

# Estimation of selectivity, growth, and natural mortality in the assessment for EBS snow crab

Cody Szuwalski and Jack Turnock

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Juneau, AK

## and SSC comments

CPT had several comments and questions related to formatting and presentation from the September meeting:

- Review SAFE guidelines to make sure required tables and figures are present
- Plot the relative proportion of new to old shell males to see how important the lack of fit to old males really is
- Plot Bayesian posterior intervals for growth parameters
- Model 0 has to be last year's accepted model

Issues will be corrected in the SAFE document presented at the September meeting. The CPT also had several suggestions for potential model runs and expanded analyses, including:

- Estimate  $M$  for mature females
- Document rationale for prior on  $M$  for immature crab
- Try starting the assessment in 1982 to check the behavior of the survey  $q$ s when the first survey s are excluded
- Apply priors to the survey  $q$ s so they are somewhat constrained
- Provide more detailed MCMC chain diagnostics
- Extract bycatch mortality from the Tanner crab directed fisheries that is currently lumped into groundfish trawl bycatch (in a table in the assessment chapter, not necessarily in the model)

# Review

CMC diagnostics and Bayesian vs. maximum likelihood

- Bayesian methods designed to produce distributions

- Diagnostics suggests some problematic population processes

Survey selectivity

- Eliminate the first 4 years of data

- Think about how to use BSFRF data

Growth

- Piece-wise models causes problems, but contributes little to the model

- Unclear what the best model is

Natural mortality

- Mature female natural mortality should be estimated

- Priors for immature natural mortality should be revisited

# Bayesian methods vs. maximum likelihood

Get a TAC that accounts for scientific uncertainty

Bayesian methods:

Estimate parameters via ML

Put parameters into projection script

Get numbers at length for the final year in the projection script with error

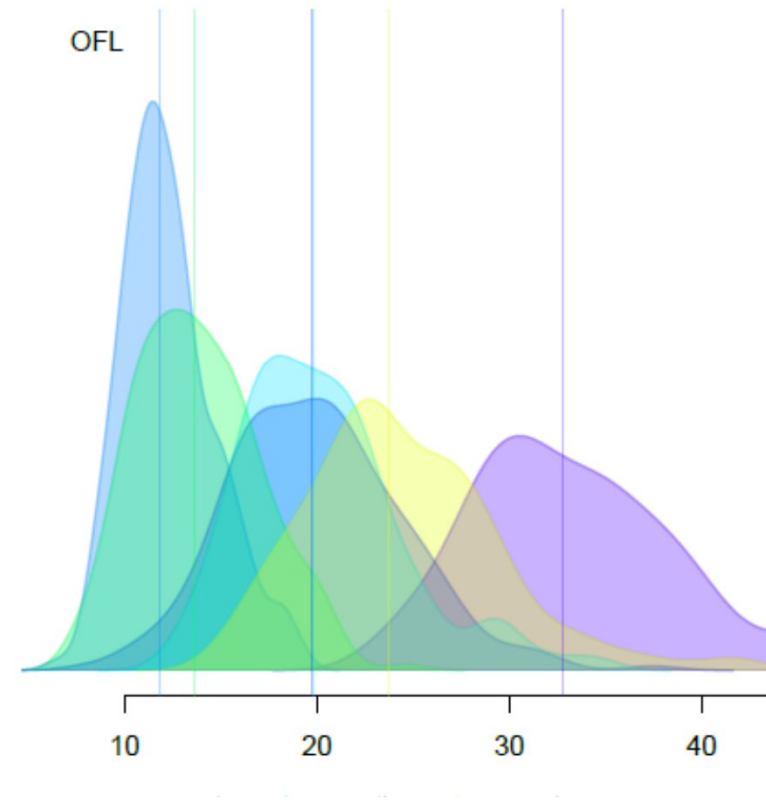
Simulate a distribution of the OFL based on the error added to the numbers at length

Compare with historical methods:

Parameter values are not perfectly known, but are assumed so.

Error added to numbers at length is arbitrary, but determines the distribution of the OFL.

Optimization was required to ensure MLEs were found





# Bayesian methods vs. maximum likelihood

Get a TAC that accounts for scientific uncertainty

Bayesian methods:

• Define a distribution for each parameter

• Do not require copying and pasting model output

• Posterior distributions of the OFL are a result of the uncertainty in parameter estimates

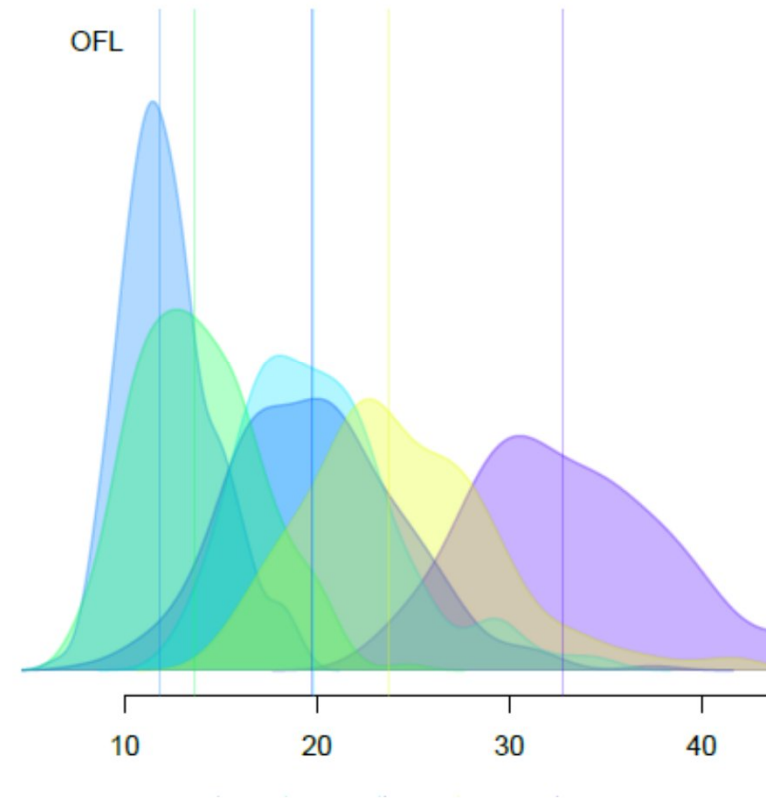
Working with Bayesian methods:

• Priors must be specified

• Time-consuming

• Knowing the model has converged is difficult (though there are many diagnostics to identify non-convergence)

• Reliant on the var/covar matrix; therefore reliant on an appropriately specified model



# Bayesian methods vs. maximum likelihood

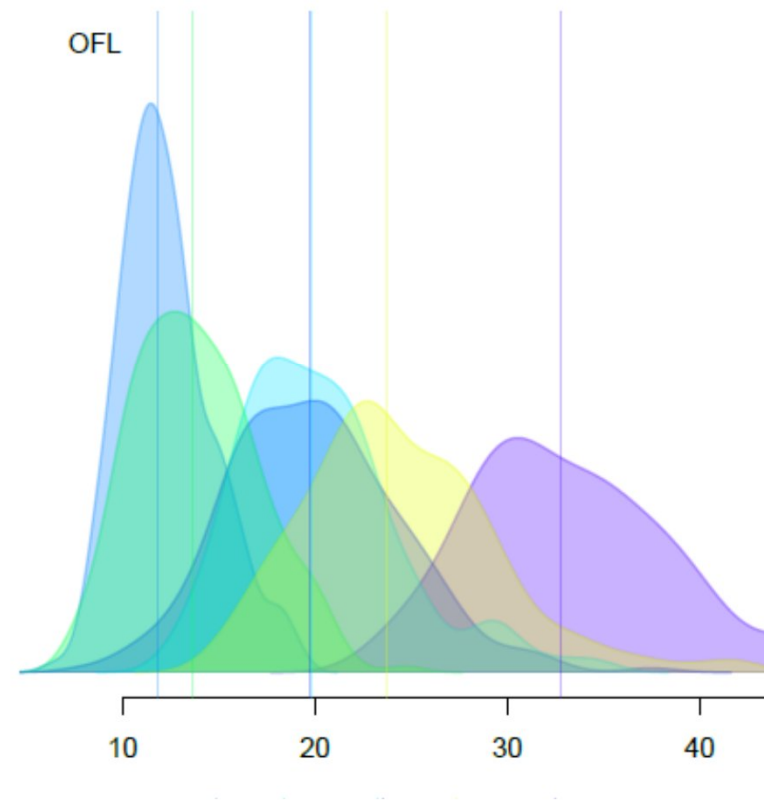
possible modification of historical methods  
avoids Bayesian methods:

simulate the OFL during the fitting of the  
model instead of in the report section and  
include the OFL as a sd\_report variable

can produce a distribution of the OFL with its  
simulated standard deviation

used this once with the 'Trim data' model,  
the model blew up.

takes forever to fit the model because  
reference points and the OFL have to be  
simulated in every step.



# MCMC diagnostics

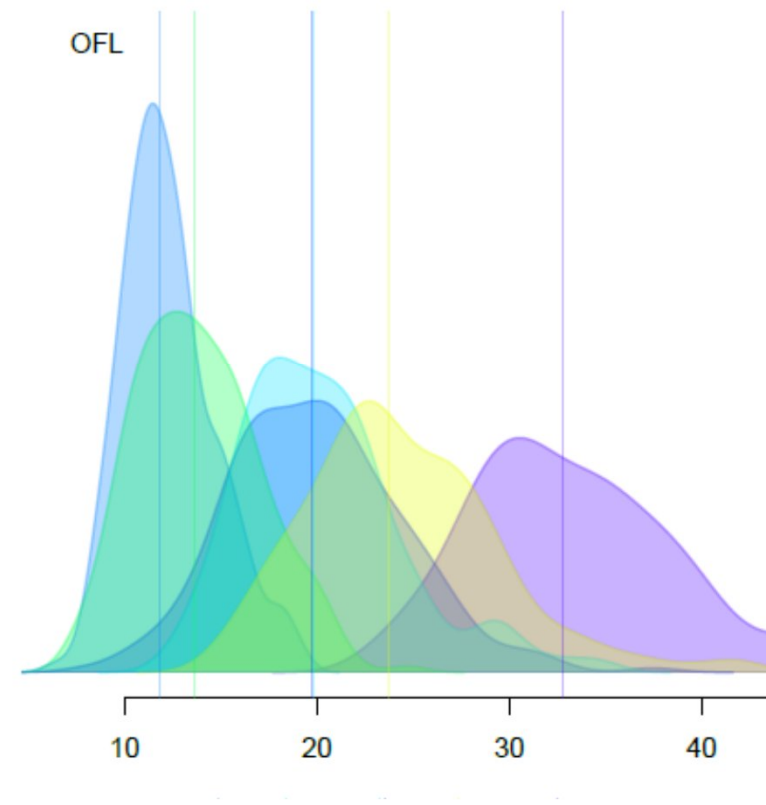
MCMC is used to 'build' the distributions of parameters and derived quantities

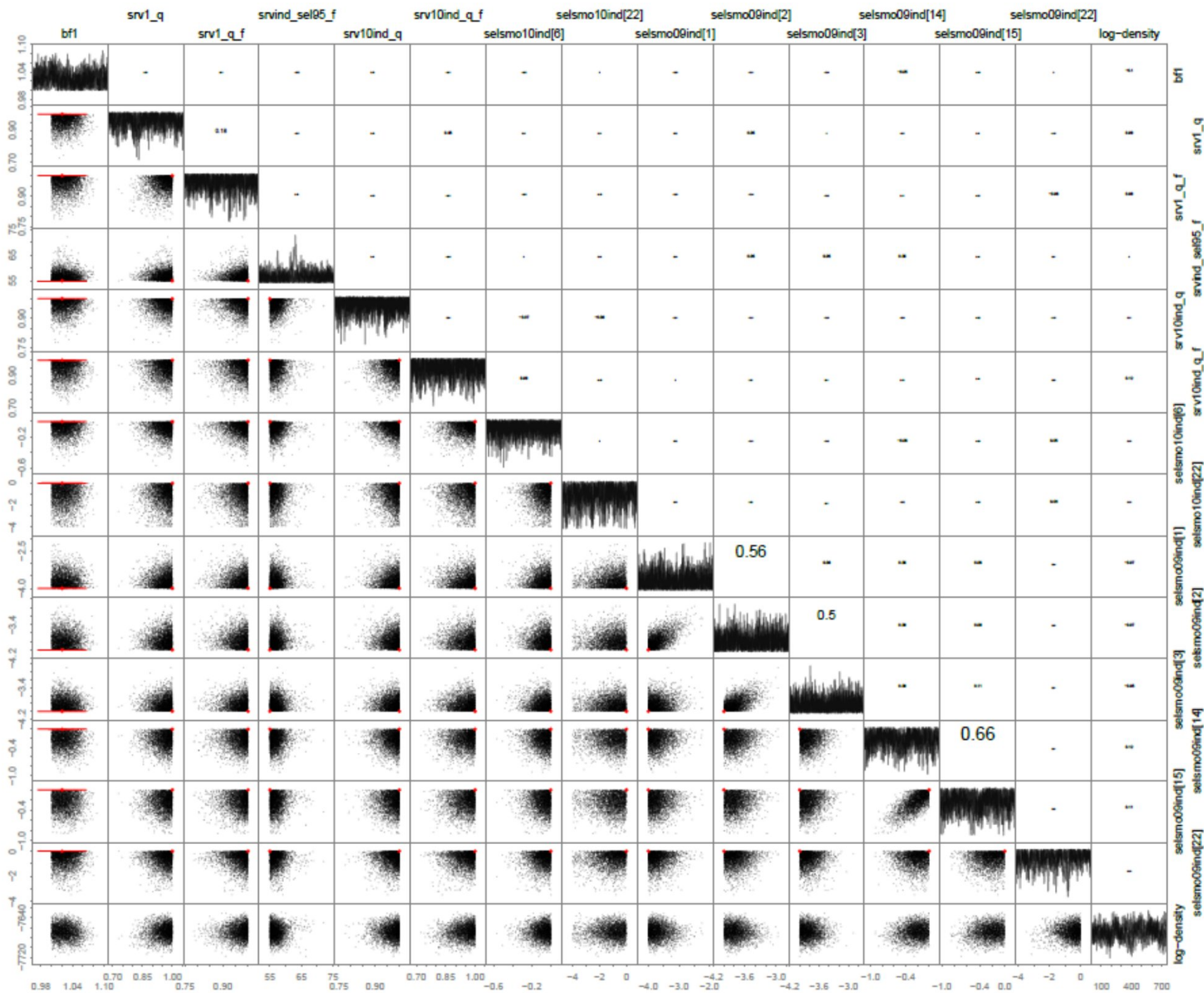
Diagnostics check for appropriate specification of the model and var/covar matrix (used to tune the posterior)

Autocorrelation in the traces for parameters and objective function (mixing)

Parameters hitting bounds

If there are problems in these diagnostics, be cautious about inference from the model









# Diagnostic summary

General processes have problem parameters

mixing:

Growth parameters

rec devs

initial numbers at length

and hitting

Growth parameters

survey selectivity (NMFS) during era 1

industry survey selectivity parameters

Start with slow mixing by using really long chains last year, but this takes a very long time

Methods for adjustment

priors on parameters hitting their bounds

reformulating the model

excluding problematic periods of data

# Estimating survey selectivity

Historical methods

Data

Model runs

Model results

Fits

OFL and reference points

Processes influenced

Recommendations



# Historical methods

## Logistic selectivity

Three eras

1974-1981: different gear

1982-1988: different area

1989-present: current

## Catchability coefficient ( $q$ )

Changes in estimates over time

Era 1 has always been fixed at 1

1985

Q: modulates the impact of catch on the survey index

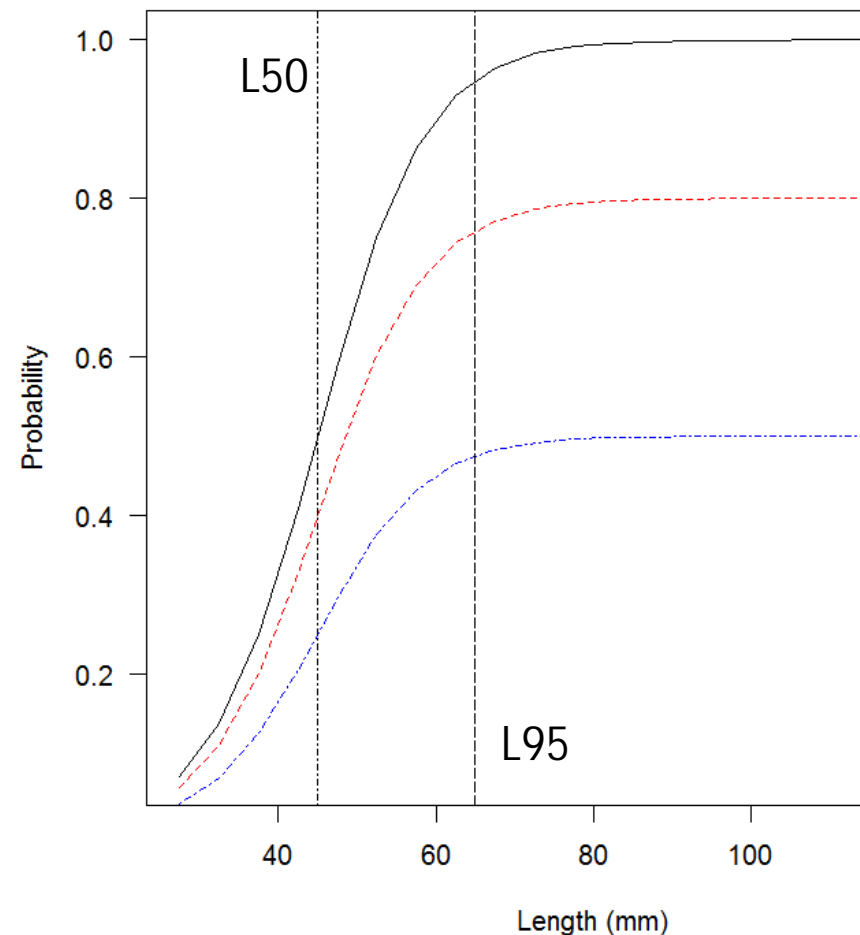
Foreign fleets were excluded starting 1980, so it's not clear if the catches are fully represented in era 1

Q: in the first era is consistently estimated on its bounds and anchors the catchability in the other era

Are the survey areas appropriately chosen?

Are there alternate sensible configurations?

$$S_l = \frac{q}{1 + e^{(-\ln(19)) \frac{L_l - L_{95}}{L_{50} - L_{95}}}}$$



# Historical methods

Logistic selectivity

Three eras

1978-1981: different gear

1982-1988: different area

1989-present: current

Logistic selectivity coefficient:  $S_l$

Change in estimates over time

Era 1 has always been fixed at 1

Why?

○ To evaluate the impact of catch on the survey index

○ Foreign fleets were excluded starting 1989, so it's not clear if the catches are fully representative

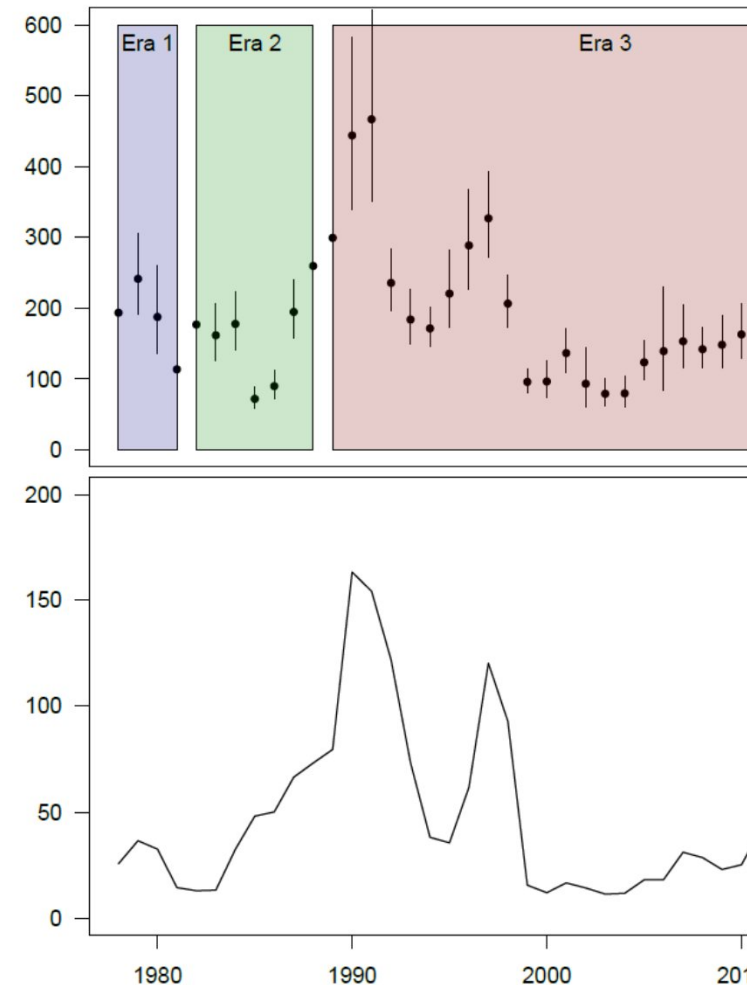
○ In the first era is consistently estimated on US outputs

and anchors the catchability in the other era

○ The survey areas are approximately constant

○ There are no other sensible configurations

$$S_l = \frac{q}{1 + e^{(-\ln(19)) \frac{L_{t+1} - \beta_{t+1}}{\beta_{t+1} - \beta_t}}}$$



# Historical methods

Logistic selectivity

Three eras

1978-1981: different gear

1982-1988: different area

1989-present: current

Catchability coefficient (q)

Changes in estimates over time

Era 1 has always been fixed at 1

Issues

○ To evaluate the impact of catch on the survey index

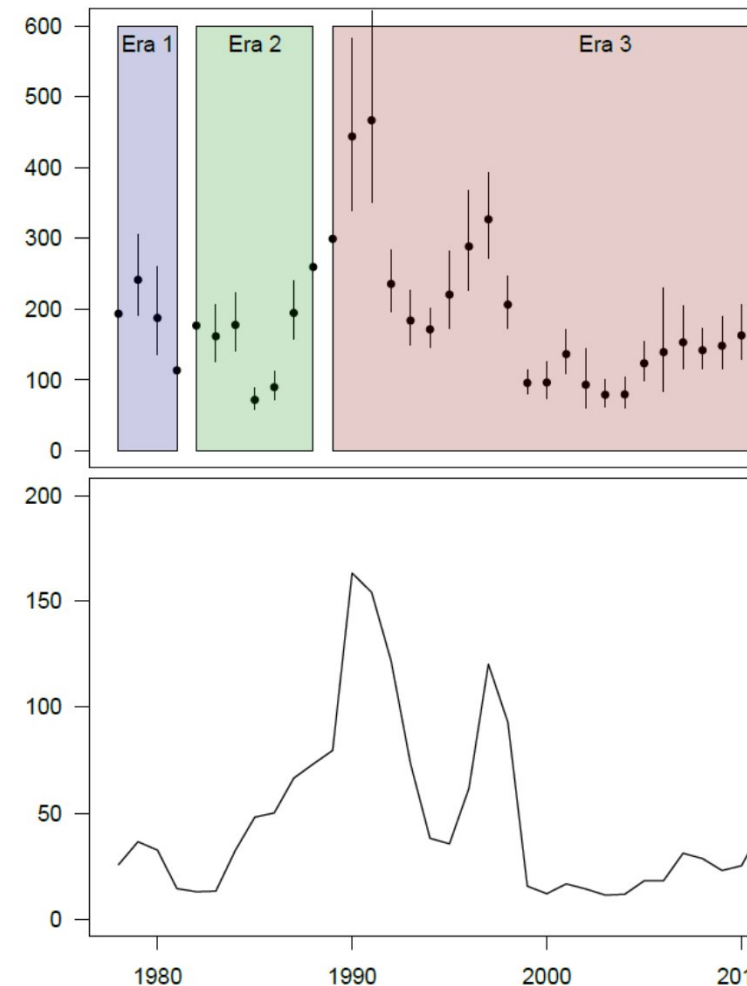
○ Foreign fleets were excluded starting 1981, so it's not clear if the catches are fully representative in the first era

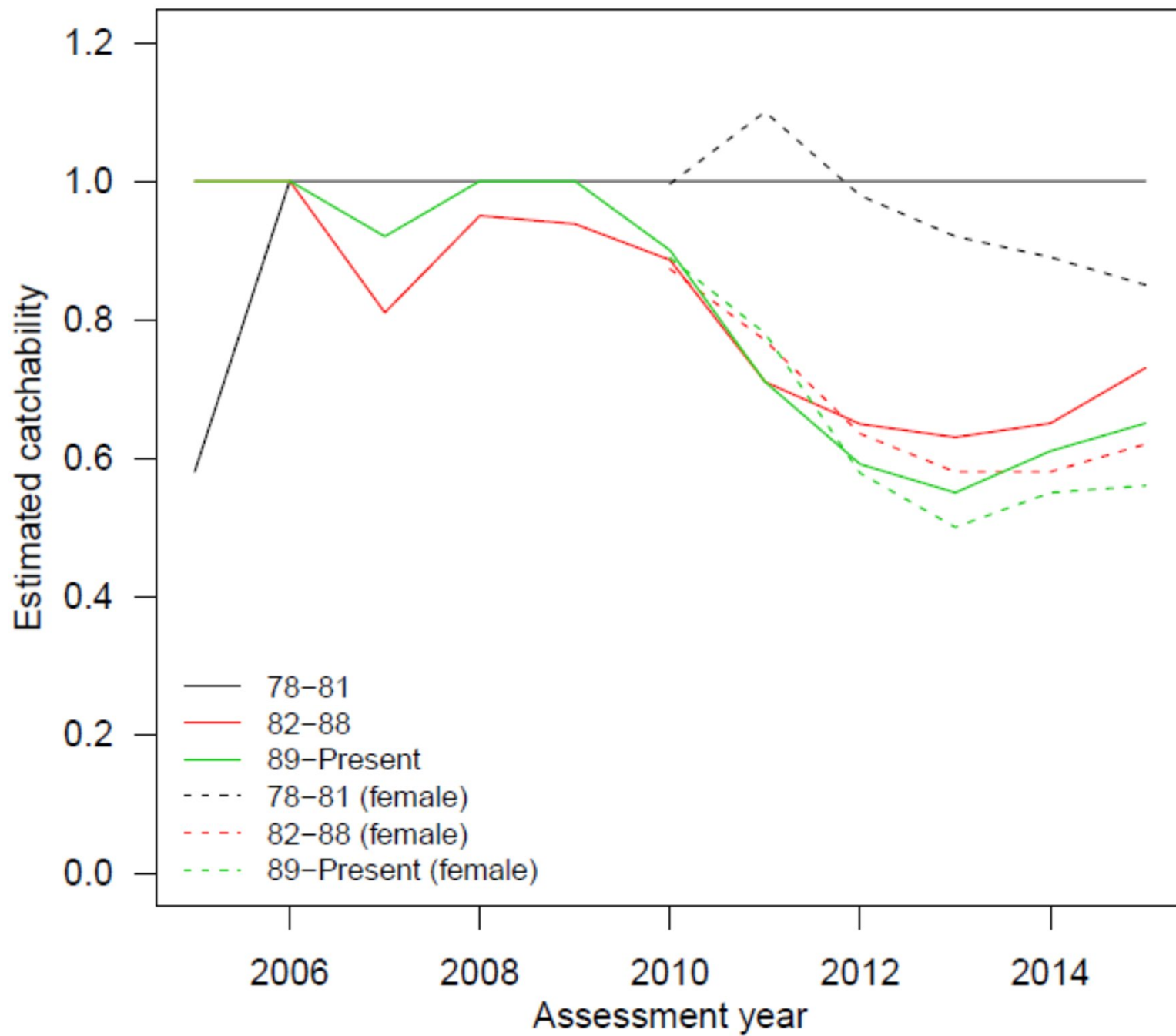
○ q in the first era is consistently estimated on its bounds and anchors the catchability in the other era

○ Is the survey area appropriately cross-sectional?

○ Are there alternate sensible configurations?

$$S_l = \frac{q}{1 + e^{(-\ln(19)) \frac{L_l - \beta_{l+1}}{\beta_{l+1} - \beta_l}}}$$





# Historical methods

Logistic selectivity

Three eras

1978-1981: different gear

1982-1988: different area

1989-present: current

Catchability coefficient ( $q$ )

Changes in estimates over time

Era 1 has always been fixed at 1

Issues

$Q$  modulates the impact of catch on the survey index

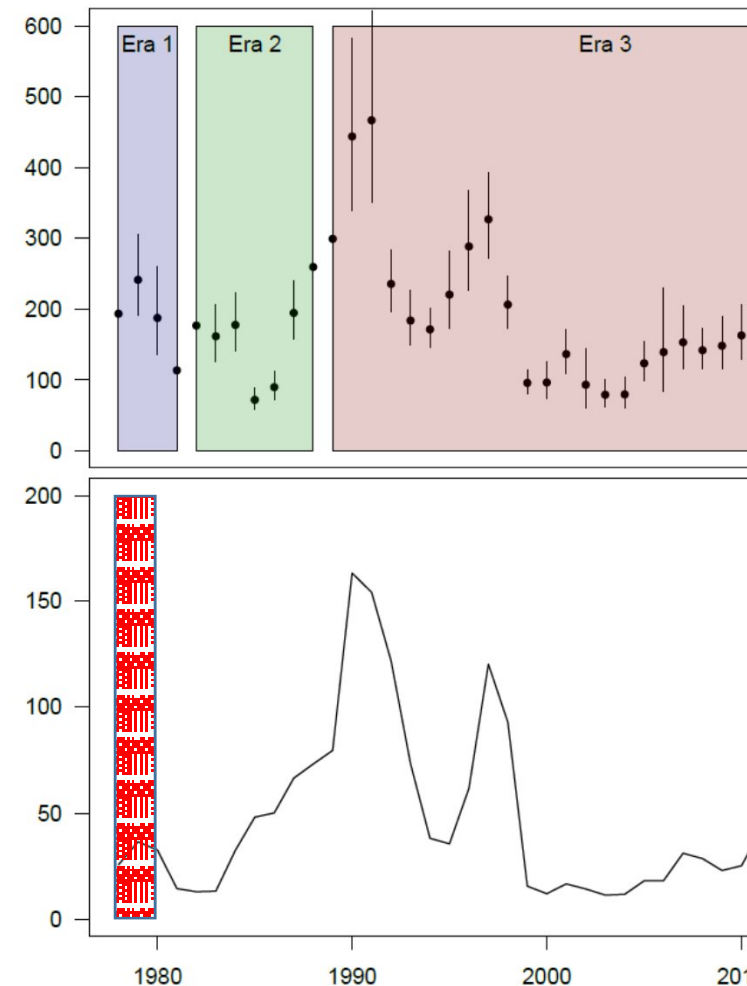
Foreign fleets were excluded starting 1980, so it's not clear if the catches are fully represented in era 1

$Q$  in the first era is consistently estimated on its bounds and anchors the catchability in the other era

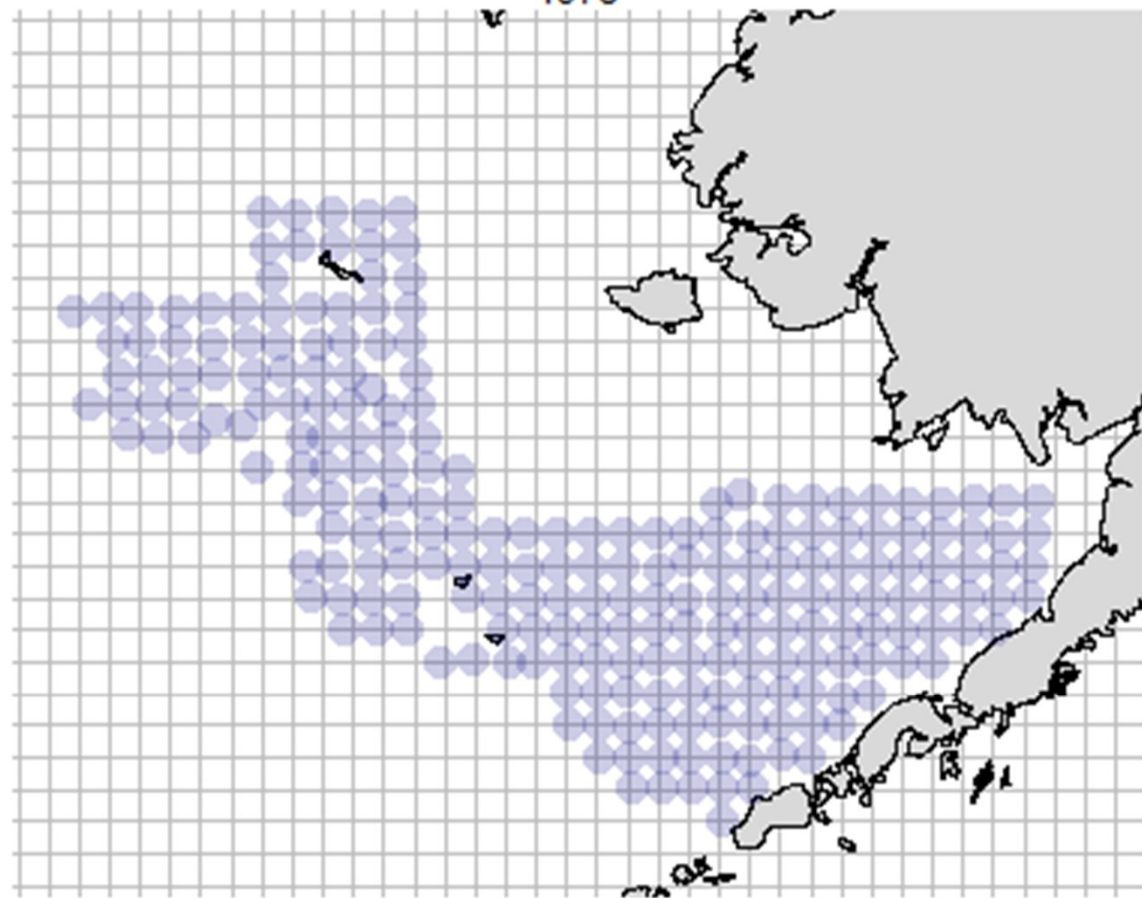
Are the survey eras appropriately chosen?

