



SEDAR 66 South Atlantic Tilefish Assessment Webinar III

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Introduction

Assessment Schedule



- 2020-11-16 to 2020-11-19: Data/Assessment Workshop
- 2020-12-01: Assessment Webinar I
- 2021-01-27: Assessment Webinar II
- 2021-02-18: Assessment Webinar III
- 2021-03-05: Assessment Report draft due
- 2021-03-19: Assessment Report comments due
- 2021-04-02: Assessment Report final due to SEDAR
- 2021-04-02: Assessment Report final due to Council

Introduction

Recap from Assessment Webinar II



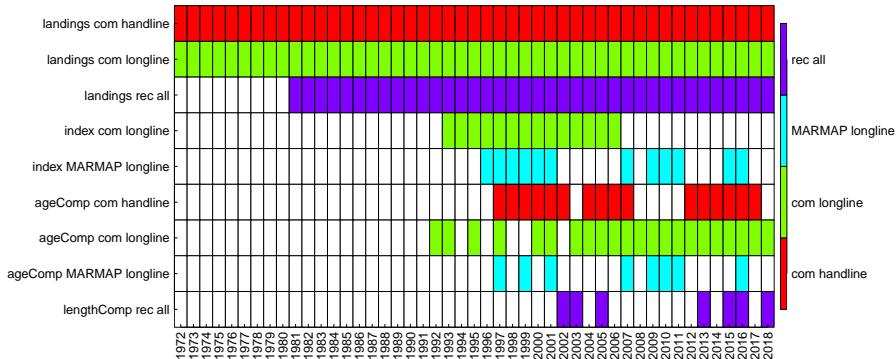
- We agreed on a base model
- We identified sensitivity runs including a new run involving alternative recruitment for the last several years of the assessment
- Meet for Webinar III to look at model diagnostics and projections



1. Random restarts
2. Likelihood profiles on steepness, natural mortality, R_0 , rec sigma, and F_{init}
3. Sensitivity analysis
4. Uncertainty analysis

Data

Summary of data being fit by the assessment



Stock assessment model

Base model configuration



- Model coded in Automatic Differentiation Model Builder (ADMB)
- Catch-at-age model used in most SEDAR assessments in the US South Atlantic, the Beaufort Assessment Model (BAM)
- Started with most updated version of BAM (SEDAR 25, 2017 Update)
- Timeline for SEDAR 66: 1972-2018

Stock assessment model

Base model configuration



- Age-structured life history
 - ▶ $(W_{\text{fishWhole}} = aL^b)$
 - ▶ $W_{\text{gonad}} = aW_{\text{fishWhole}}^b$
 - ▶ $W_{\text{fishWhole}} = aW_{\text{fishGutted}}$
 - ▶ $TL = VB(\text{age})$
 - ▶ $TL_{\text{female}} = VB(\text{age})$
 - ▶ Age-dependent female maturity
 - ▶ Age-dependent natural mortality
- Match landings time series
- Fit indices of abundance time series
- Fit age compositions
- Fit length compositions
- Estimate recruitment deviations
- Estimate fleet specific fishing mortality (average and time series of deviations)
- Estimate Beverton-Holt stock-recruit parameters
- Estimate selectivity functions
- Calculate biological reference points and stock status

Stock assessment model

Initialization



- Initial fishing mortality in 1972 (F_{init}) is fixed at 0.01
- Deviations from equilibrium age structure in 1972 are fixed at zero

Stock assessment model

Recruitment



Stock-recruit function

- Beverton-Holt stock-recruit function
- R_0 (unfished age-1 recruitment) is being estimated
- Steepness (h) is fixed at 0.84 (had a tendency to go to upper bound)
- Rec sigma (σ) is fixed at 0.6 (had a tendency to go to a lower bound)

Recruitment deviations

- Age composition data spans 1992-2018
- Most commercial ages are for age 6-15
- Recruitment deviations estimated from 1982-2011
- No additional constraints on recruitment

Stock assessment model

Selectivity



- Commercial handline landings
 - ▶ Logistic
 - ▶ Two time blocks: 1972 – 2008, 2009 – 2018
- Commercial longline landings
 - ▶ Logistic
 - ▶ Two time blocks: 1972 – 2008, 2009 – 2018
- Recreational landings
 - ▶ Logistic
 - ▶ One time block 1972 – 2018
- MARMAP horizontal longline survey index
 - ▶ Logistic
 - ▶ One time block 1996 – 2016

Stock assessment model

Comparison to previous assessment



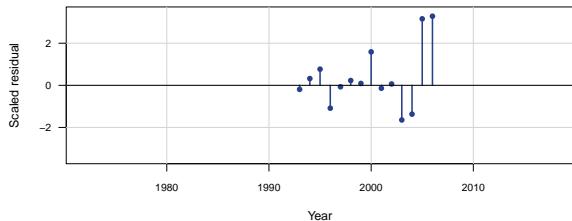
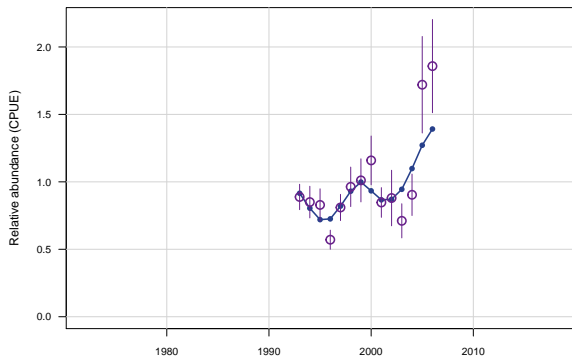
Model configuration

1. Beverton-Holt rec sigma (σ) is fixed
2. Selectivity of commercial handline and longline include two time blocks (1972-2007, 2008-2018)
3. Length and age compositions were fit using Dirichlet multinomial likelihoods, compared with robust multinomial likelihoods used in the SEDAR 25, 2016 Update
4. Modeling ages 1-20 in population compared with 1-25 in SEDAR 25, 2016 Update

Stock assessment model

Fit to indices

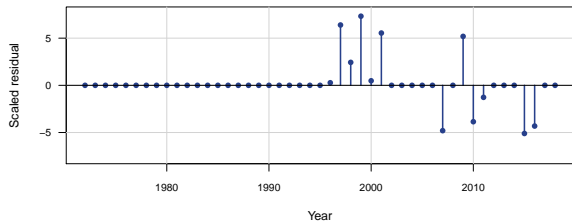
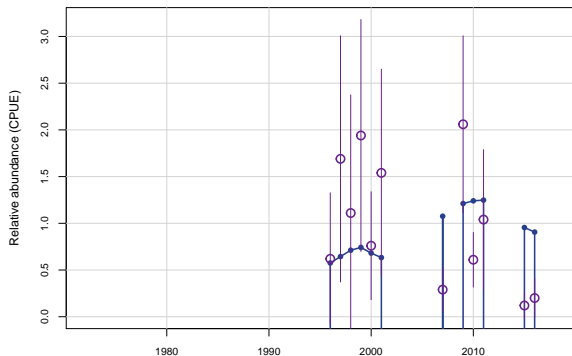
Commercial longline index



Stock assessment model

Fit to indices

MARMAP horizontal longline survey index



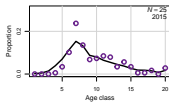
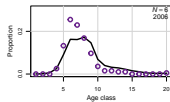
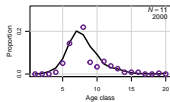
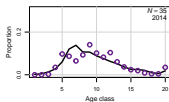
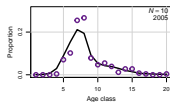
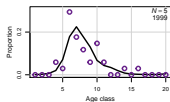
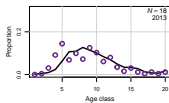
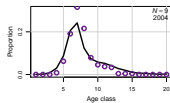
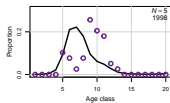
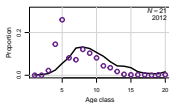
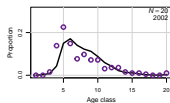
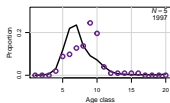
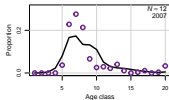
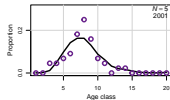
Stock assessment model

