

#### Investigating spatial variation in growth of bigeye in the Pacific based on annual ageing data

Presenter: Paige Eveson

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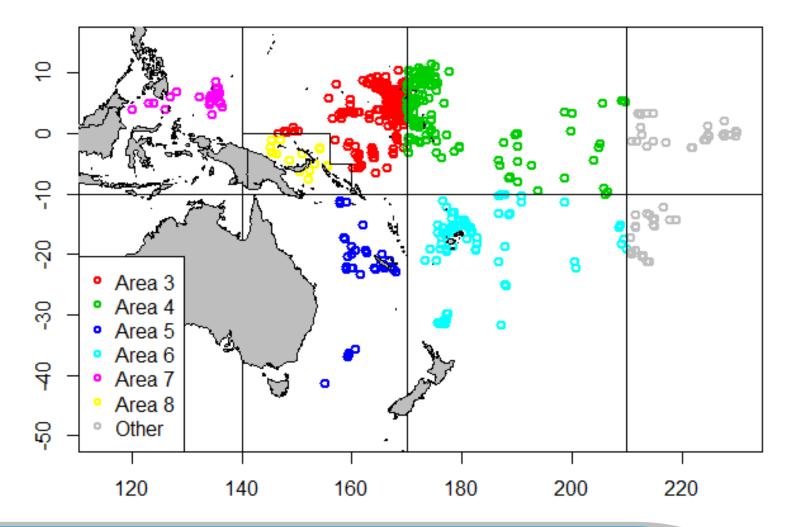
WORKSHOP ON AGE AND GROWTH OF BIGEYE AND YELLOWFIN TUNAS IN THE PACIFIC OCEAN, JANUARY 23-25, 2019, LA JOLLA, CALIFORNIA, USA

# Investigate spatial variation in growth

- Differences in growth observed between EPO and WPO
- Could be due to:
  - differences in otolith age estimates
  - true spatial differences in growth
- Can investigate spatial differences using all the annual ageing data we have for the Pacific from otoliths read using the same technique and reader



### **Spatial management areas**



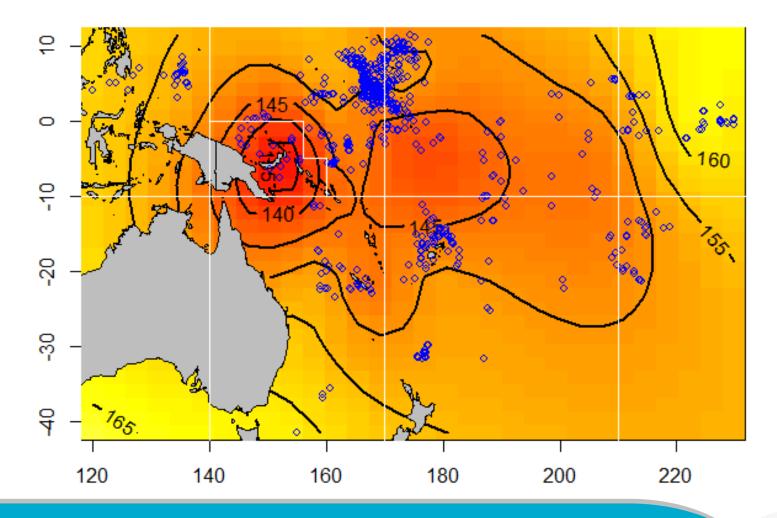
# Investigate spatial variation in growth

- Fit GAM to length data with age and (lat,lon) as smooth terms: Length ~ s(Age) + s(Lat,Lon)
- Output from R:
- Approximate significance of smooth terms:
- edf Ref.df F p-value
- s(Decimal\_age) 7.528 8.468 786.925 <2e-16 \*\*\*
- s(Lon,Lat) 25.814 28.312 6.895 <2e-16 \*\*\*
- R-sq.(adj) = 0.909 Deviance explained = 91.2%
- GCV = 83.008 Scale est. = 80.789 n = 1285
- So spatial term came out highly significant
- Can plot spatial map of predicted length at any specified age