Are there alternate sensible configurations?

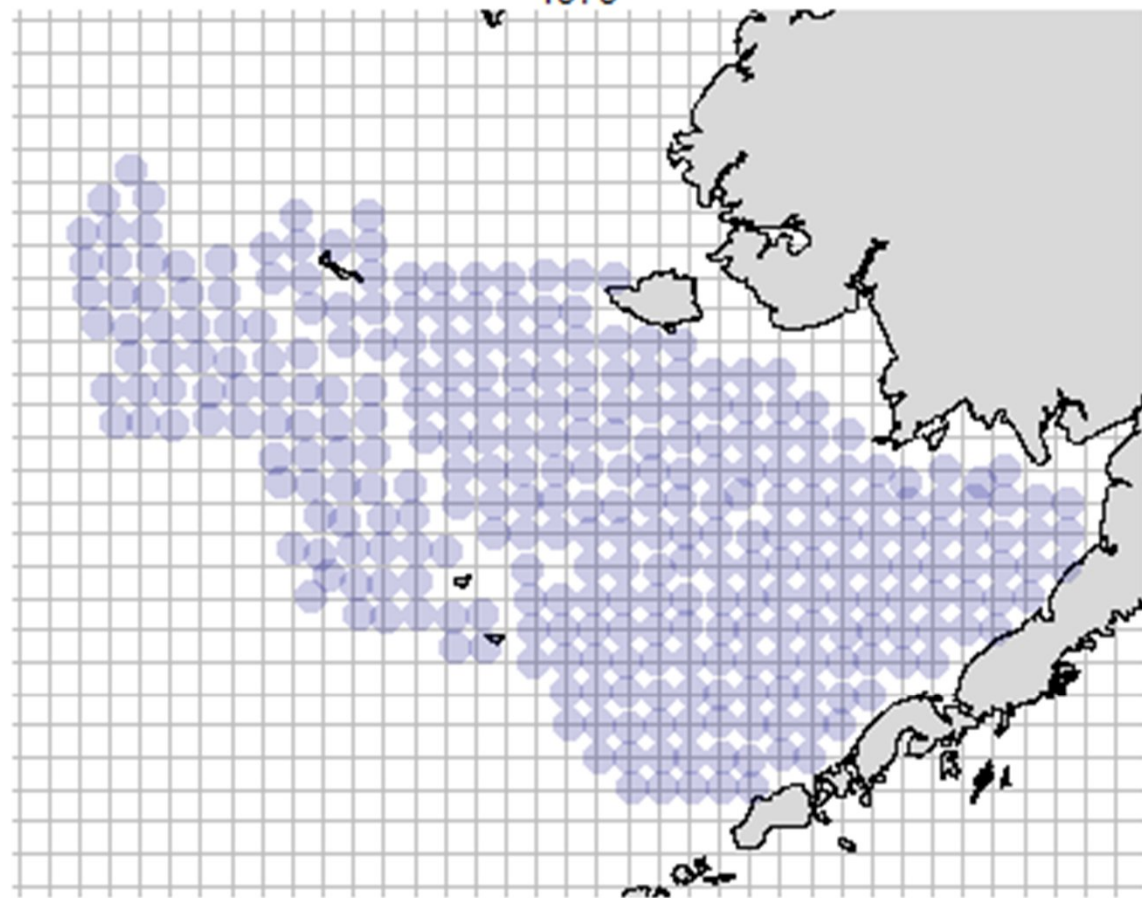
$$S_l = \frac{q}{1 + e^{(-\ln(19)) \frac{L_{??} - \beta_{??}}{\beta_{??} - \beta_{??}}}}$$



