

Analysis of Size-Frequency Data (aka Mean Length Approach)

SEDAR 26
October, 2011



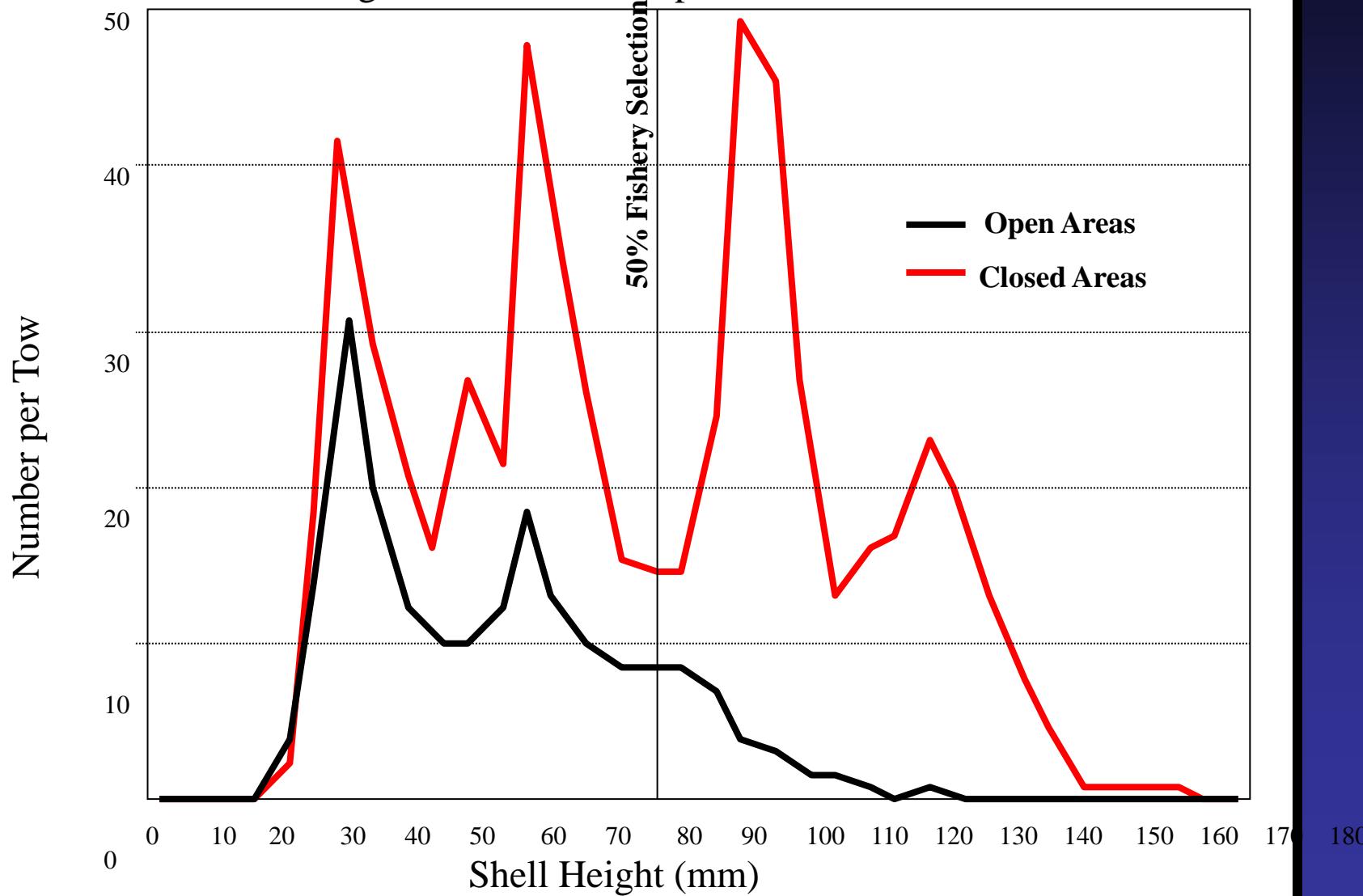
**NOAA
FISHERIES
SERVICE**

Todd Gedamke (SEFSC)

Early Use of Length Based Methods—late 1800's

(Petersen, then Fulton, Baranov and others)

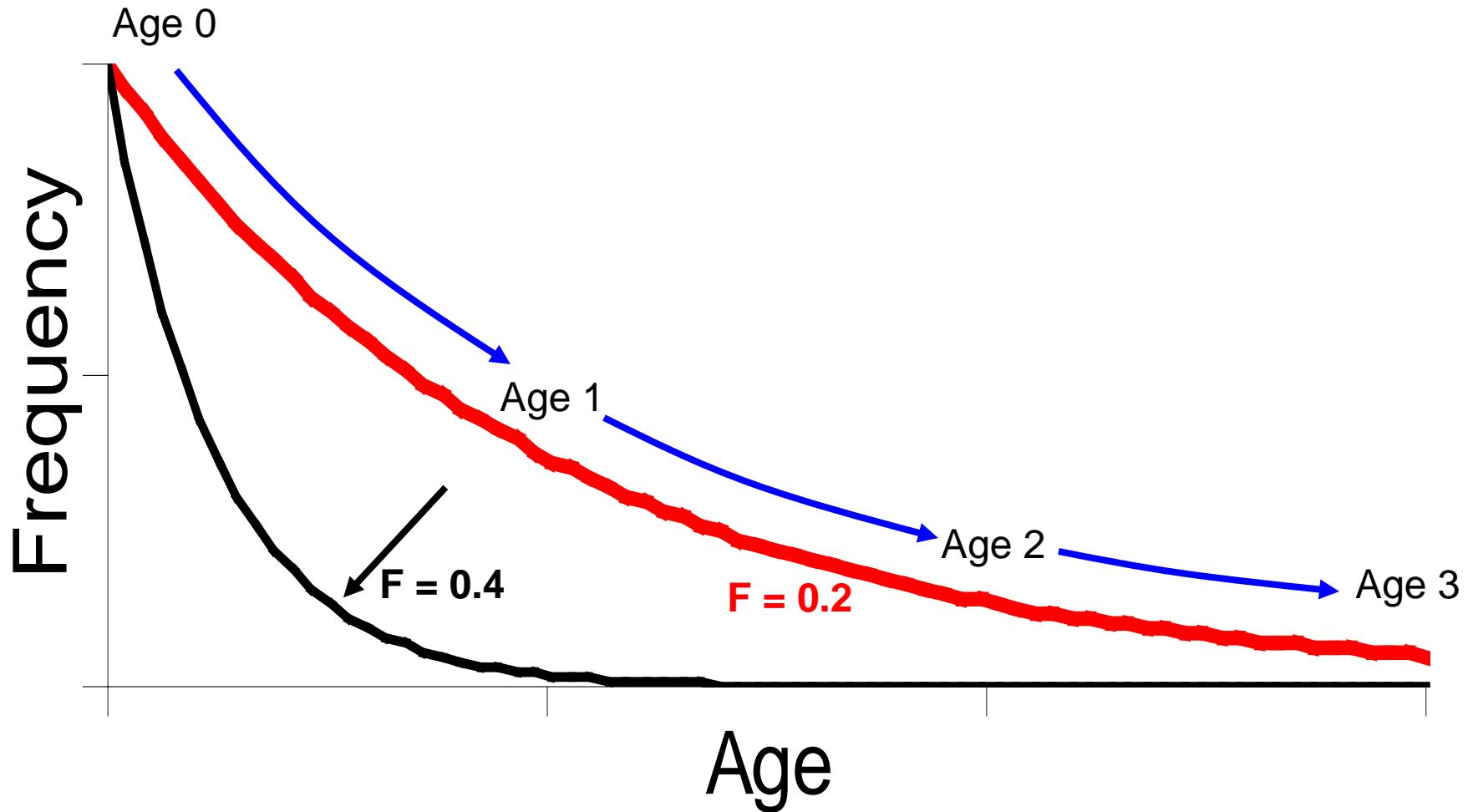
Georges Bank Sea Scallops-1998



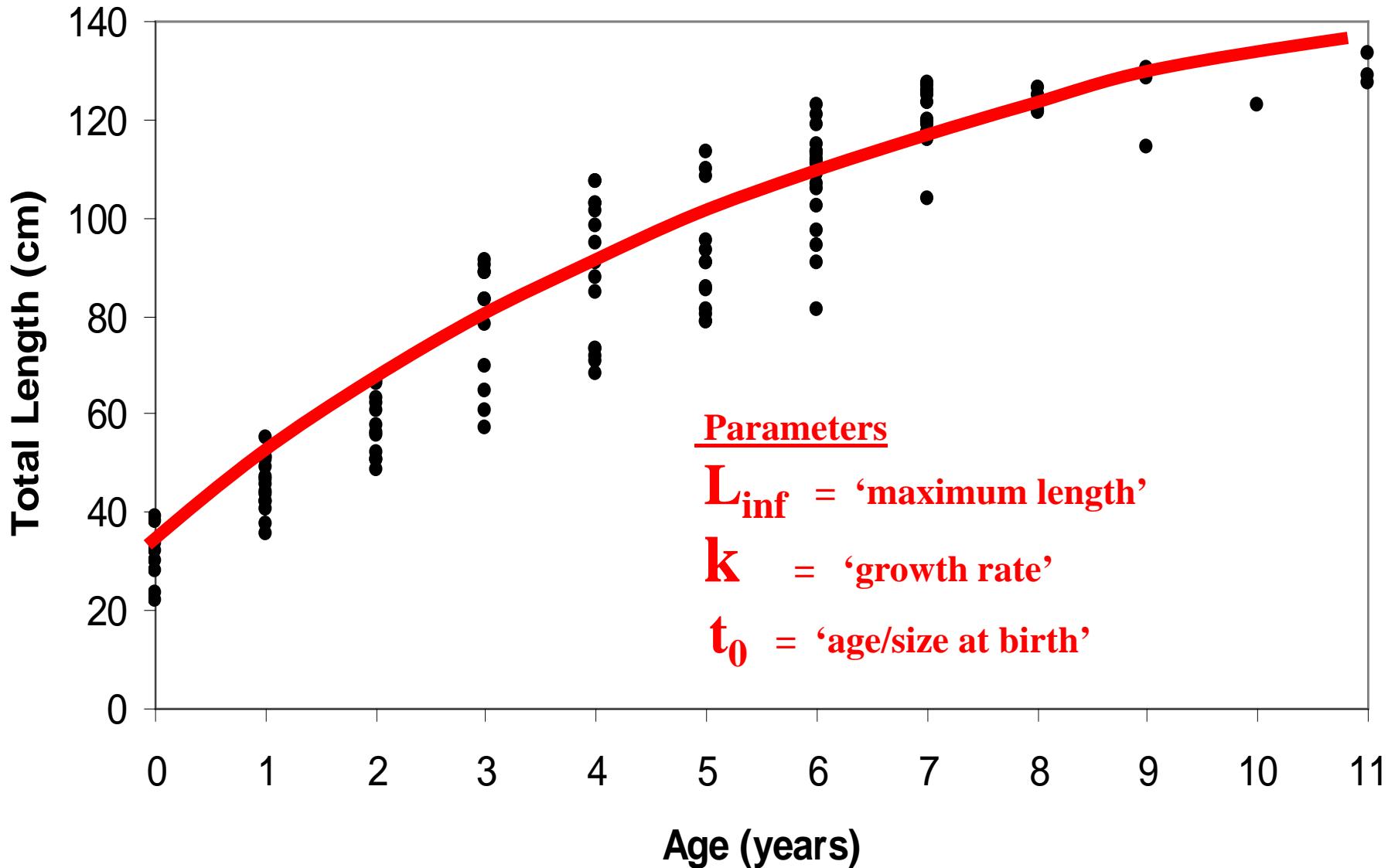
What determines mean length?

$$\bar{L} = \frac{\int_{t_c}^{\infty} N_t L_t dt}{\int_{t_c}^{\infty} N_t dt} = \frac{\sum \text{Number of animals at age } t \times \text{Length at age } t}{\sum \text{Number of animals at age } t}$$

More Fishing → Less Older/Larger Fish



Von Bertalanffy Growth Function



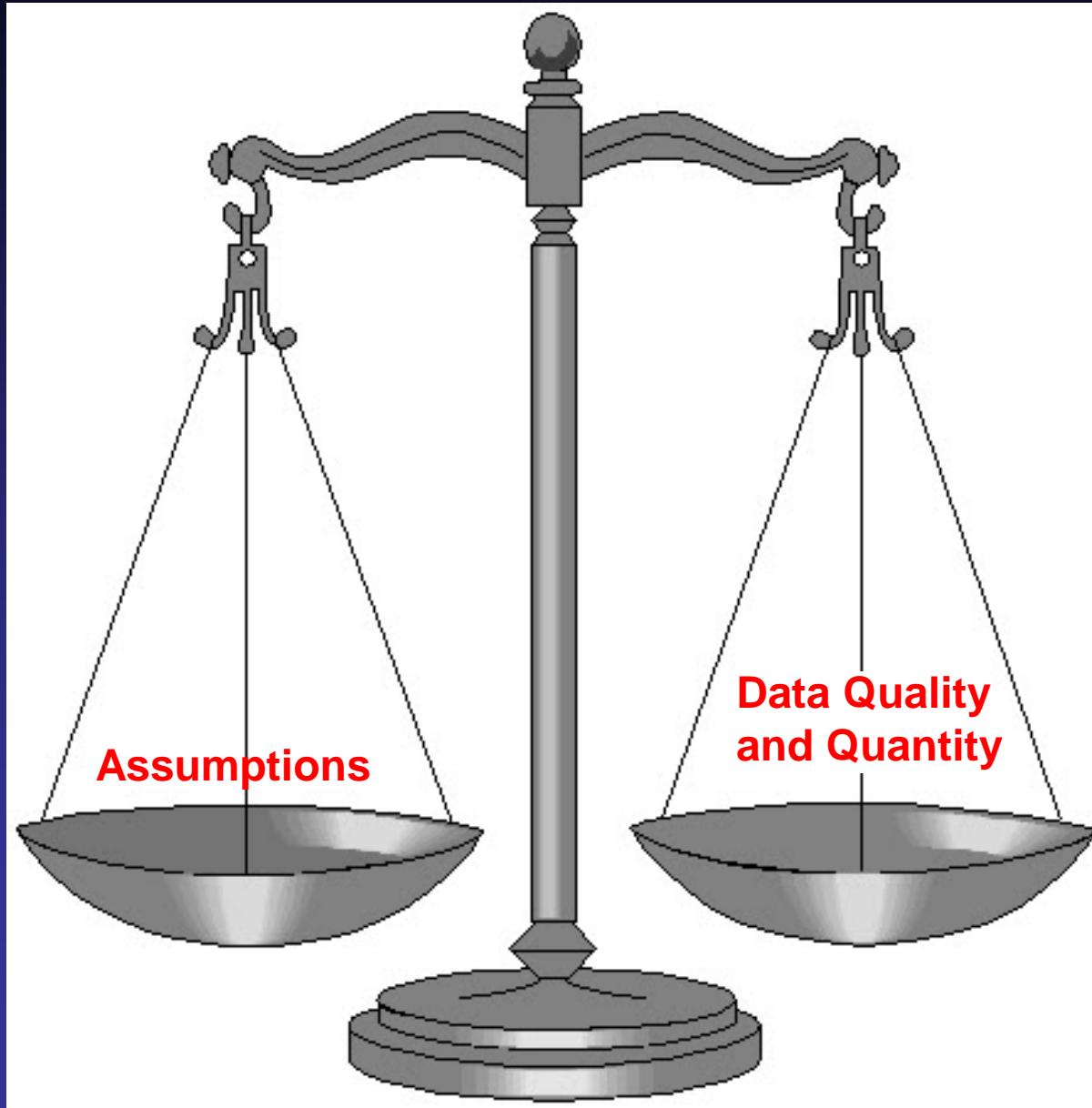
Beverton-Holt mean length mortality estimator

$$Z = \frac{K(L_{\infty} - \bar{L})}{\bar{L} - L_c}$$

Diagram illustrating the components of the Beverton-Holt mean length mortality estimator:

- total mortality (points to the term Z)
- mean length (points to the term \bar{L})
- length at which all animals are fully vulnerable to gear (points to the term L_c)
- growth rate (points to the coefficient K)
- maximum length (points to the term L_{∞})

The Trade-off!



Beverton-Holt mean length mortality estimator

5 assumptions:

$$Z = \frac{K(L_\infty - \bar{L})}{\bar{L} - L_c}$$

1. Asymptotic growth, K and L_∞ known & constant over time.
2. No individual variability in growth.
3. ‘Constant’ & continuous recruitment over time.
4. Mortality constant with age (eg. Selectivity, M).
5. Mortality constant over time → Population in equilibrium (mean length reflects mortality)

Von Bertalanffy Growth Function

Total Length (cm)

140

120

100

80

60

40

0



3



Beverton-Holt mean length mortality estimator

5 assumptions:

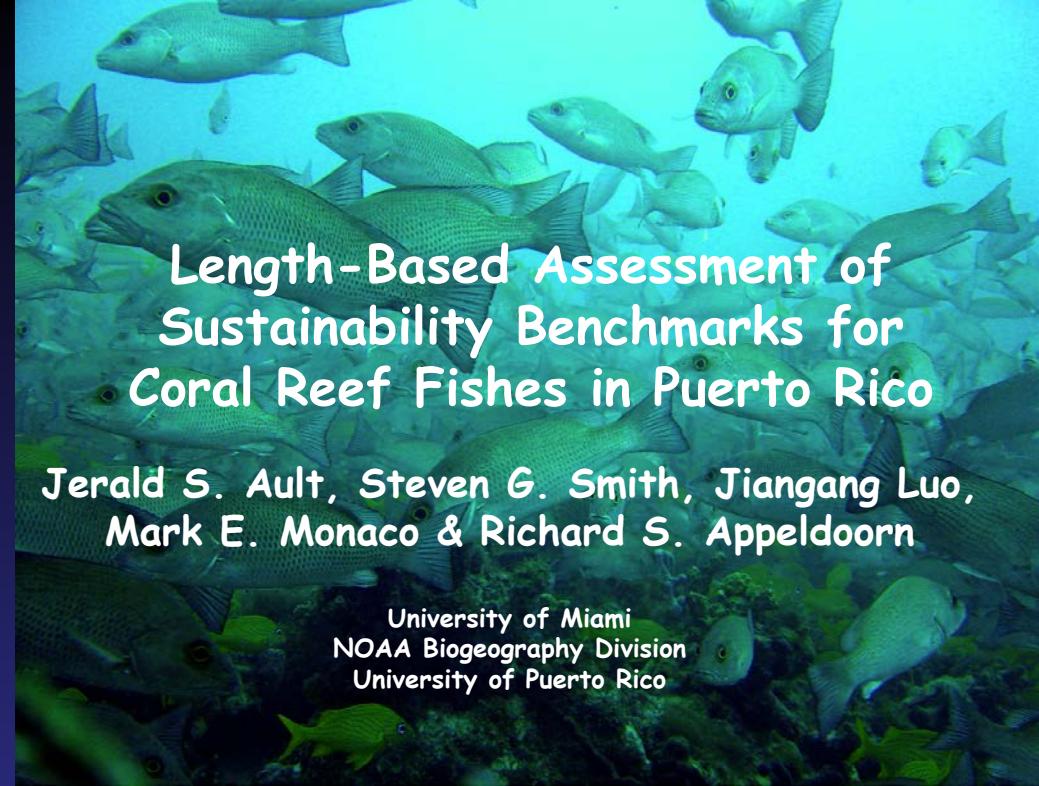
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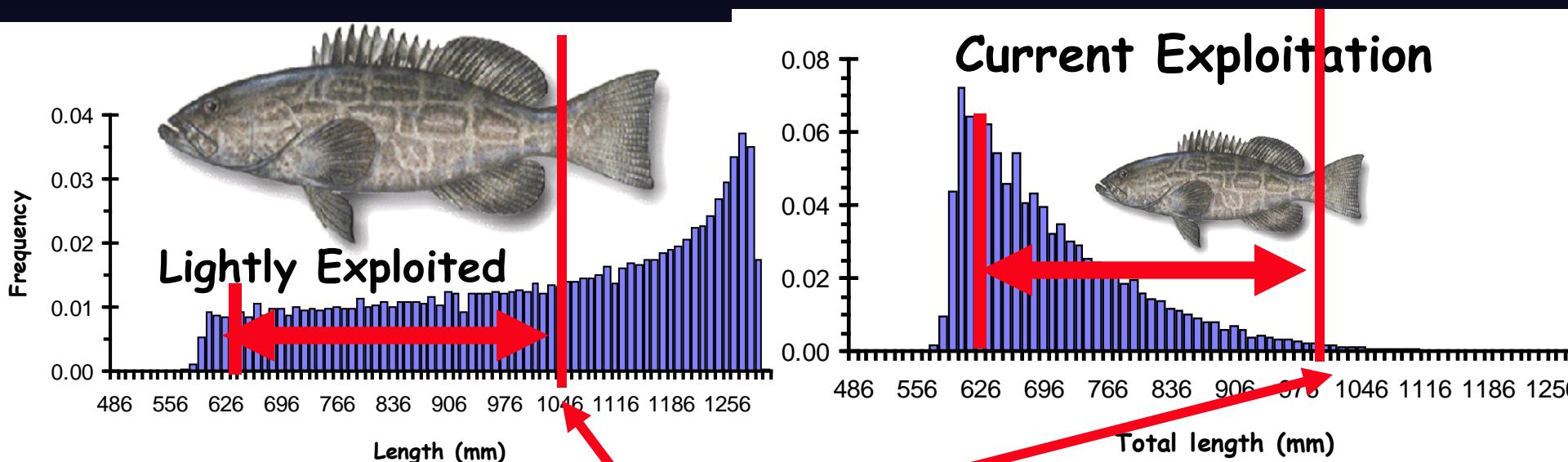
Modifications of the B-H Mortality Estimator

5 assumptions:

1. Asymptotic growth, K and L_{∞} known & constant over time.
2. No individual variability in growth.
3. ‘Constant’ & continuous recruitment over time.
4. **Mortality constant with age (eg. Selectivity, M).**
5. Mortality constant over time → Population in equilibrium (mean length reflects mortality)



Ehrhardt-Ault mortality estimators



$$\bar{L}(t) = \frac{\int_{a_c}^{a_\lambda} F(t) N(a, t) L(a, t) da}{\int_{a_c}^{a_\lambda} F(t) N(a, t) da}$$

Estimating Mortality from Mean Lengths in Non-equilibrium Situations

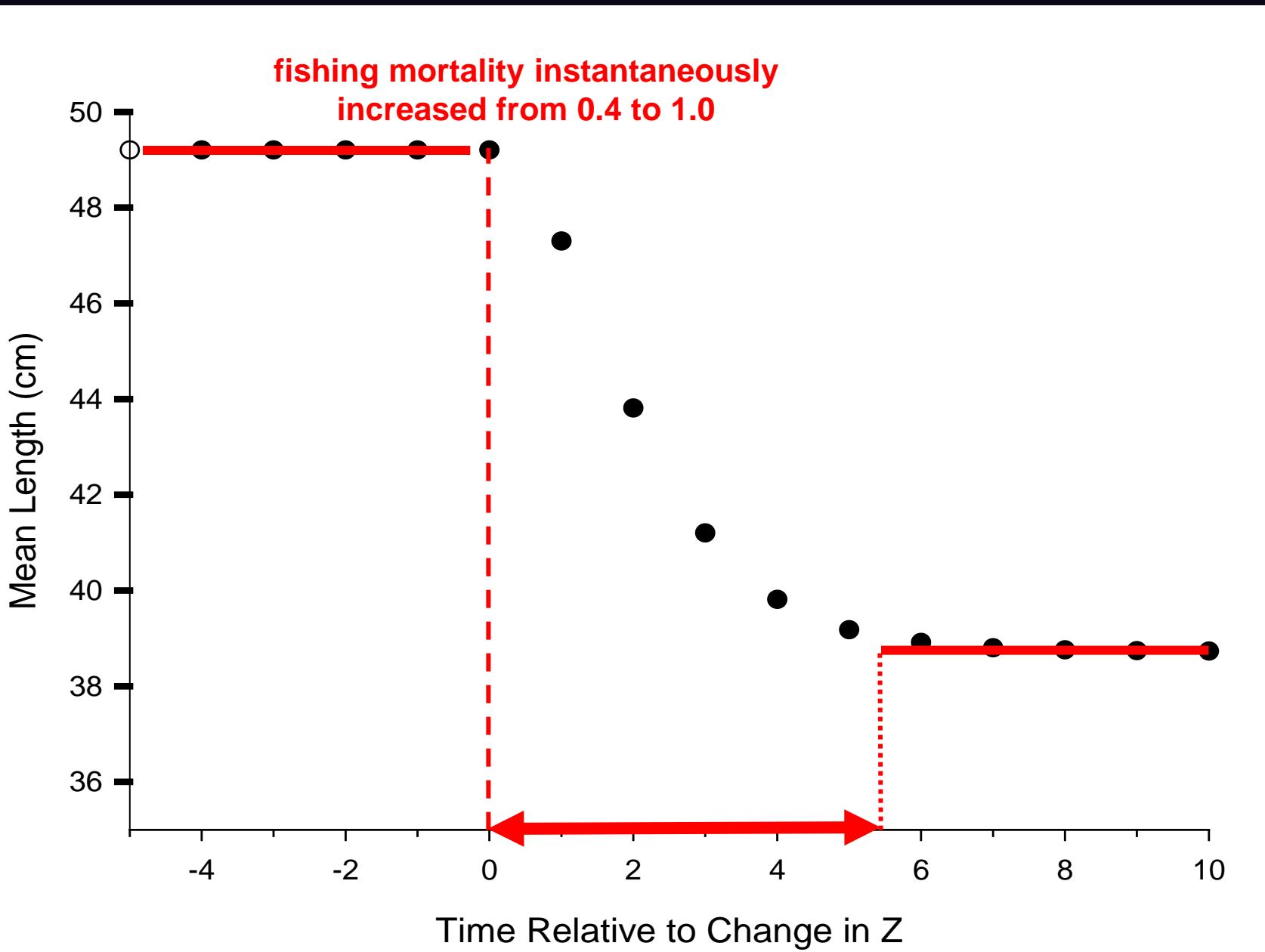
Gedamke and Hoenig (2006)

Beverton-Holt mean length mortality estimator

$$Z = \frac{K(L_\infty - \bar{L})}{\bar{L} - L_c}$$

5 assumptions:

1. Asymptotic growth, K and L_∞ known & constant over time.
2. No individual variability in growth.
3. ‘Constant’ & continuous recruitment over time.
4. Mortality constant with age (eg. Selectivity, M).
5. Mortality constant over time → Population in equilibrium (mean length reflects mortality)



$$\bar{L} = \frac{\int_{t_c}^g N_o \exp(-Z_2(t-t_c)) L_t dt + \int_g^\infty N_o \exp(-Z_2 d) \exp(-Z_1(t-g)) L_t dt}{\int_{t_c}^g N_o \exp(-Z_2(t-t_c)) dt + \int_g^\infty N_o \exp(-Z_2 d) \exp(-Z_1(t-g)) dt}$$

In English:

Mean Length =

$$\frac{\text{SUM # of younger animals at each age (Z}_2\text{ only)} \bullet \text{Length at age} + \text{SUM # of older animals at each age (both Z}_1\text{ and Z}_2\text{)} \bullet \text{Length at age}}{\text{SUM # of younger animals at each age (Z}_2\text{ only)} + \text{SUM # of older animals at each age (both Z}_1\text{ and Z}_2\text{)}}$$

$$\bar{L} = \frac{\int_{t_c}^g N_o \exp(-Z_2(t-t_c)) L_t dt + \int_g^\infty N_o \exp(-Z_2 d) \exp(-Z_1(t-g)) L_t dt}{\int_{t_c}^g N_o \exp(-Z_2(t-t_c)) dt + \int_g^\infty N_o \exp(-Z_2 d) \exp(-Z_1(t-g)) dt}$$

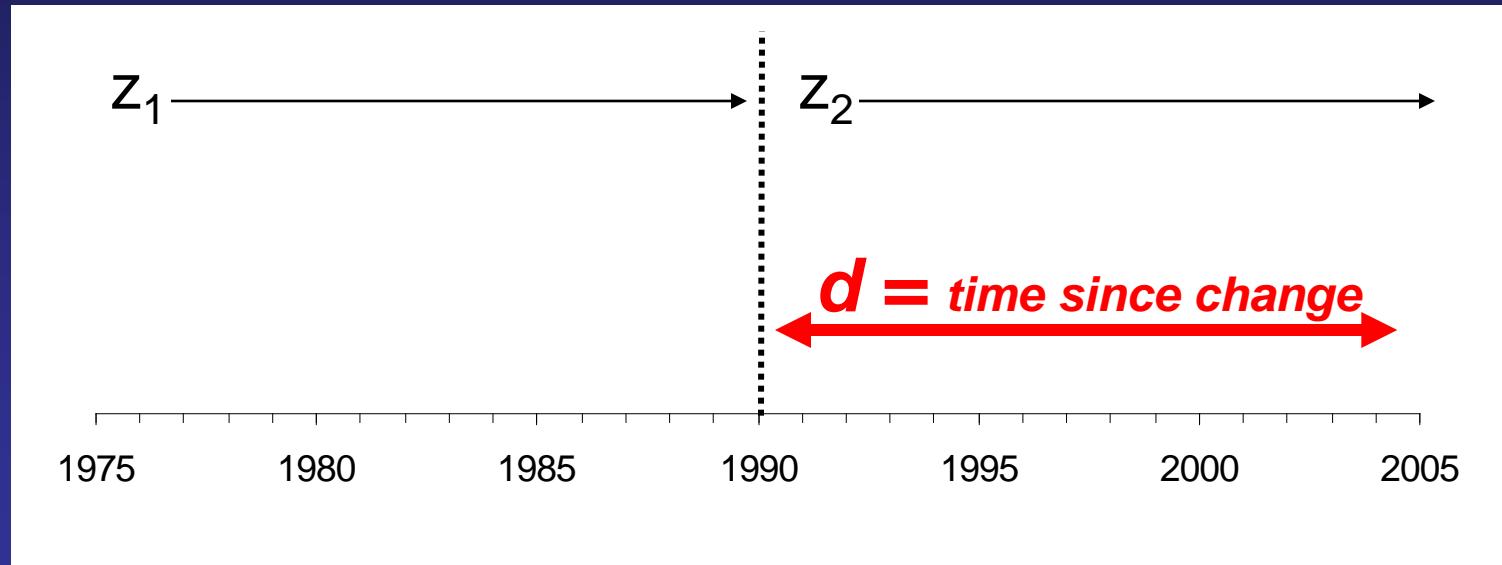
In English:

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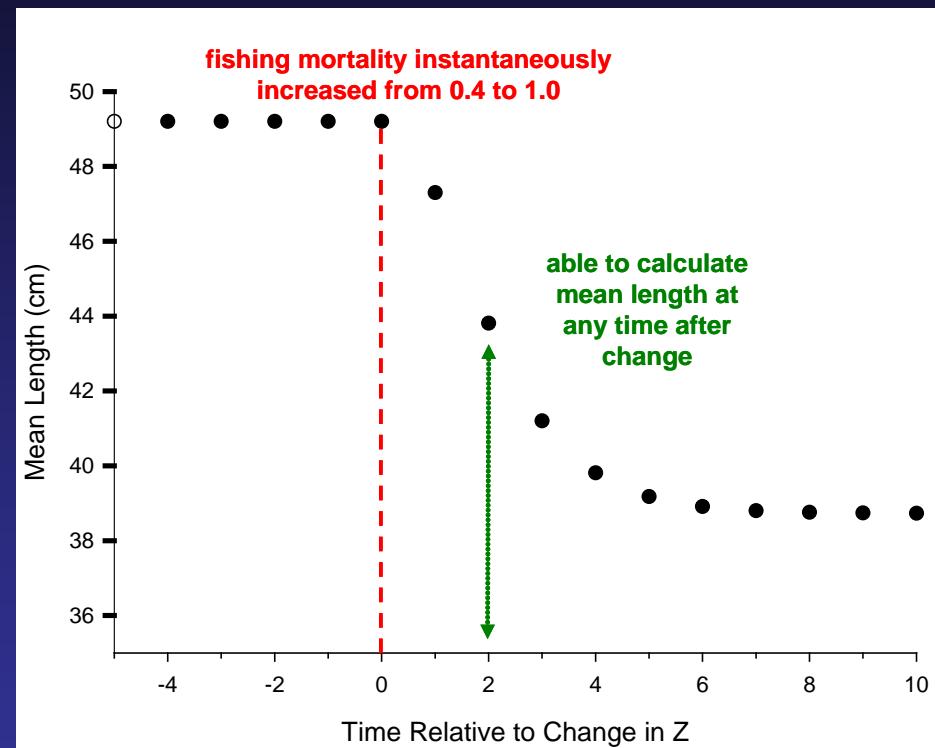
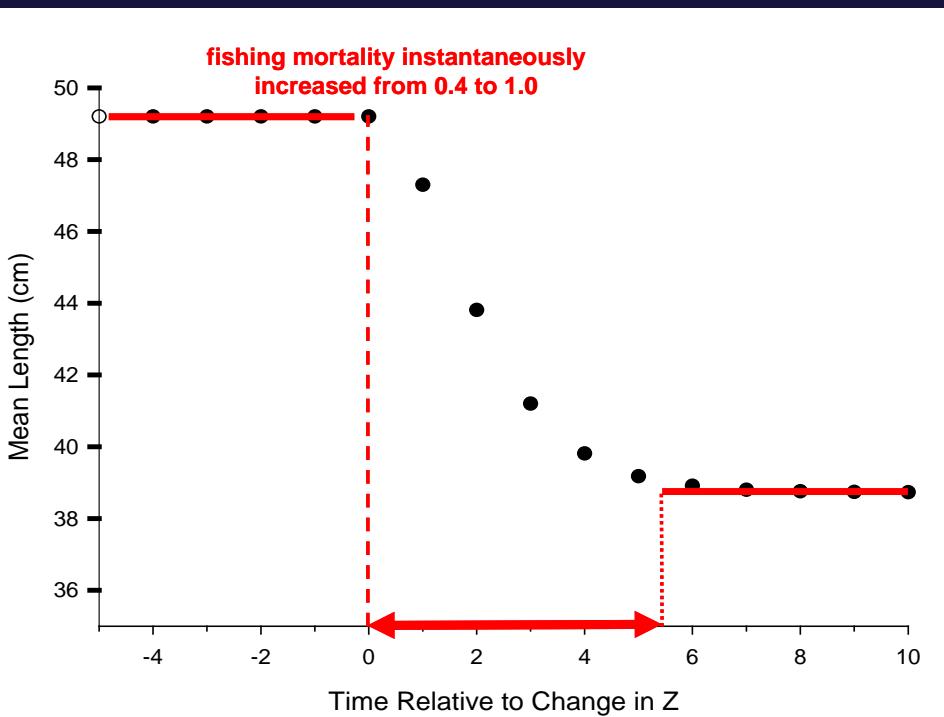
Simplified Equation (at least relatively)

$$\bar{L} = L_{\infty} - \frac{Z_1 Z_2 (L_{\infty} - L_c) \{Z_1 + K + (Z_2 - Z_1) \exp(-(Z_2 + K)d)\}}{(Z_1 + K)(Z_2 + K)(Z_1 + (Z_2 - Z_1) \exp(-Z_2 d))}$$



Estimating Mortality from Mean Lengths in Non-equilibrium Situations

Gedamke and Hoenig (2006)



$$\bar{L} = L_{\infty} - \frac{Z_1 Z_2 (L_{\infty} - L_c) \{ Z_1 + K + (Z_2 - Z_1) \exp(-(Z_2 + K)d) \}}{(Z_1 + K)(Z_2 + K)(Z_1 + (Z_2 - Z_1) \exp(-Z_2 d))}$$

Assumption 5

Population in equilibrium (enough time elapsed after change in mortality that mean length reflects new mortality).

