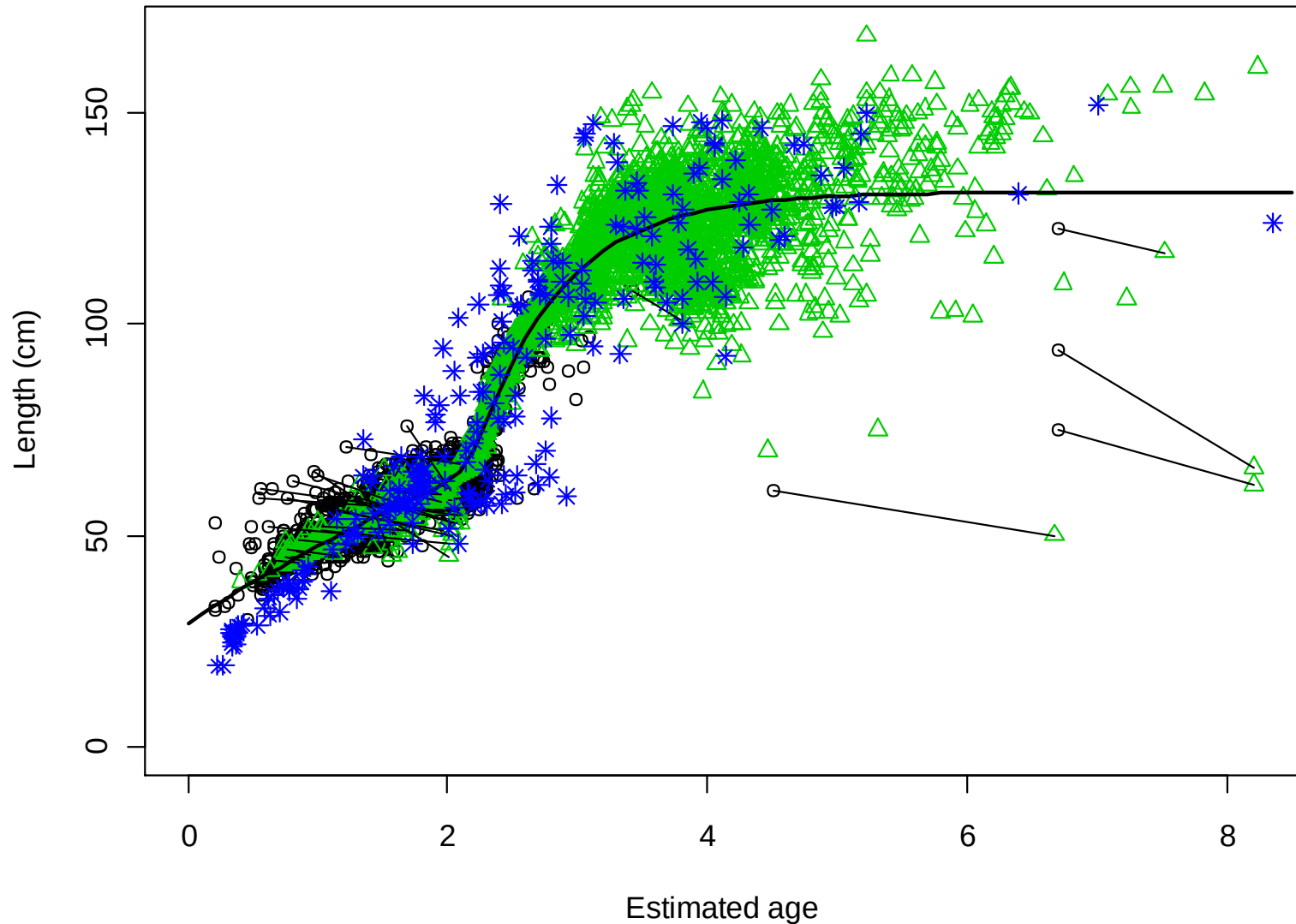
Model	μ_∞	σ_∞	k1	k2	α	β	$\mu.A$	$\sigma.A$	$\sigma.tag$	a0	$\sigma.oto$
1	132.3	10.1	0.20	1.51	3.5	15.6	1.06	0.14	3.9	-1.3	11.9
2	145[#]	10.4	0.15	0.94	3.8	20 [^]	1.18	0.13	4.3	-1.7	12.8