1978

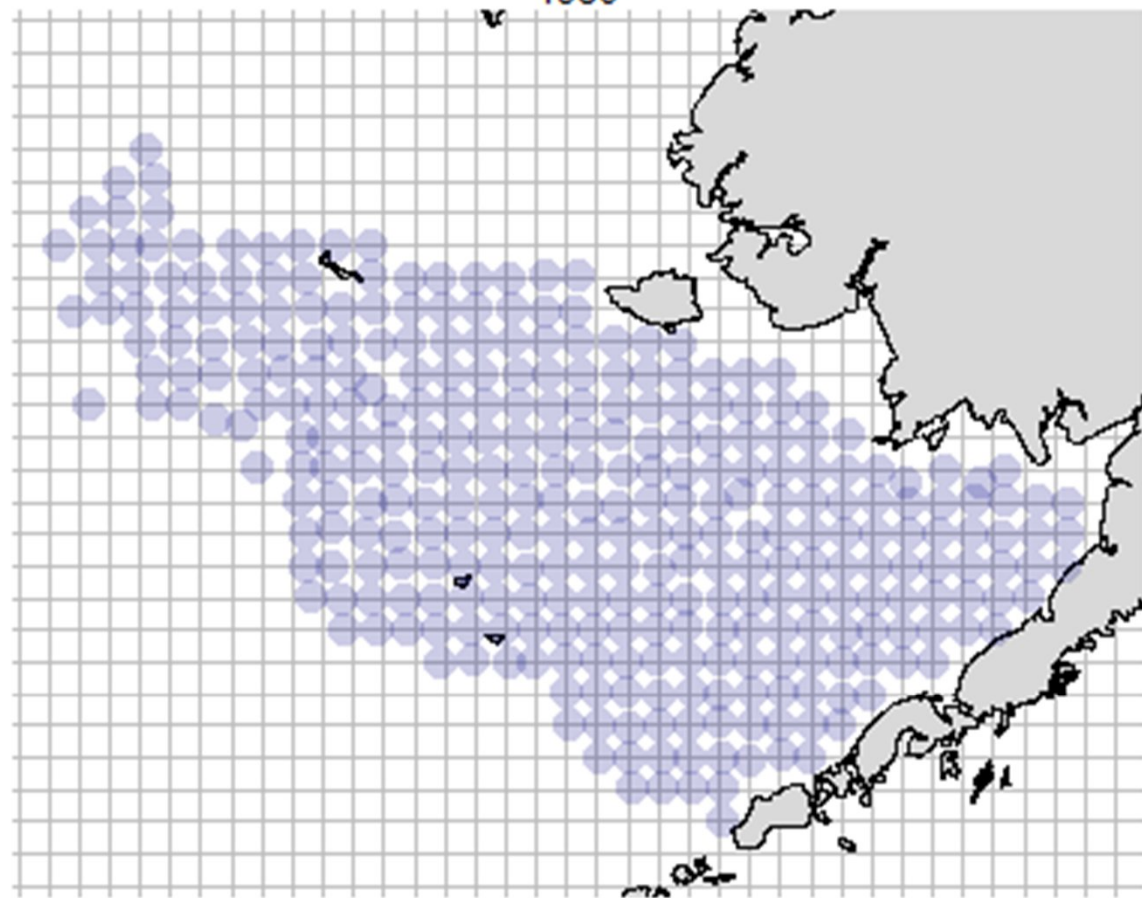


1979

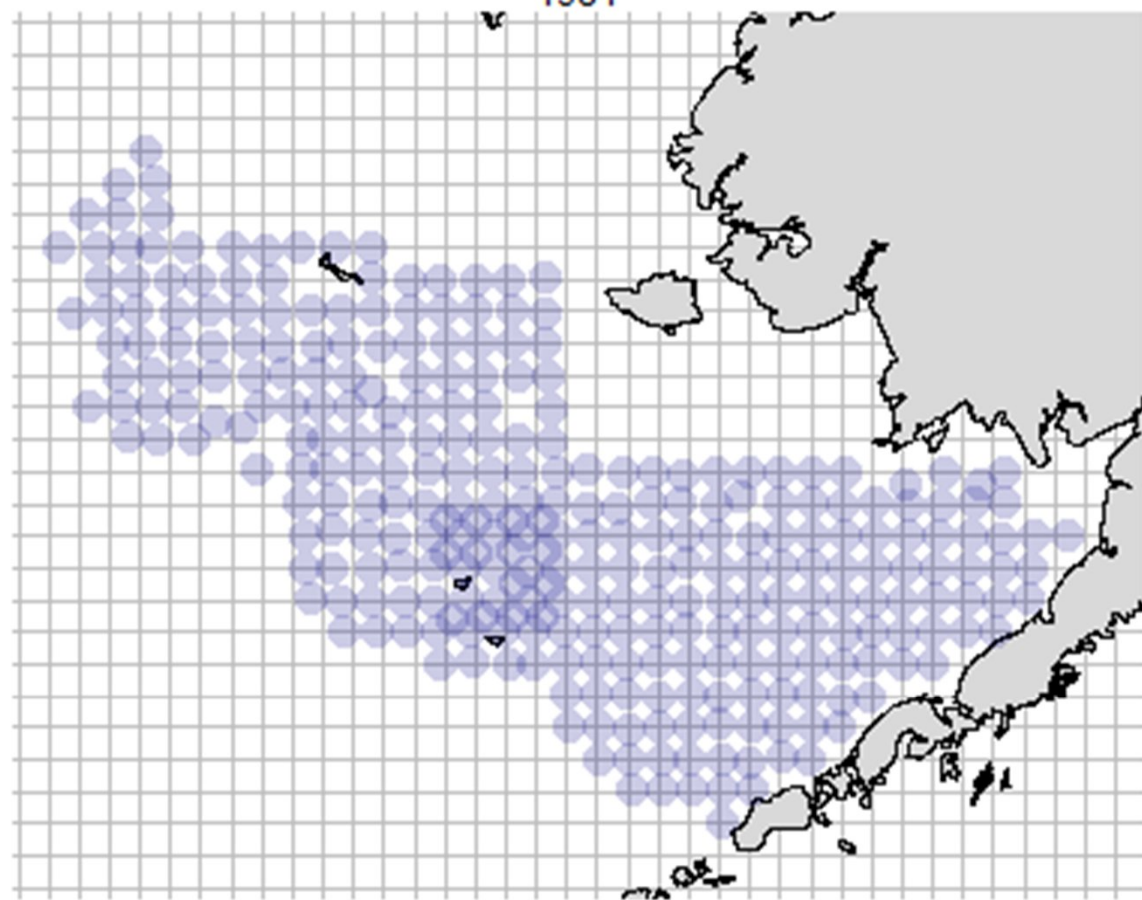




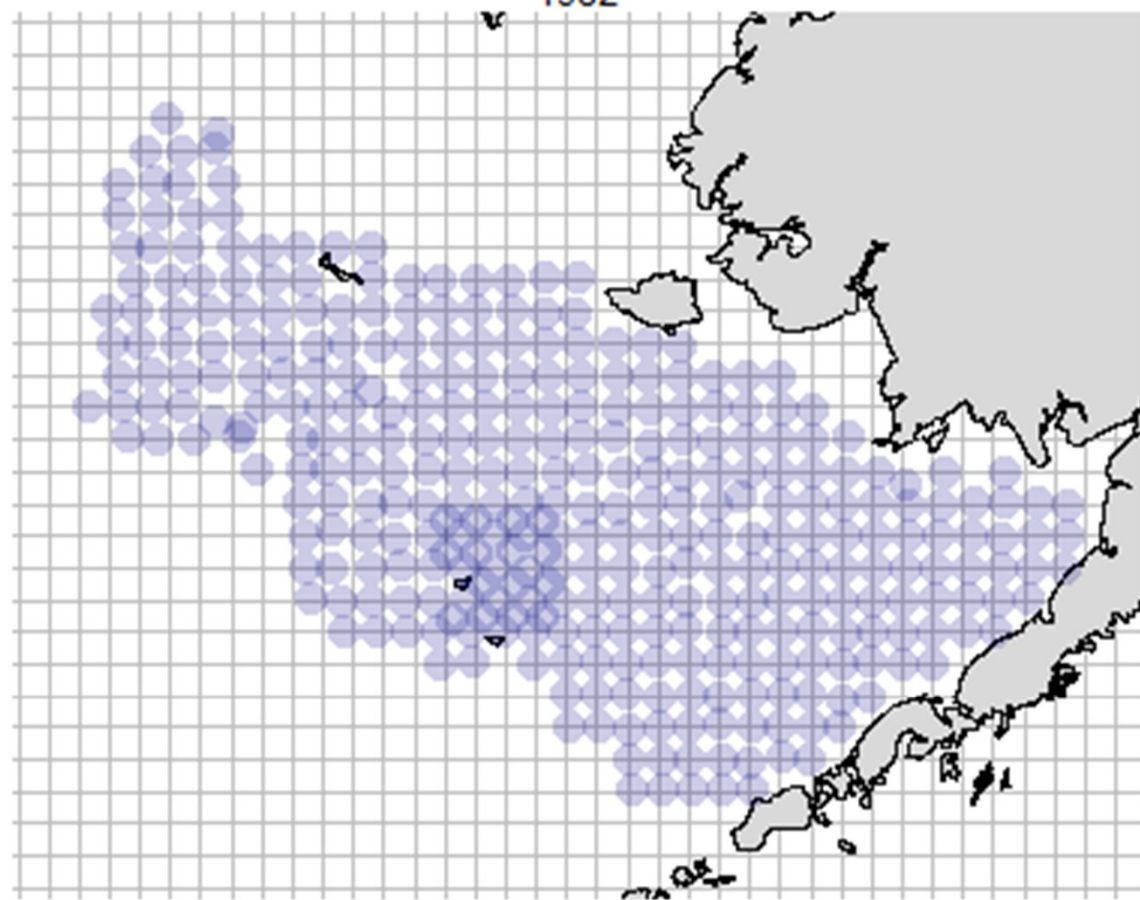
1980



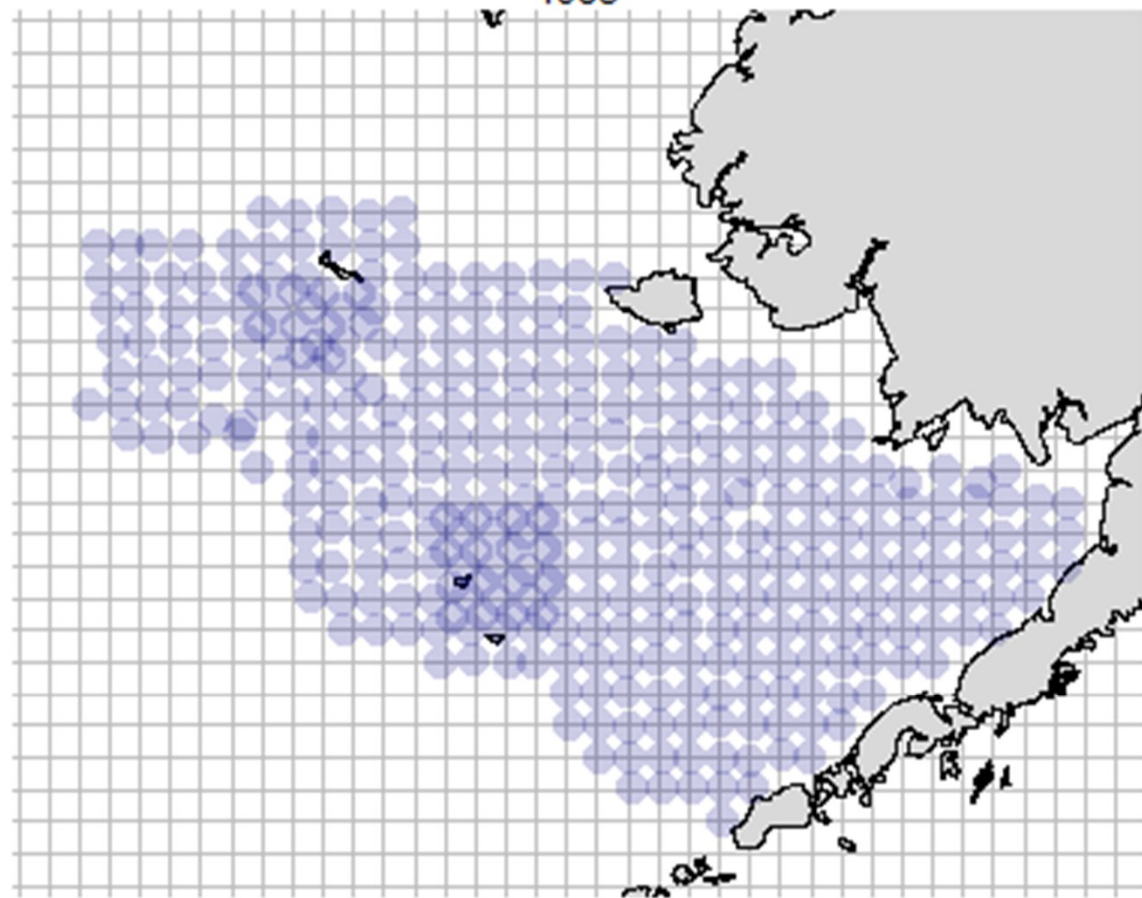
1981



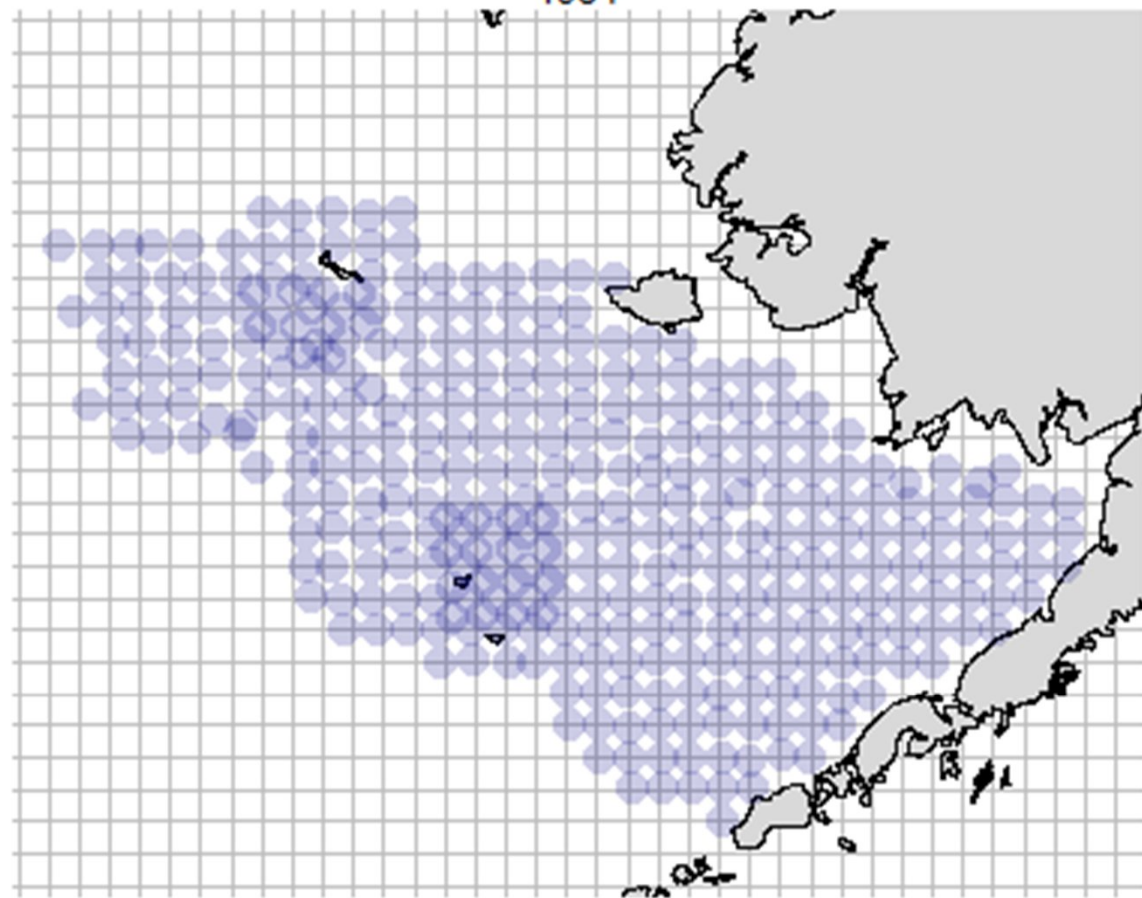
1982



1983

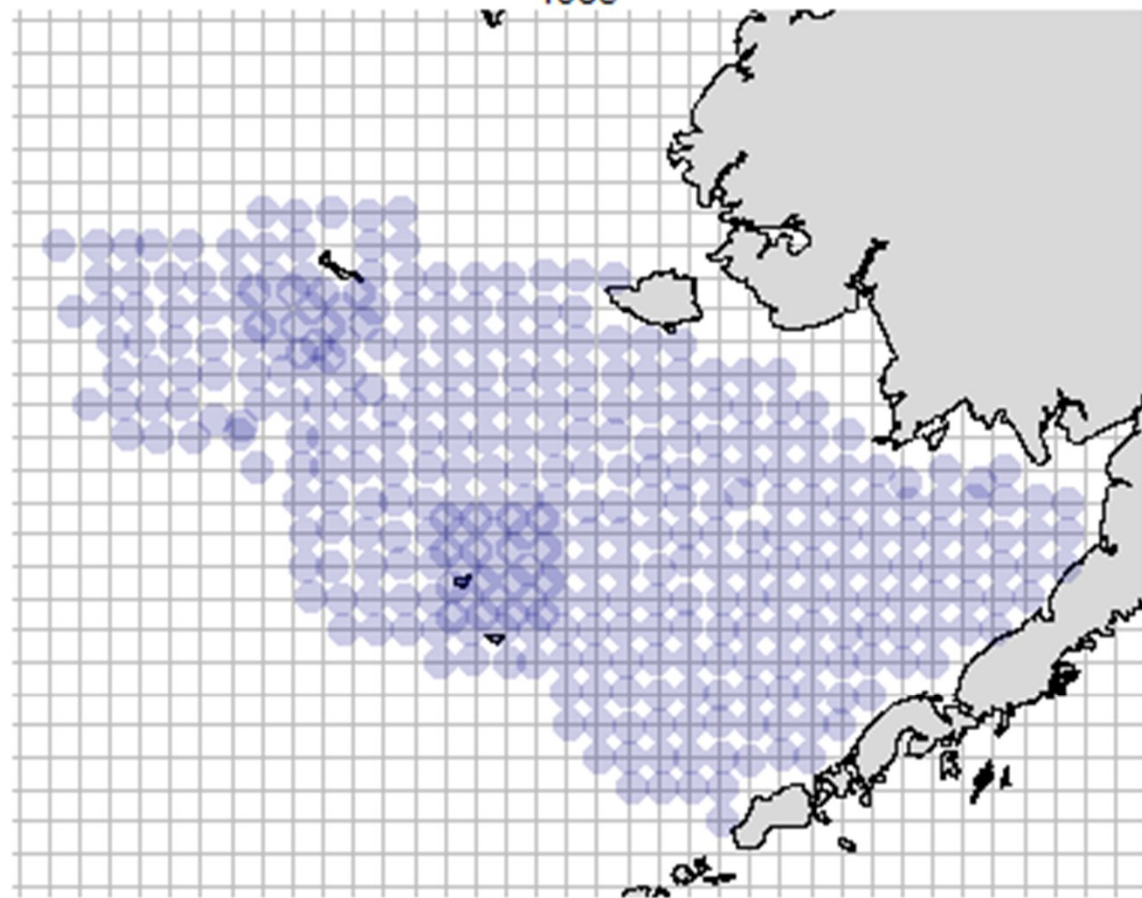


1984

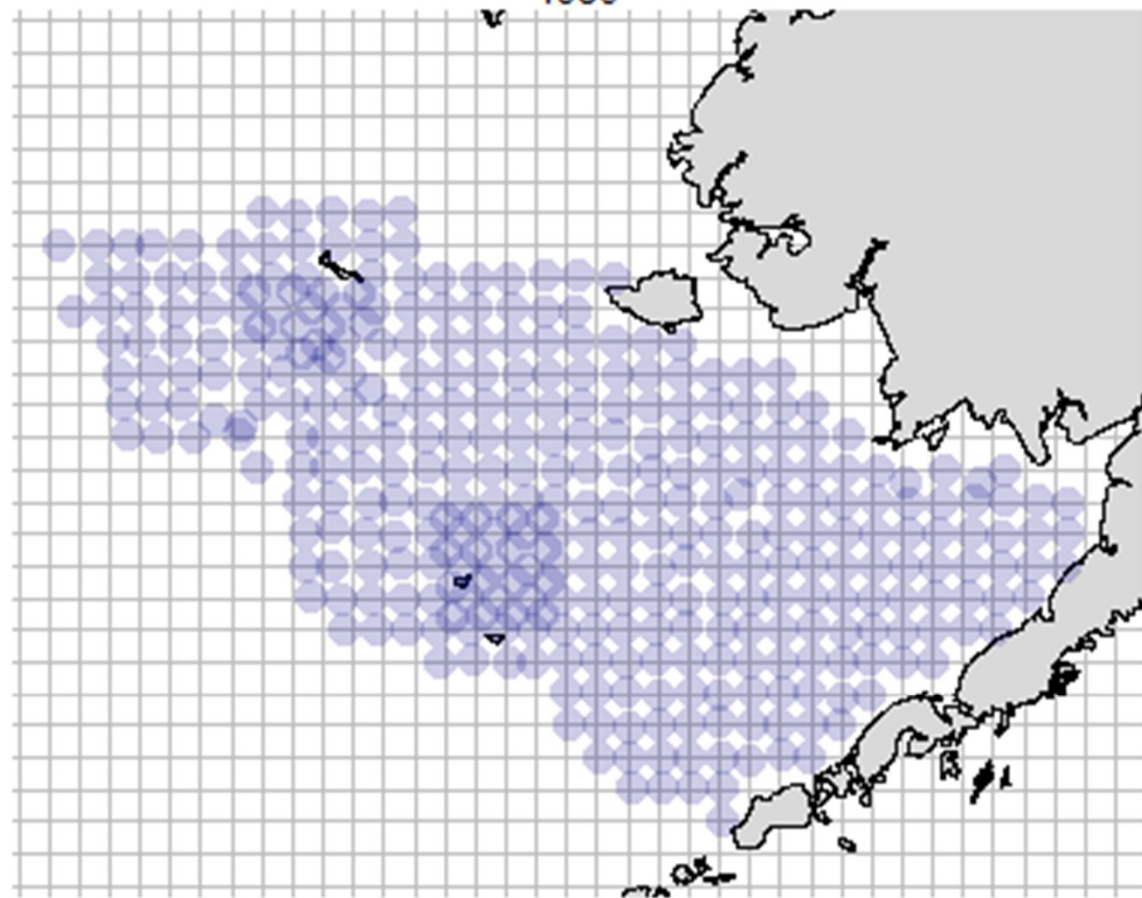




1985

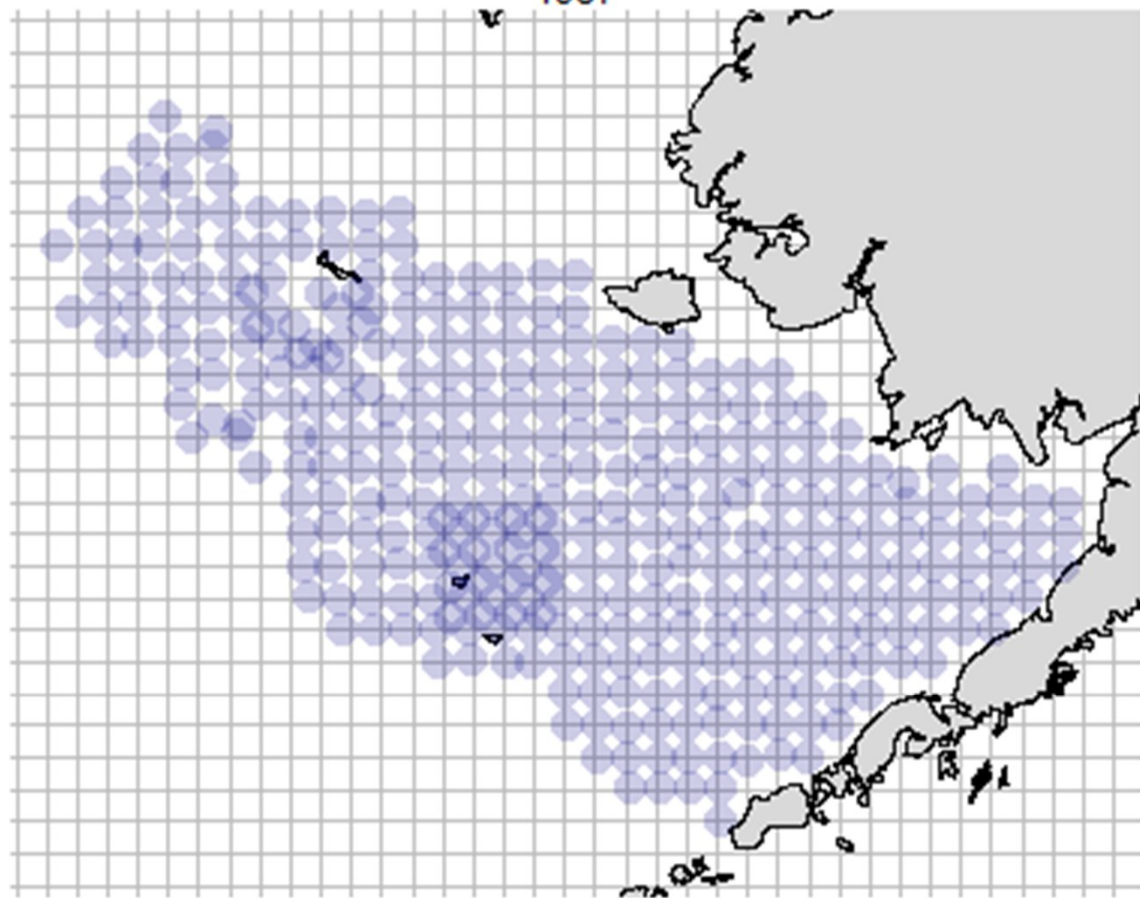


1986

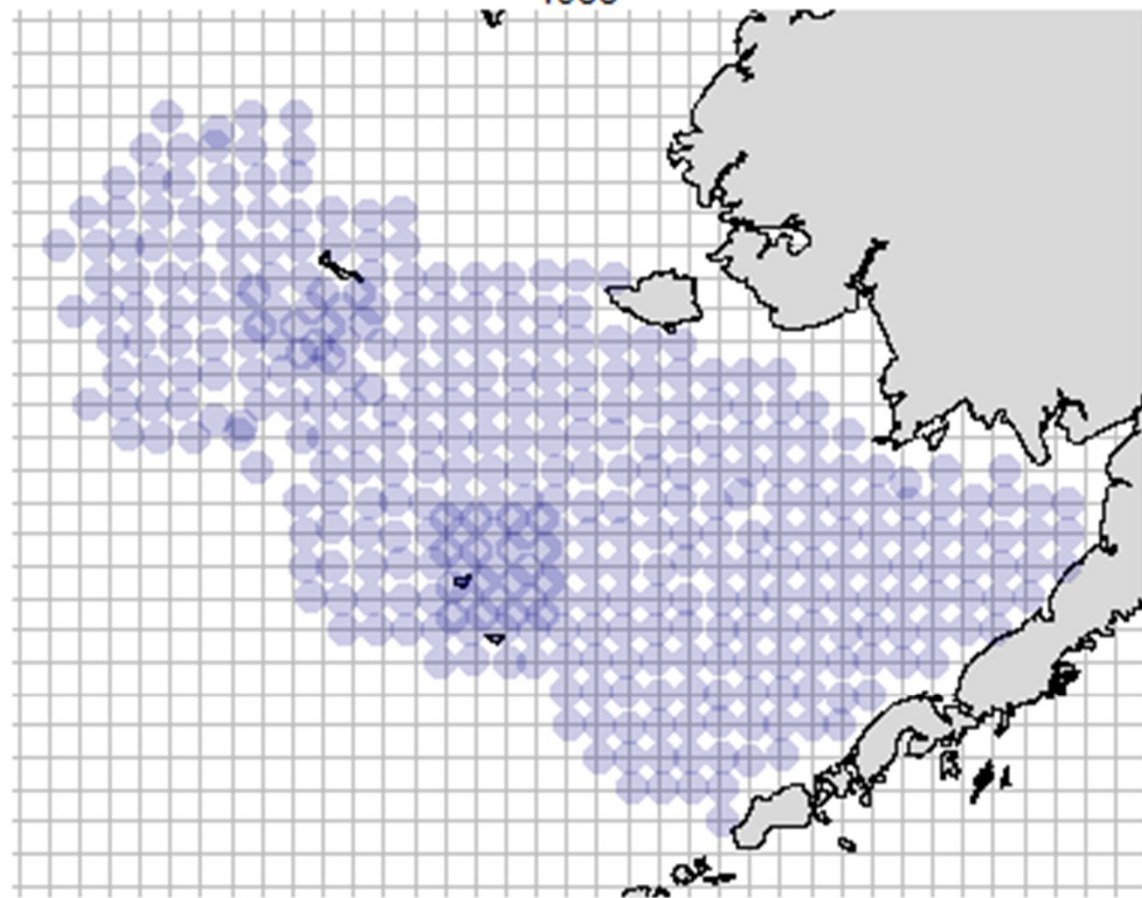




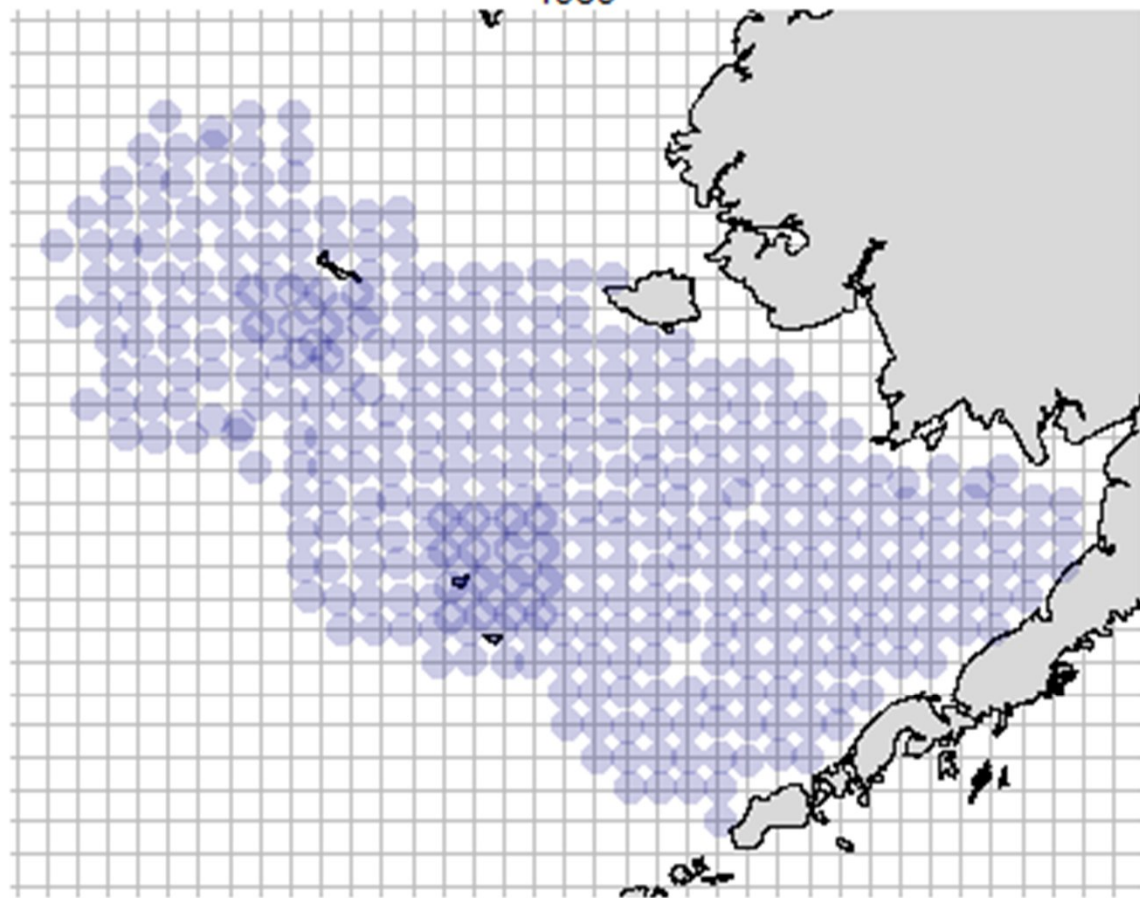
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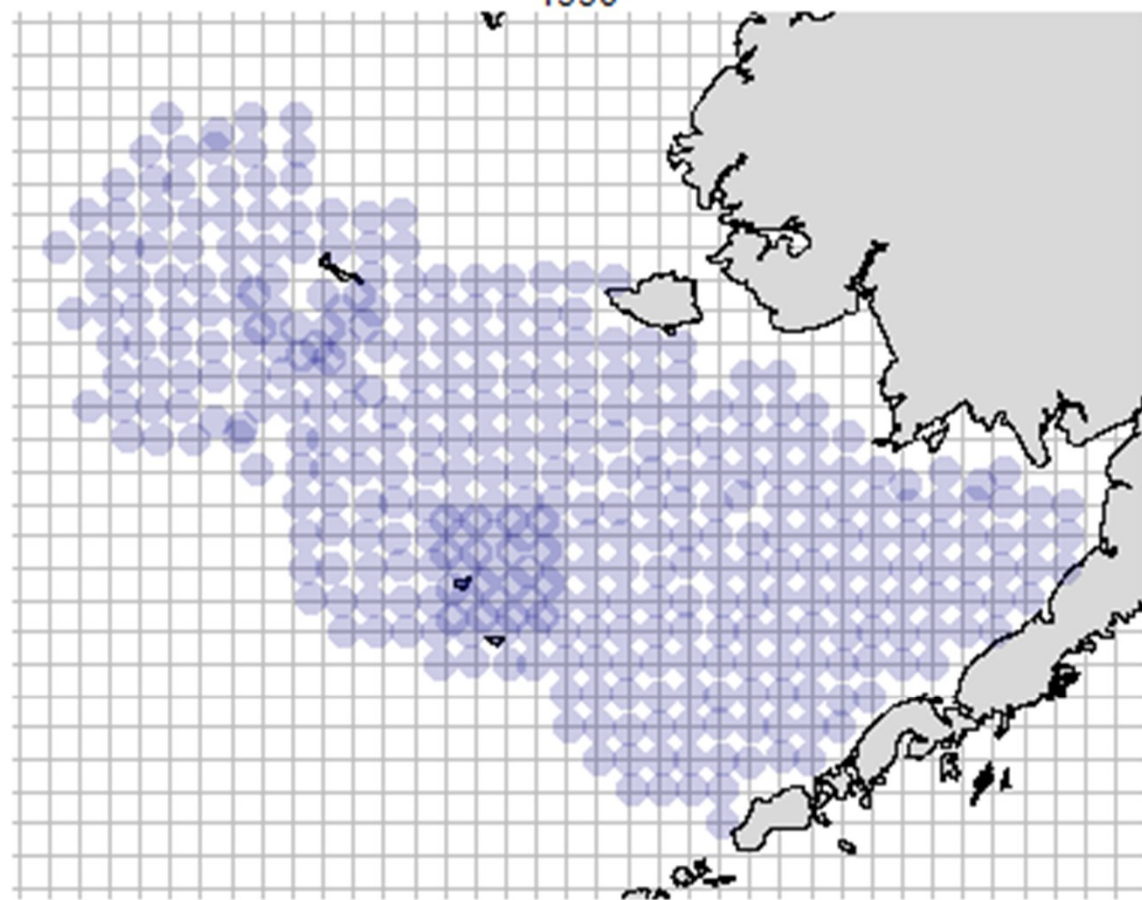
1988



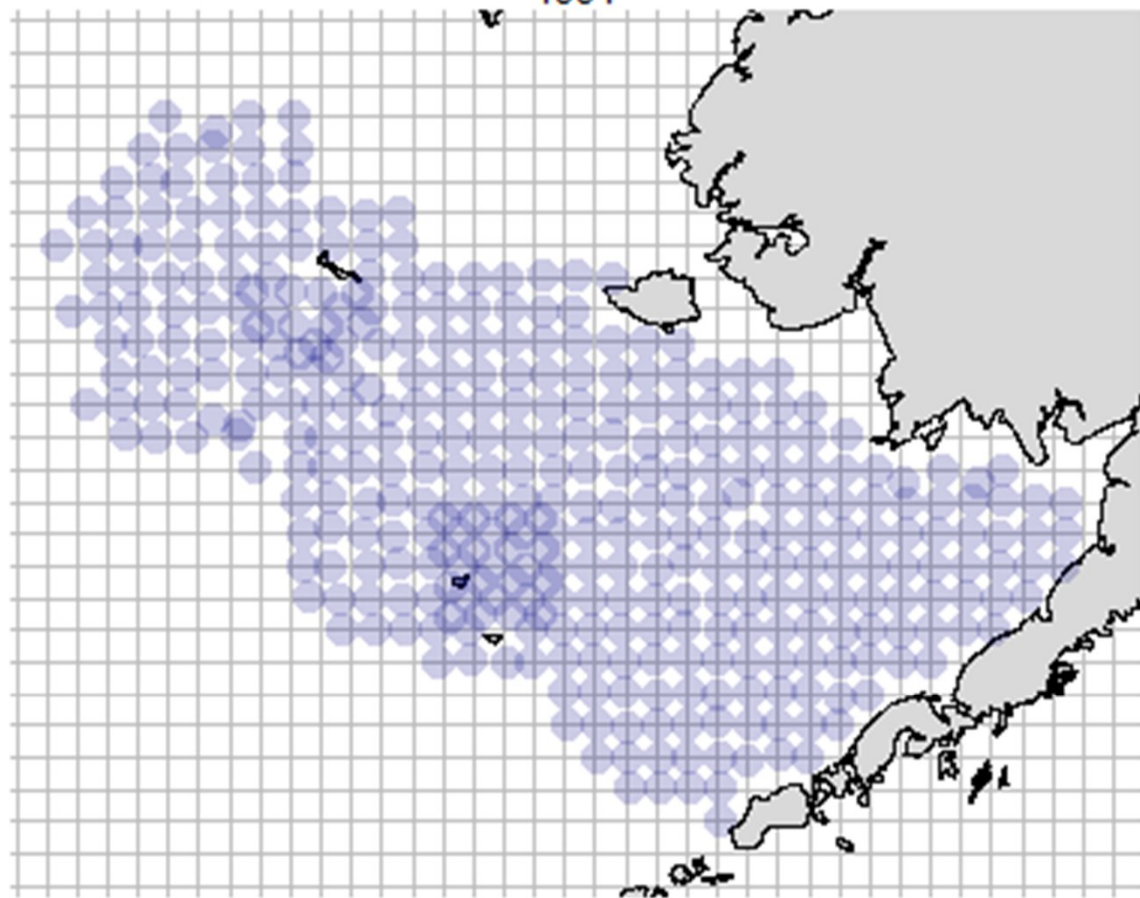
1989



1990

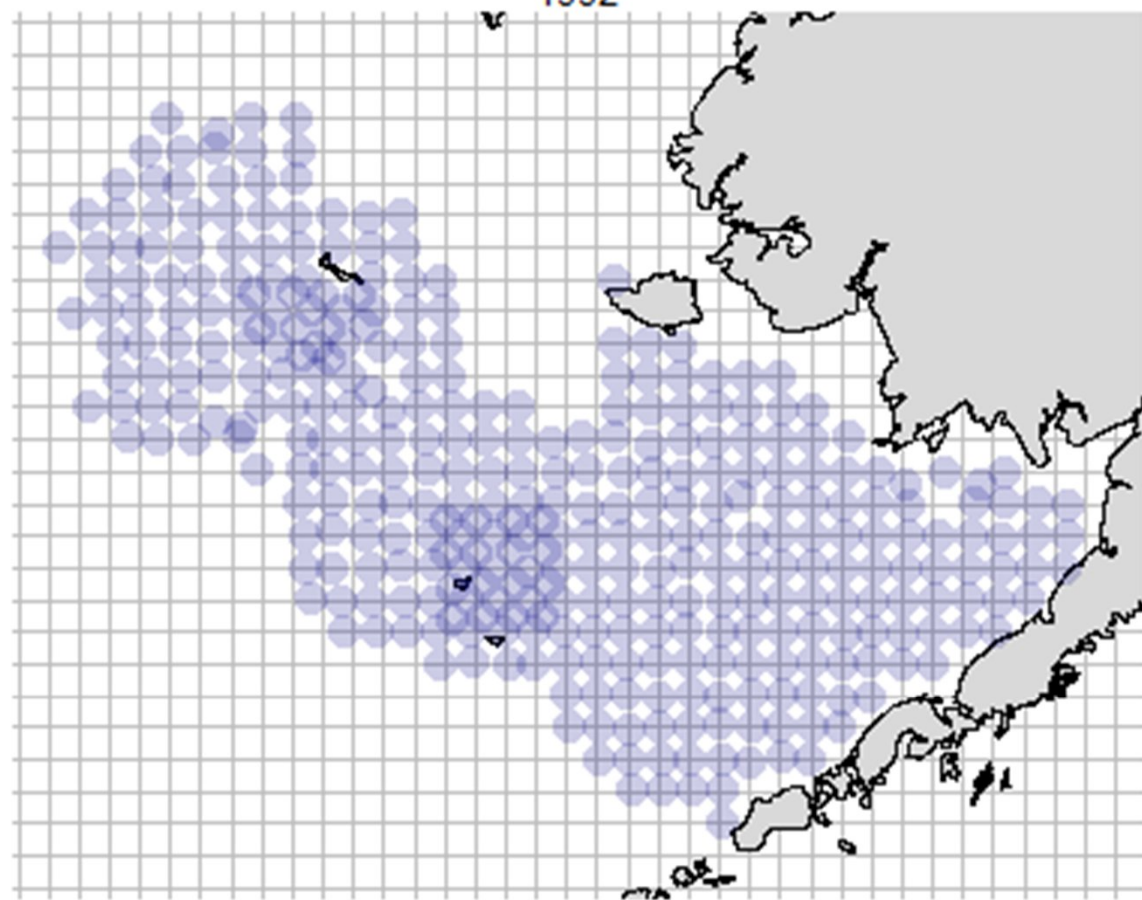


1991

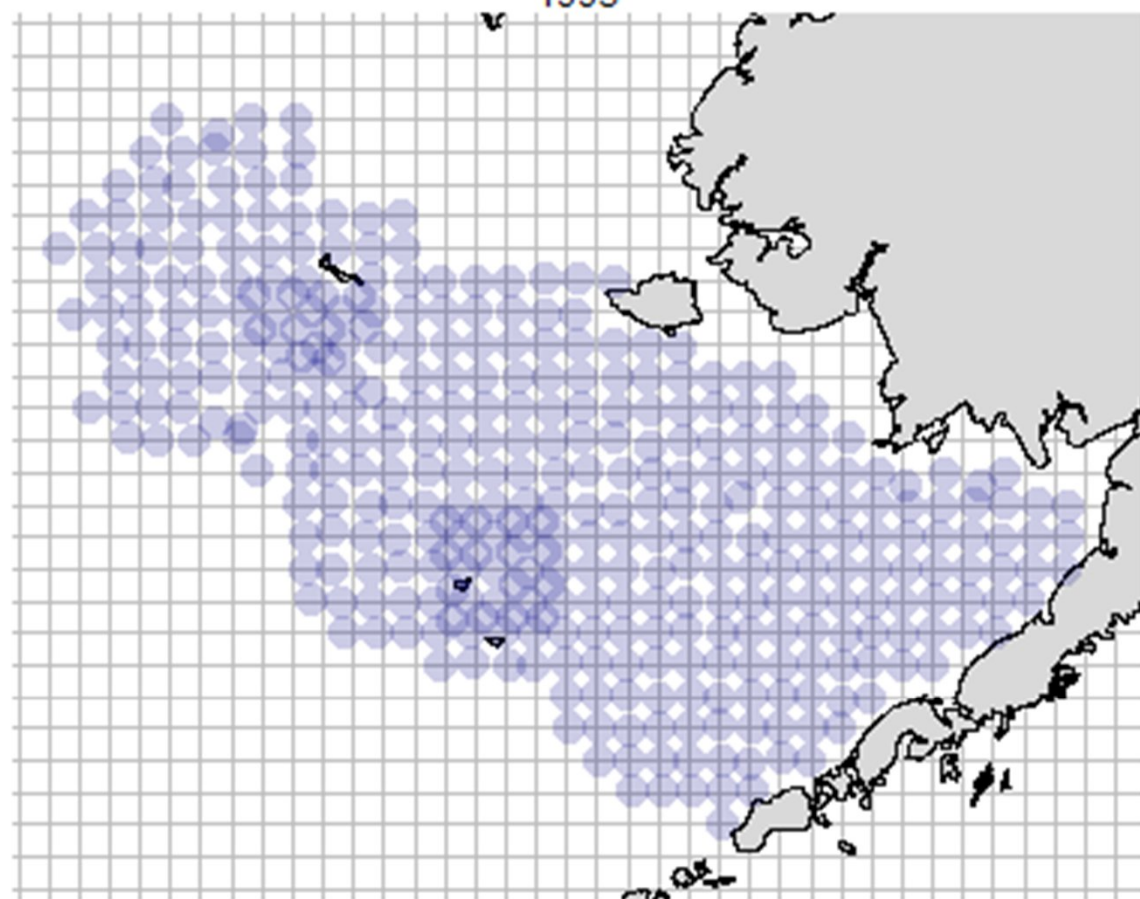




1992

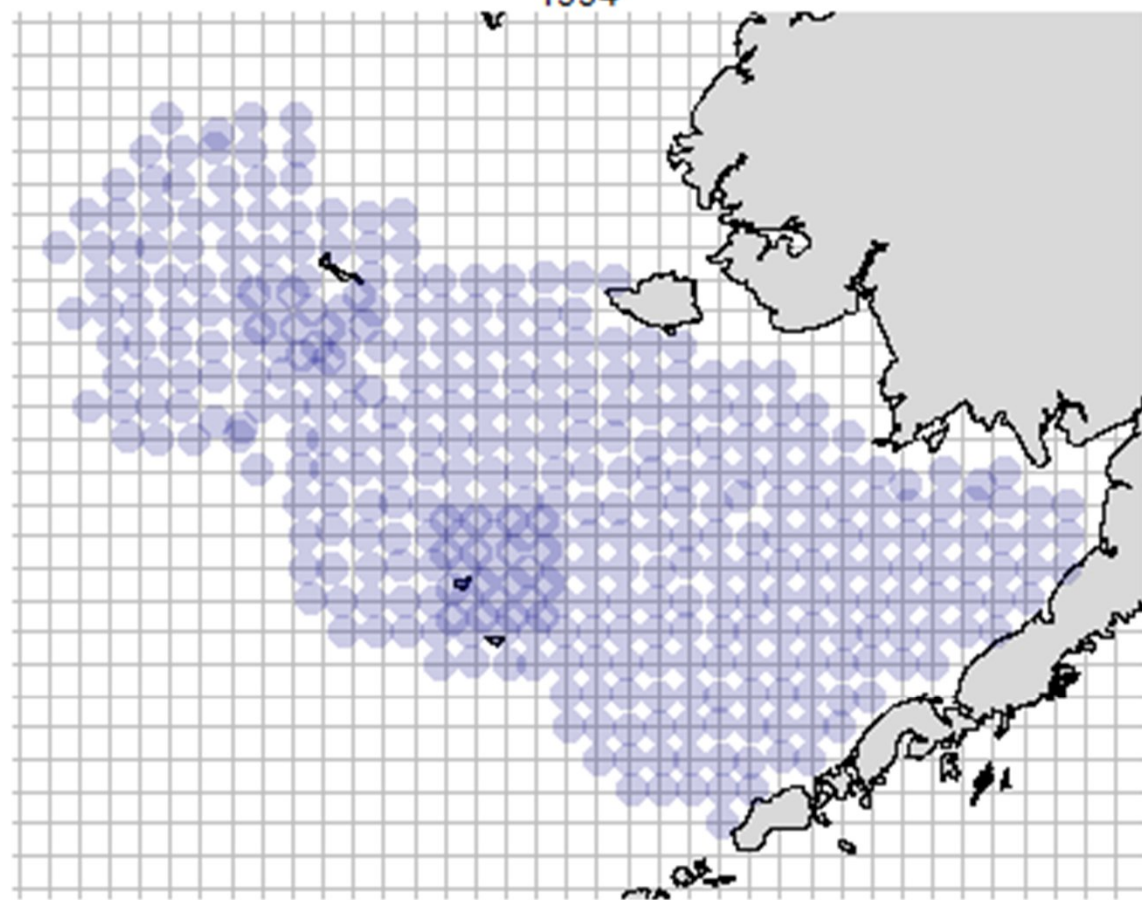


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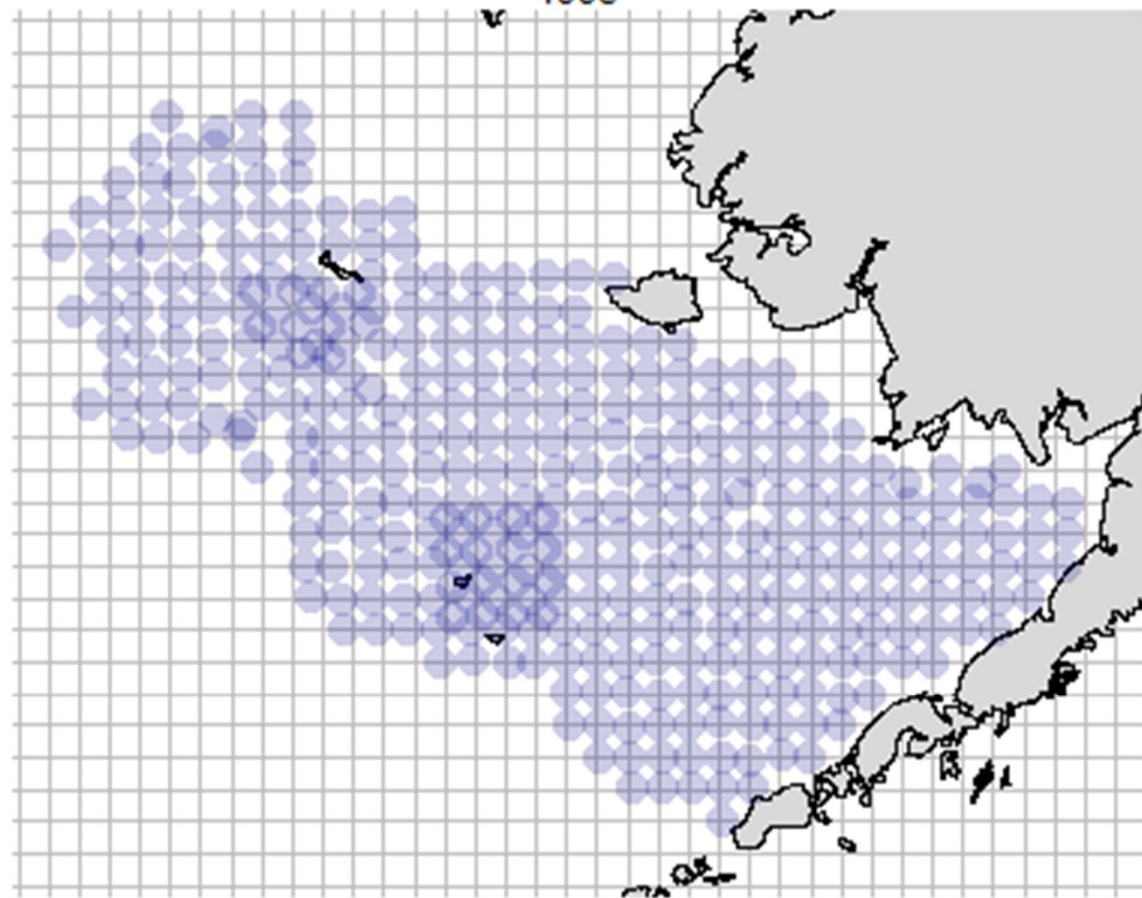




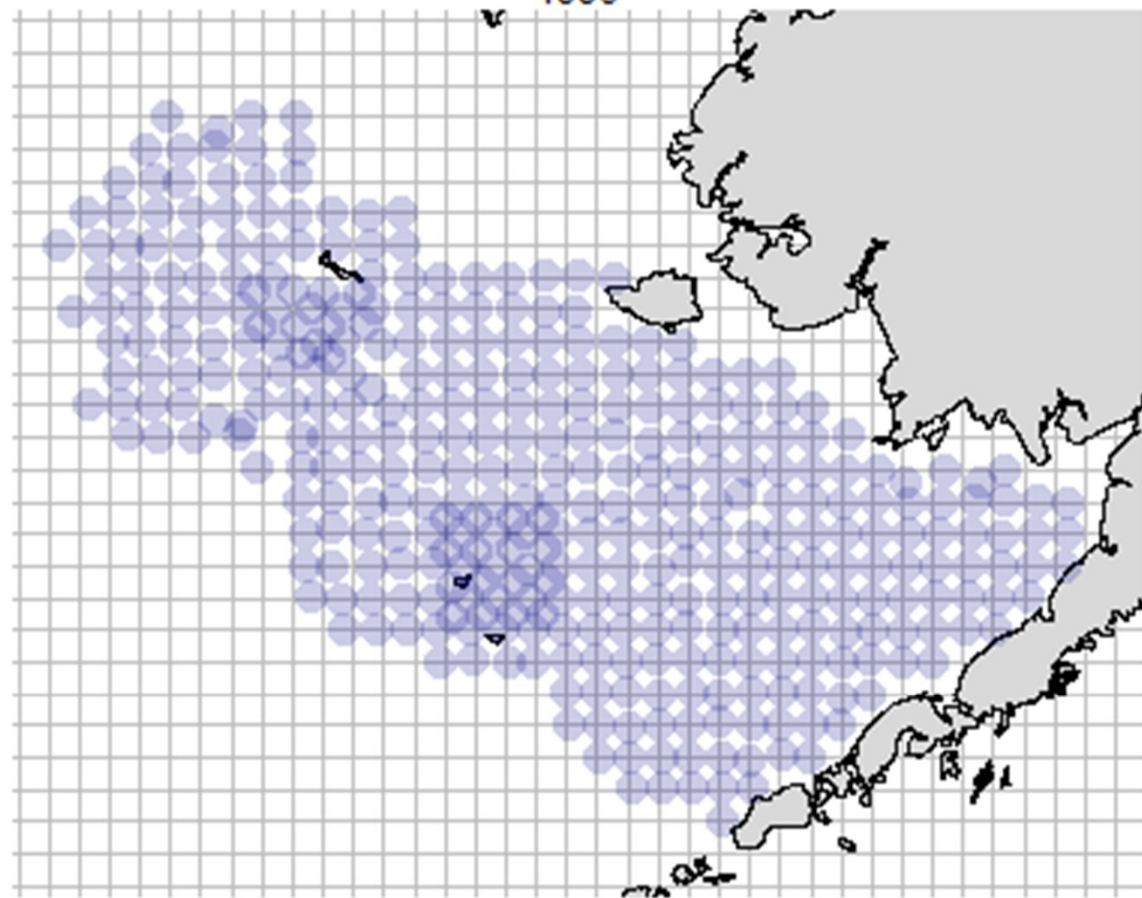
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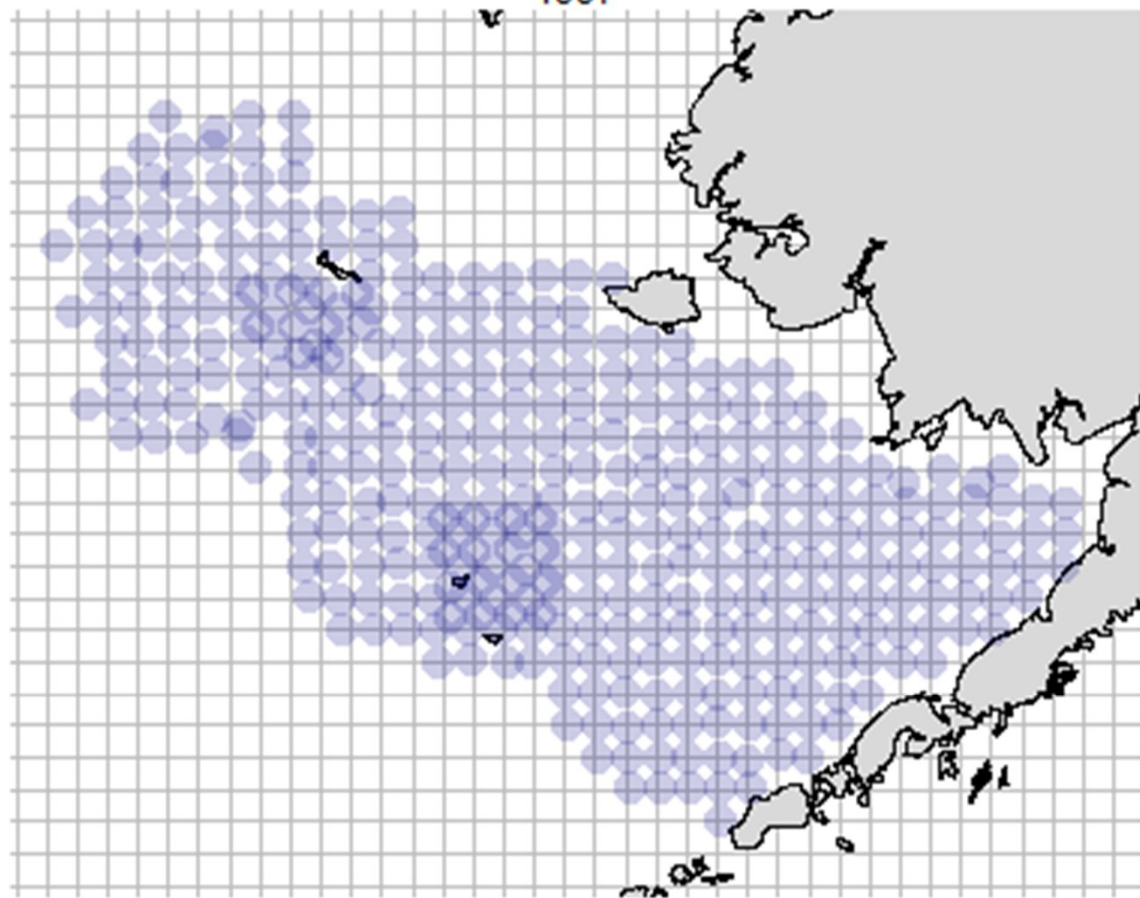
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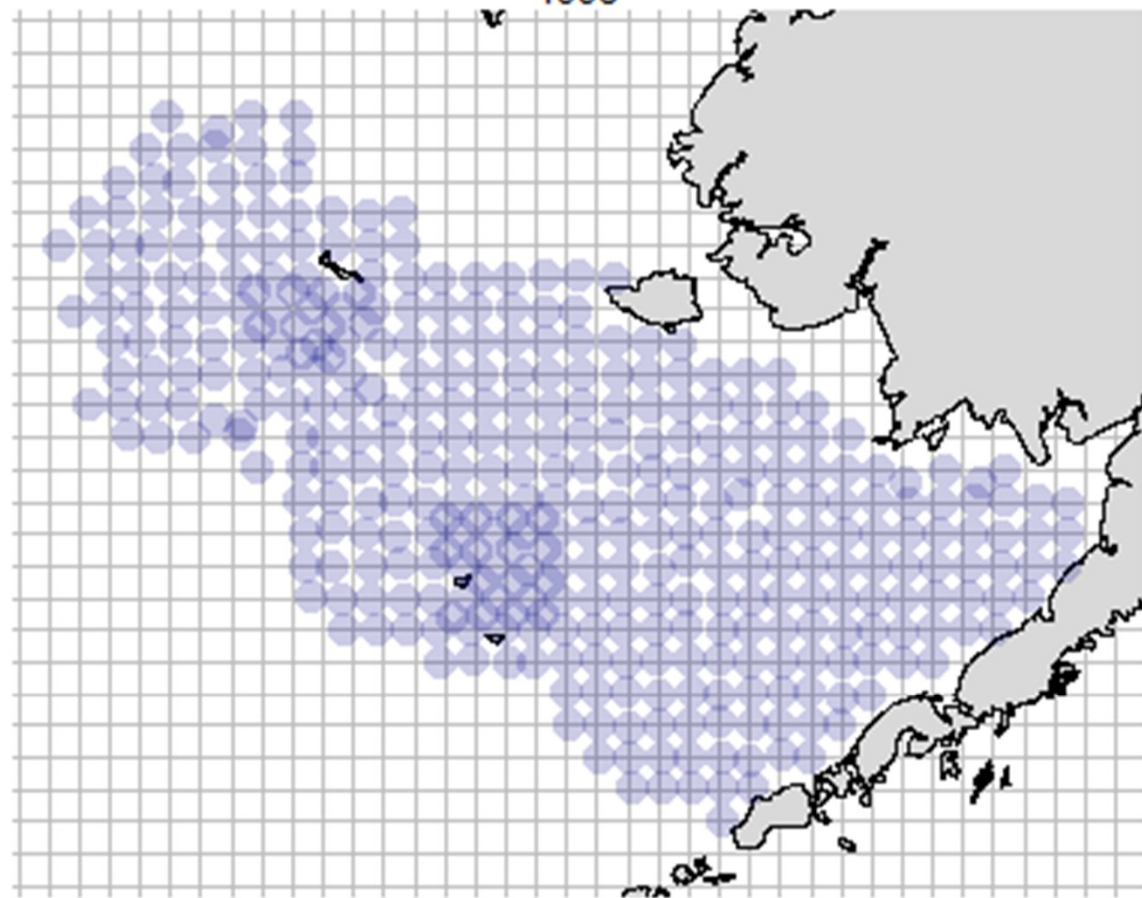
1996