Fit to age and length compositions



Commercial handline ages: 1997 – 2015

↓ acomp.ch ↓



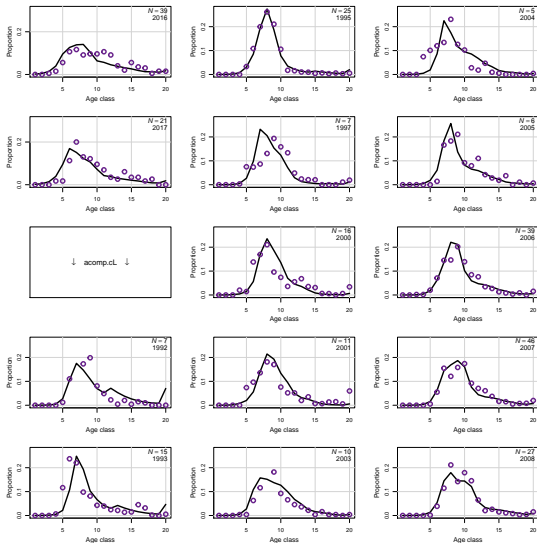


Stock assessment model

Fit to age and length compositions

Commercial handline ages: 2016 – 2017

Commercial longline ages: 1992 – 2008



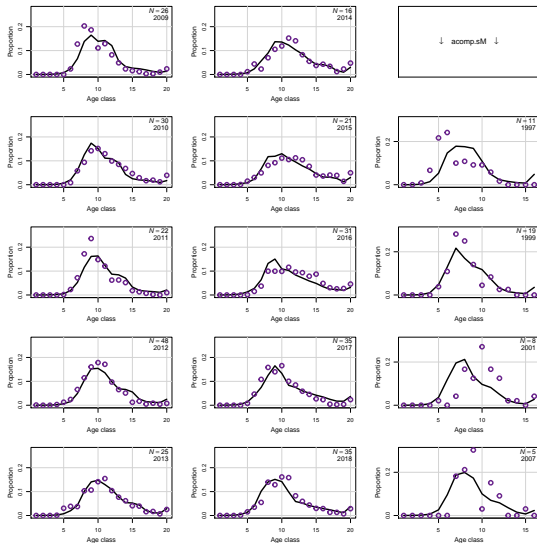


Stock assessment model

Fit to age and length compositions

Commercial longline ages: 2009 – 2018

MARMAP longline survey ages: 1997 – 2007



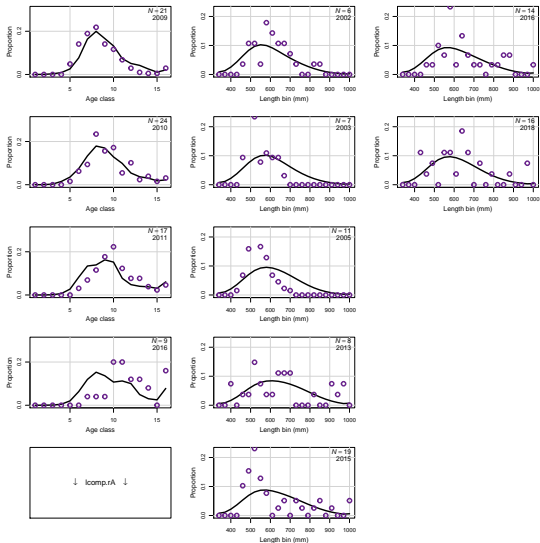


Stock assessment model

Fit to age and length compositions

MARMAP longline survey ages: 2009 – 2016

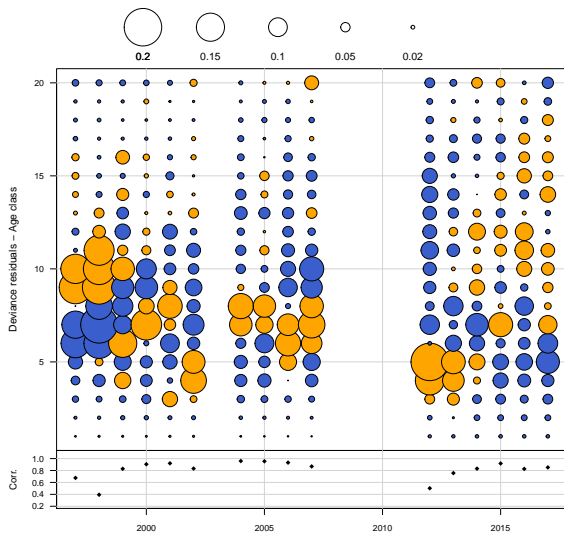
Recreational lengths: 2002 – 2018



Stock assessment model

Fit to age and length compositions: residuals

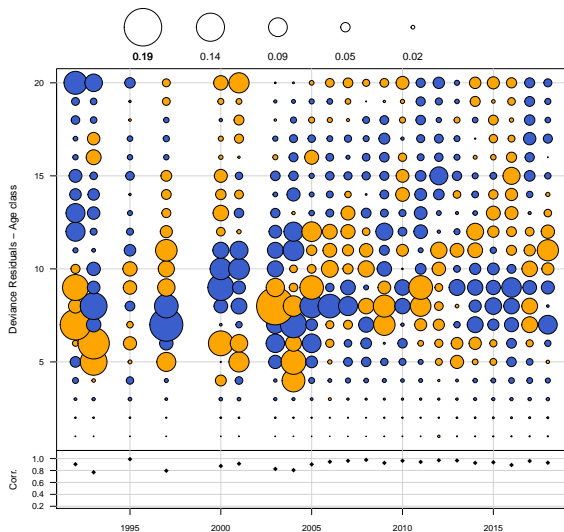
Commercial handline ages



Stock assessment model

Fit to age and length compositions: residuals

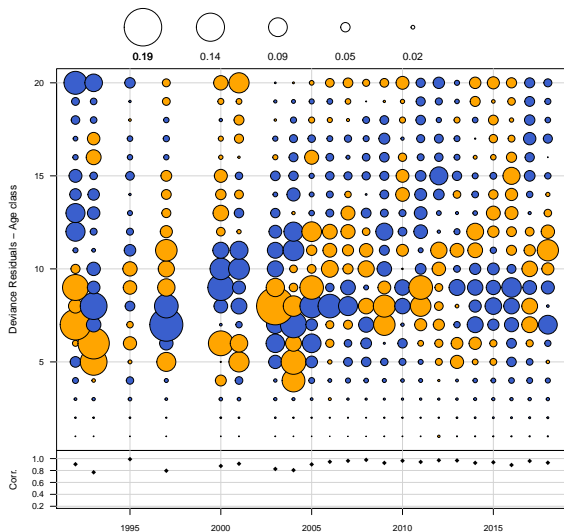
Commercial longline ages



Stock assessment model

Fit to age and length compositions: residuals