Hard to meet in the real world!

Years to “reach” equilibrium after change in mortality

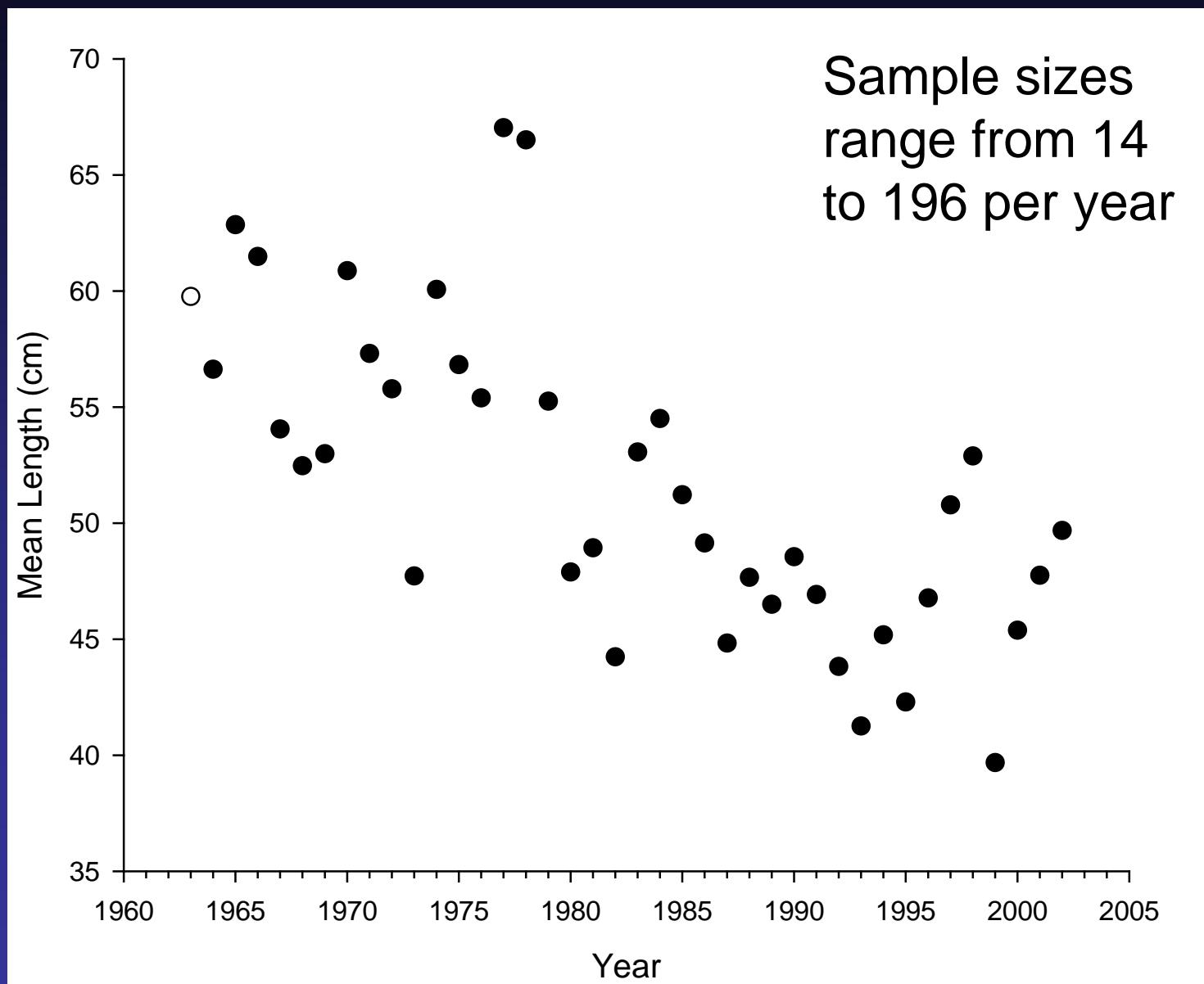
	Z_2				
	0.1	0.3	0.5	0.7	0.9
Z_1	-	14	11	9	7
0.3	26	-	6	6	6
0.5	28	8	-	4	4
0.7	29	10	5	-	3
0.9	29	10	6	3	-

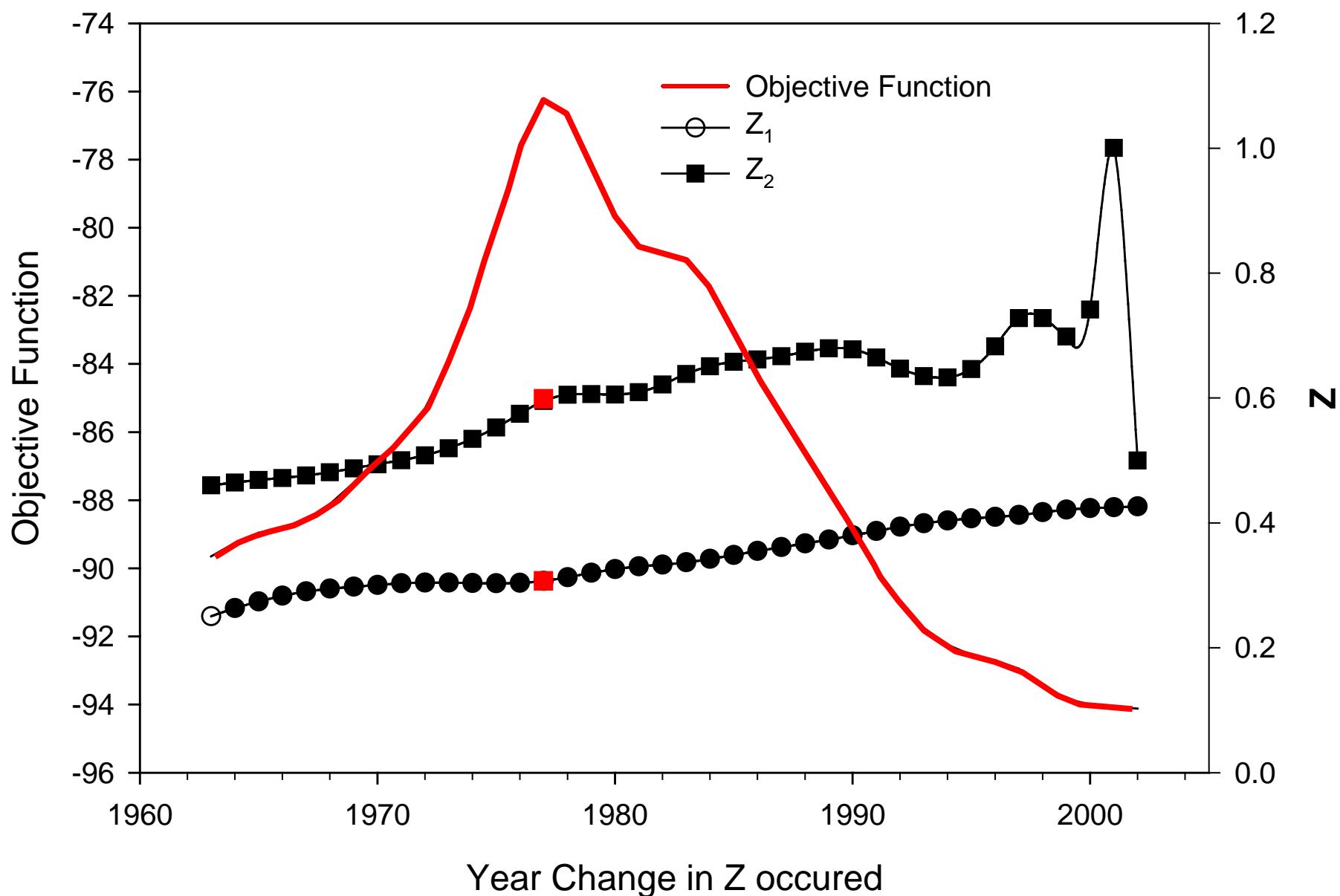
Application to the Assessment of Goosefish *(Lophius americanus)*

data: mean lengths over time, sample sizes
maximum likelihood estimation

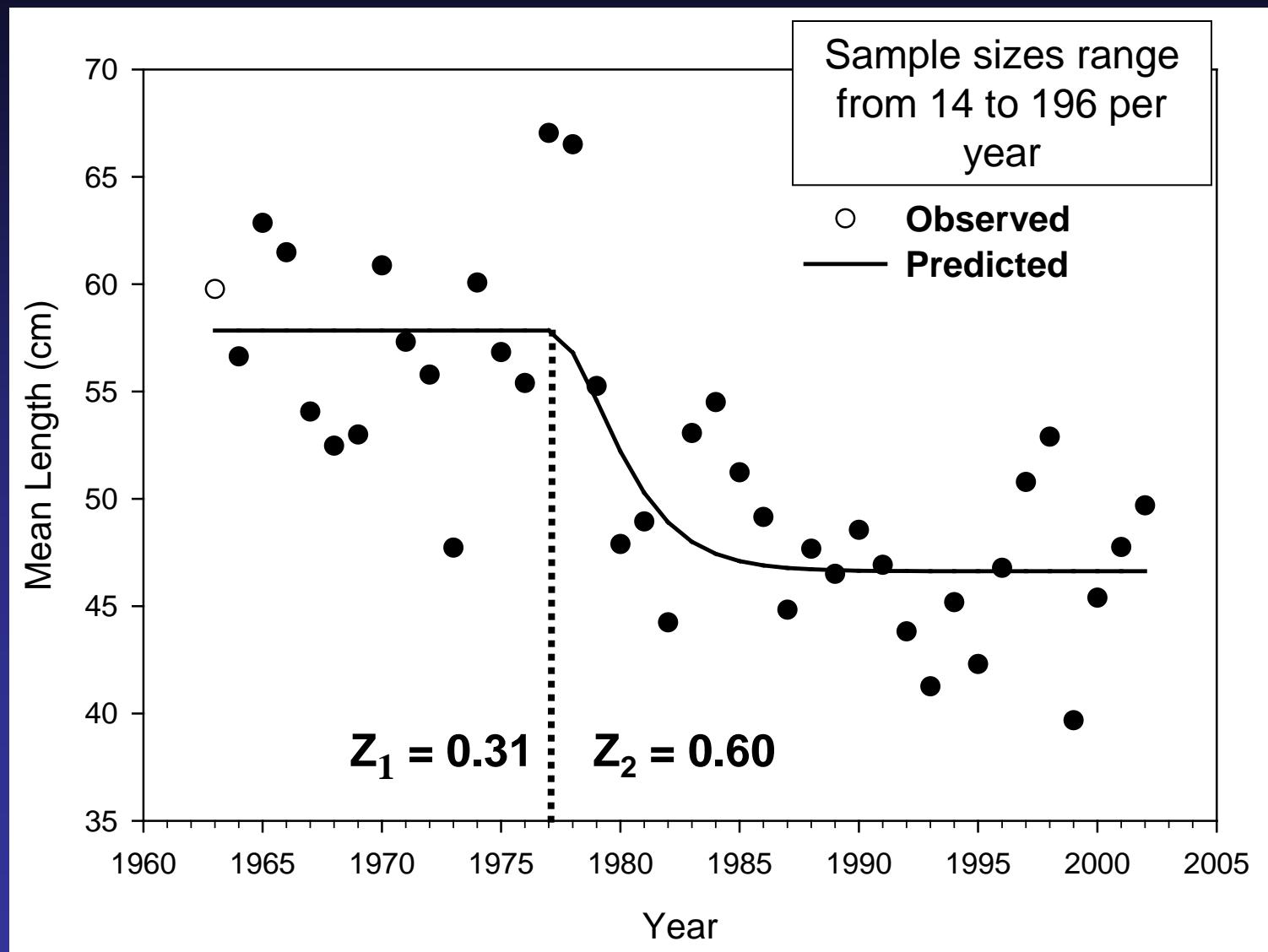


Goosefish Mean Length Data - Southern Management Region - NEFSC Fall Groundfish Survey

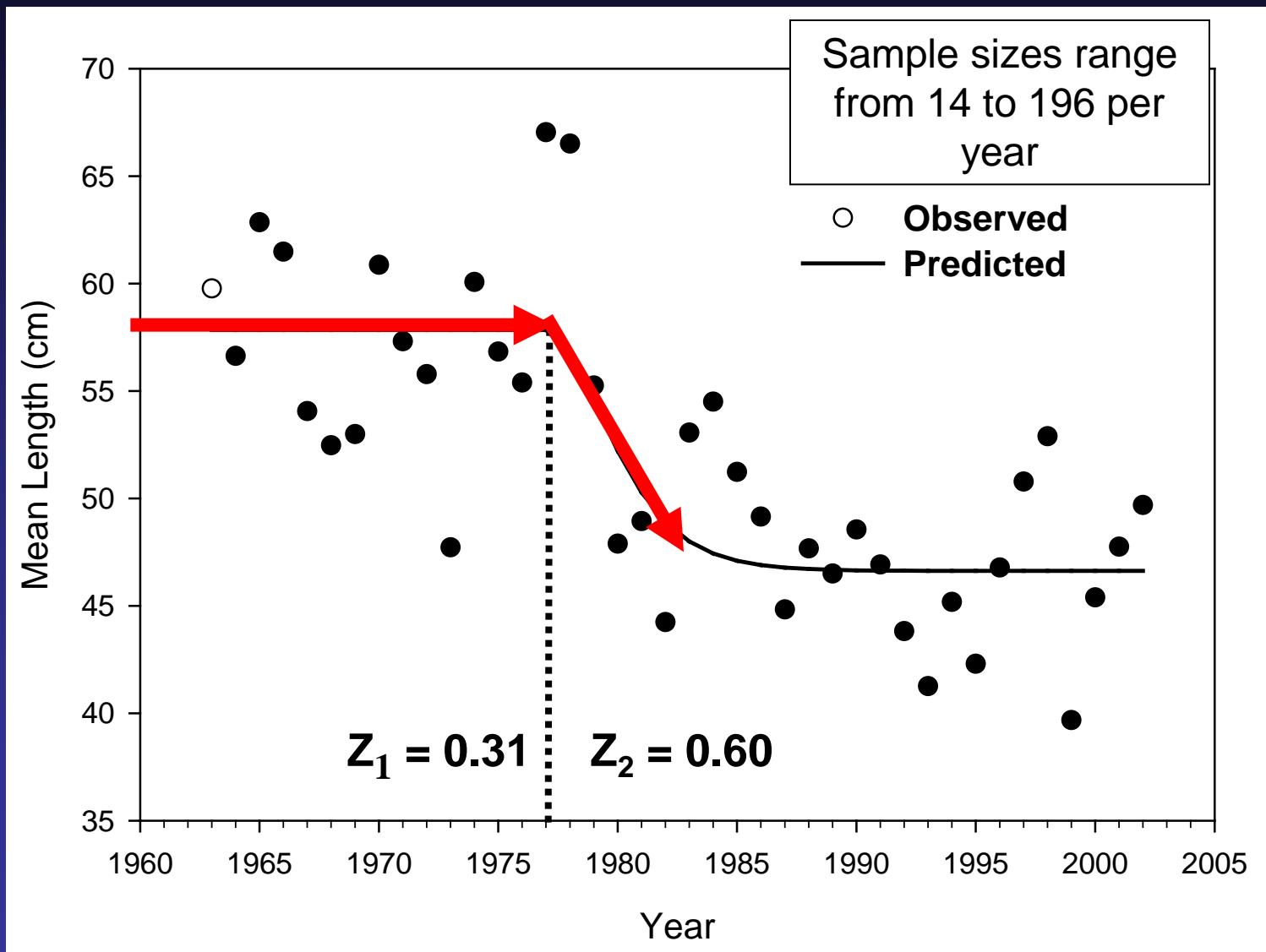


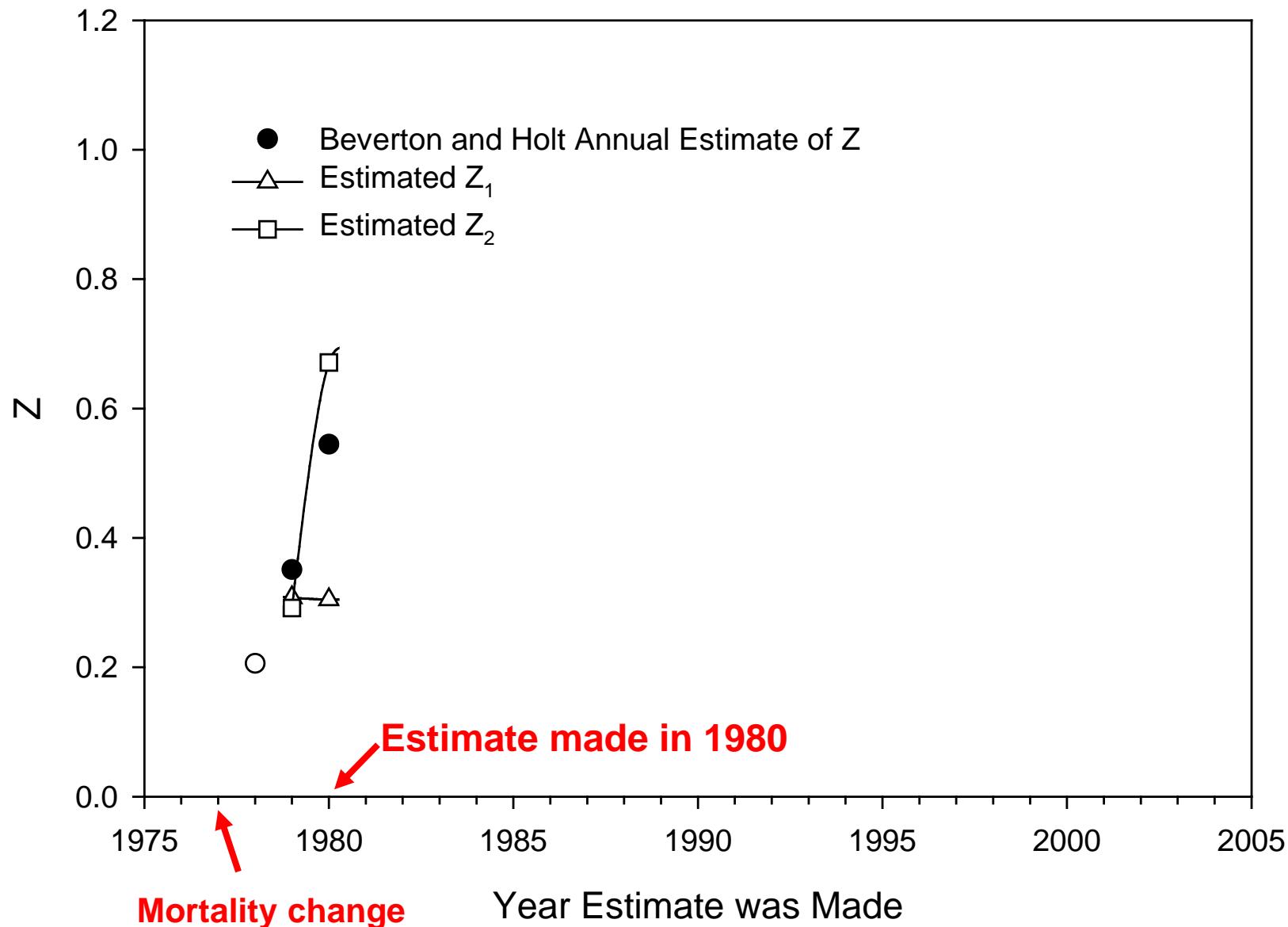


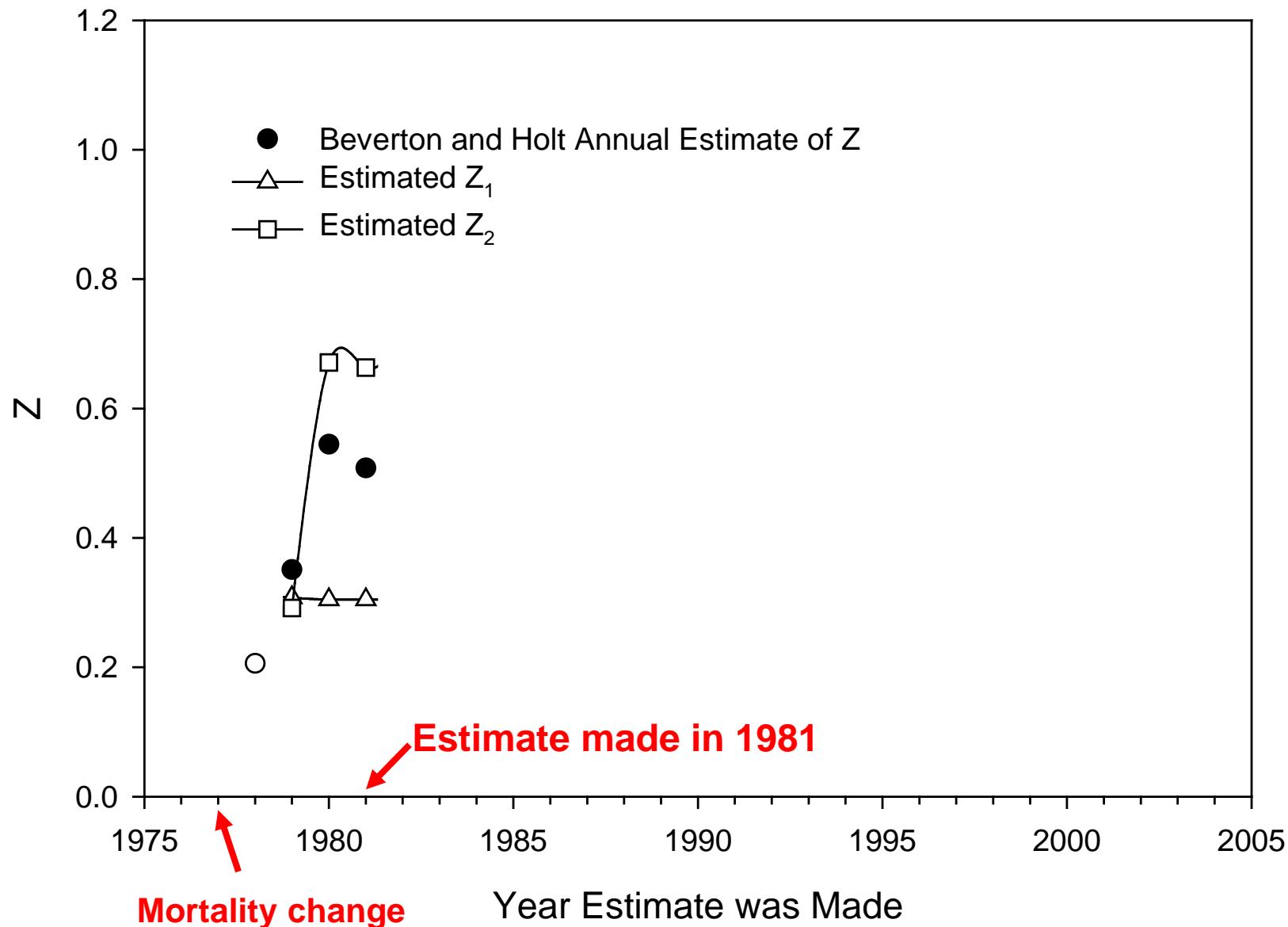
Mortality Estimates for Goosefish - Southern Management Region - NEFSC Fall Groundfish Survey

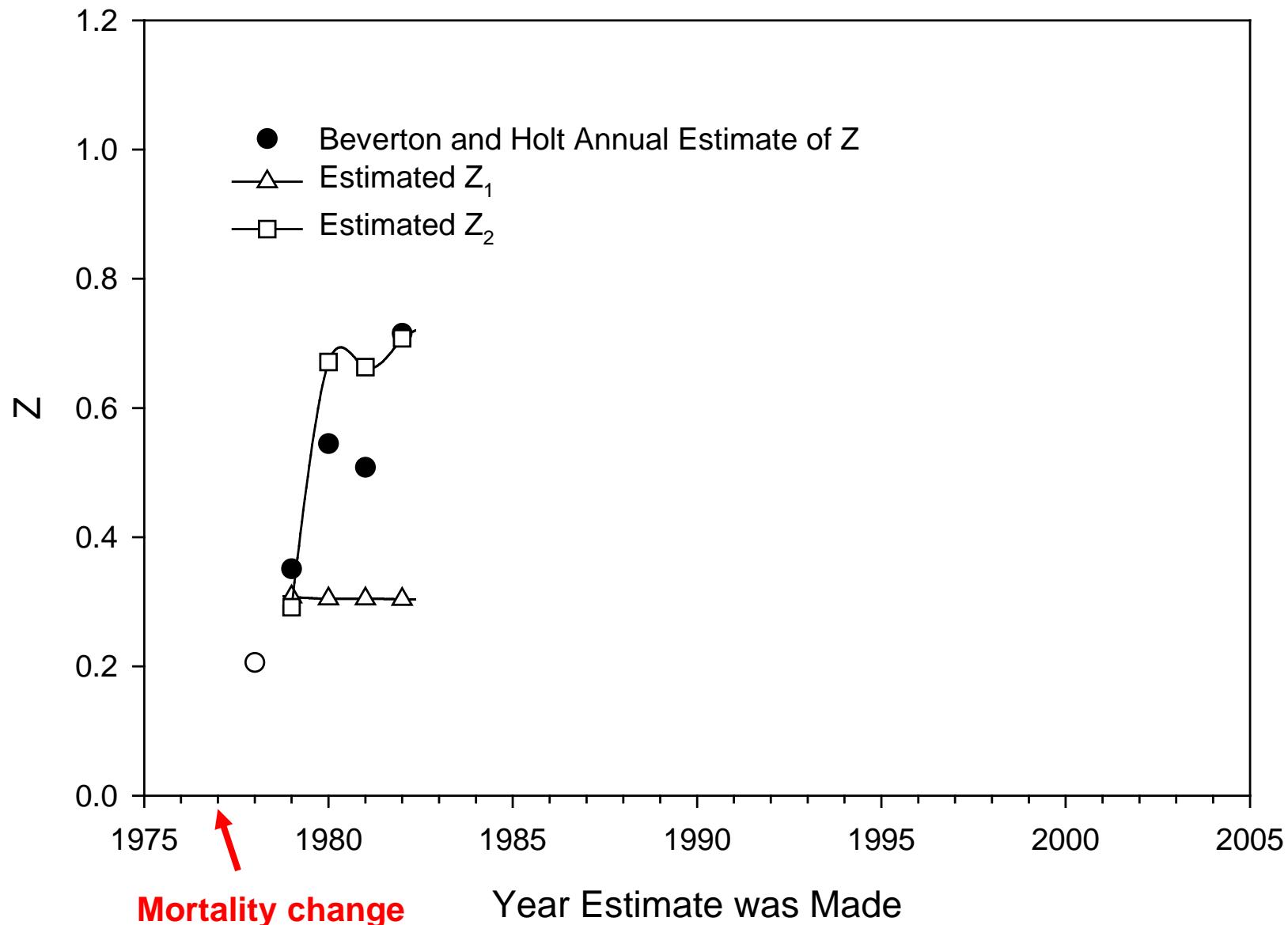


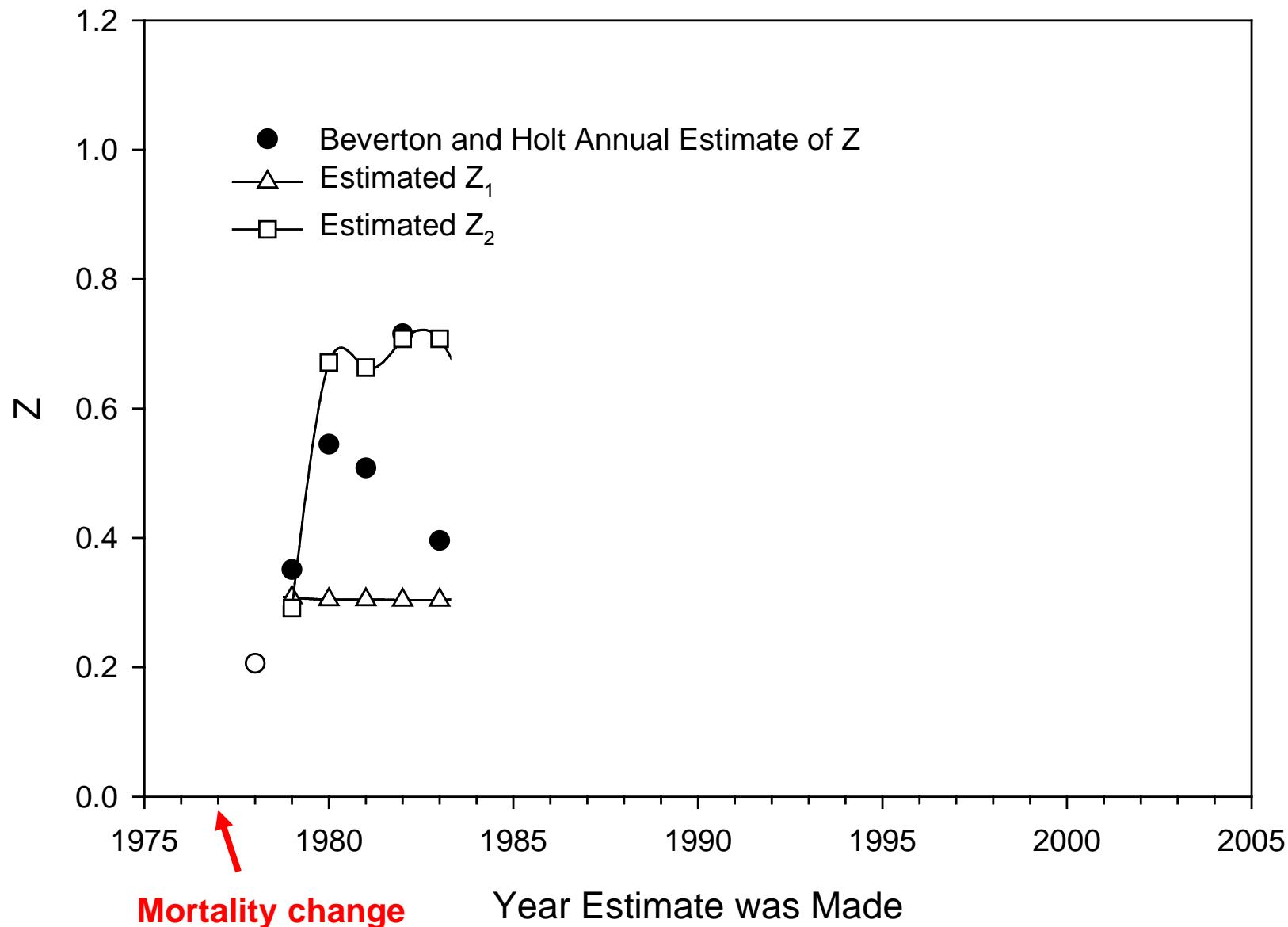
Mortality Estimates for Goosefish - Southern Management Region - NEFSC Fall Groundfish Survey

