

# Monte Carlo Simulation of Selectivity and Maturity at Age in a Length- Based Age-Structured Model

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# Objectives

**Investigate the effects of simulating uncertainty in length at age in an age-structured model**

1. Verify expected outcome of including uncertainty  
(parametric bootstrapping)
2. Discuss remaining questions for simulation methods  
(Rescale  $sel_a$  and  $mat_a$  to max of 1?)

# Monte Carlo simulation

## **Completed Work:**

**Step 1) Model numbers at age with simulated uncertainty in recruitment**

**Step 2) Model  $sel_L$  and  $mat_L$  and simulate uncertainty in length at age**

**Step 3) Transform simulated uncertainty in length at age into simulated uncertainty in  $sel_a$  and  $mat_a$**

## Ongoing Work:

Step 4) Project numbers at age ahead with fixed (assumed) parameter values for most life history and simulated uncertainty in recruitment and length at age

Step 5) Summarize the distributions of derived variables for stock status

Step 6) Repeat – evaluate model sensitivity to assumed fixed parameter values

Step 7) Risk analysis – summarize results over a range of “what if scenarios”

# Numbers at age

$$N_{a,t+1} = \begin{cases} N_{r,t+1}; & a = r \\ N_{a-1,t} e^{(-M_{a-1} - sel_{a-1} F_t)}; & a > r \text{ to } A-1 \\ (N_{A-1,t} + N_{A,t}) e^{(-M_A - sel_A F_t)}; & a = A \end{cases} \quad (\text{Quinn and Deriso, 1999})$$

$$C_{a,t} = \frac{sel_a F_t}{M_a + sel_a F_t} N_{a,t} \left(1 - e^{(-M_a - sel_a F_t)}\right) \quad (\text{Quinn and Deriso, 1999; e.g., Wetzel and Punt, 2011a})$$

$$\tilde{N}_{a=r,t} = \frac{\alpha E_t}{1 + \beta E_t} e^{(-0.5\sigma_R^2 + \varepsilon_t)}; \quad \varepsilon_t \sim N(0; \sigma_R^2) \quad (\text{Quinn and Deriso, 1999; Brooks and Powers, 2007; Brooks et al., 2010})$$

$$E_t = \sum_{a=r}^A m_a f_a \rho_{a,t} N_{a,t} \quad (\text{Punt and Walker 1998; cf, Simpfendorfer et al. 2000})$$

# Length-based $sel_L$ and $mat_L$

$$sel_L = \left( \frac{1}{1-\gamma} \right) \left( \frac{1-\gamma}{\gamma} \right)^\gamma \left( \frac{e^{\alpha\gamma(\beta-L)}}{1+e^{\alpha(\beta-L)}} \right) \quad (\text{Thompson, 1994; Sigler, 1999})$$

$$m_L = \frac{m_\infty}{1+e^{-k(L-\gamma)}} \quad (\text{Quinn and Deriso, 1999})$$

$$\tilde{L}_a = L_\infty \left( 1 - e^{-k(a-t_0)} \right) + \varepsilon_{L_a}; \quad \varepsilon_{L_a} \sim N(0; \sigma_{L_a}^2) \quad (\text{Quinn and Deriso, 1999})$$

where

$L_\infty$  = asymptotic (maximum) length,

$k$  = the growth rate parameter,

$t_0$  = the age when an individual would have been at  $L = 0$ ,

$\varepsilon_{L_a} \sim N(\mu_{L_a}; \hat{\sigma}_{L_a}^2)$ , where  $\mu_{L_a}$  was the predicted LVB length at each age  $a$  without uncertainty,

$\hat{\sigma}_{L_a}^2$  was the assumed variance in length at each age  $a$  required to achieve a constant CV.

# Age-length transition matrix

$$\phi_{a,l} = \begin{cases} \Phi\left(\frac{L'_{\min} - \tilde{L}_a}{\sigma_{L_a}}\right); & l = 1 \\ \Phi\left(\frac{L'_{l+1} - \tilde{L}_a}{\sigma_{L_a}}\right) - \Phi\left(\frac{L'_l - \tilde{L}_a}{\sigma_{L_a}}\right); & l > 1 \text{ to } l < A_l \\ 1 - \Phi\left(\frac{L'_{\max} - \tilde{L}_a}{\sigma_{L_a}}\right); & l = A_l \end{cases}$$

(Wetzel and Punt, 2011; Methot and Wetzel, in Press)

$$\left. \begin{aligned} sel_a &= \sum_{l=1}^{A_l} \phi_{a,l} sel_l \\ m_a &= \sum_{l=1}^{A_l} \phi_{a,l} m_l \end{aligned} \right\}$$

? Citations (difficult to interpret)

1. Verify expected outcome (parametric bootstrapping)

2. Verify methods (rescale to max of 1?)

# Verify expected outcome (parametric bootstrap)

- *Draw a random length at each age (1-50)*

- a)  $\tilde{L}_a = L_\infty(1 - e^{-k(a-t_0)}) + \varepsilon_{L_a}; \quad \varepsilon_{L_a} \sim N(0; \sigma_{L_a}^2), a = (r, r+1, \dots, A)$

- b)  $\tilde{L}_a = L_\infty(1 - e^{-k(a-t_0)})e^{\varepsilon_{L_a}} +; \quad \varepsilon_{L_a} \sim N(0; \sigma_{L_a}^2), a = (r, r+1, \dots, A)$