*Estimate on lower bound

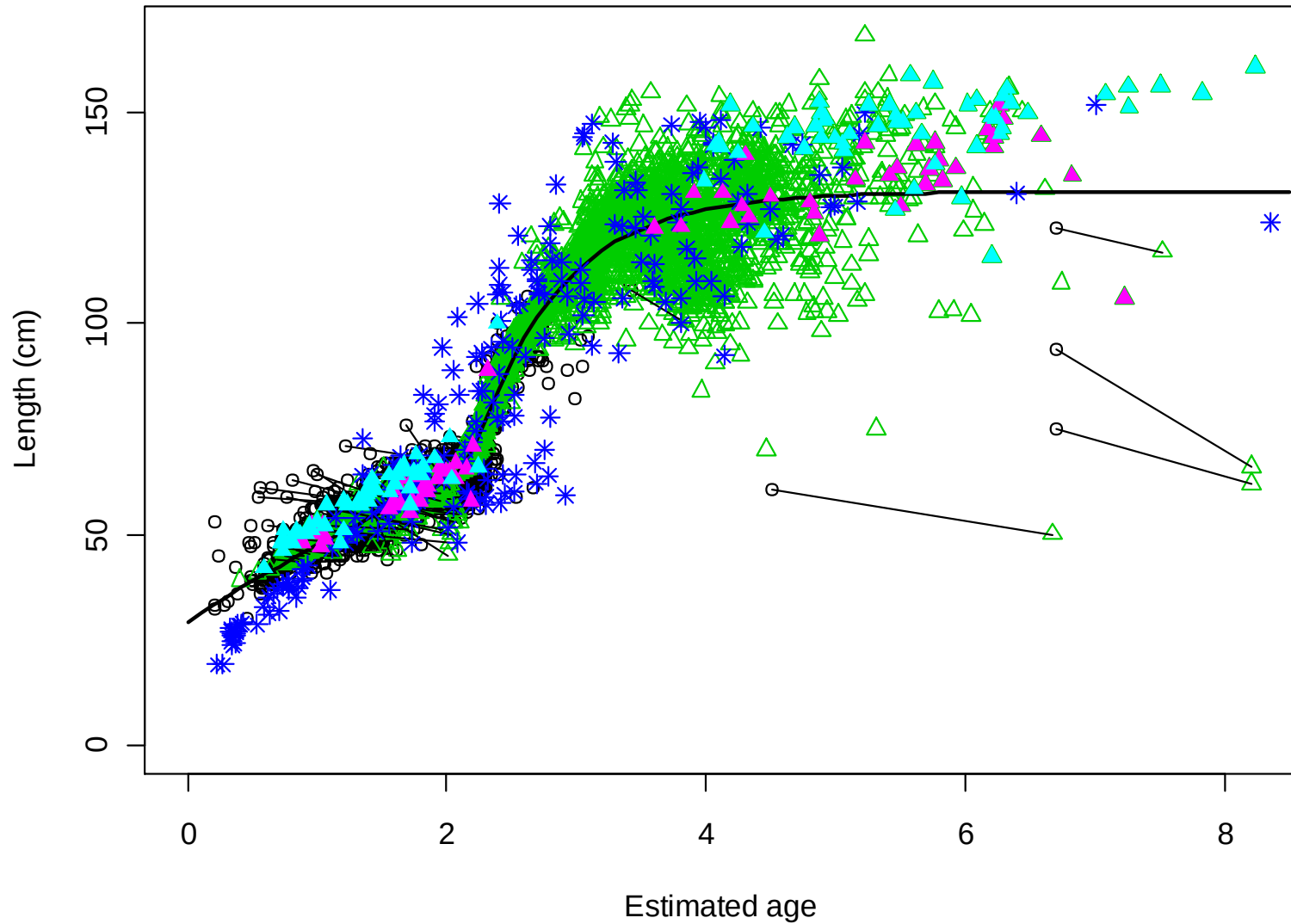
[^]Estimate on upper bound

[#] **Fixed**

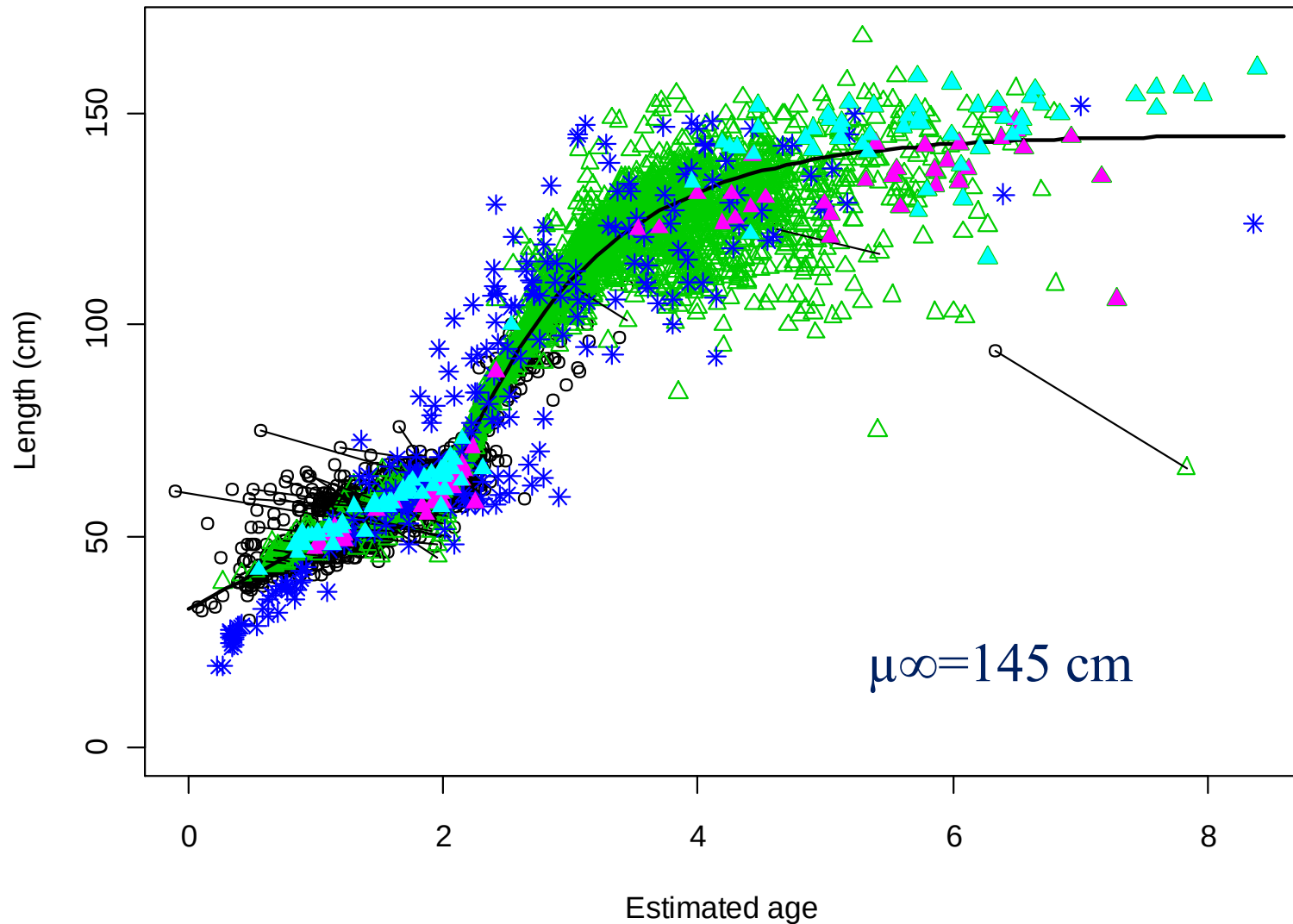