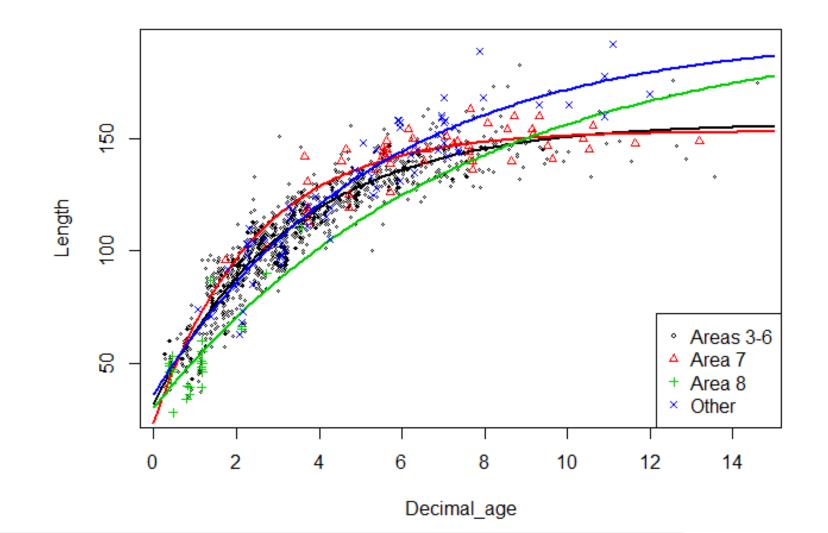
### **Spatial GAM results**

#### Predicted length at age 10





#### VB fit to combined areas based on spatial GAM





# Discussion

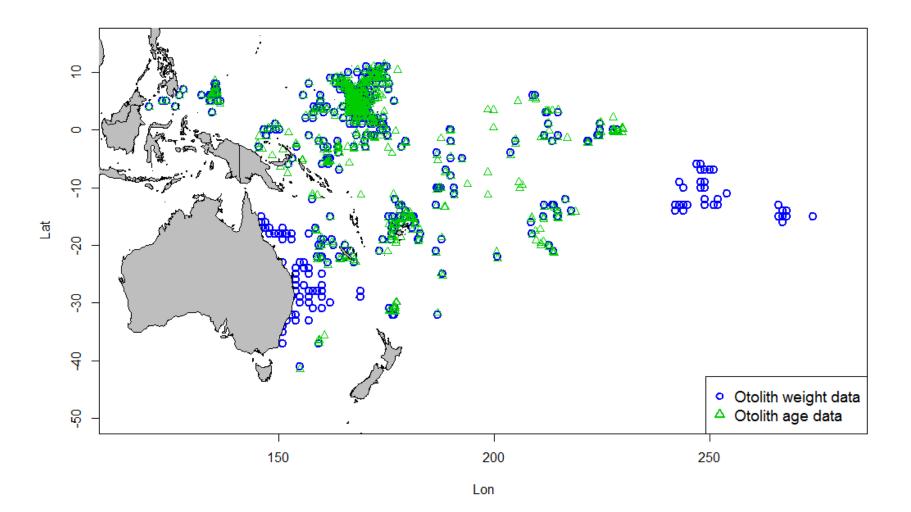
- Fairly strong evidence of spatial variation in growth across Pacific
- However, limited sample sizes and age/length ranges by regions make full assessment difficult
- Good to get otolith data from further east to see if trend toward larger fish continues
- Otolith weight data may be able to help with last two points...



# Using otolith weight to investigate spatial variation in growth

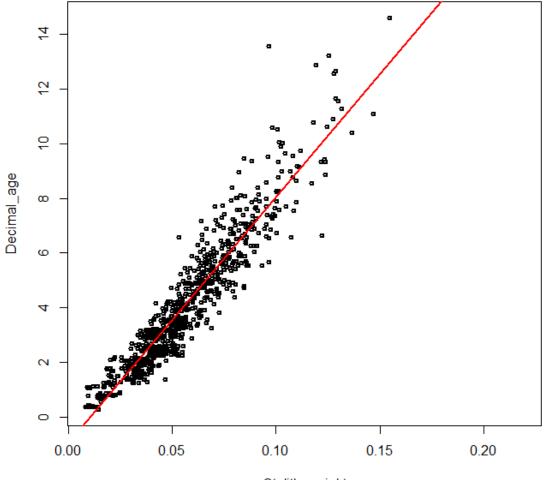
- Often have otolith weight information from samples without age estimates
- Correlation between otolith weight and fish age is often strong
- Therefore, may be useful to use otolith weight in place of age in spatial GAMs to increase sample sizes and spatial range of data
- Tried this approach for bigeye in the Pacific