- *Determine  $sel_L$  and  $mat_L$  at simulated length*

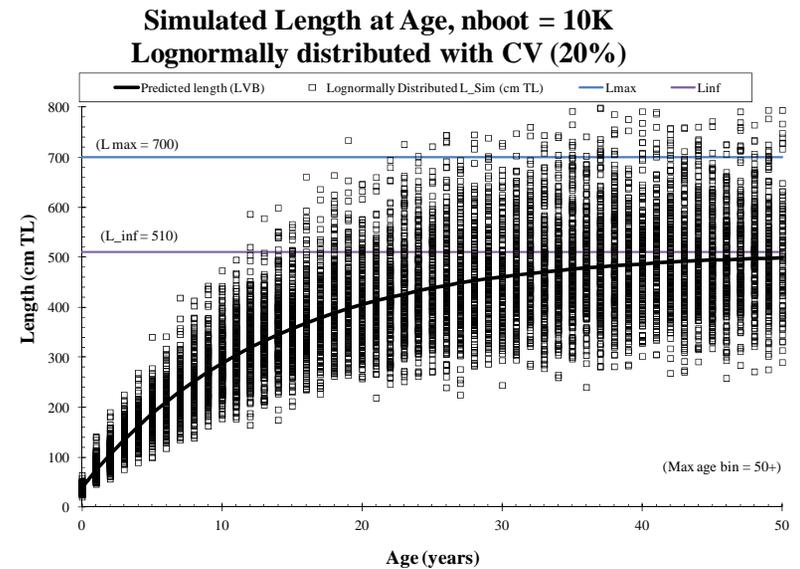
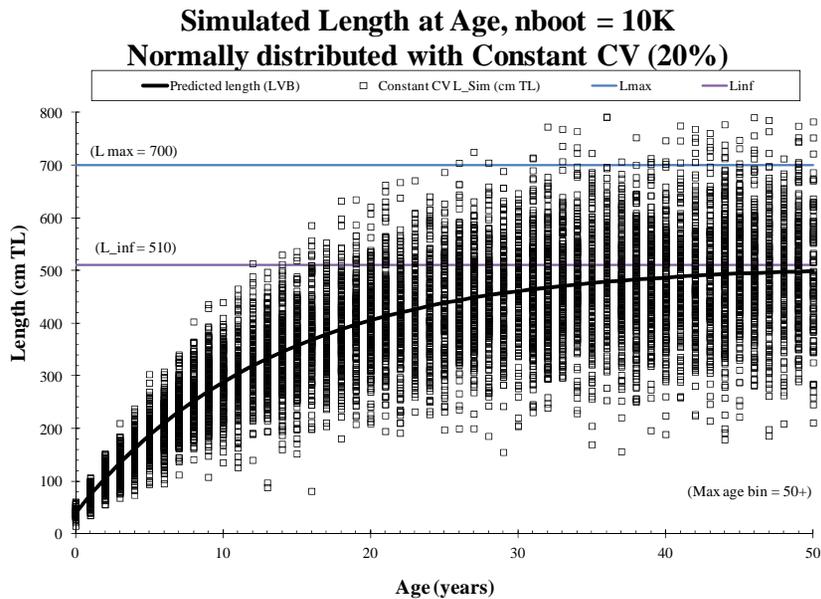
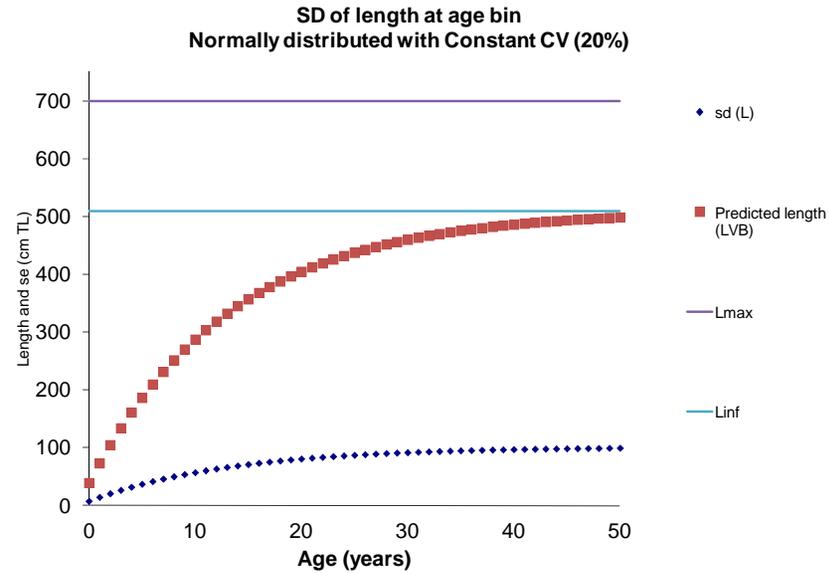
- $sel_{\tilde{L}_a} = \left( \frac{1}{1-\gamma} \right) \left( \frac{1-\gamma}{\gamma} \right)^\gamma \left( \frac{e^{\alpha\gamma(\beta-\tilde{L}_a)}}{1+e^{\alpha(\beta-\tilde{L}_a)}} \right)$
  - $m_{\tilde{L}_a} = \frac{m_\infty}{1+e^{-k(\tilde{L}_a-\gamma)}}$