YFT Model 1: fitted growth curve



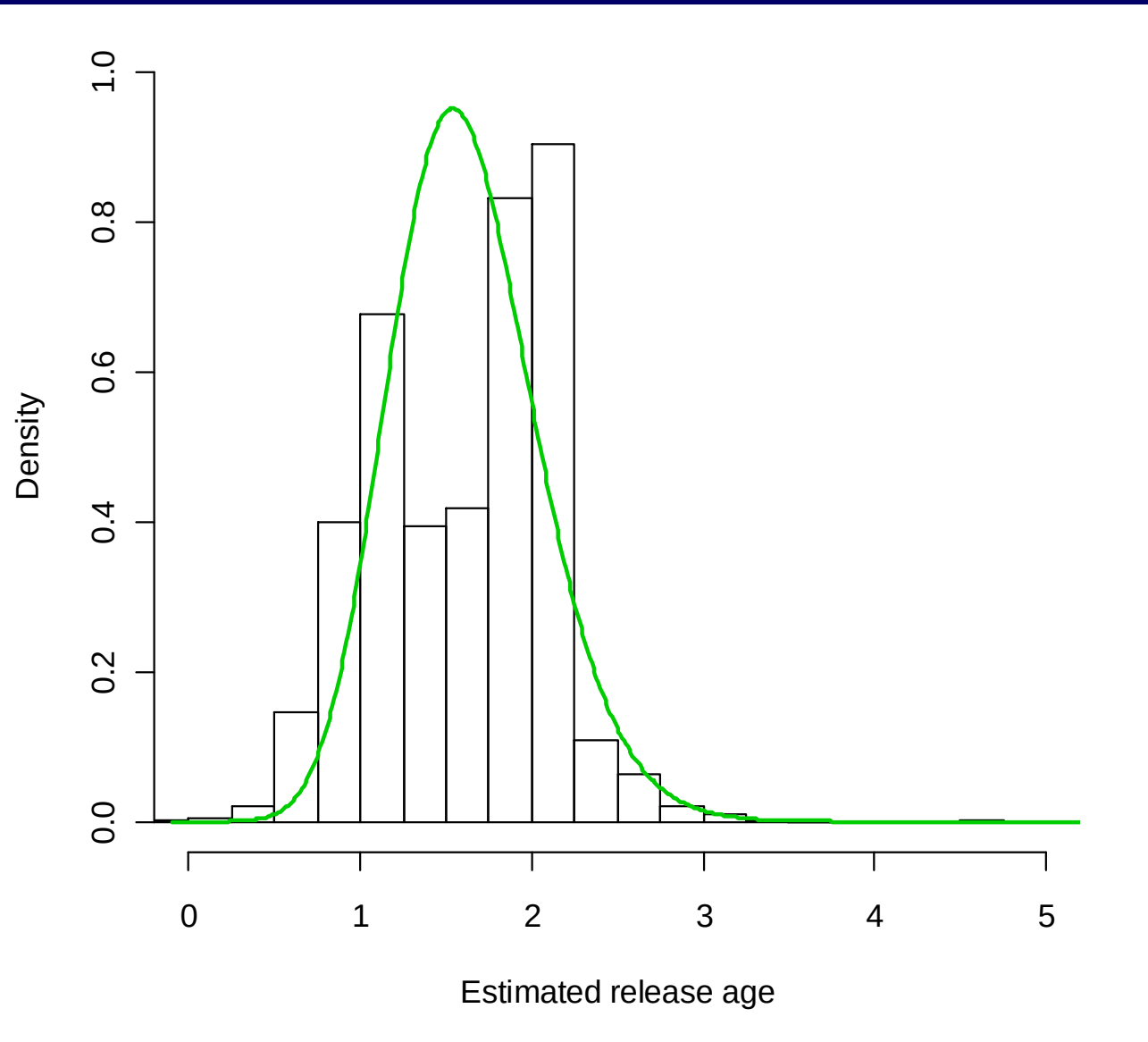
YFT Model 1: fitted growth curve



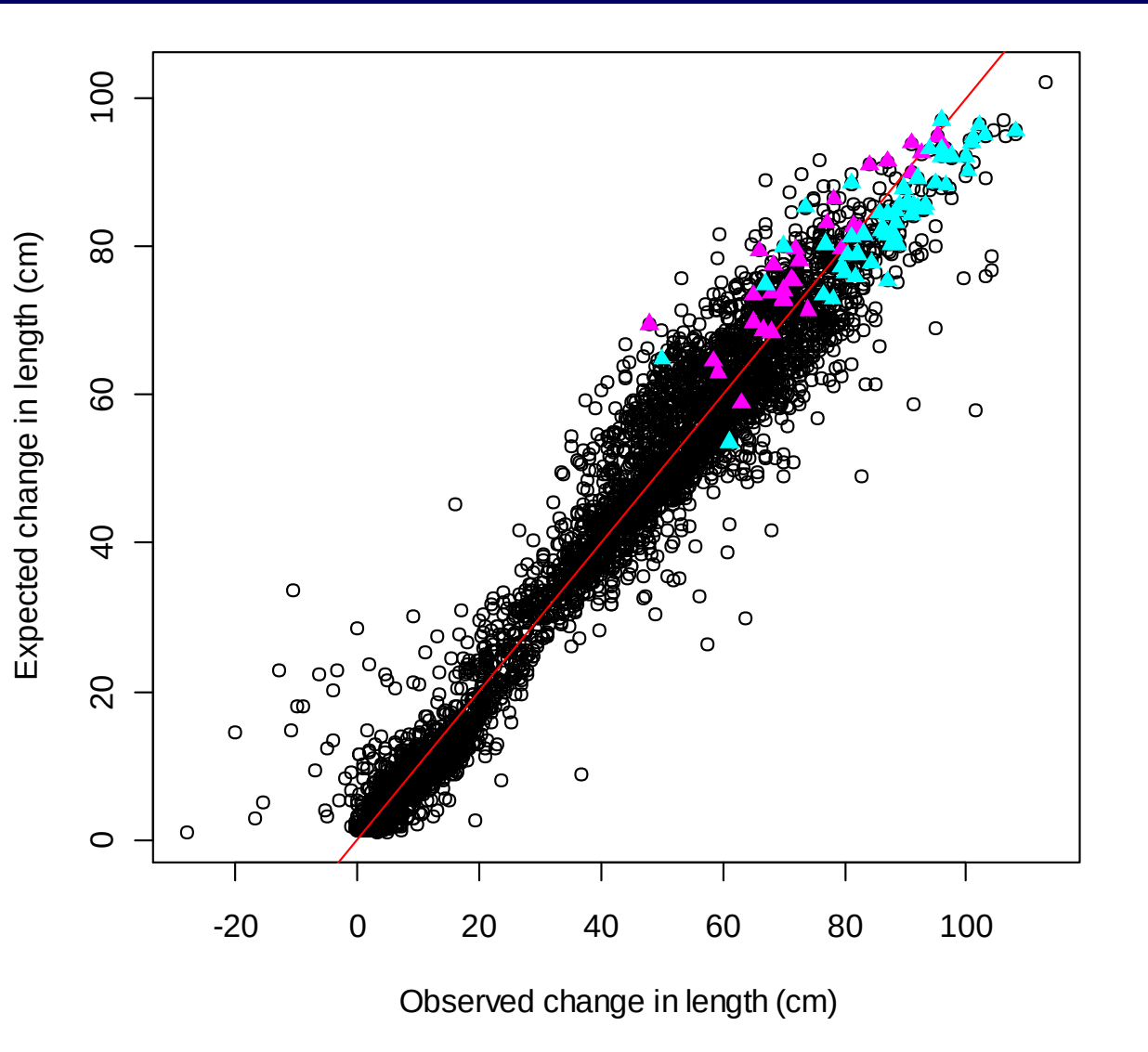