1997

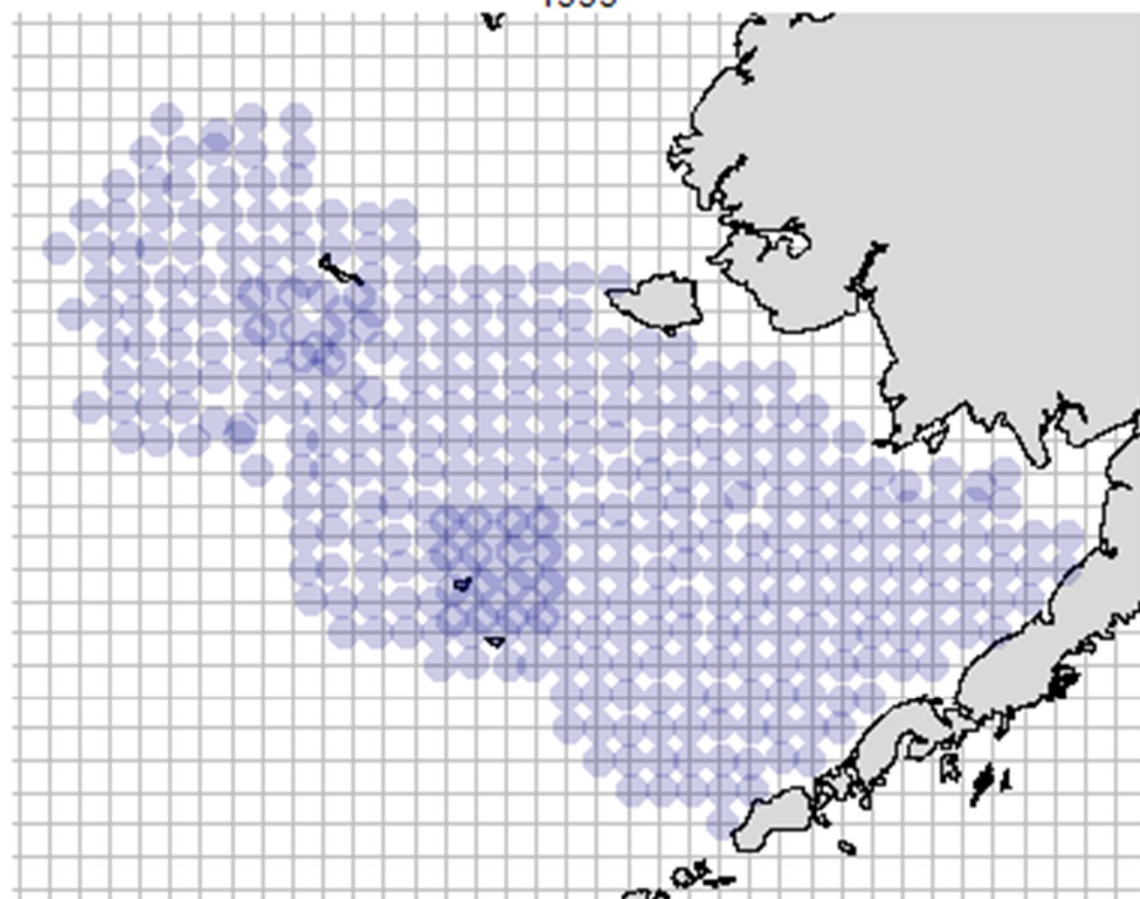


1998

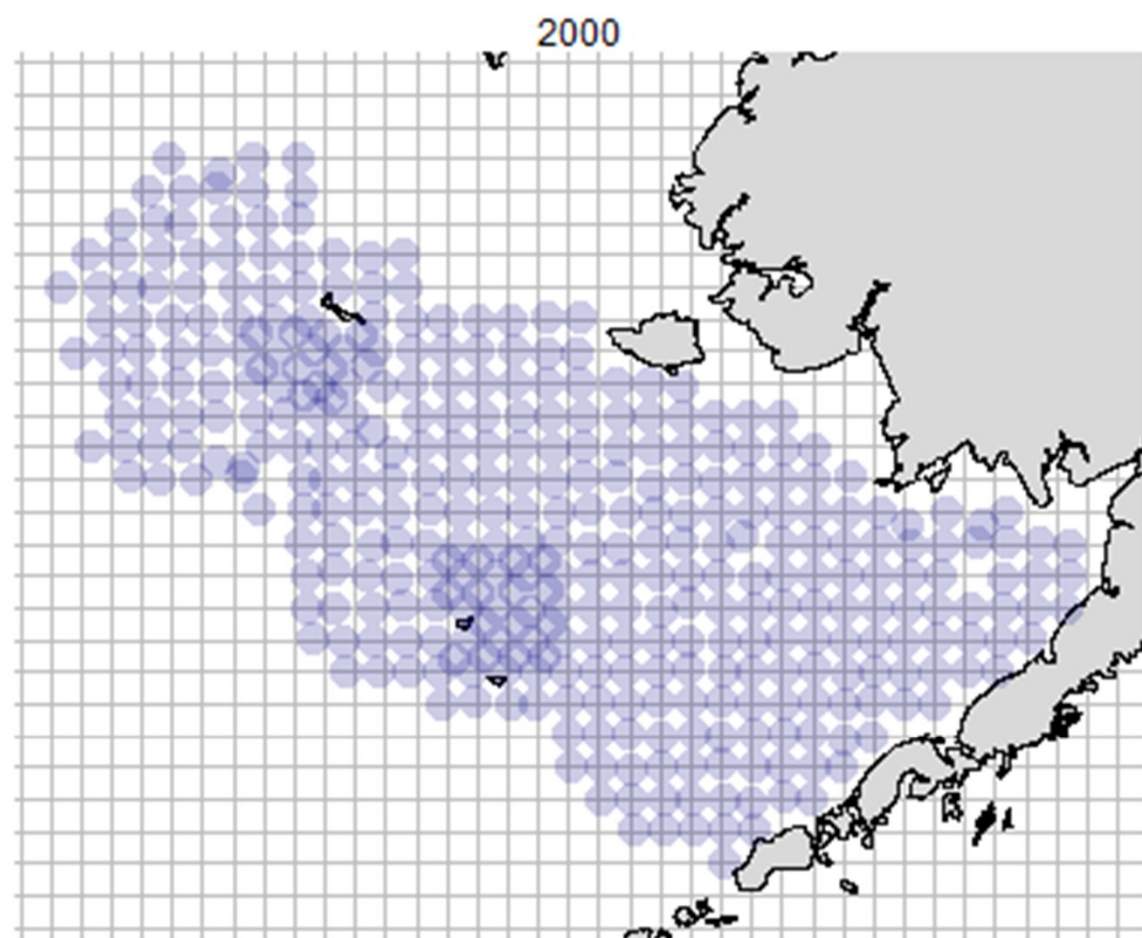




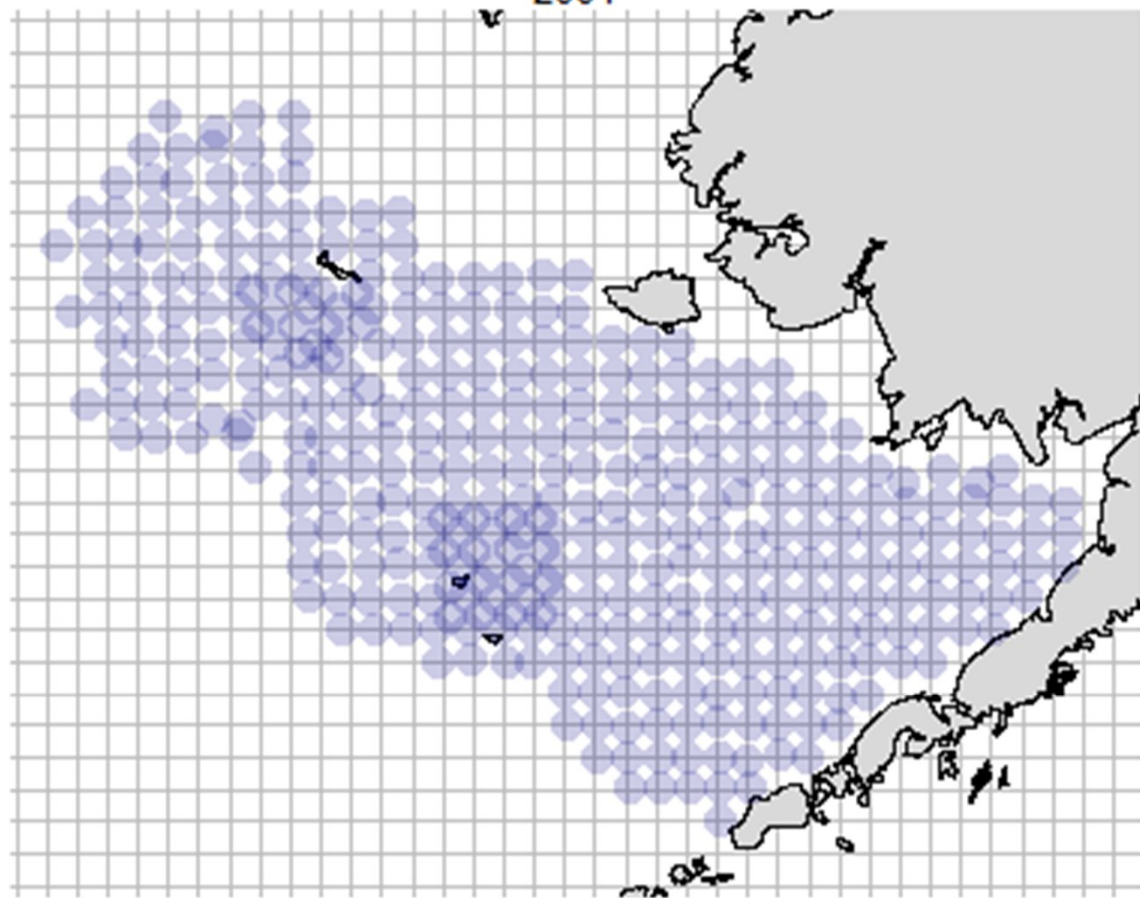
1999



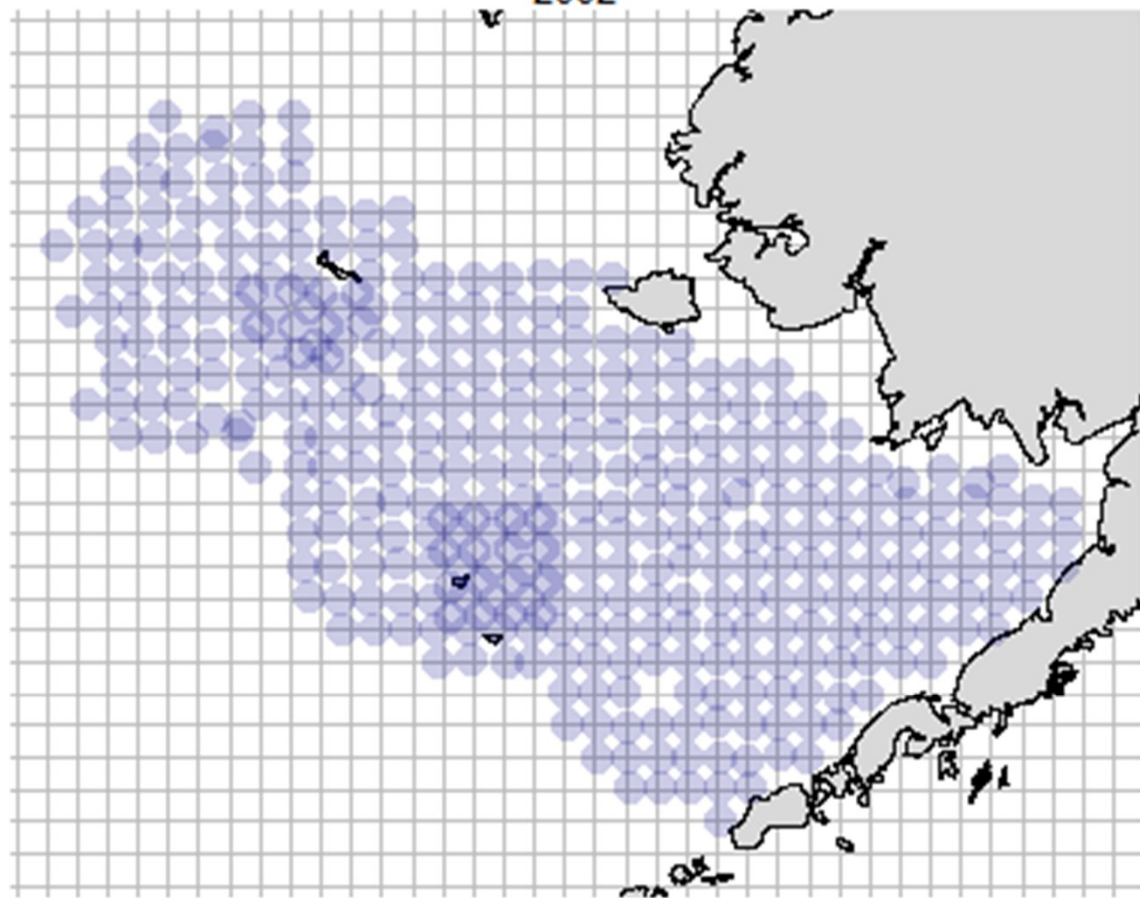




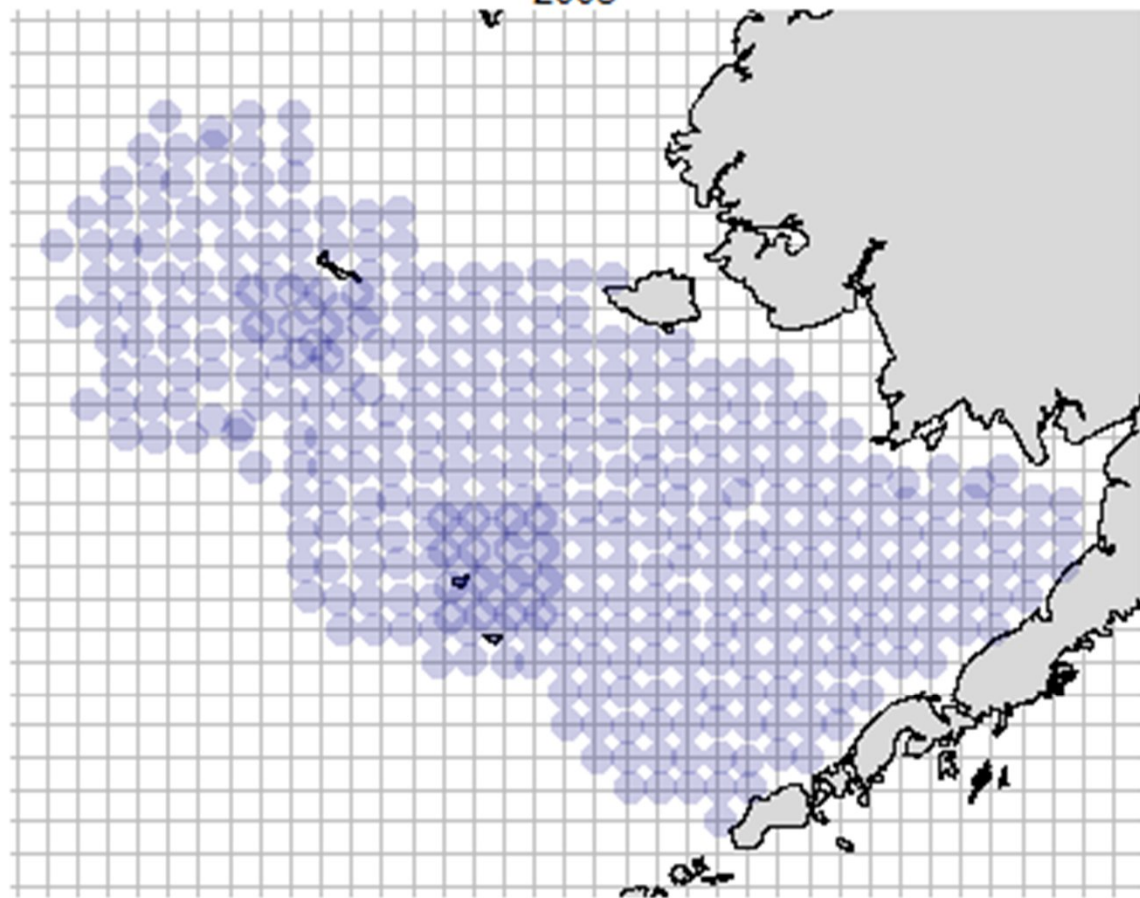
2001



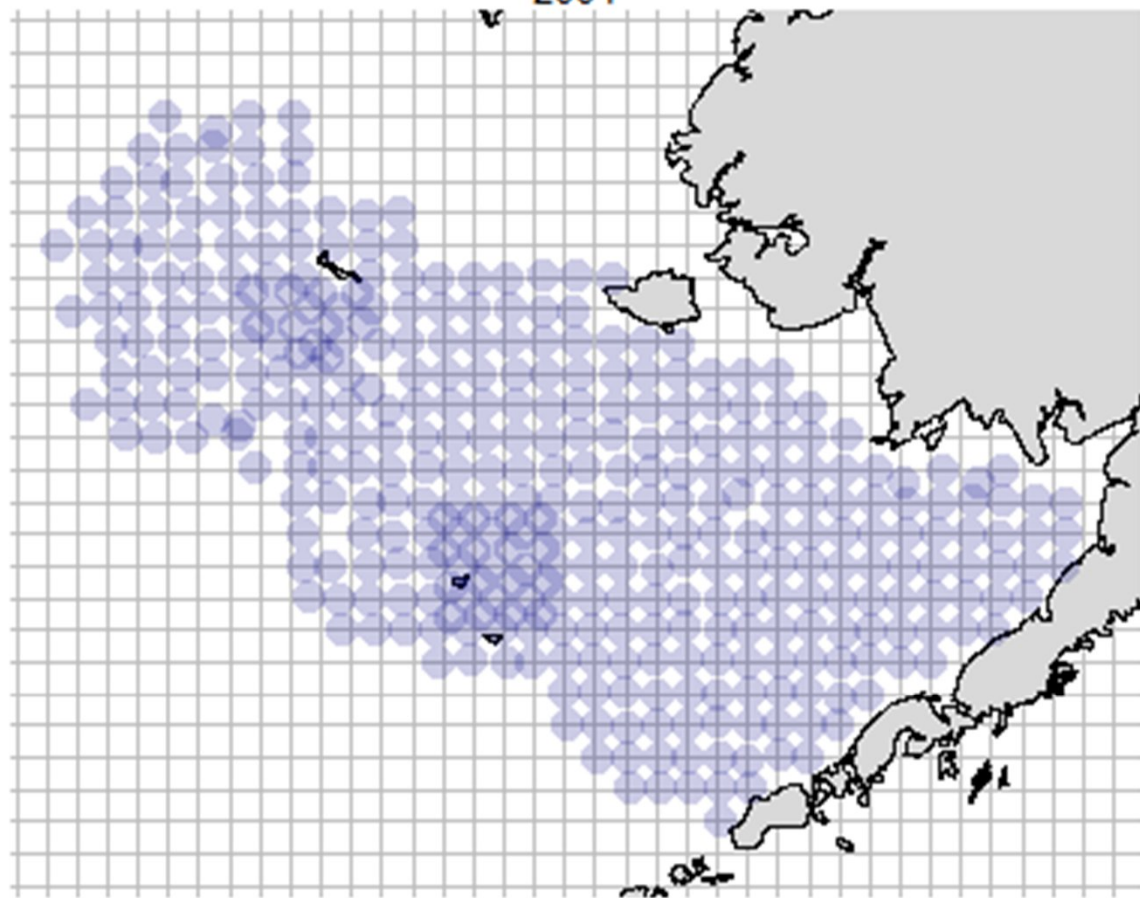
2002



2003

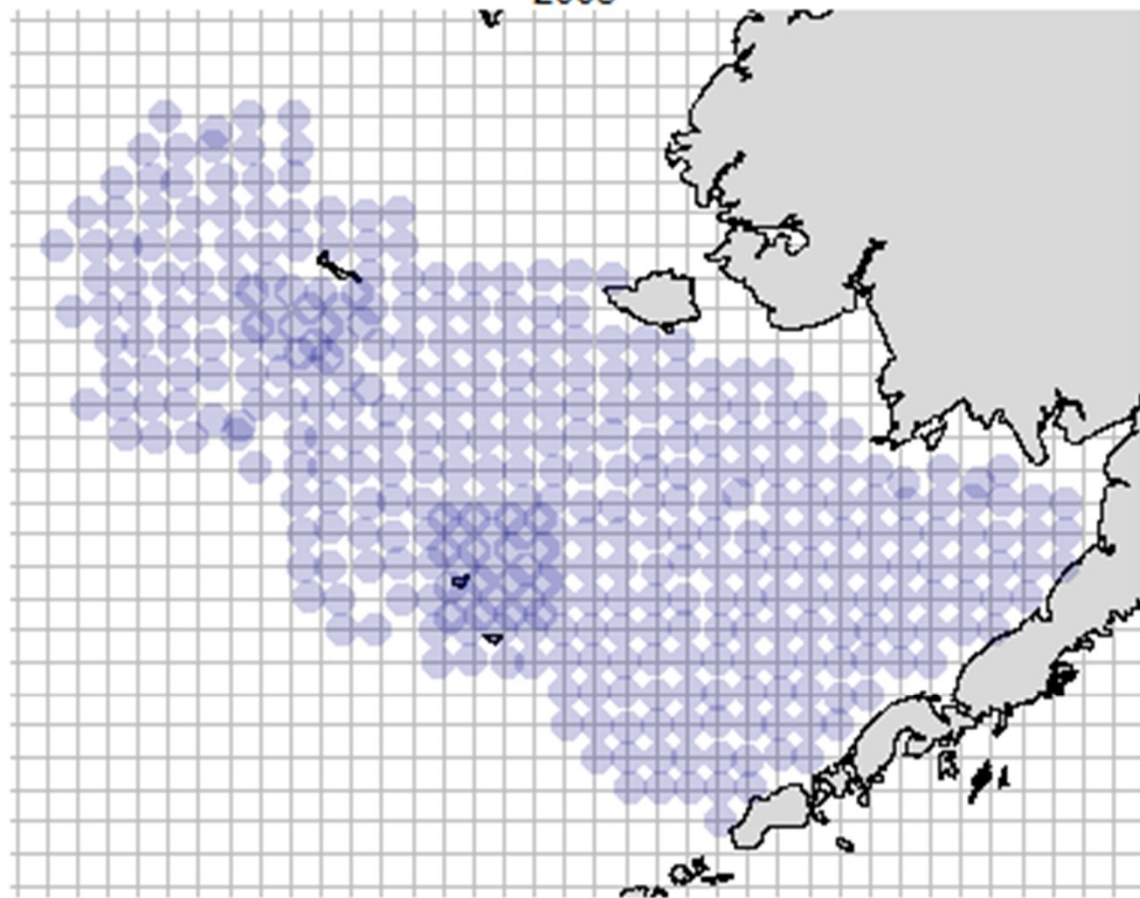


2004



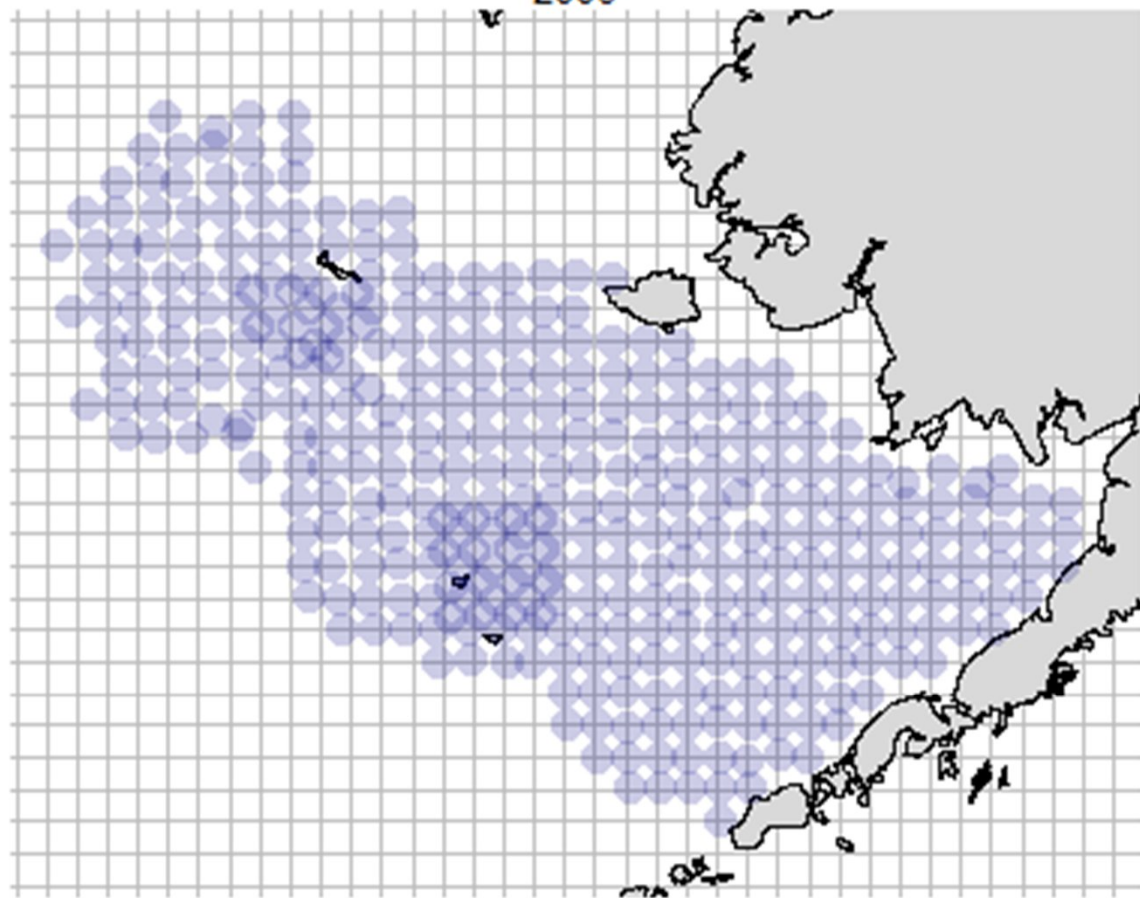


2005

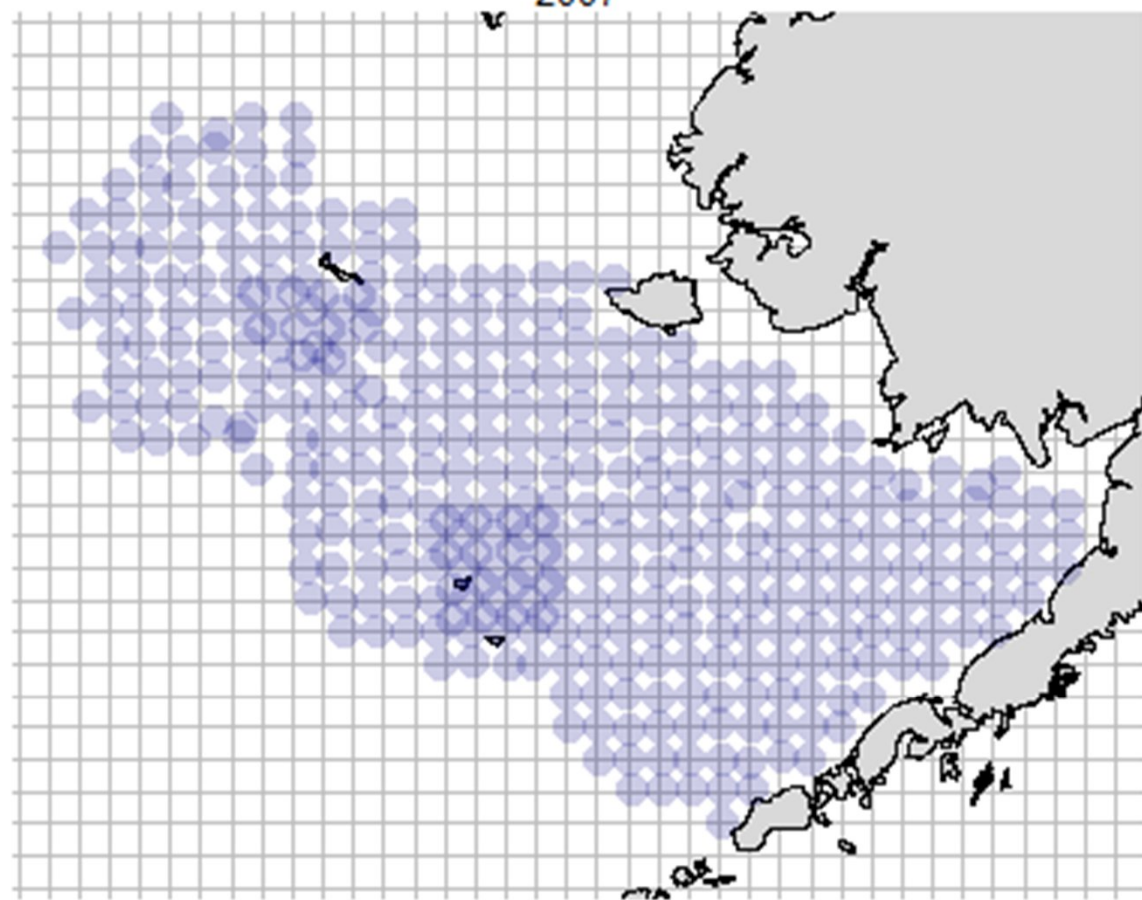




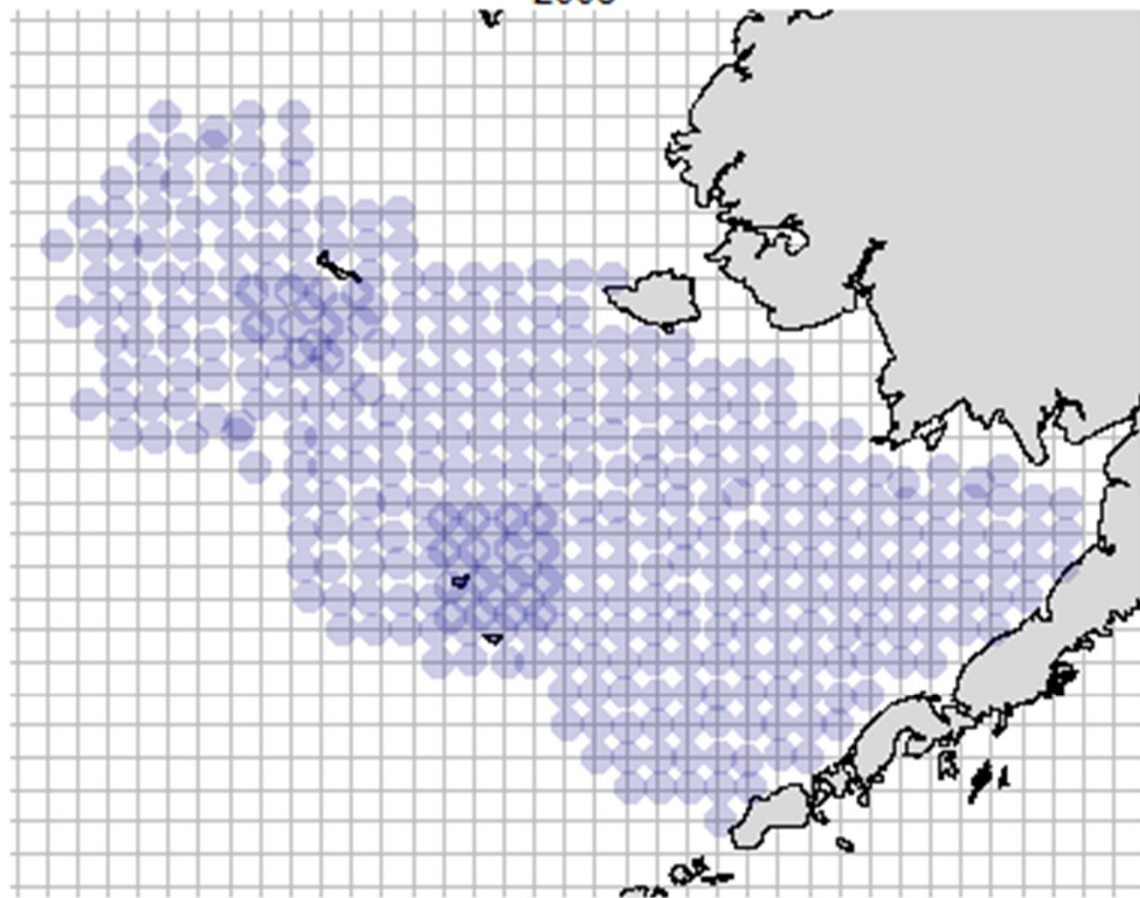
2006



