### Summer School in Quantitative Fisheries Stock Assessment

## Day 1: Biomass dynamic models

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Unlimited population growth

Limited population growth

Fishing

Estimation

Important points

Summary

### Outline

### Unlimited population growth

Limited population growth

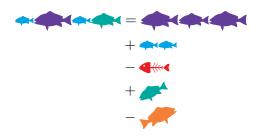
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## Demographic variables



$$N_{t+1} = N_t + B_t - D_t + I_t - E_t$$

#### where

•  $N_t$  is the population number at time t

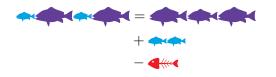


- $B_t$  is the number of births
- $D_t$  is the number of deaths
- $I_t$  is the number immigrating
- $E_t$  is the number emigrating





# Closed population assumption



$$N_{t+1} = N_t + B_t - D_t + \mathcal{Y}_k - \mathcal{Y}_k$$

### Birth and death rates

### Define constant per-capita rates of:

birth

$$b = \frac{}{}$$

death

# Population growth

Can now write

$$N_{t+1} = N_t + bN_t - dN_t$$
$$= N_t + (b - d)N_t$$

Define constant per-capita population growth rate

$$r = (b - d)$$

So

$$N_{t+1} = N_t + rN_t$$
$$= (1+r)N_t$$

# Population growth

Here

$$b = \frac{}{}$$

$$d = \frac{4}{3}$$

$$r = (b - d) = \frac{2}{3} - \frac{1}{3} = \frac{1}{3}$$

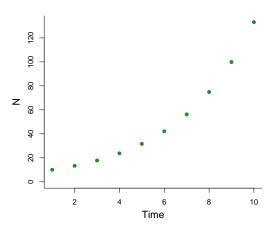
# Population growth

$$\begin{array}{lcl} \textit{N}_{1} & = & \textit{N}_{0} \left( 1 + \frac{1}{3} \right) \\ \textit{N}_{2} & = & \textit{N}_{1} \left( 1 + \frac{1}{3} \right) = \textit{N}_{0} \left( 1 + \frac{1}{3} \right) \times \left( 1 + \frac{1}{3} \right) \\ \textit{N}_{3} & = & \textit{N}_{2} \left( 1 + \frac{1}{3} \right) = \textit{N}_{0} \left( 1 + \frac{1}{3} \right) \times \left( 1 + \frac{1}{3} \right) \times \left( 1 + \frac{1}{3} \right) \\ \vdots & \vdots & \vdots \\ \textit{N}_{t} & = & \textit{N}_{0} \left( 1 + \frac{1}{3} \right)^{t} \end{array}$$

### Geometric population growth

$$N_t = N_0 \left(1 + r\right)^t$$

## Geometric population growth



"Population, when unchecked, increases in a geometrical ratio".

## Geometric population growth

Populations will be limited though through competition for limited:

- Food
- Space
- Light
- . . .

Need to include effect of competition on the population growth rate, i.e., density dependence

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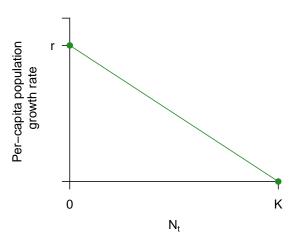
Important points

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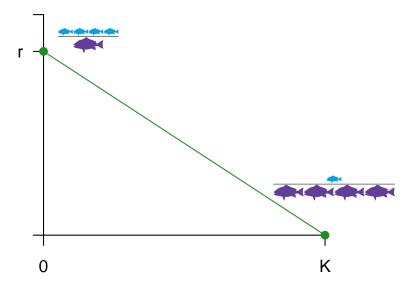
For geometric population growth:

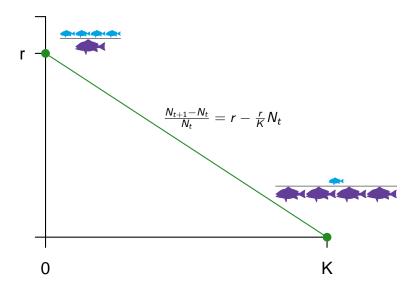
$$\begin{array}{rcl}
N_{t+1} & = & N_t + rN_t \\
\frac{N_{t+1} - N_t}{N_t} & = & r
\end{array}$$

Now let the per-capita growth rate be a **decreasing** function of population size, i.e., not constant



Increase in per-capita rate of population growth  $\underline{\text{compensates}}$  for reduced population size





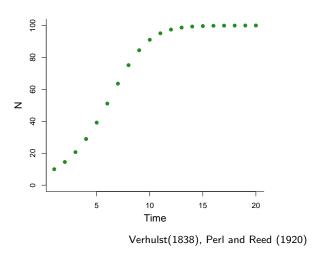
$$\frac{N_{t+1} - N_t}{N_t} = r - \frac{r}{K} N_t$$

$$N_{t+1} = N_t + rN_t \left(1 - \frac{N_t}{K}\right)$$

- Termed the discrete logistic population model
- Cornerstone of initial theory of compensation
- Often used in stock assessment of harvested populations
- Note  $N_t$  not disaggregated by age (would be  $N_{a,t}$ )