#### **Otolith weight vs age sample locations**





## **Otolith weight vs age**

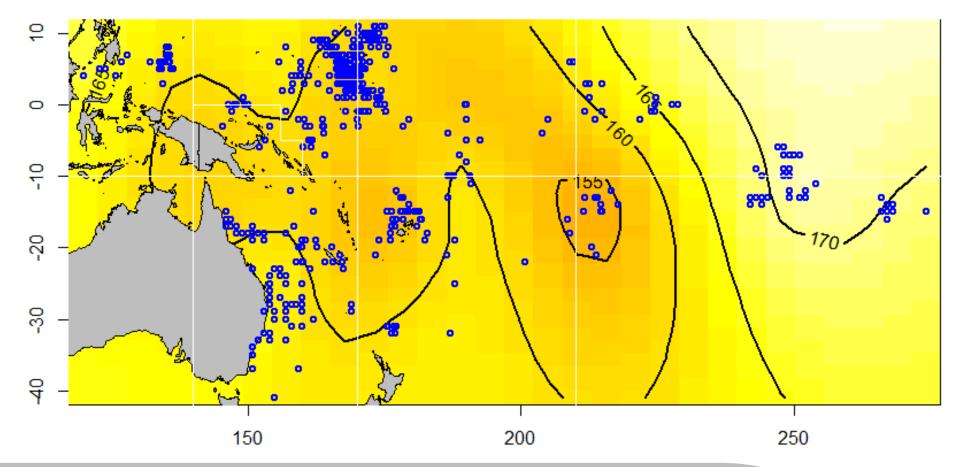


Otolith\_weight



# Spatial GAM results using otolith weight

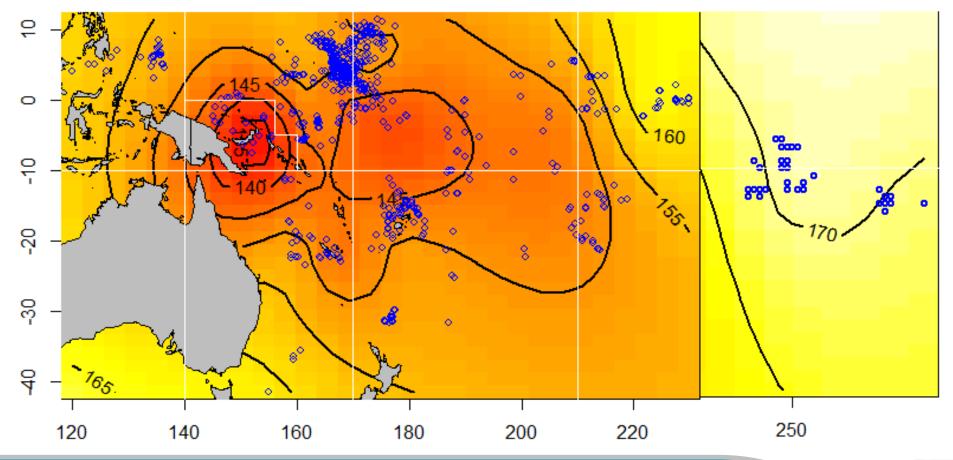
#### Predicted length at otolith weight 0.122





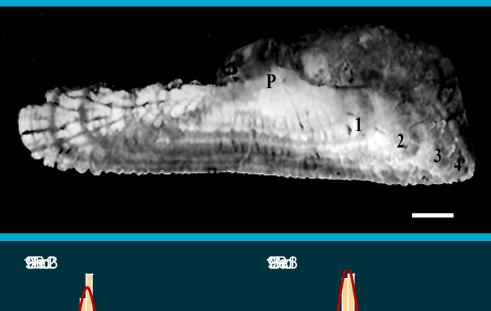
# Spatial GAM results using otolith weight

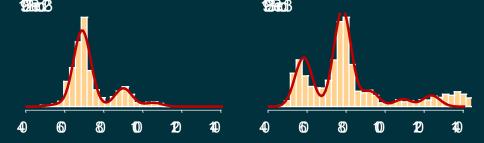
#### Predicted length at otolith weight 0.122











# Integrating multiple data sets into growth models

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# Why an integrated model?

- A single source of data does not generally contain complete growth information
- Different data sources contain information about different portions of the life cycle and different aspects of growth
- Using multiple data sources enables us to check for biases and inconsistencies



# **Sources of growth information**

Three common data sources:

- Release and recapture length and time at liberty data from tag-recapture experiments
- Direct age data obtained from hard-parts (e.g. otoliths) and fish length
- Length-frequency data from commercial catches



# LEP maximum likelihood approach

- Develop a likelihood for each data set separately
- Multiply likelihoods together to obtain an overall likelihood to be optimized (assuming data sets are independent)
- Requires growth function to be of form L<sub>∞</sub>\*f(age,θ) (e.g. for VB, f = 1-exp{-k\*(age-t0)}, θ={k,t0})
- To allow for individual variability in growth, model  $L_\infty$  as a random variable (optional)



#### **Tag-recapture component**

- Model the joint density of release and recapture lengths
- <u>Age at release unknown</u>
- Model it as a random variable, A
- For many species, reasonable to assume  $A \sim \log N(\mu_A, \sigma_A^2)$



# **Otolith component**

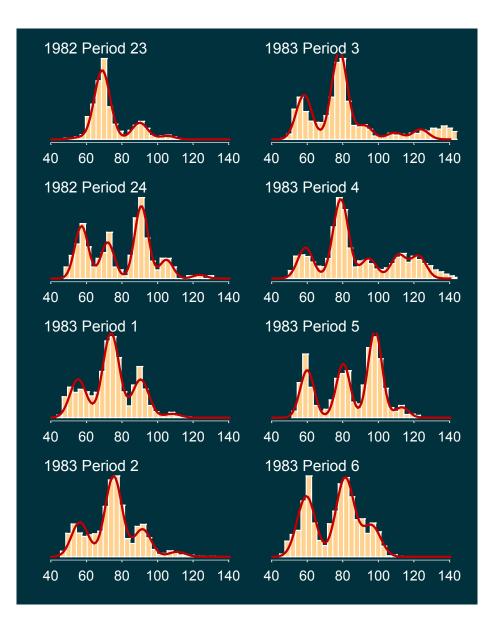
- Assign an age to each fish based on the number of annual or daily increments
- In case of annual increments, estimate decimal age using an assumed average date of birth and date of band formation for the population
- Model fish length as a function of age, treating the ages as exact