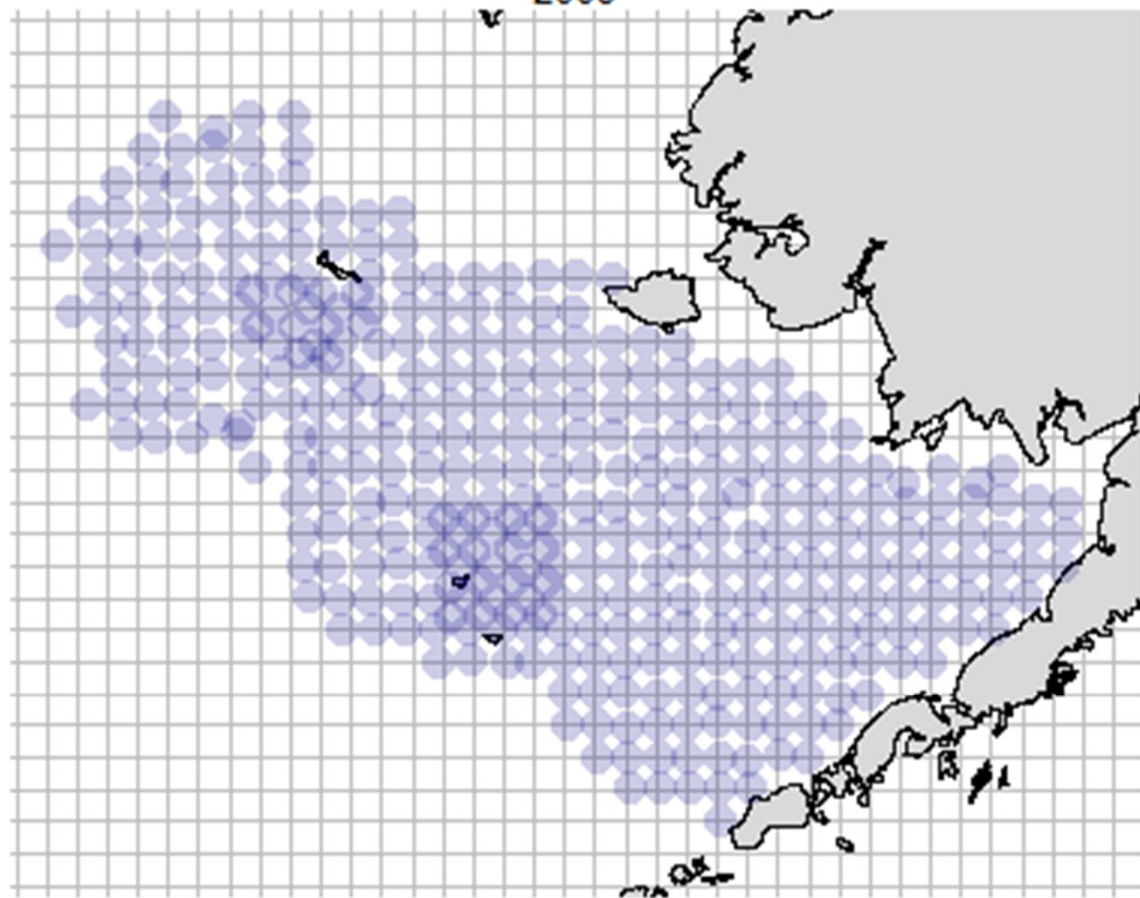
2007



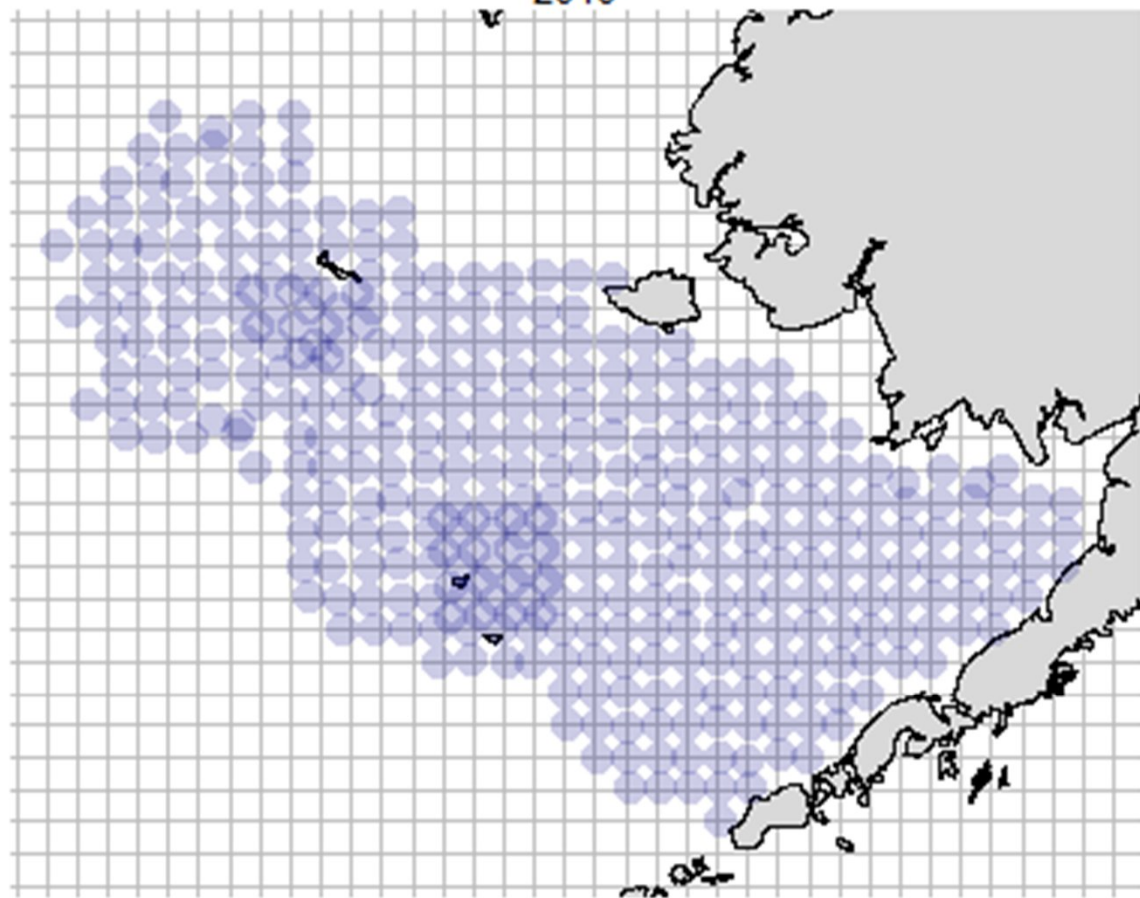
2008



2009

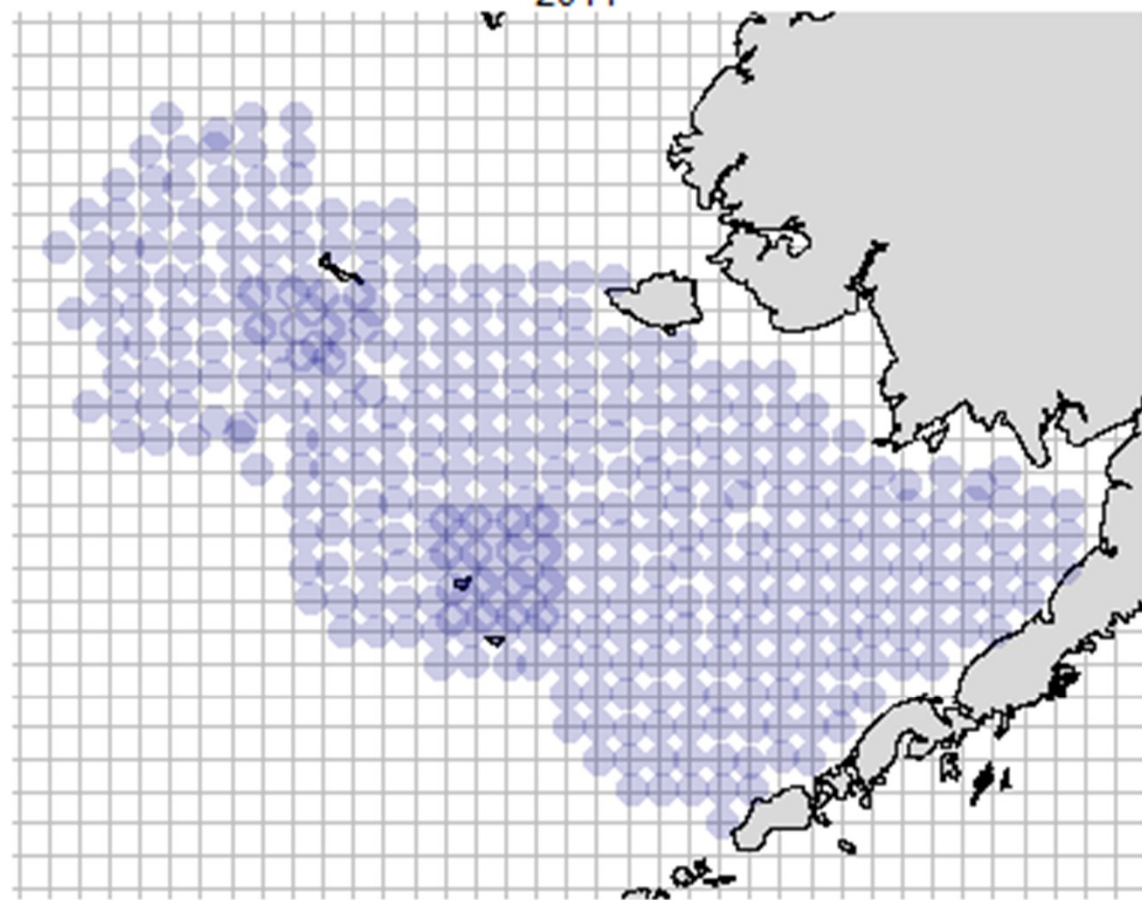


2010



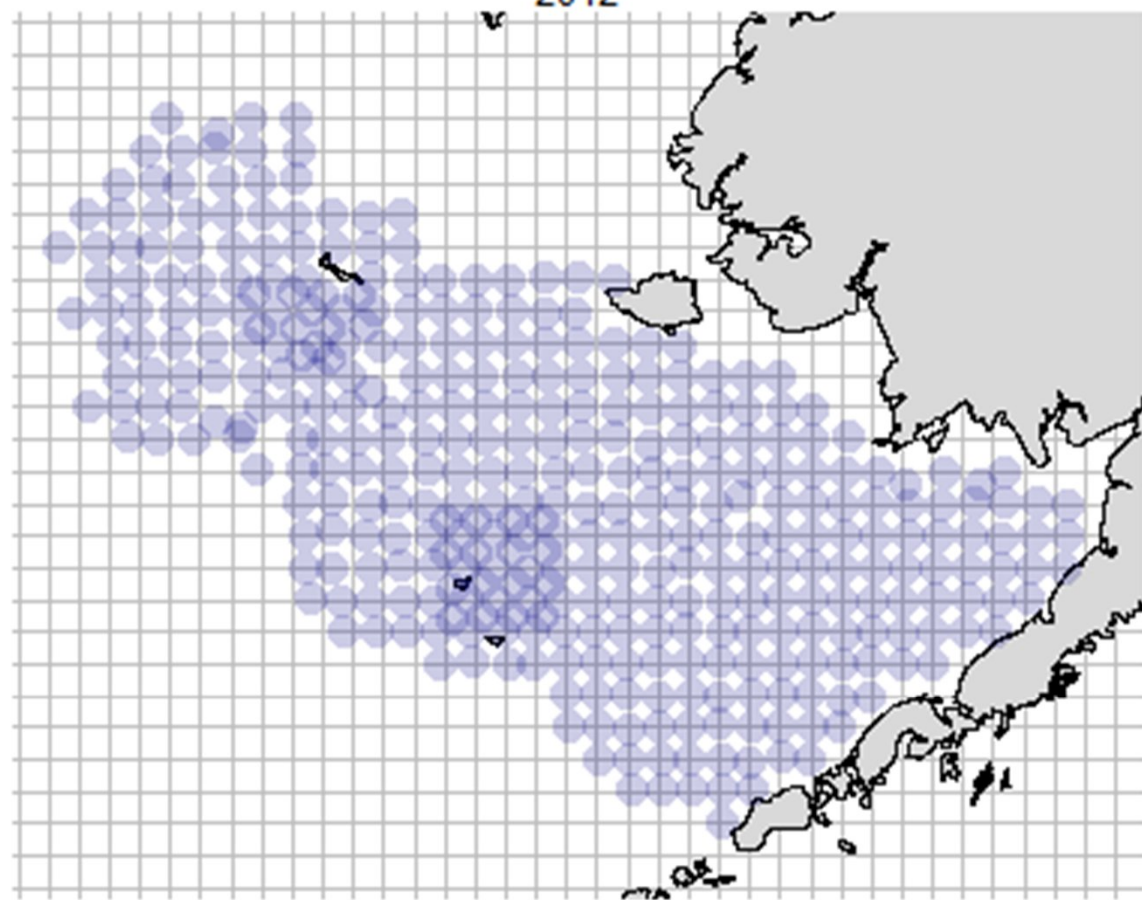


2011

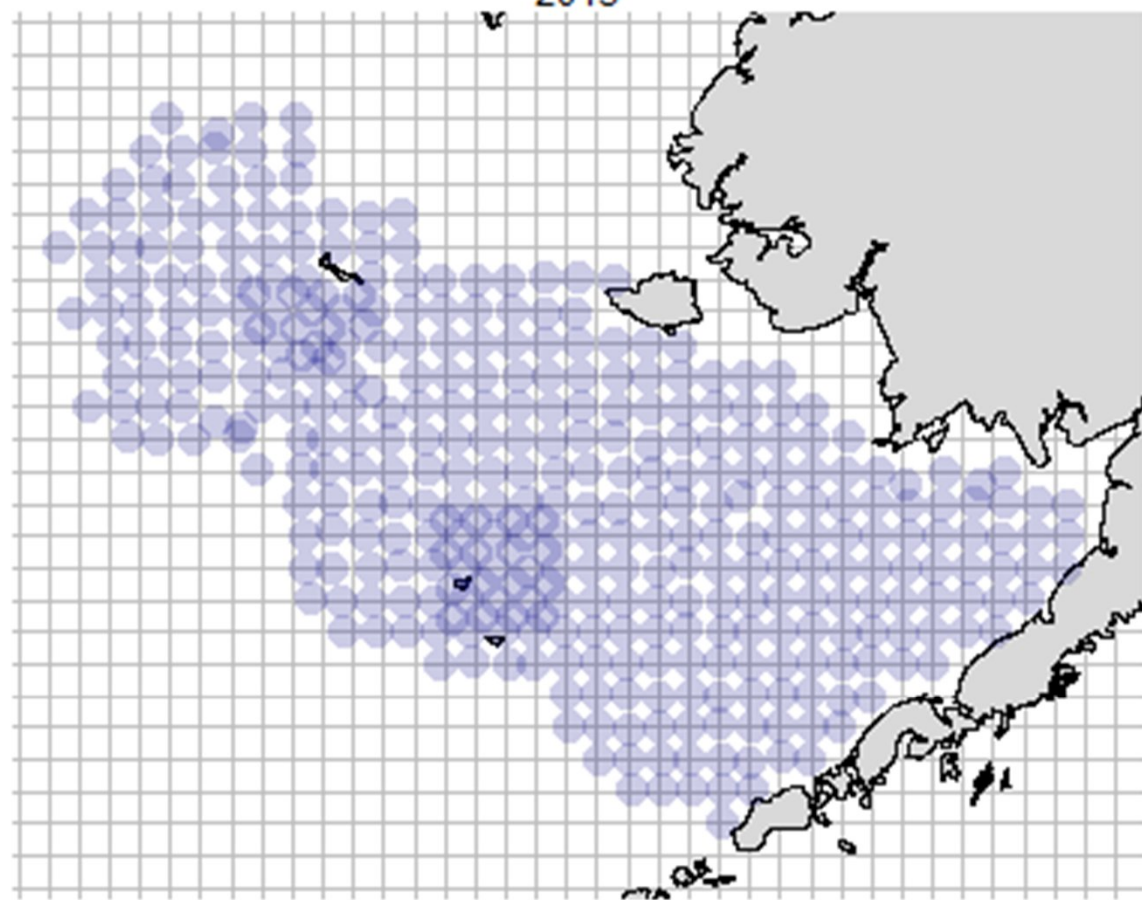




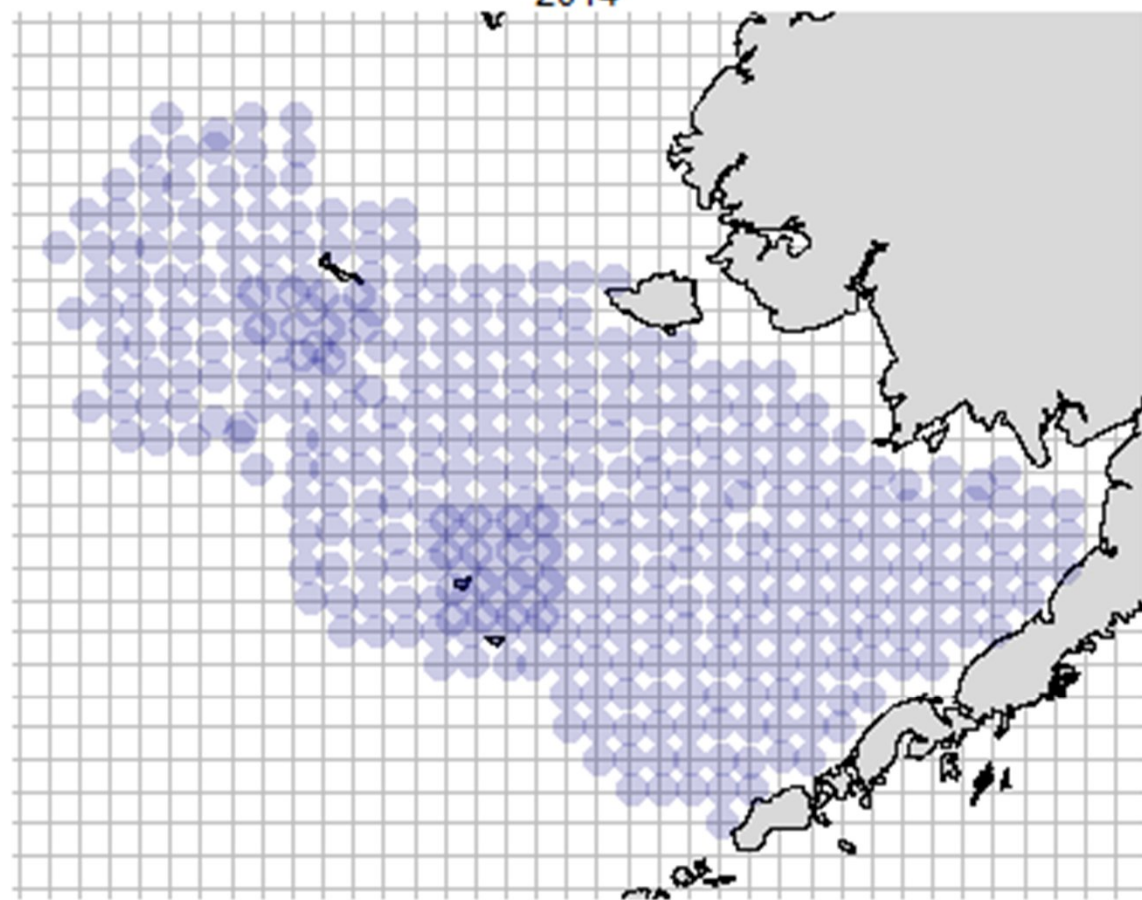
2012



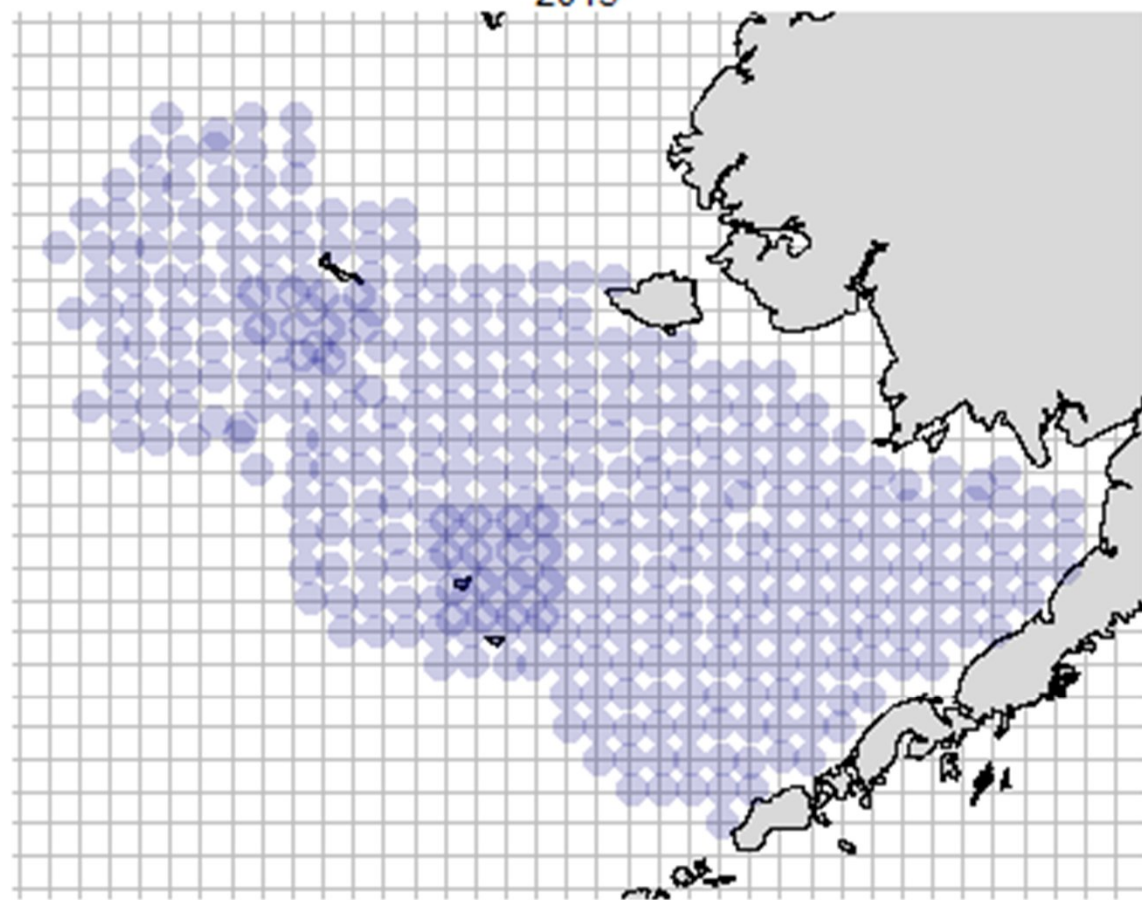
2013



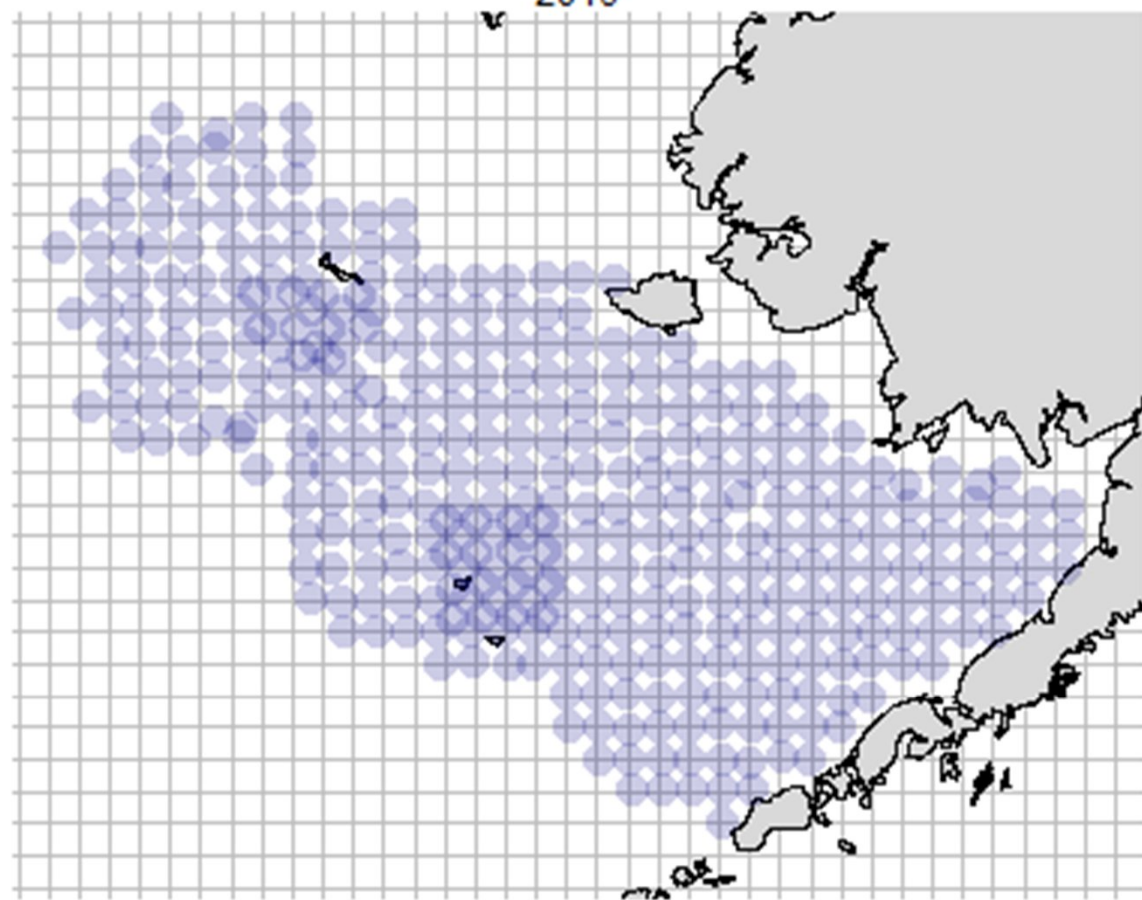
2014



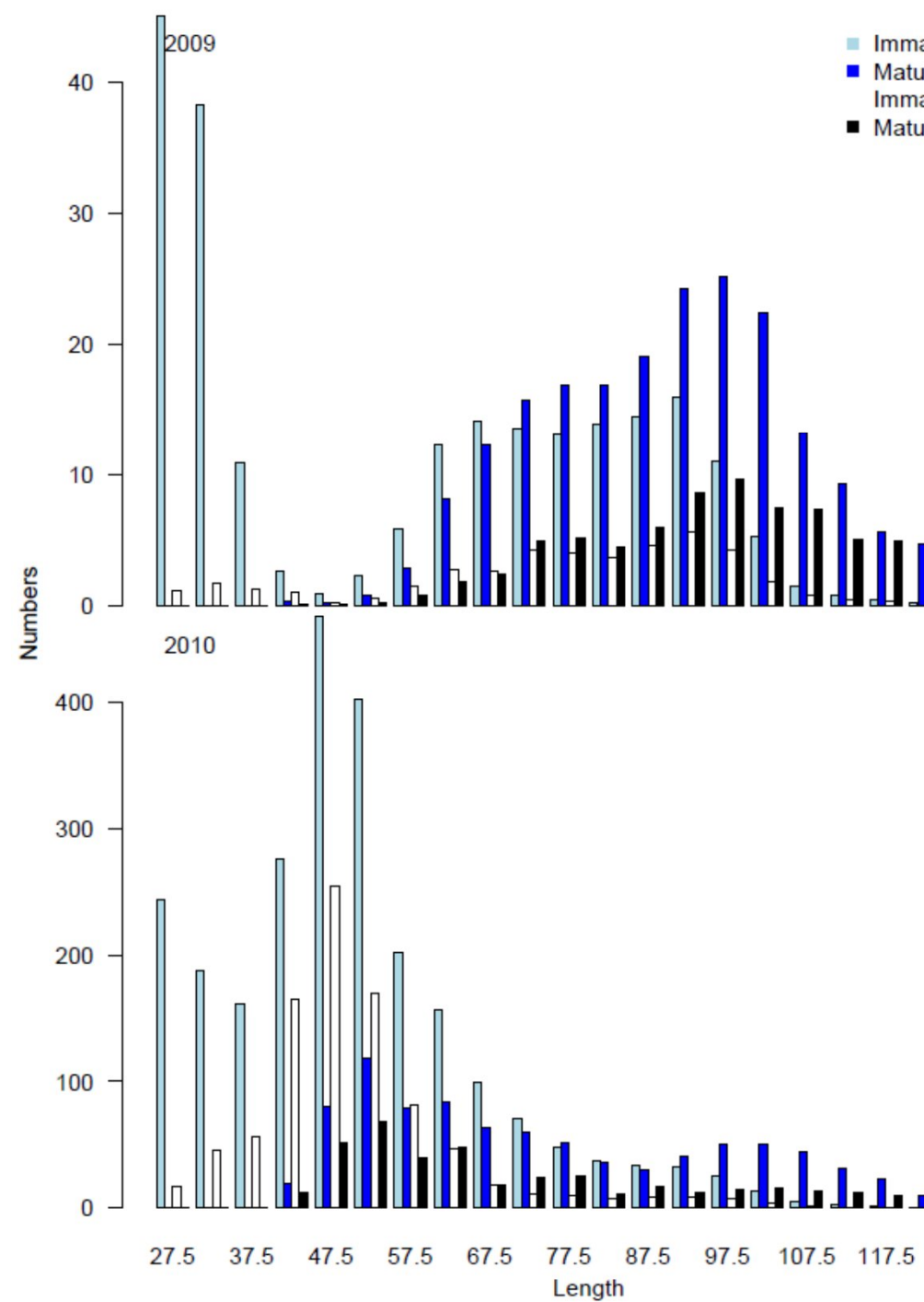
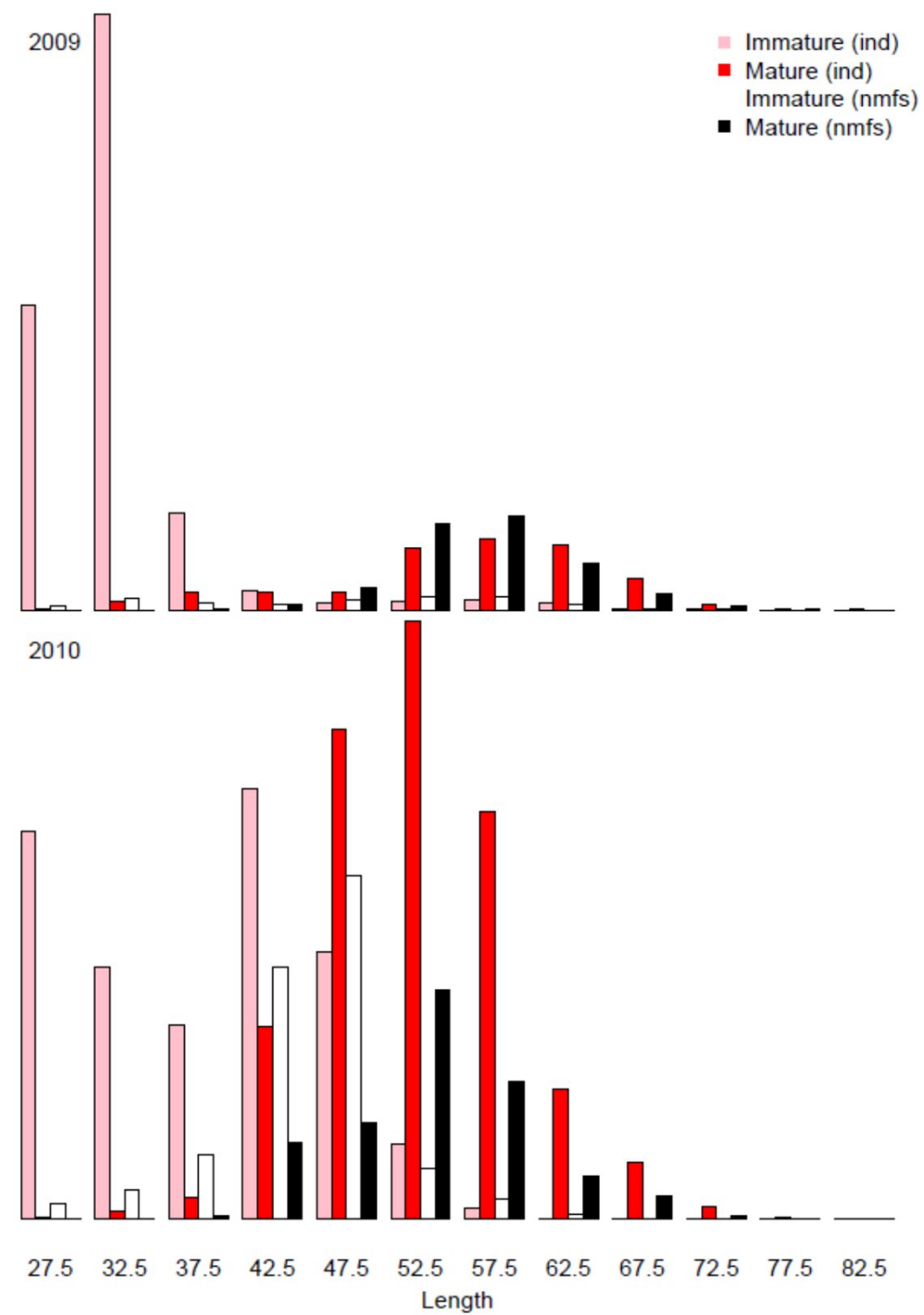
2015