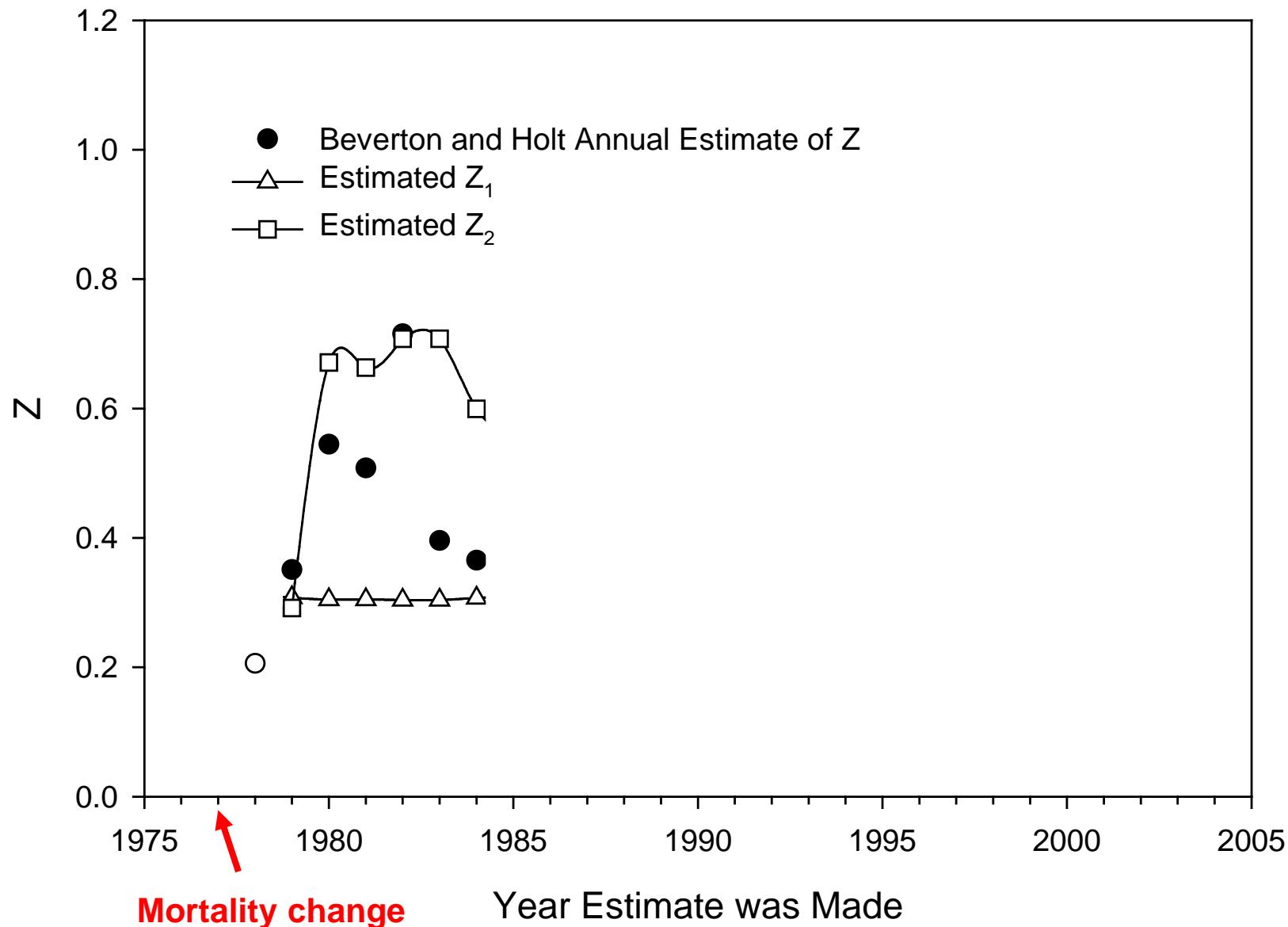


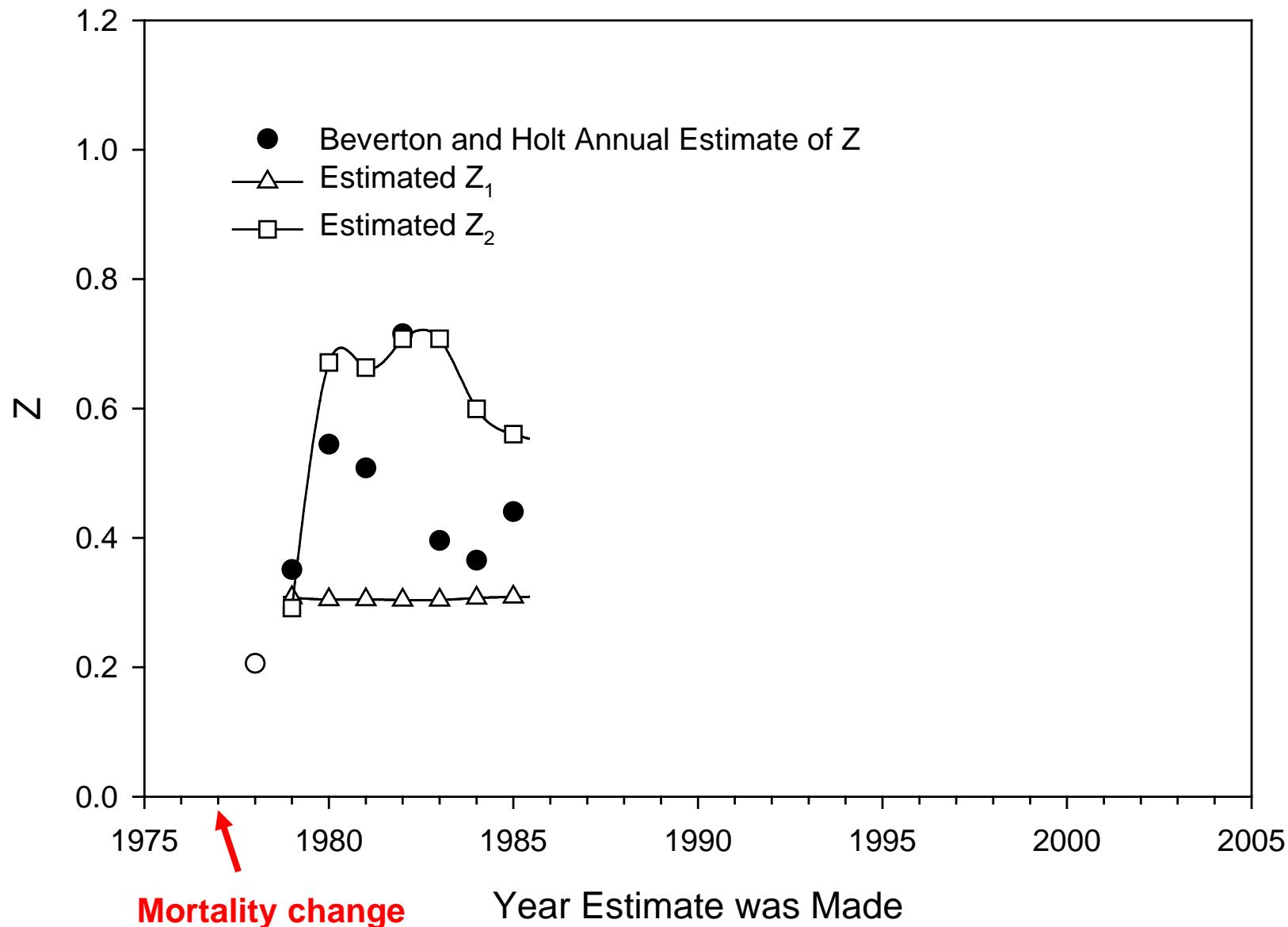


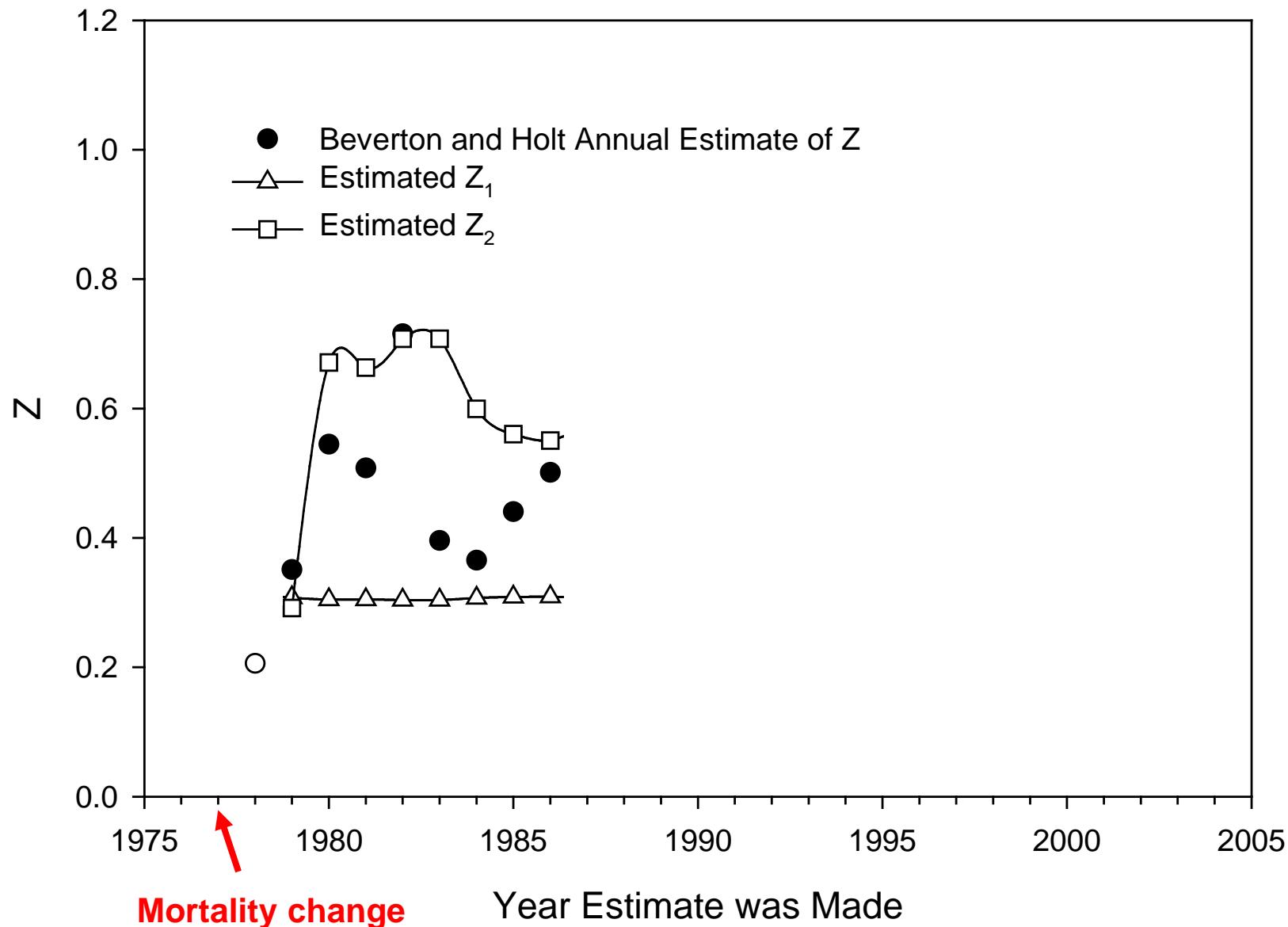


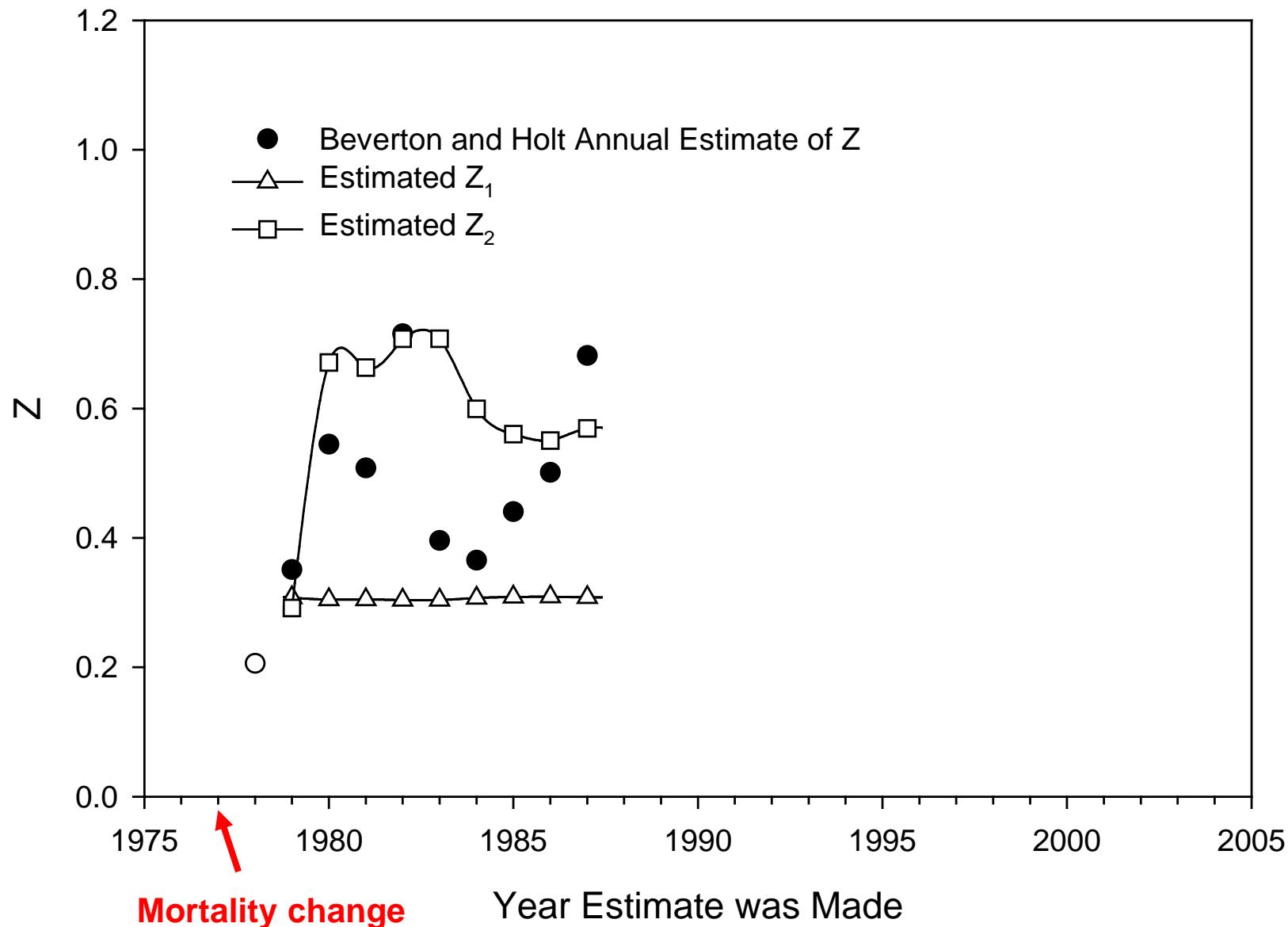


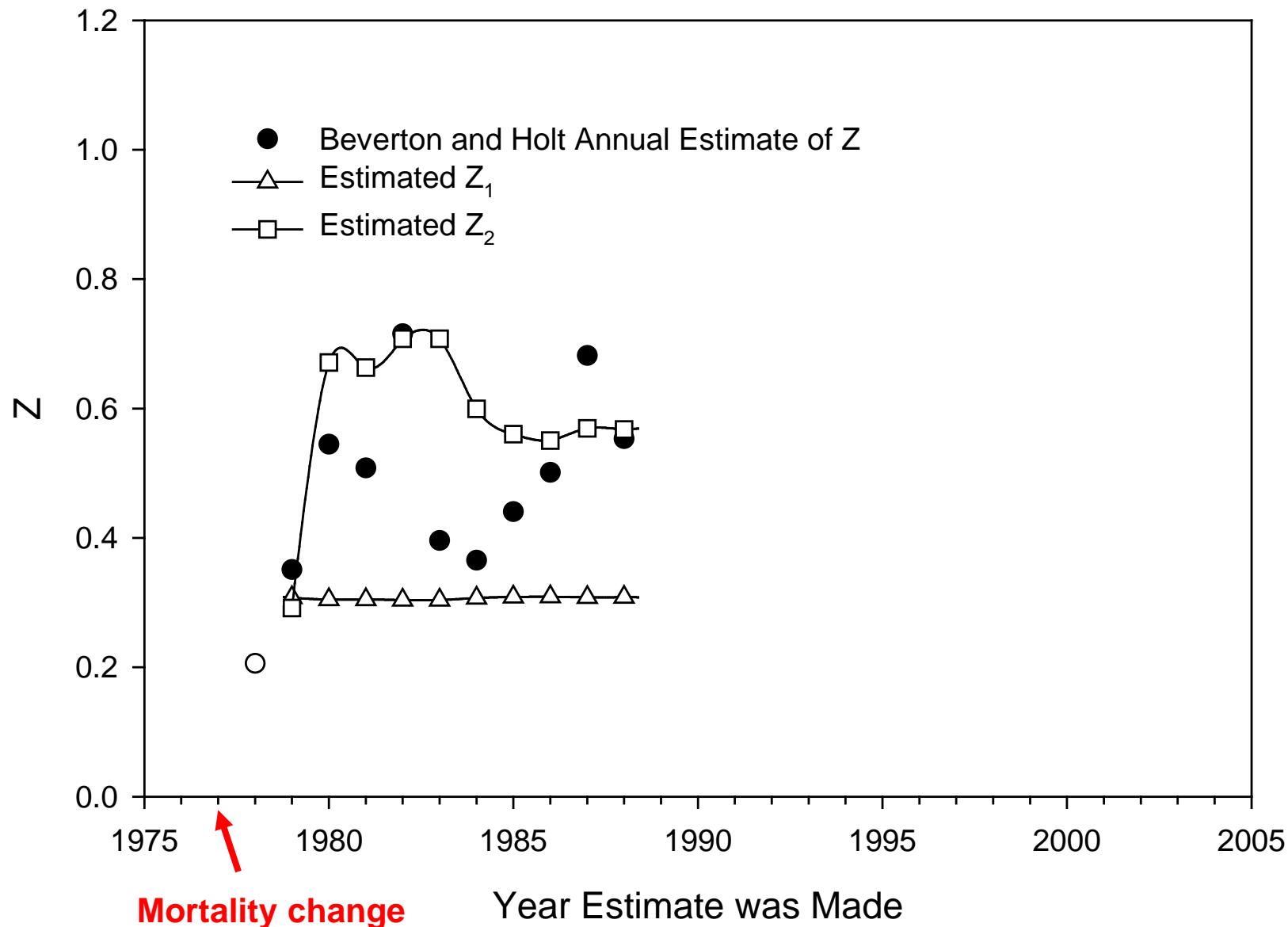


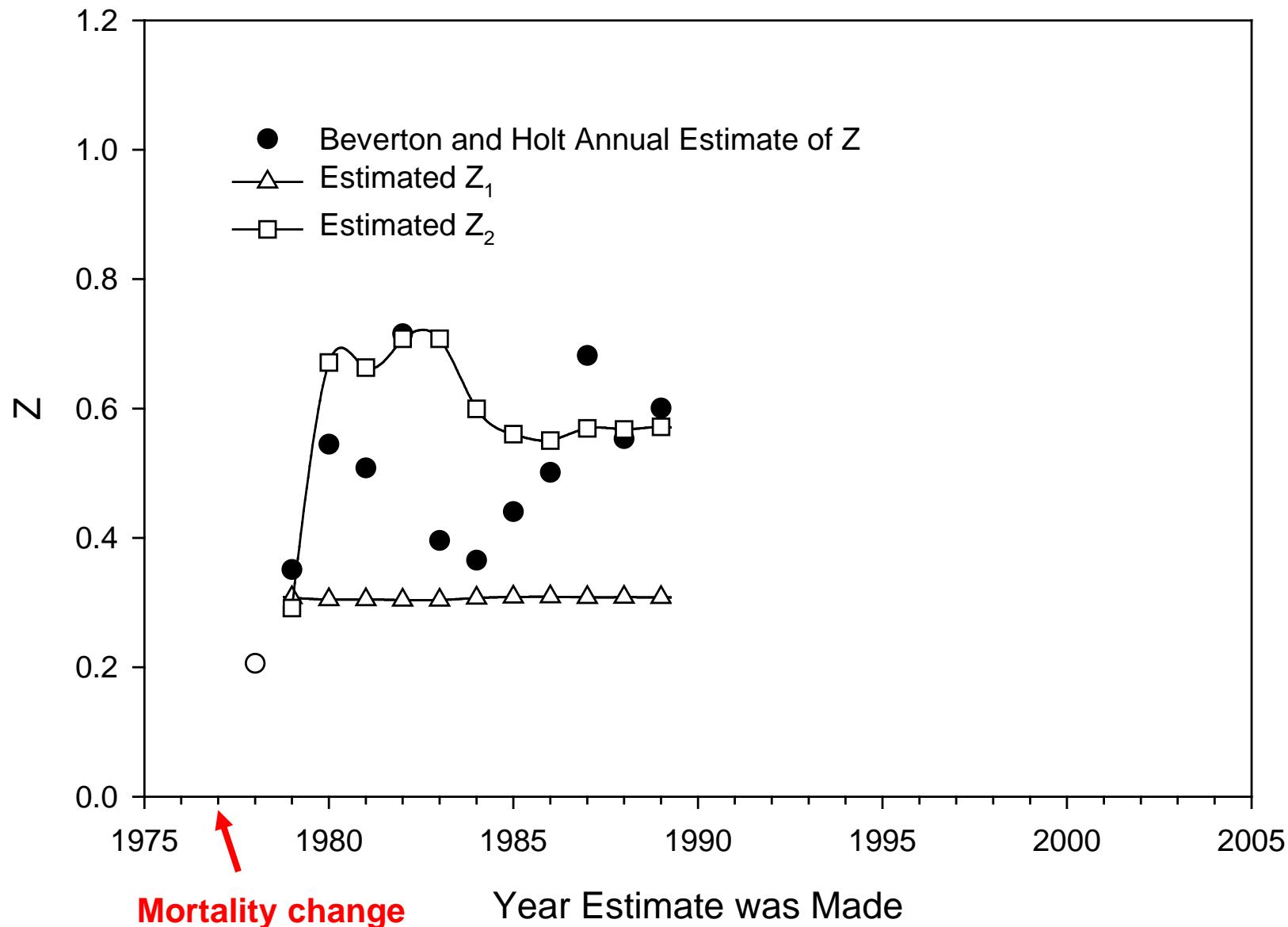


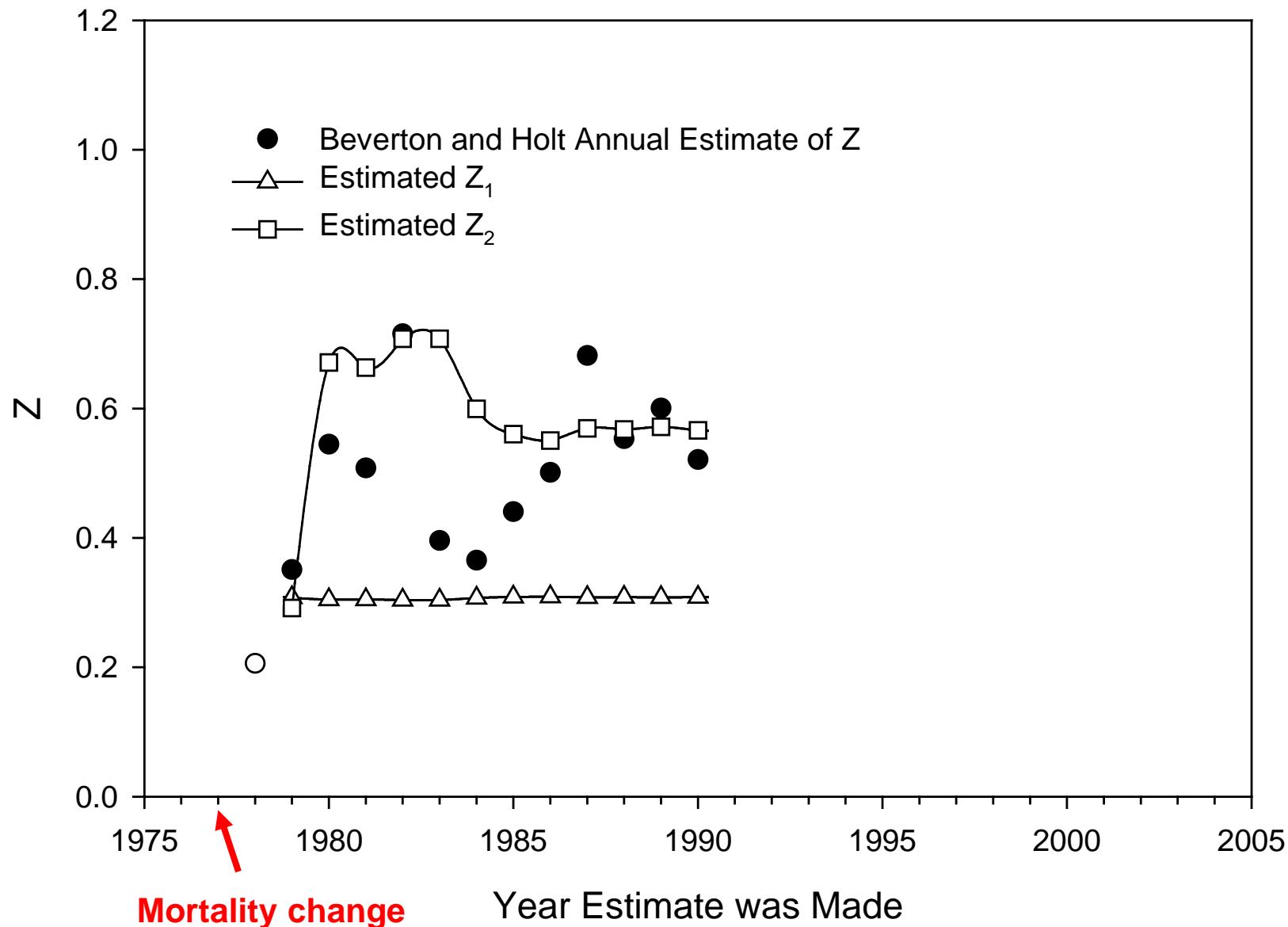


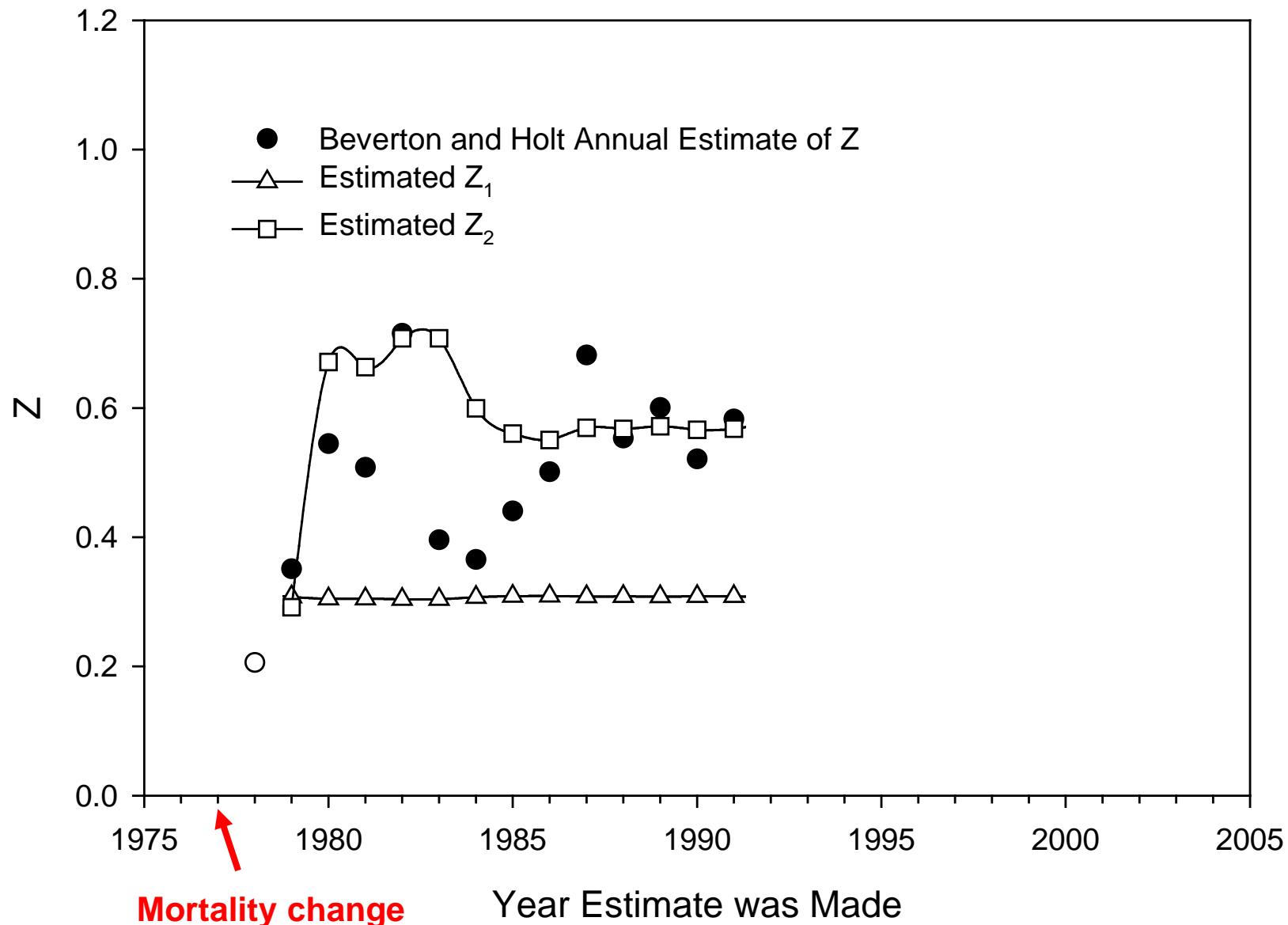


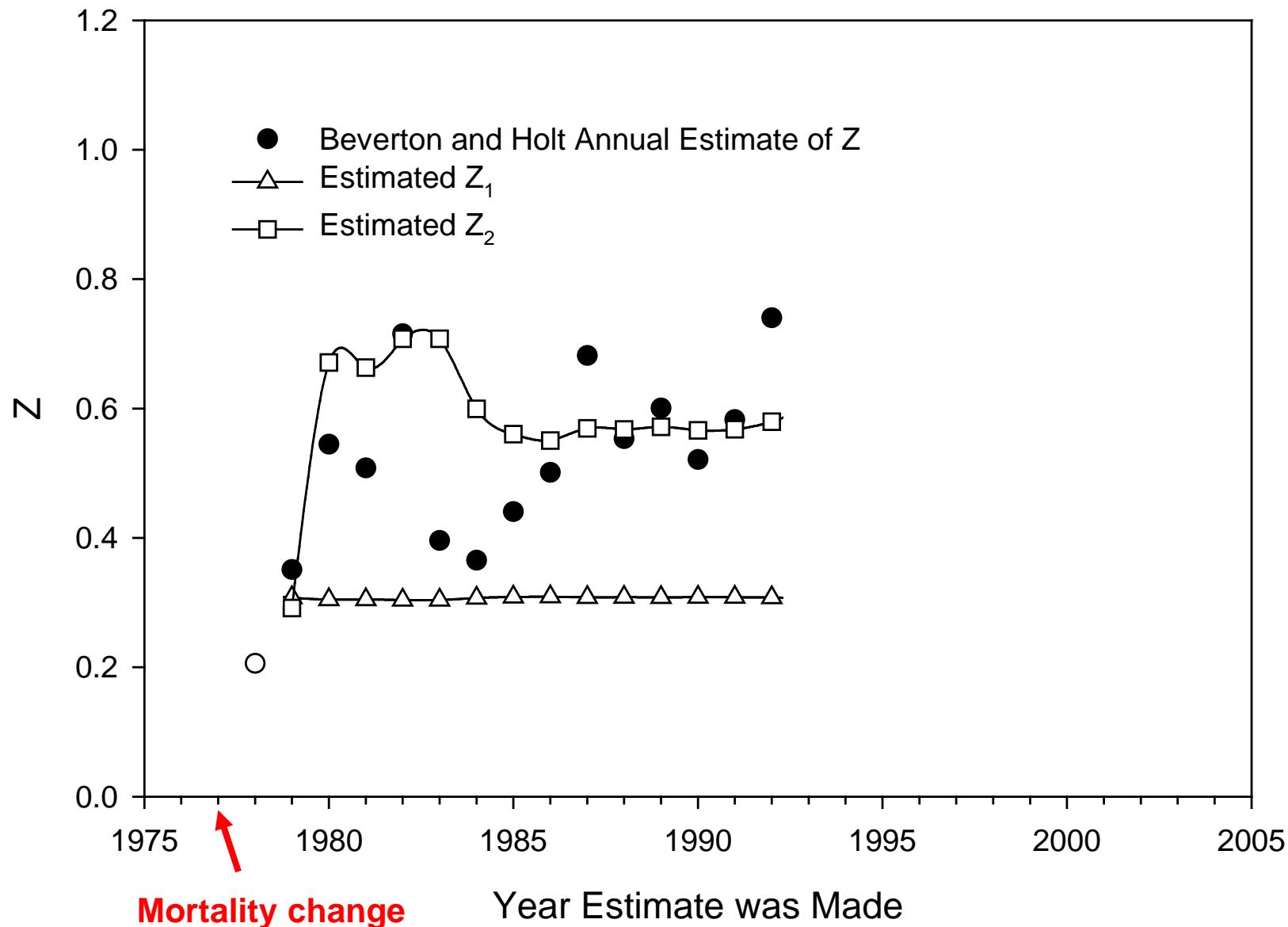


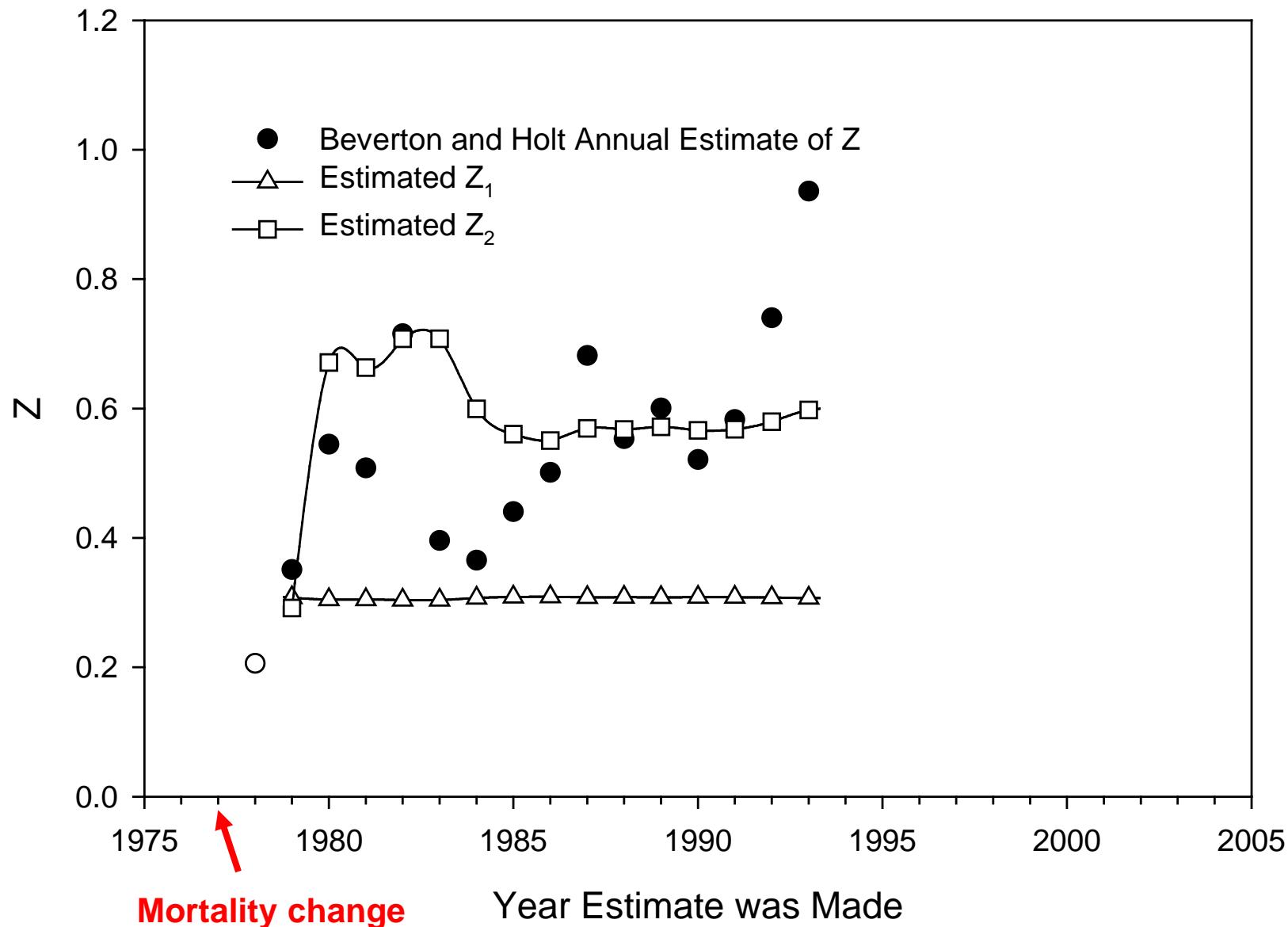


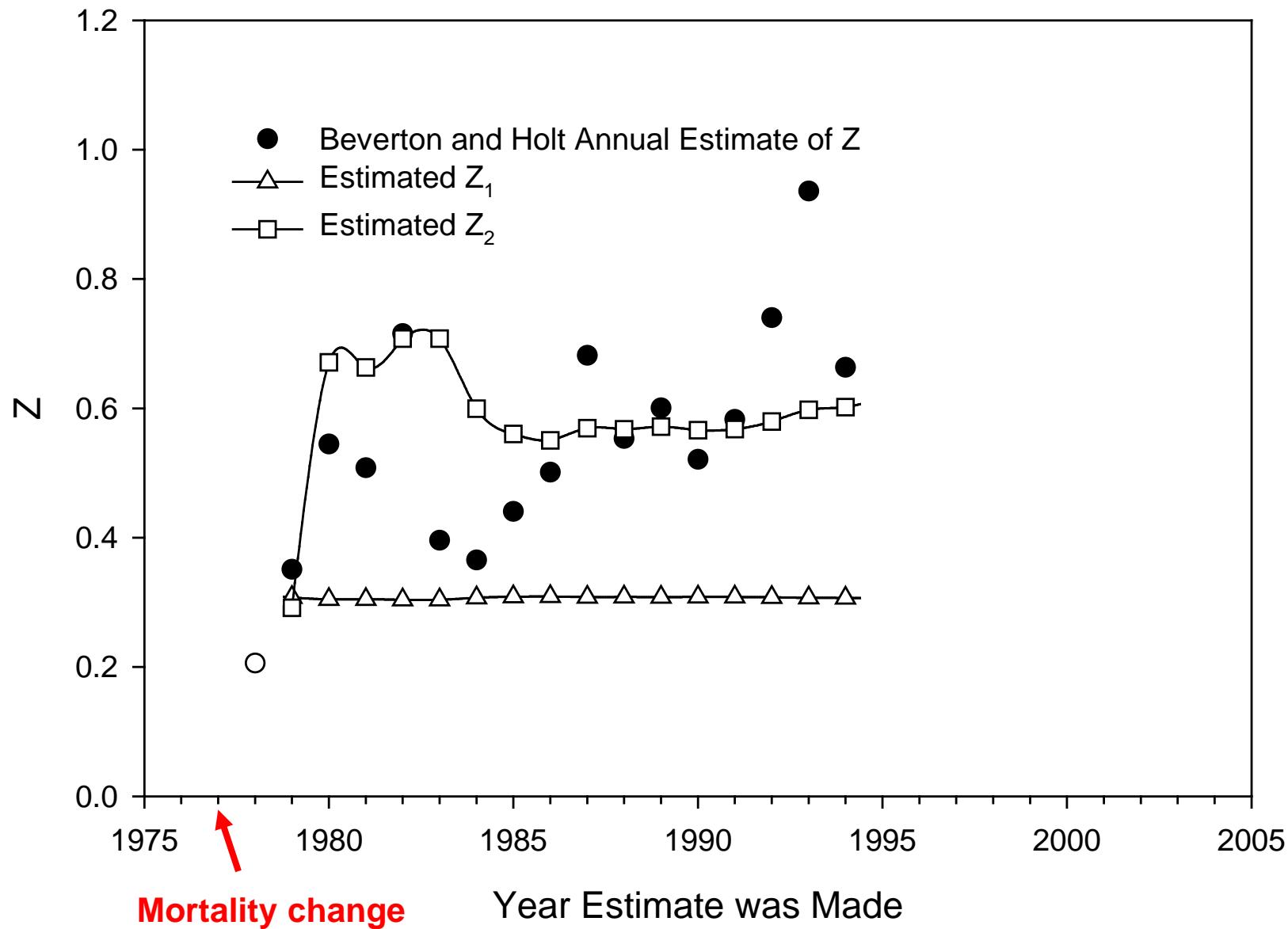


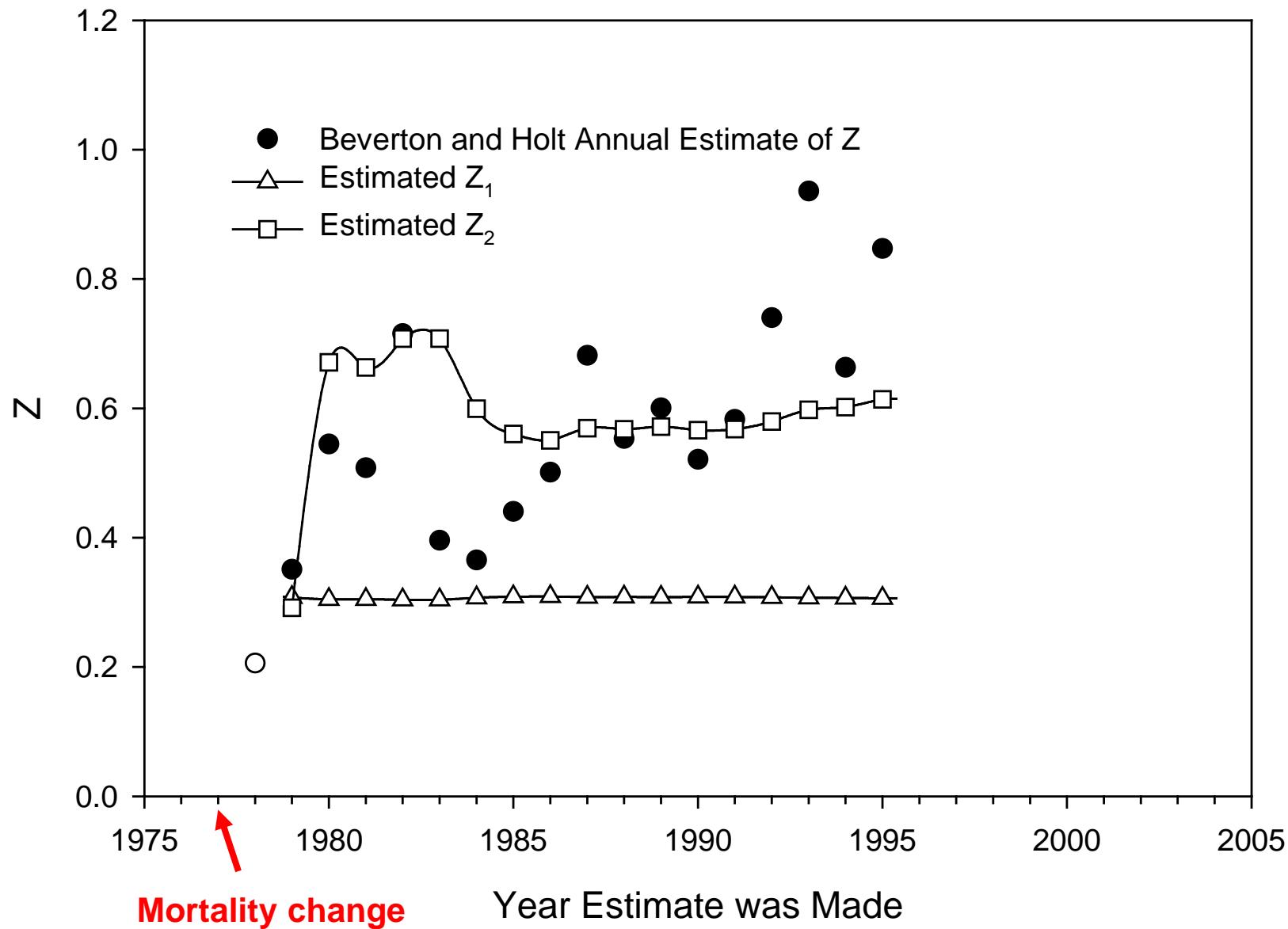


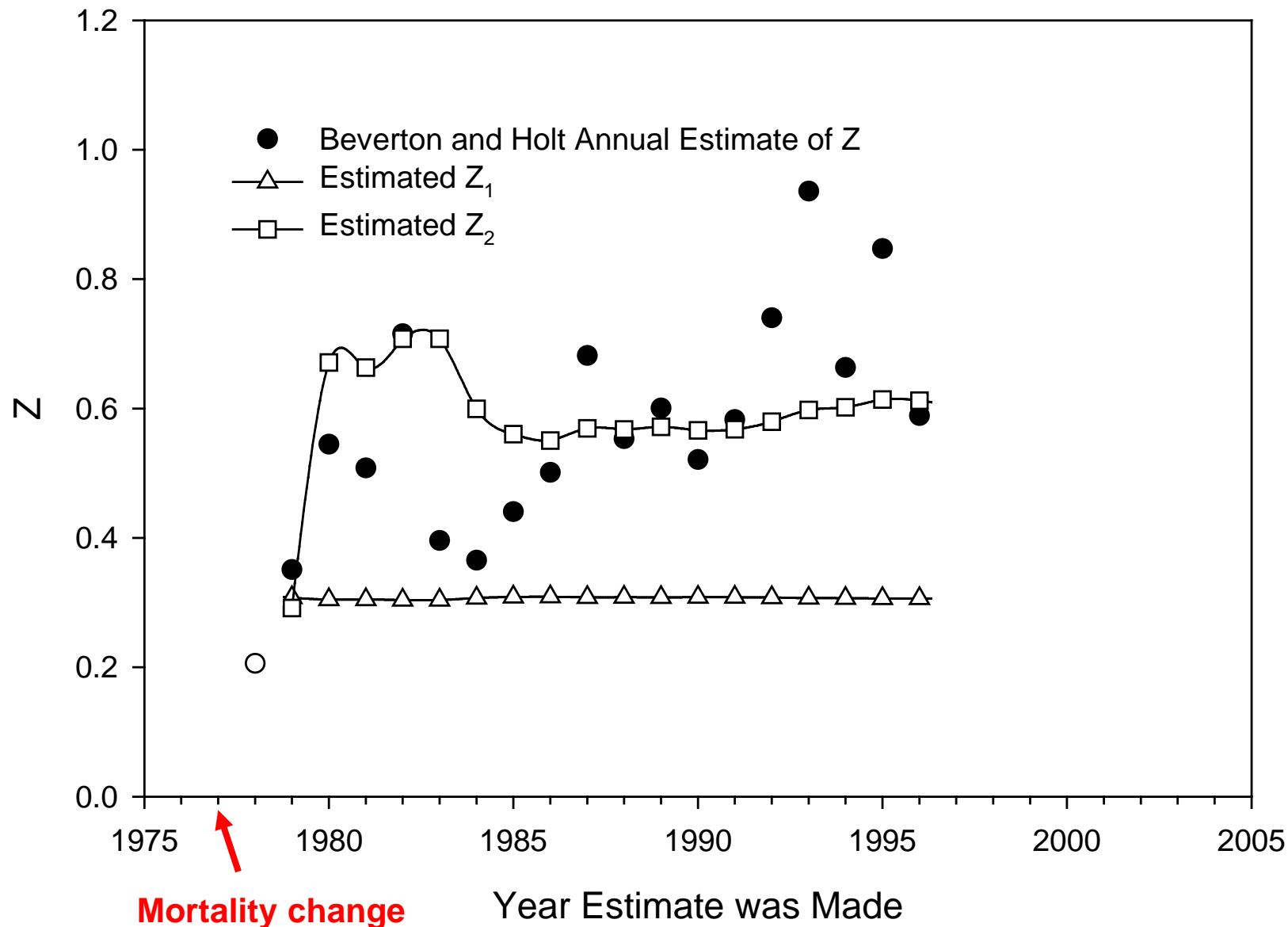


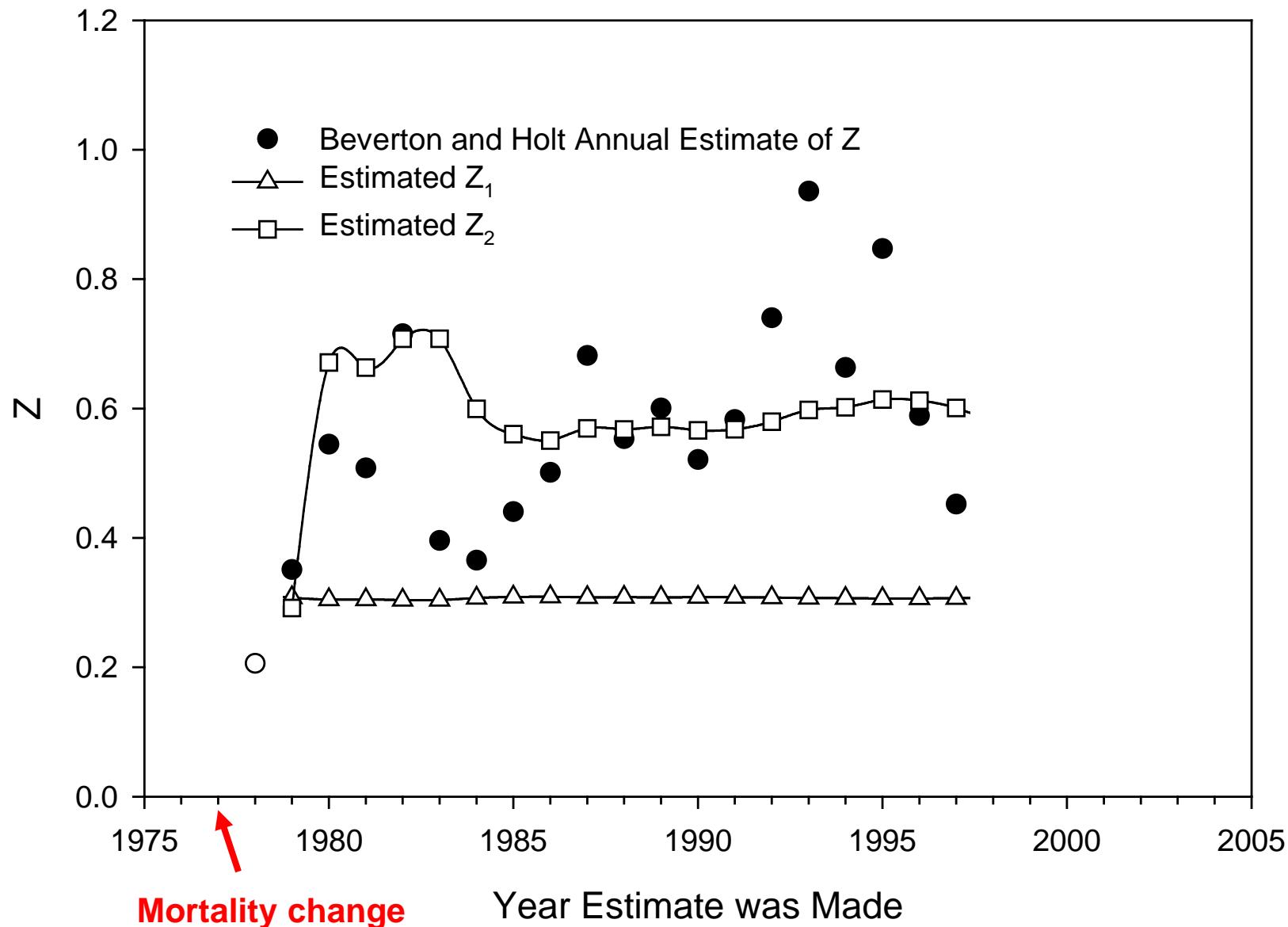


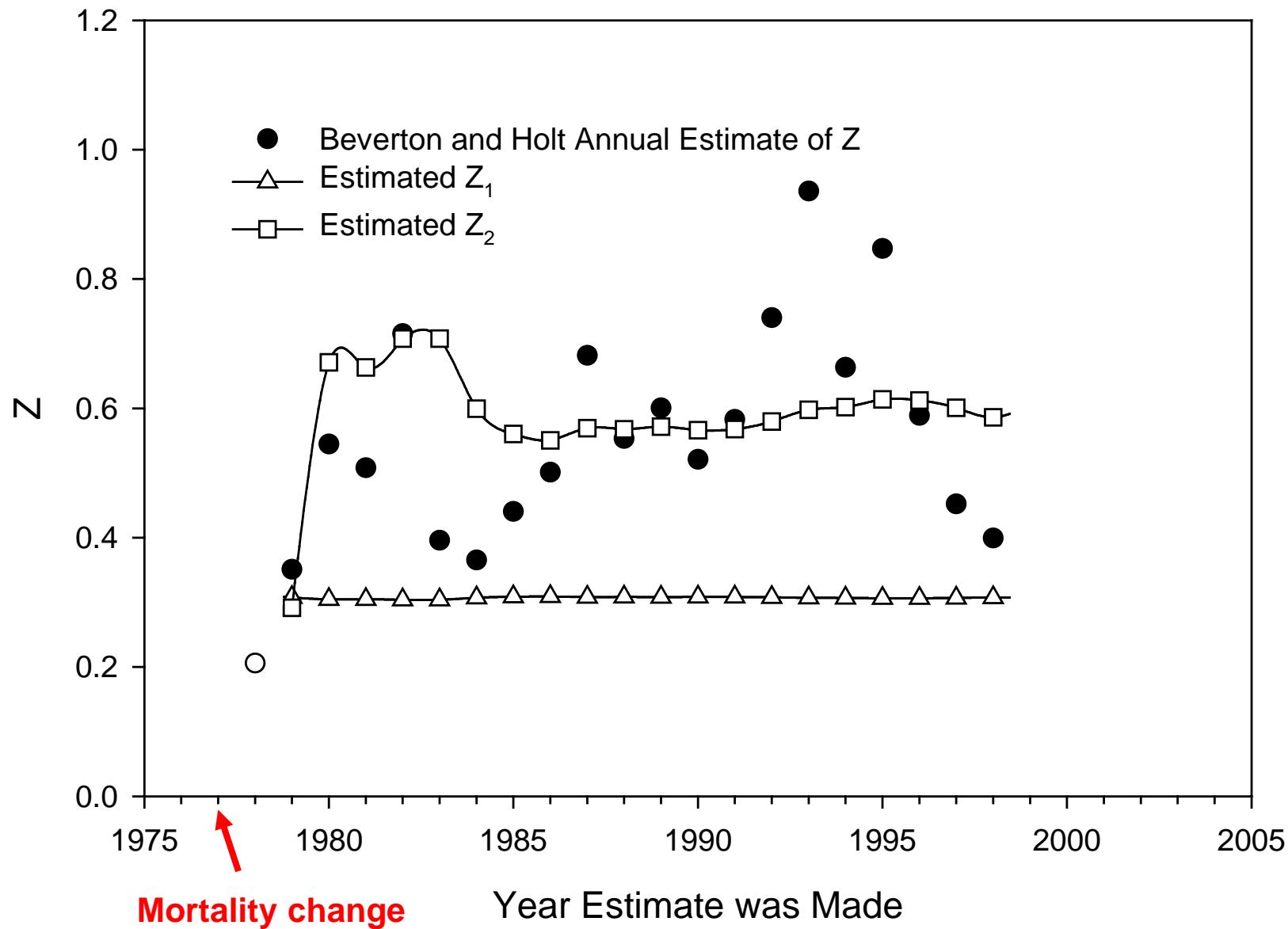


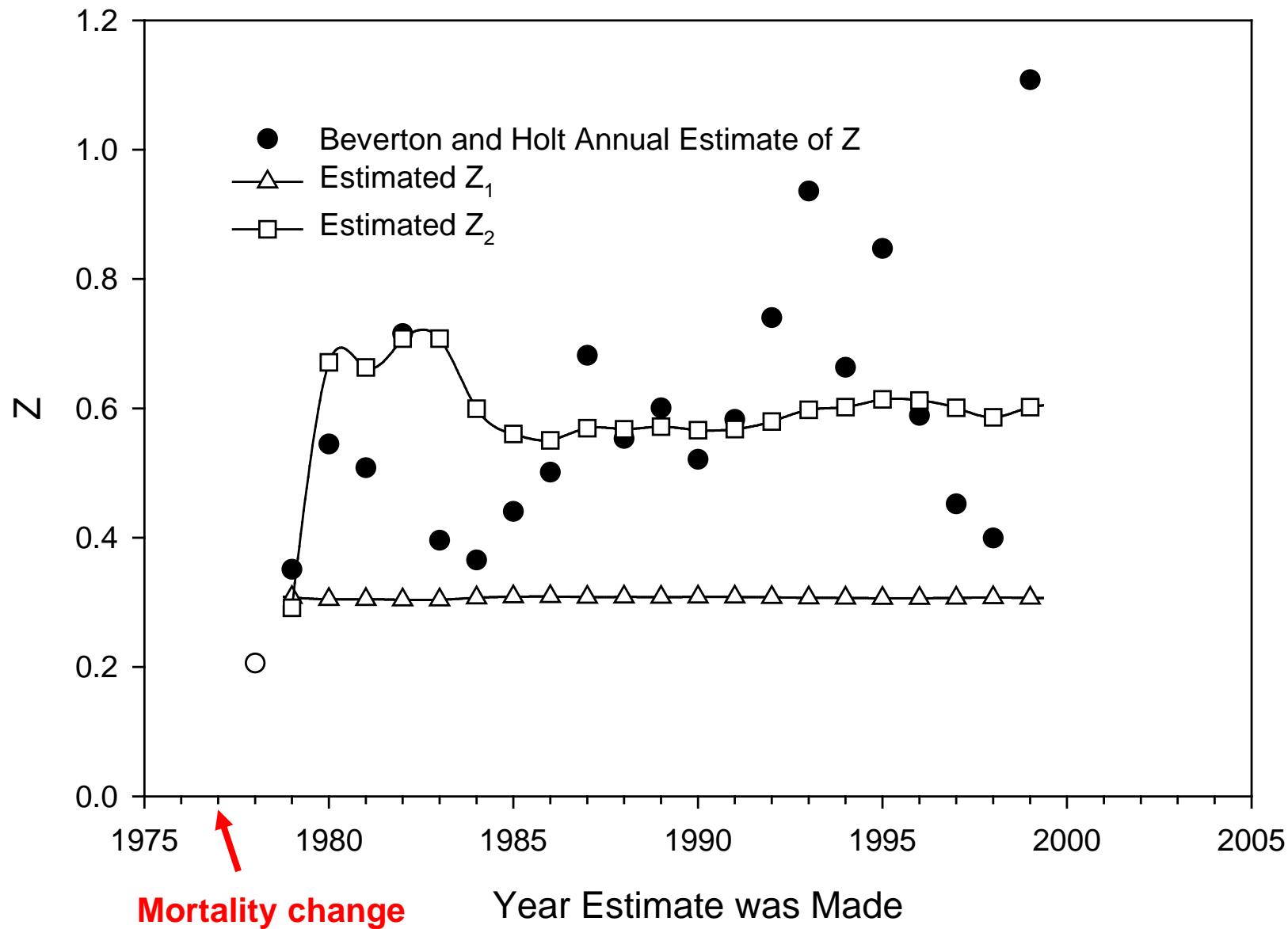


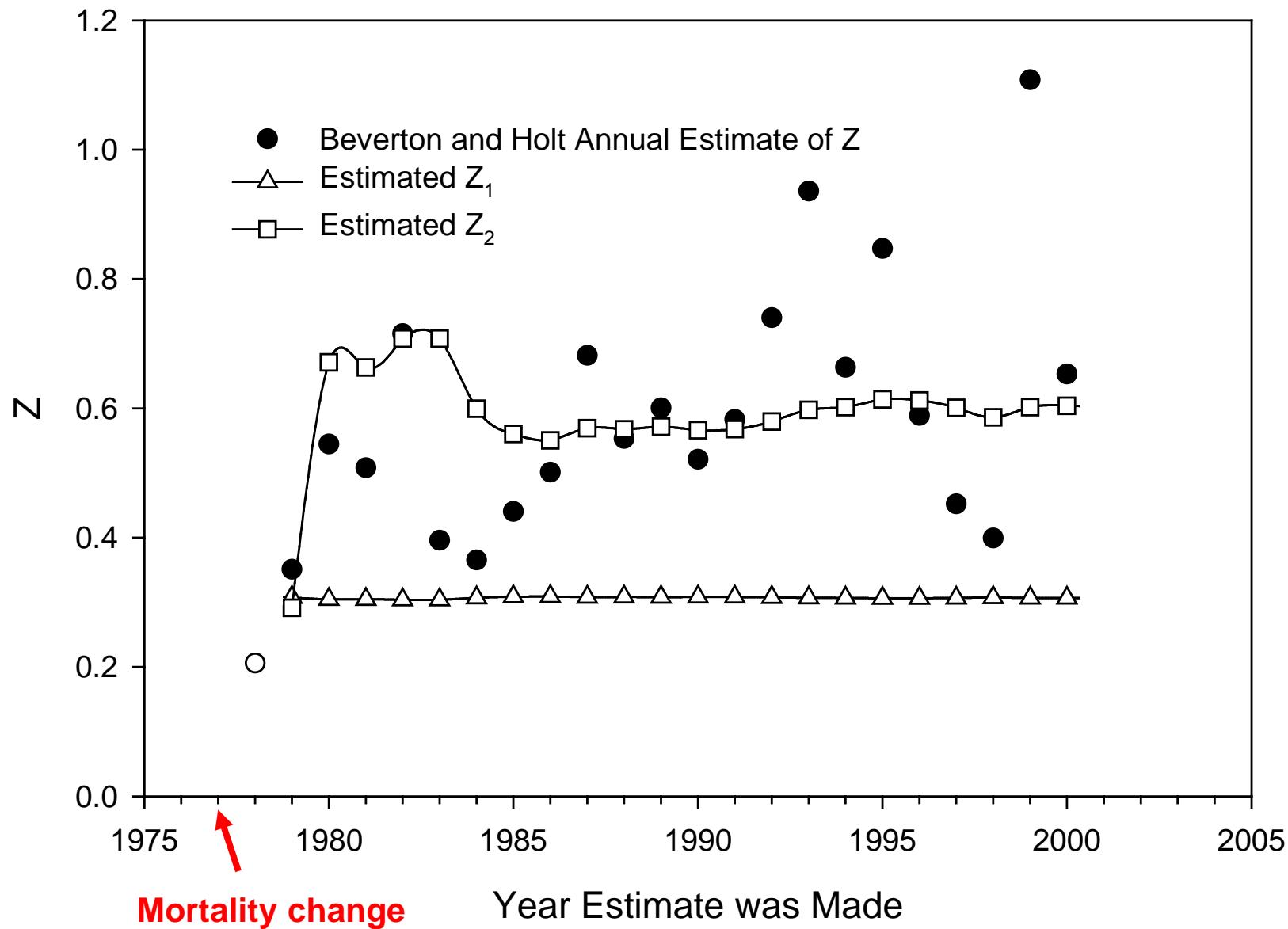


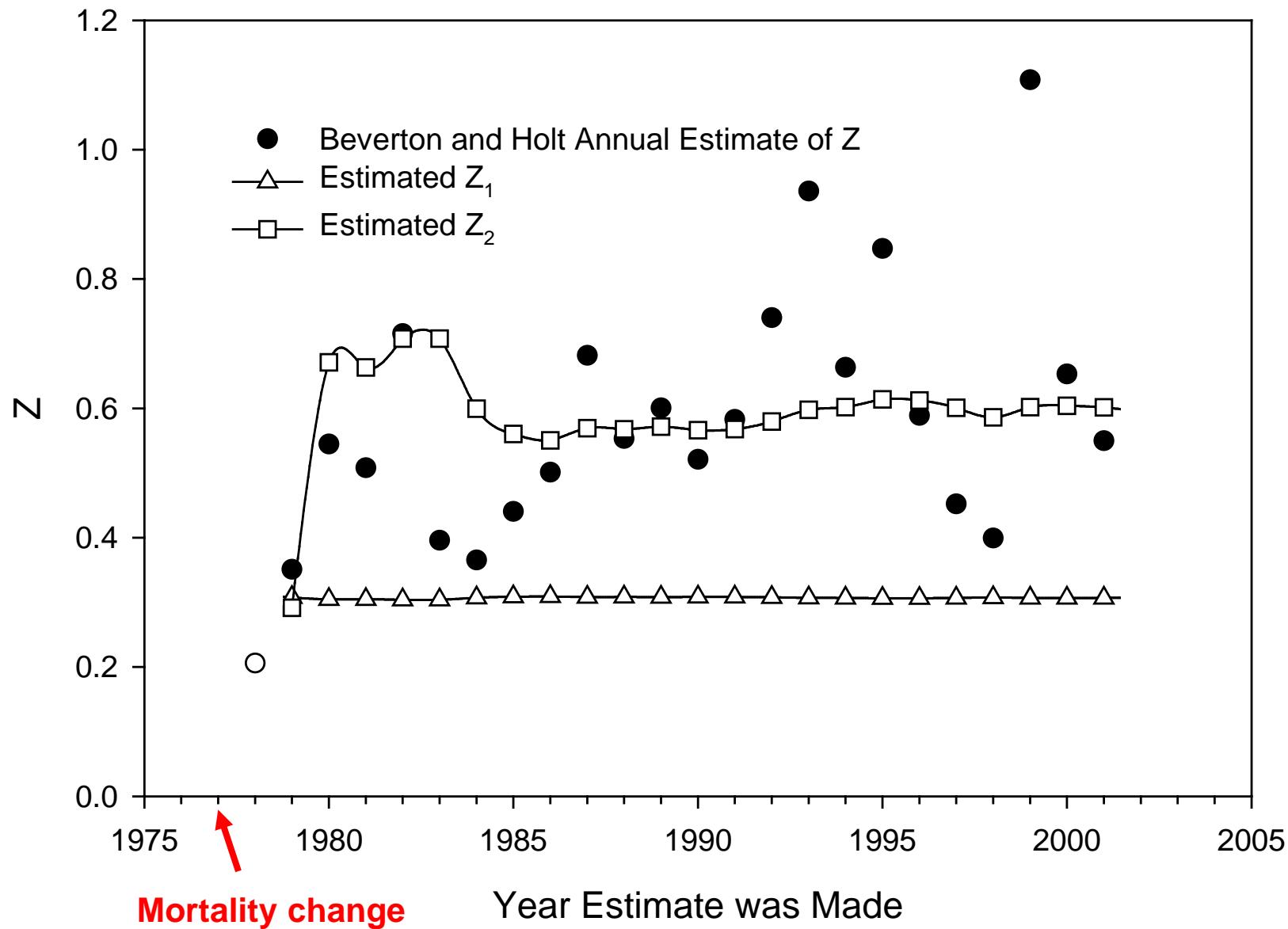


