



Overview of Artificial Cohort Method for Estimating Zooplankton Production in the Ocean

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PICES 2018 Annual Meeting, October 25, 2018, Yokohama, Japan

Outline

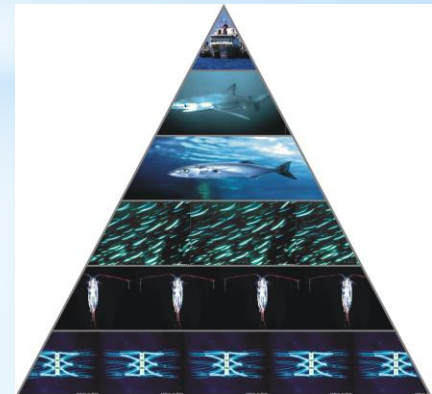
- Classical methods for measuring zooplankton production
- Artificial Cohort (AC) methods and Caveats
- Utility and practicality of AC methods
- Status of AC research in the USA

Zooplankton Production

- Limited spatial-temporal resolution of zooplankton production
- A bottleneck for estimating zooplankton production

$$P_{2nd} = \sum_{i=1}^n G_i \times B_i$$

- Lack of consensus on practical methods for measuring zooplankton growth (Hirst et al. 2005, Kimmerer et al. 2007, Liu et al. 2013)