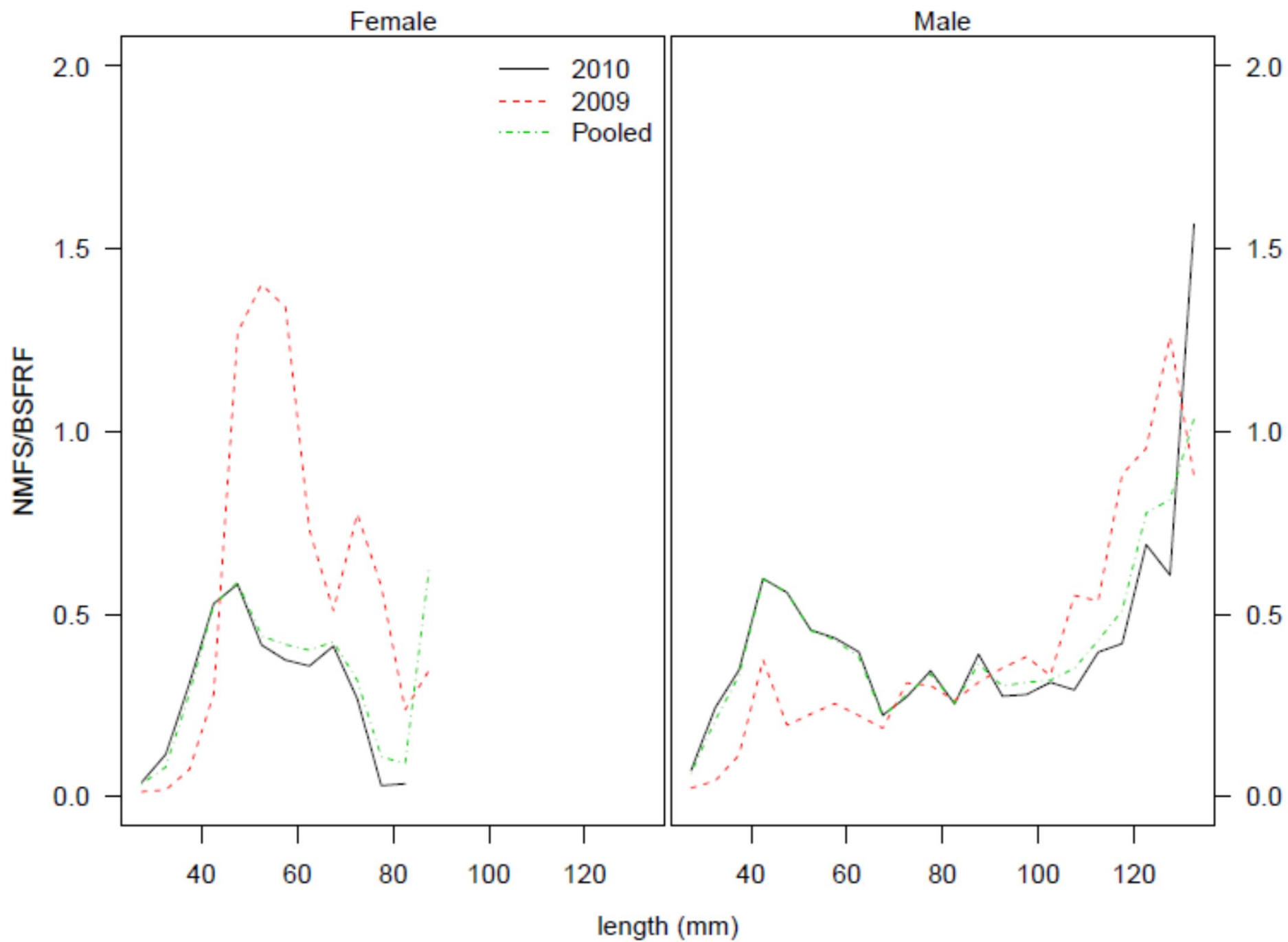
2016











# Model runs

## 'Trim data'

Excludes all data from 1978-1981, start model in 1982

Explores problem of anchoring of  $q$  and bound hitting parameters

## 'Fixed obs sel'

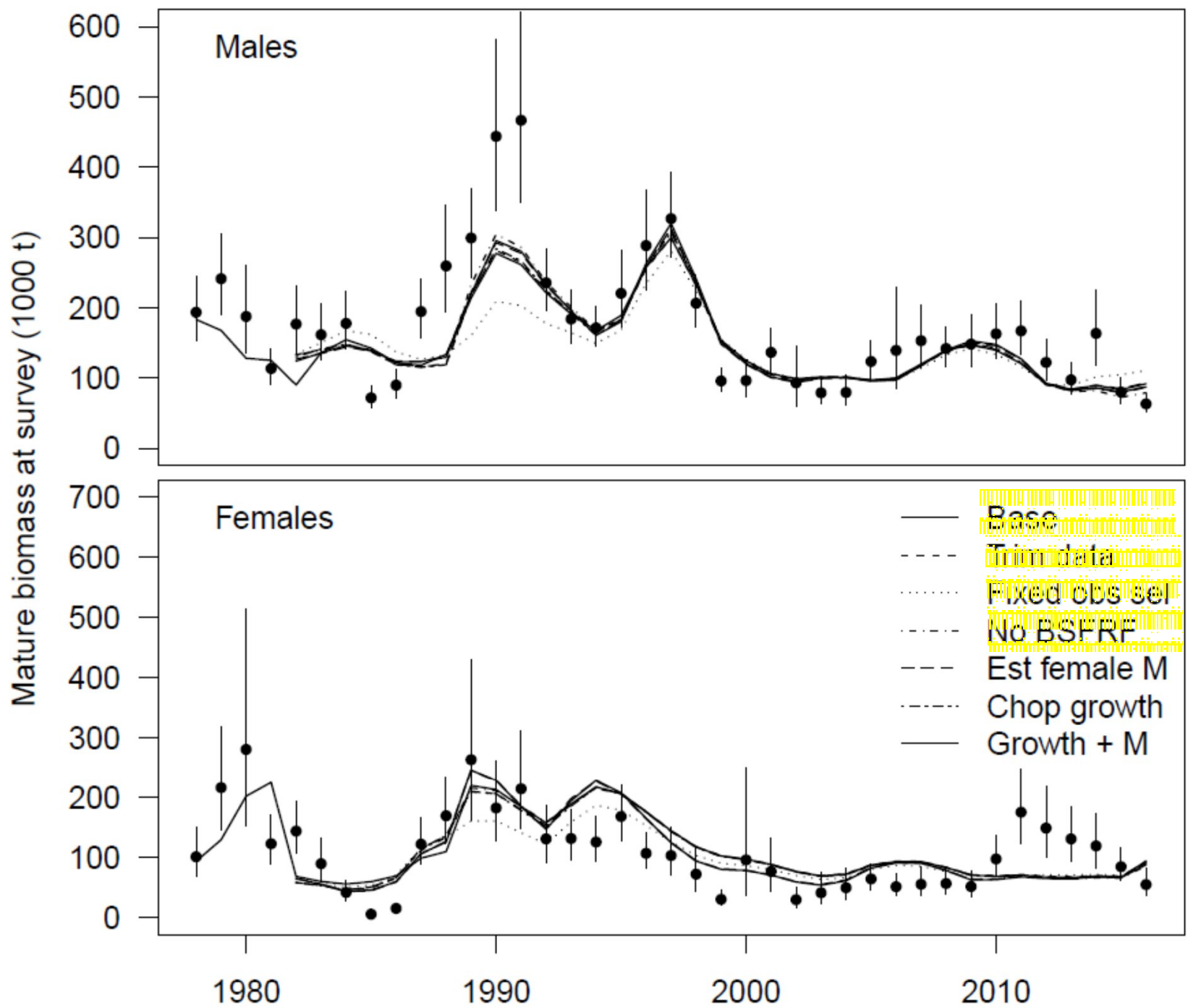
'Trim data' + fixing survey selectivity in era 2 and era 3 to selectivity inferred from BSFRF data

Explores implications of BSFRF data

## 'BSFRF'

'Trim data' + setting the weights for the BSFRF likelihood components to 0

Explores the impact of the BSFRF data on model output



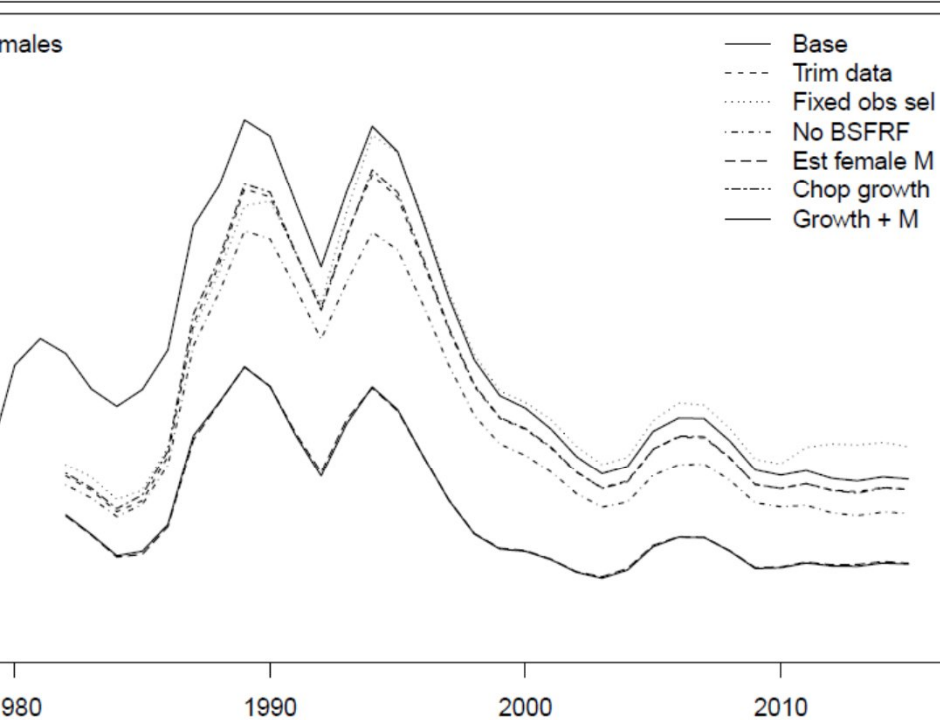
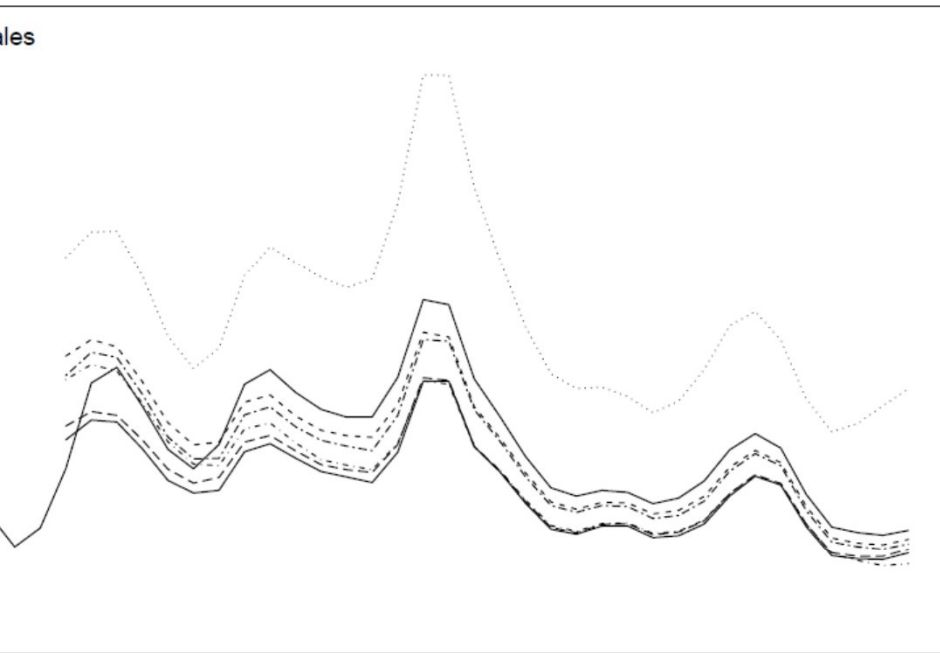
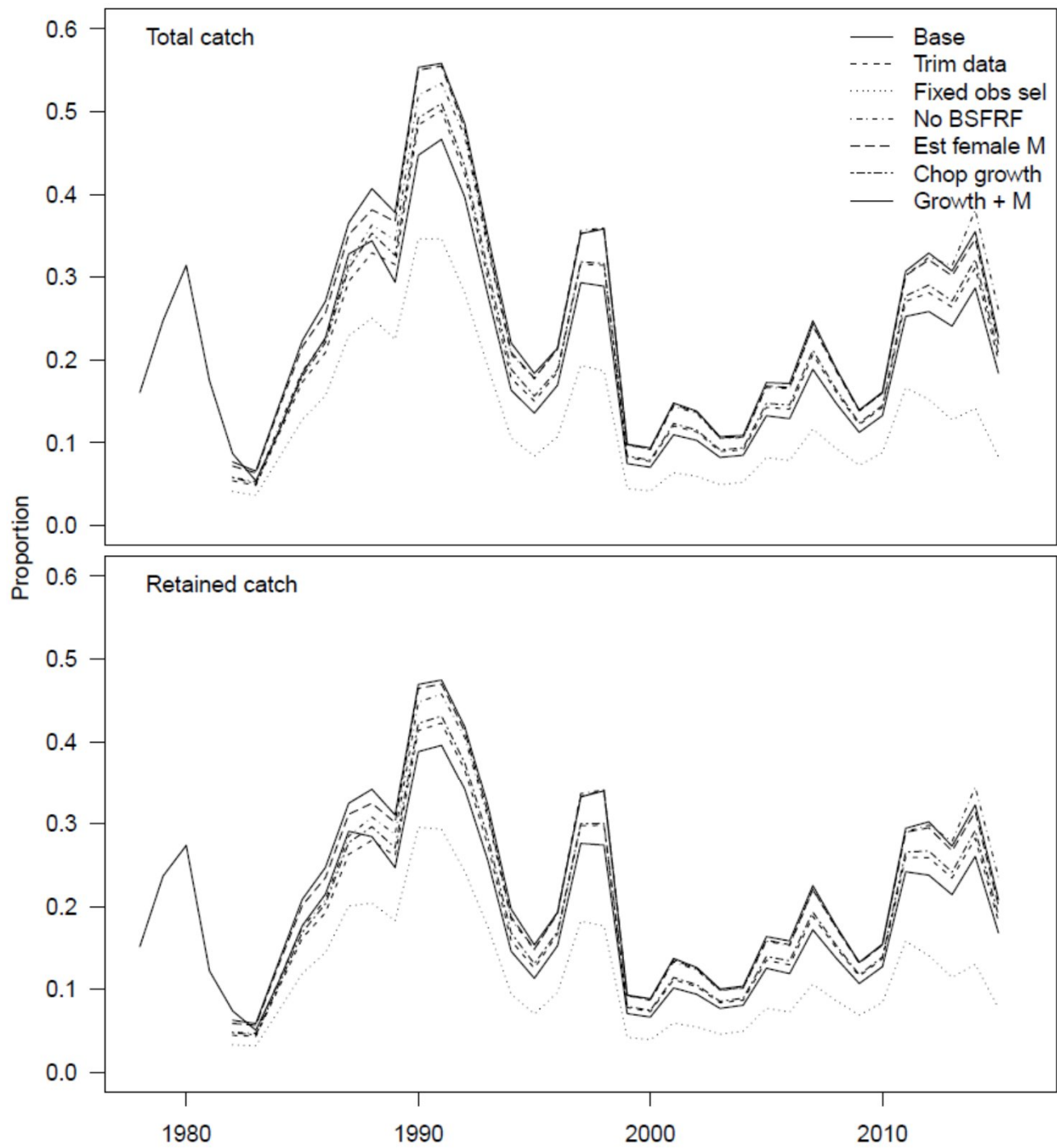
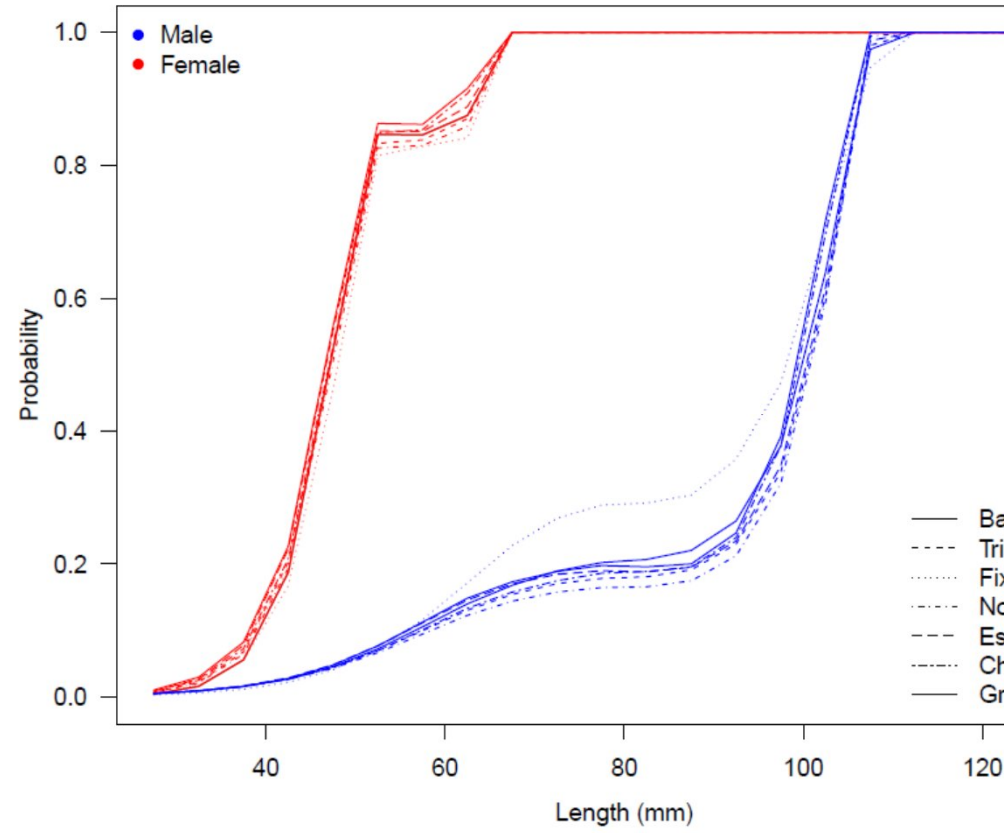
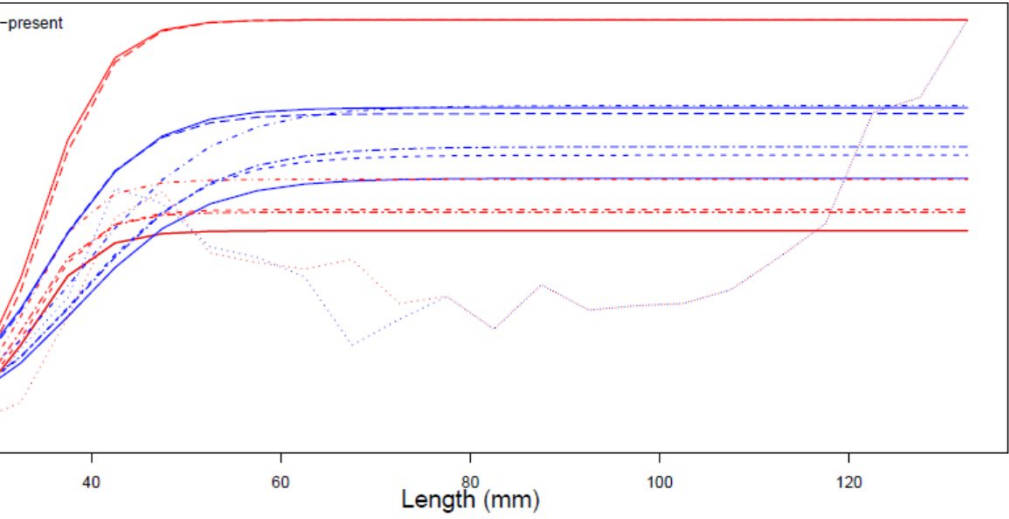
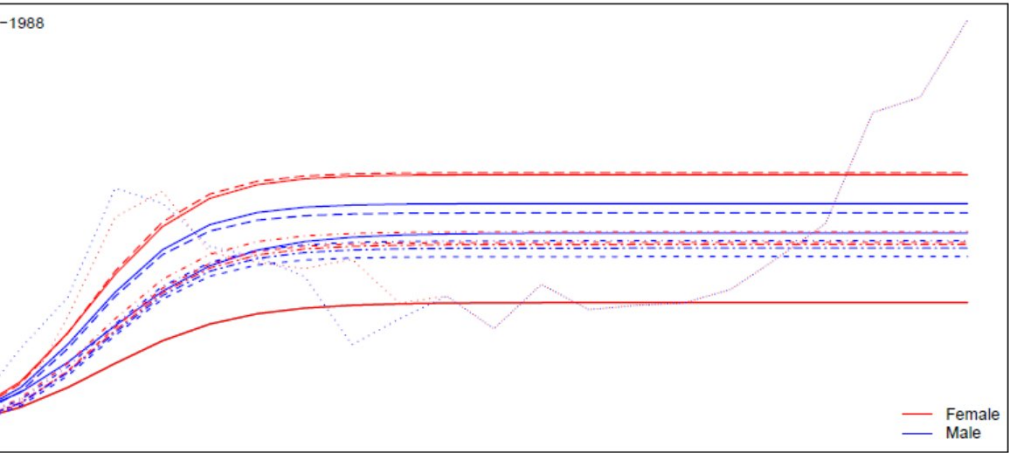
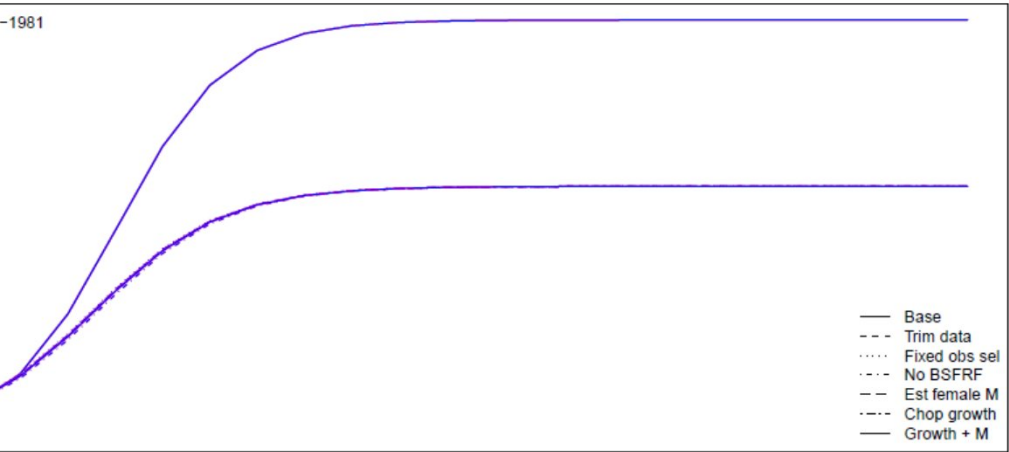


Table 1: Changes in management quantities for each scenario considered. Reported quantities are the MLEs because running MCMCs for every model was prohibitively time-consuming. The MLEs for scenarios in which MCMCs were performed are very close to the medians of the posterior distributions.

