

An age-structured assessment model for yelloweye rockfish (*Sebastes ruberrimus*) in Southeast Alaska Outside Waters

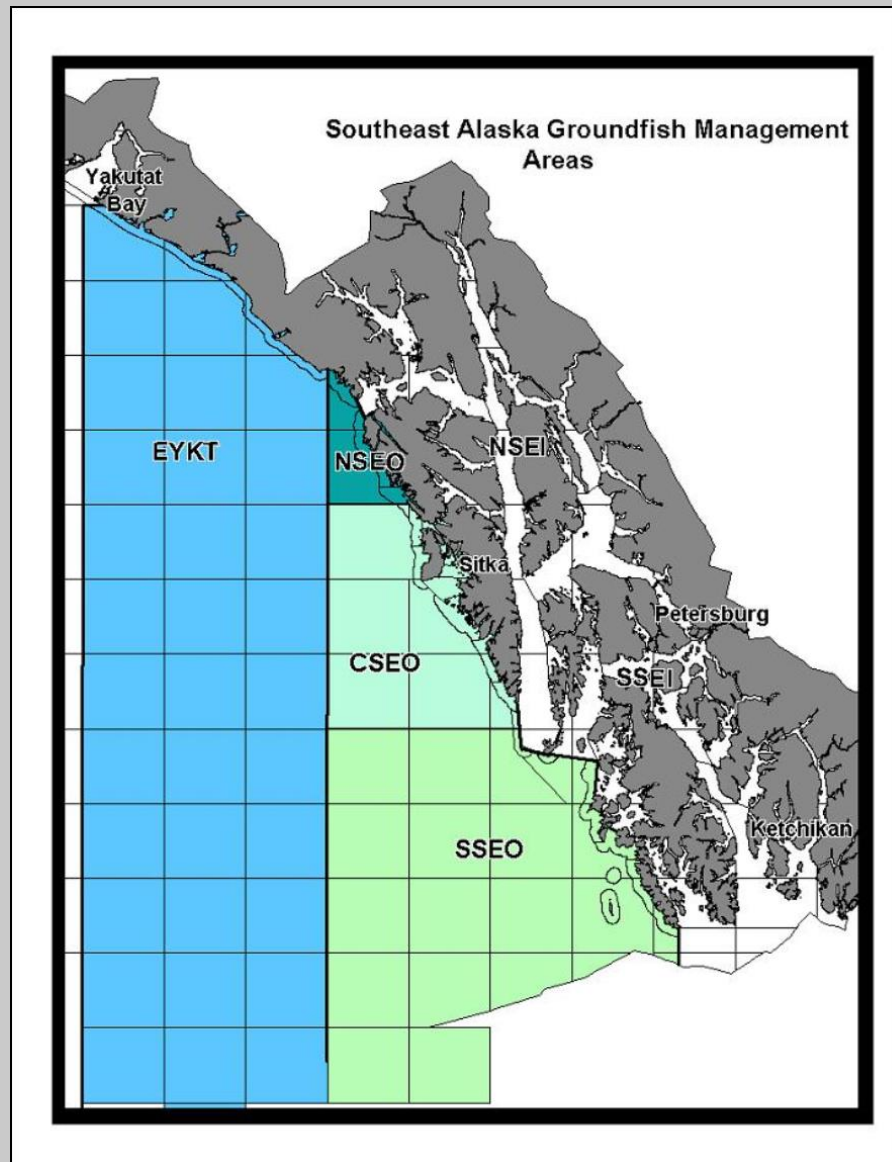


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Southeast Alaska Outside Waters





Changes to model data & structure



Data updated through 2015

1. Total annual catch:
Commercial fishery, sport fishery, halibut fishery bycatch
2. Age composition:
Commercial fishery, halibut fishery bycatch
3. Density:
ROV survey

Structural changes

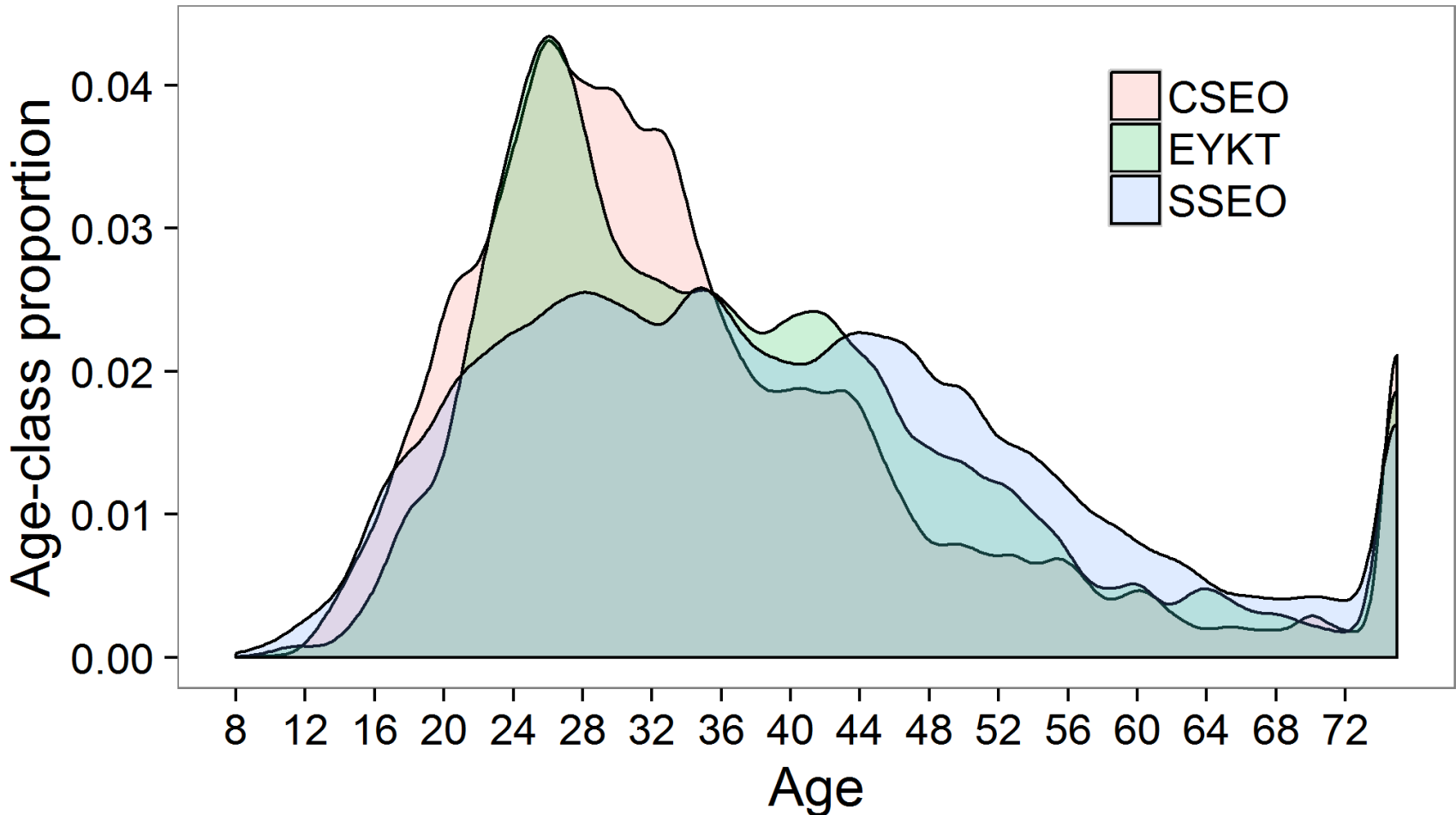
1. Terminal plus-class changed from 97+ to 75+
2. Natural mortality is estimated
3. CPUE scaled
4. Lower 90% CI for model-estimated biomass, F_{xx} ,
and ABC used when evaluating potential harvest levels
5. Additional sigma parameter for density from last year's
assessment removed due to confounding with estimating
natural mortality



Changes to model structure



- Plus-class changed from 97+ to 75+
- number of age classes was reduced
- proportion of individuals in the plus-class did not exceed any sub-plus-class age proportion





Four model structures



Model 1:

1. Regionally-distinct data and likelihood;
2. Asymptotic fishery selectivity-at-age

Model 2:

1. Regionally-distinct data and likelihood;
2. Common parameters:
 - a. natural mortality
 - b. commercial fisheries catchability
 - c. IPHC survey catchability
3. Asymptotic fishery selectivity-at-age

Model 3:

1. Regionally-distinct data and likelihood;
2. Common parameters:
 - a. natural mortality
 - b. commercial fisheries catchability
 - c. IPHC survey catchability
3. Dome-shaped fishery selectivity-at-age option

Model 4: (global)

1. Data and likelihood merged over regions;
2. Common parameters:
 - a. natural mortality
 - b. commercial fisheries catchability
 - c. IPHC survey catchability
 - d. mean age-8 recruitment
 - e. mean year-1 abundance
 - f. sigma for year-1 abundance deviation vector
 - g. mean full-recruitment fishing mortality
 - h. selectivity curve parameters
 - i. annual deviation vectors for recruitment, abundance, and fishing mortality
3. Asymptotic fishery selectivity-at-age



Four model structures



Alternative structures

Multivariate logistic likelihood for age composition

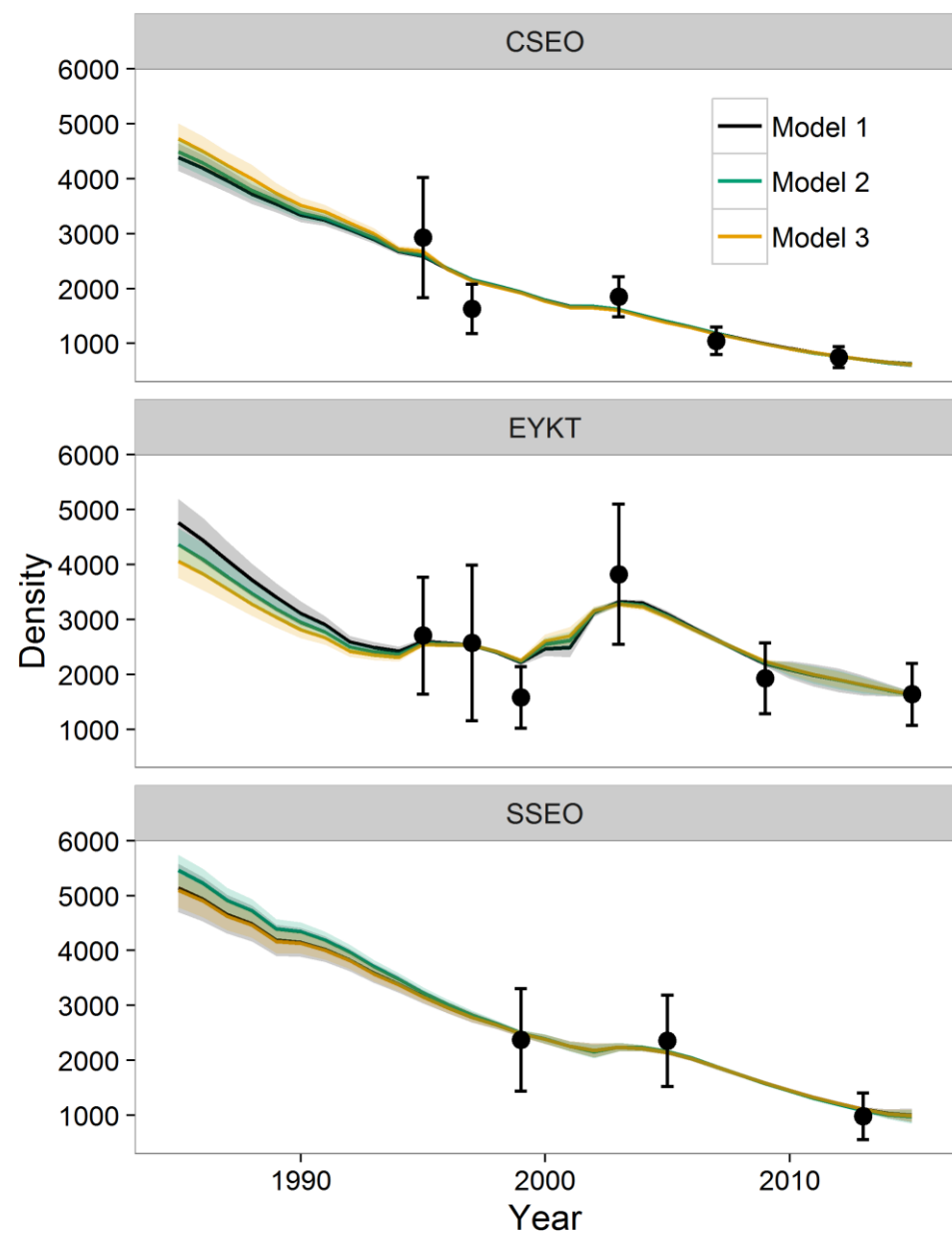
Partitioning global dataset to fit regional likelihoods