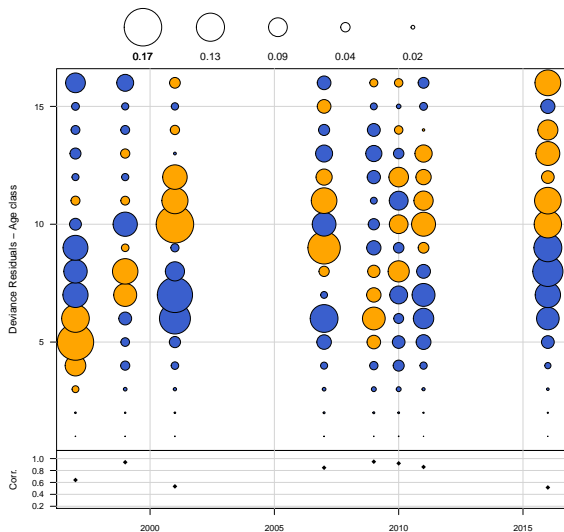
Commercial longline ages



Stock assessment model

Fit to age and length compositions: residuals

MARMAP longline survey ages

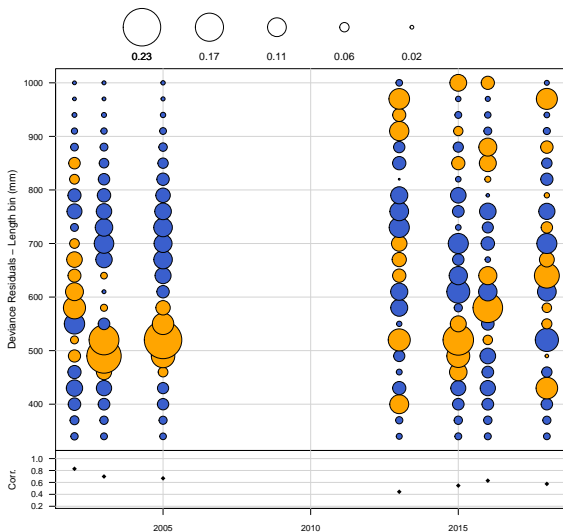




Stock assessment model

Fit to age and length compositions: residuals

Recreational lengths

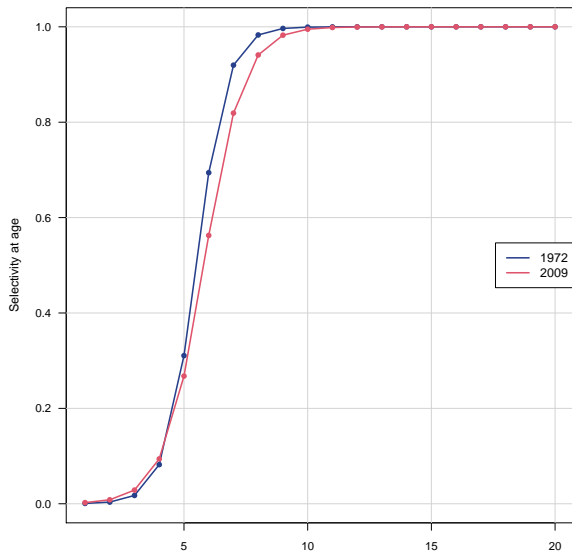


Stock assessment model

Estimated selectivity



Commercial handline landings

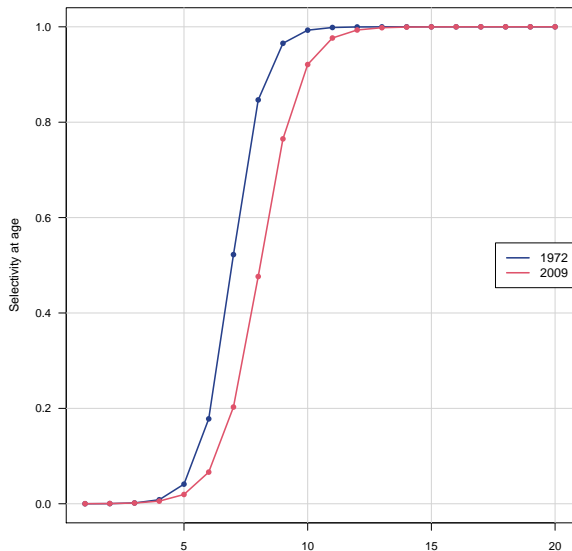


Stock assessment model

Estimated selectivity



Commercial longline landings

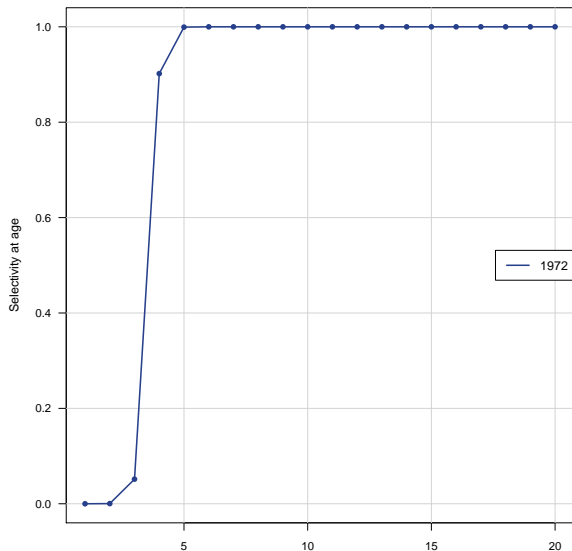


Stock assessment model

Estimated selectivity



Recreational landings

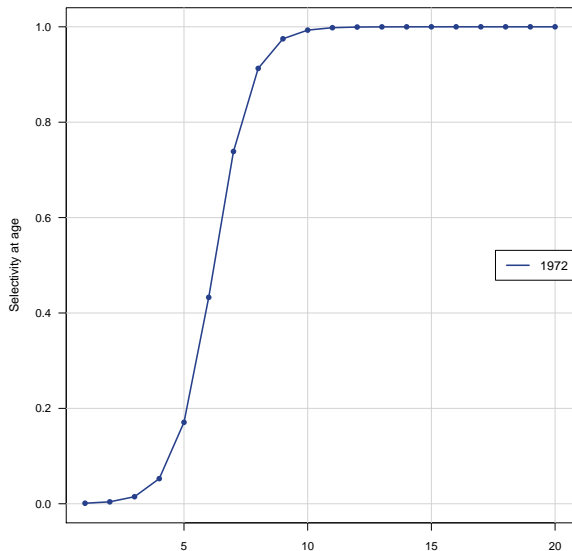


Stock assessment model

Estimated selectivity



MARMAP longline survey



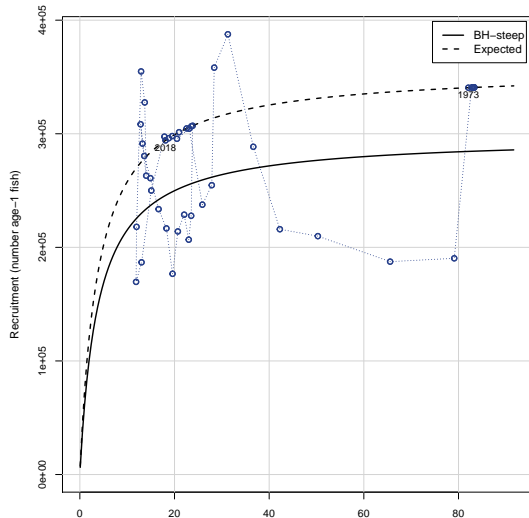
Stock assessment model

Spawner-recruit curve

$$h = 0.84$$

$$R_0 = 283,600$$

$$\sigma_R = 0.6$$

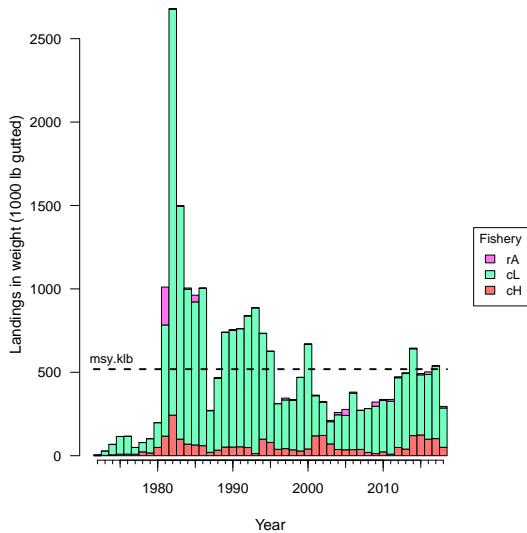


Stock assessment model

Landings



Landings (1000 lb)