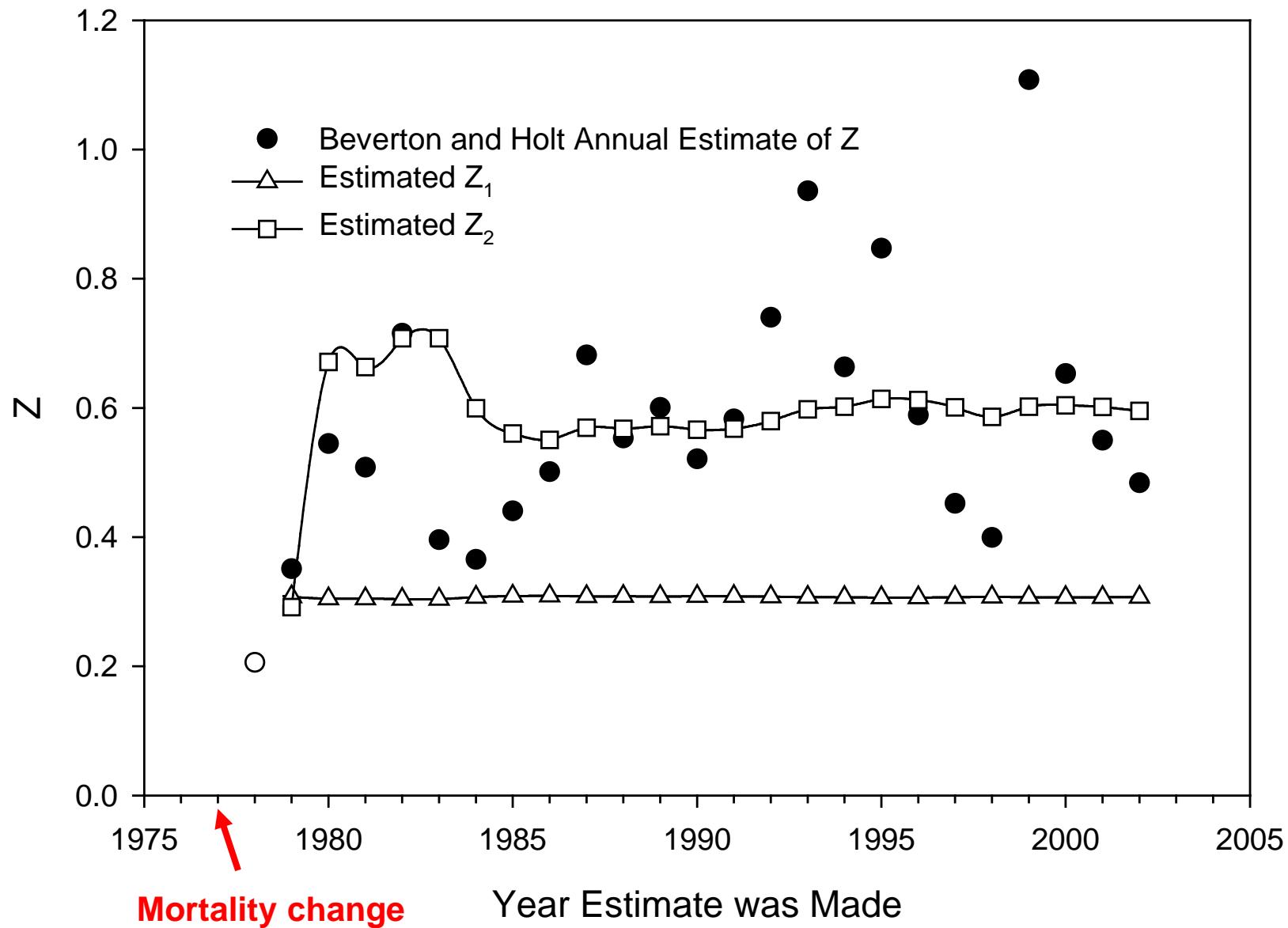


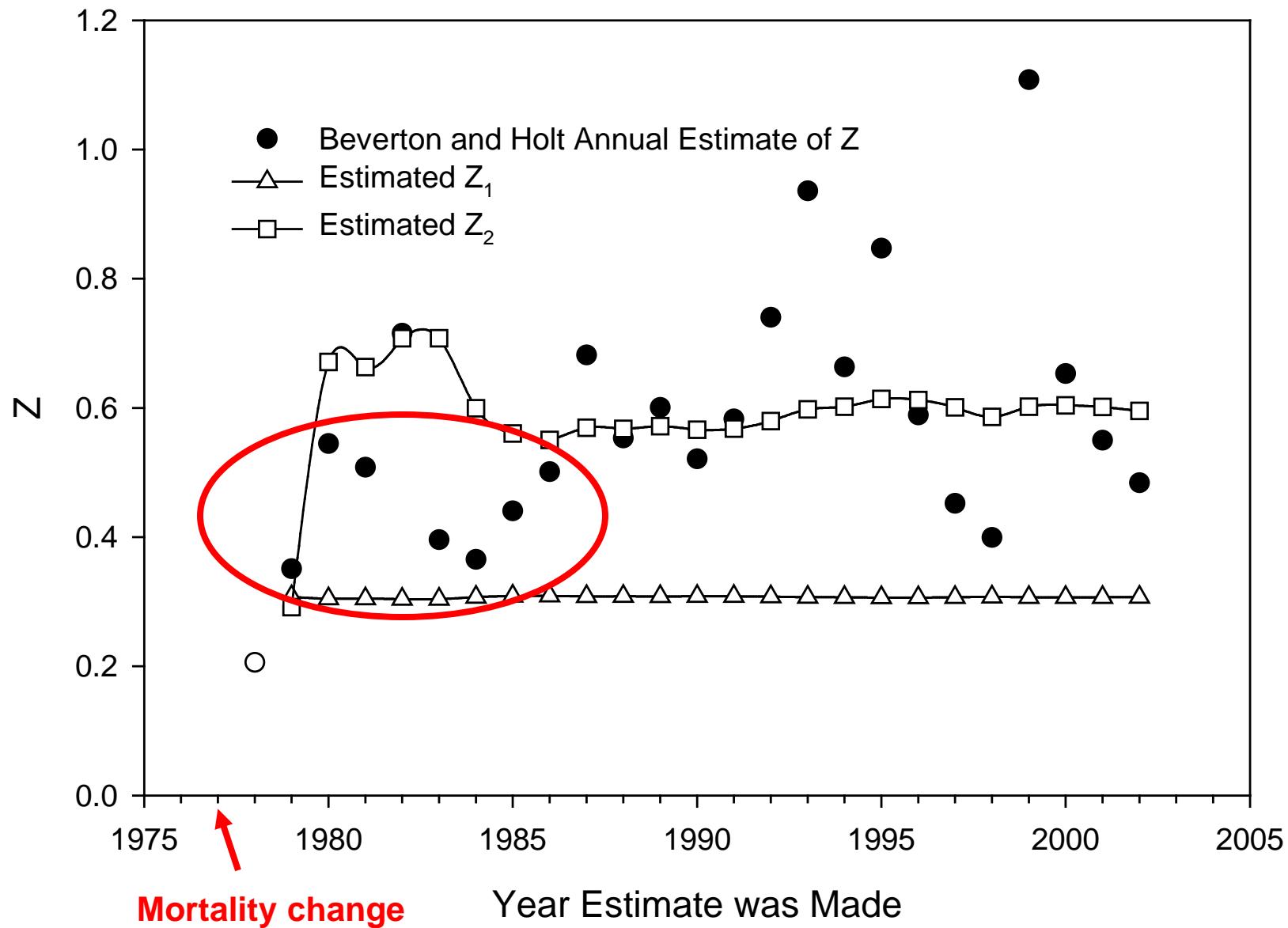




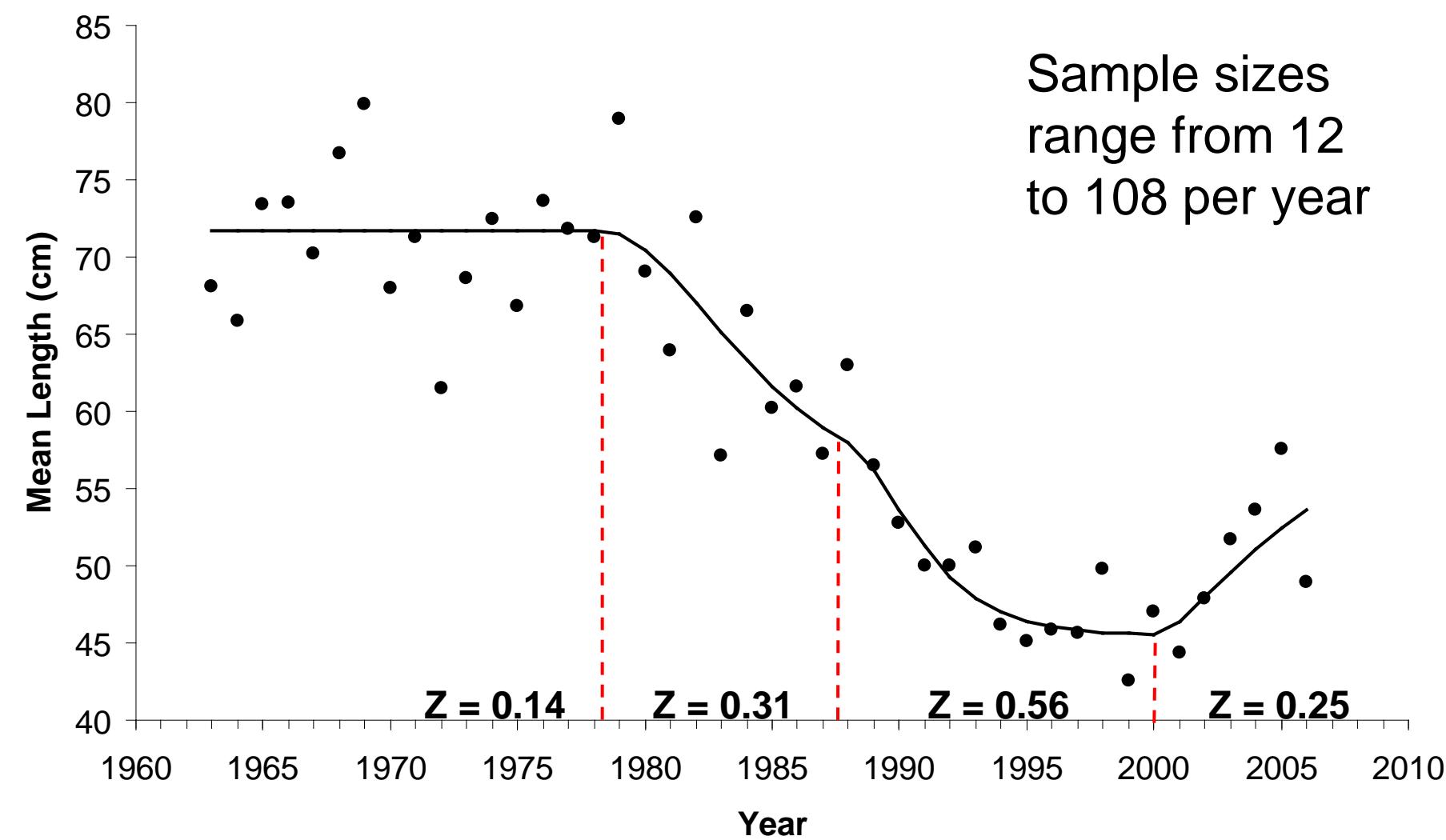








Goosefish Mortality Estimates--Northern Management Region NEFSC Fall Groundfish Survey





General Data

Mean Length

Growth

Optimization

Grid Search

NOAA's National Marine Fisheries Service



NOAA Fisheries Toolbox

Survival Estimation in Non-Equilibrium Situations

Version 1.3

Input File

None Selected

Model ID

First Year in Data

SET

Last Year in Data

Run Mode

 Single Run Grid Search

1/24/2009

1:20 PM

GUI Programming:
Alan Seaver

Moral Support:
Chris Legault

Programming Support:
Brian Linton

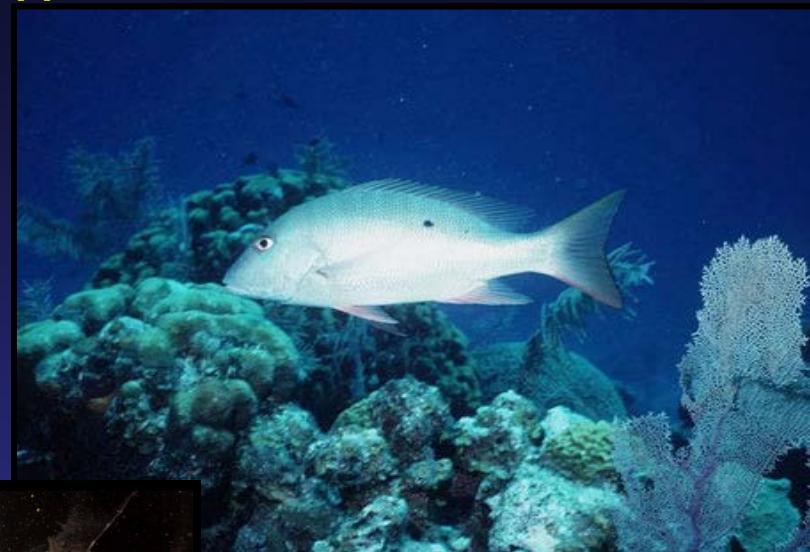
Variations of Non-equilibrium formulation

Estimating Mortality from Changes in Mean Lengths and Catch Rates in Non-equilibrium Conditions