Spawner-recruit curves

Global recruitment partitioned into region-specific recruitment

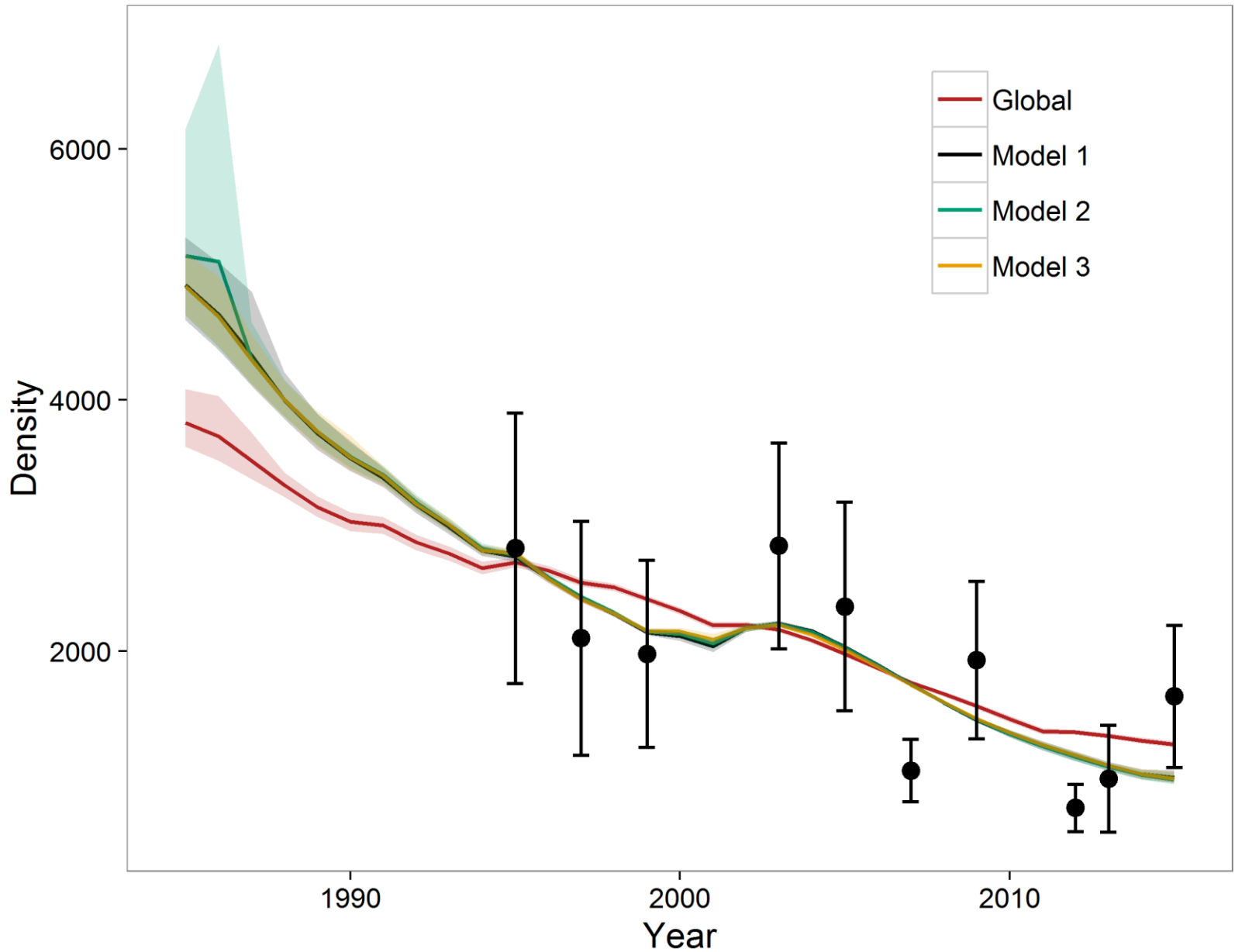


Results: Regional density



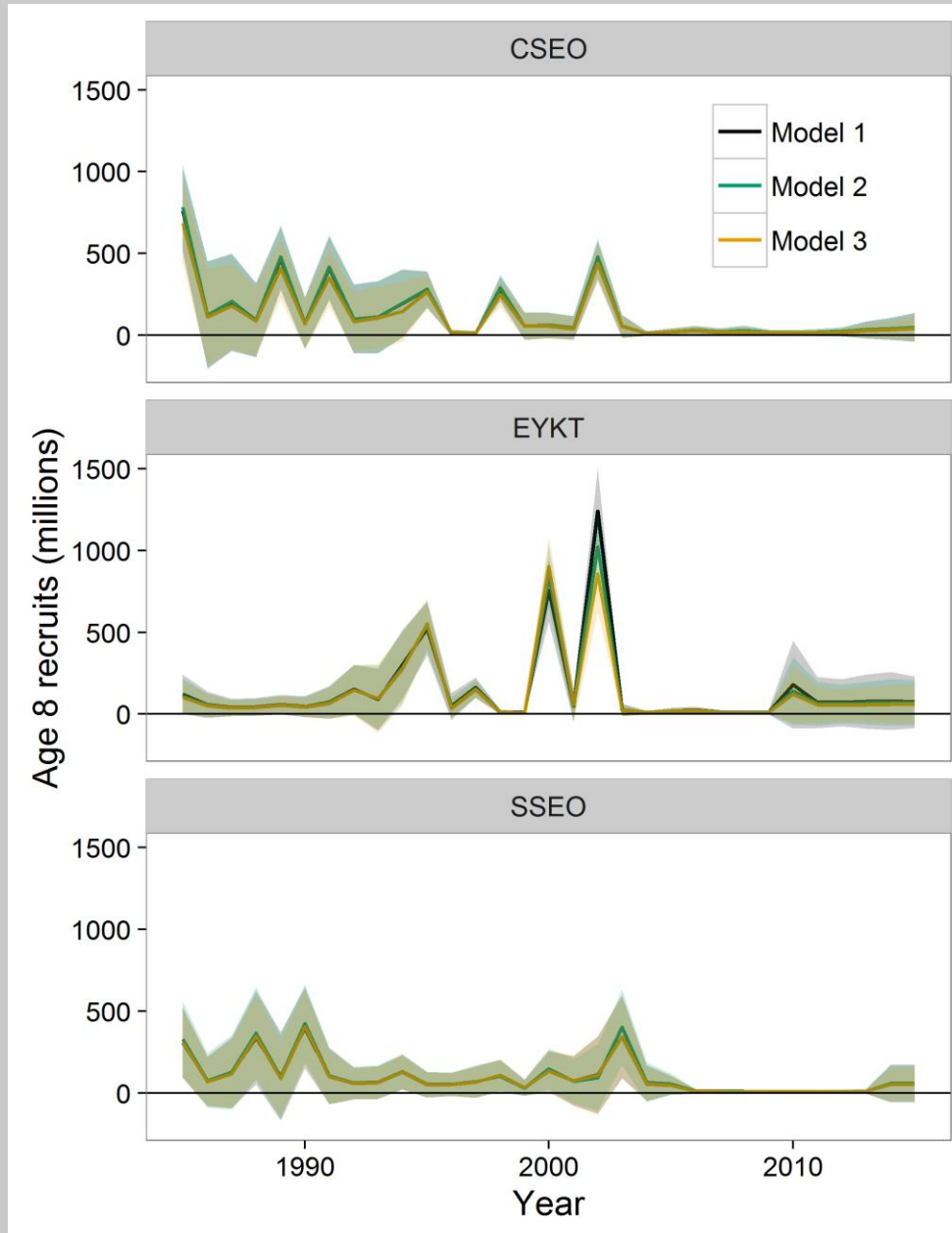


Results: Total density



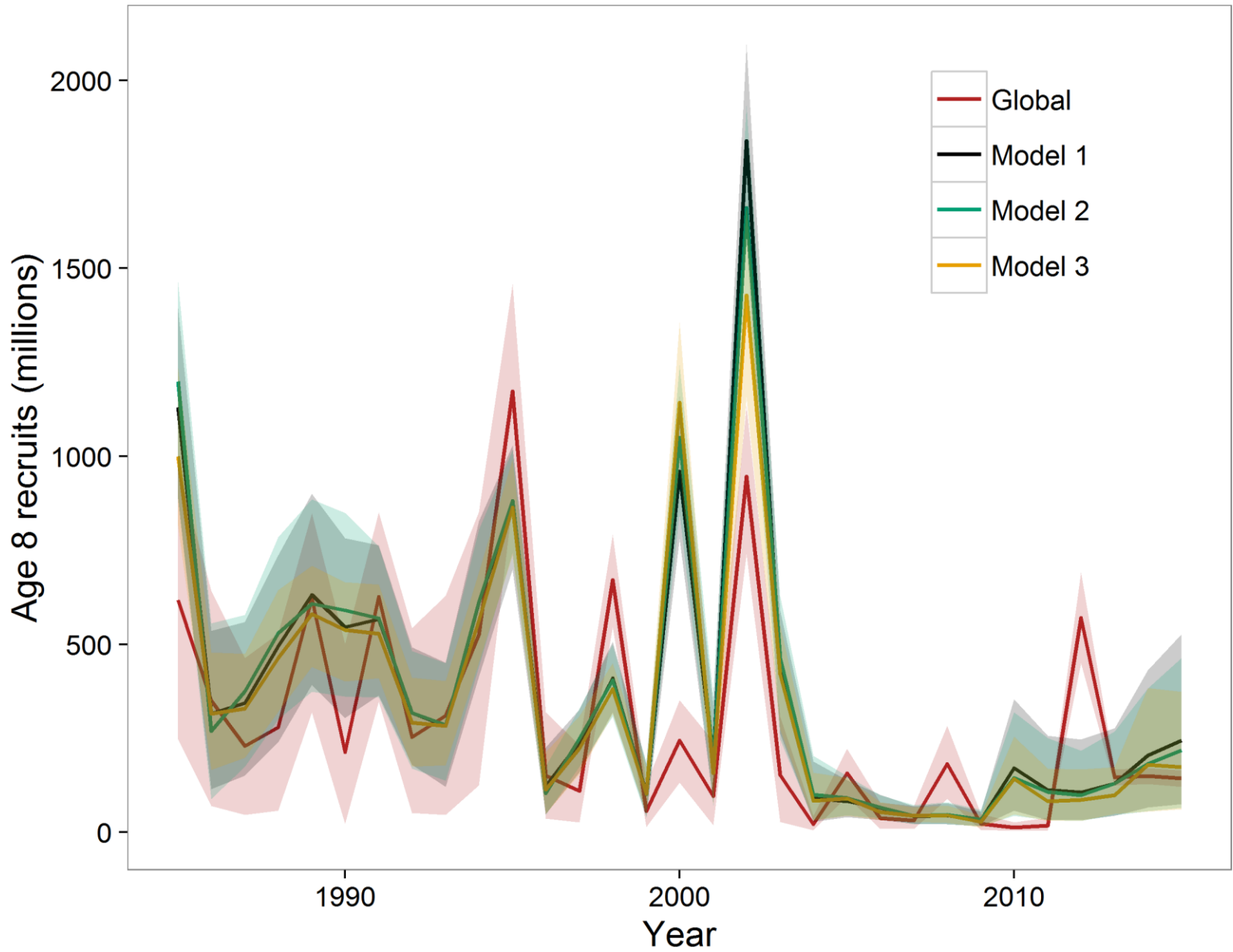


Results: Regional recruitment



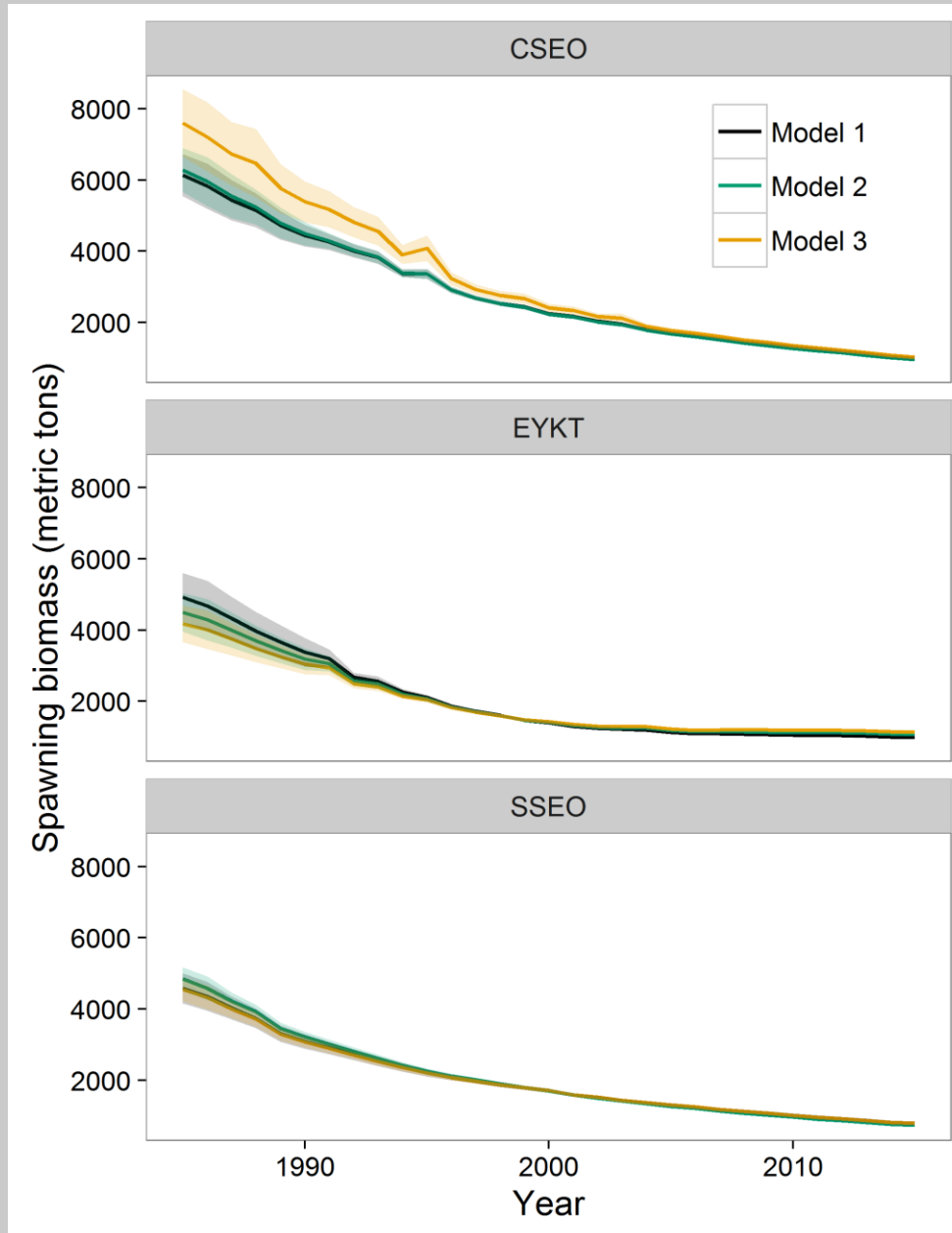


Total recruitment



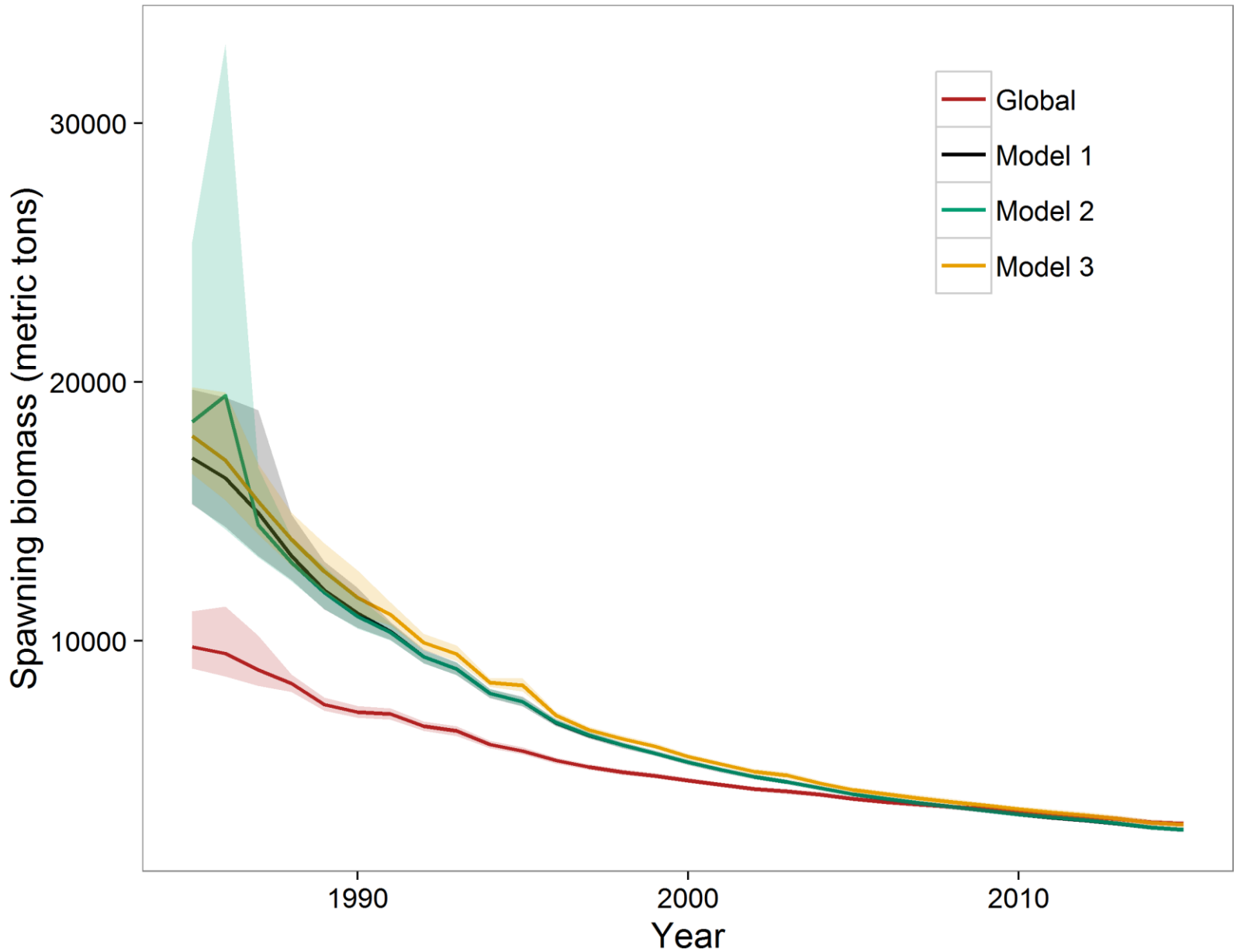


Results: Regional spawning biomass



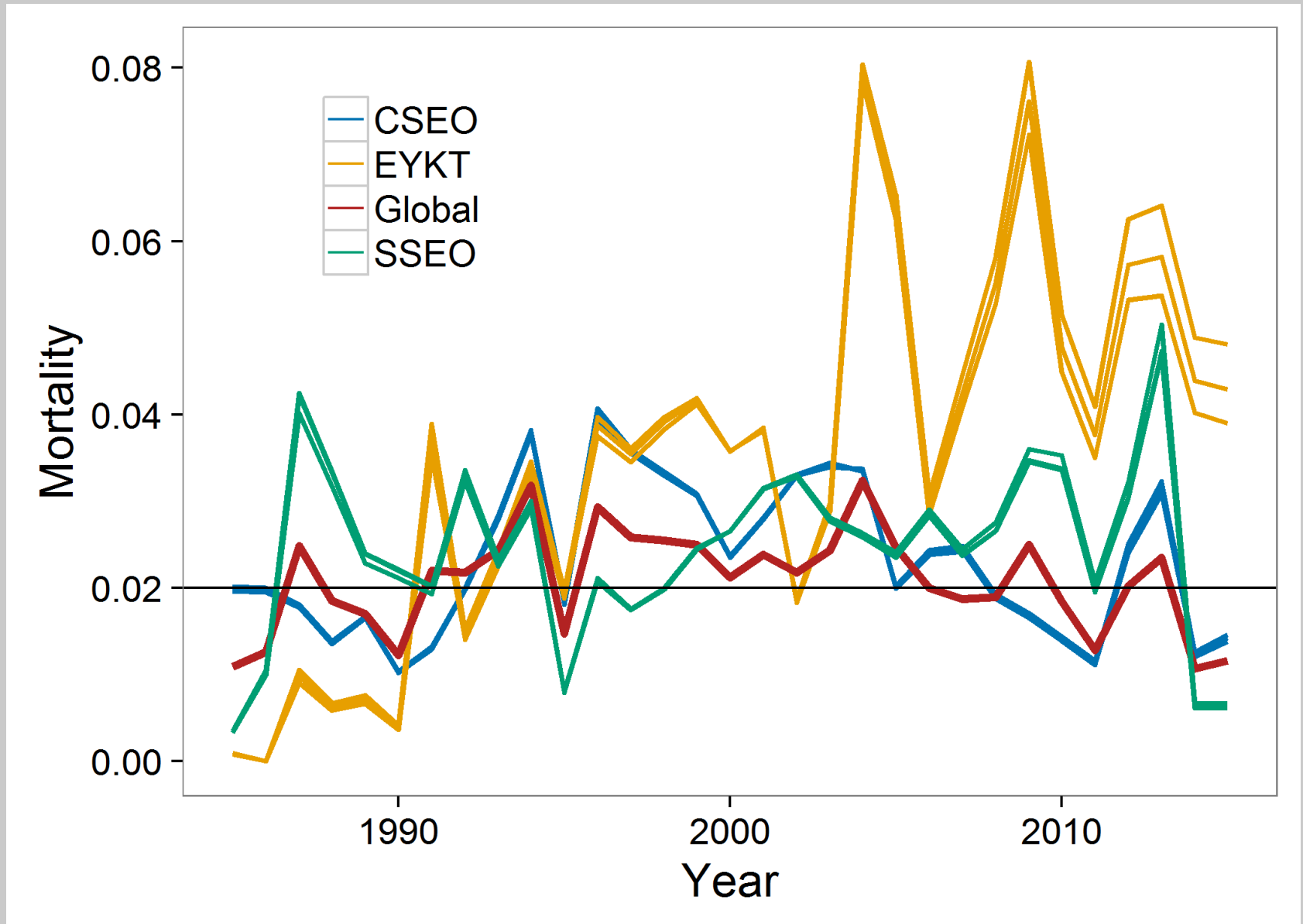


Result: Total spawning biomass



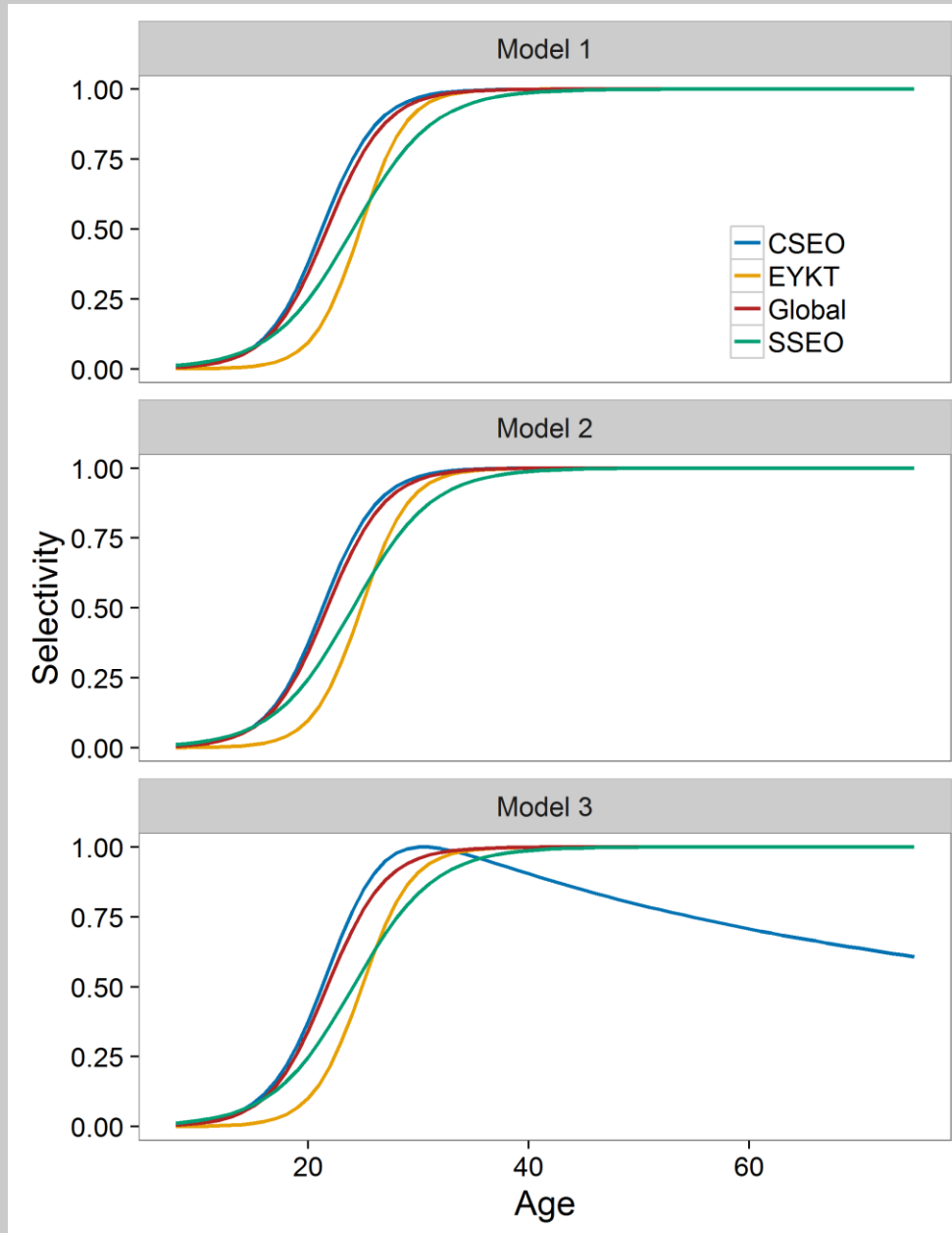