- *Assign  $sel_L$  and  $mat_L$  to each age (1-50)*

- $sel_{\tilde{a}} = sel_{\tilde{L}_a}$
  - $m_{\tilde{a}} = m_{\tilde{L}_a}$

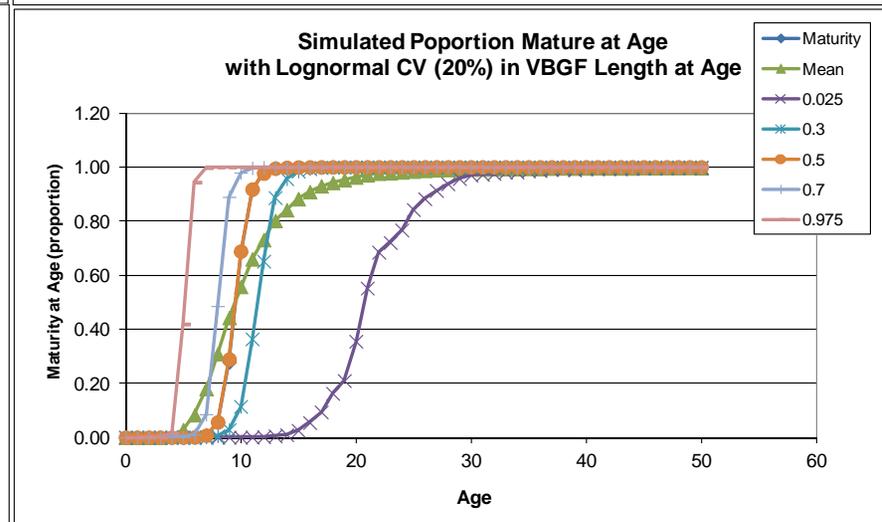
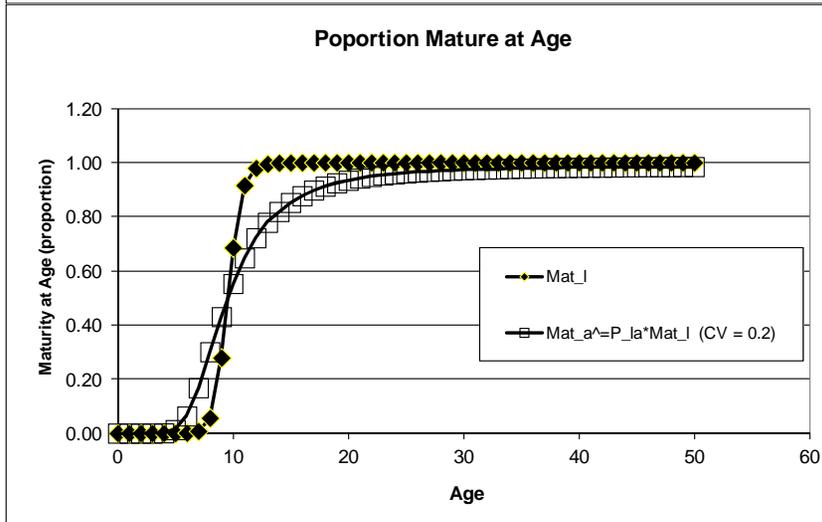
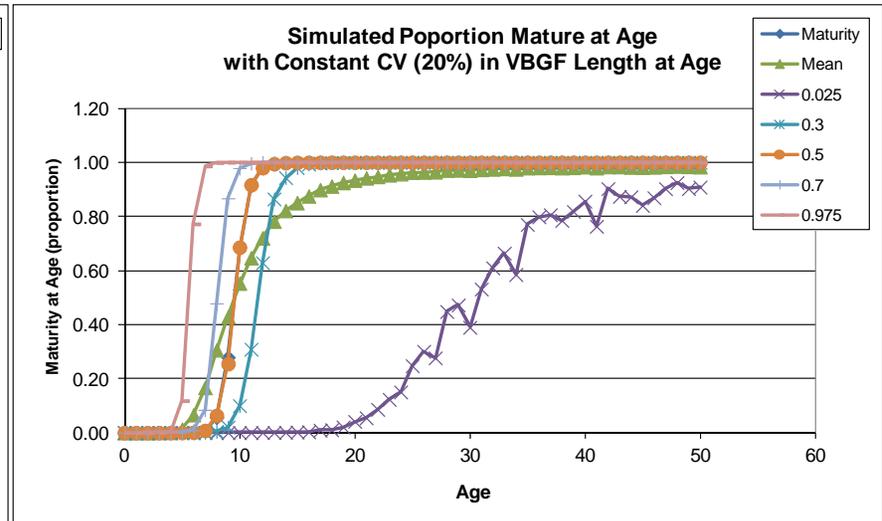
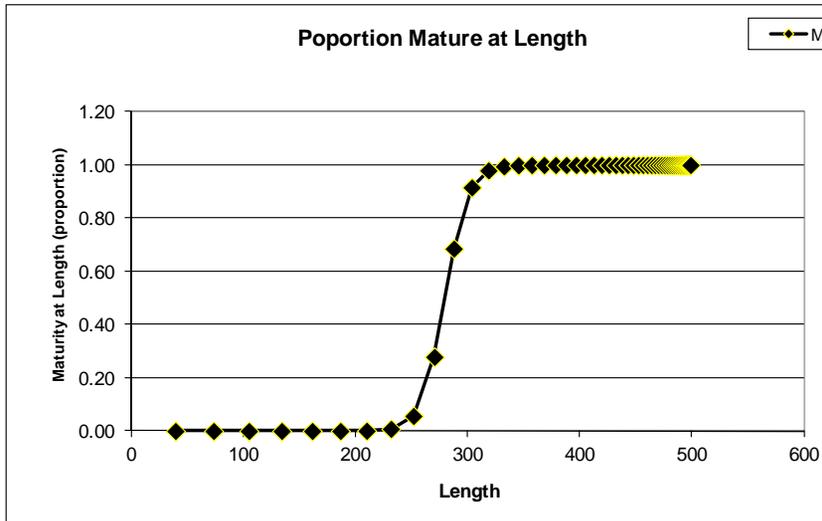
- *Repeat ( $n=10,000$ ) and summarize*