Model	MMB	B35	F35	FOFL
Base	92.09	152.3	1.91	1.14
Trim data	89.8	152.3	1.42	0.81
Fixed obs sel	221.6	215.3	5.49	2.62
No BSFRF	60.86	142.3	1.17	0.56
Est female M	74.29	139.6	1.21	0.68
Chop growth	79.57	149.8	1.34	0.75
Growth + M	70.89	137.4	1.17	0.64

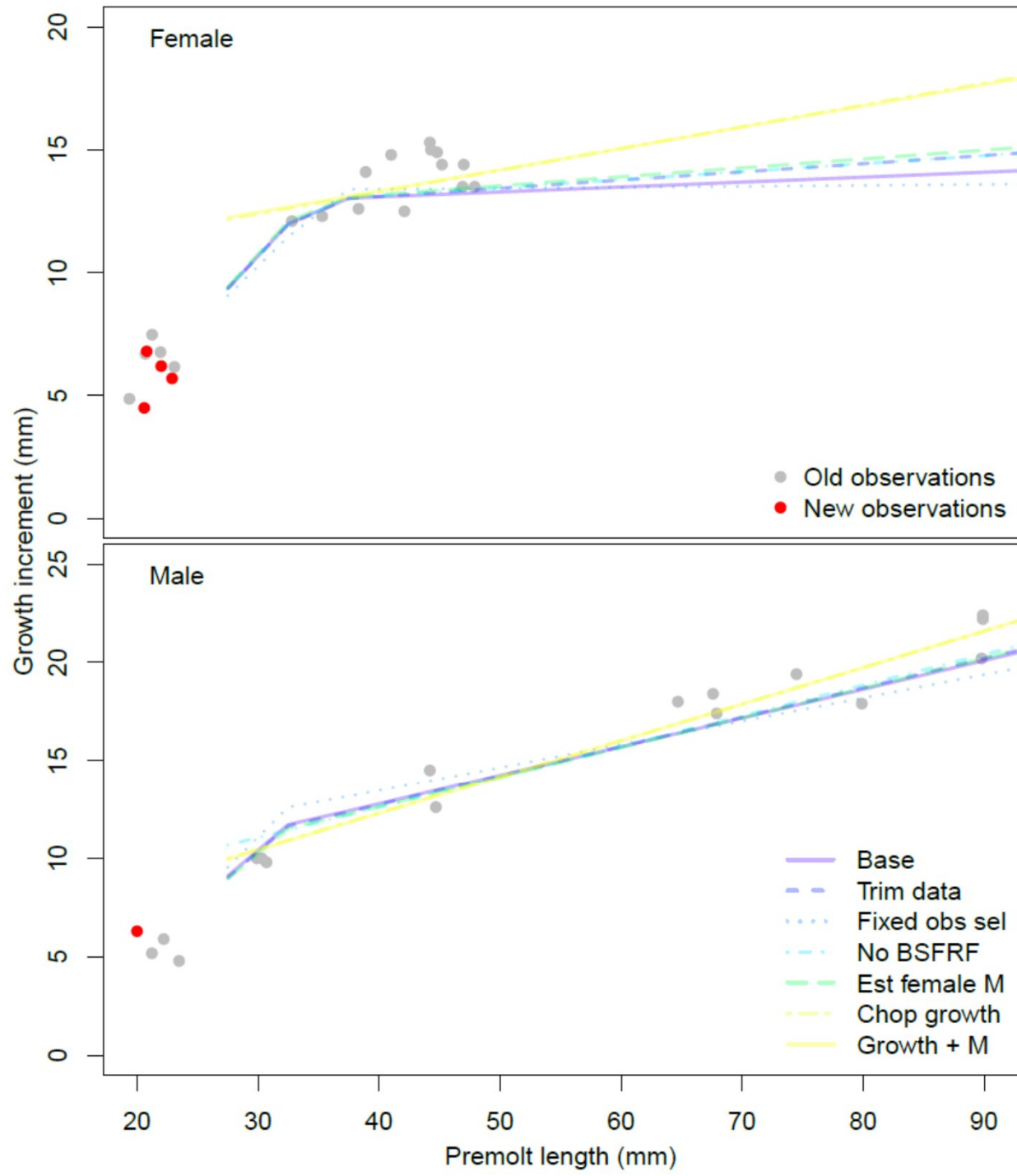




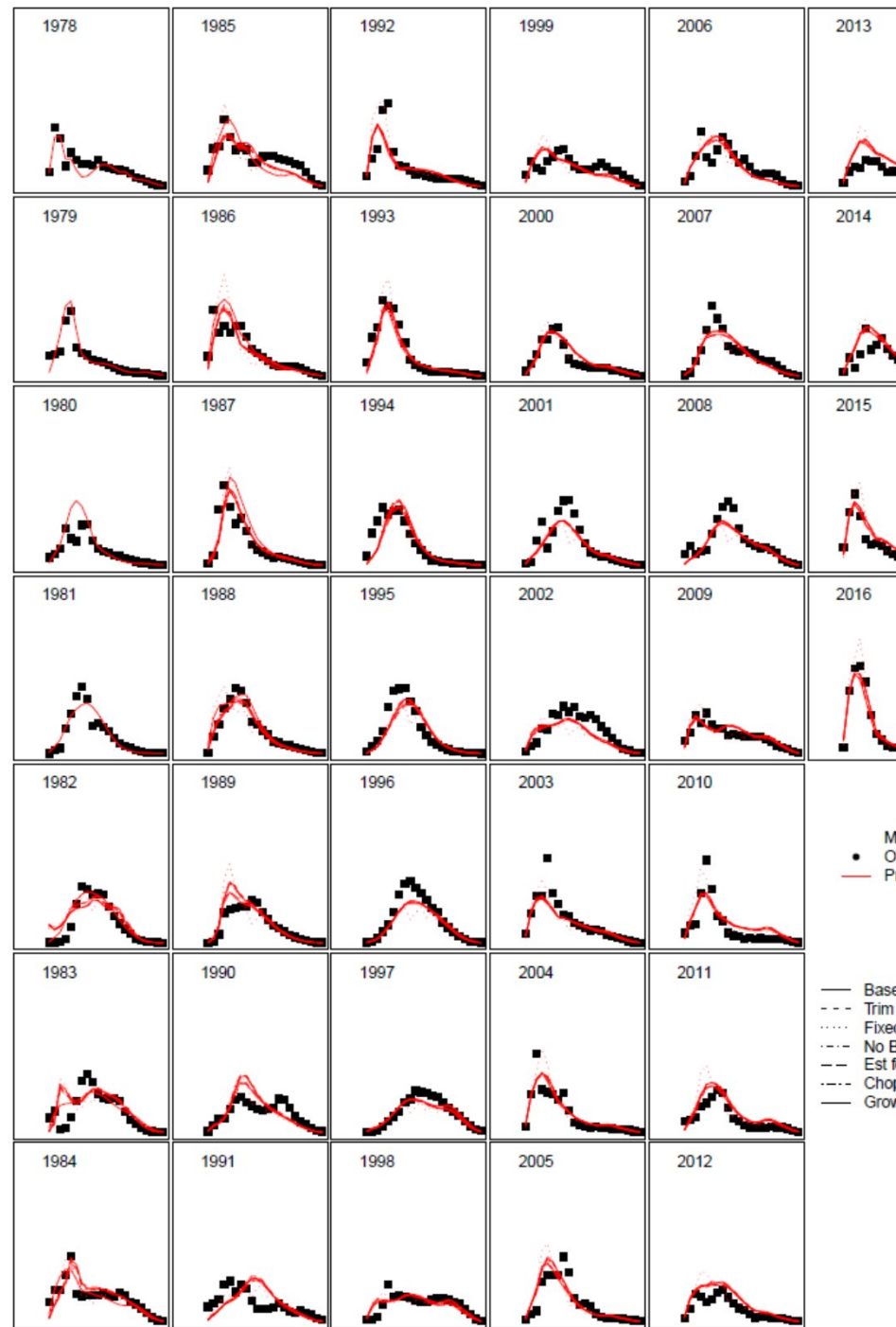
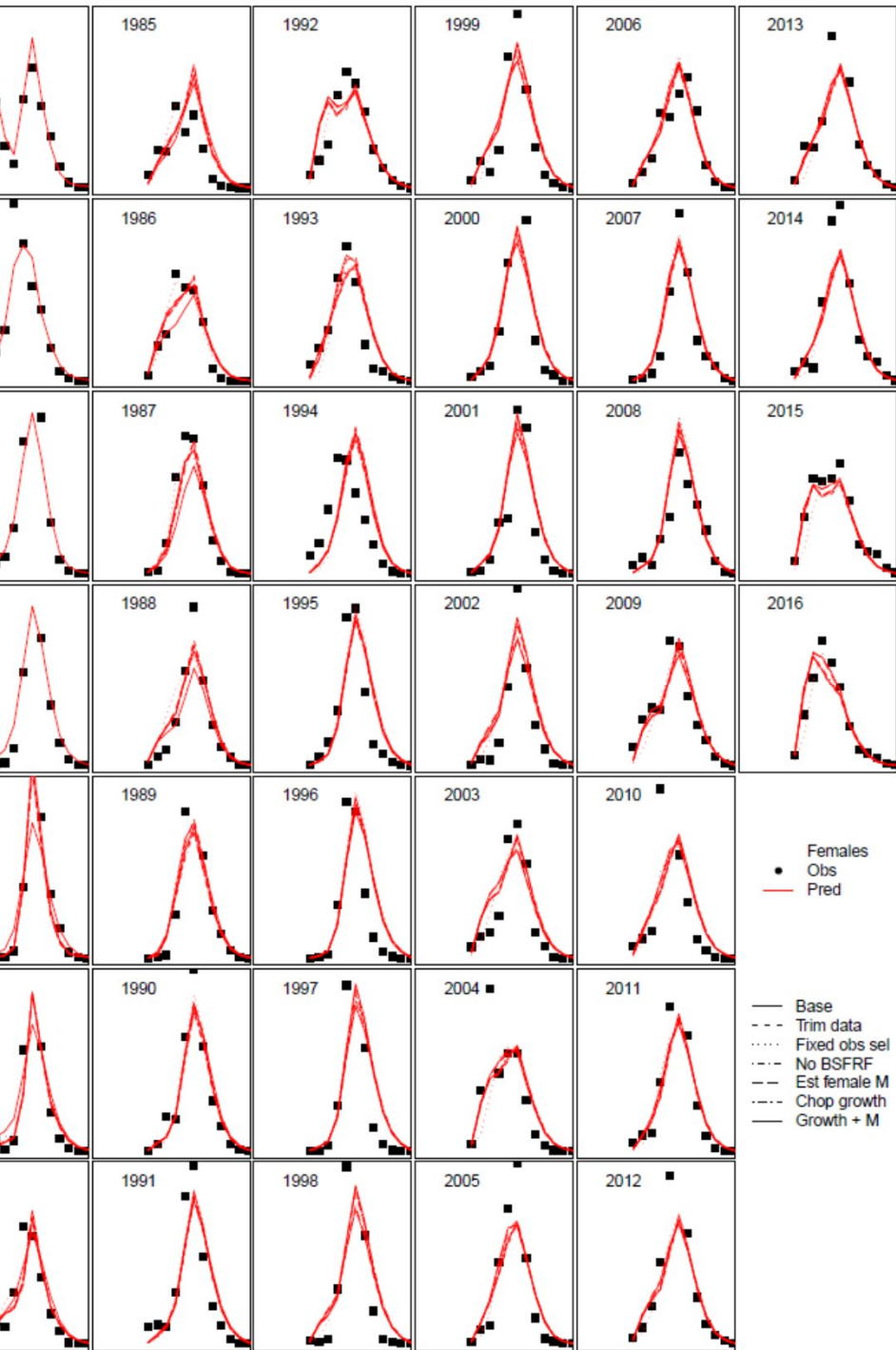


Parameter	Base	Trim data	Fixed obs sel	No BSFRF	Est female M	Chop growth	Growth + M
srv1_q	1	0.6	0.6	0.6	0.6	0.6	0.6
srv1_q_f	1	0.6	0.6	0.6	0.6	0.6	0.6
srv1_sel95	59.89	61.19	61.19	61.64	61.71	61.34	61.39
srv1_sel50	42.66	41.18	41.18	41.69	41.7	41.32	41.37
srv2_q	0.49	0.43	0.43	0.47	0.54	0.45	0.56
srv2_q_f	0.32	0.46	0.46	0.49	0.63	0.43	0.63
srv2_sel95	61.3	57.05	57.05	57.32	58.24	58.55	58.01
srv2_sel50	41.32	41.18	41.18	40.84	39.82	41.35	39.36
srv3_q	0.62	0.68	0.68	0.79	0.77	0.7	0.79
srv3_sel95	57.24	57.63	57.63	59.43	49.53	59.37	50.62
srv3_sel50	38.42	38.59	38.59	38.78	34.78	39.15	34.94
srv3_q_f	0.49	0.54	0.54	0.62	1	0.54	1
srv3_sel95_f	43.09	43.42	43.42	42.85	45.03	42.34	44.8
srv3_sel50_f	33.27	33.47	33.47	32.97	34.03	32.94	34.27

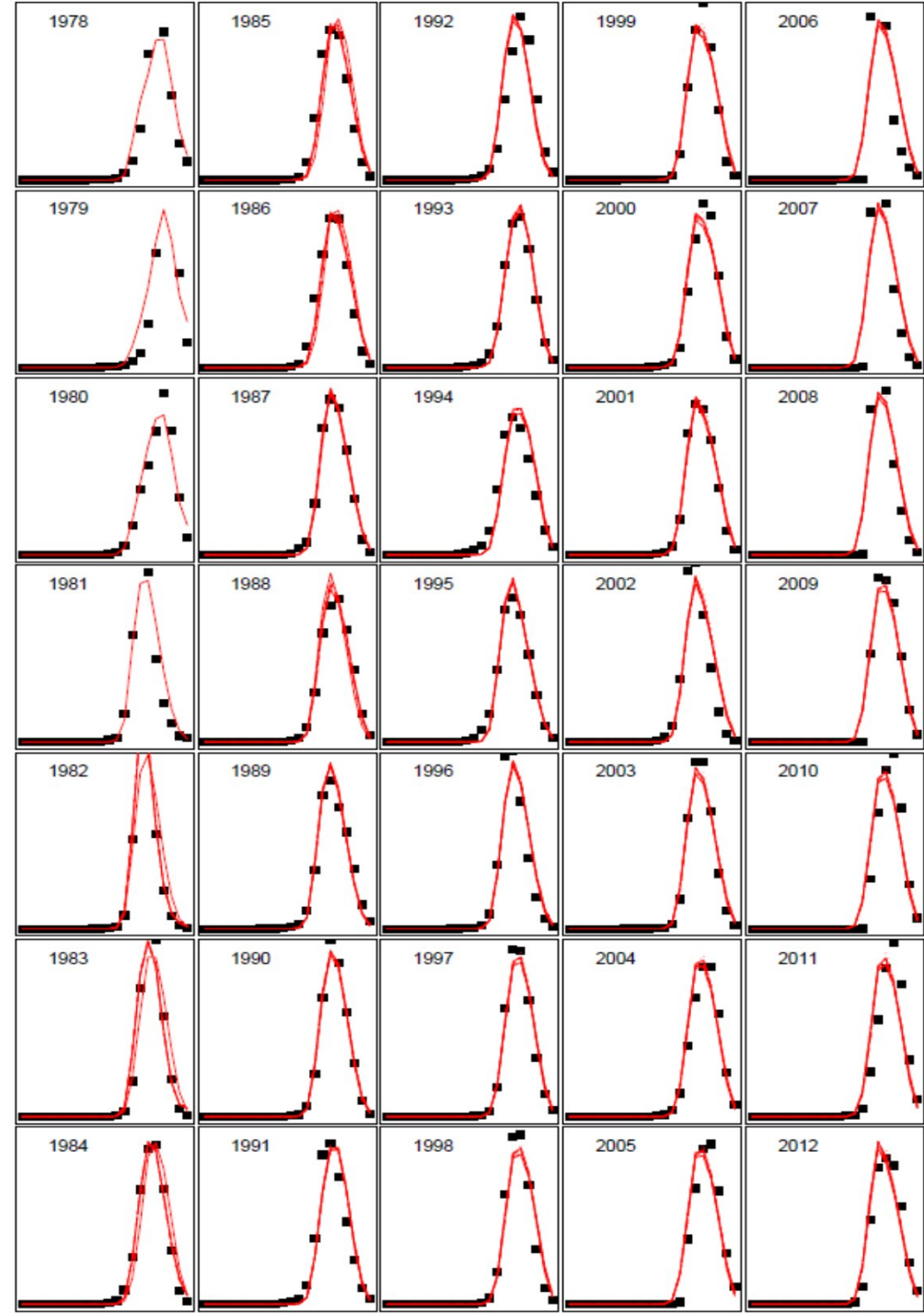
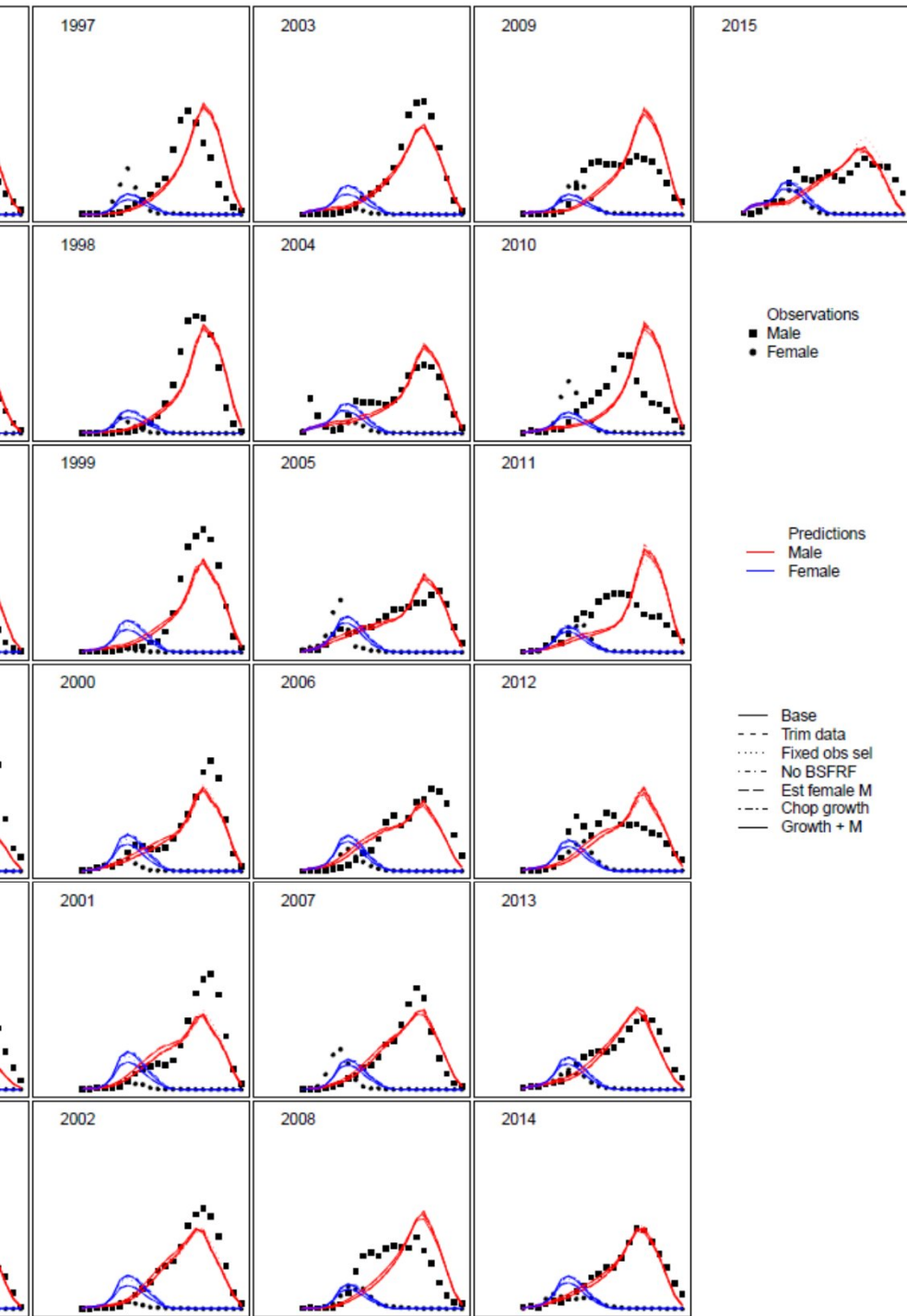
Parameter	Base	Trim data	Fixed obs sel	No BSFRF	Est female M	Chop growth	Growth + M
Mmult_imat	1.8	1.81	1.22	1.74	1.28	1.81	1.27
Mmult	1.13	1.08	1.13	1.06	1.15	1.09	1.16
Mmultf					1.41		1.42



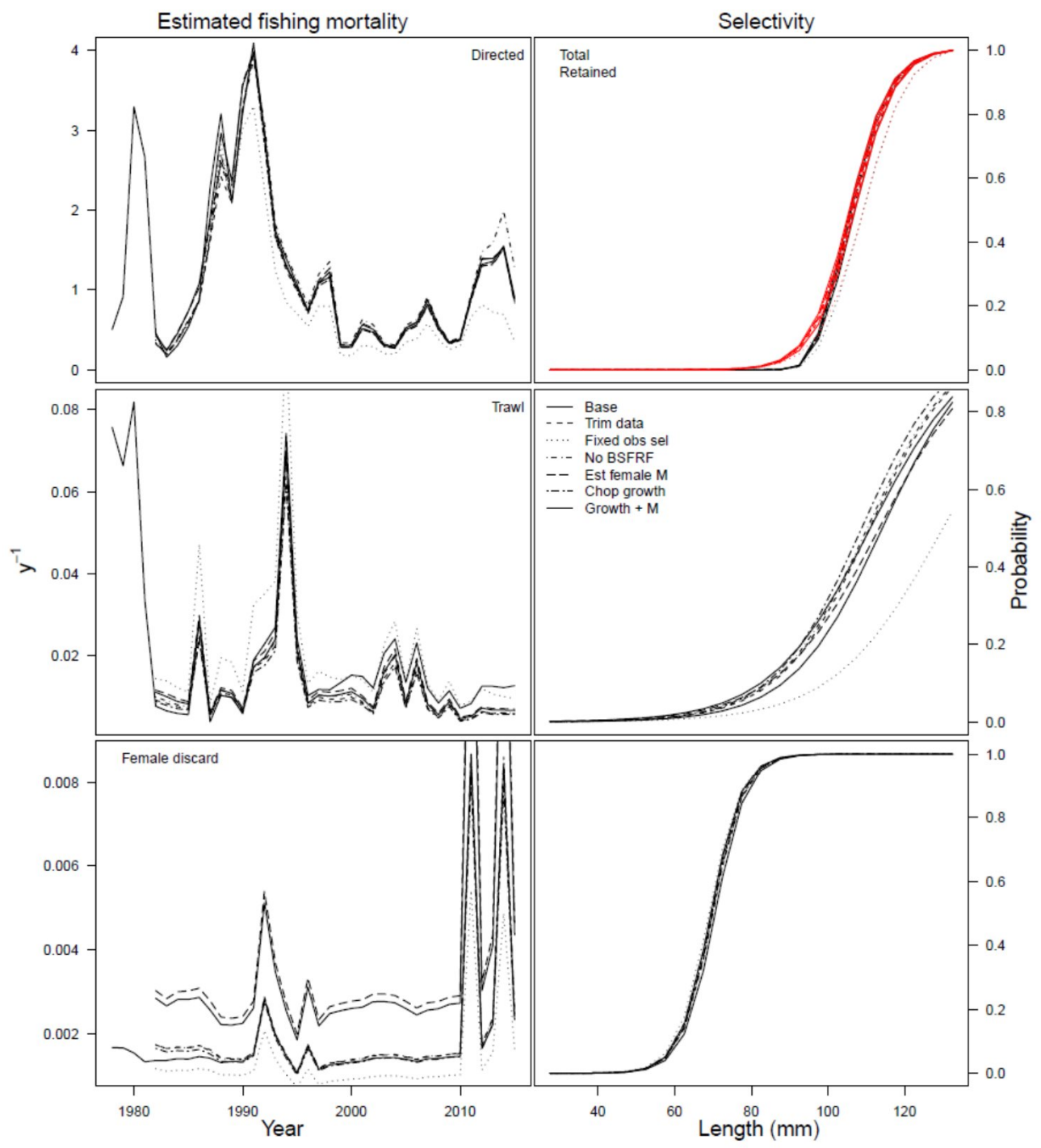
Parameter	Base	Trim data	Fixed obs sel	No BSFRF	Est female M	Chop growth	Growth + M
af	-5.08	-5.06	-4.1	-5.08	-5.09	-5.09	-5.07
am	-5.74	-5.83	-7.48	-12.2	-5.61	-4.89	-4.64
bf	1.53	1.52	1.48	1.53	1.53	1.69	1.69
bm	1.54	1.54	1.62	1.83	1.53	1.19	1.19
b1	1.15	1.15	1.12	1.16	1.15	1.25	1.25
bf1	1.02	1.03	1	1.03	1.04	1.15	1.15
deltam	32.2	32.25	32.37	27.47	32.19	32.33	32.66
deltaf	34.37	34.29	36.51	34.33	34.13	34.30	34.30

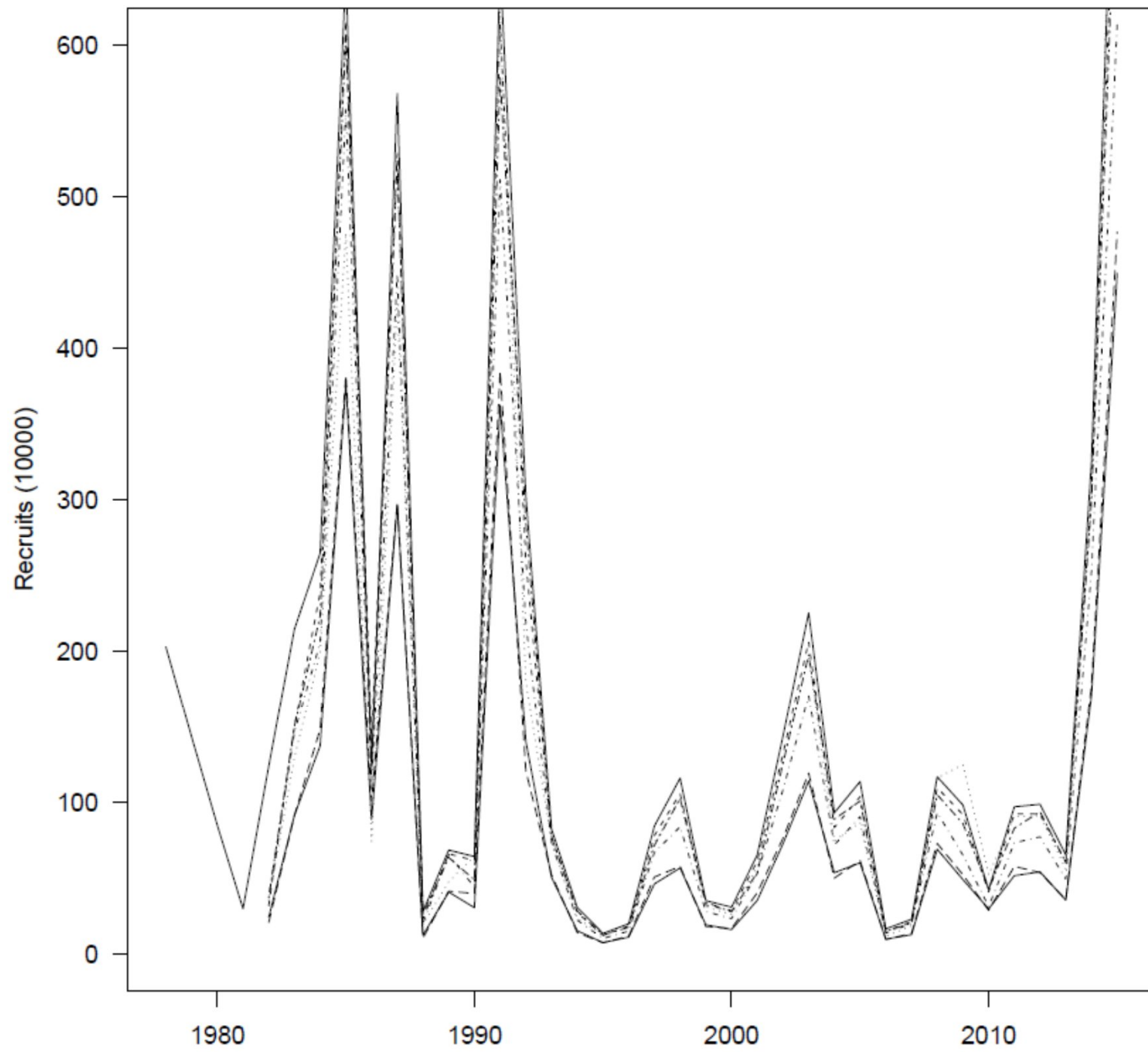


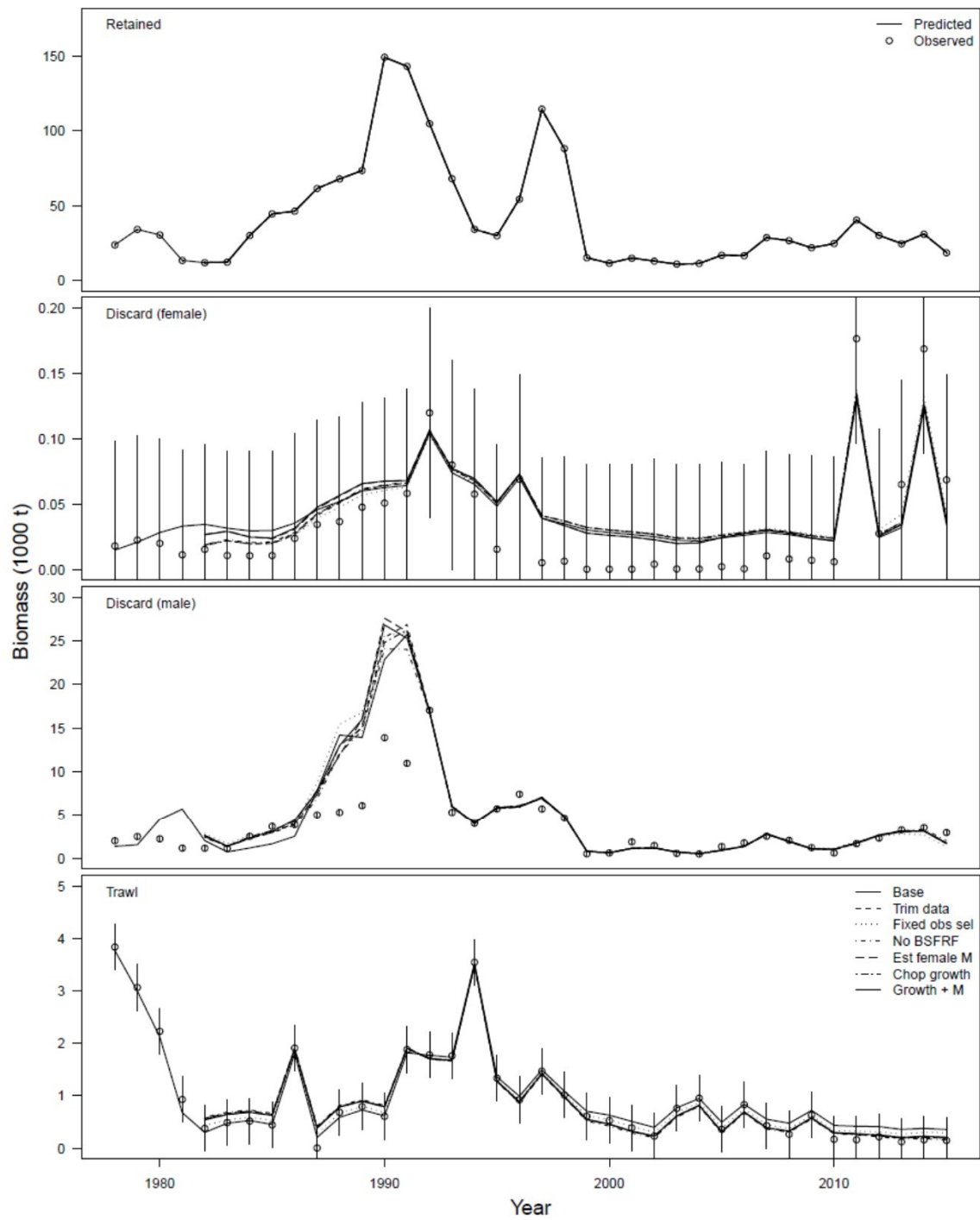












# Model results

## 'm data'

OFL decreased compared to 'Base', primarily because  $q$  in era 3 increased

Decrease in estimated probability of maturing and natural mortality for  $F_{35}$  brought down

## 'fixed obs sel'

OFL increased (a lot) compared to 'Base' from a decrease in  $q$ , probability of maturing, and growth

Much worse fits to survey MMB, survey selectivity much lower than any of the estimated scenarios

## 'BSFRF'

OFL decreased compared to 'Base', primarily because  $q$  in era 3 increased

Shifts breakpoint in growth

$F_{35}$  decreases relative to 'Base' due to decreases in natural mortality and probability of maturing

Table 1: Changes in management quantities for each scenario considered. Reported quantities are the MLEs because running for every model was prohibitively time-consuming. The scenarios in which MCMCs were performed are very close to the medians of the posterior distributions.