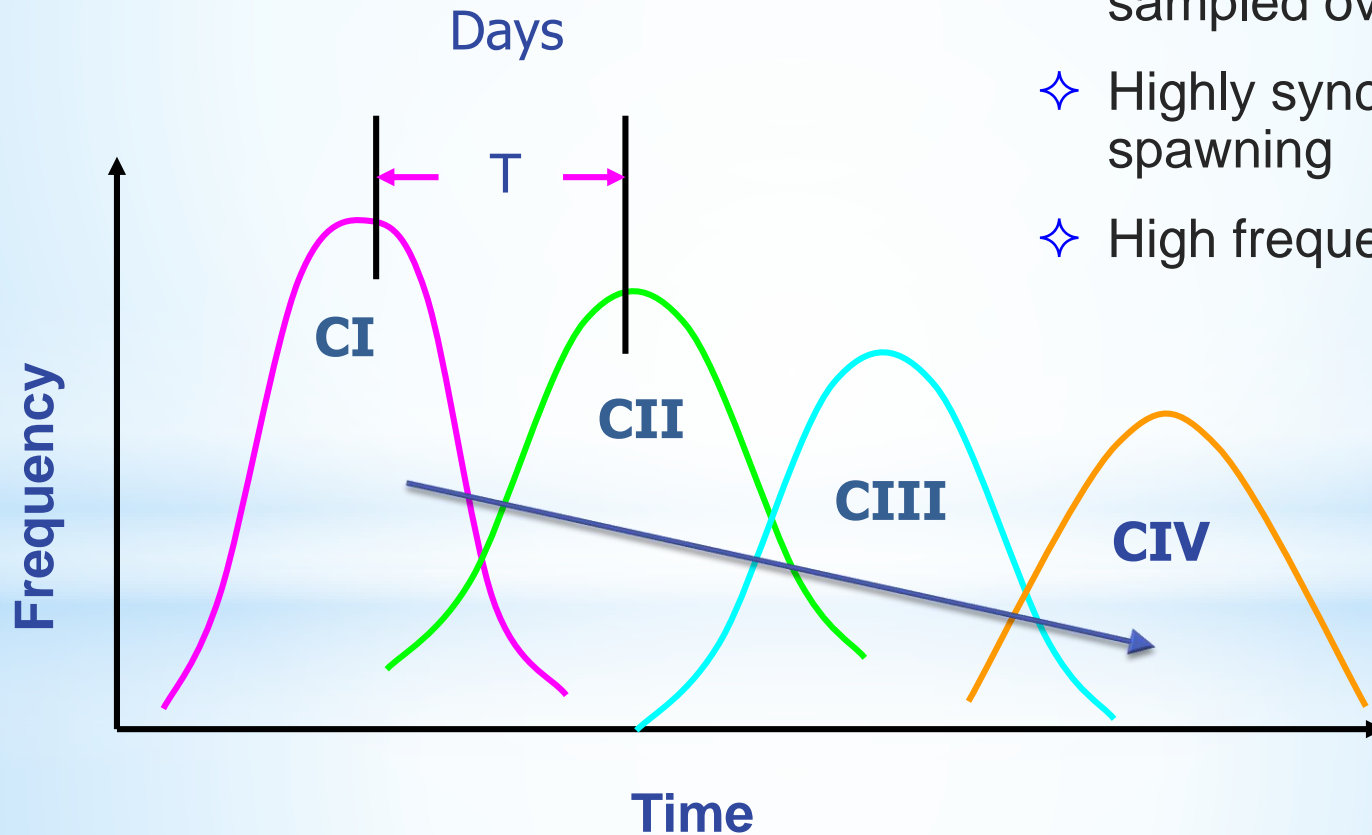


Methodological Challenge

Natural cohort method

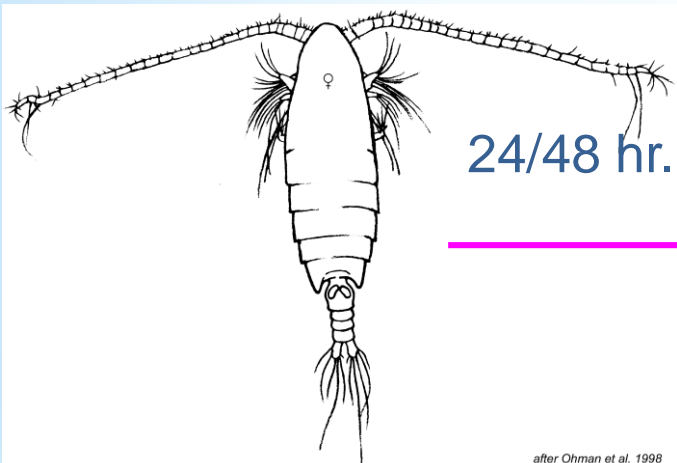
Assumptions:

- ✧ Same population to be sampled over time
- ✧ Highly synchronized spawning
- ✧ High frequency sampling



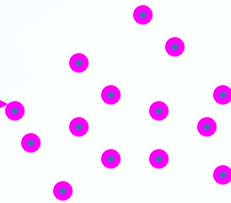
Egg production rate

Female



24/48 hr.

Eggs



Specific Egg
Production (SEP)
(% of female
weight produced
as eggs per day)