- Draw a random length at each age (1-50)



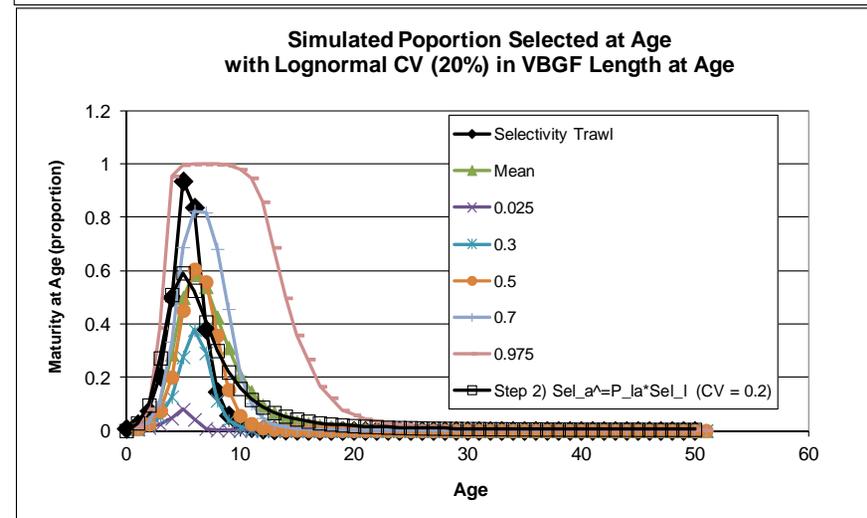
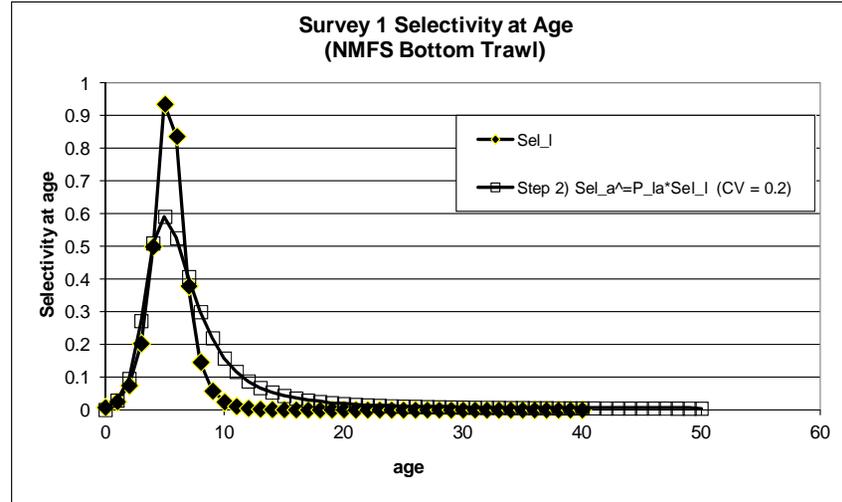
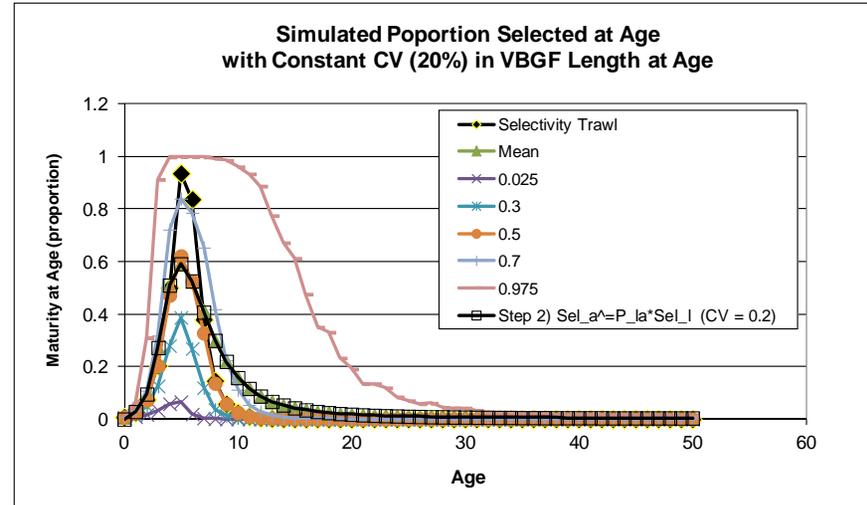
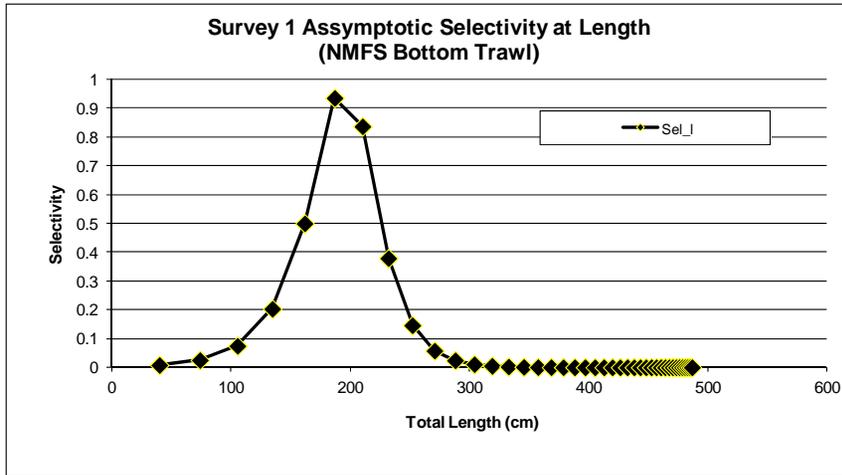
□ Matrix  $m_a = \sum_{l=1}^{A_l} \phi_{a,l} m_l$