YFT Model 2: fitted growth curve



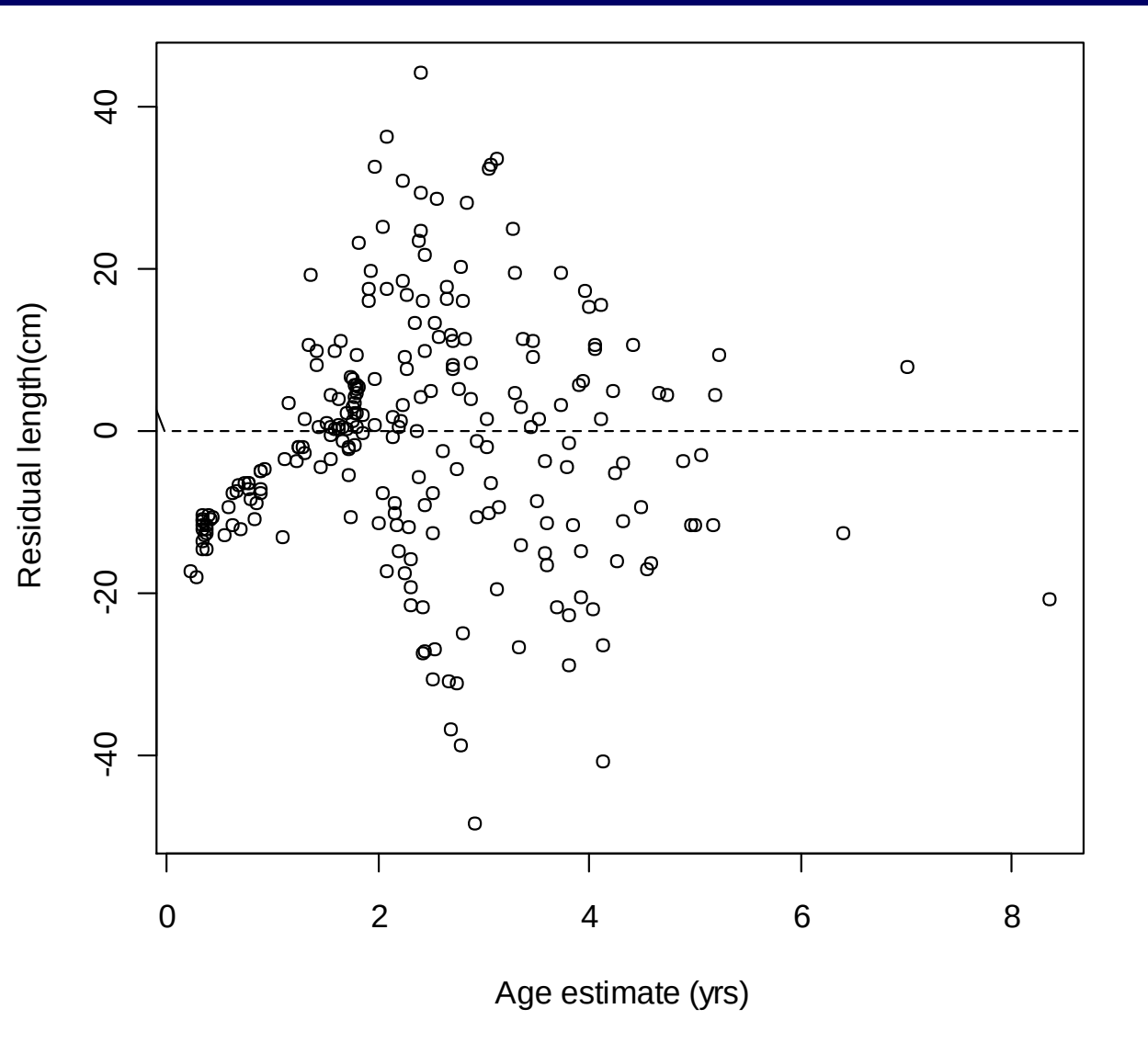
YFT Model 2: tag-recapture fit