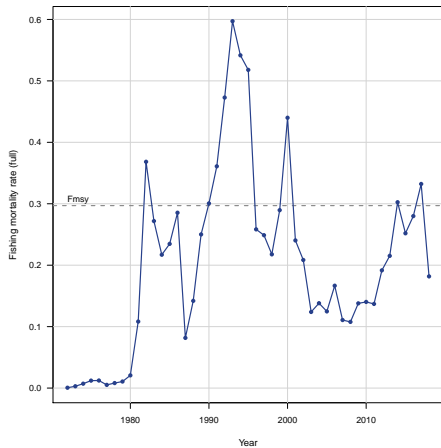


Stock assessment model

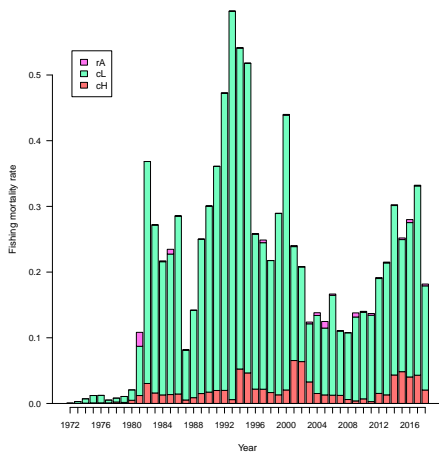
Benchmark time series



F -full



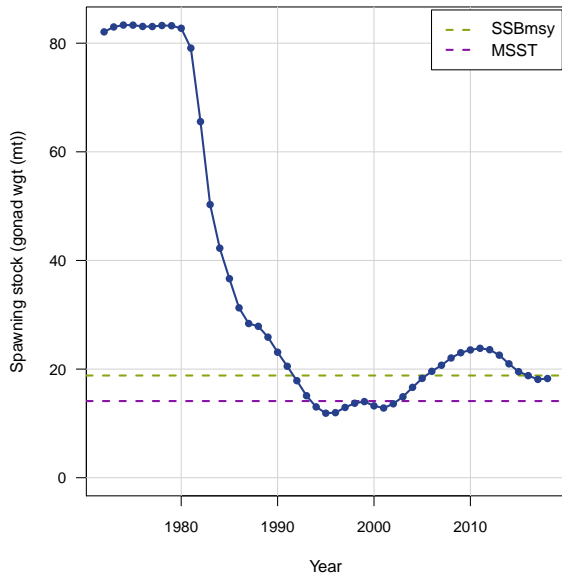
F by fleet



Stock assessment model

Benchmark time series

Spawning stock biomass (SSB)

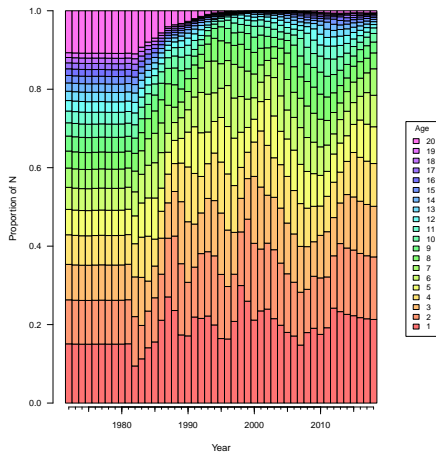


Stock assessment model

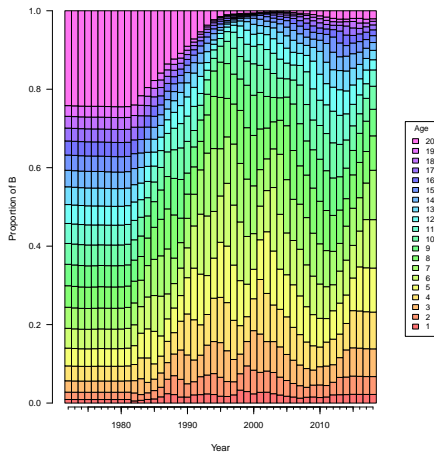
Benchmark time series



Numbers-at-age (proportions)

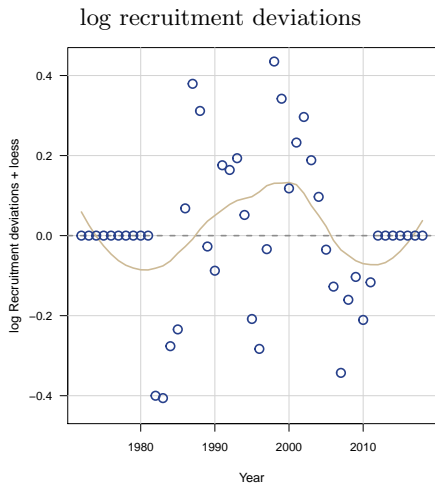
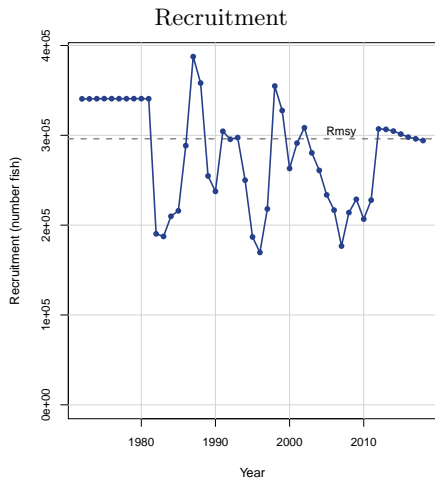


Biomass-at-age (proportions)



Stock assessment model

Benchmark time series



Stock assessment model

Management quantities



| Quantity | Units | Estimate | Median | SE |
|-------------------------|----------------|----------|--------|-------|
| F_{MSY} | y^{-1} | 0.297 | 0.255 | 0.183 |
| $85\%F_{MSY}$ | y^{-1} | 0.252 | 0.217 | 0.156 |
| $75\%F_{MSY}$ | y^{-1} | 0.223 | 0.191 | 0.138 |
| $65\%F_{MSY}$ | y^{-1} | 0.193 | 0.166 | 0.119 |
| $F_{20\%}$ | y^{-1} | 0.381 | 0.423 | 0.132 |
| $F_{30\%}$ | y^{-1} | 0.204 | 0.225 | 0.054 |
| $F_{40\%}$ | y^{-1} | 0.129 | 0.141 | 0.029 |
| B_{MSY} | metric tons | 2283.3 | 2515.7 | 463.2 |
| SSB_{MSY} | gonad wgt (mt) | 18.8 | 21.2 | 7.6 |
| MSST | gonad wgt (mt) | 14.1 | 15.9 | 5.7 |
| MSY | 1000 lb gutted | 518.8 | 507.2 | 79.1 |
| R_{MSY} | 1000 fish | 296.1 | 331.1 | 103.1 |
| $L_{85\%MSY}$ | 1000 lb gutted | 516.6 | 504.7 | 79.8 |
| $L_{75\%MSY}$ | 1000 lb gutted | 511.9 | 499.2 | 81.2 |
| $L_{65\%MSY}$ | 1000 lb gutted | 503.3 | 489.3 | 83.3 |
| $F_{2016-2018}/F_{MSY}$ | — | 0.864 | 1.049 | 1.755 |
| $SSB_{2018}/MSST$ | — | 1.294 | 1.116 | 0.63 |
| SSB_{2018}/SSB_{MSY} | — | 0.97 | 0.837 | 0.473 |



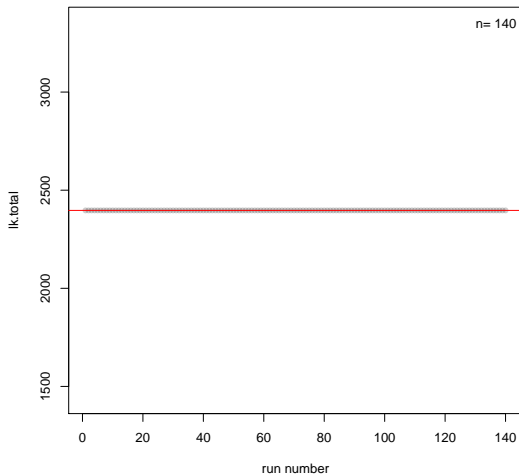
- This procedure is repeated many times and the results are aggregated to see if the model is sensitive to the starting parameter values, over a reasonable range
- The model should find the same solution despite randomly changing the starting values
- Starting values for estimated point parameters are randomly sampled from uniform distributions $\pm 50\%$ of the starting values from the base model
- Parameter vectors like the F and recruitment deviations, and dirichlet multinomial parameters, which usually start at zero, were not randomized
- The model was rerun with these new starting values