□ Bootstrap  $m_{\tilde{a}} = m_{\tilde{L}_a}$



□ *Matrix*  $sel_a = \sum_{l=1}^{A_l} \phi_{a,l} sel_l$

□ *Bootstrap*  $sel_{\tilde{a}} = sel_{\tilde{L}_a}$



# Conclusions

- **Mean of bootstrap approximated results from transition matrix**

- $$\text{mean}(sel_{\tilde{L}_a}) \cong sel_a = \sum_{l=1}^{A_l} \phi_{a,l} sel_l$$

- **Median of bootstrap approximated results from back-transformed von Bertalanffy growth curve**

- $$\text{median}(sel_{\tilde{L}_a}) \cong sel_a = sel_{L_a},$$

- where 
$$L_a = L_\infty (1 - e^{-k(a-t_0)})$$

- **To do**

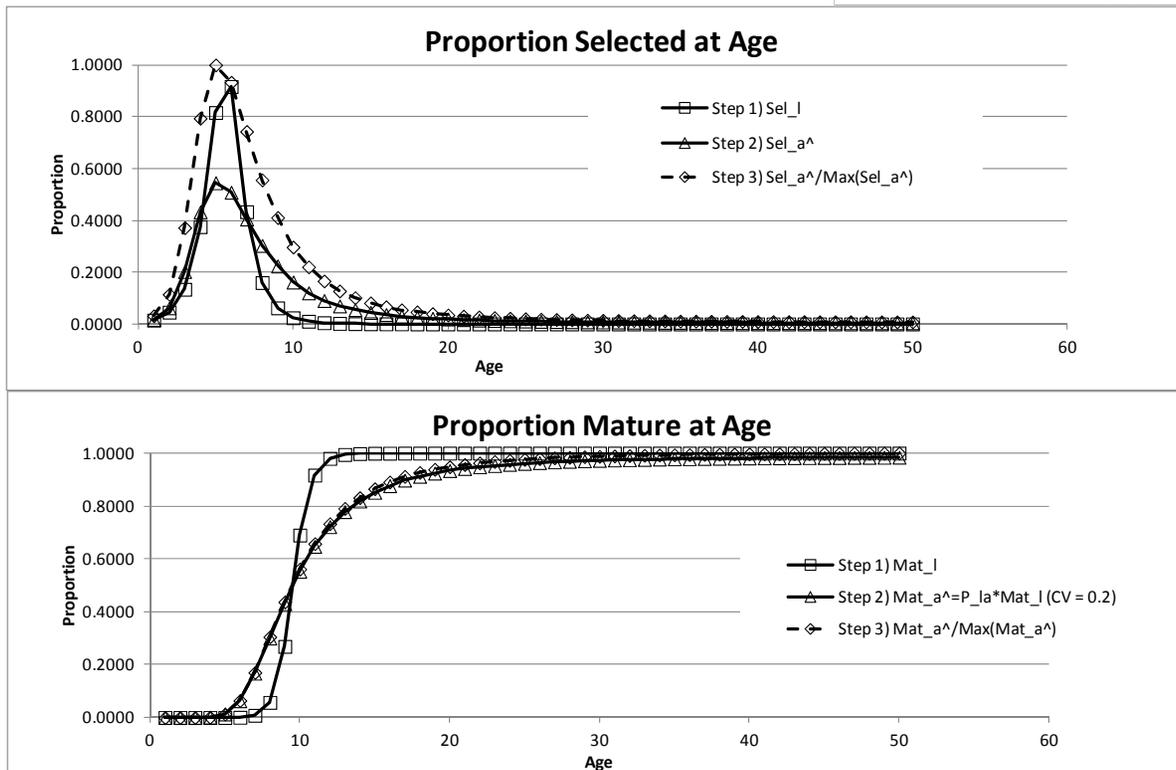
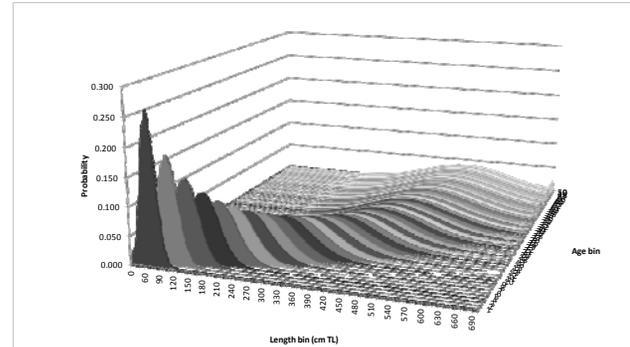
- Rescale to 1? 
$$sel_a = \sum_{l=1}^{A_l} \phi_{a,l} sel_l$$