YFT Model 2: tag-recapture fit



YFT Model 3: otolith fit



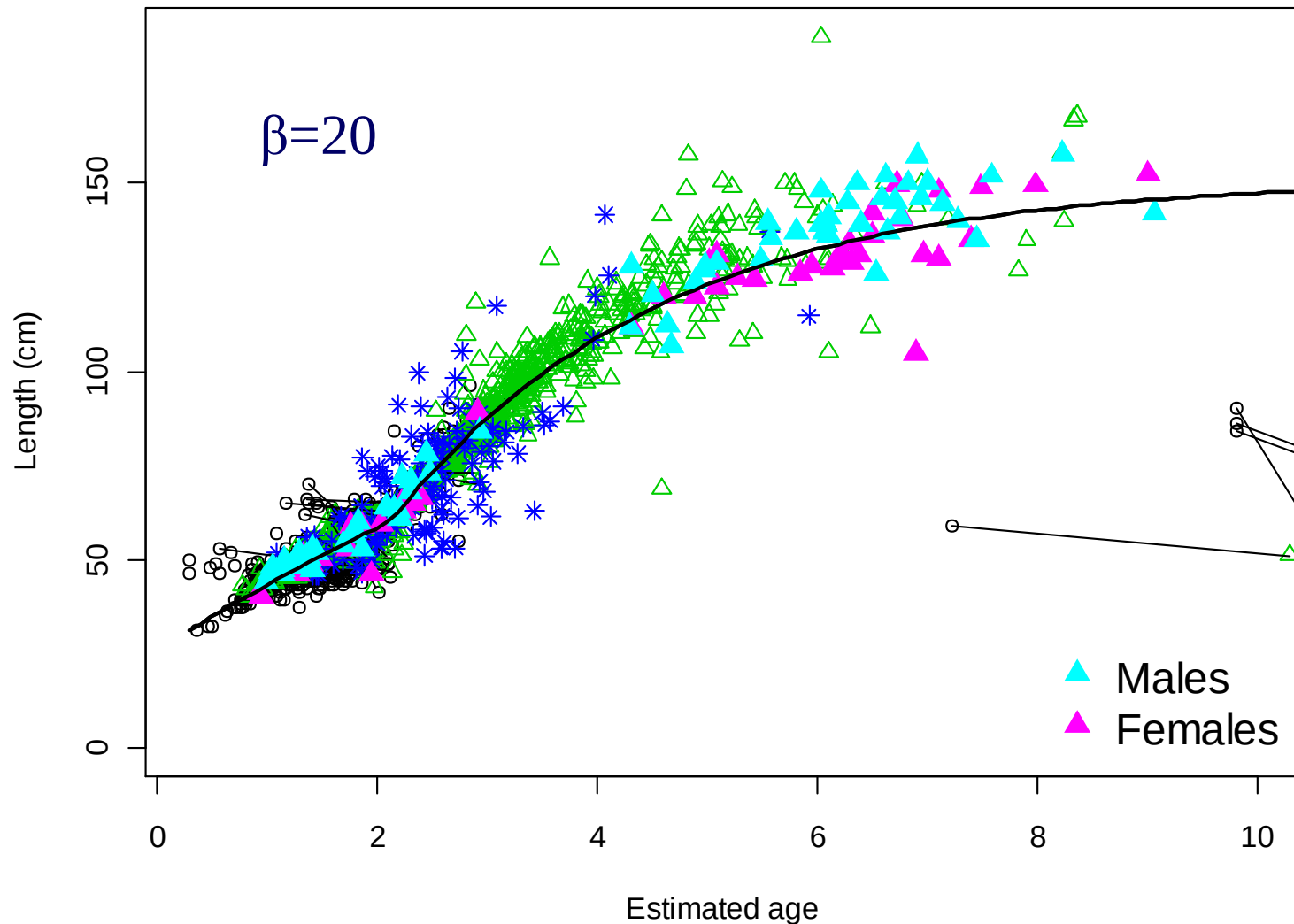
Bigeye results

Model	μ_∞	σ_∞	k1	k2	α	β	$\mu.A$	$\sigma.A$	$\sigma.tag$	a0	$\sigma.oto$
1	145.5	9.5	0.02	0.55	17.1	2.3	2.8	0.03	2.9	-15*	10.5
2	150.9	11.7	0.15	0.41	3.4	20#	1.02	0.14	3.1	-1.2	7.8