Application to the Mutton Snapper Assessment



Lutjanus analis



Lutjanus analis



Lophius americanus

Mutton Snapper Photos reprinted from
<http://www.flmnh.ufl.edu/fish/gallery/descript/muttonsnapper/muttonsnapper.html>.

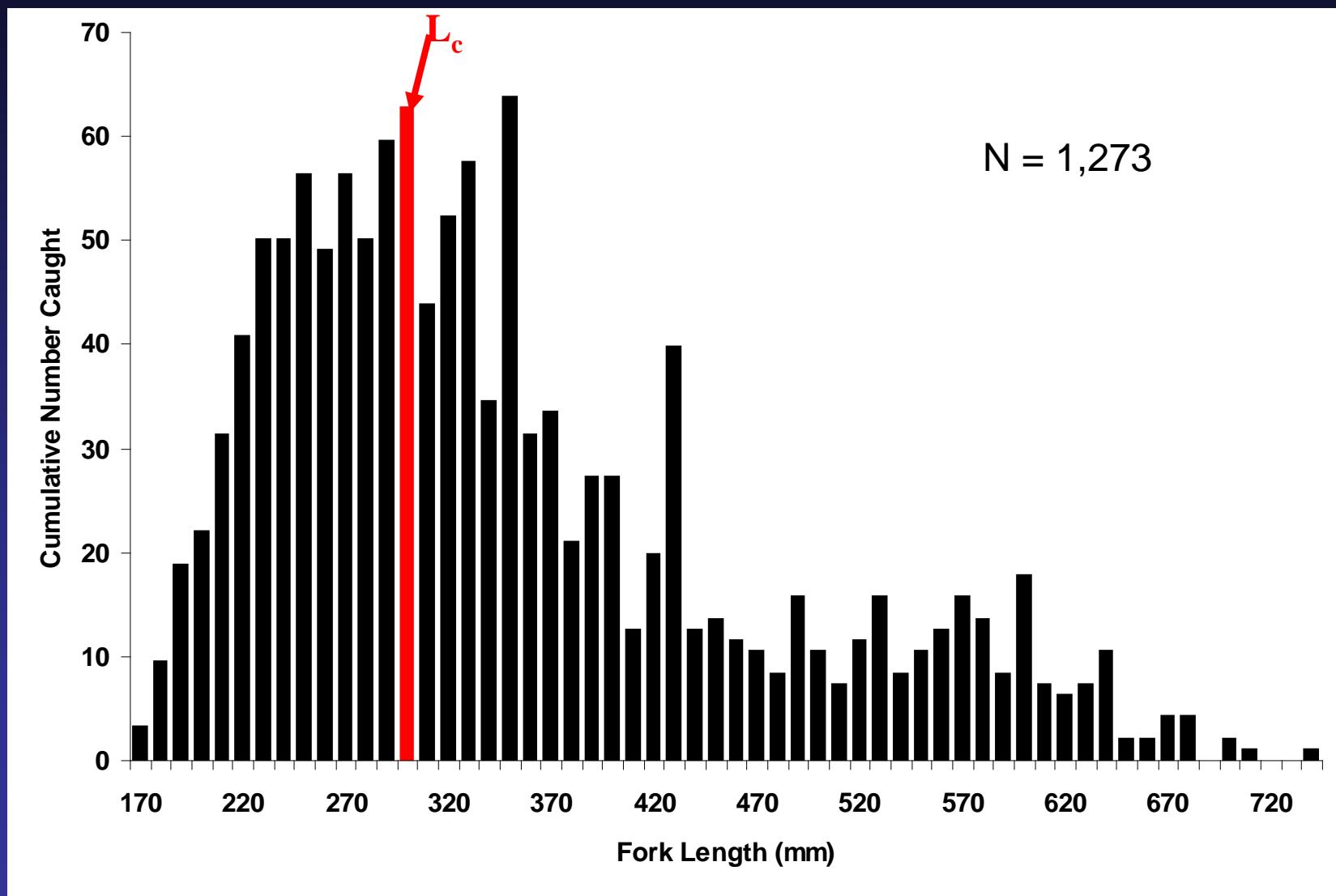
Todd Gedamke¹, Clay E. Porch¹ and John Hoenig²

¹National Marine Fisheries Service, Southeast Fisheries Science Center, Miami, Florida

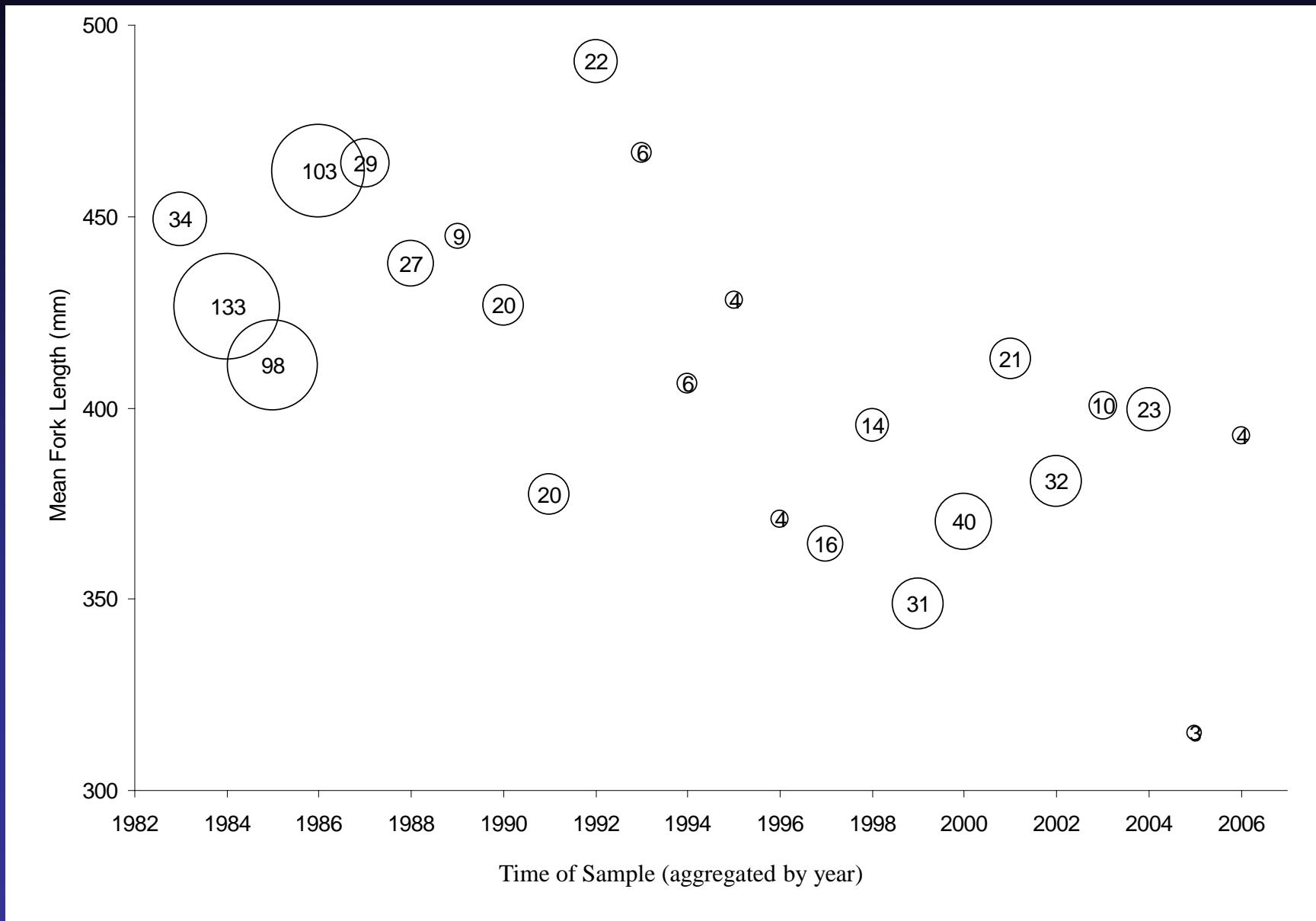
²Virginia Institute of Marine Science, Gloucester Point, Virginia



Cumulative plot of all individuals in the trap fishery of Puerto Rico (1983– 2006).

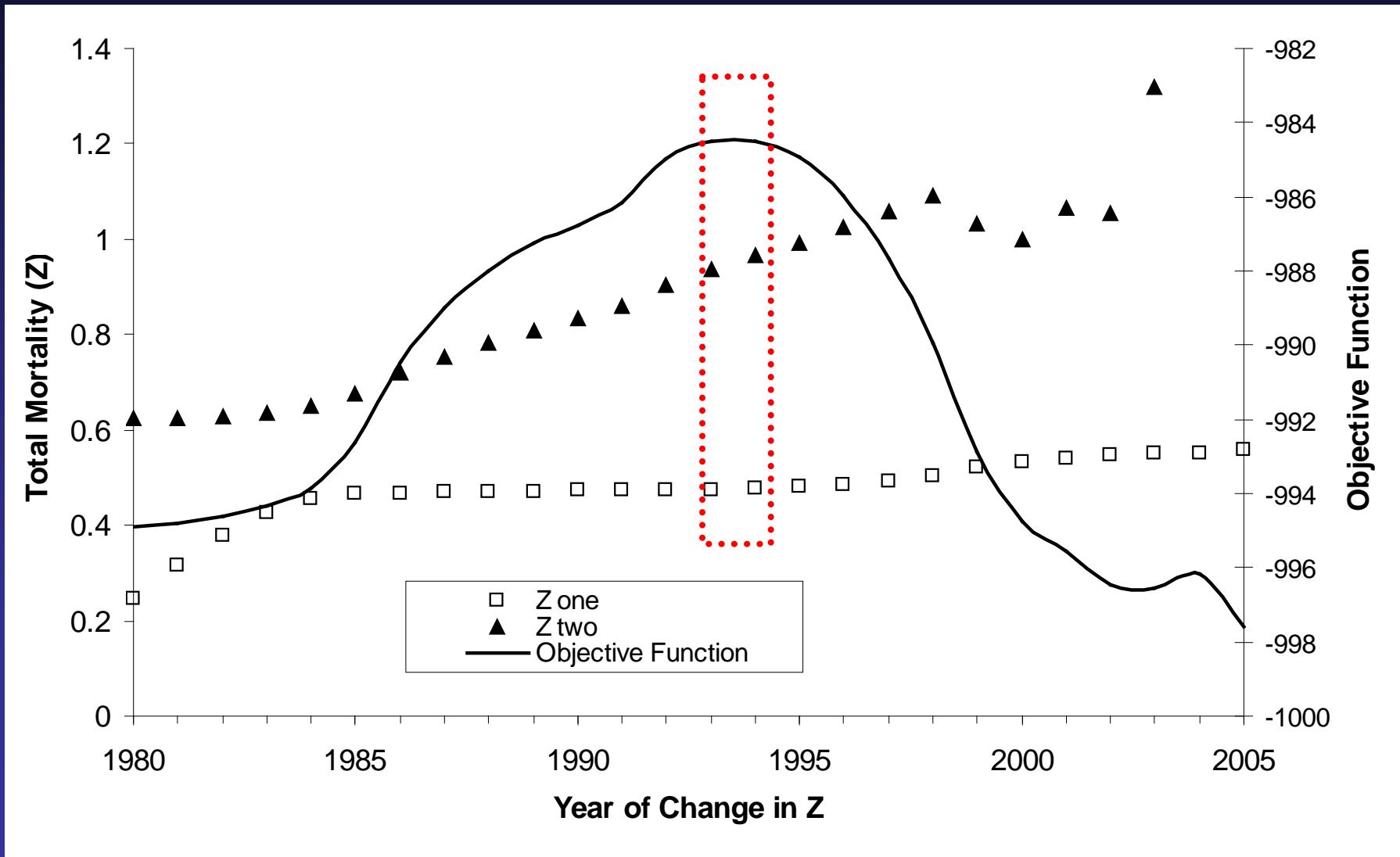


Mean Length calculated by year. Sample numbers for each year have been indicated by both bubble size and number.