- $$sel_a / \max(sel_a)?$$



# Rescale to 1?

- Rescale  $sel_a$  and  $mat_a$  to max of 1?
- Literature not clear
- Rules of thumb?



$$sel_a = \sum_{l=1}^{A_l} \phi_{a,l} sel_l$$

$$sel_a / \max(sel_a)?$$

$$m_a = \sum_{l=1}^{A_l} \phi_{a,l} m_l$$

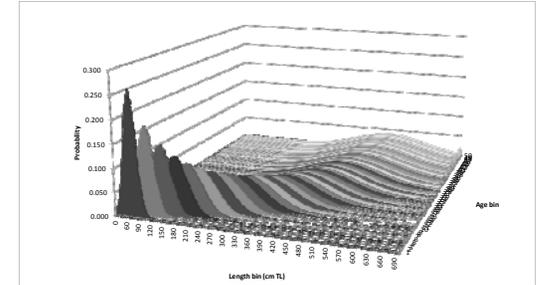
$$m_a / \max(m_a)?$$

# Example: Proportions at length

$$p_{\ell_i, a_j} = \begin{cases} \int_{L_{Max}}^{+\infty} \frac{1}{\sqrt{2\pi}\hat{\sigma}_j} e^{-\left[\frac{(L_i - \bar{L}_j)^2}{2(\hat{\sigma}_j)^2}\right]} dL & \text{For } L_i = L_{Max} \\ \int_{L_i}^{L_i + \Delta_i} \frac{1}{\sqrt{2\pi}\hat{\sigma}_j} e^{-\left[\frac{(L_i - \bar{L}_j)^2}{2(\hat{\sigma}_j)^2}\right]} dL & \text{For } L_{Min} < L_i < L_{Max} \\ \int_{-\infty}^{L_{Min}} \frac{1}{\sqrt{2\pi}\hat{\sigma}_j} e^{-\left[\frac{(L_i - \bar{L}_j)^2}{2(\hat{\sigma}_j)^2}\right]} dL & \text{For } L_i = L_{Min} \end{cases}$$

e.g., Methot, 1990, and 2000,  
e.g., their stage 1 model case  
4 transition matrix; e.g.,  
Haddon, 2011, their length-  
to-length transition matrix  
for a stage based model

$$\begin{bmatrix} \hat{n}_{\ell_1} & \hat{n}_{\ell_2} & \cdots & \hat{n}_{\ell_m} \end{bmatrix} = \begin{bmatrix} n_{a_1} & n_{a_2} & \cdots & n_{a_n} \end{bmatrix} \bullet \begin{bmatrix} p_{\ell_1, a_1} & p_{\ell_2, a_1} & \cdots & p_{\ell_m, a_1} \\ p_{\ell_1, a_2} & \vdots & & \\ \vdots & & & \\ p_{\ell_1, a_n} & & & p_{\ell_m, a_n} \end{bmatrix}$$



where each element  $\hat{n}_{\ell_i}, i = 1, 2, 3, \dots, m$  of  $N_{\ell}$  is calculated as

$$\hat{n}_{\ell_i} = n_{a_1} \times p_{\ell_i, a_1} + n_{a_2} \times p_{\ell_i, a_2} + \dots + n_{a_n} \times p_{\ell_i, a_n}$$

$$\hat{n}_{\ell_i} = \sum_{j=1}^n n_{a_j} \times p_{\ell_i, a_j}$$

Rescale as a proportion by dividing by the sum (~pdf) [standard is to rescale before multiplying by matrix]

# Example: Proportions at age

$$\hat{Sel}_a = P_{\ell,a} \bullet Sel_l$$

(e.g., Methot, 1990, 2000; their stage 2 model – size specific availability; Coleraine Users Manual e.g., their page 8),

where

$P_{\ell,a}$  is the  $(n \times m)$  length-at-age transition matrix (equation 2),

$Sel_l$  is a  $(1 \times m)$  vector of the selectivity at length by bin  $\ell_i, i=1,2,3,\dots,m$ , and

$\hat{Sel}_a$  is a  $(1 \times n)$  vector of the simulated selectivity at age by bin  $a_j, j=1,2,3,\dots,n$ .

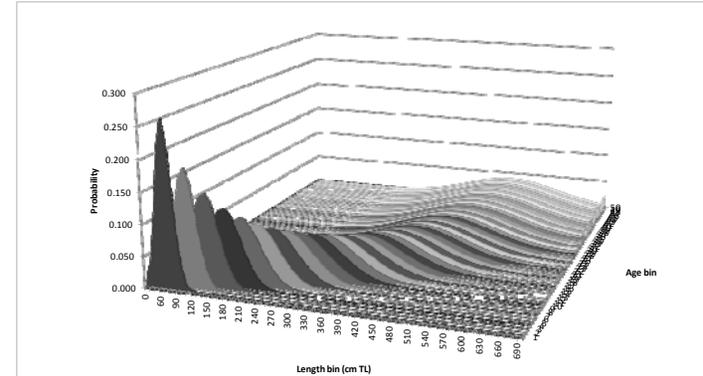
An example of the  $(n \times 1) = (n \times m)(m \times 1)$  matrix multiplication is