Random restarts

Results



- Likelihood values from random restart runs



Likelihood profiles

Methods



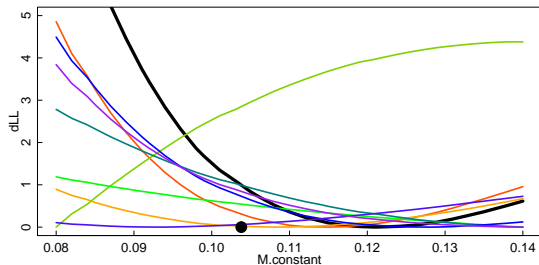
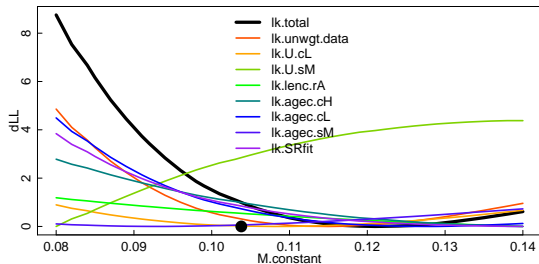
- Likelihood profiles on steepness, natural mortality, R_0 , rec sigma, and F_{init}
- This procedure is repeated multiple times for each parameter of interest over a range of fixed values above and below the base model value
- The model was rerun with these fixed values
- Plotting likelihood versus the fixed parameter values allows us to investigate sources of information informing each parameter

Likelihood profiles

Results



• M

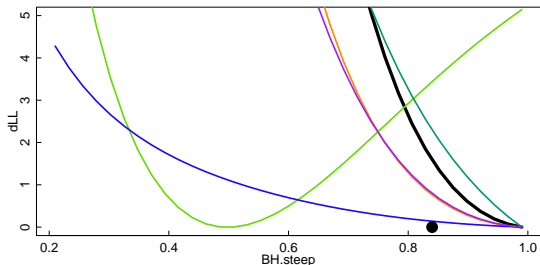
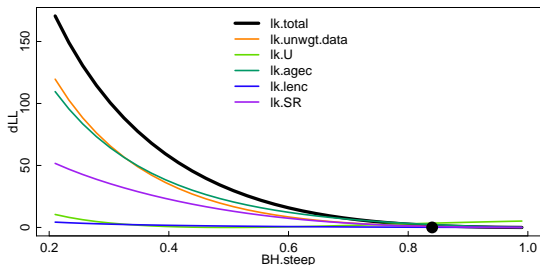


Likelihood profiles

Results



- steepness

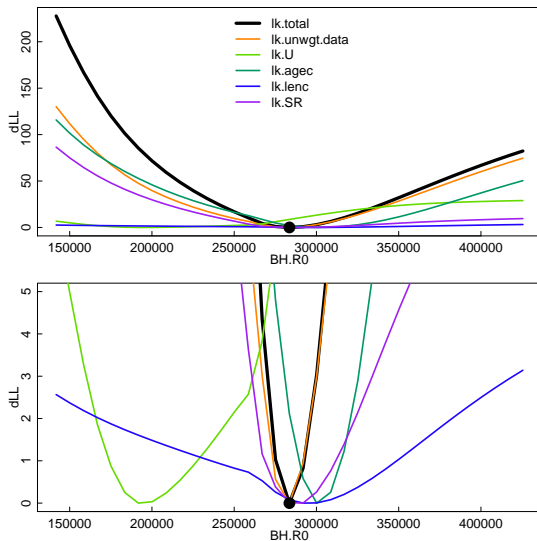


Likelihood profiles

Results



• R_0

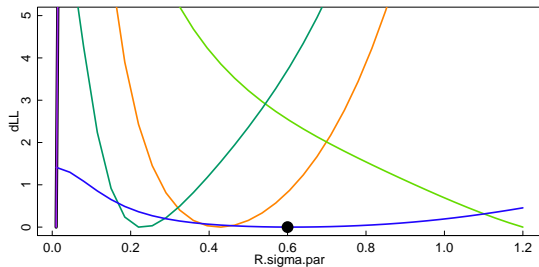
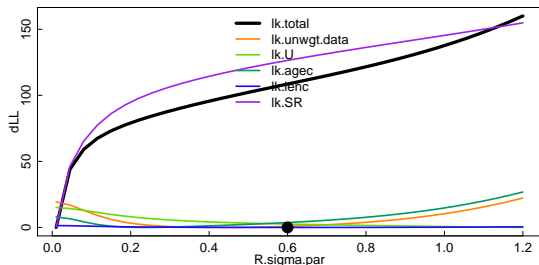


Likelihood profiles

Results



• R_0

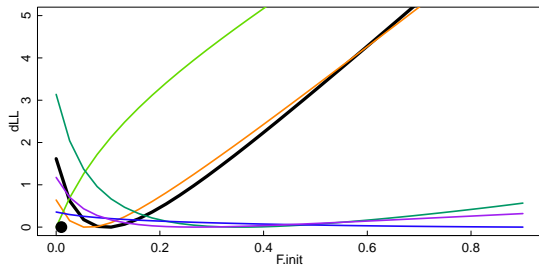
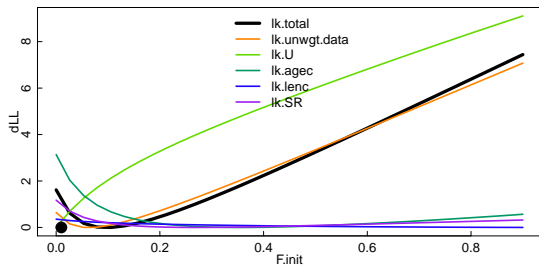


Likelihood profiles

Results



• R_0



Monte Carlo Bootstrap (MCB) uncertainty ensemble

Methods



- The Monte-Carlo Bootstrap (MCB) analysis is a process of randomizing data inputs and fixed parameters that go into the assessment model
- 4200 sets of randomized inputs were drawn and the assessment model is run each time
- Outputs from each run are summarized.

Monte Carlo Bootstrap (MCB) uncertainty ensemble

Methods



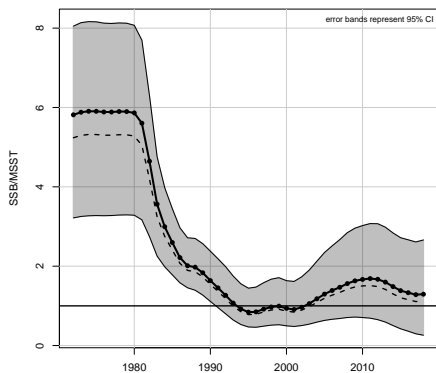
- Landings resampled from log-normal distributions defined by values provided by data providers
 - ▶ Recreational (MRIP) landings CVs based on a new method developed by NMFS staff in Miami
- Indices resampled from log-normal distributions defined by values provided by data providers
- Length and age composition data resampled, with replacement
- Natural mortality estimates were sampled from uniform distribution (0.08 – 0.14) and then used to rescale age-varying M