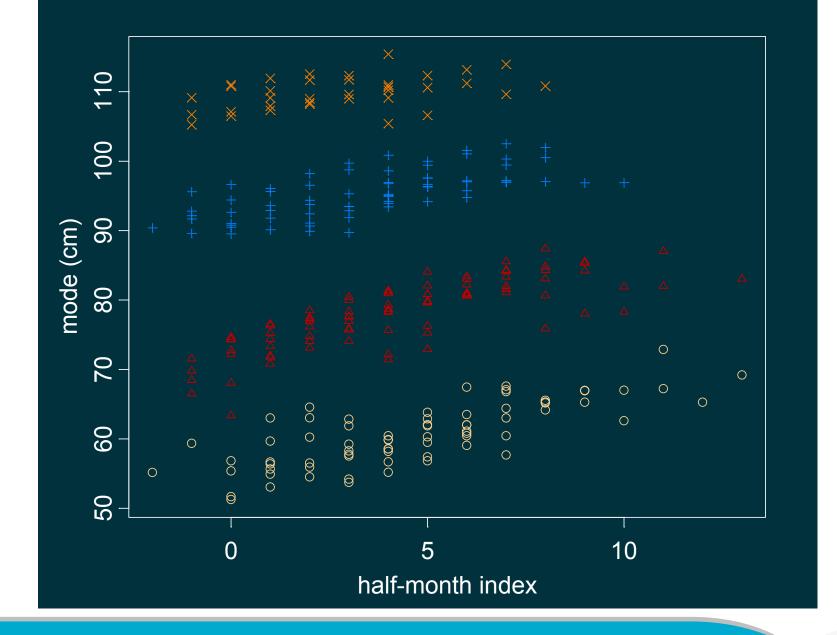
## **Length-frequency component**

- Step 1: Decompose the length-frequency distributions into modes using a Gaussian mixture model
- Step 2: Model mean fish length using the estimated modes and their standard errors obtained in Step 1









# Investigating compatibility of otolith and tag-recapture data using simulations

Simulate VB growth data with parameter values:

- n.tag = 500
- n.oto = 500
- Linf=200
- sig.Linf=5
- k=0.4
- t0= -0.5
- sig.err=5



### Simulate tag-recapture data

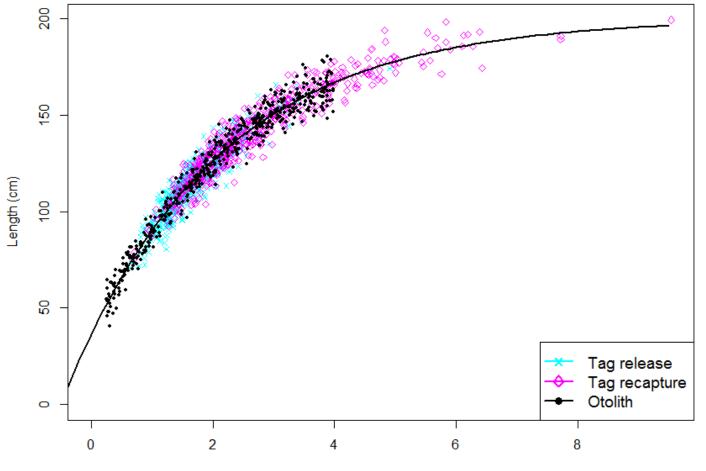
- TAL = rgamma(n.tag, 1, 1) # time at liberty
- a1 = rlnorm(n.tag, 0.5, 0.3) + a0 # release age
- $a^2 = a^1 + TAL$  # recapture age
- Linf.tag = rnorm(n.tag, Linf, sig.Linf) # fish-specific Linf
- L1 = Linf.tag\*f(a1-t0,k) + rnorm(n.tag, 0, sig.err) # release length
- L2 = Linf.tag\*f(a2-t0,k) + rnorm(n.tag, 0, sig.err) # recap length

# Simulate otolith data

- a.oto = runif(n.oto, 0.25, 4)
- Linf.oto = rnorm(n.oto, Linf, sig.Linf)
- L.oto = Linf.oto\*f(a.oto-t0,k) + rnorm(n.oto, 0, sig.err)