$$\begin{bmatrix} \hat{sel}_{a_1} \\ \hat{sel}_{a_2} \\ \vdots \\ \hat{sel}_{a_n} \end{bmatrix} = \begin{bmatrix} p_{\ell_1,a_1} & p_{\ell_2,a_1} & \dots & p_{\ell_m,a_1} \\ p_{\ell_1,a_2} & & & \\ \vdots & & & \\ p_{\ell_1,a_n} & & & p_{\ell_m,a_n} \end{bmatrix} \bullet \begin{bmatrix} sel_{\ell_1} \\ sel_{\ell_2} \\ \vdots \\ sel_{\ell_m} \end{bmatrix}$$

where each element  $\hat{sel}_{a_j}, j=1,2,3,\dots,n$  of  $\hat{Sel}_a$  is calculated as

$$\hat{sel}_{a_j} = p_{\ell_1,a_j} \times sel_{\ell_1} + p_{\ell_2,a_j} \times sel_{\ell_2} + \dots + p_{\ell_m,a_j} \times sel_{\ell_m}$$

$$\hat{sel}_{a_j} = \sum_{i=1}^m p_{\ell_i,a_j} \times sel_{\ell_i}$$

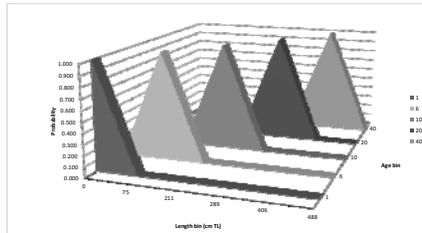


# Ex. Proportions at age (CV = 0.001)

- 5 age bins and 6 length bins low (CV = 0.0001)
- $a = (1, 6, 10, 20, 40)$  (yrs)
- $L = (74+1, 210+1, 288+1, 405+1, 487+1)$  (cm TL)

Example 1. Five age bins with low uncertainty:  $a_j = (1, 6, 10, 20, 40)$ ,  $\ell_j = (74+1, 210+1, 288+1, 405+1, 487+1)$  cm TL, CV = 0.0001.

Assumed true age (a)	Rounded to nearest integer		Upper bound (e.g., <75)						sum	
	LVB	sd(L)	age	Lower bound (e.g., >=0)	75	211	289	406		488
1	74	0.0074	1	1.000	0.000	0.000	0.000	0.000	0.000	1.000
6	210	0.0210	6	0.000	1.000	0.000	0.000	0.000	0.000	1.000
10	288	0.0288	10	0.000	0.000	1.000	0.000	0.000	0.000	1.000
20	405	0.0405	20	0.000	0.000	0.000	1.000	0.000	0.000	1.000
40	487	0.0487	40	0.000	0.000	0.000	0.000	1.000	0.000	1.000

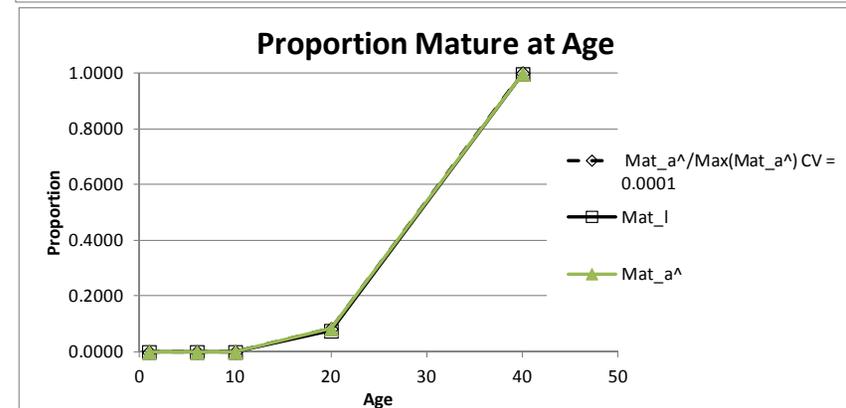
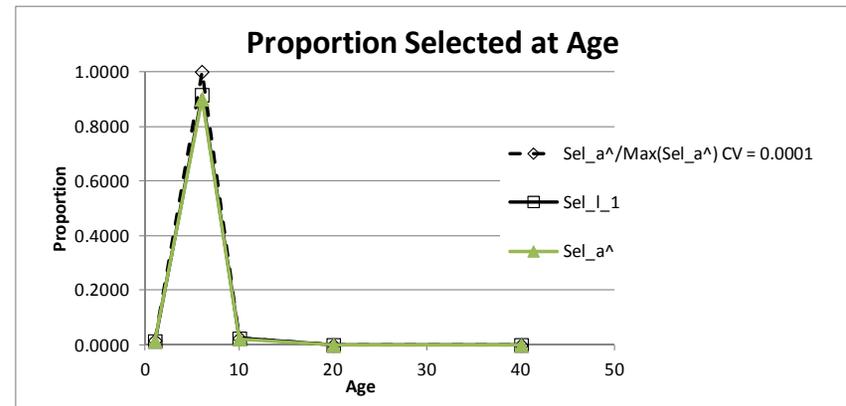


a	Rescaled to Max 1					
	Pa=PSL	LVB	PSL	sd(L)	Check CV	PL^a/P_la^a*Pa
1	0.0068	73.6003	0.0068	0.0074	0.0001	0.0070
6	0.9689	210.0671	0.9689	0.0210	0.0001	1.0000
10	0.0238	287.8043	0.0238	0.0288	0.0001	0.0246
20	0.0000	405.0422	0.0000	0.0405	0.0001	0.0000
40	0.0000	486.5807	0.0000	0.0487	0.0001	0.0000