Monte Carlo Bootstrap (MCB) uncertainty ensemble

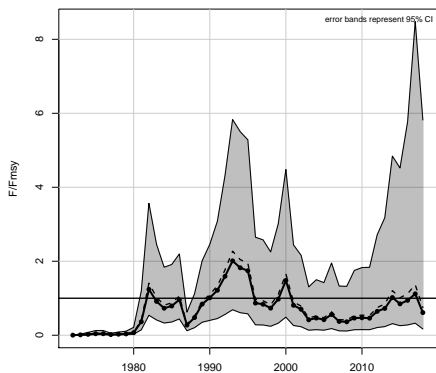
Results



$SSB/MSST$

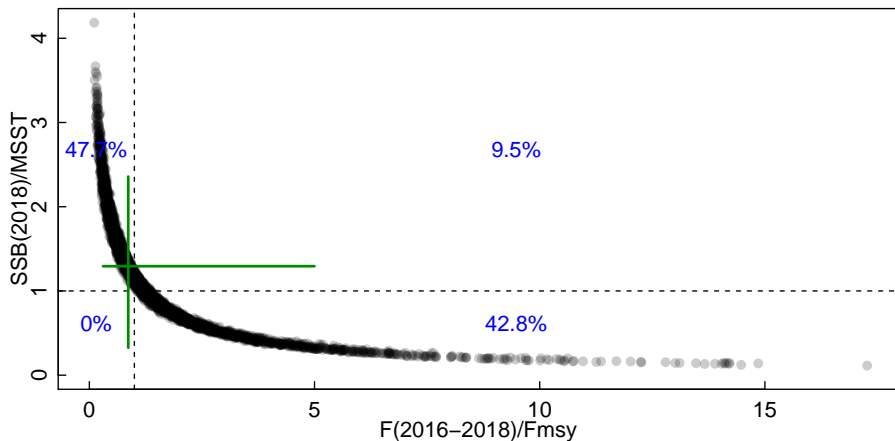


F/F_{msy}





Uncertainty in stock and fishery status

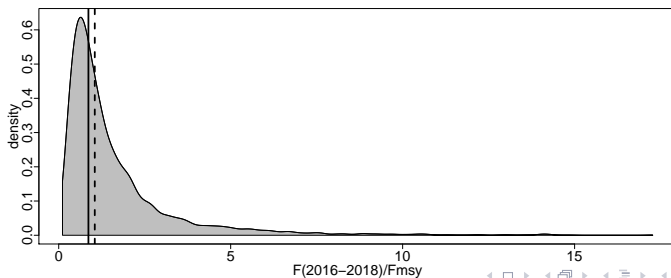
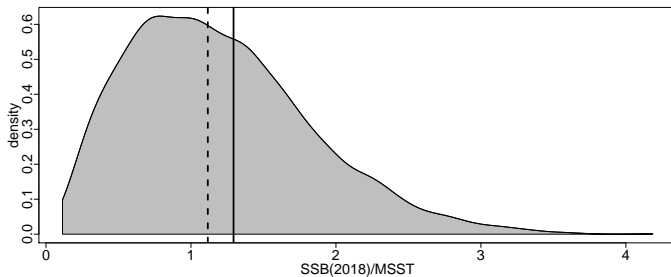


Monte Carlo Bootstrap (MCB) uncertainty ensemble

Results



Uncertainty in stock and fishery status



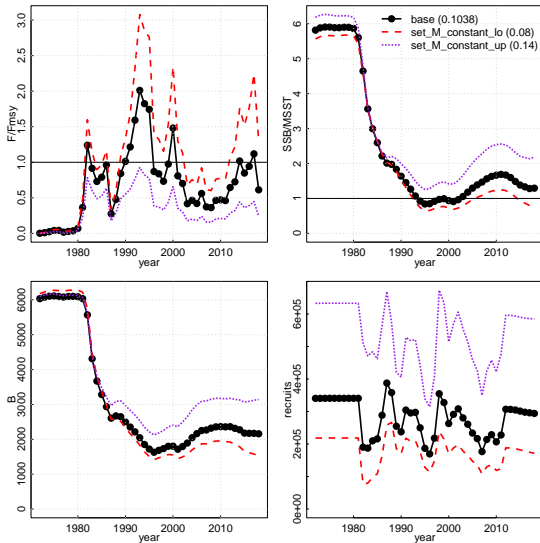


- S1-S2: Low/high values of natural mortality ($M = 0.08, 0.14$)
- S3-S4: Low/high values of steepness ($h = 0.74, 0.94$)
- S5-S6: Higher values of initial F ($F_{init} = 0.053, 0.106$). Values associated with minimum of likelihood profile (lkmin) and half that value (0.5lkmin).
- S7-S8: Down/upweight MARMAP longline index: $1/10\times, 10\times$
- S9: Use alternate recruitment estimates for years at the end of the assessment (2012-2018) where recruitment deviations were not estimated, based on geometric mean recruitment deviation from the last six years where recruitment deviations were estimated (2006-2011)

Sensitivity analysis

Results

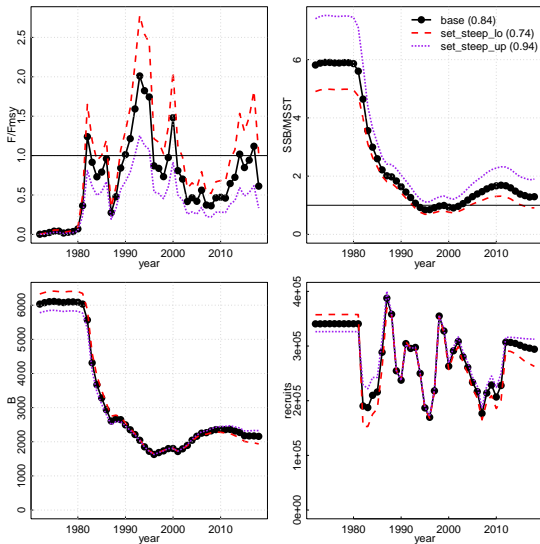
S1-S2: Low/high values of natural mortality ($M = 0.08, 0.14$)



Sensitivity analysis

Results

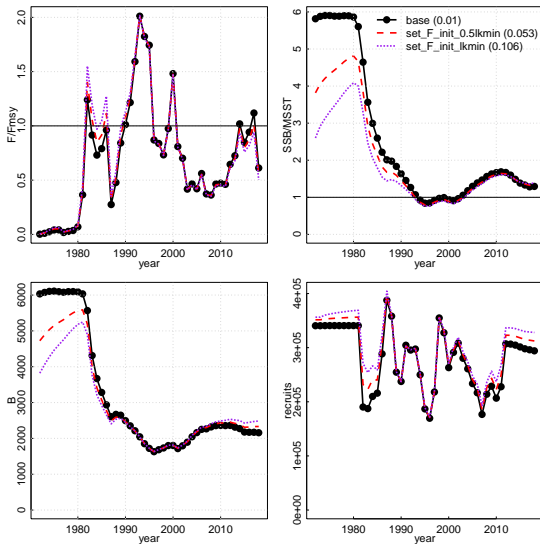
S3-S4: Low/high values of steepness ($h = 0.74, 0.94$)



Sensitivity analysis

Results

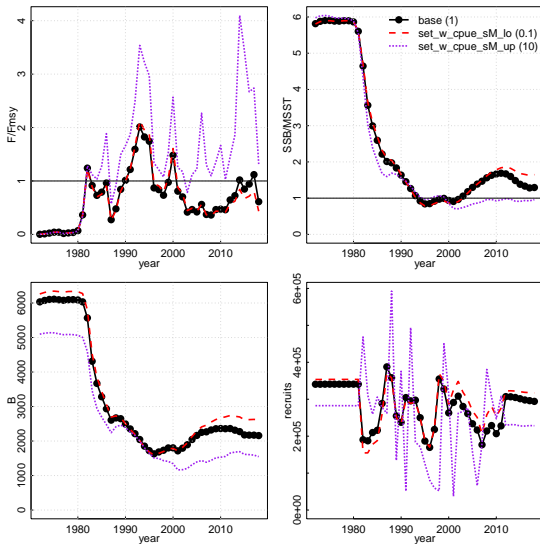
S5-S6: Higher values of initial F ($F_{init} = 0.053, 0.106$).



Sensitivity analysis

Results

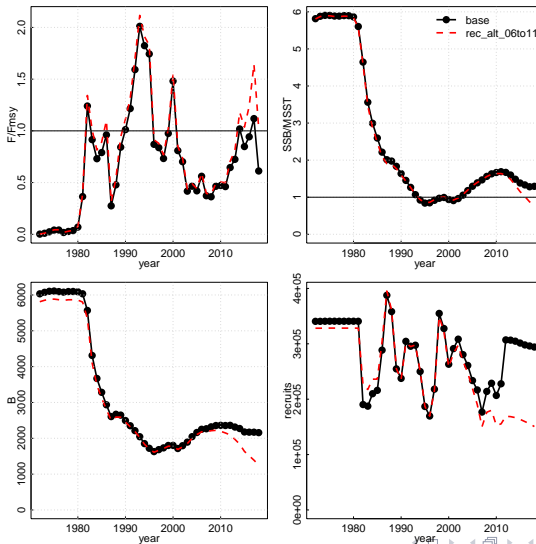
S7-S8: Down/upweight MARMAP longline index: $1/10\times$, $10\times$