Results: Full-recruitment fishing mortality



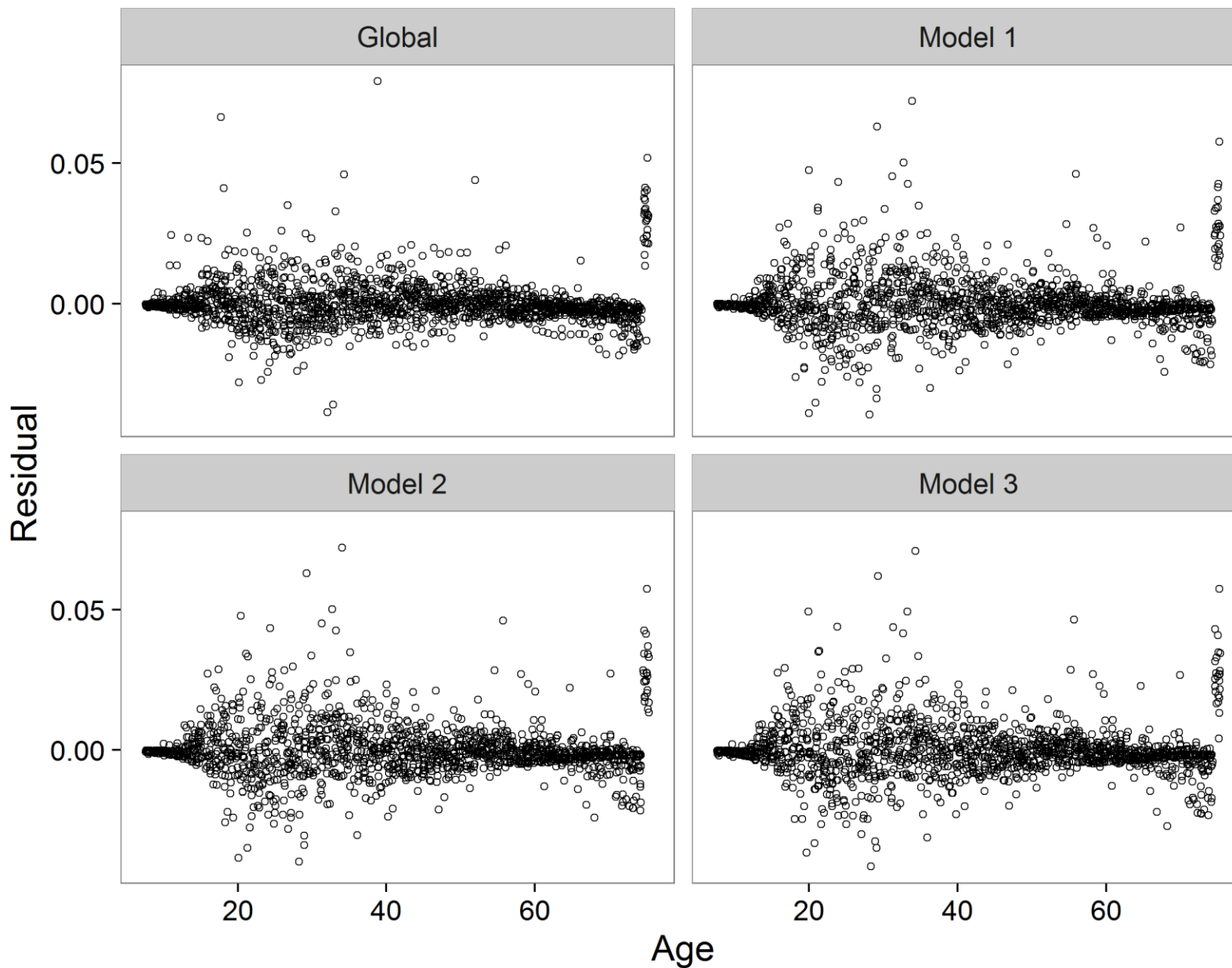


Results: Fishery selectivity



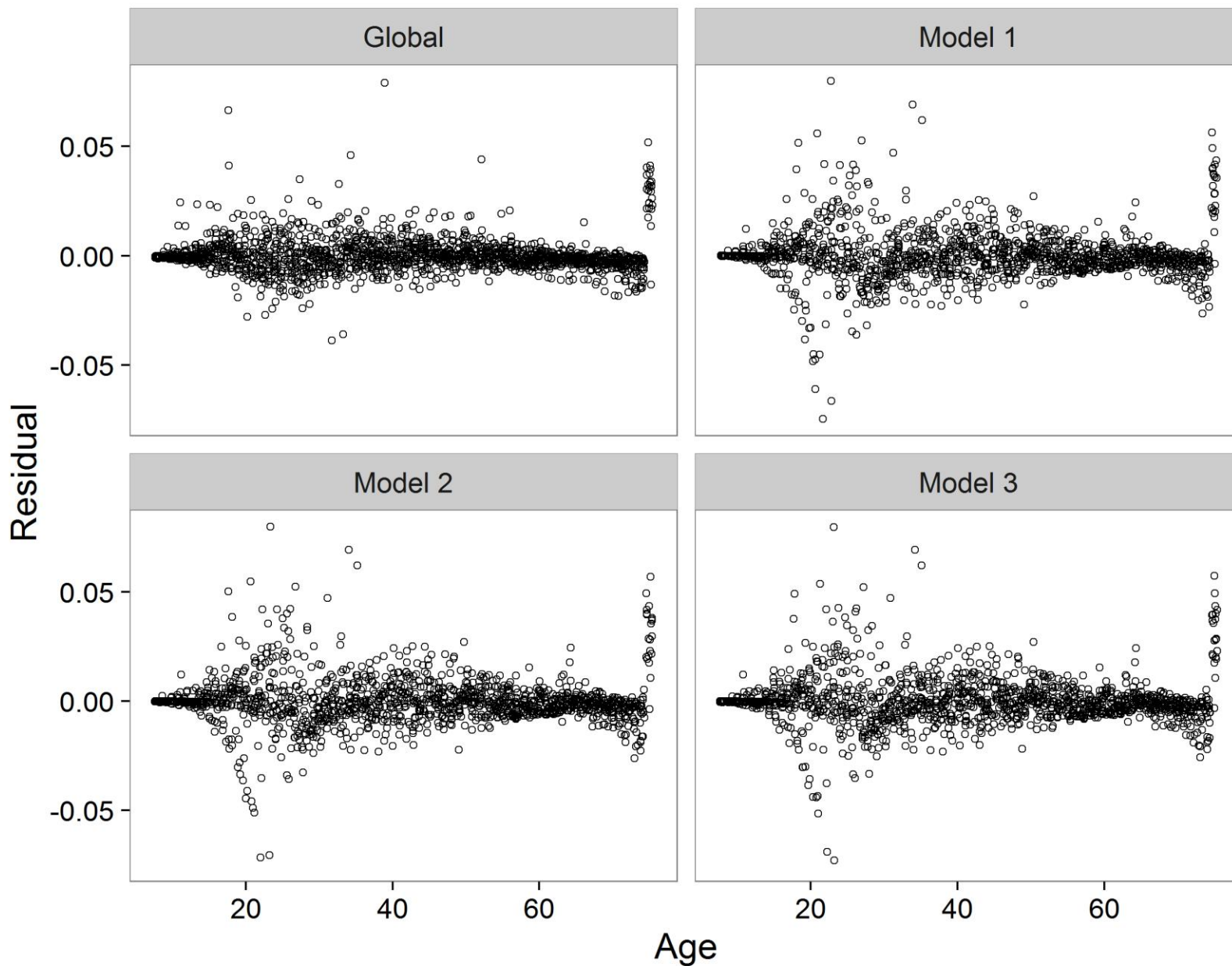


Results: Catch-age residuals - CSEO



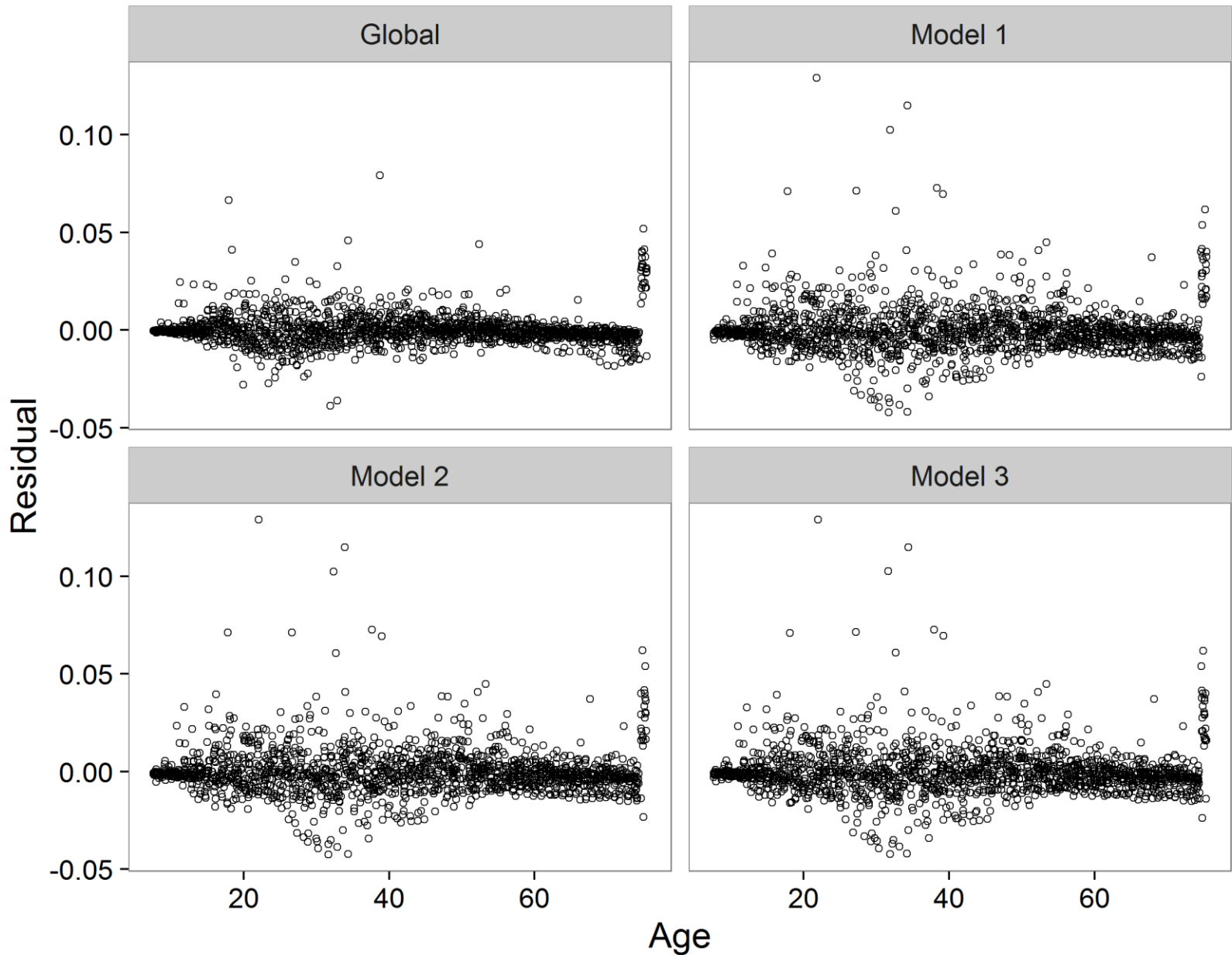


Results: Catch-age residuals - SSEO



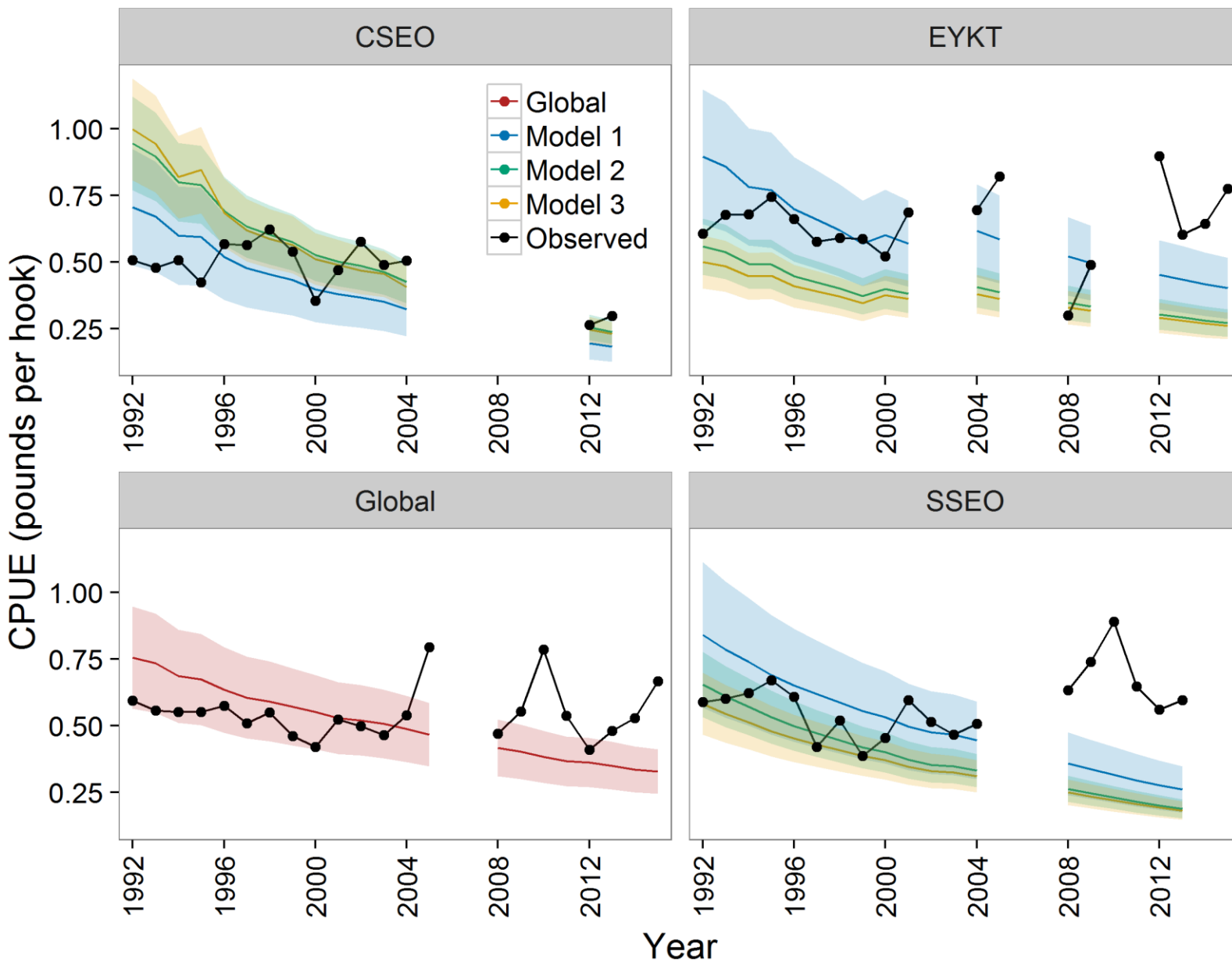


Results: Catch-age residuals - EYKT



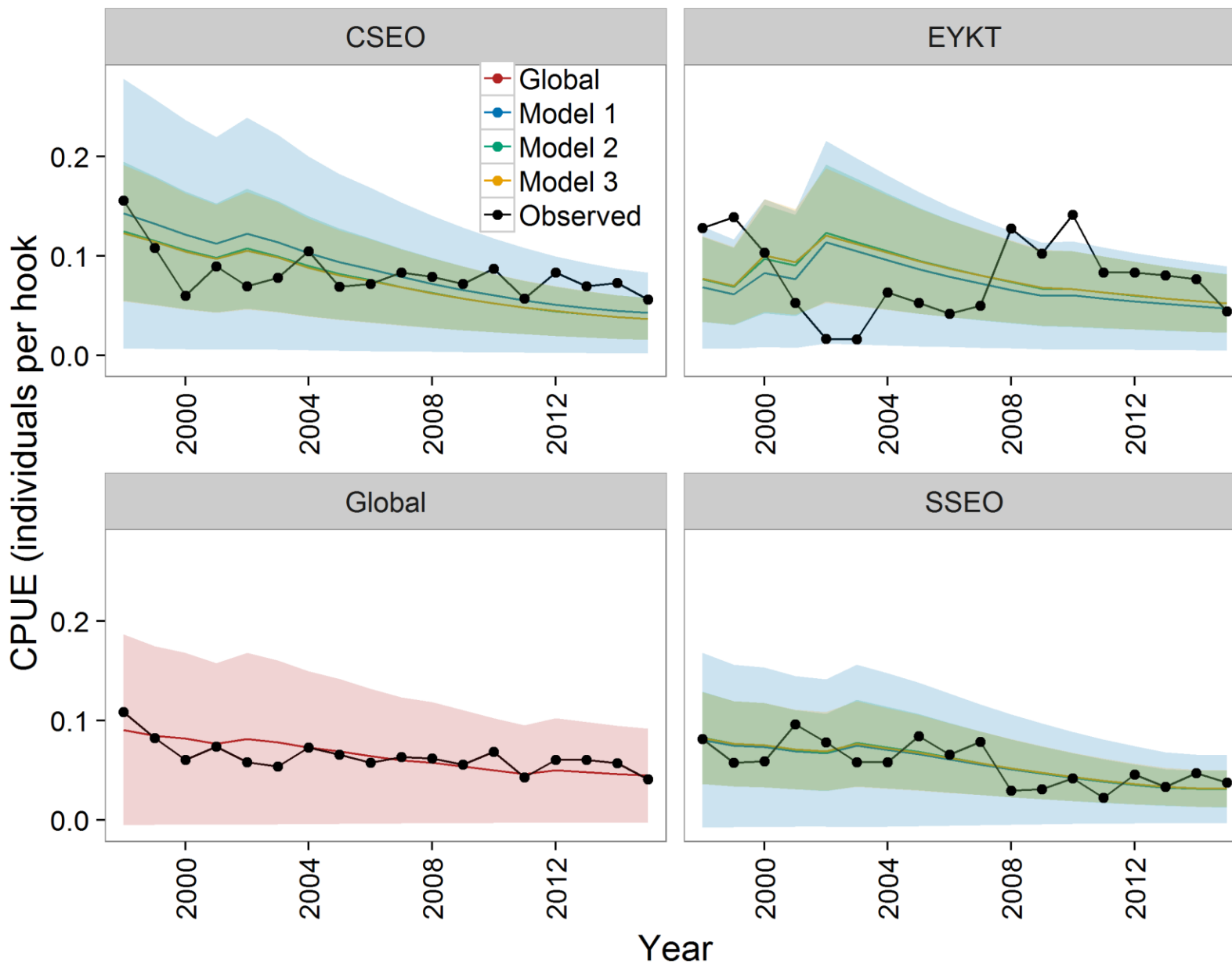


Results: Commercial fisheries CPUE





Results: IPHC survey CPUE





Model Results: Shared parameters



Natural mortality M

| Model 1 | Model 2 | Model 3 | Model 4 |
|---------------|---------|---------|---------|