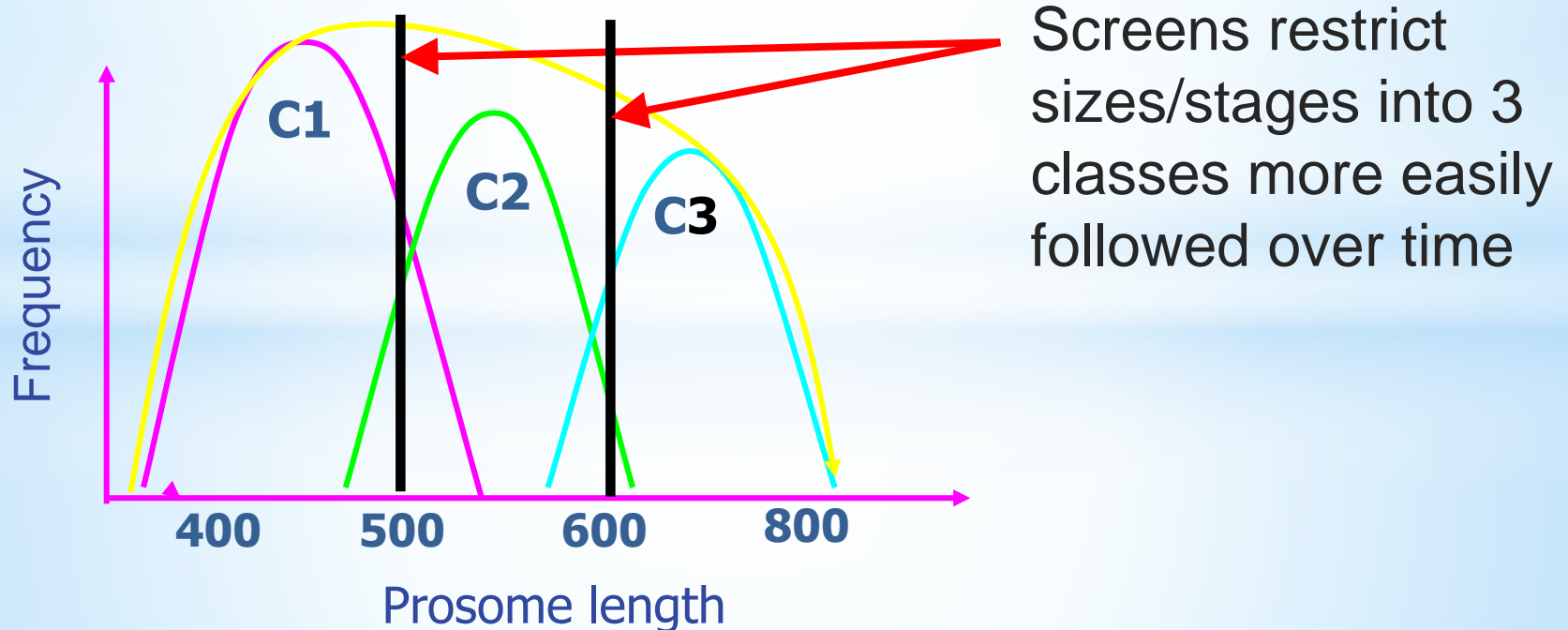
Courtesy of Ohman et al. 1998

Assumptions

- SEP not influenced by incubation
- Somatic growth of all preceding stages is equivalent to female SEP
- Female remains in steady-state

Artificial Cohort Method (Kimmerer & McKinnon 1987)

- uses sieves of varying mesh sizes to separate the copepod community into different size classes (size-fractionated cohorts)
- also can be done by manually picking specific stages (cleaned and sorted cohorts)



Caveats of AC methods

Assumptions

- Close coupling between moulting and growth (Hart 1990, Peterson et al. 1991)
- Handing does not stimulate moulting, “moulting burst” (Miller et al. 1991)

Limitations

- No control over number and kinds of species
- Differential growth among stages and species
- Different developmental stage isolated for different species. Incubation times to be appropriate for species and stages of interest

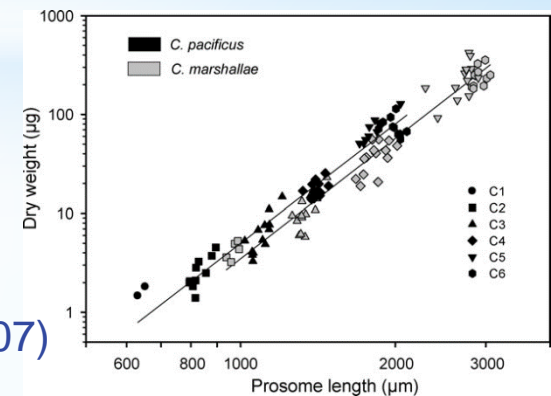
Methodological considerations of AC

➤ Age structure within stages

Ideally uniform age structure within stages assumed (Hirst et al. 2005). In practice growth often labeled by stage, not age

➤ Biomass increment of growth

Direct measurement of biomass (Kimmerer et al. 2007), or using a length-weight relationship (Liu & Hopcroft 2006a&b, 2007&08)



Liu & Hopcroft (2007)

Applications of AC

Table 1. Applications of the Kimmerer & McKinnon (1987) artificial cohort method. Temp.: ambient mean temperature or range; MR: experiments that measured molting rates alone and not growth; C: copepodite stages; N: naupliar stages