Sensitivity analysis

Results

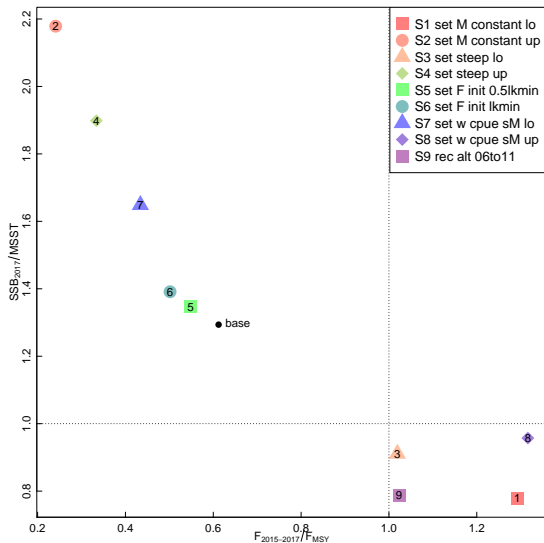
S9: Use alternate recruitment estimates for years at the end of the assessment (2012-2018)



Sensitivity analysis

Results

Stock and fishery status of sensitivity runs



Retrospective analysis

Methods

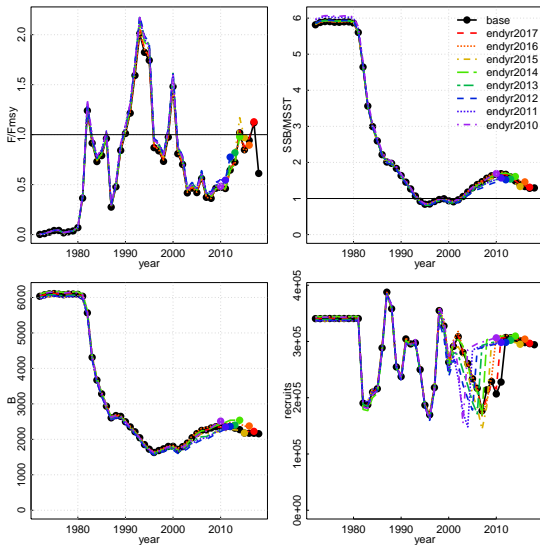


- Methods similar to sensitivity analysis
- Data in assessment model were truncated to new terminal years of 2010 – 2017
- The base model was rerun with truncated data
- Results of retrospective runs were plotted together to look for patterns in terminal year values

Retrospective analysis

Results

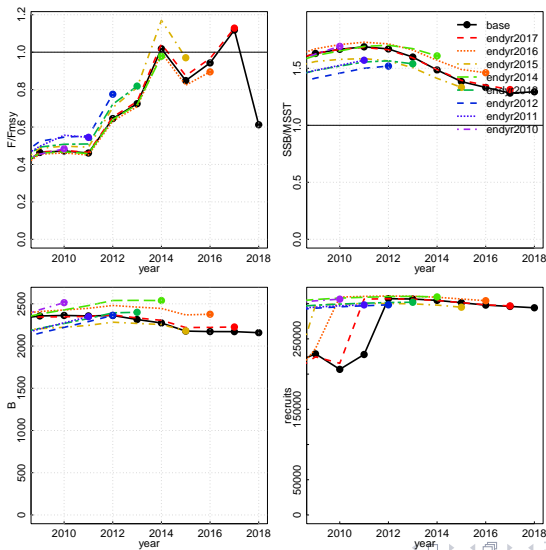
Terminal years of retrospective runs: 2010 – 2017



Retrospective analysis

Results

Terminal years of retrospective runs: 2010 – 2017
(zoomed in on recent years)





- Projections were constructed as specified in the ToRs
- Projections were made to 2027, with projected fishing level changes beginning in 2022.
- Fishing mortality for 2019-2021 was set at $F_{\text{current}} = 0.2566$ (geometric mean F from 2016-2018)
- Projections at fixed F from 2022 – 2027
- To determine OFL:
 - ▶ F based on $P^* = 50\%$
 - ▶ $F = F_{\text{MSY}}$
- To determine ABC:
 - ▶ F based on $P^* = 30\%$
 - ▶ $F = 75\%F_{\text{MSY}}$



- Scenario 1-4: $F = F_{\text{current}}$ from 2019 to 2021
- Scenario 1: $F = F_{P_{50\%}^*}$ from 2022 to 2027
- Scenario 2: $F = F_{\text{MSY}}$ from 2022 to 2027
- Scenario 3: $F = F_{P_{30\%}^*}$ from 2022 to 2027
- Scenario 4: $F = 75\%F_{\text{MSY}}$ from 2022 to 2027



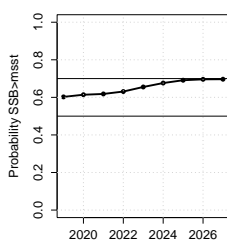
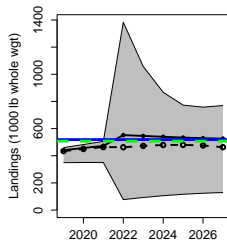
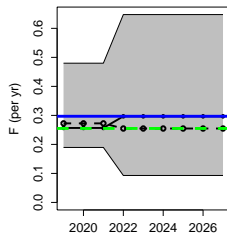
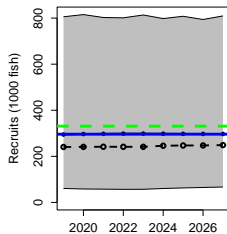
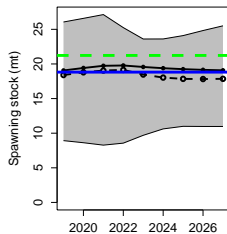
- The following slides show plots of SSB, landings, recruits, dead discards, F and the probability that $SSB > MSST$ for projections
 - ▶ In all panels except the bottom right:
 - ★ solid lines with solid circles = base model values
 - ★ dashed lines with open circles = median values from projections
 - ★ thin solid lines = 5th and 95th percentiles of projections
 - ★ Solid horizontal blue lines = MSY-related quantities from the base model
 - ★ dashed horizontal green lines medians = MSY-related quantities median from projections
 - ▶ In the bottom right panel, the curve represents the proportion of projection replicates for which SSB has reached the replicate-specific MSST.

Projections

Results



Scenario 1: $F = F_{P*_{50\%}}$ from 2022 to 2027

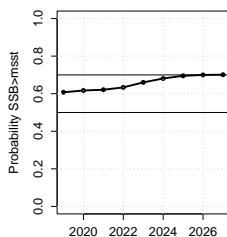
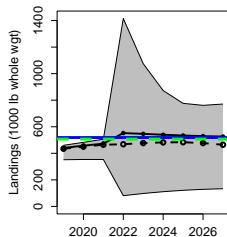
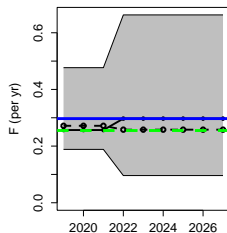
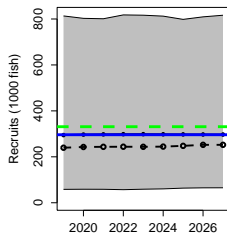
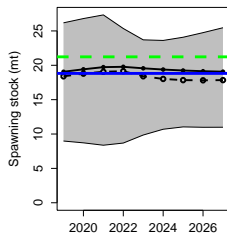


Projections

Results



Scenario 2: $F = F_{MSY}$ from 2022 to 2027

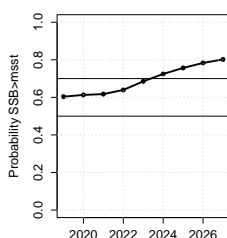
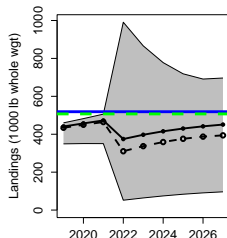
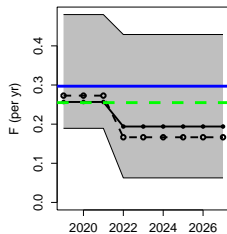
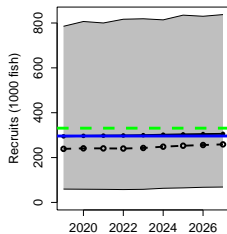
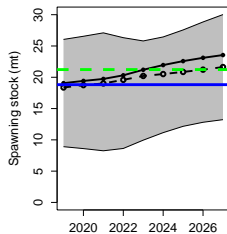