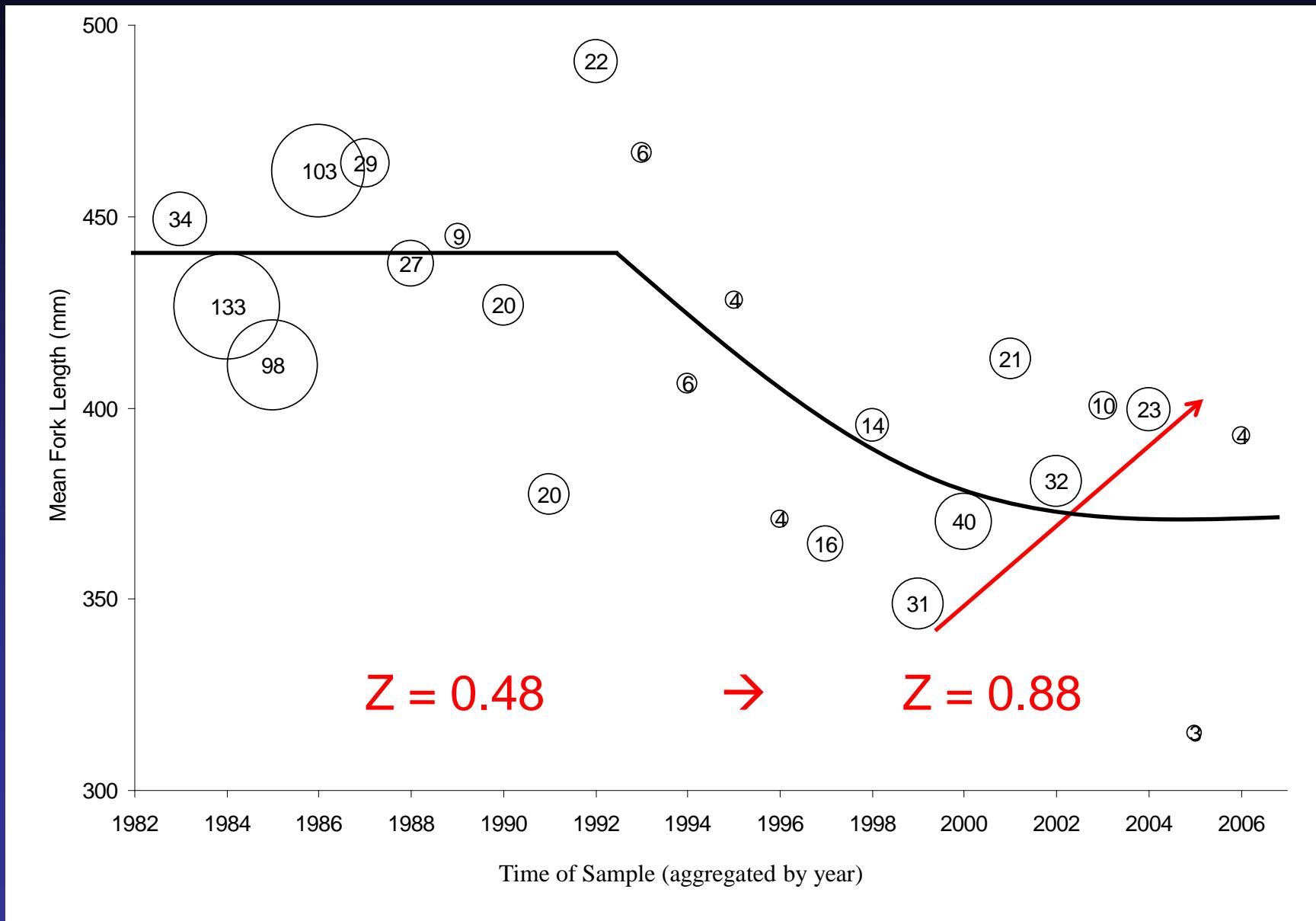


Results of grid search. Likelihood function is weighted by the sample size.

$Z_{\text{one}} = 0.48 \text{ yr}^{-1} \rightarrow Z_{\text{two}} = 0.97 \text{ yr}^{-1}$ in 1993.

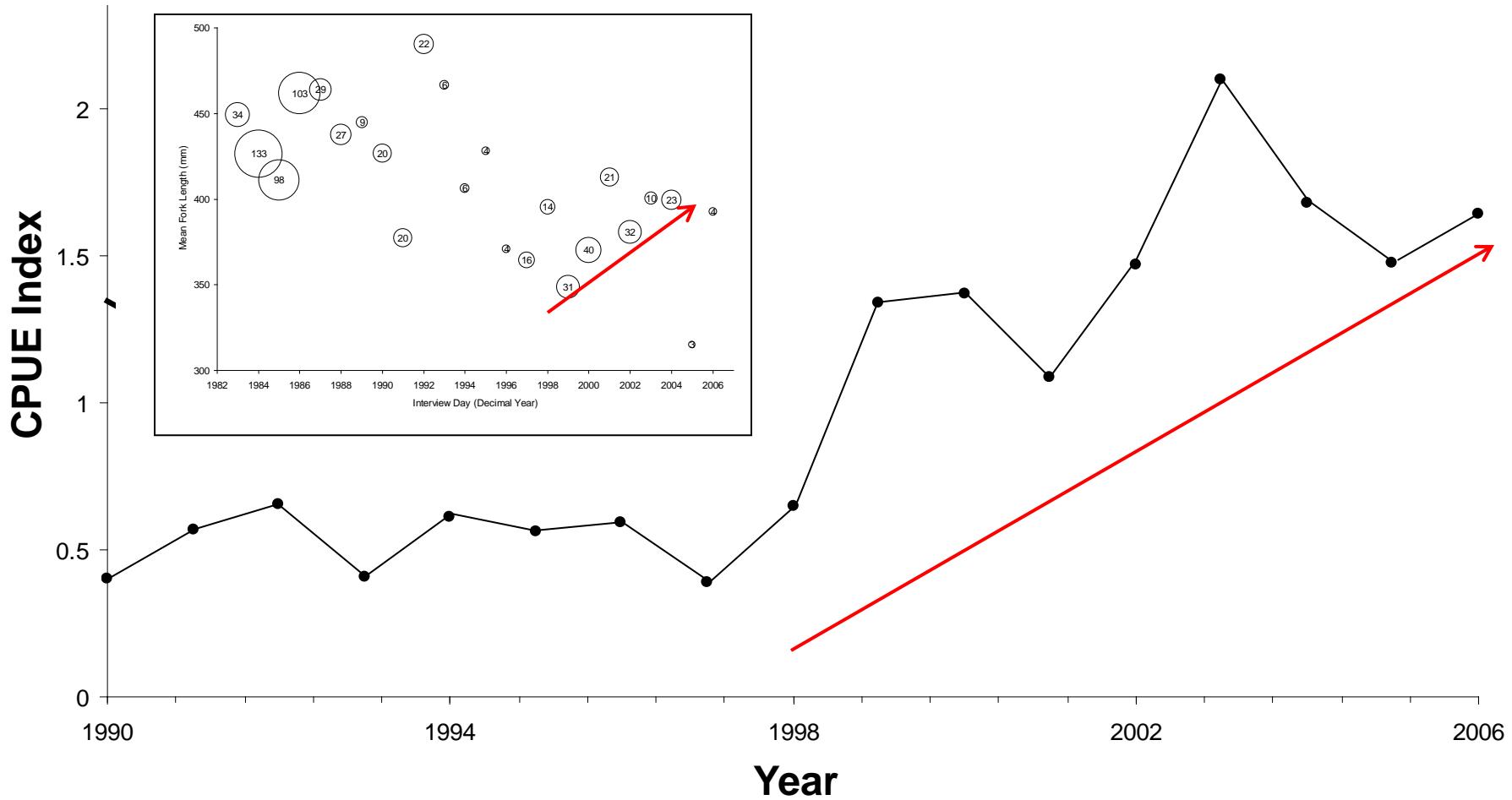


Mean Length calculated by year. Sample numbers for each year have been indicated by both bubble size and number.



Integrating Catch Rates into the Mean Length Analysis

Standardized CPUE Indices—Mutton Snapper Pot Fishery—Puerto Rico



Estimating Mortality from Changes in Mean Lengths and Catch Rates in Non-equilibrium Conditions

Application to the Mutton Snapper Assessment

Final Step of Derivation:

If q^* is assumed to be constant,

$$\frac{I_1}{I_2} = \frac{Z_2}{Z_1}$$

Key Point: Only able to get information on relative changes in total mortality rates from catch indices

e.g. $2 * Z = 0.5 * I$



Todd Gedamke¹, Clay E. Porch¹ and John Hoenig²

¹National Marine Fisheries Service, Southeast Fisheries Science Center, Miami, Florida

²Virginia Institute of Marine Science, Gloucester Point, Virginia



Multispecies

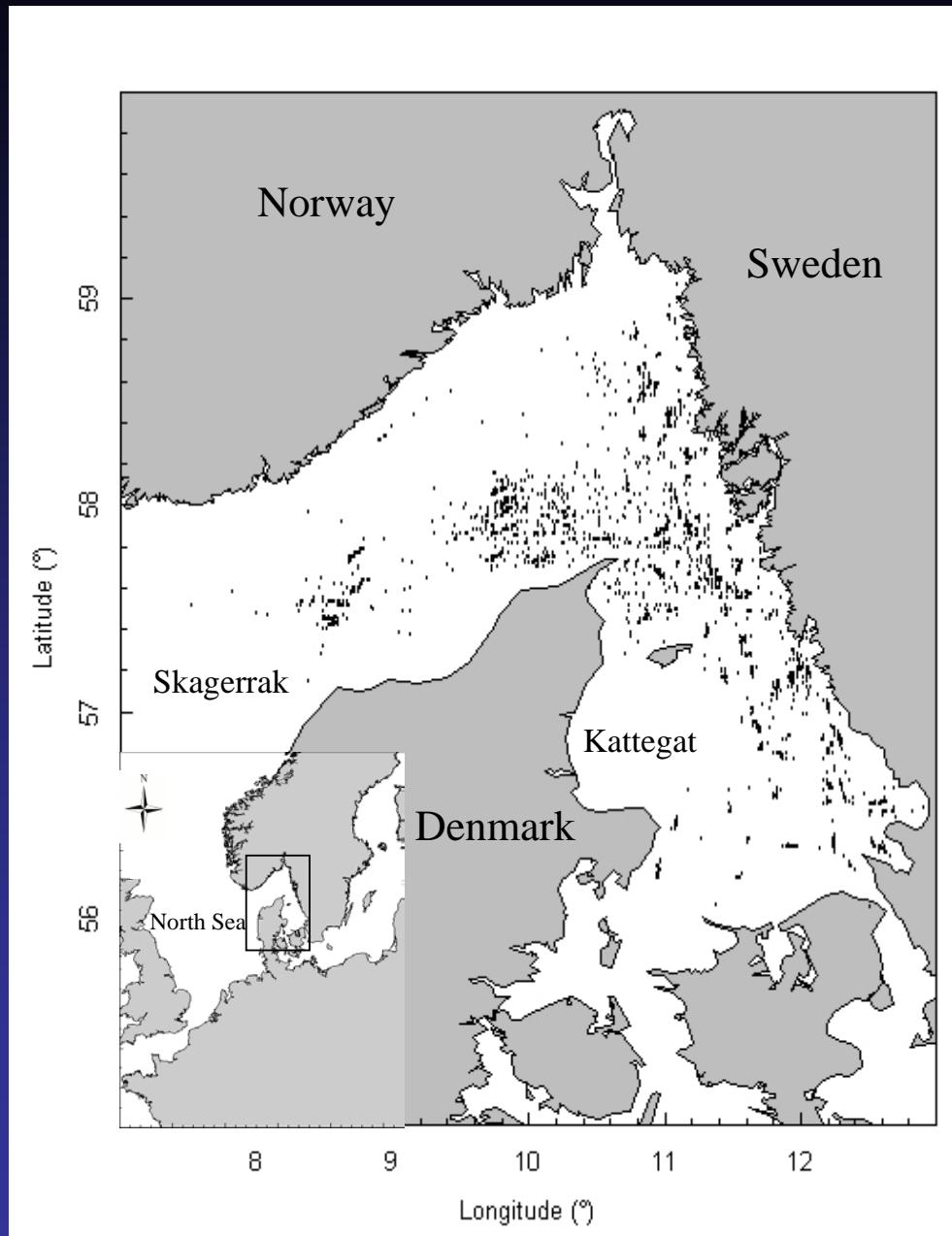
A century of variability of body size in a flatfish community: implications for management

Massimiliano Cardinale*, Andrea Belgrano, Valerio Bartolino, Todd Gedamke, Hans Linderholm

Massimiliano Cardinale, Andrea Belgrano, Valerio Bartolino,
Institute of Marine Research, Swedish Board of Fisheries, P.O.
Box 4, SE-453 21 Lysekil, Sweden

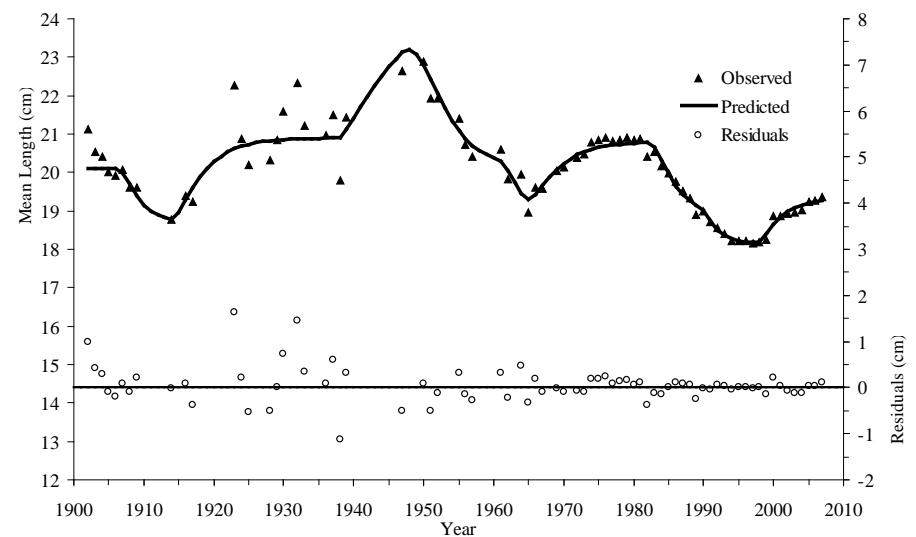
Todd Gedamke, National Oceanic and Atmospheric
Administration, National Marine Fisheries Service, 75 Virginia
Beach Drive, Miami, Florida, 33149, USA

Hans Linderholm, Regional Climate Group, Department of
Earth Sciences, University of Gothenburg, Gothenburg,
Sweden

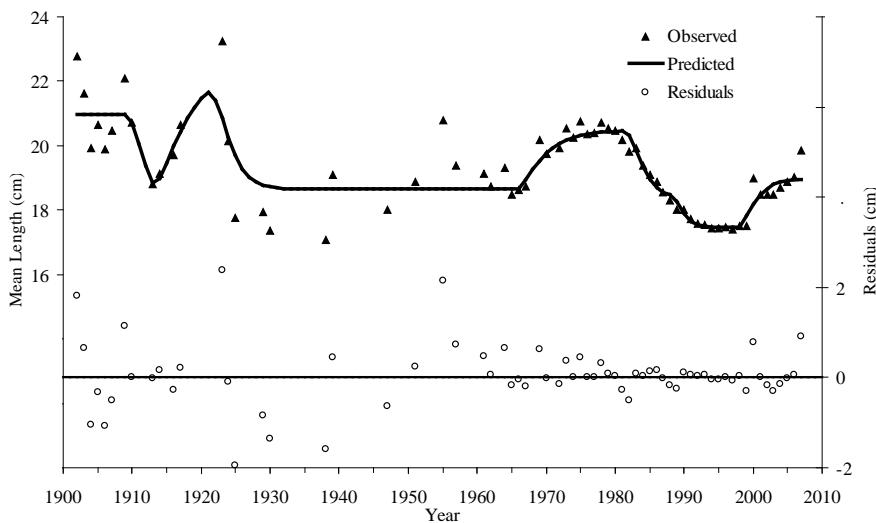


Dab, Long Rough Dab, Turbot, Plaice

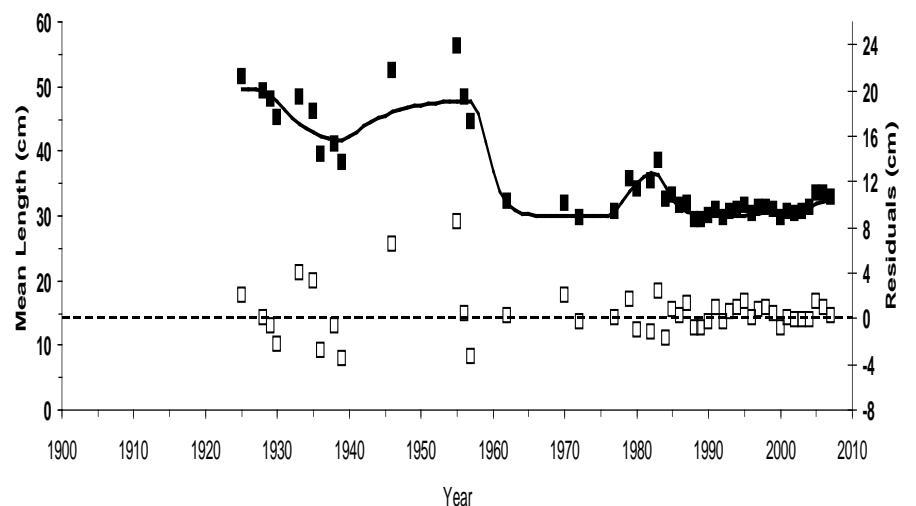
Dab - 9 changes



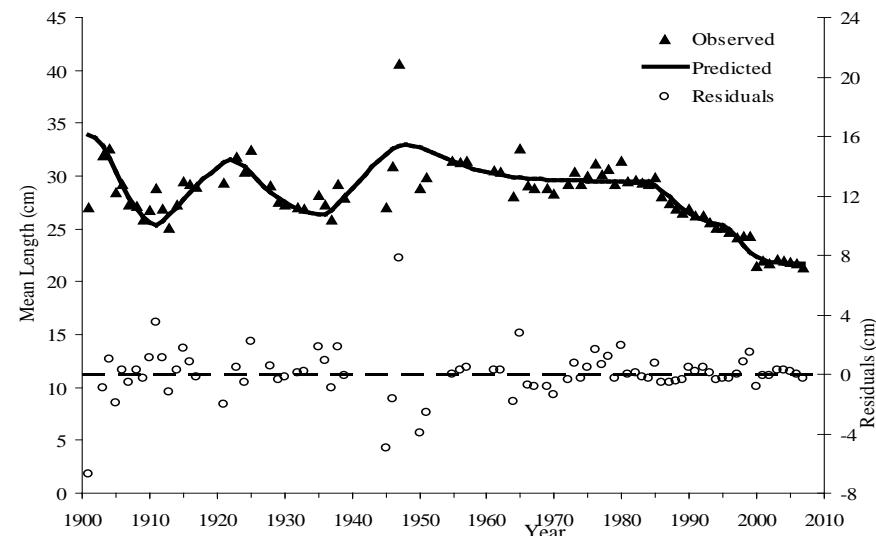
Long rough dab - 7 changes



Turbot - 7 changes

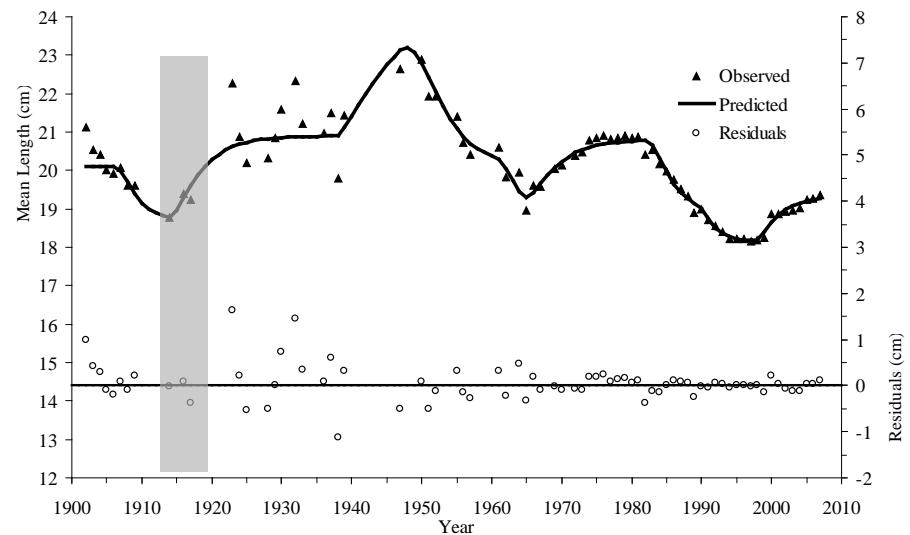


Plaice-7 changes

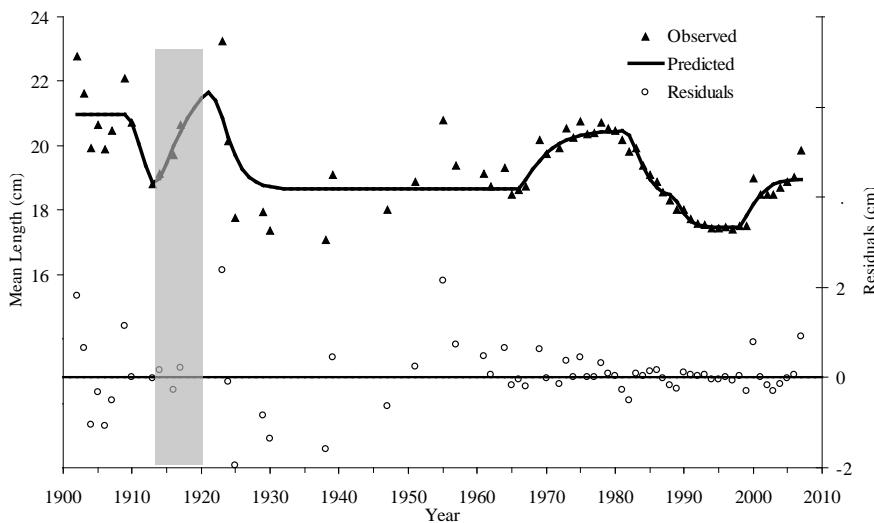


World War I

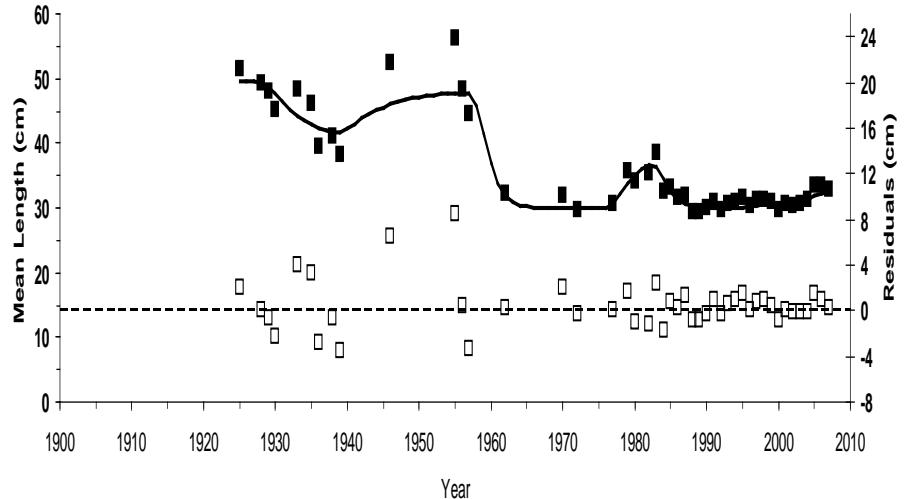
Dab - 9 changes



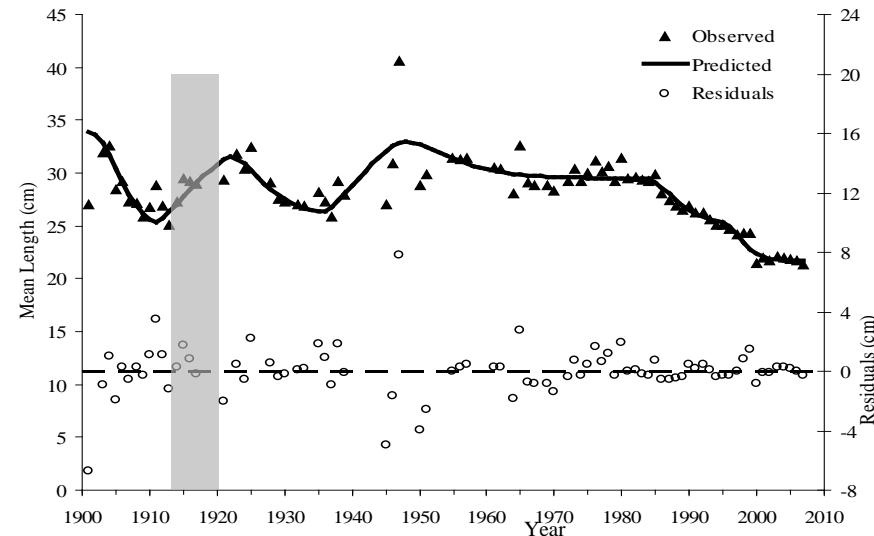
Long rough dab - 7 changes



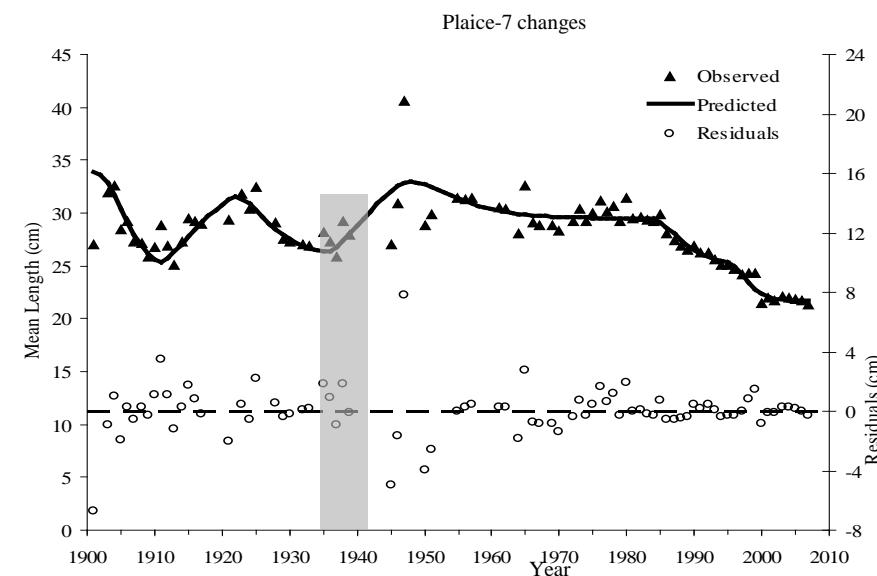
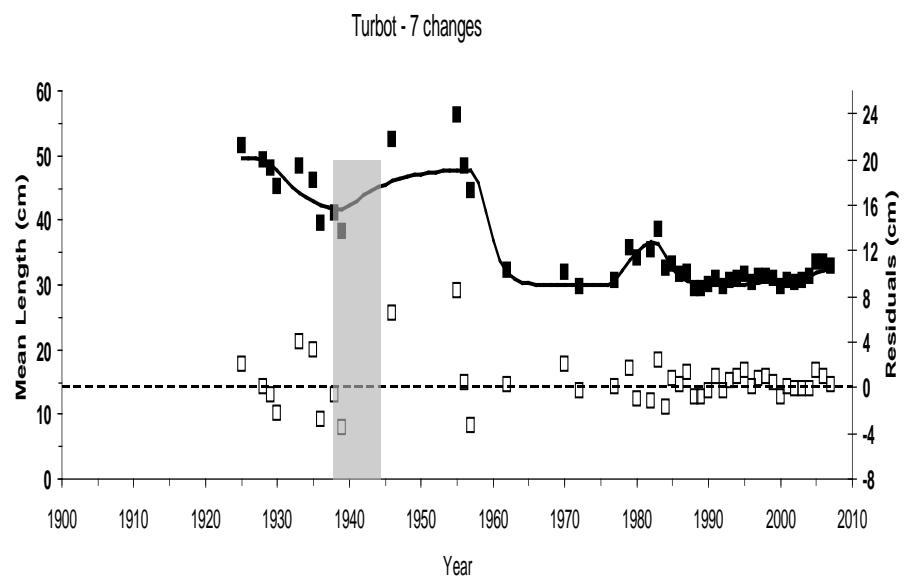
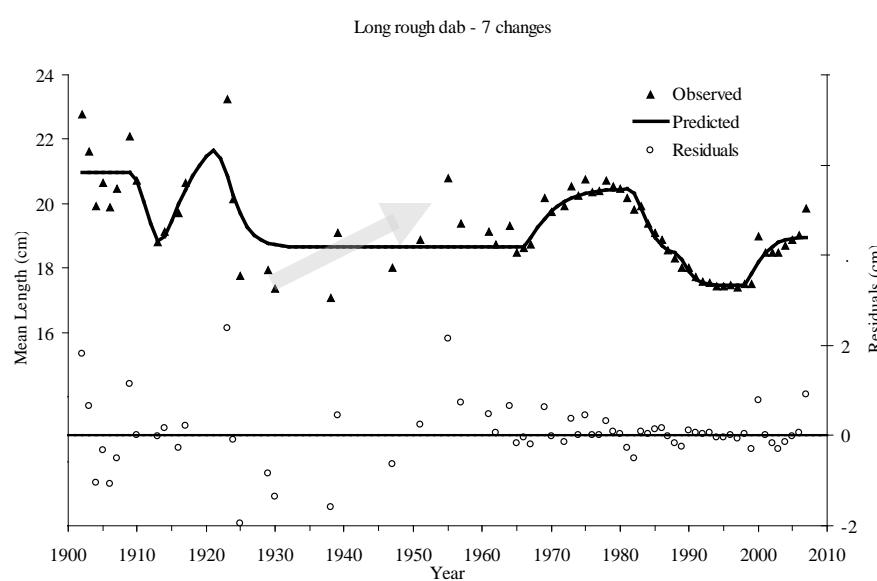
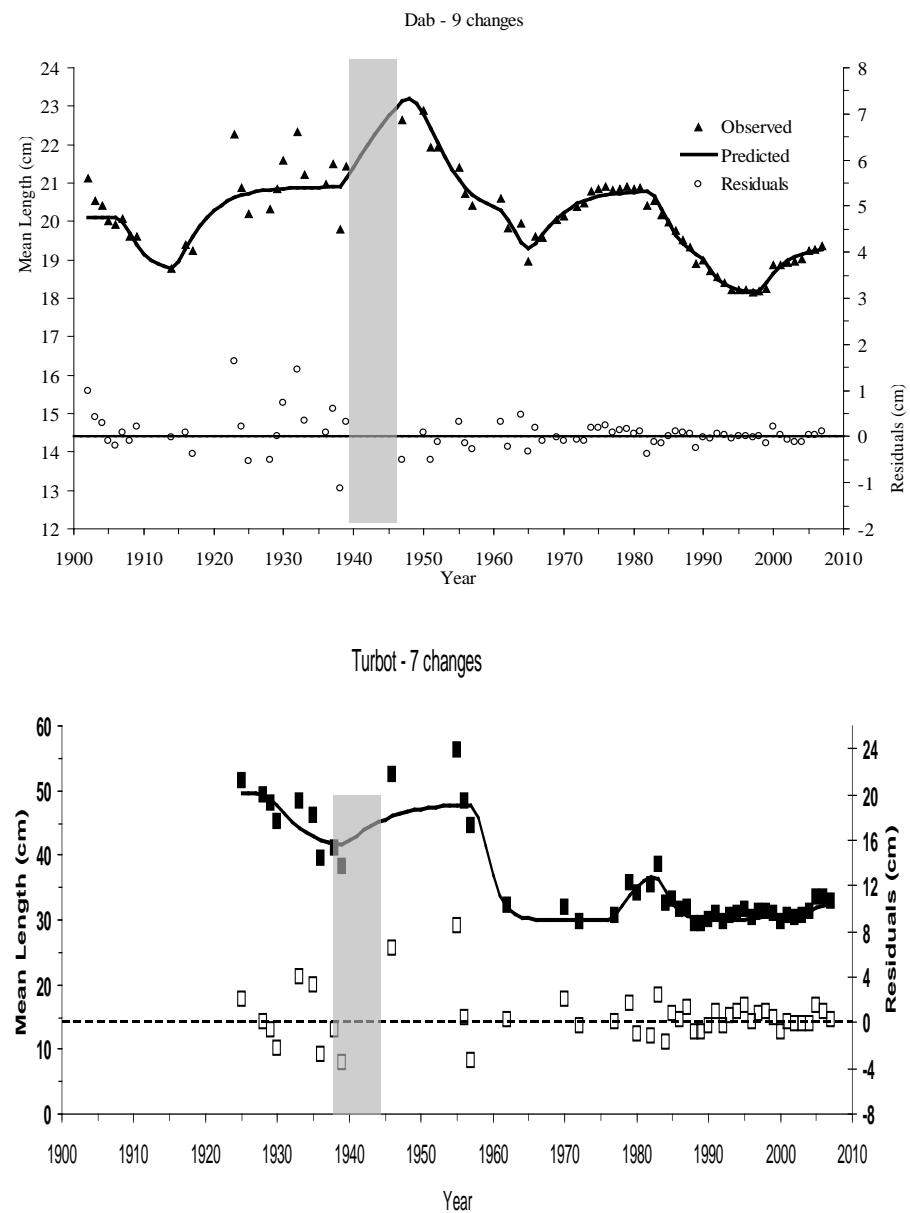
Turbot - 7 changes