Grand Total						
Sum	1.000				Sum	1.032
Max	0.969				Max	1.000
Rescaled to Max 1						
a	Pa=PML	LVB	PML	sd(L)	Check CV	PL^a/P_la^a*Pa
1	0.0000	73.60027	0.0000	0.00736	0.0001	0.0000
6	0.0000	210.0671	0.0000	0.021007	0.0001	0.0000
10	0.0000	287.8043	0.0000	0.02878	0.0001	0.0000
20	0.0762	405.0422	0.0762	0.040504	0.0001	0.0764
40	0.9965	486.5807	0.9965	0.048658	0.0001	1.0000

Grand Total		
Sum	1.073	Sum
Max	0.997	Max
		1.076
		1.000

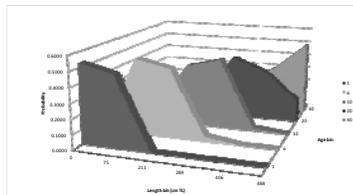


# Ex. Proportions at age (CV = 0.2)

- 5 age bins and 6 length bins low (CV = 0.2)
- $a = (1, 6, 10, 20, 40)$  (yrs)
- $L = (74+1, 210+1, 288+1, 405+1, 487+1)$  (cm TL)

Example 2. Five age bins with moderate uncertainty:  $a_j = (1, 6, 10, 20, 40)$ ,  $\ell_j = (74+1, 210+1, 288+1, 405+1, 487+1)$  cm TL, CV = 0.2.

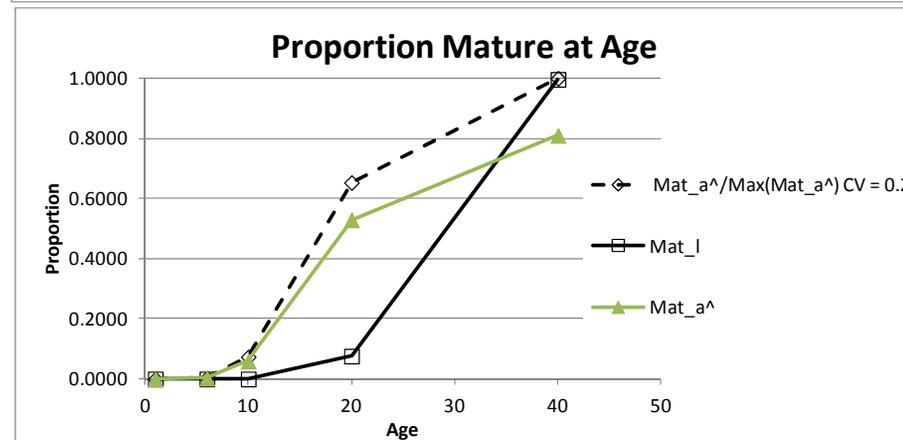
Assumed true age (a)	Round to nearest integer LVB	sd(L)	Check CV	age	Upper bound (e.g., <75)						sum
					Lower bound (e.g., >=0)	75	211	289	406	488	
1	74	14.7201	0.1989	1	0.5271	0.4729	0.0000	0.0000	0.0000	0.0000	1.000
6	210	42.0134	0.2001	6	0.0007	0.5088	0.4605	0.0300	0.0000	0.0000	1.000
10	288	57.5609	0.1999	10	0.0001	0.0904	0.4164	0.4729	0.0199	0.0003	1.000
20	405	81.0084	0.2000	20	0.0000	0.0083	0.0678	0.4288	0.3423	0.1528	1.000
40	487	97.3161	0.1998	40	0.0000	0.0023	0.0187	0.1817	0.3015	0.4959	1.000



Rescaled to Max 1							
a	Pa=PSL	LVB	PSL	sd(L)	Check CV	PL^=P_la*Pa	PL^/Max(PL^)
1	0.0068	73.6003	0.0068	14.7201	0.2000	0.0042	0.0084
6	0.9689	210.0671	0.9689	42.0134	0.2000	0.4984	1.0000
10	0.0238	287.8043	0.0238	57.5609	0.2000	0.4561	0.9152
20	0.0000	405.0422	0.0000	81.0084	0.2000	0.0404	0.0810
40	0.0000	486.5807	0.0000	97.3161	0.2000	0.0005	0.0010
Grand Total	Sum	1.000			Sum	1.000	2.006
	Max	0.969			Max	0.498	1.000

Rescaled to Max 1							
a	Pa=PML	LVB	PML	sd(L)	Check CV	PL^=P_la*Pa	PL^/Max(PL^)
1	0.0000	73.6003	0.0000	14.7201	0.2000	0.0000	0.0000
6	0.0000	210.0671	0.0000	42.0134	0.2000	0.0029	0.0089
10	0.0000	287.8043	0.0000	57.5609	0.2000	0.0238	0.0728
20	0.0762	405.0422	0.0762	81.0084	0.2000	0.2137	0.6545
40	0.9965	486.5807	0.9965	97.3161	0.2000	0.3265	1.0000
Grand Total	Sum	1.073			Sum	0.567	1.736
	Max	0.997			Max	0.327	1.000



# Questions?



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