Projections

Results



Scenario 3: $F = F_{P*_{30\%}}$ from 2022 to 2027

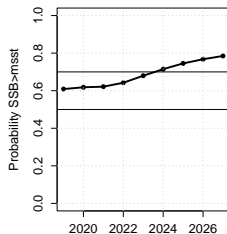
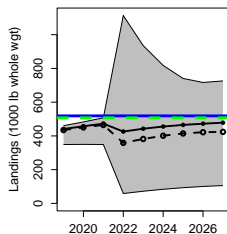
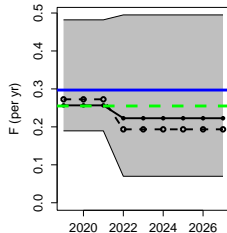
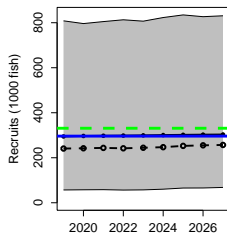
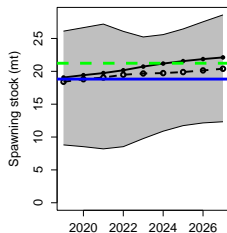


Projections

Results



Scenario 4: $F = 75\%F_{MSY}$ from 2022 to 2027



Projections

Results



Scenario 1: $F = F_{P^*_{50\%}}$ from 2022 to 2027

| Year | R_b | R_{med} | F_b | F_{med} | S_b (mt) | S_{med} (mt) | L_b (n) | L_{med} (n) | L_b (w) | L_{med} (w) | $P(> MSST)$ |
|------|-------|-----------|-------|-----------|------------|----------------|-----------|---------------|-----------|---------------|-------------|
| 2019 | 295 | 241 | 0.26 | 0.27 | 19 | 18 | 54 | 54 | 441 | 434 | 0.602 |
| 2020 | 297 | 241 | 0.26 | 0.27 | 19 | 19 | 57 | 56 | 457 | 450 | 0.614 |
| 2021 | 298 | 242 | 0.26 | 0.27 | 20 | 19 | 59 | 57 | 473 | 464 | 0.618 |
| 2022 | 298 | 241 | 0.30 | 0.25 | 20 | 19 | 68 | 57 | 552 | 462 | 0.631 |
| 2023 | 299 | 241 | 0.30 | 0.25 | 20 | 18 | 67 | 58 | 546 | 473 | 0.655 |
| 2024 | 298 | 246 | 0.30 | 0.25 | 19 | 18 | 67 | 59 | 540 | 479 | 0.676 |
| 2025 | 298 | 247 | 0.30 | 0.25 | 19 | 18 | 66 | 59 | 534 | 480 | 0.691 |
| 2026 | 297 | 247 | 0.30 | 0.25 | 19 | 18 | 66 | 58 | 529 | 476 | 0.696 |
| 2027 | 297 | 249 | 0.30 | 0.25 | 19 | 18 | 65 | 57 | 526 | 463 | 0.697 |

Projections

Results



Scenario 2: $F = F_{MSY}$ from 2022 to 2027

| Year | R_b | R_{med} | F_b | F_{med} | S_b (mt) | S_{med} (mt) | L_b (n) | L_{med} (n) | L_b (w) | L_{med} (w) | $P(> MSST)$ |
|------|-------|-----------|-------|-----------|------------|----------------|-----------|---------------|-----------|---------------|-------------|
| 2019 | 295 | 240 | 0.26 | 0.27 | 19 | 18 | 54 | 54 | 441 | 434 | 0.608 |
| 2020 | 297 | 243 | 0.26 | 0.27 | 19 | 19 | 57 | 56 | 457 | 450 | 0.617 |
| 2021 | 298 | 243 | 0.26 | 0.27 | 20 | 19 | 59 | 57 | 473 | 464 | 0.621 |
| 2022 | 298 | 244 | 0.30 | 0.26 | 20 | 19 | 68 | 58 | 552 | 468 | 0.633 |
| 2023 | 299 | 244 | 0.30 | 0.26 | 20 | 18 | 67 | 59 | 546 | 477 | 0.660 |
| 2024 | 298 | 244 | 0.30 | 0.26 | 19 | 18 | 67 | 59 | 540 | 482 | 0.681 |
| 2025 | 298 | 248 | 0.30 | 0.26 | 19 | 18 | 66 | 59 | 534 | 483 | 0.695 |
| 2026 | 297 | 252 | 0.30 | 0.26 | 19 | 18 | 66 | 58 | 529 | 477 | 0.700 |
| 2027 | 297 | 252 | 0.30 | 0.26 | 19 | 18 | 65 | 57 | 526 | 464 | 0.701 |

Projections

Results



Scenario 3: $F = F_{P_{30\%}^*}$ from 2022 to 2027

| Year | R_b | R_{med} | F_b | F_{med} | S_b (mt) | S_{med} (mt) | L_b (n) | L_{med} (n) | L_b (w) | L_{med} (w) | $P(> MSST)$ |
|------|-------|-----------|-------|-----------|------------|----------------|-----------|---------------|-----------|---------------|-------------|
| 2019 | 295 | 239 | 0.26 | 0.27 | 19 | 18 | 54 | 54 | 441 | 434 | 0.604 |
| 2020 | 297 | 241 | 0.26 | 0.27 | 19 | 19 | 57 | 56 | 457 | 450 | 0.613 |
| 2021 | 298 | 241 | 0.26 | 0.27 | 20 | 19 | 59 | 57 | 473 | 464 | 0.617 |
| 2022 | 298 | 240 | 0.19 | 0.17 | 20 | 20 | 46 | 38 | 375 | 310 | 0.639 |
| 2023 | 300 | 243 | 0.19 | 0.17 | 21 | 20 | 48 | 41 | 397 | 337 | 0.685 |
| 2024 | 302 | 249 | 0.19 | 0.17 | 22 | 21 | 50 | 43 | 416 | 359 | 0.724 |
| 2025 | 303 | 253 | 0.19 | 0.17 | 23 | 21 | 52 | 45 | 430 | 376 | 0.757 |
| 2026 | 305 | 256 | 0.19 | 0.17 | 23 | 21 | 53 | 46 | 442 | 387 | 0.783 |
| 2027 | 306 | 259 | 0.19 | 0.17 | 24 | 22 | 54 | 46 | 451 | 394 | 0.802 |

Projections

Results



Scenario 4: $F = 75\%F_{MSY}$ from 2022 to 2027

| Year | R_b | R_{med} | F_b | F_{med} | S_b (mt) | S_{med} (mt) | L_b (n) | L_{med} (n) | L_b (w) | L_{med} (w) | $P(> MSST)$ |
|------|-------|-----------|-------|-----------|------------|----------------|-----------|---------------|-----------|---------------|-------------|
| 2019 | 295 | 241 | 0.26 | 0.27 | 19 | 18 | 54 | 54 | 441 | 434 | 0.609 |
| 2020 | 297 | 242 | 0.26 | 0.27 | 19 | 19 | 57 | 56 | 457 | 450 | 0.618 |
| 2021 | 298 | 244 | 0.26 | 0.27 | 20 | 19 | 59 | 57 | 473 | 465 | 0.621 |
| 2022 | 298 | 242 | 0.22 | 0.19 | 20 | 19 | 52 | 44 | 426 | 359 | 0.642 |
| 2023 | 300 | 244 | 0.22 | 0.19 | 21 | 20 | 54 | 47 | 443 | 382 | 0.680 |
| 2024 | 301 | 247 | 0.22 | 0.19 | 21 | 20 | 55 | 49 | 455 | 401 | 0.715 |
| 2025 | 302 | 253 | 0.22 | 0.19 | 22 | 20 | 56 | 50 | 465 | 414 | 0.745 |
| 2026 | 303 | 255 | 0.22 | 0.19 | 22 | 20 | 57 | 50 | 473 | 422 | 0.767 |
| 2027 | 303 | 257 | 0.22 | 0.19 | 22 | 20 | 58 | 50 | 478 | 424 | 0.785 |

Summary and conclusions



- This assessment indicates that Tilefish are currently not overfished or experiencing overfishing
- However MCB analysis suggests substantial uncertainty in status of the stock or fishery
- Recruitment is low toward the end of the time series which could be problematic for the stock if that trend continues
- The probability of being above *MSST* only exceeds 50% in all years of all projections