Model	MMB	B35	F35	FOFL
Base	92.00	152.3	1.01	1.14
Trim data	83.8	152.3	1.42	0.81
Fixed obs sel	221.6	215.3	2.40	2.62
No BSFRF	60.86	142.3	1.17	0.56
Est female M	74.29	139.6	1.21	0.68
Chop growth	79.57	149.8	1.34	0.75
Growth + M	70.89	137.4	1.17	0.64

# Recommendations

m data'

Adopt exclusion of all data from 1978-1981, start model in 1982

Rationale: Including this era artificially anchors catchability at 1. Given the uncertainties around expected changes under different survey gear and the potential for catch to be missing in the early years, the risks outweigh the benefits

Era 3 should start in 1988 instead of 1989

'ked obs sel' & "No BSFRF"

Neither of these should be adopted, but were used illustratively

Think harder about how to do incorporate extra survey

Issues:

- problems with variables hitting bounds
- Large disconnect between 'observed' selectivity and estimated
- Stock is at it's lowest, in spite of an assumption of a higher  $q$  than implied by the industry surveys—changing this assumption to the 'observed' would have exacerbated the decline in MMB

# atural mortality

urrent:

Immature M (male and female):

- 0.41 (estimated)
- Prior =  $N(0.23, 0.154)$

Mature female M:

- 0.23 (fixed)

Mature male M:

- 0.26 (estimated)
- Prior =  $N(0.23, 0.054)$

ues

Mature male M was higher than female, which was biologically questionable

No natural mortality specific data (e.g. tagging data)

Has a large impact on reference points

Poorly documented rationale



# del runs

t female M'

'Trim data' + estimating mature female M + setting the prior for immature crab equal to prior for mature crab

Potentially corrects for flip-flop of M between sexes and corrects the prior to conform to the rationale of M being based on longevity

# Model results

## Prim data'

### Immature M (male and female):

- 0.29 (estimated, decreased from 0.41)
- Prior =  $N(0.23, 0.054)$  [sd decreased from 0.154]

### Mature female M:

- 0.32 (estimated, increased from 0.23)
- Prior =  $N(0.23, 0.054)$

### Mature male M:

- 0.26 (estimated; did not change)
- Prior =  $N(0.23, 0.054)$

Survey catchability is pegged at 1 for females when estimating mature M  
OFL decreased compared to 'Base', primarily because  $q$  in era 3 increased  
Decrease in estimated probability of maturing for brought down F35%

Parameter	Base	Trim data	Fixed obs sel	No BSFRF	Est female M	Chop growth	Growth + M
srv1_q	1	60.13	60.13	60.13	60.13	60.13	60.13
srv1_q_f	1	60.13	60.13	60.13	60.13	60.13	60.13
srv1_sel95	59.89	60.13	60.13	60.74	60.74	60.34	60.59
srv1_sel50	42.66	41.18	41.18	41.63	41.7	41.35	41.34
srv2_q	0.49	0.43	0.43	0.47	0.54	0.45	0.56
srv2_q_f	0.32	0.46	0.46	0.49	0.63	0.46	0.63
srv2_sel95	61.3	57.05	57.05	57.32	55.24	58.05	56.01
srv2_sel50	41.32	41.18	41.18	40.84	39.82	41.35	39.86
srv3_q	0.62	0.68	0.68	0.79	0.77	0.7	0.79
srv3_sel95	57.24	57.63	57.63	59.43	49.53	59.37	50.62
srv3_sel50	38.42	38.59	38.59	38.78	34.78	39.15	34.94
srv3_q_f	0.49	0.54	0.54	0.62	1	0.54	1
srv3_sel95_f	43.09	43.42	43.42	42.85	45.23	42.84	44.8
srv3_sel50_f	33.27	33.47	33.47	32.97	34.73	32.94	34.27

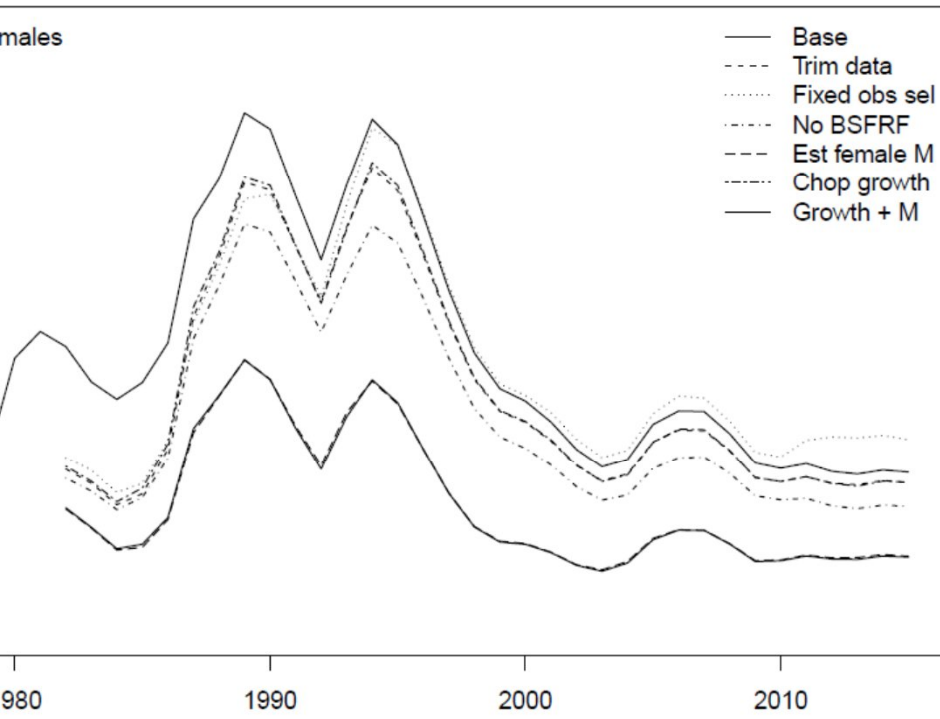
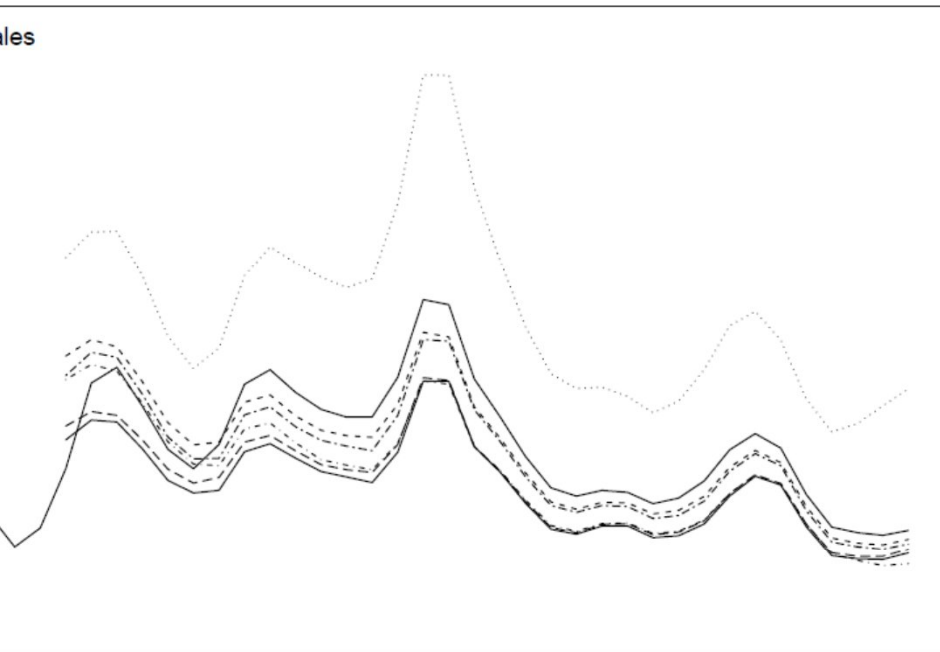


Table 1: Changes in management quantities for each scenario considered. Reported quantities are the MLEs because running MCMC for every model was prohibitively time-consuming. The MLEs for scenarios in which MCMCs were performed are very close to the medians of the posterior distributions.

Model	MMB	B35	F35	FOFL
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Chop growth	79.57	149.8	1.34	0.75
Growth + M	70.89	137.4	1.17	0.64



# Recommendations

It makes sense to estimate mature female natural mortality and corrects the relationship between mature male and mature female  $M$ , but now survival for females hits its bound of 1.

Mature natural mortality is now flipped, but this is a result of placing the prior on immature  $M$  that is placed on mature  $M$ .



# Growth model and available data

Current: Piece-wise linear model

Estimated parameters

data points

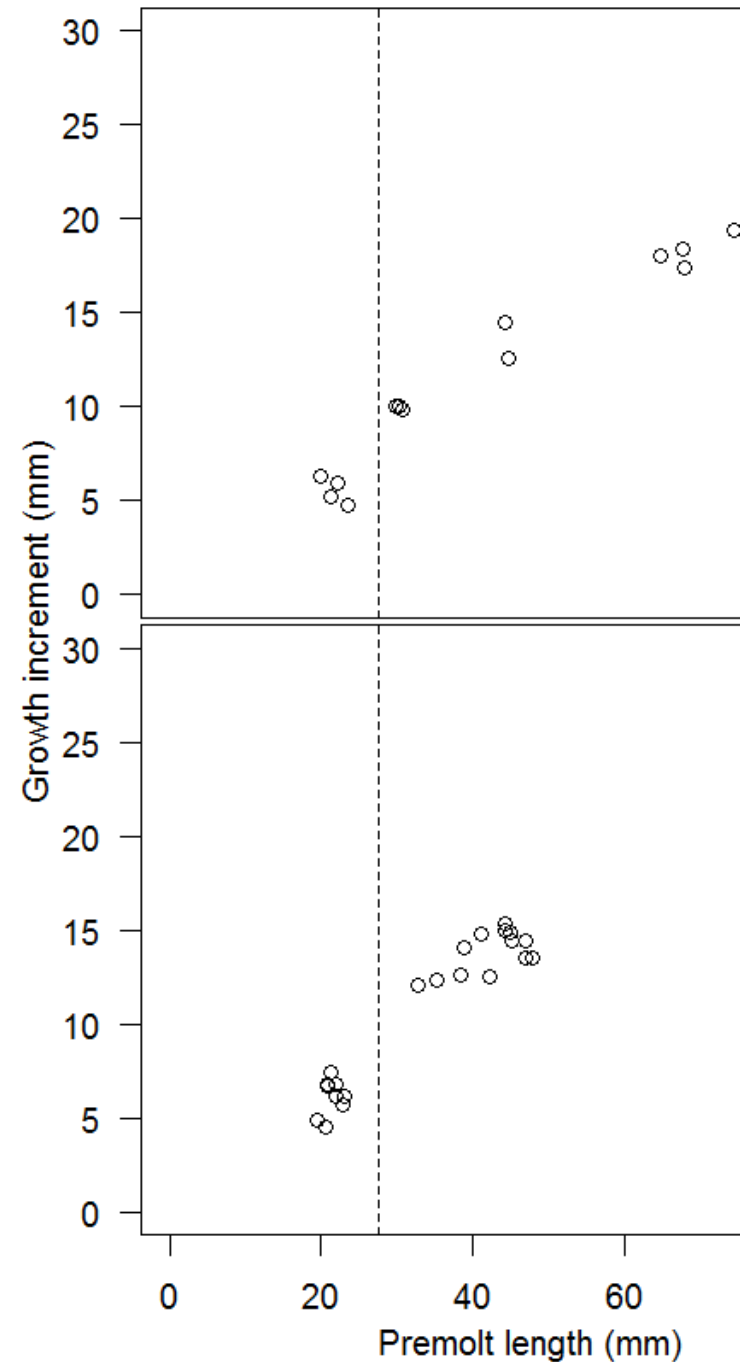
Issues

No data where the breakpoint, resulting in poor estimation

Data beneath the breakpoint impacts the model little

Growth parameters hit bounds and are generally poorly behaved

What model should be used for growth?



# Growth model and available data

Current: Piece-wise linear model

Estimated parameters

data points

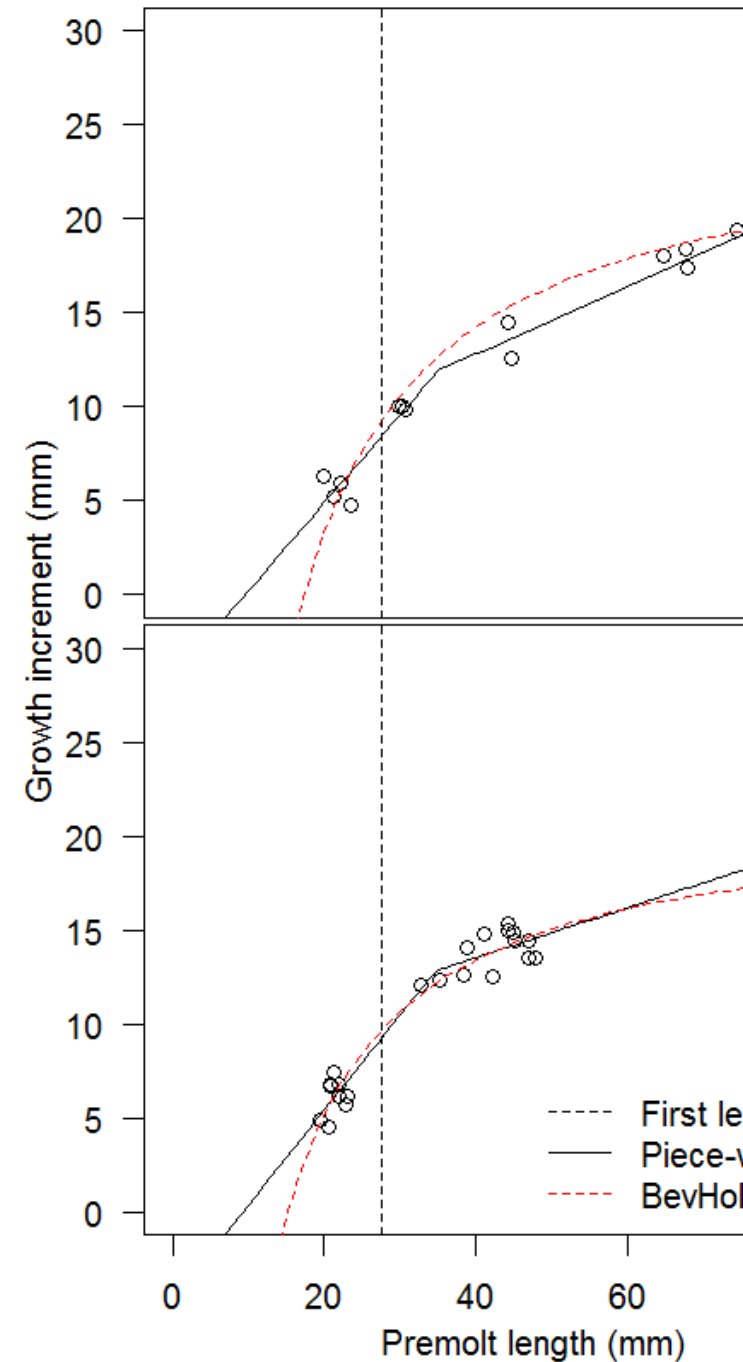
UBS

No data where the breakpoint, resulting in poor estimation

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What model should be used for growth?



# Growth model and available data

Current: Piece-wise linear model

Estimated parameters

data points

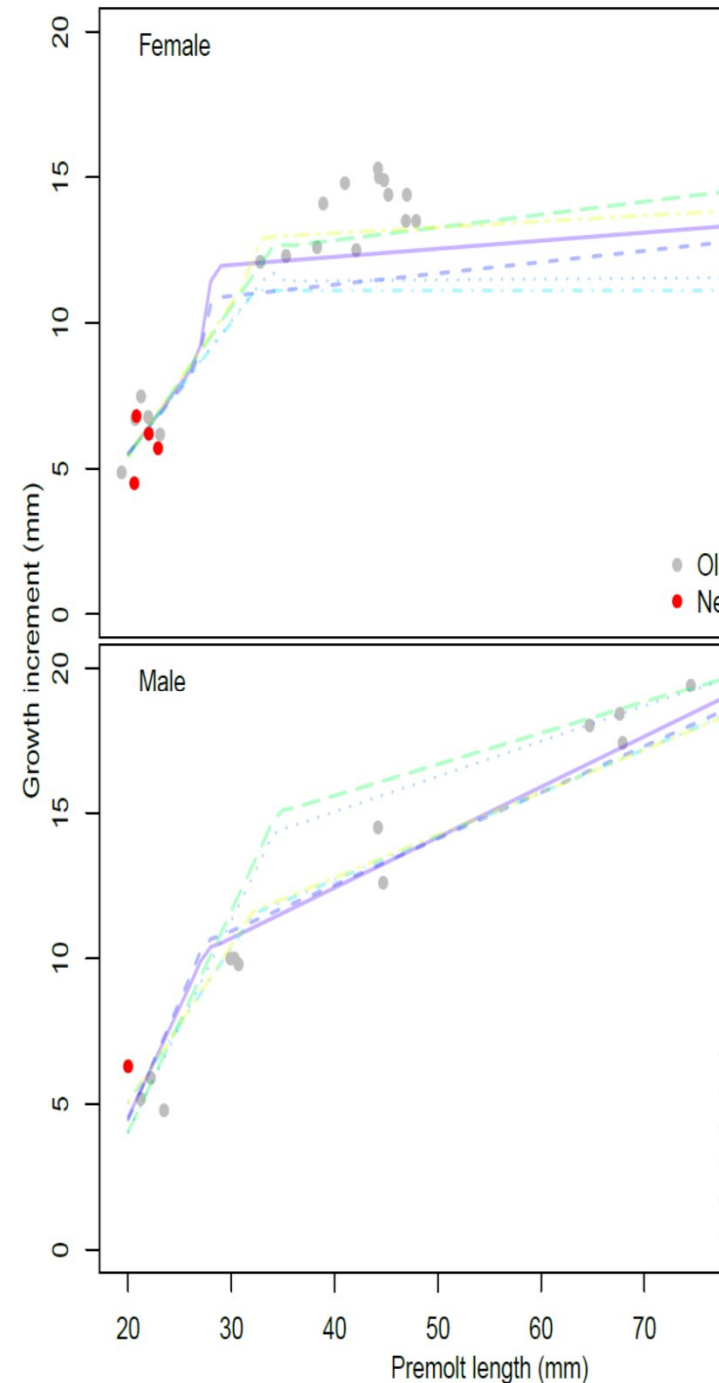
Issues

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Data beneath the breakpoint impacts the model little

Growth parameters hit bounds and are generally poorly behaved

What model should be used for growth?



# Model runs

'Chop growth'

"Trim data" + excludes all growth data with a premolt size of  $<27.5$  mm, then estimates only a linear model for both males and females

Explores problem of bound hitting and poorly estimated growth parameters

Growth + M'

'Chop growth' + estimating mature female M

Growth and natural mortality are somewhat confounded

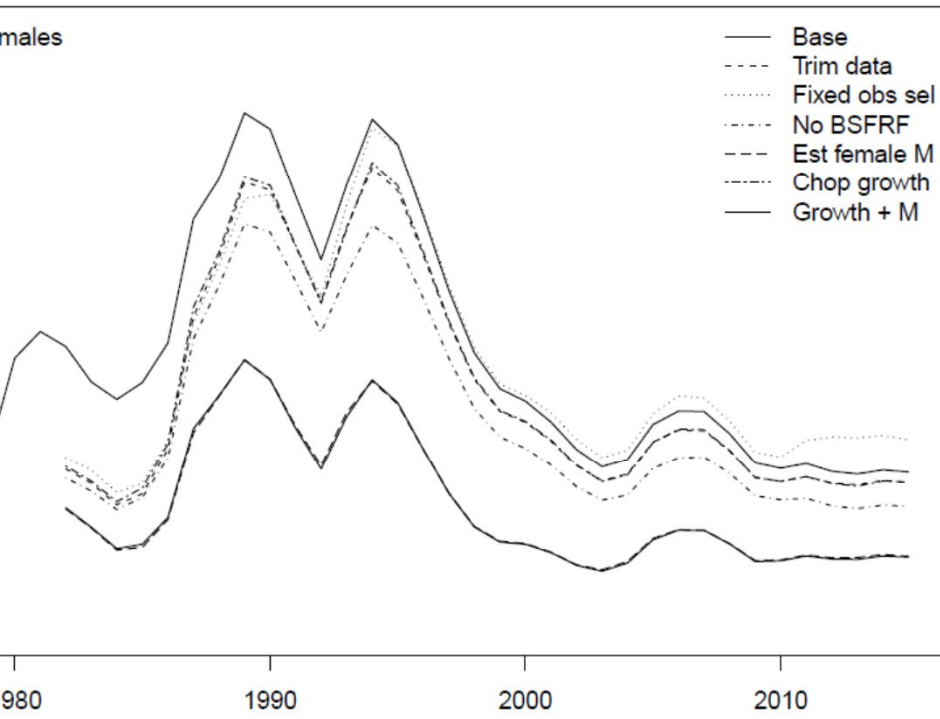
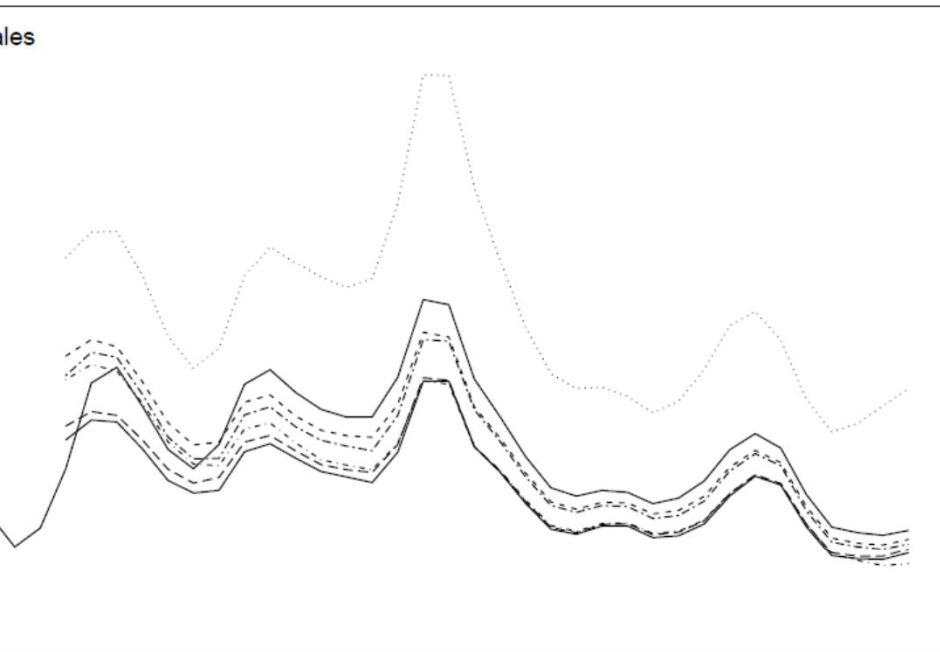
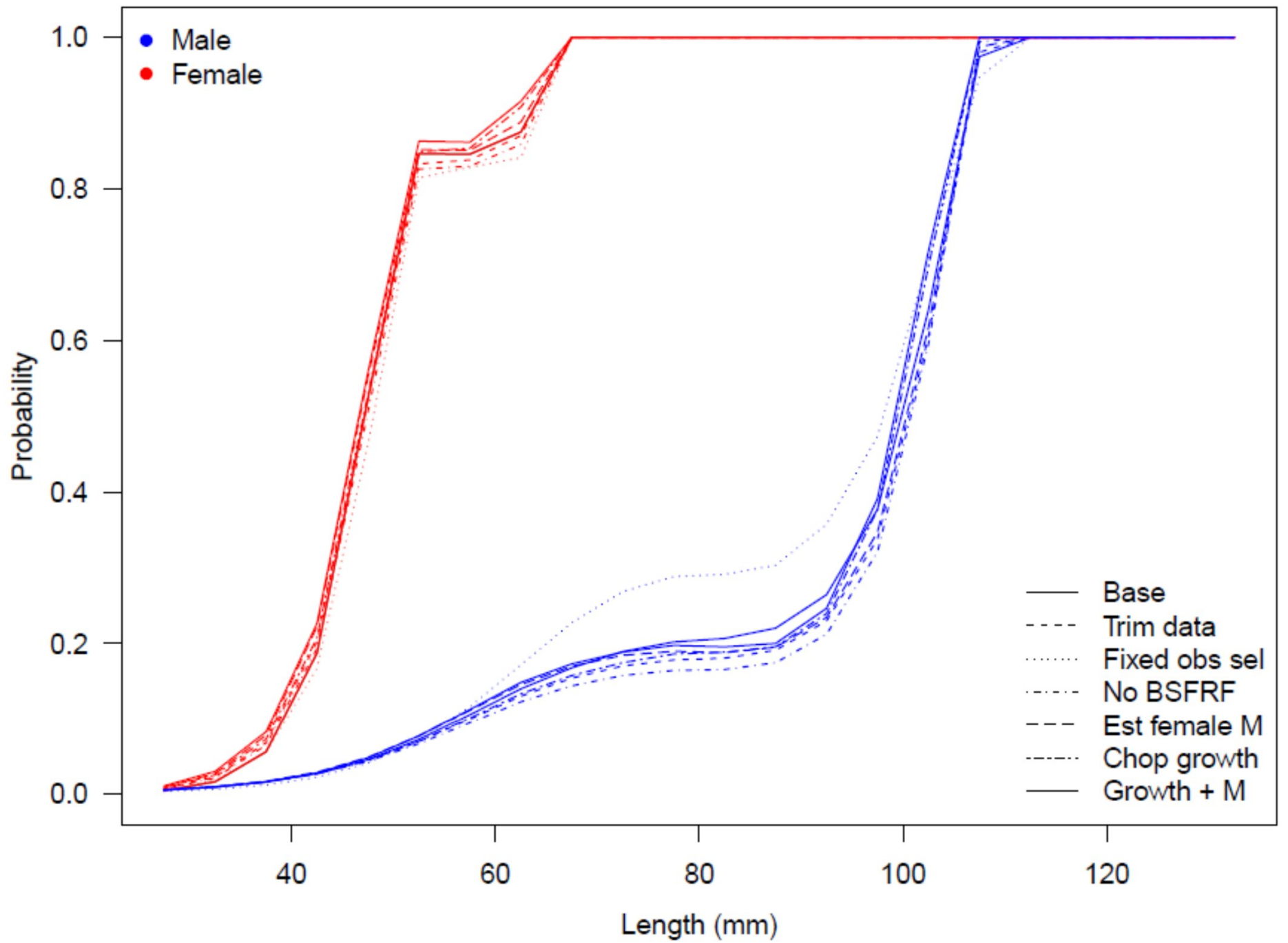


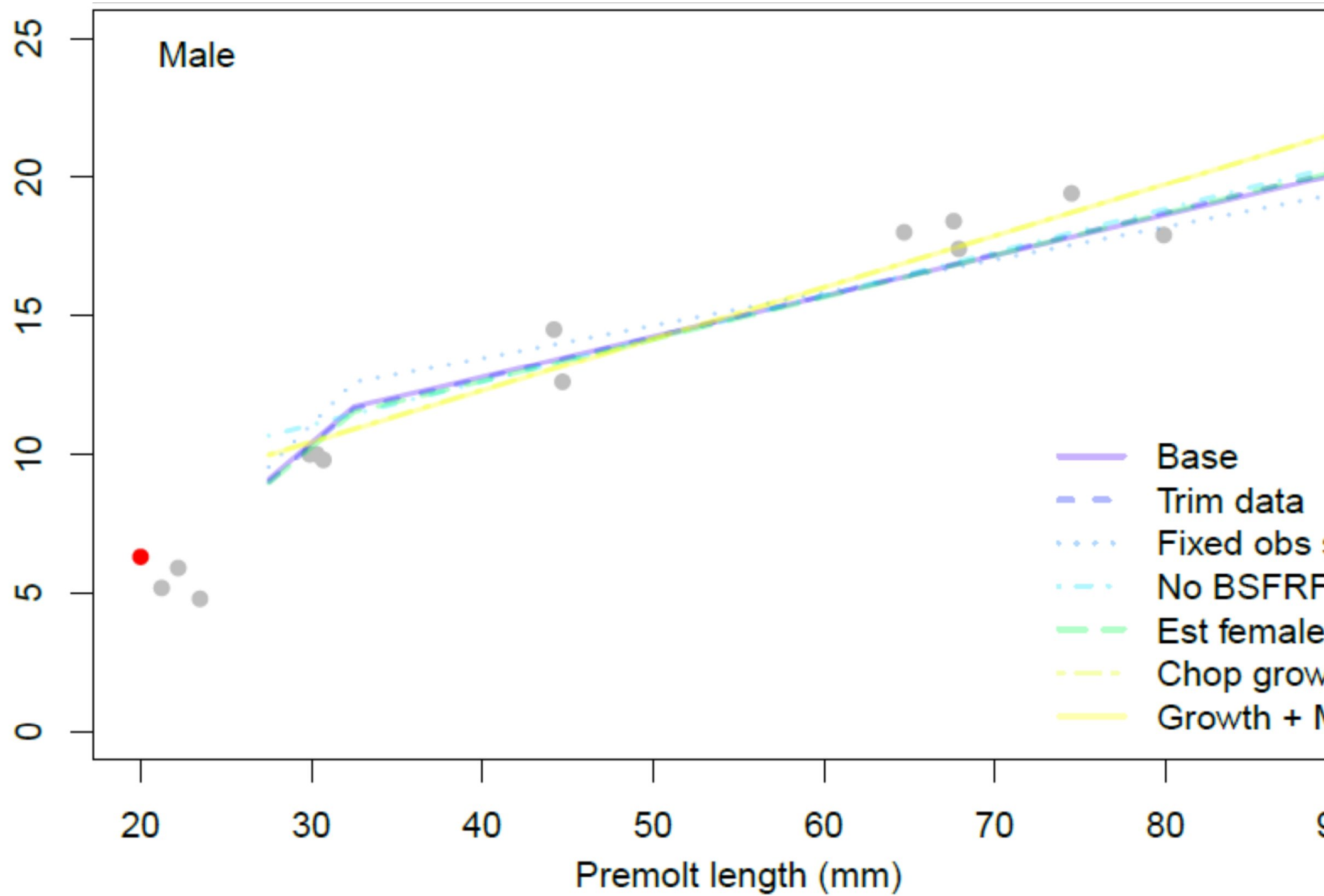
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Parameter	Base	Trim data	Fixed obs sel	No BSFRF	Est female M	Chop growth	Growth + M
srv1_q	1						
srv1_q_f	1						
srv1_sel95	59.89						
srv1_sel50	42.66						
srv2_q	0.49	0.43		0.47	0.54	0.45	0.56
srv2_q_f	0.32	0.46		0.49	0.63	0.46	0.63
srv2_sel95	61.3	57.05		57.32	55.24	58.05	56.01
srv2_sel50	41.32	41.18		40.84	39.82	41.35	39.86
srv3_q	0.62	0.68		0.79	0.77	0.7	0.79
srv3_sel95	57.24	57.63		59.43	49.53	59.37	50.62
srv3_sel50	38.42	38.59		38.78	34.78	39.15	34.94
srv3_q_f	0.49	0.54		0.62	1	0.54	1
srv3_sel95_f	43.09	43.42		42.85	45.23	42.84	44.8
srv3_sel50_f	33.27	33.47		32.97	34.73	32.94	34.27

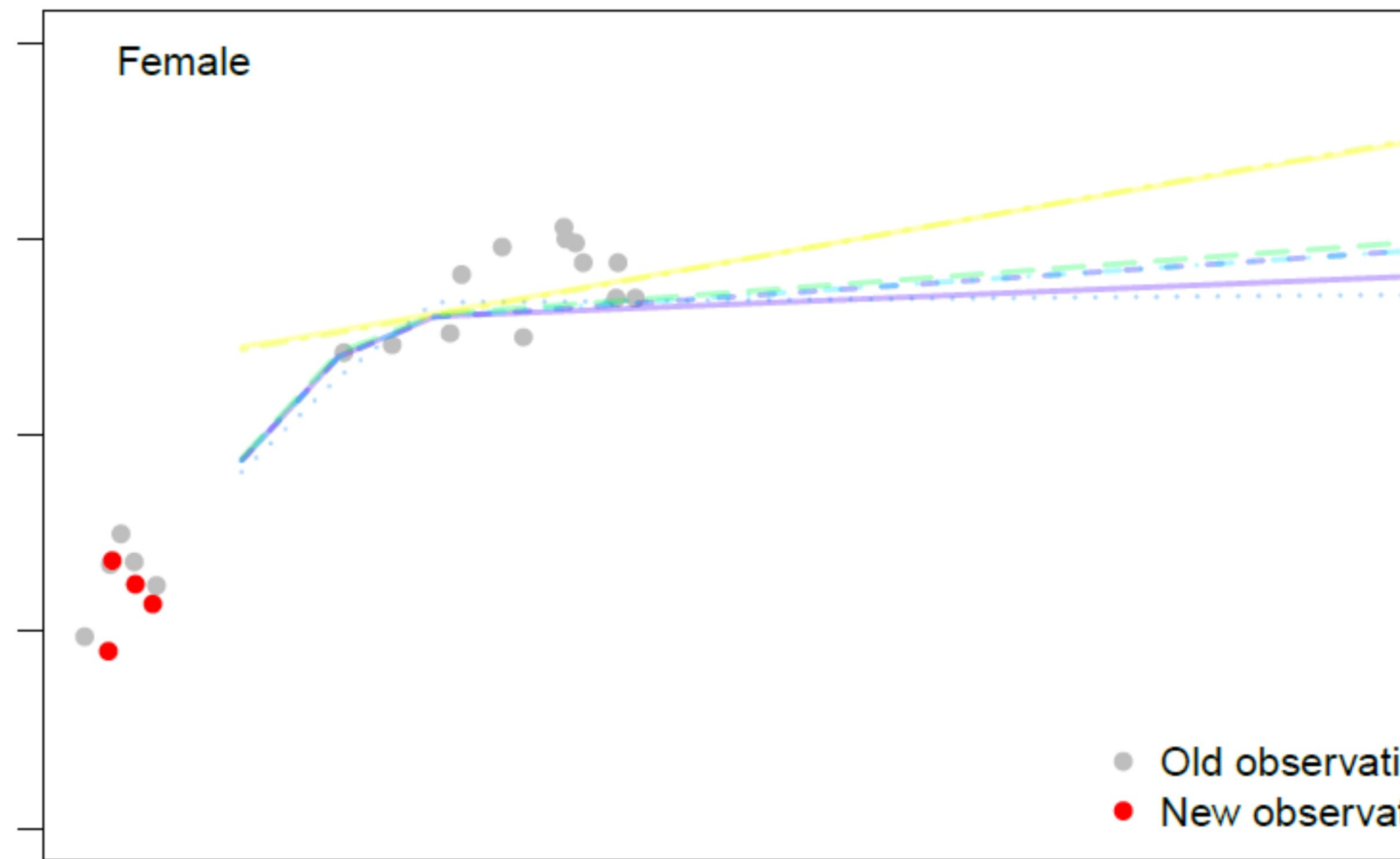






Female

- Old observati
- New observa



# Model results

Growth + M'

'Growth + M' is the 'synthesis' of all the changes—excludes 1978-1981 data, estimates mature female M, eliminates problem parameters from growth

BUT, new problems arise:

- Survey q (females era 3) is now estimated at 1
- Survey q (males era 3) is now estimated at 0.79, which is a large increase over the 'Base'

# Recommendations summary

Select a method for computing a distribution of the OFL based on uncertainty in the data and parameter estimates

Use 1978-1981 data

Start era 3 in 1988

Use a model selection approach to identify a model other than the piece-wise linear models for growth?

Estimate mature female  $M$ , but whack-a-mole era 3 survey  $q$ ?