*Estimate on lower bound

Fixed

Bigeye Model 2: fitted growth curve

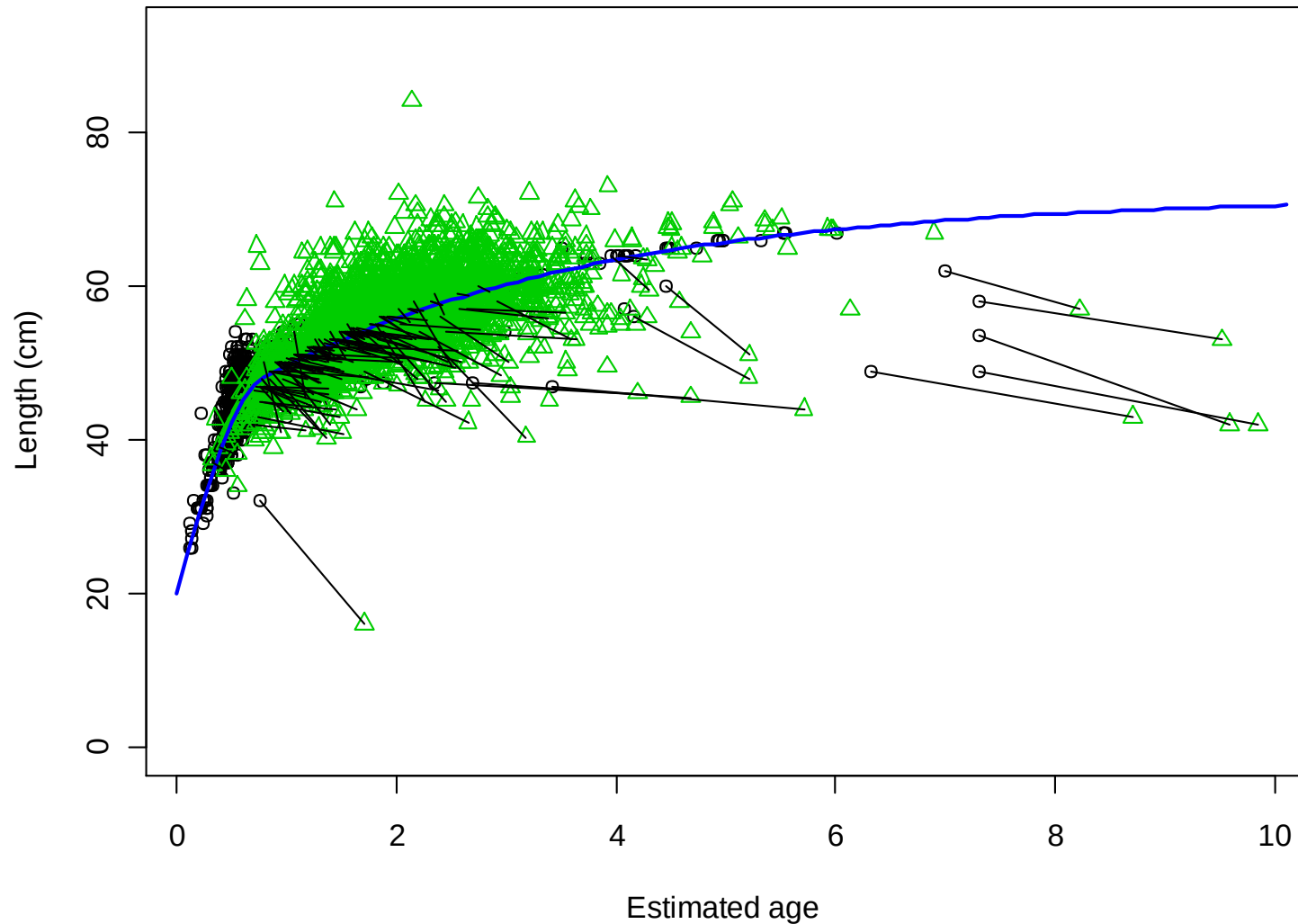


Skipjack results

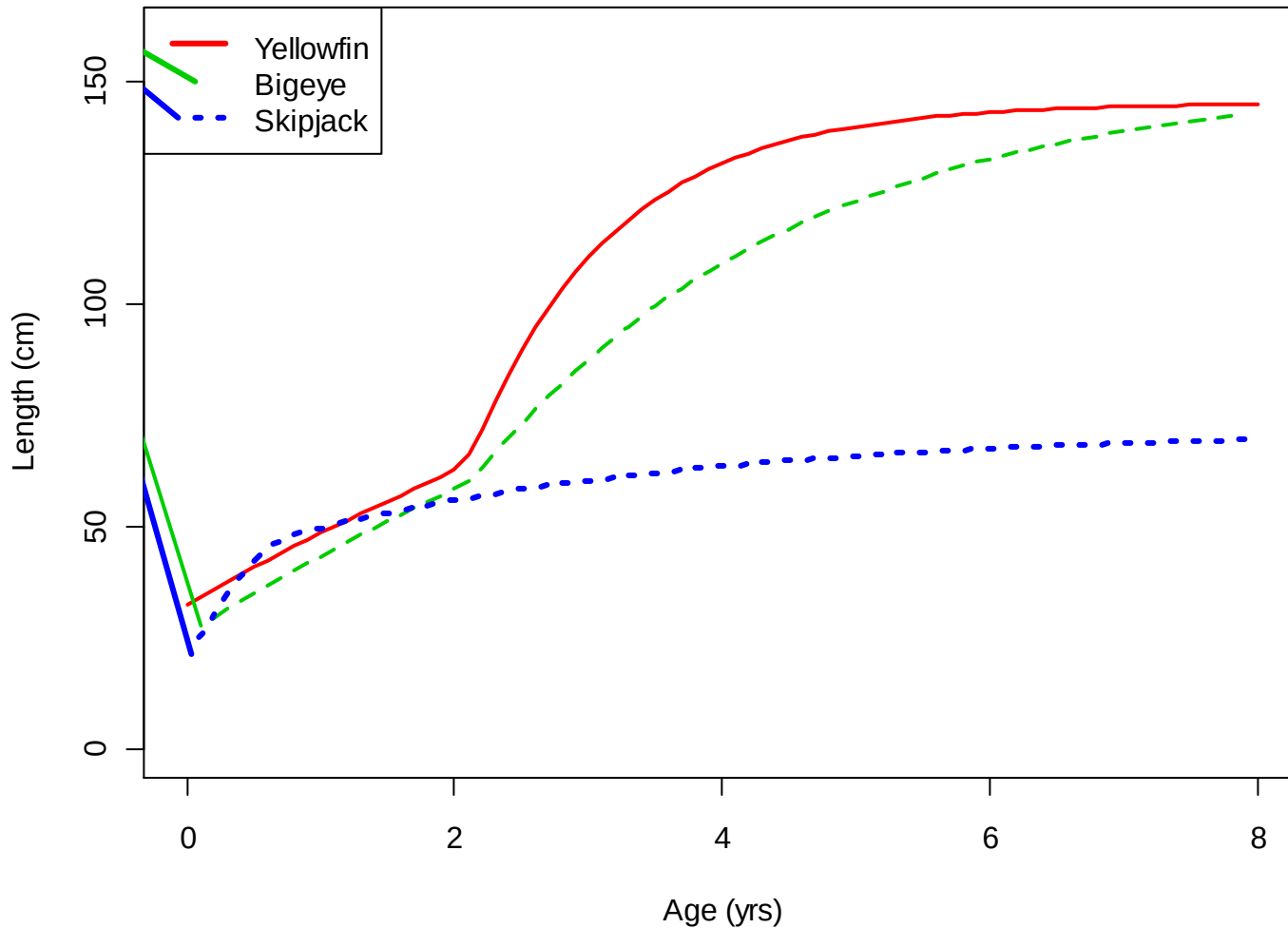
Model	μ_{∞}	σ_{∞}	k1	k2	α	β	$\mu.A$	$\sigma.A$	$\sigma.tag$	a0*
1	71.6	4.9	1.12	0.33	0.95	24.8	0.22	0.23	1.7	-0.3

* Calculated after fitting the model so that $L(0) = 20\text{cm}$
(since we do not get an estimate of a_0 from the model with tag-recapture data alone)

Skipjack Model 1: fitted growth curve



Comparison of mean growth curves for the three species



Future Work

- Incorporate error in the ages of the direct aging data
- Obtain realistic standard errors for parameter estimates
 - difficulty getting from inverse Hessian
 - likelihood profiles more reliable but very slow
- Allow for growth parameters to change over time (some work done on this)