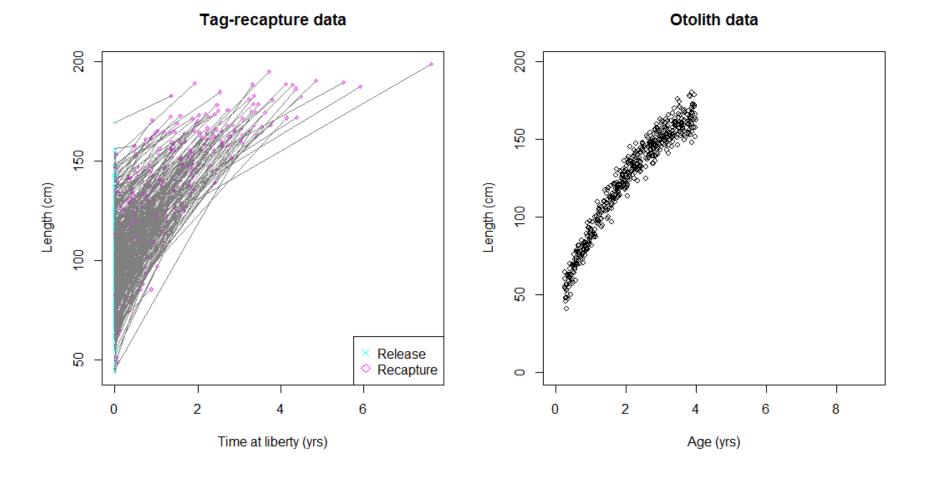
#### Simulated data sets



Age (yrs)



#### Simulated data sets

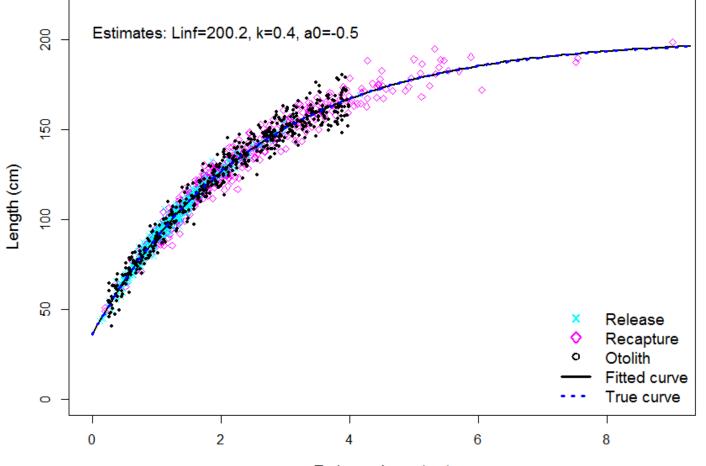


# Assessing model fit for tag-recapture data

- To calculate fitted release and recapture lengths requires an estimated value of A for each fish
- Obvious approach is to use mean of posterior distribution for A conditioned on a fish's release and recapture lengths
- However, this approach yields biased estimates of A
- Thus, we use alternative approximately conditionally unbiased estimator proposed in Laslett et al. (2004)



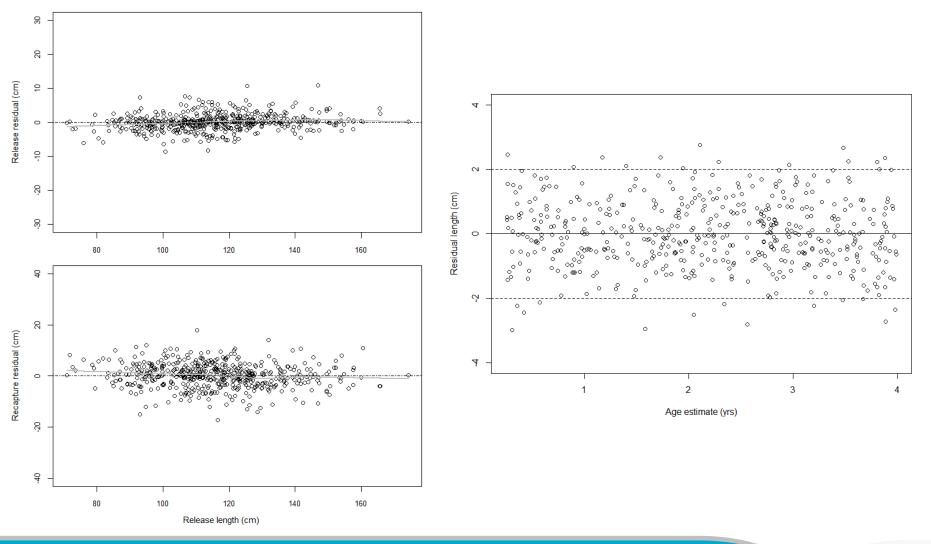
#### Integrated model results: fitted curve



Estimated age (yrs)