Plaice-7 changes

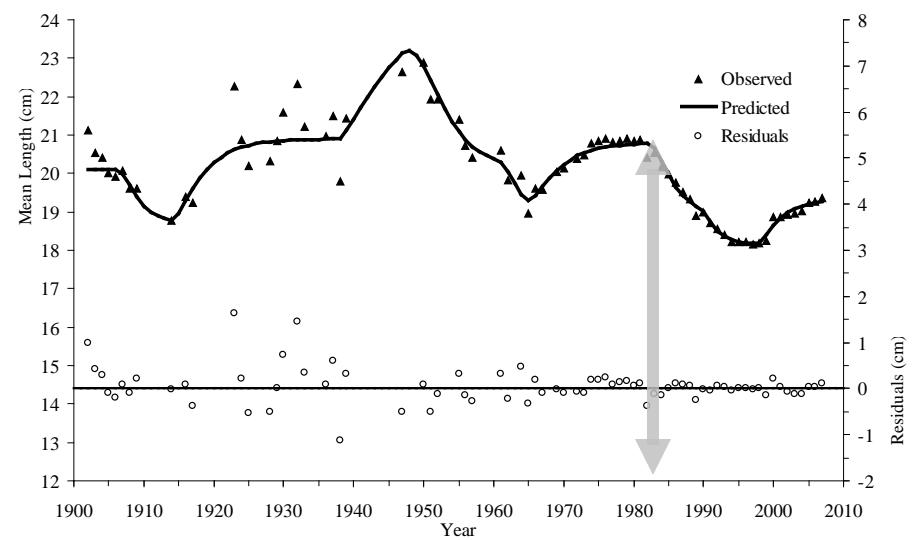


World War II

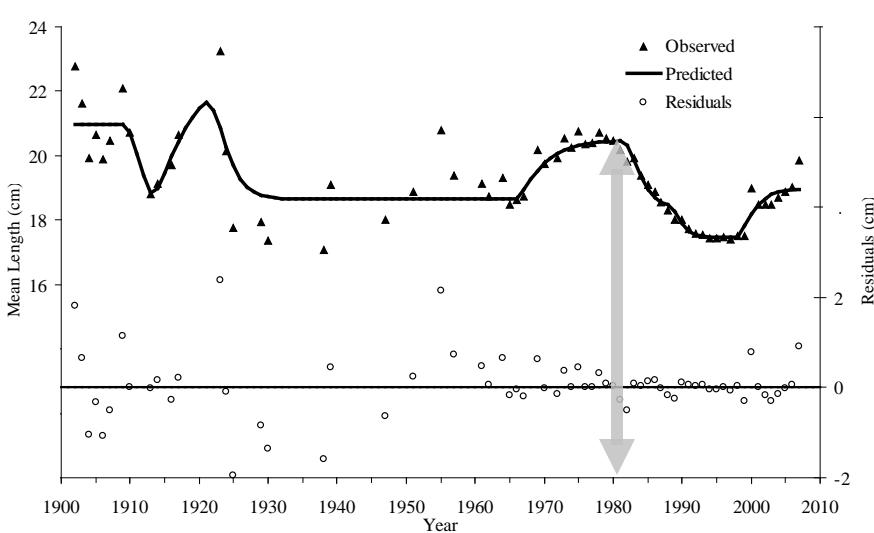


Early 1980's – Declining Mean Lengths

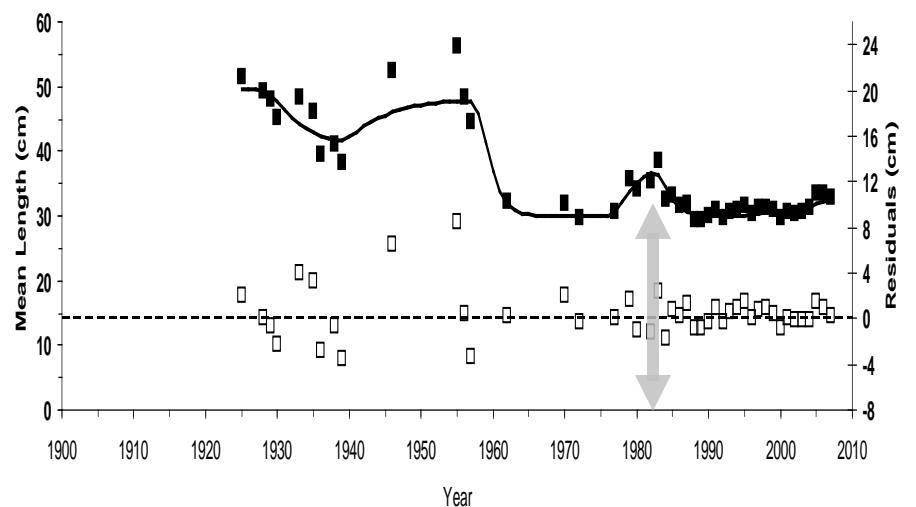
Dab - 9 changes



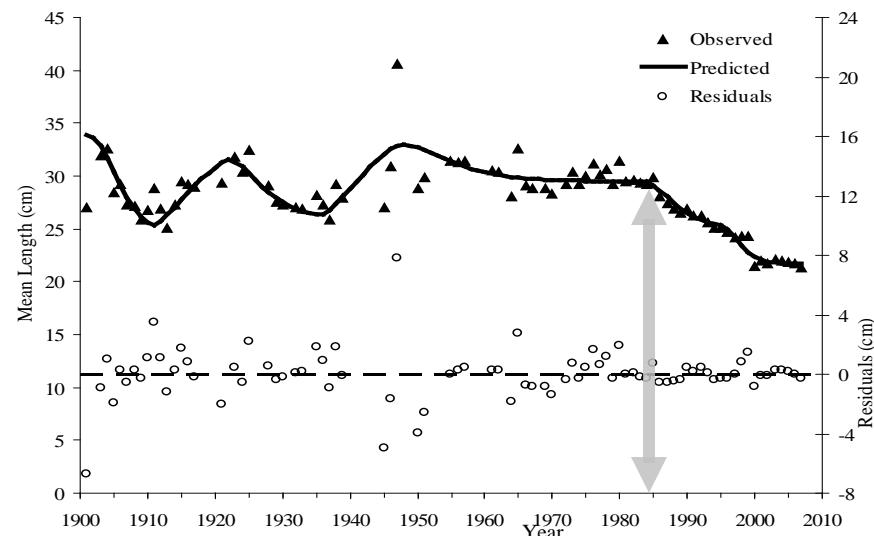
Long rough dab - 7 changes



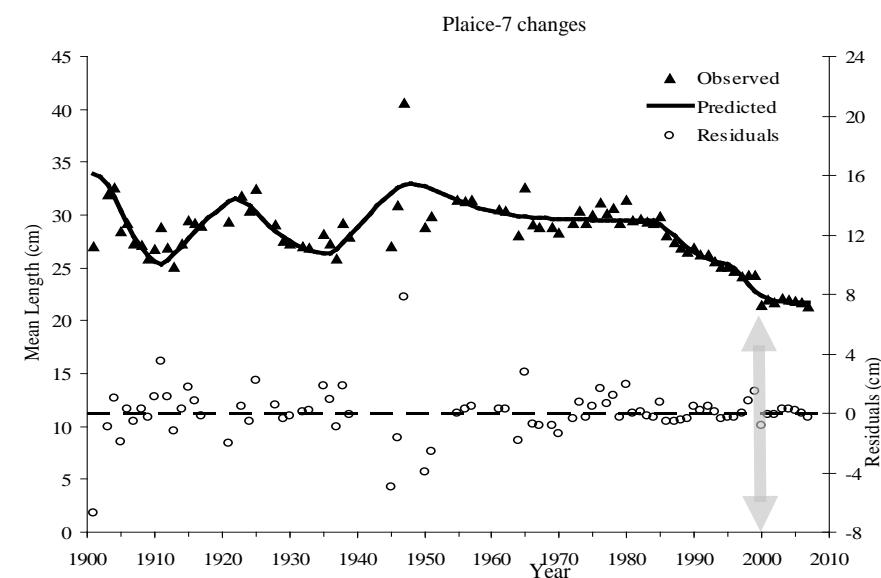
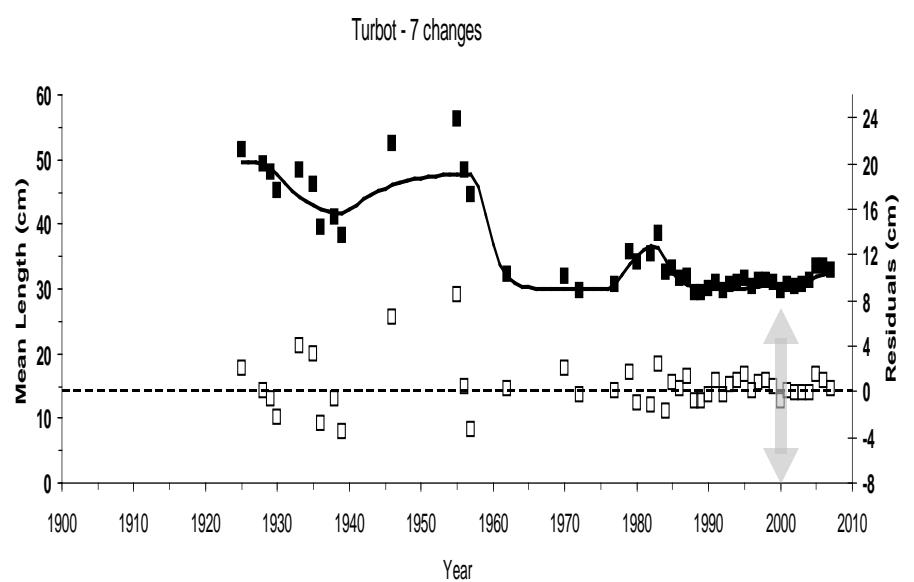
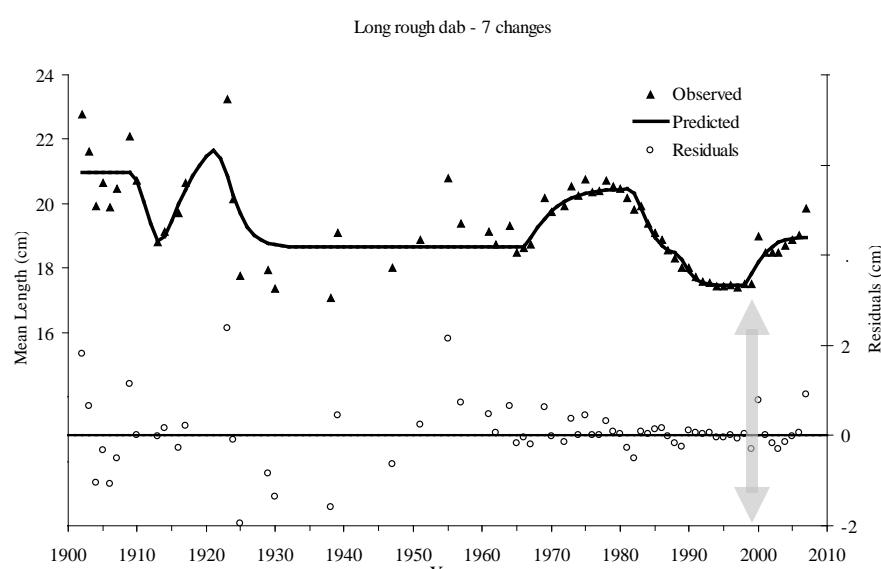
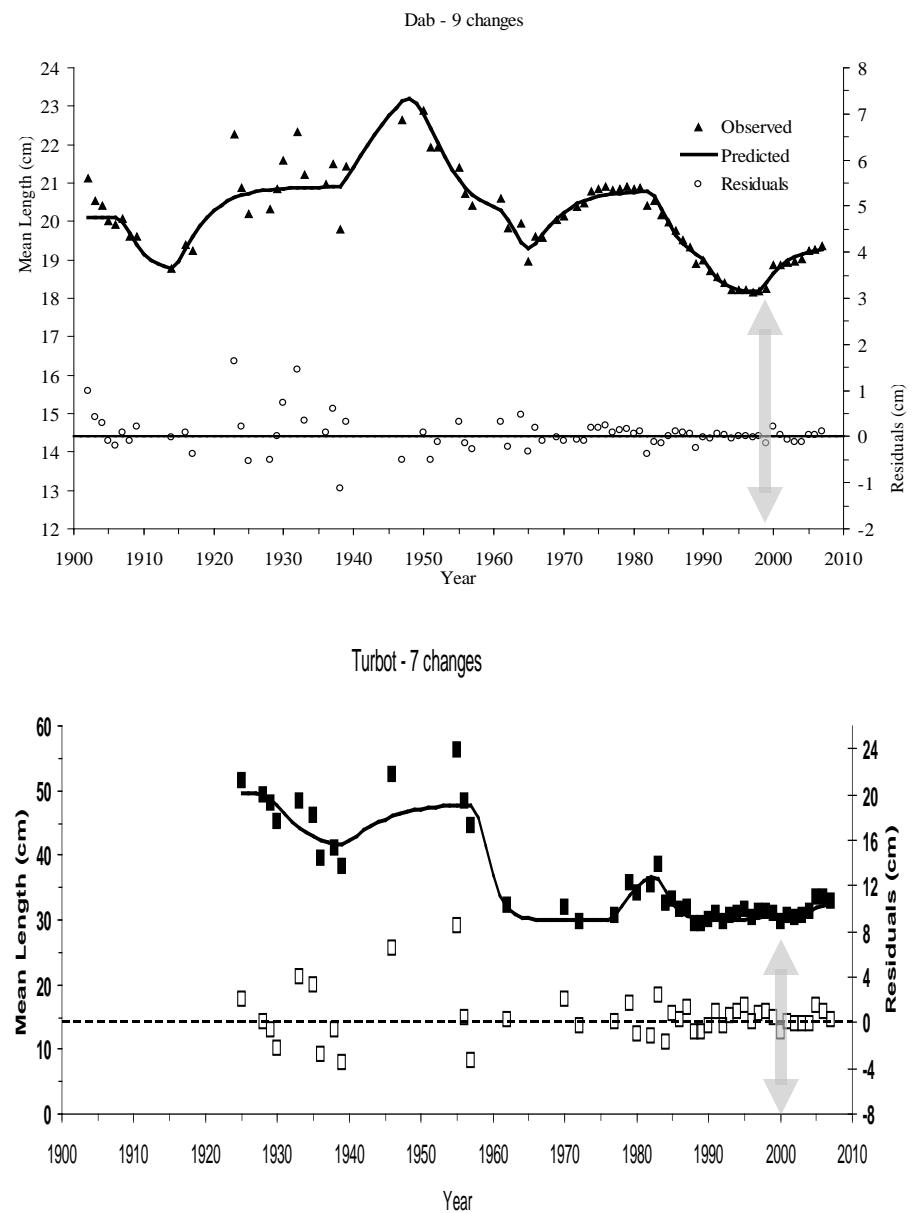
Turbot - 7 changes



Plaice-7 changes



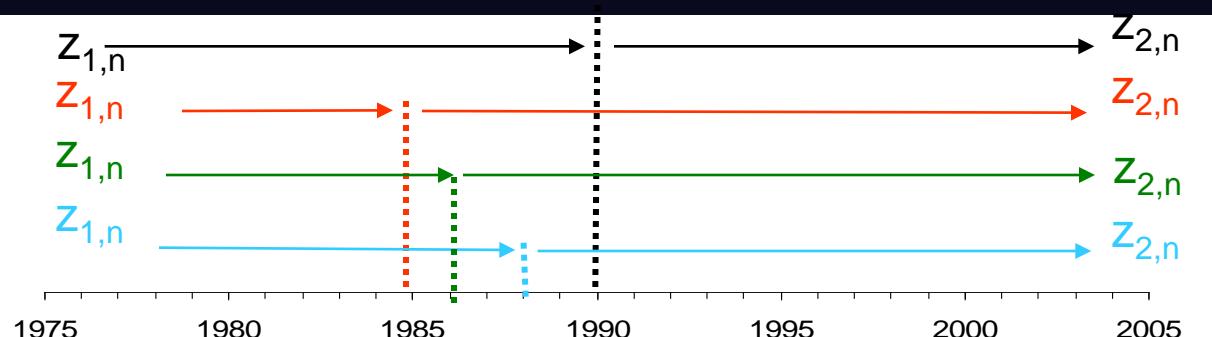
Another Change ~2000 - Management?



MULTI-SPECIES APPROACH

- Similar to single species model with modified assumptions
- Uses information from multiple species to increase sample size
- Assumes that species used from ‘complexes’ are subject to similar patterns of effort

Multi-Species Model Development



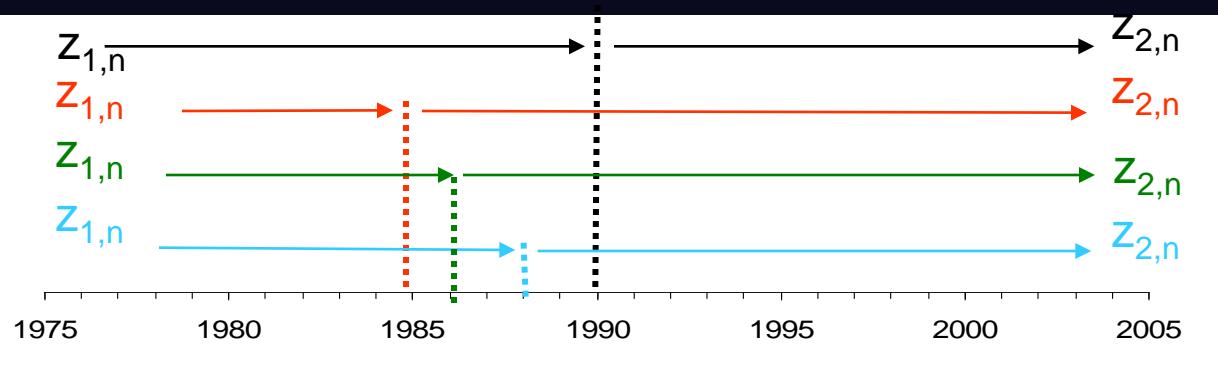
Single Species Model:

4 parameters =
(Z_1 , Z_2 , ChangeYear, σ)

4 species = 16 parameters

+ 4 parameters for each additional species

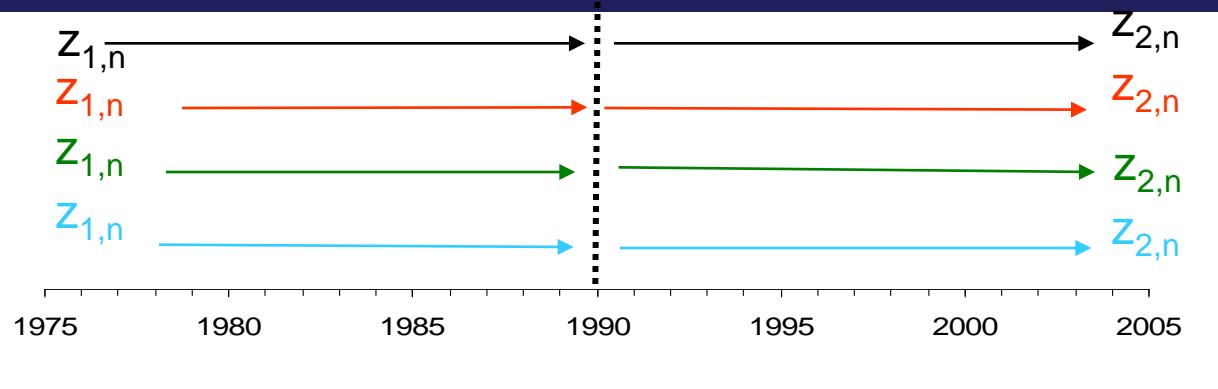
Multi-Species Model Development



Single Species Model:

4 parameters =
(Z_1 , Z_2 , ChangeYear, σ)

4 species = 16 parameters



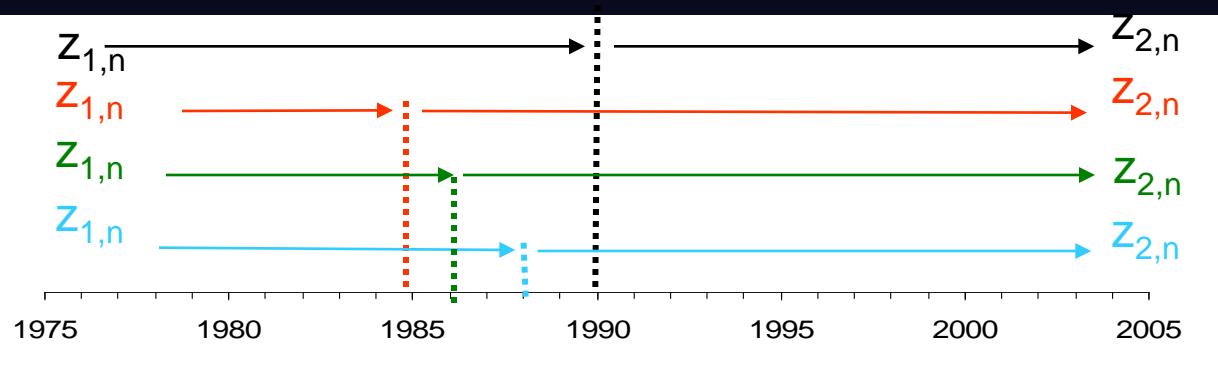
Multi-Species Model 1:

4 parameters =
(Z_1 , Z_2 , ChangeYear, σ)

Assume common ChangeYear
for multiple species → 4
species = 13 parameters

+ 3 parameters for each additional species

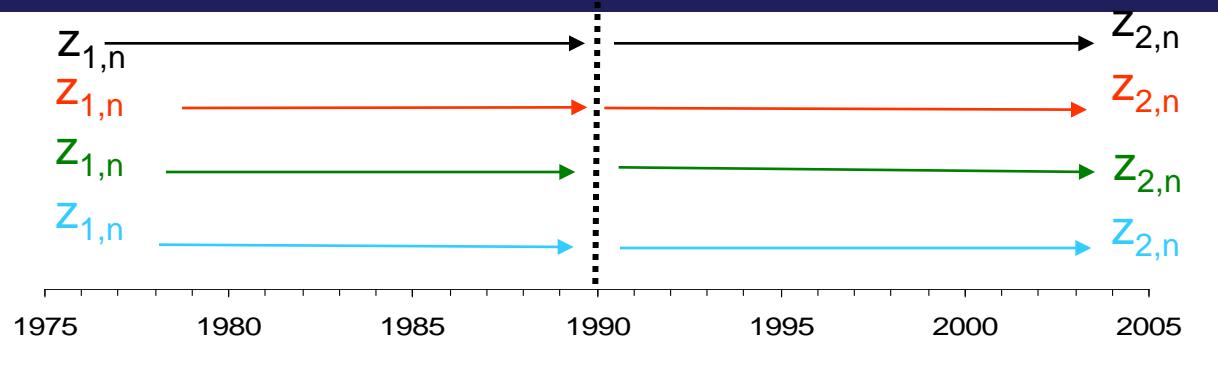
Multi-Species Model Development



Single Species Model:

4 parameters =
(Z_1 , Z_2 , ChangeYear, σ)

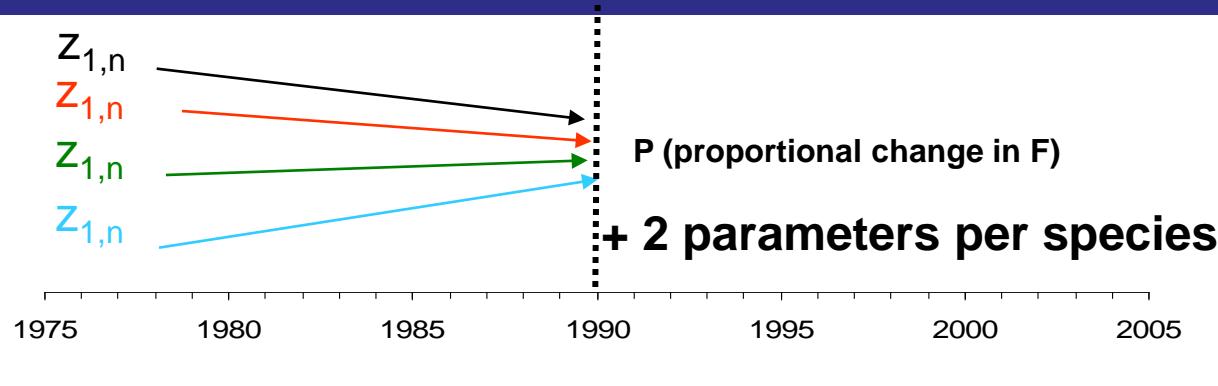
4 species = 16 parameters



Multi-Species Model 1:

4 parameters =
(Z_1 , Z_2 , ChangeYear, σ)

Assume common ChangeYear
for multiple species → 4
species = 13 parameters

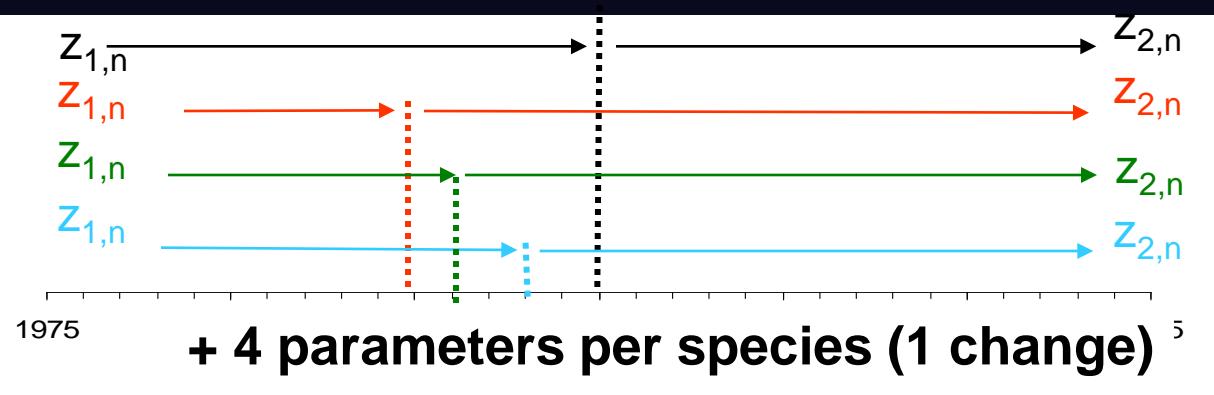


Multi-Species Model 2:

4 parameters =
(Z_1 , Z_2 , ChangeYear, σ)

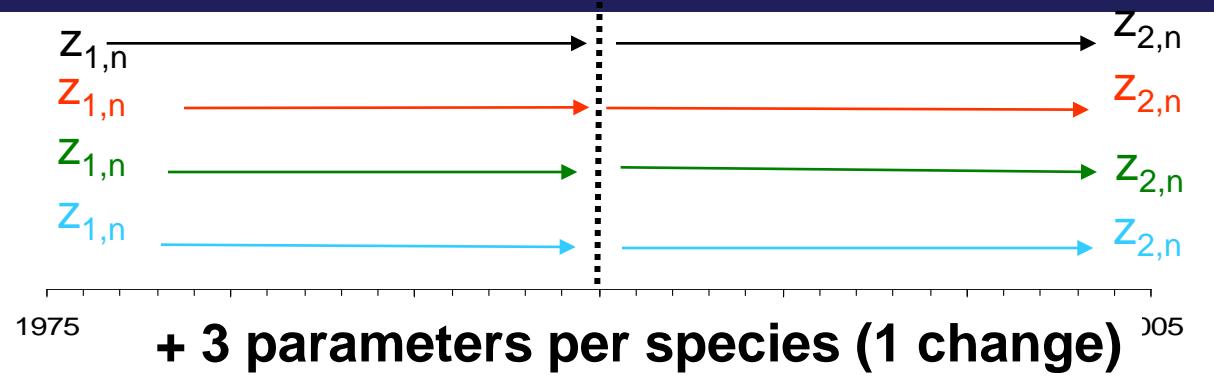
Assume common ChangeYear
and proportional change in F
→ 4 species = 10 parameters

Multi-Species Model Development



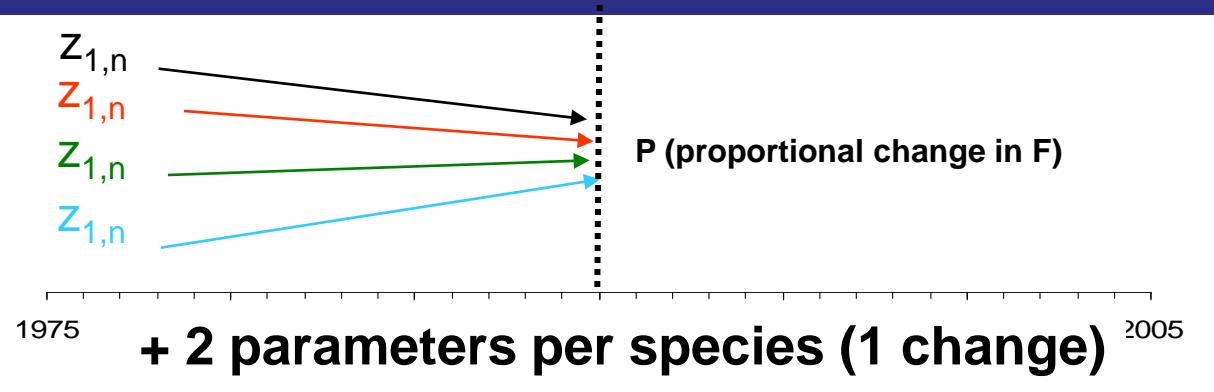
For 3 ChangeYears and 4 species:

32 Parameters



Assume common ChangeYear

23 Parameters

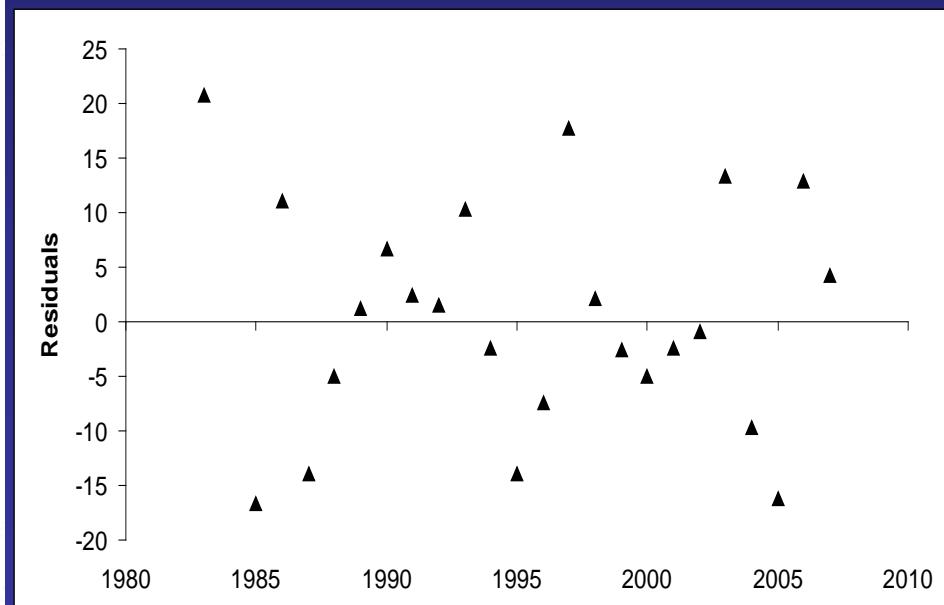
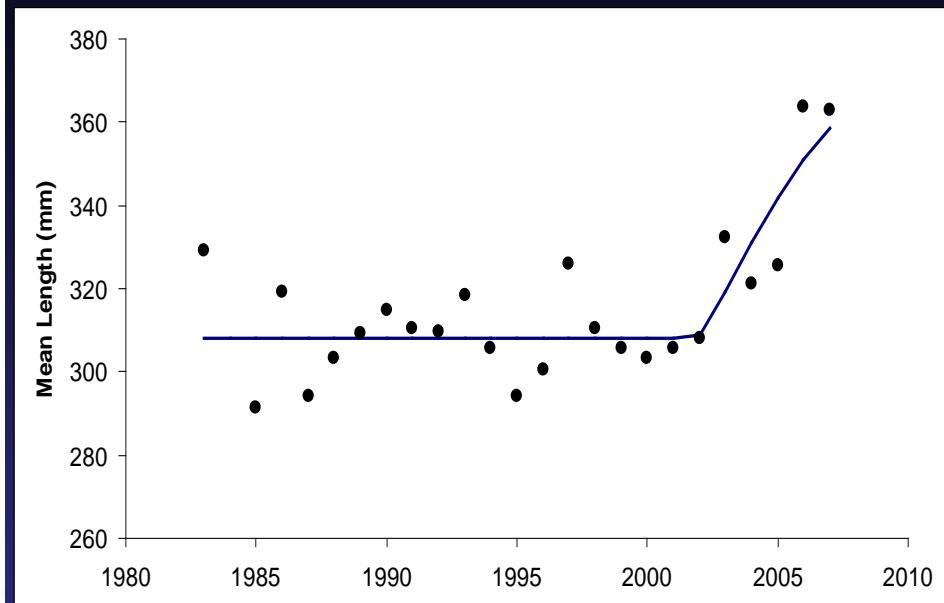
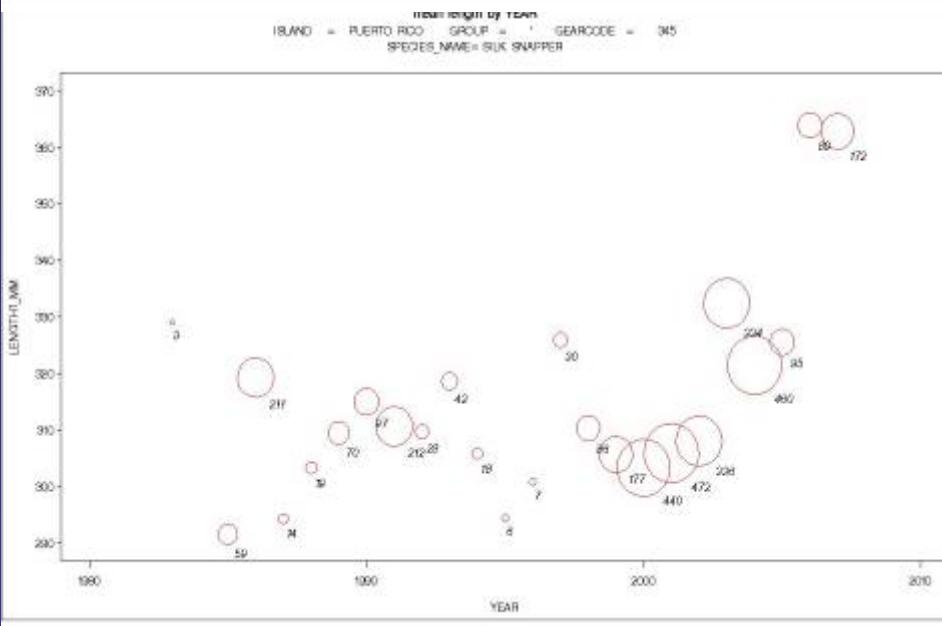
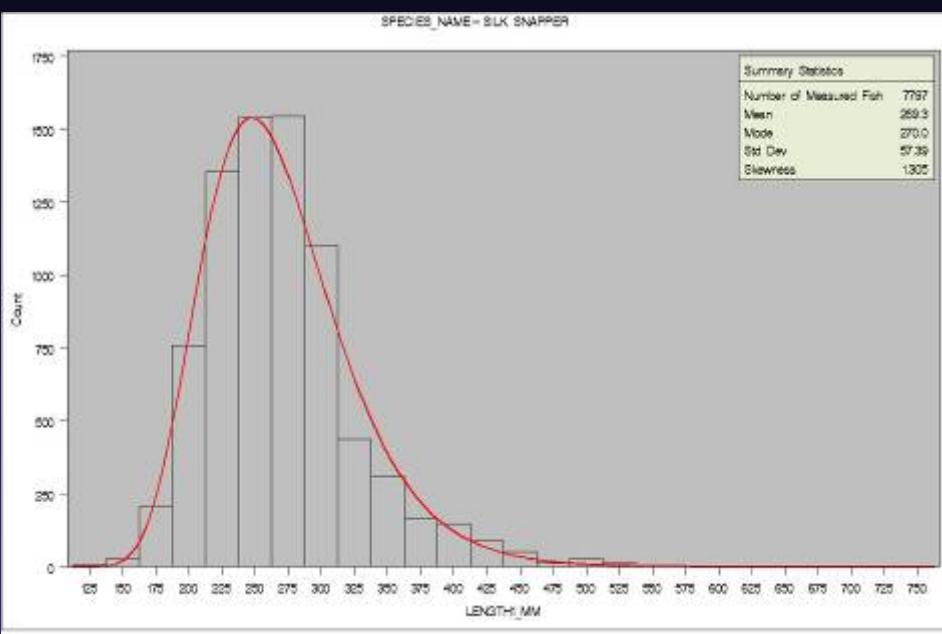


Assume common ChangeYear and proportional change in F

14 Parameters

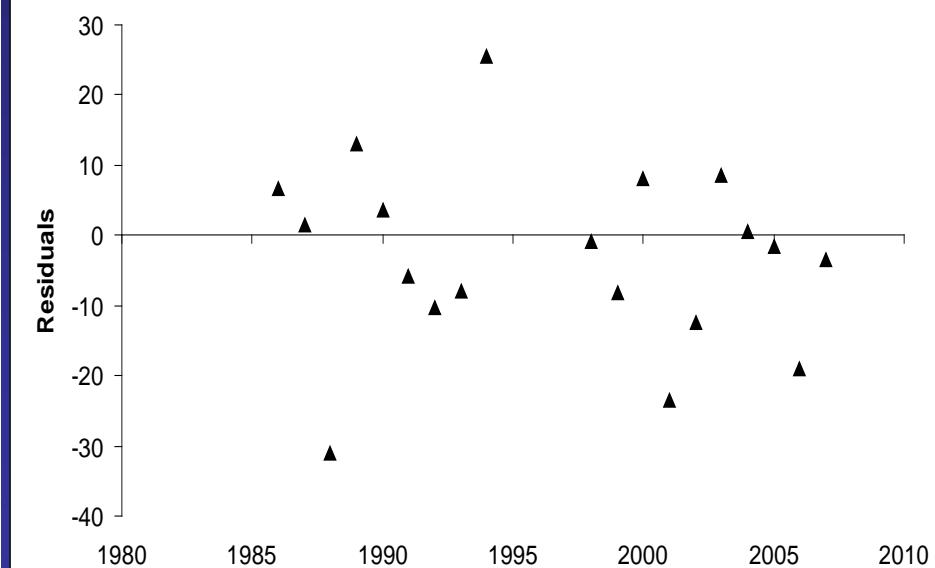
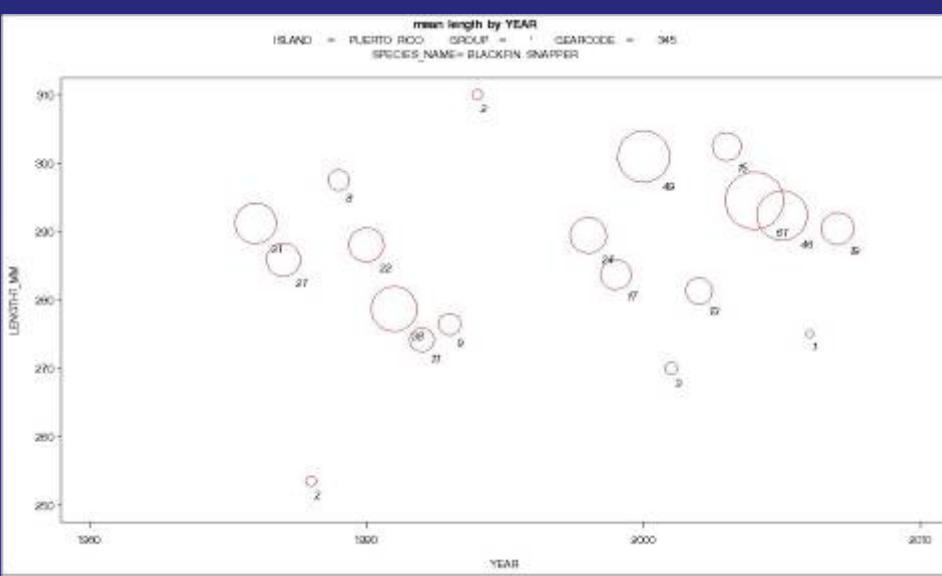
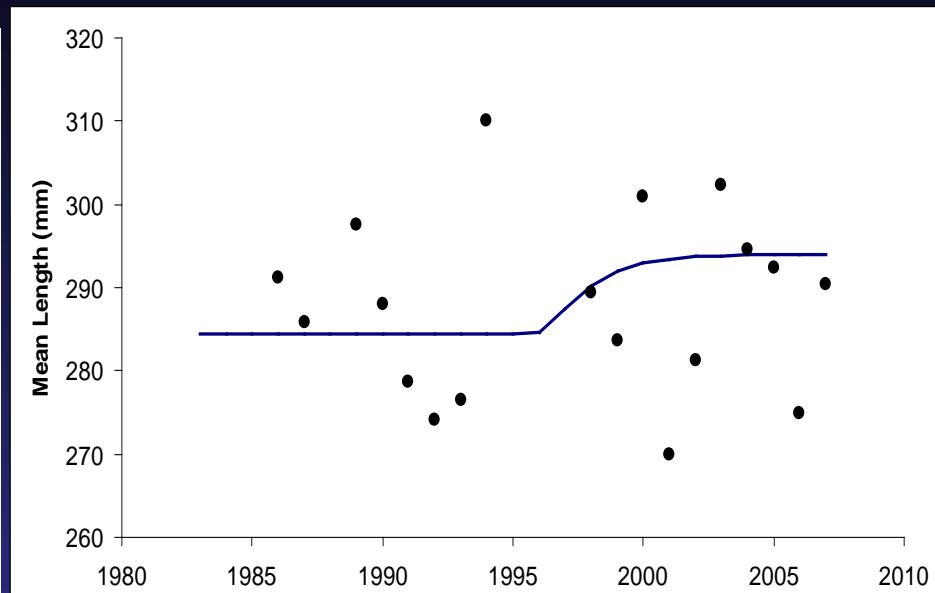
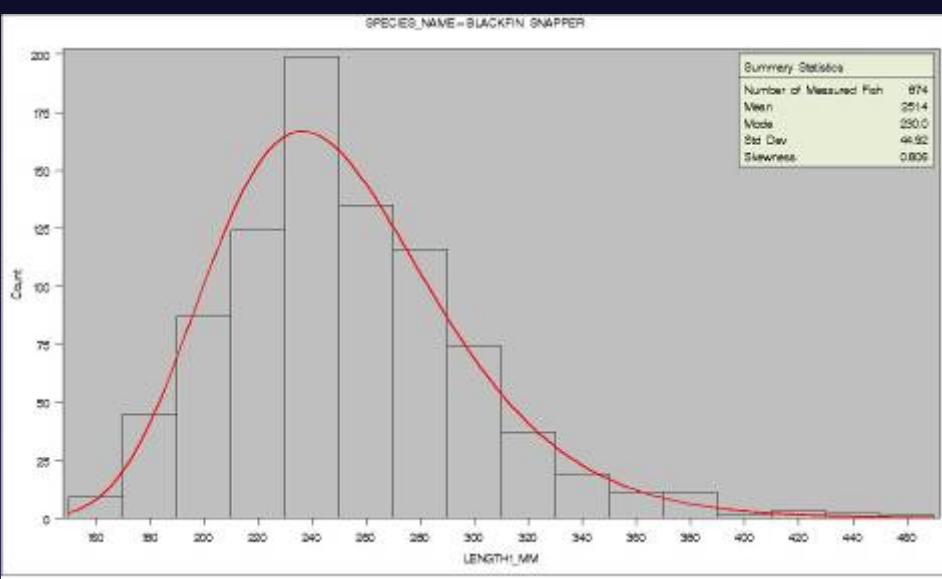
Puerto Rico – Silk Snapper – Traps

$Z = 1.12 \rightarrow 0.31$ in 2001.8



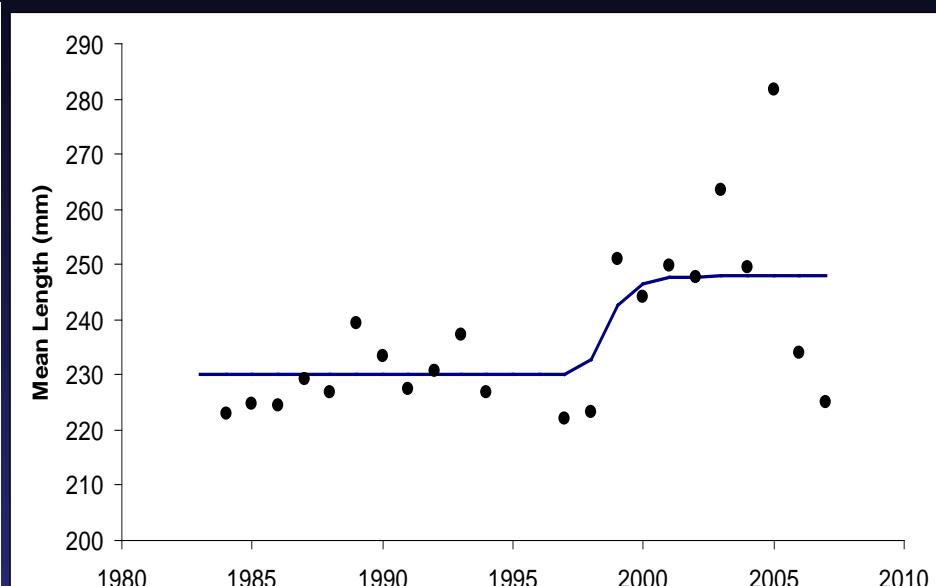
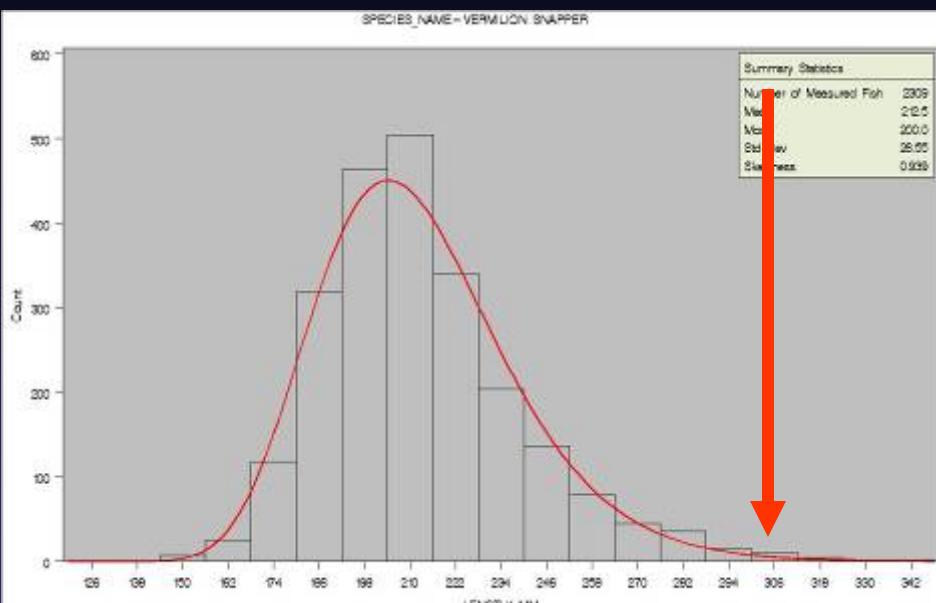
Puerto Rico – Blackfin Snapper – Traps

$Z = 1.13 \rightarrow 0.86$ in 1995.8

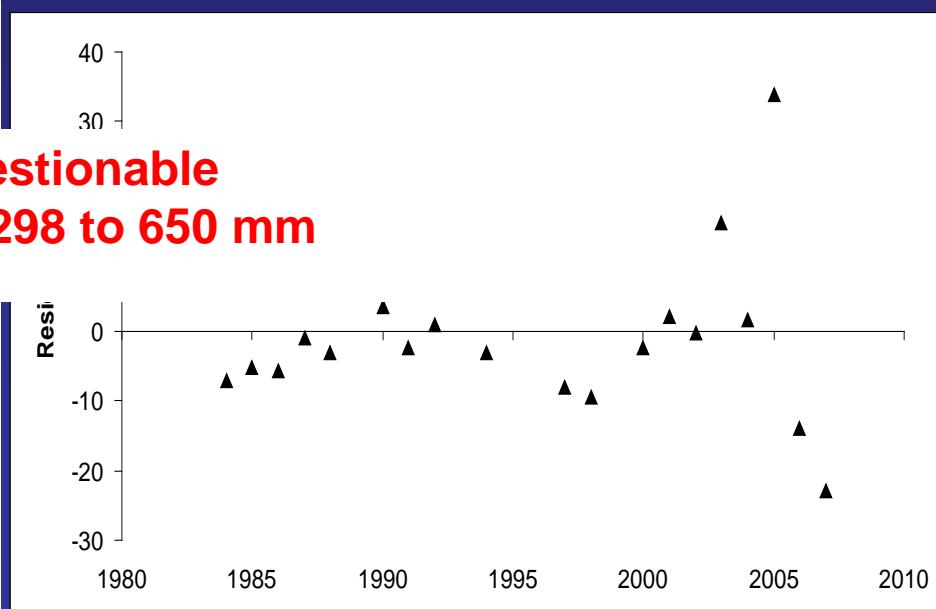
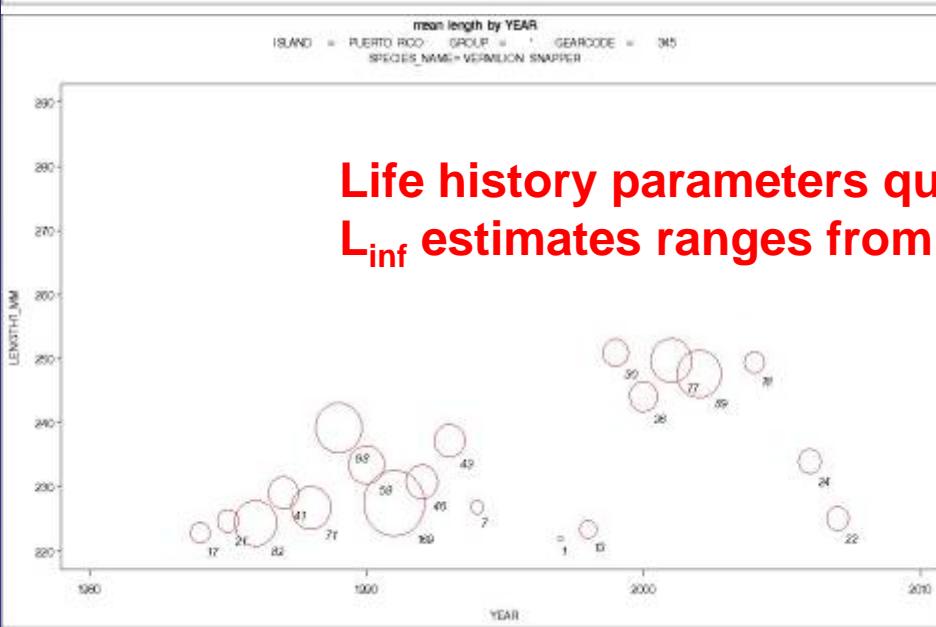


Vermillion Snapper

Z = 3.46 → 1.64 in 1997.7

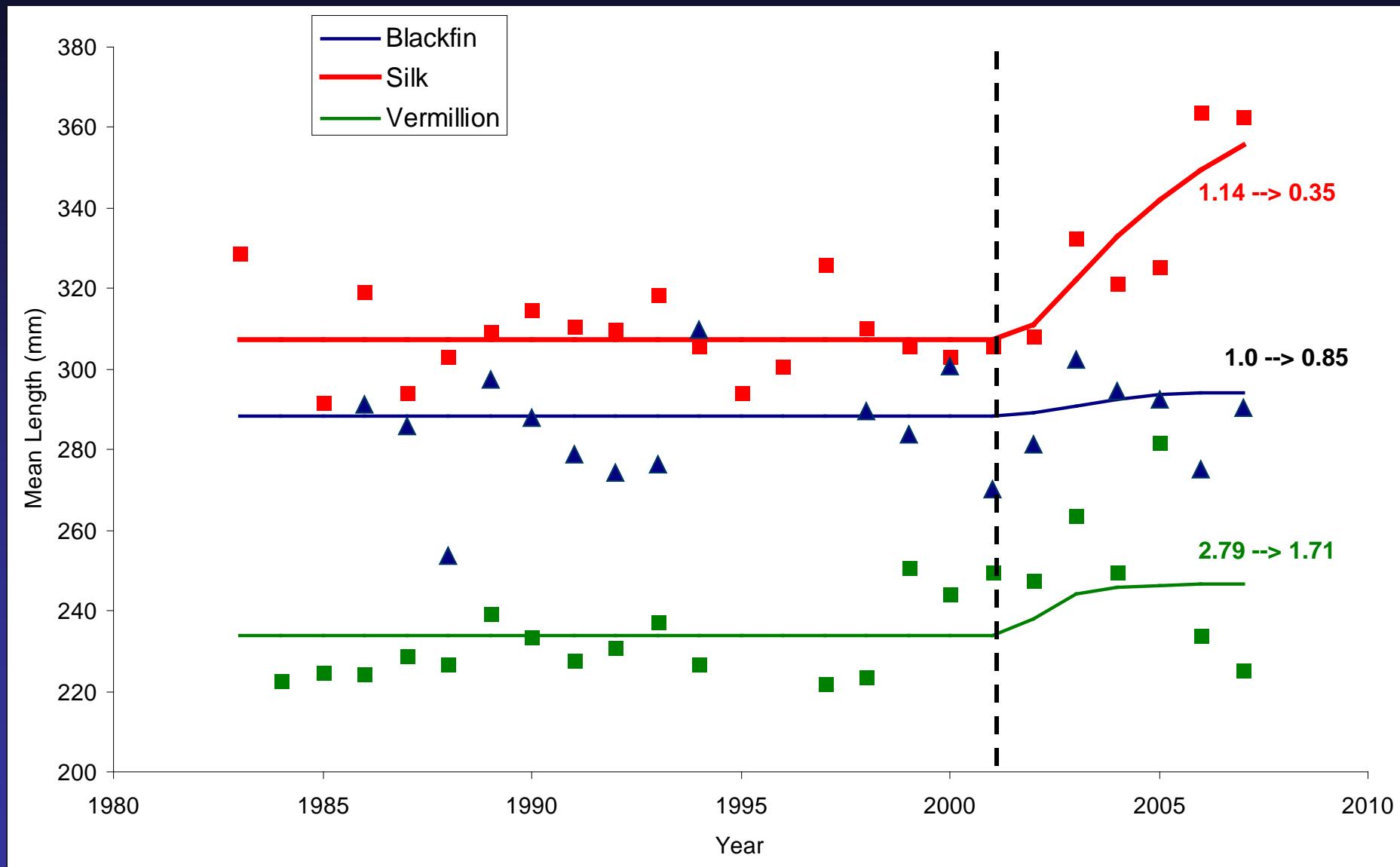


Life history parameters questionable L_{inf} estimates ranges from 298 to 650 mm



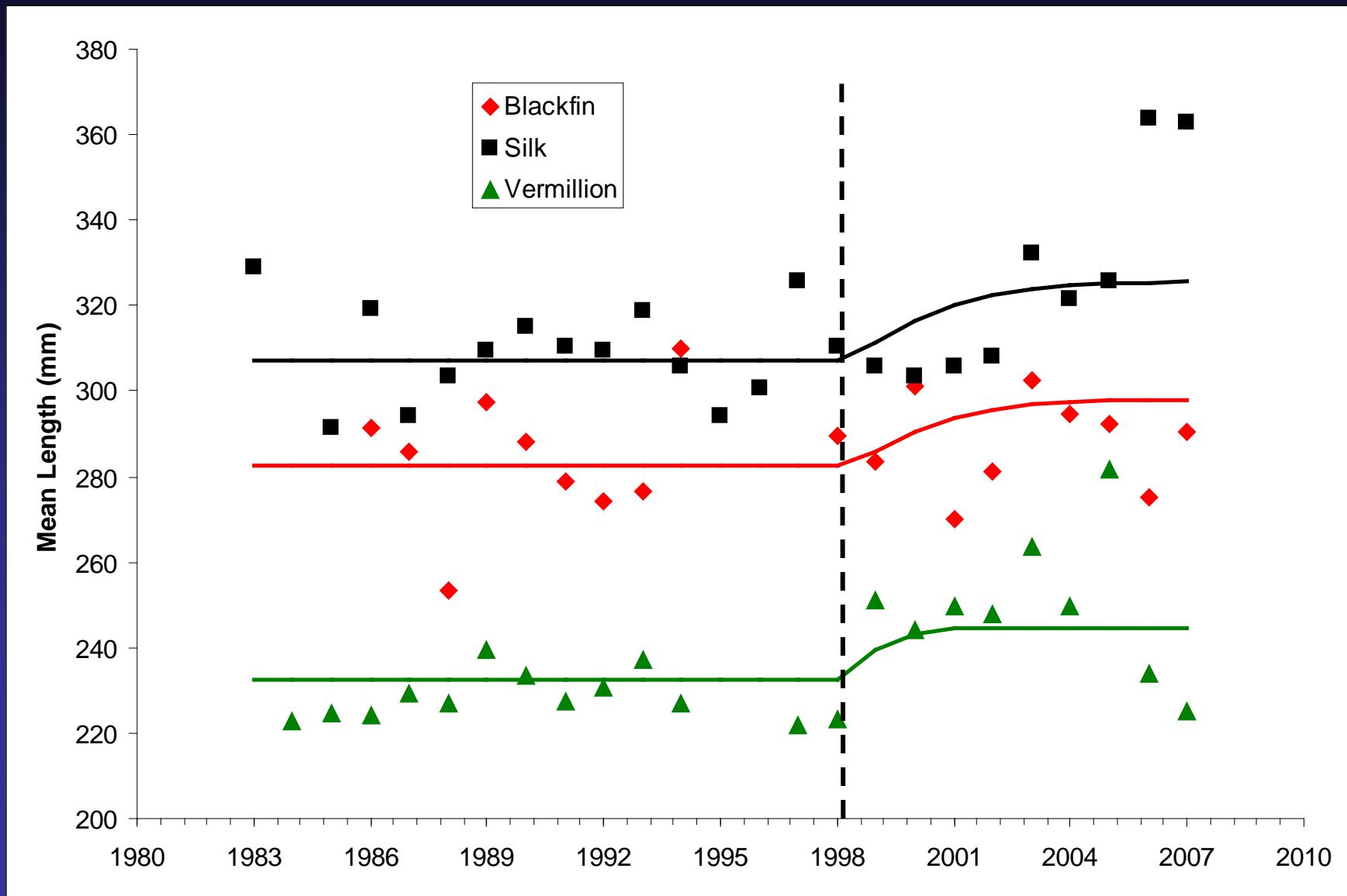
Multispecies - Individual Z's estimated

Estimated Common Year of Change = 2001.4



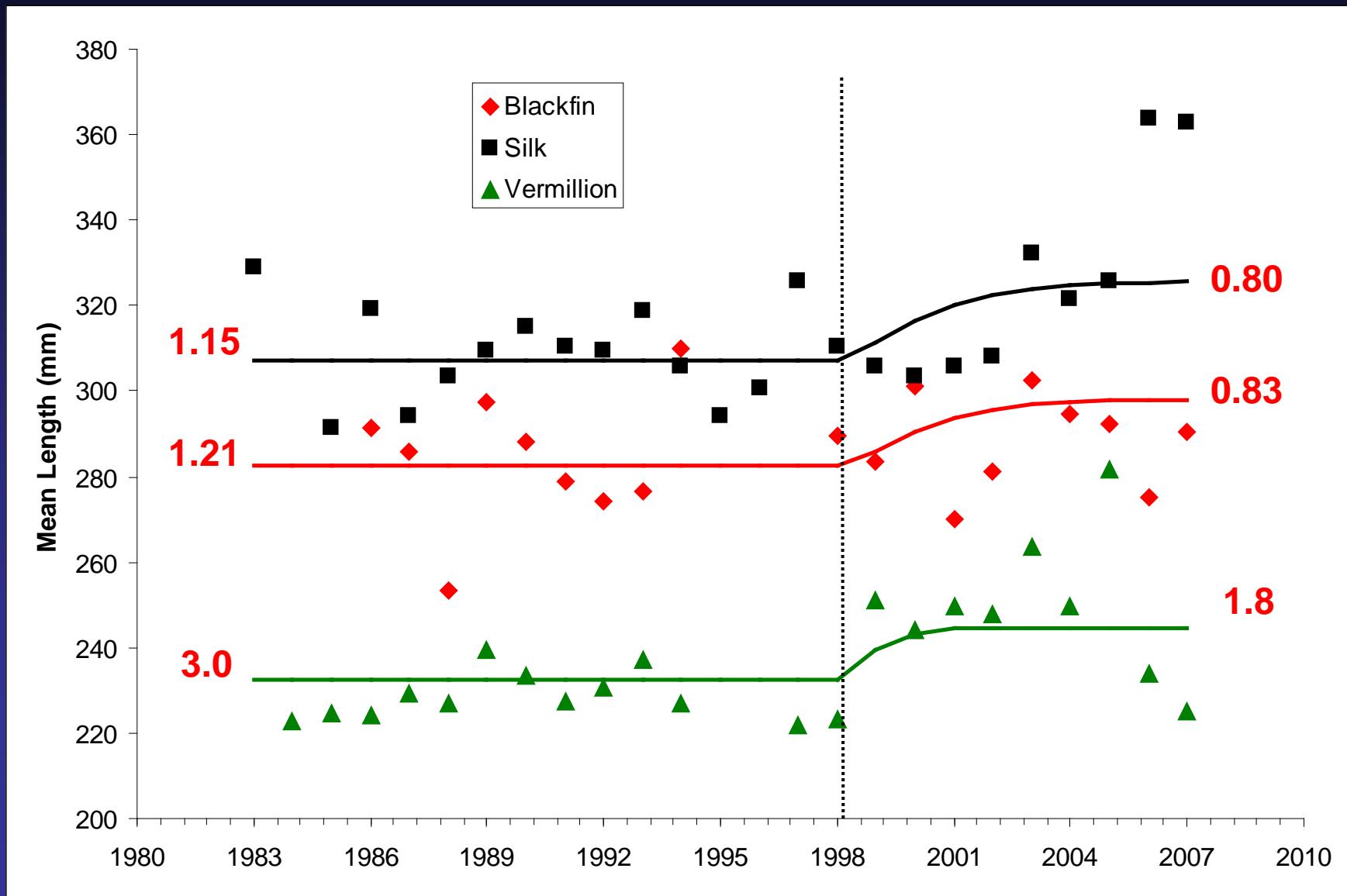
Multispecies w/ Common Proportional Change in F

Common Year of Change = 1998.03
Common Proportional change in F = 0.57



Multispecies w/ Common Proportional Change in F

Common Year of Change = 1998.03
Common Proportional change in F = 0.57



Model Selection

AIC (Akaike's information criterion)

- Evaluates balance between number of parameters in model and the fit to the data (likelihood value).
- Penalty for increased number of parameters
- Lower AIC values better → Reduction of ~5 indicates strong support for model

$$AICc = AIC + \frac{2 \times Nparam \times (Nparam+1)}{(NumObs - Nparam-1)}$$

Model	Gear	Depth	Number of Parameters	Likelihood Value	AIC	AICc
Single species	345/610	<80	12	441.2	906.4	911.4
Multispecies (individual Z's)	345/610	<80	10	432.9	885.7	889.1
Multispecies (proportional change)	345/610	<80	8	445.0	905.9	908.1
Single species	345	All	12	398.1	820.3	825.3
Multispecies (individual Z's)	345	All	10	363.1	746.1	749.6
Multispecies (proportional change)	345	All	8	370.6	757.1	759.3

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