| CSEO – 0.0831 | | | |
| SSEO – 0.0804 | 0.0850 | 0.0798 | 0.0791 |
| EYKT – 0.0915 | | | |

Commercial fishery CPUE catchability

| Model 1 | Model 2 | Model 3 | Model 4 |
|---------------|---------|---------|---------|
| CSEO – 0.0697 | | | |
| SSEO – 0.1233 | 0.0927 | 0.0858 | 0.0341 |
| EYKT – 0.1431 | | | |

Full-recruitment F_{45}

| Model 1 | Model 2 | Model 3 | Model 4 |
|---------------|---------|---------|---------|
| CSEO – 0.1203 | 0.1263 | 0.111 | |
| SSEO – 0.1562 | 0.1736 | 0.154 | 0.1331 |
| EYKT – 0.3271 | 0.2636 | 0.2225 | |

IPHC survey CPUE catchability

| Model 1 | Model 2 | Model 3 | Model 4 |
|---------------|---------|---------|---------|
| CSEO – 0.0464 | | | |
| SSEO – 0.0396 | 0.0405 | 0.0406 | 0.0117 |
| EYKT – 0.0363 | | | |



Model Results: Comparisons



Deviance Information Criterion

DIC values for all models from 2,000,000 MCMC iterations, saving every 100th

MODEL ONE

| | |
|--------------------------------|----------------|
| Expectation of log-likelihood | 11797 |
| Expectation of theta | 13421 |
| Number of estimated parameters | 439 |
| Effective number of parameters | -1624 |
| DIC | 10173.5 |