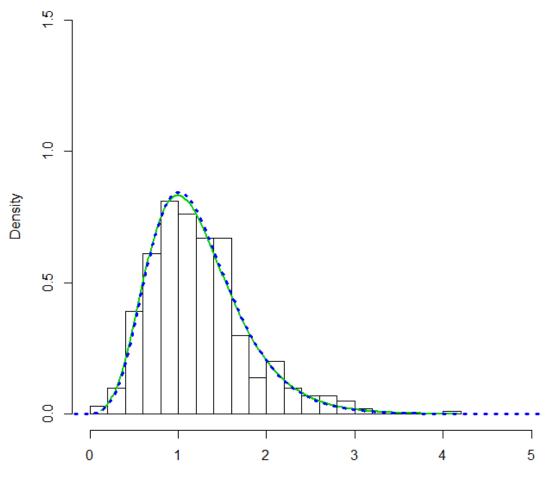


#### **Integrated model results: residuals**





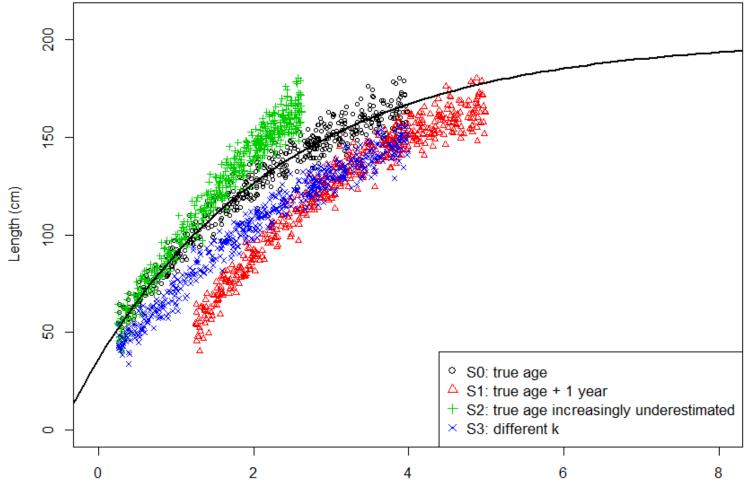
#### **Estimated release age distribution**



Estimated release age (yrs)

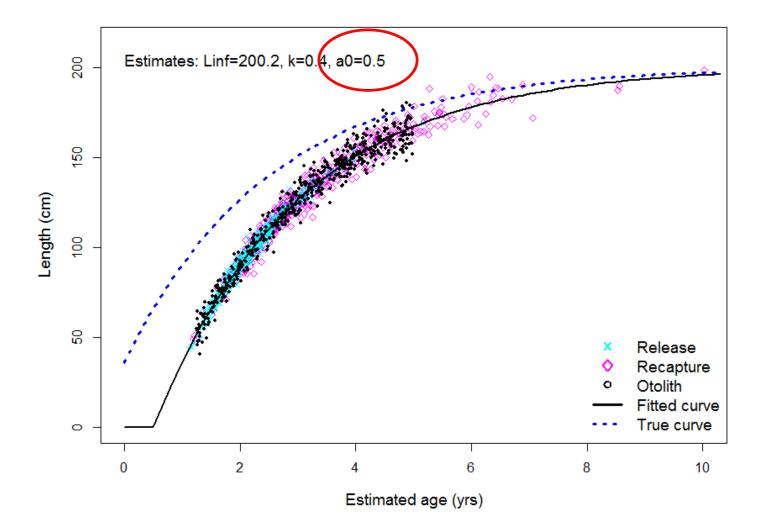


#### Alternative scenarios for otolith data



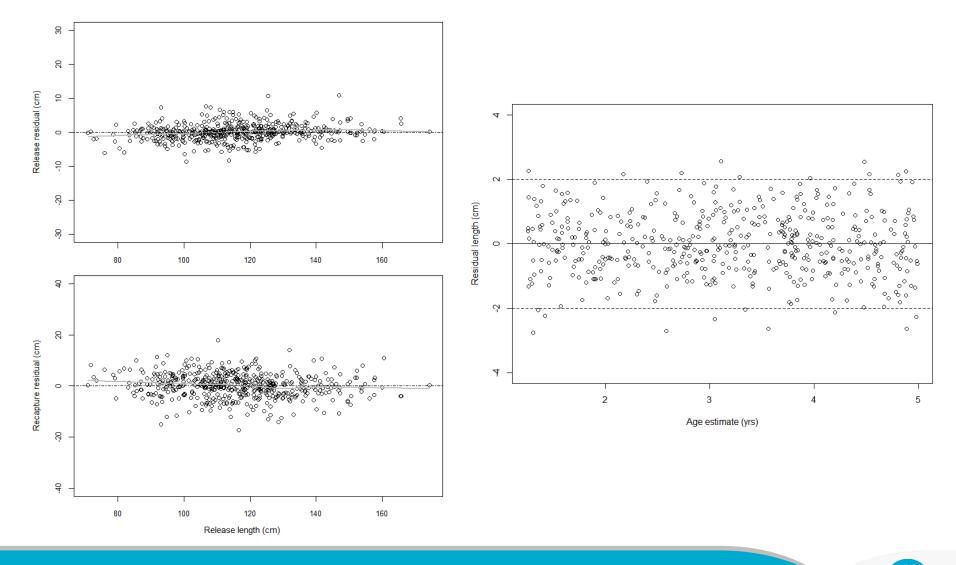
Age (yrs)