Location Species	Temp. (°C)	Weight method	Growth rate (d ⁻¹)	Incubation period (h)	Source
Australia <i>Acartia fancetti</i>	11–22	Mean weight of stage	0.025–0.26	26–50	Kimmerer & McKinnon (1987)
Skagerrak <i>Centropages typicus</i> <i>Temora longicornis</i> <i>Paracalanus parvus</i> <i>Pseudocalanus</i> spp. <i>Calanus finmarchicus</i> <i>Acartia longiremis</i>	16–17	Length–weight regression	0.24–0.77 0.15–0.56 0.16–0.48 0.12–0.35 0.01–0.14 0.15–0.24	24	Peterson et al. (1991)
Norway <i>Temora longicornis</i>	18	Length–weight regression	0.00–0.32	1 sample (24 h) ⁻¹	Hernández-León et al. (1995)
Canary Islands <i>Oncaea</i> sp.	29				
Jamaica <i>Oikopleura dioica</i> <i>Parvocalanus crassirostris</i>		Length–weight regression	1.38–3.12 0.24–0.79	120 1 sample (24 h) ⁻¹	Hopcroft & Roff (1995)
Jamaica <i>Acartia</i> spp. <i>Centropages velificatus</i> <i>Paracalanus aculeatus</i> <i>Parvocalanus crassirostris</i> <i>Temora turbinata</i> <i>Corycaeus</i> spp. <i>Oithona nana</i> <i>Oithona simplex</i>	28	Length–weight regression	0.25–1.43 0.70–1.00 0.25–1.26 0.44–1.08 0.34–1.23 0.10–0.36 0.40–0.91 0.17–0.53	120 1 sample (24 h) ⁻¹	Hopcroft et al. (1998b) Hopcroft & Roff (1998b,c)
France <i>Acartia bifilosa</i>		Length–weight regression	0.03–0.14	72	Irigoien & Castel (1995)
Agulhas Bank <i>Calanus agulhensis</i>	17–18	Length–weight regression	0.19–0.46	24	Peterson & Hutchings (1995)
North Sea <i>Temora longicornis</i> <i>Pseudocalanus elongatus</i>	6–16	–	MR MR	24	Klein Breteler et al. (1998)
Plymouth <i>Calanus helgolandicus</i>	15	–	MR	48	Shreeve et al. (1998)
Alboran Sea <i>Centropages typicus</i>	17	Volume–weight relationship	<0.01–0.27	24–26	Calbet et al. (2000)
Georges Bank <i>Calanus finmarchicus</i>		Volume–weight relationship	C: –0.09 to 0.31 N: –0.07 to 0.20	48	Campbell et al. (2001a)
Indian Ocean Mixed calanoid guild Mixed cyclopoid guild	21–31	Volume–weight relationship	C: 0.38 N: 0.43 C: 0.28 N: 0.38	48 24 48 24	McKinnon & Duggan (2003)
Great Barrier Reef Mixed calanoid guild Mixed cyclopoid guild	22–30	Volume–weight relationship	C: 0.12–0.53 C: 0.16–0.48	48 (2 expts sampled also after 24 h)	McKinnon et al. (2005)
Gulf of Alaska <i>Neocalanus Flemingi/ plumchrus</i>	5–9	Length–weight regression	0.01–0.28	120	Liu & Hopcroft (2006a)
Gulf of Alaska <i>Metridia pacifica</i>	5–14	length–weight regression	0.01–0.28	96 or 120	Liu & Hopcroft (2006b)
Gulf of Alaska <i>Calanus marshallae</i> , <i>C. pacificus</i>	5–14	Length–weight regression	0.03–0.30	96 or 120	H. Liu & R. R. Hopcroft (unpubl.)