MODEL TWO

| | |
|--------------------------------|--------------|
| Expectation of log-likelihood | 11814 |
| Expectation of theta | 13482 |
| Number of estimated parameters | 433 |
| Effective number of parameters | -1667 |
| DIC | 10147 |

MODEL THREE

| | |
|--------------------------------|--------------|
| Expectation of log-likelihood | 11724 |
| Expectation of theta | 11787 |
| Number of estimated parameters | 441 |
| Effective number of parameters | -63 |
| DIC | 11661 |

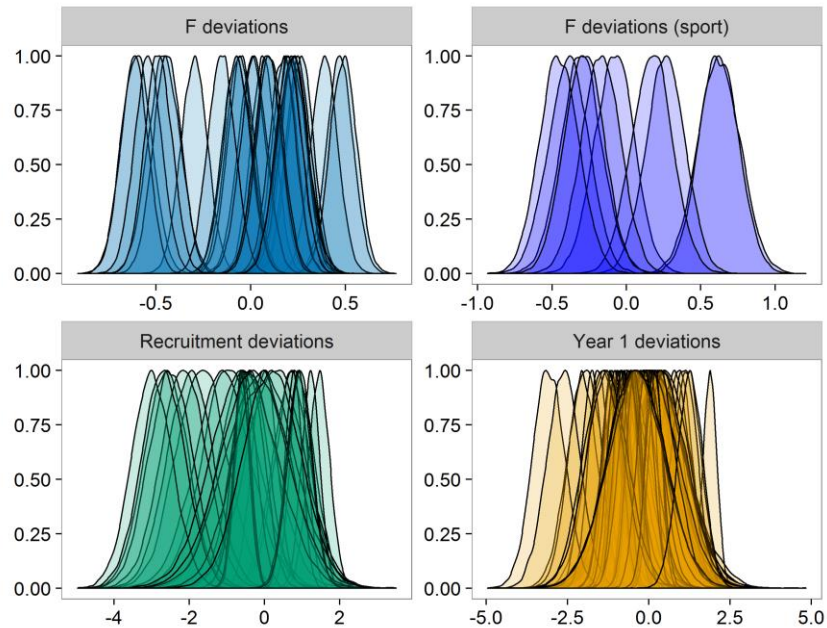
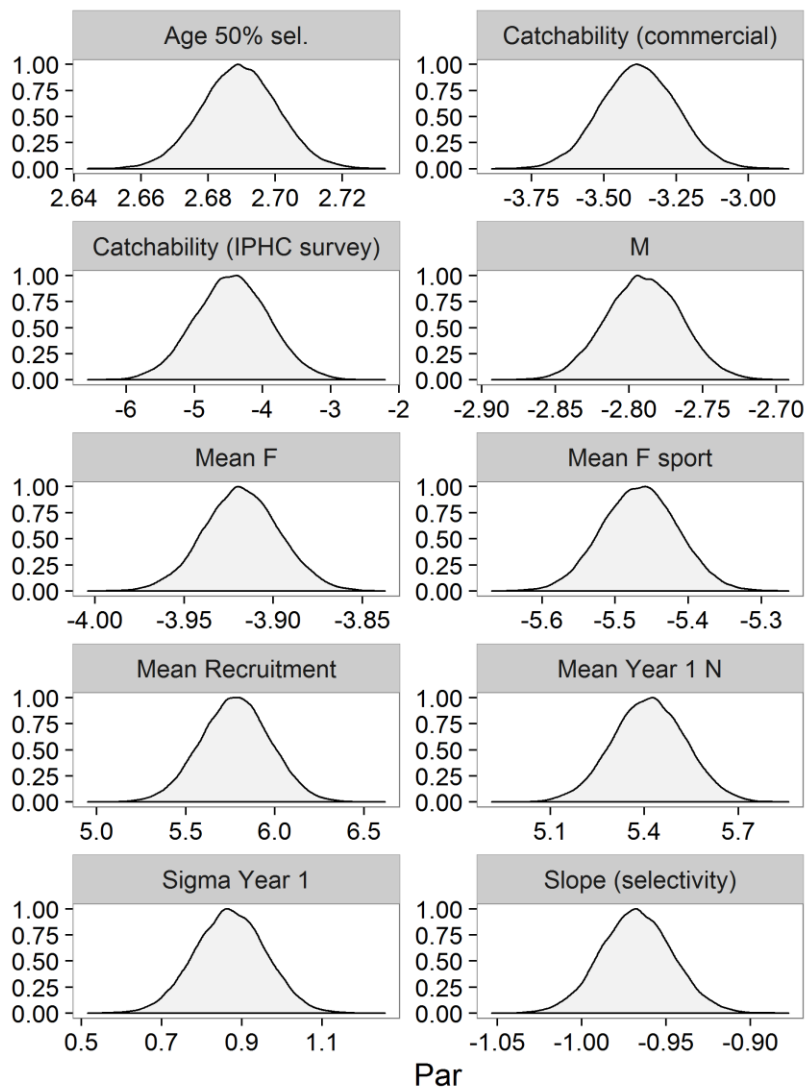
MODEL FOUR (Global)

| | |
|--------------------------------|-------------|
| Expectation of log-likelihood | 9743 |
| Expectation of theta | 10374 |
| Number of estimated parameters | 149 |
| Effective number of parameters | -632 |
| DIC | 9111 |



Global model evaluation

20,000 parametric bootstrap draws:
Full parameter space explored; no bound constraints

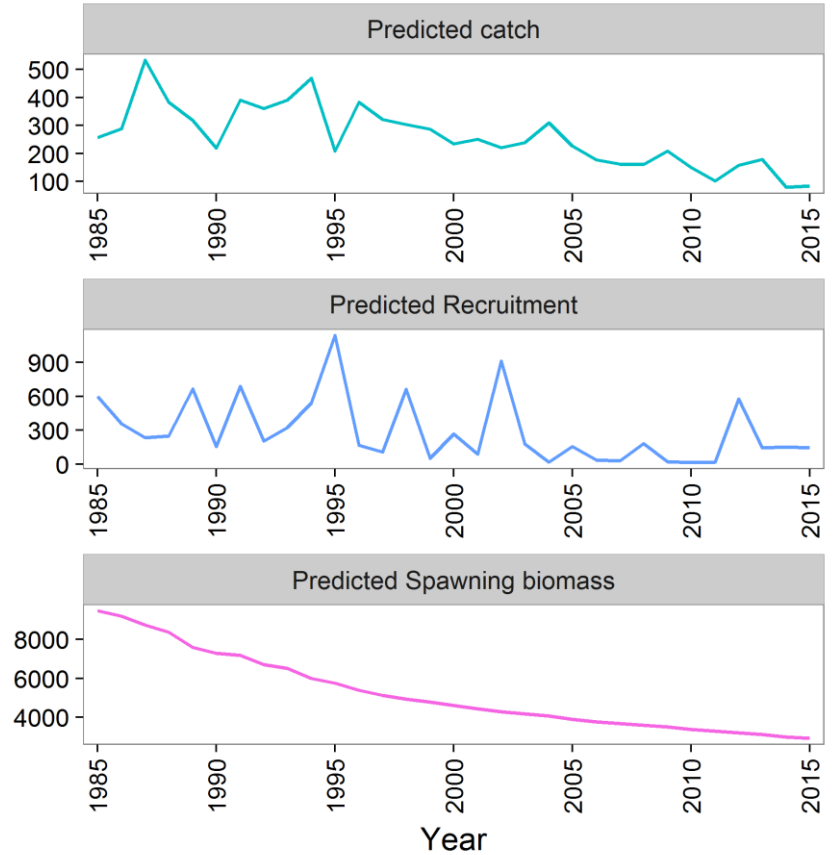
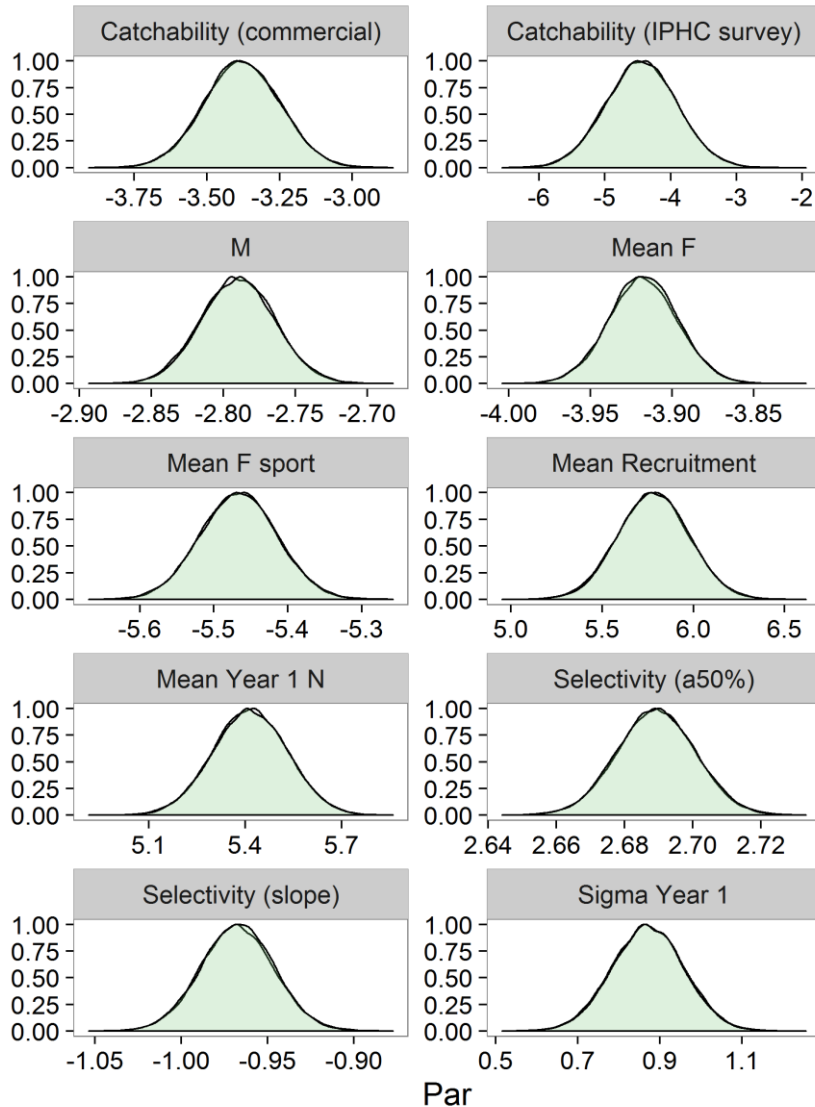