#### **S1 results:** true otolith age +1

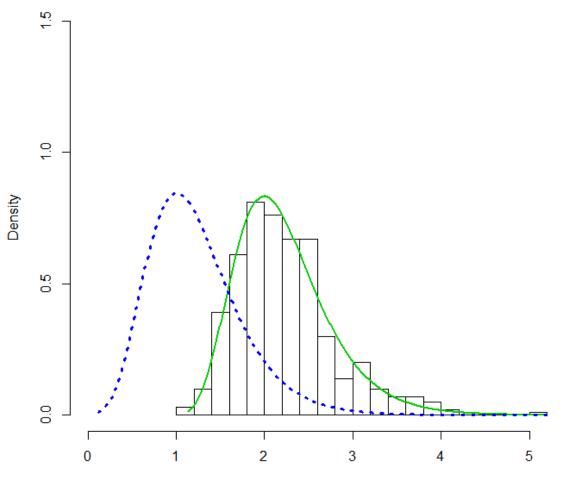




### S1 results: true otolith age +1



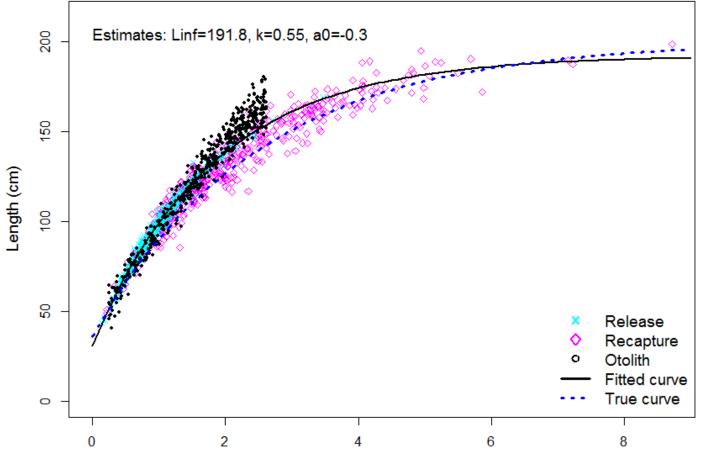
#### **S1 results:** true otolith age +1



Estimated release age (yrs)

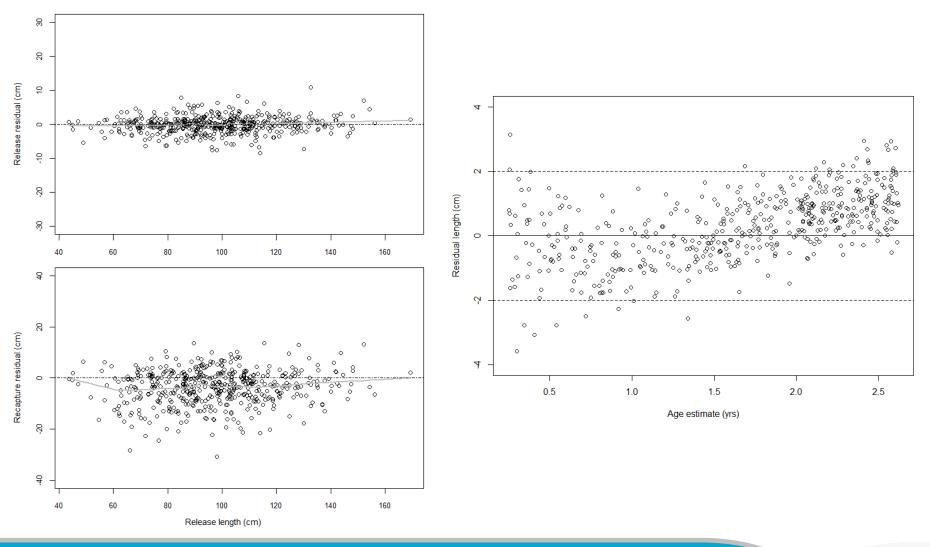


#### S2 results: true age increasingly underestimated



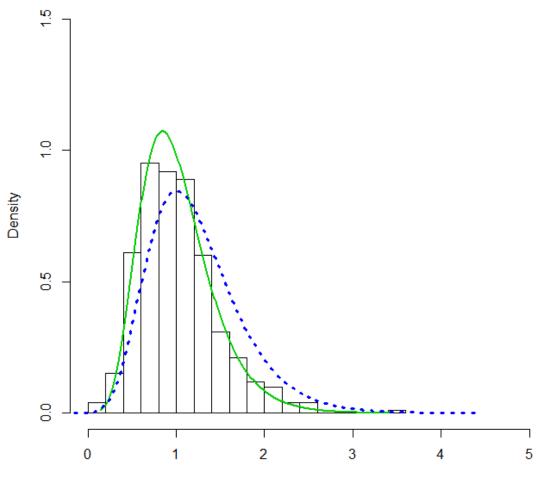
Estimated age (yrs)