By 2008,
18 studies,
14 regions,
31 species and 4
copepod guilds

Kimmerer et al. (2007)

Deployment of the AC experiments



Size-fractionated vs. manually picked cohorts

Comparison of Two Methods

Single-Stage Method

Pros:

- 1: known development stages for given species
- 2: exact stage-specific growth data available

Cons:

- 1: apt for few large species
- 2: labor intensive and requires suitable working condition at sea

Screen-Filter Method

Pros:

- 1: routinely deployed at sea
- 2: simultaneously applicable to whole copepod community

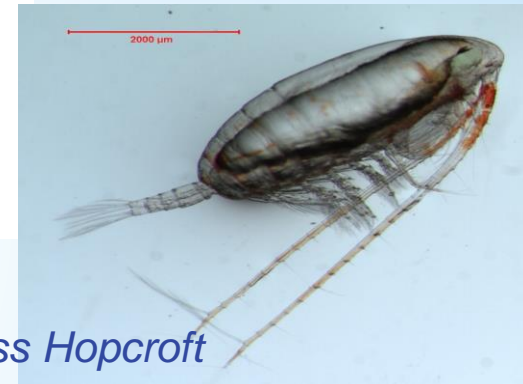
Cons:

- 1: mixed development stages
- 2: exact stage-specific growth data unavailable

Liu et al. (2013)

Stage Error in Screen-Filter Data

Forcing numerical values calculated from mixed stages to represent the single stage to facilitate comparison to the single-stage method



Photos Courtesy of Russ Hopcroft

Size-fractionated vs. manually picked cohorts

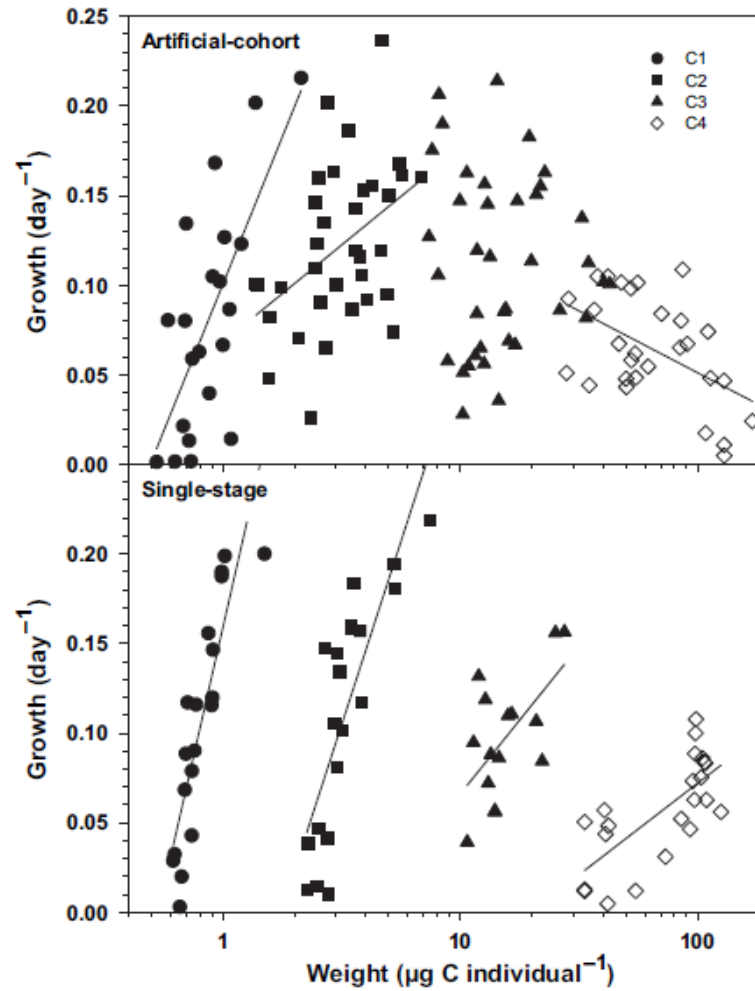
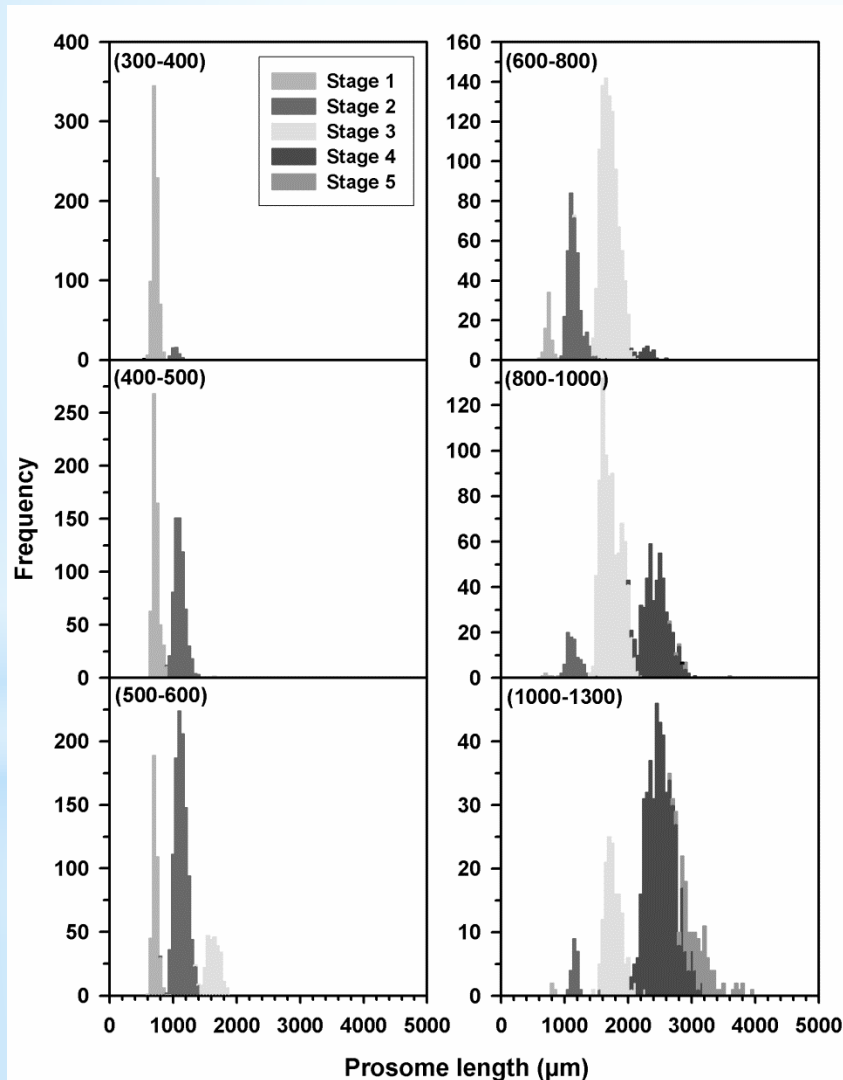