Global model evaluation



Self-test

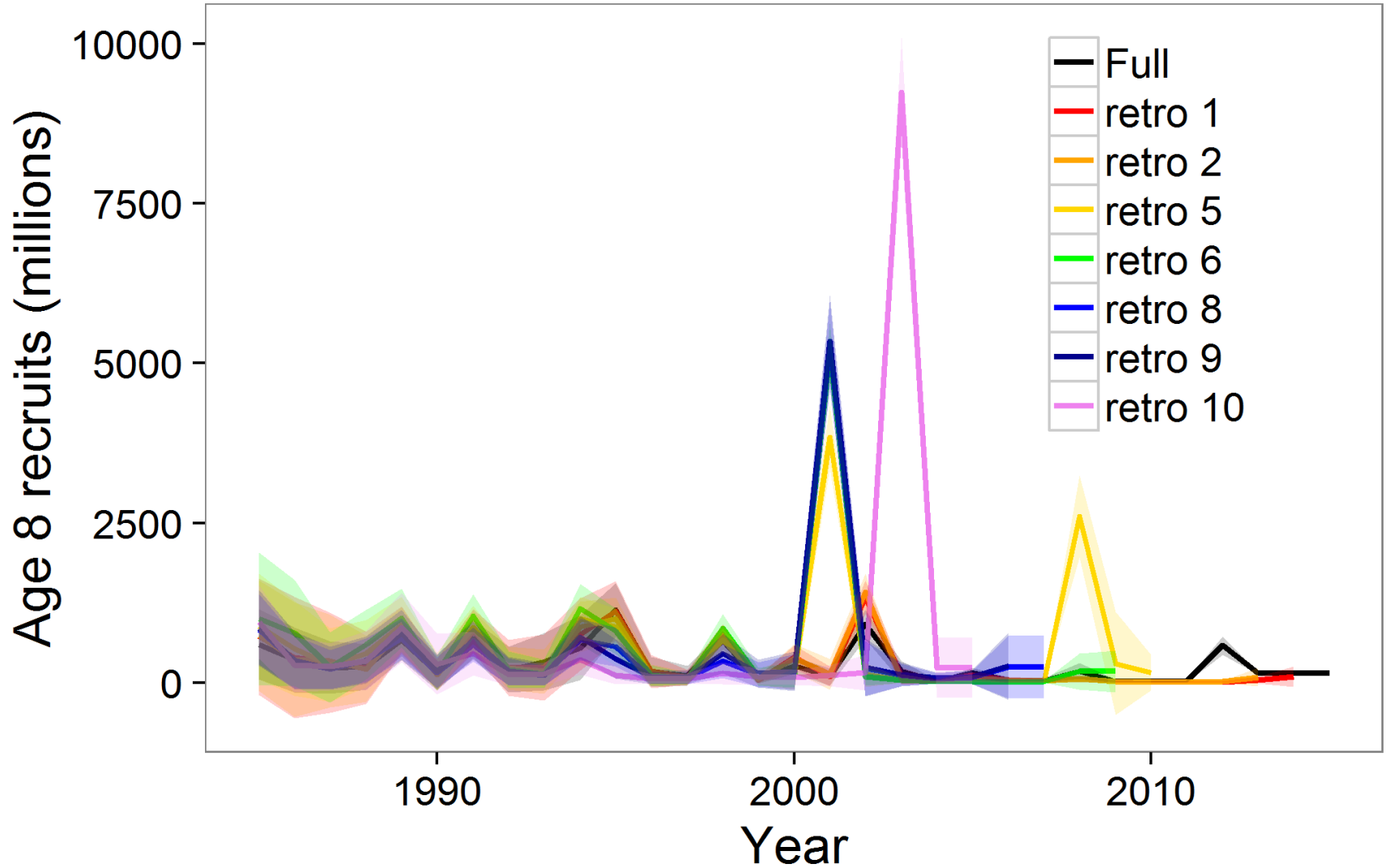




Global model evaluation



Retrospective analysis: age 8 recruitment

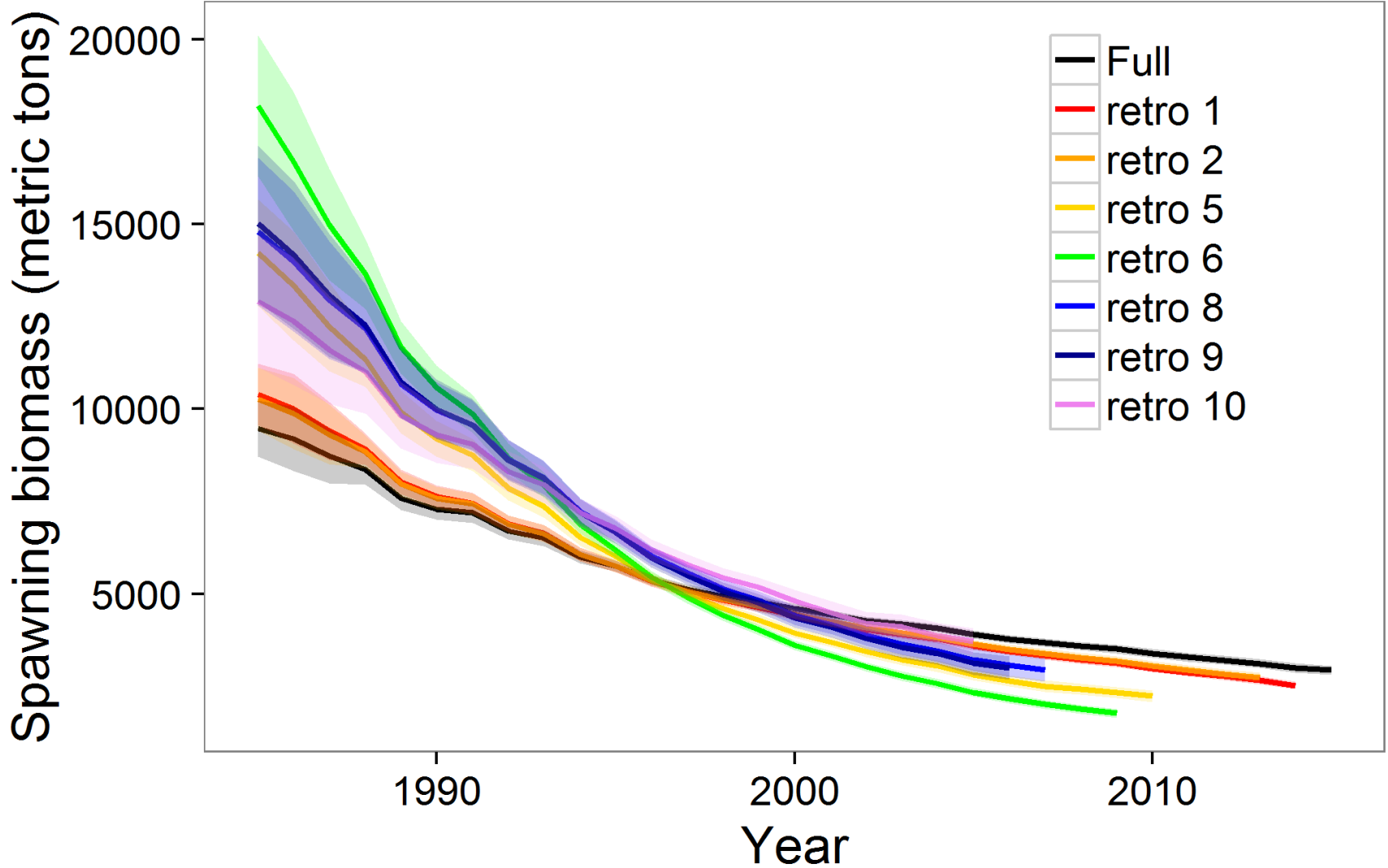




Global model evaluation



Retrospective analysis: spawning biomass





Global model evaluation



Estimating natural mortality

Confounded with extra variance term

M goes to zero

$$L = 0.5\ln(2\pi) + \ln(\sigma_{dens} + \sigma_{+}) + 0.5 \frac{(\ln(obs_den) - \ln(pred_den))^2}{2(\sigma_{dens}^2 + \sigma_{+}^2)}$$

$$\sigma_{dens}^2 = \log(1 + \sigma_{distance} / obs_den^2) \quad (\text{Burnham et al. 1987})$$

1. Evaluated root mean-squared error (RMSE) for density surveys inside model structure with no extra variance term;
2. Used the fixed RMSE as additional variance term

$$\sigma_{dens}^2 = \log(1 + (\sigma_{distance} + rmse) / obs_den^2)$$

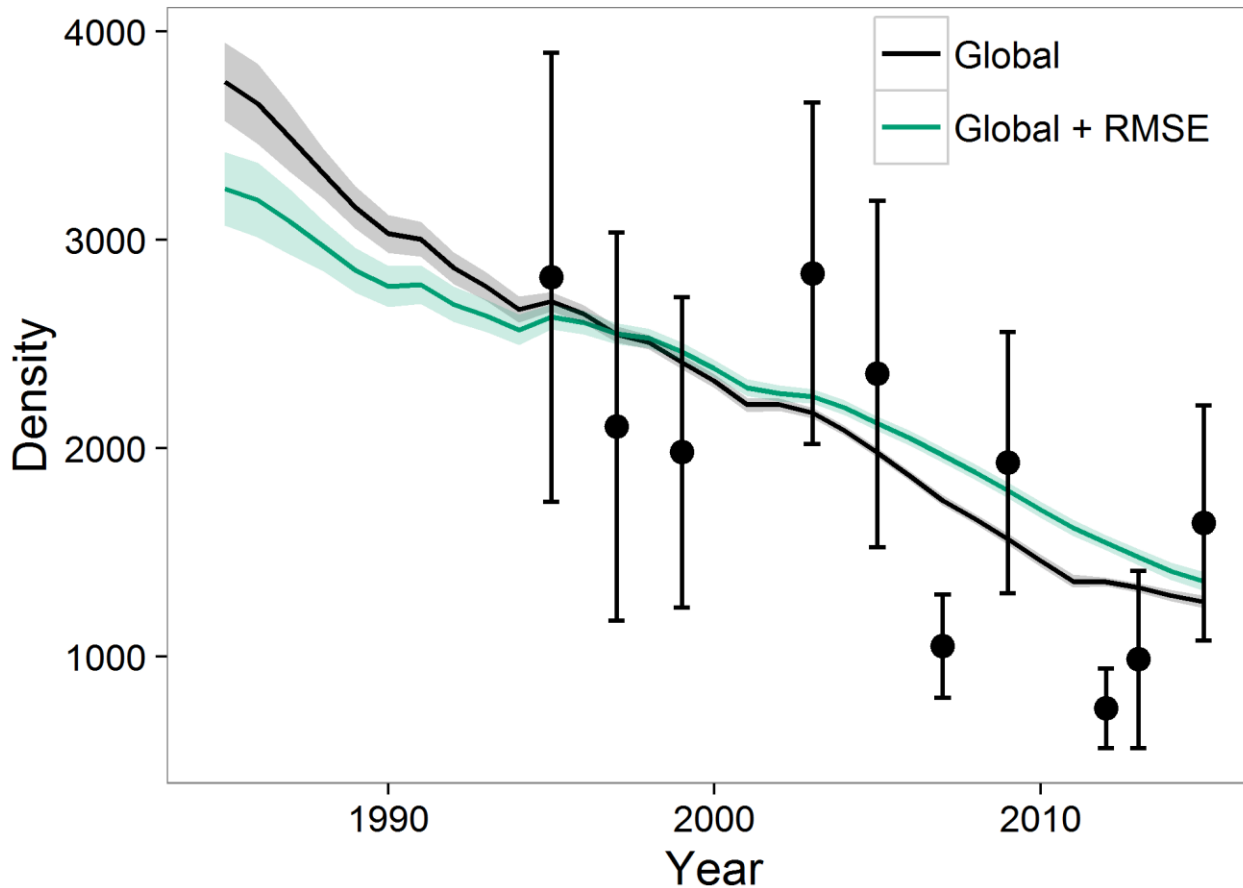


Global model evaluation



DIC values for all models from 2,000,000 MCMC iterations, saving every 100th

| RMSE Global model | | Global model | |
|--------------------------------|-------------|--------------------------------|-------------|
| Expectation of log-likelihood | 6644 | Expectation of log-likelihood | 9743 |
| Expectation of theta | 6928 | Expectation of theta | 10374 |
| Number of estimated parameters | 149 | Number of estimated parameters | 149 |
| Effective number of parameters | -283 | Effective number of parameters | -632 |
| DIC | 6361 | DIC | 9111 |