#### S2 results: true age increasingly underestimated





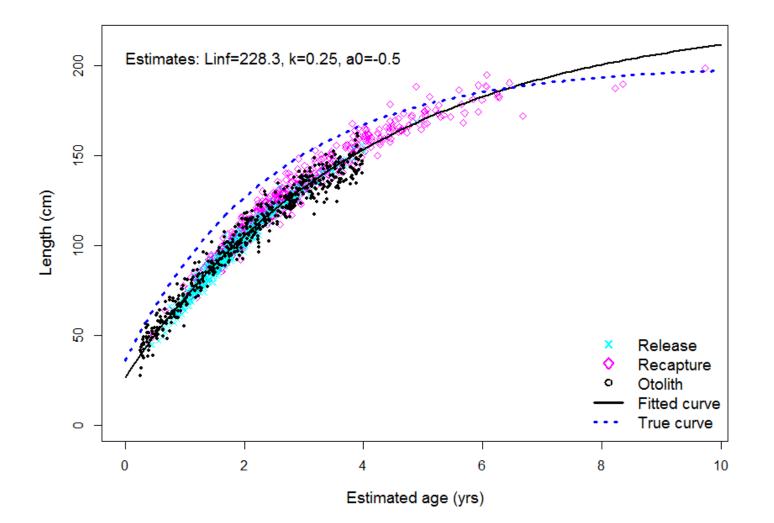
#### S2 results: true age increasingly underestimated



Estimated release age (yrs)

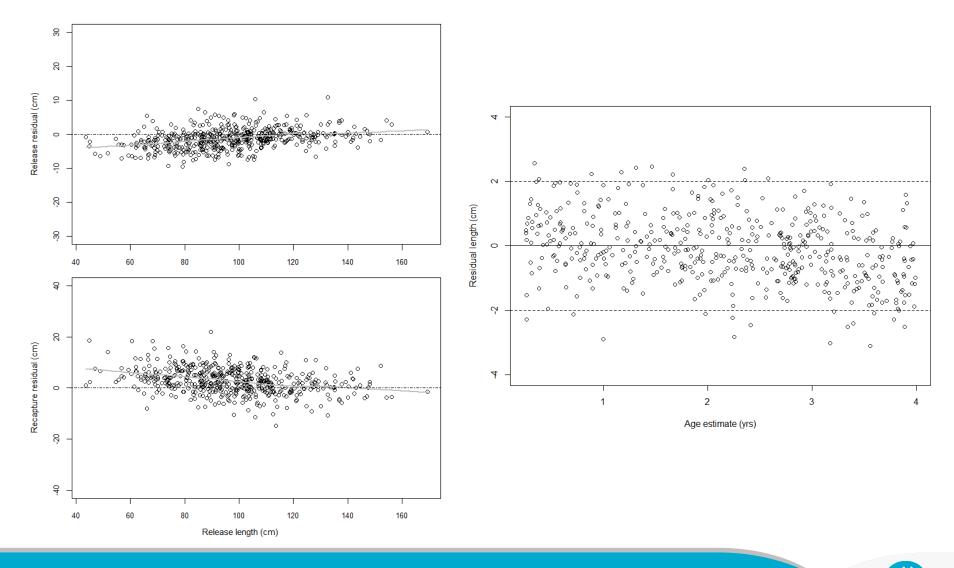
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#### S3 results: different k for otolith data

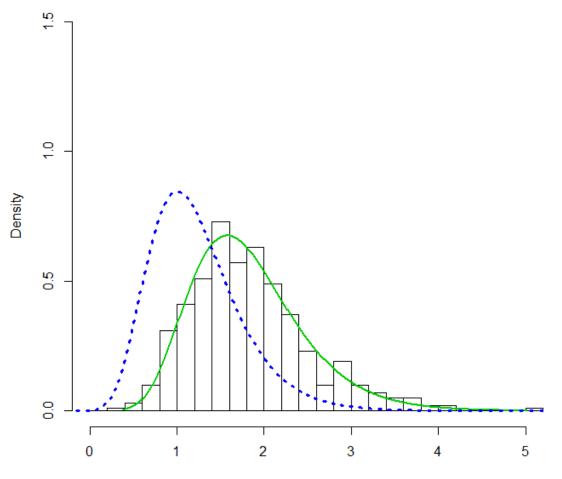




#### S3 results: different k for otolith data



#### S3 results: different k for otolith data



Estimated release age (yrs)



# Discussion

- Can be many benefits to having multiple data sources, but only if they are unbiased
- Might expect residuals to reveal problems, but not always obvious when tag-recapture data are involved (due to ability of model to manipulate release age estimates)
- Ability to determine biases/incompatibility in tag-recapture and otolith data sets will depend on:
  - nature of the biases (e.g. not possible to detect with Scenario 1)
  - sample sizes and age/length ranges of data sets