Fig. 6. Relationship between growth rate of *Neocalanus flemingeri/plumidus* and the body weight ($\mu\text{g C individual}^{-1}$) within early copepodite stages estimated by artificial-cohort and single-stage methods in the northern Gulf of Alaska.

Mixed stages in size-fractionated cohorts




$$\bar{C} = \sum_{i=1}^5 (fc)_i / N$$

C: stage (1,2,3,4,5)

f: stage frequency

N: total observations of target species at each mesh size


Quantifying the stage error in size-based cohorts



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Statistical modeling of copepod growth rates: Comparisons for data collections using the artificial cohort (AC) method

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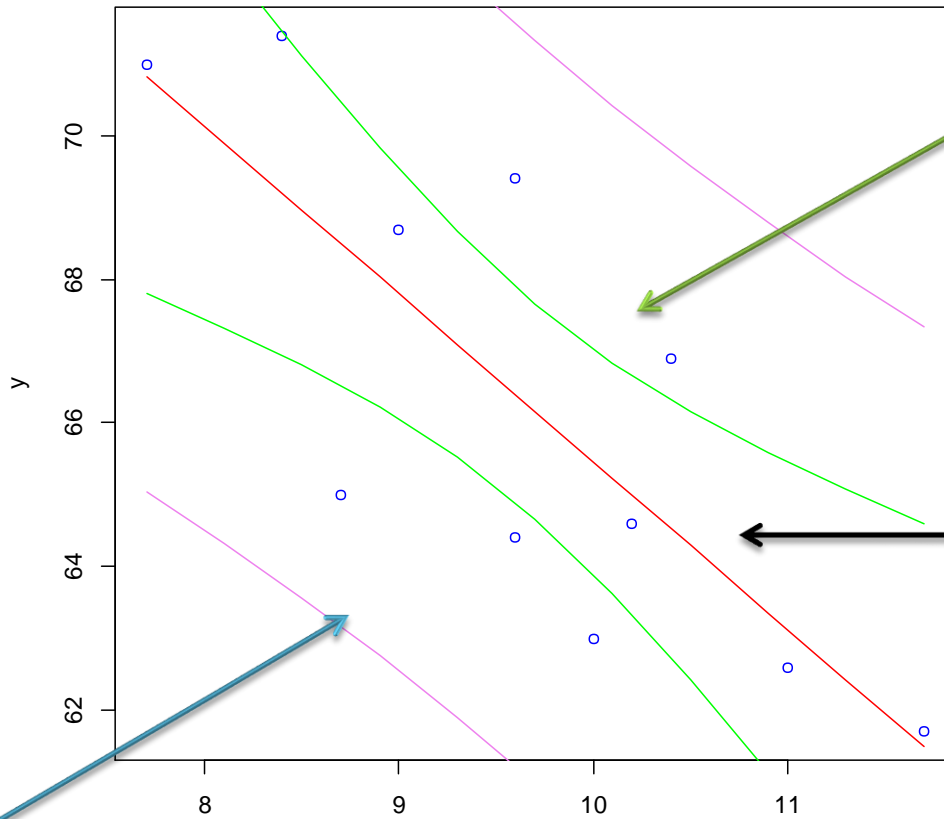
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<p>$S_j(\cdot)$: univariate smoothing function for $i = 1, 2$ $\epsilon_i \sim \text{iid } N(0, \sigma^2)$</p> <p>Model 2 $Y_i = \beta_1 X_{i1} + \beta_2 X_{i2} + \beta_3 X_{i3} + \beta_4 X_{i4} + \beta_5 \log_{10}(X_{i5}) + \beta_6 X_{i6}/(X_{i6} + \beta_7) + \epsilon_i$ The X_{ij} is the same as in Model 1 ($j = 1, \dots, 6$), and, as before, the $\epsilon_i \sim \text{iid } N(0, \sigma^2)$</p> <p>Model 3 $Y_i \sim \text{normal}(\mu_i, \sigma^2)$ $\mu_i = (\beta_1 + \Delta_1 \times \text{ind})X_{i1} + (\beta_2 + \Delta_2 \times \text{ind})X_{i2} + (\beta_3 + \Delta_3 \times \text{ind})X_{i3} + (\beta_4 + \Delta_4 \times \text{ind})X_{i4} + \beta_5 \log_{10}(X_{i5}) + \beta_6 X_{i6}/(X_{i6} + \beta_7)$ The X_{ij} is the same as Model 1 ($j = 1, \dots, 6$) Δ_j ($j = 1, 2, 3, 4$) the stage error for stage 1 to stage 4</p>	<table style="width: 100%; border-collapse: collapse;"> <tr> <td style="width: 50%; vertical-align: top;"> <ol style="list-style-type: none"> 1. Nonlinear Model 2. Parametric approach 3. Prediction </td> <td style="width: 50%; vertical-align: top;"> <p>Same as Model 1</p> </td> </tr> <tr> <td style="width: 50%; vertical-align: top;"> <ol style="list-style-type: none"> 1. Bayesian model 2. Parameters considered as random variables 3. Posterior estimation for parameters based on data available </td> <td style="width: 50%; vertical-align: top;"> <p>Modeling growth rates and discerning the stage errors in the screen-filter data</p> </td> </tr> </table>	<ol style="list-style-type: none"> 1. Nonlinear Model 2. Parametric approach 3. Prediction 	<p>Same as Model 1</p>	<ol style="list-style-type: none"> 1. Bayesian model 2. Parameters considered as random variables 3. Posterior estimation for parameters based on data available 	<p>Modeling growth rates and discerning the stage errors in the screen-filter data</p>
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- Stage errors estimated not significantly different from zero
- The stages labeled with the practical technique in the screen-filter method not statistically different from that identified by manually picking