Natural mortality

Global: 0.791

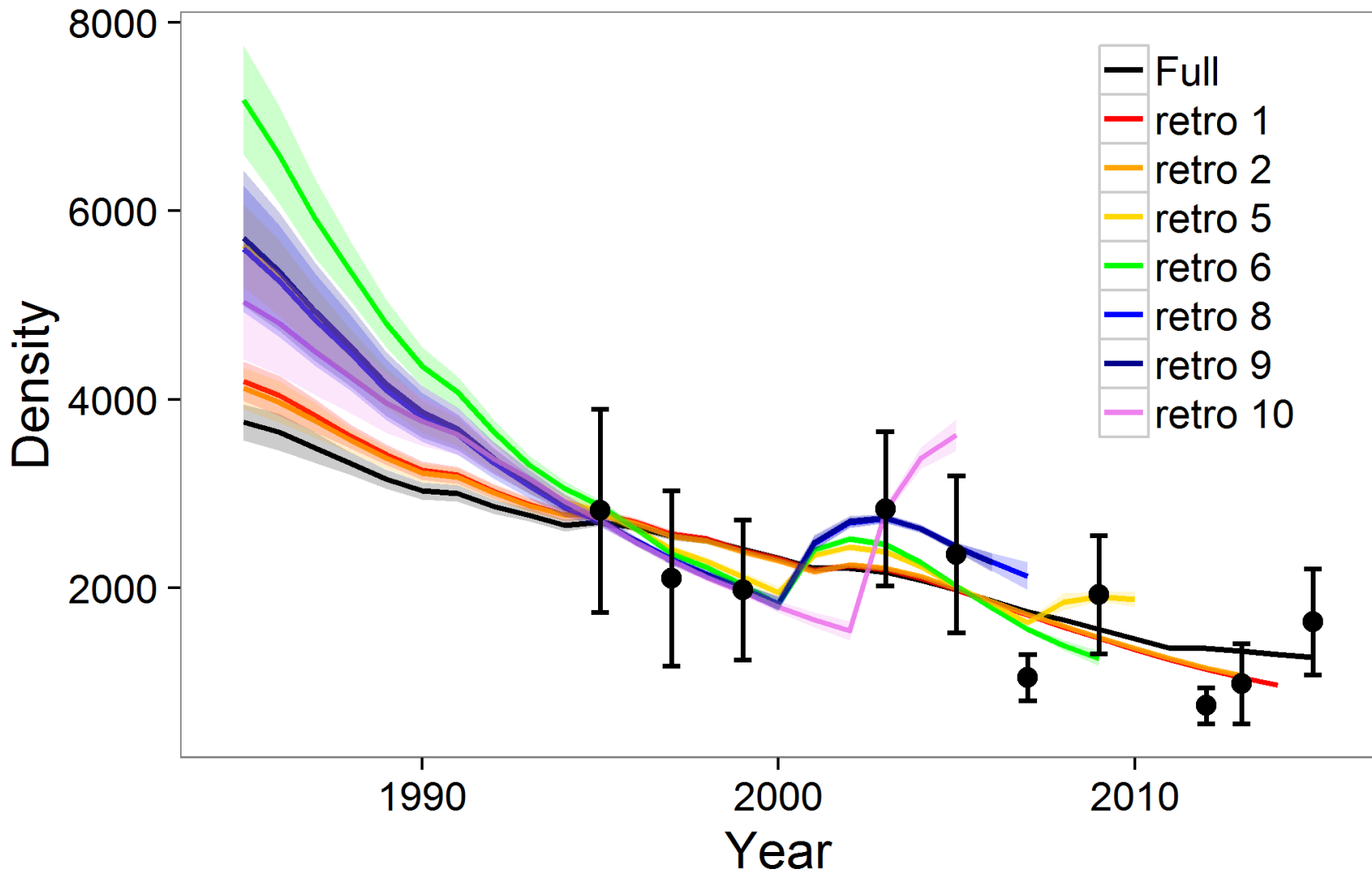
RMSE: 0.467



Model Results: Comparisons



Global model: density retrospective

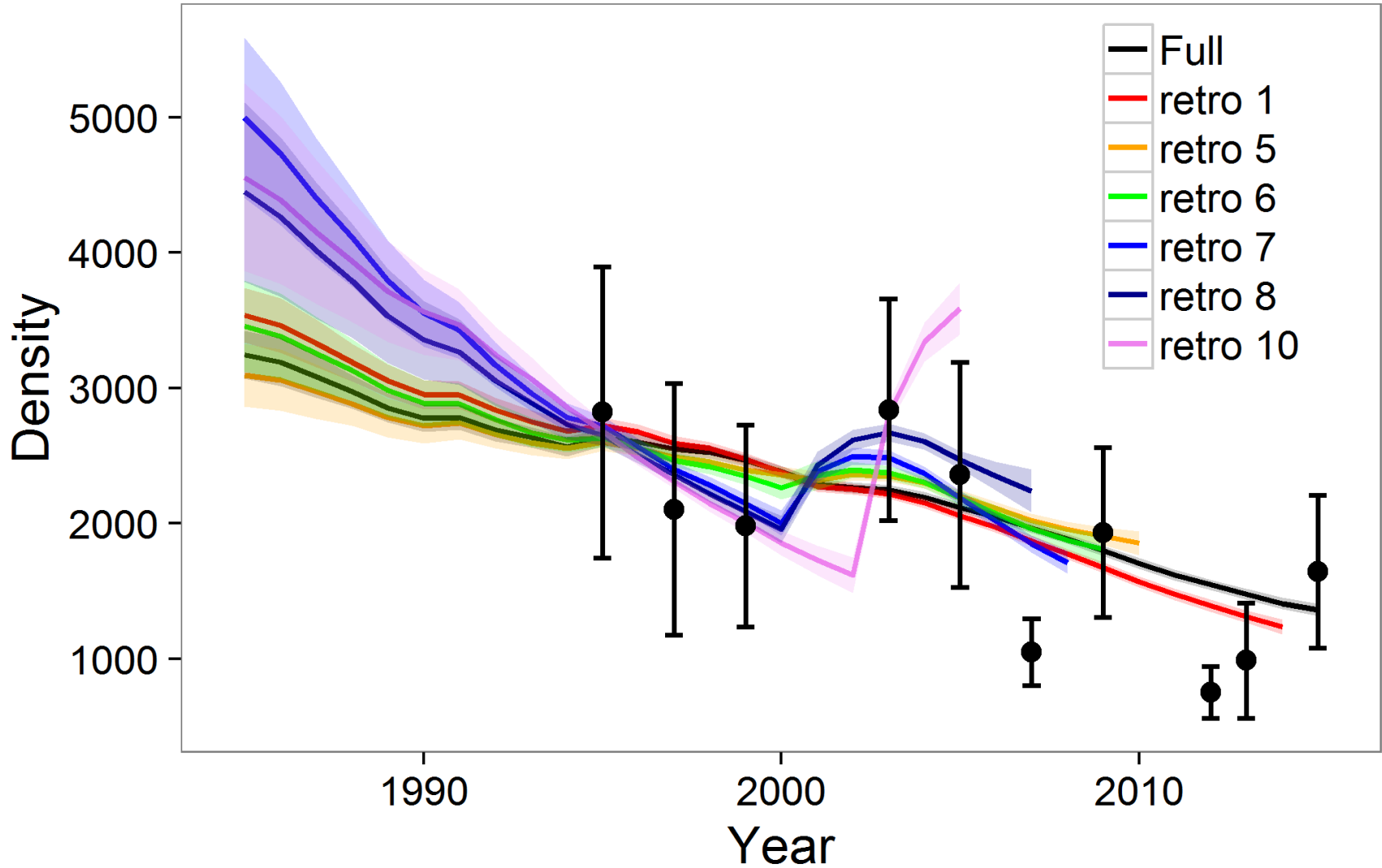




Model Results: Comparisons

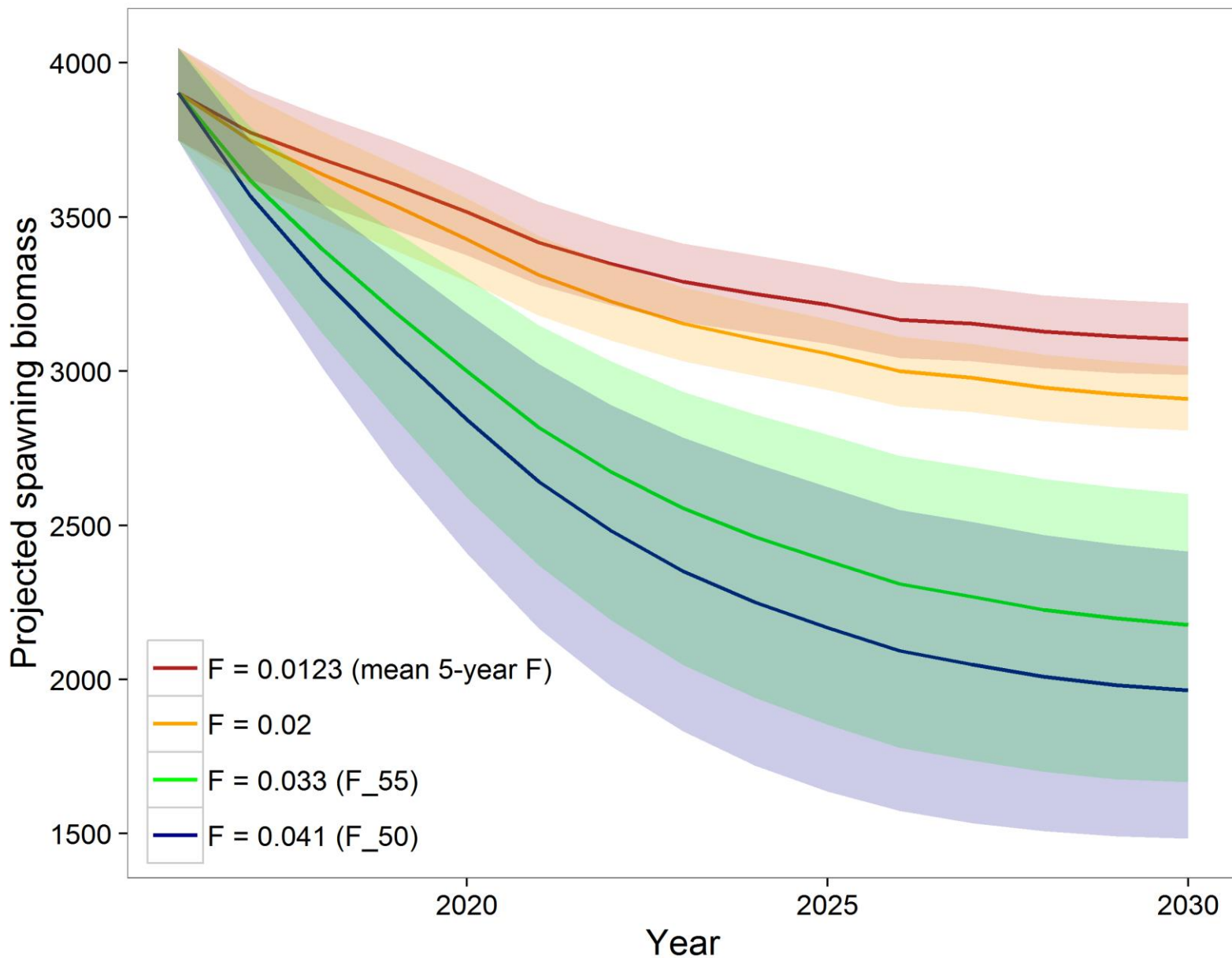


RMSE global model retrospective: density





Spawning biomass projections





Model Recommendation



| F level | Biomass (metric tons) | ABC | ABC (metric tons) |
|---|--------------------------------|-----------------|--------------------------|
| F_{45} (0.060) | L 90% CI (11,317) | Point-estimate | 554 |
| F_{50} (0.049) | L 90% CI (11,317) | Point-estimate | 454 |
| F_{55} (0.041) | L 90% CI (11,317) | Point-estimate | 382 |
| L 90% CI of F_{45} (0.032) | L 90% CI (11,317) | Point-estimate | 309 |
| L 90% CI of F_{50} (0.027) | L 90% CI (11,317) | Point-estimate | 253 |
| L 90% CI of F_{55} (0.022) | L 90% CI (11,317) | Point-estimate | 207 |
| F_{45} (0.060) | Point-estimate (11,697) | L 90% CI | 314 |
| F_{50} (0.049) | Point-estimate (11,697) | L 90% CI | 263 |
| F_{55} (0.041) | Point-estimate (11,697) | L 90% CI | 216 |
| CURRENT ABC ($F = 0.02$, assumes no selectivity) | | | 218 |

If the RMSE-modified global model is accepted for purposes of management advice, the author recommends reducing harvest levels to F_{55} and using the lower 90% confidence interval of the model-estimated ABC to set catch levels, which produces an ABC level for 2016 of **216** metric tons, which is essentially equivalent to the ABC of **218** metric tons under current management methods.



Priorities



1. Determine best approach for incorporating density uncertainty;
2. Re-analyze ADF&G survey data for global model;
3. Explore alternative methods for ROV survey – adaptive-cluster sampling for relative density zones across habitat