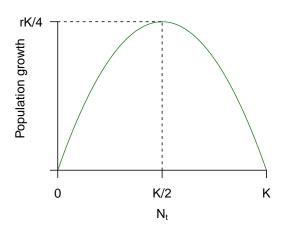
## Limited population growth

### Logistic model



### Overall population growth rate

Population growth (per-capita  $\times$  population size) has an optimum



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# Russell (1931)

### Russell's Fishery equation:

$$S_2 = S_1 + (R + G) - (M + F)$$

### where

- S is the stock size
- R is recruitment of new individuals
- G is somatic growth
- M is natural mortality
- *F* is fishing mortality

# Russell (1931)

Can cast in terms of our simple population model

$$B_{t+1} = B_t + R_t - D_t - C_t$$

#### where here

- B<sub>t</sub> is now the biomass of the population (numbers times mass)
- R<sub>t</sub> is the biomass of new recruits
- $D_t$  is the biomass dying naturally
- C<sub>t</sub> is mass caught (catch)

### Graham-Schaefer biomass dynamic model

Difference (non-continuous) version

$$B_{t+1} = B_t + rB_t \left( 1 - \frac{B_t}{K} \right) - C_t$$

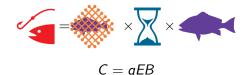
Often called just the Schaefer model. Widely-used model in fisheries assessment

### Note

The Schaefer biomass dynamic model can be derived simply from a linear decrease in the per-capita growth rate over abundance with catch subtracted.

### A first catch equation

### Catch can be modelled as:



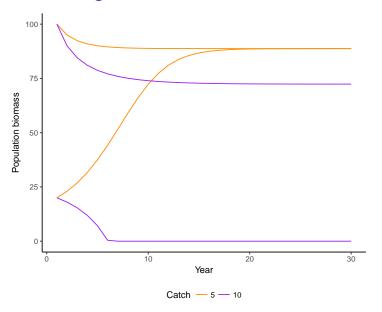
### Where

- C is the catch
- q is catchability, defined as the proportion of the population removed per unit effort
- E is effort (e.g., fishing days, kW days, hooks)

### What is a sustainable catch?

Q: For a given r and K parameters what values of catch are sustainable?

## Catch and starting biomass



## Catch and starting biomass

### Note

Sustainability of a given catch depends on what level of biomass you start from.

Better to work with a rate of removal of the population

## Fishing mortality

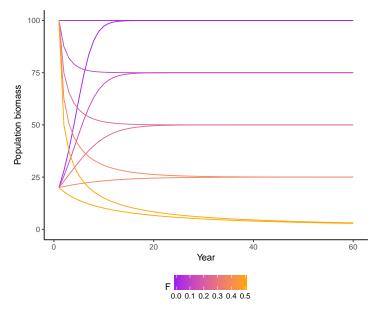
#### Note

Catchability and effort are often combined into the fishing mortality

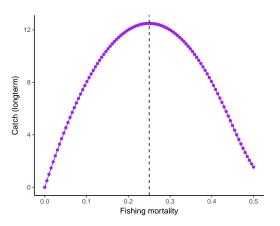
$$F = qE$$

which is usually expressed as an instantaneous rate but used here to denote the proportion of the population removed by fishing each year (often termed the harvest rate, sometimes denoted u).

# What is a sustainable rate of fishing mortality?



## Try different values for the harvest rate



- $F_{MSY} = \frac{r}{2}$  in the Schaefer model
- K/2 termed  $B_{MSY}$  long-term biomass obtained if fishing at  $F_{MSY}=r/2$

More on reference points later . . .



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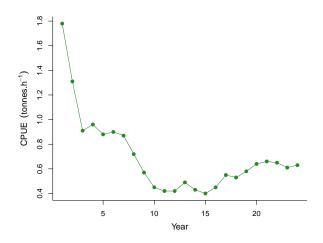
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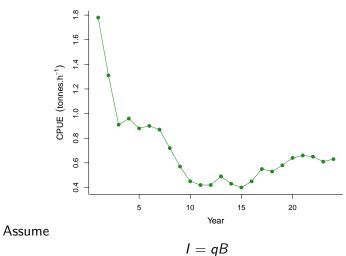
Need to estimate the parameters of the biomass dynamic model from:

- Biomass index *I*, e.g.
  - Survey index
  - Standardised CPUE index
- Catch C

### Biomass index



### Biomass index



Index is proportional to biomass

## Predicting the biomass index

Say we guessed the parameters, we would predict the index values via

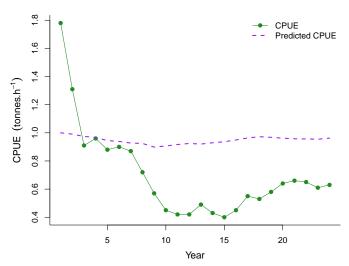
$$egin{aligned} \hat{B}_{t=1} &= \hat{B}_0 \ \hat{B}_{t+1} &= \hat{B}_t + \hat{r}\hat{B}_t \left(1 - rac{\hat{B}_t}{\hat{K}}\right) - C_t \ \hat{I}_t &= \hat{q}\hat{B}_t \end{aligned}$$