Uncertainty for using the L-W equations

Example for Simple Linear Regression



$$a + bx^* \pm t_{n-2, \alpha} \sqrt{S^2_{a+bx^*}}$$

$$S_{a+\beta x^*} = S_e \sqrt{\frac{1}{n} + \frac{(x^* - \bar{x})^2}{S_{xx}}}$$

$$\hat{y} = a + bx$$

$$a + bx^* \pm t_{n-2, \alpha} \sqrt{S^2_e + S^2_{a+bx^*}}$$

Inferences of a prediction value $a + \beta x^*$ using the L-W equations

- The estimated standard deviation of the statistic $a + \beta x^*$

is

$$S_{a+\beta x^*} = S_e \sqrt{\frac{1}{n} + \frac{(x^* - \bar{x})^2}{S_{xx}}}$$

- When the assumptions are met, the probability distribution of is t-distribution with $df=(n-2)$

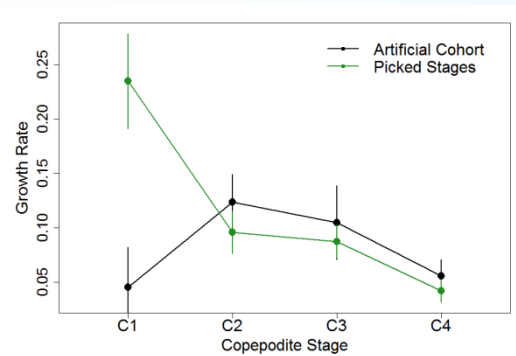
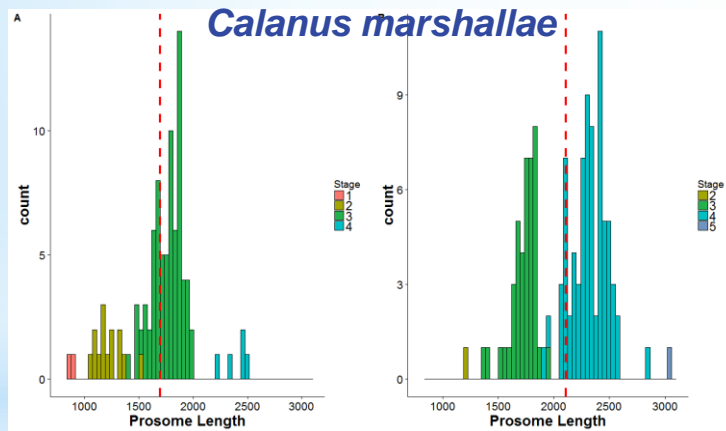
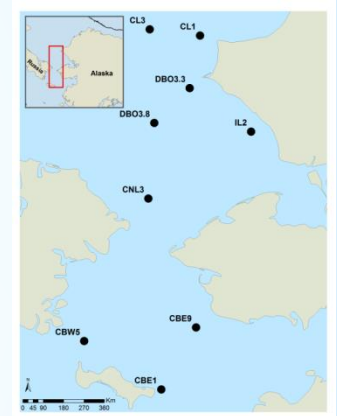
$$t = \frac{a + bx^* - (\alpha + \beta x^*)}{S_{a+\beta x^*}}$$

- Confidence interval for $a + \beta x^*$ has form:

$$a + bx^* \pm t_{n-2, \alpha} S_{a+bx^*}$$

Current AC research in the USA

- Field AC work ongoing in the Bering Sea and Chukchi Sea on *Calanus marshallae* and *Pseudocalanus* spp. (Hopcroft)



- Modeling research on AC (Liu)
- Others

Recap

- Size-based AC method is mostly practical for measuring growth rates of zooplankton with caveats
- Sized-based method can be a substitute of the manually picked methods for measuring stage-specific growth rates
- A potential way to tackle the uncertainty using L-W relationship for measuring biomass
- Methods of theoretically perfect and practically operational

Acknowledgements



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Thank You