#### Note

This is an observation error approach

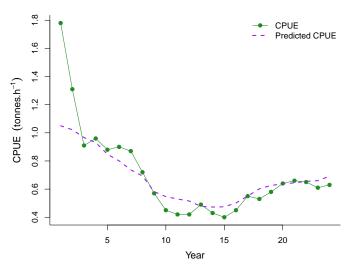
## Predicting the biomass index

Say we guessed  $\hat{r} = 0.5, \ \hat{K} = 10,000, \hat{B}_0 = 10,000, \ \hat{q} = 0.0001$ 



# Another guess

Say we guessed  $\hat{r} = 0.3$ ,  $\hat{K} = 3,500$ ,  $\hat{B}_0 = 3,500$ ,  $\hat{q} = 0.0003$ 



#### **Estimation**

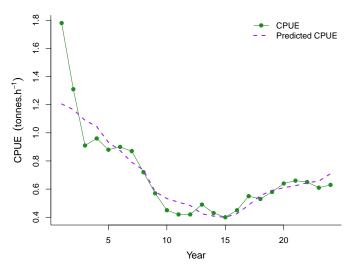
Like to do better than guess, so we estimate Define the likelihood

$$L(r, K, B_0, q, \sigma | \mathbf{I}) = \prod_{t=1}^{T} N(\ln(I_t) - \ln(\hat{I}_t), \sigma^2)$$

which we maximise with respect to the parameters to obtain the maximum likelihood estimates

### Maximum likelihood estimates: observation error

$$\hat{r} = 0.37, \ \hat{K} = \hat{B}_0 = 2,823, \ \hat{q} = 0.00042, \ \hat{\sigma} = 0.125$$



## Process error approach

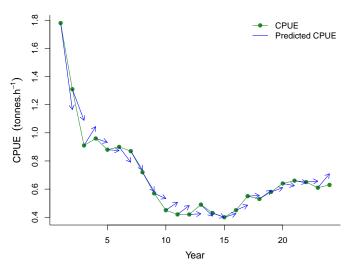
Rather than predict all the way through, a process error only model assumes the observations are perfect and the only error is in the biomass dynamic model

#### Note

A process error only approach assumes only source of error is in the biomass dynamics, not the observations.

# Maximum likelihood estimates: process error

$$\hat{r} = 0.31, \ \hat{K} = \hat{B}_0 = 3,577, \ \hat{q} = 0.00025, \ \hat{\sigma} = 0.1$$



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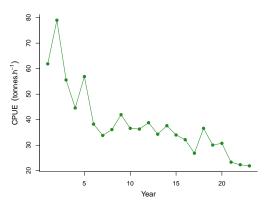
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#### Contrast

Need historical variation in stock size and fishing pressure to properly estimate the parameters of the biomass dynamic model. Beware of one-way trip data like this



- r estimation requires points at low stock size and low effort
- K estimation requires high stock sizes and low effort
- Ideally there would be variation in both



#### Alternative forms

A common alternative form is the Pella-Tomlinson model



$$B_{t+1} = B_t + \frac{r}{n}B_t\left(1 - \left(\frac{B_t}{K}\right)^n\right) - C_t$$

# State space formulation

Switch to spacing out slides

# State space formulation

Recently, much interest in estimation including  $\underline{both}$  observation and process error (spict)

Process equations: 
$$B_{t+1} = \left(B_t + rB_t\left(1 - \frac{B_t}{K}\right) - C_t\right)e^{\eta_t}$$
  $F_{t+1} = F_t e^{\epsilon_t}$  Observation equations:  $I_t = qB_t e^{\varepsilon_t}$ 

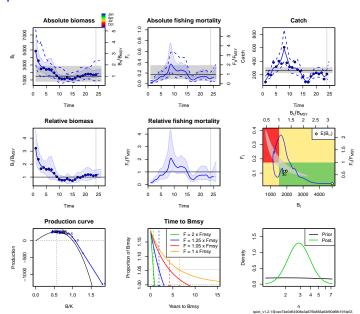
 $C_t = F_t B_t e^{\zeta_t} (\text{roughly})$ 

where  $\eta_t, \varepsilon_t$  are process and measurement errors on biomass and the index;  $\epsilon_t, \zeta_t$  are process and measurement errors on fishing

#### Note

spict also includes a rich output but care needed in terms of the number of parameters estimated and given data quality. Make sure to check parameter correlation and diagnostics.

## State space formulation



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- Reference points straightforward to obtain from a good fit
- Advanced models provide options for multiple indices, sources of error, asymmetric production
- Critical evaluation of the assessment fit and diagnostics very important with aggregate biomass dynamic models

### References

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Meyer, R., and Millar, R. B. (1999). BUGS in Bayesian stock assessments. Canadian Journal of Fisheries and Aquatic Sciences, 56(6), 1078-1087.

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