

# USE OF PARAMETRIC BOOTSTRAPPING TO GAUGE THE SIGNIFICANCE OF RETROSPECTIVE PATTERNS IN STOCK ASSESSMENTS

MEAGHAN D. BRYAN AND COLE MONNAHAN

ALASKA FISHERIES SCIENCE CENTER (AFSC)

RESOURCE ECOLOGY AND FISHERIES MANAGEMENT (REFM)

STATUS OF STOCKS AND MULTISPECIES ASSESSMENTS (SSMA)

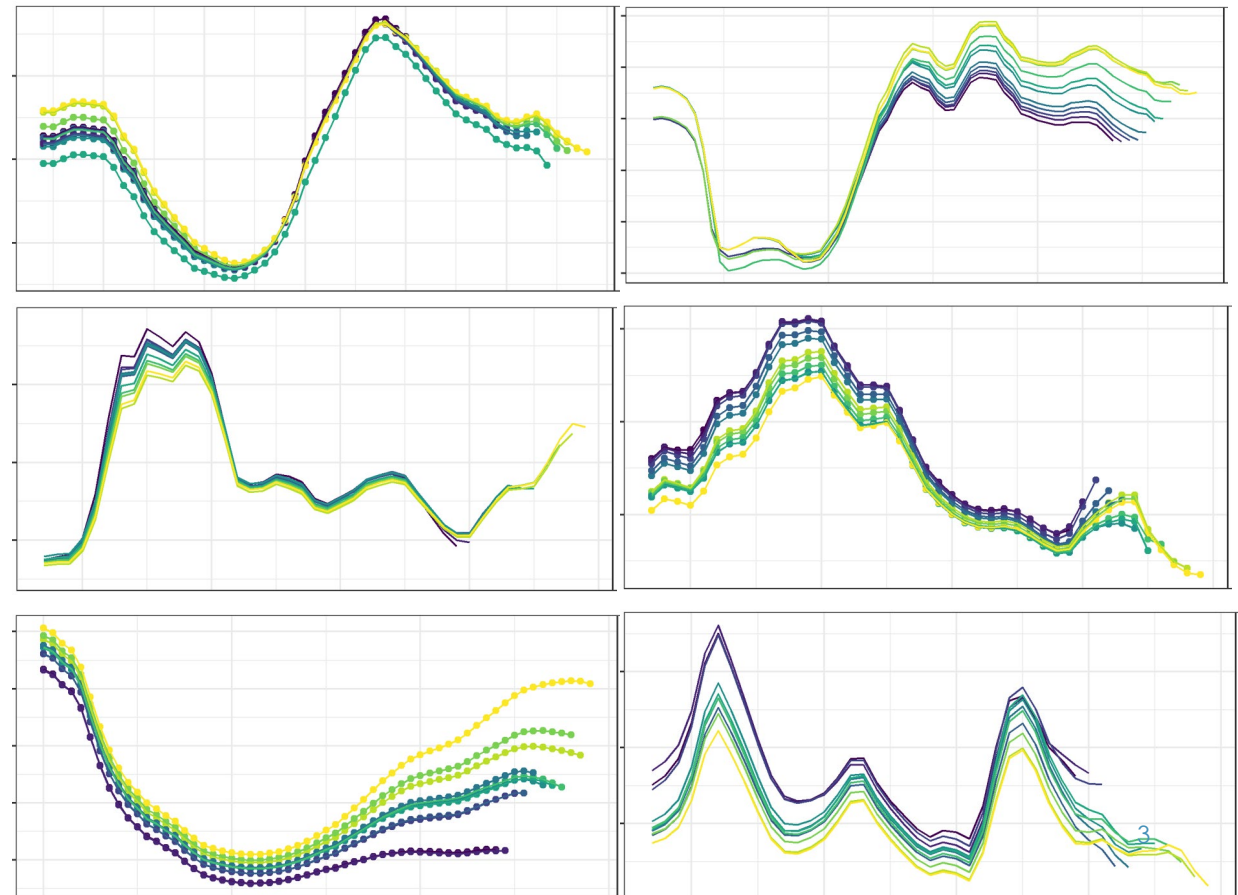


# OUTLINE

- Why use retrospective patterns as a assessment model diagnostic?
- Current method used to determine retrospective pattern significance?
- Propose a new approach
  - Describe the procedure
  - Present results from two case-studies
- Provide guidance about when to use the proposed approach

# RETROSPECTIVE PATTERN AS A DIAGNOSTIC

- Measure internal consistency of an assessment model as new data are added
- Lack of consistency indicates some model misspecification
  - Used, in addition to other evidence, to justify changes in our model structure
    - E.g., time-varying growth, natural mortality, selectivity, catchability

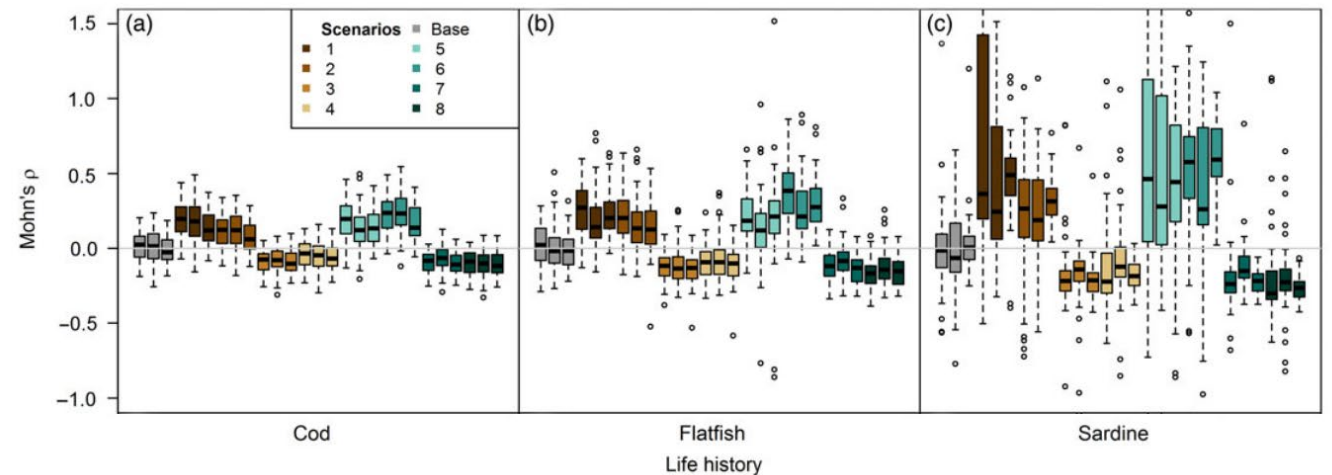


# RISK TABLE

|                          | Considerations   |  |  |  |
|--------------------------|--|--|--|--|
|                          | Assessment-related   | Population dynamics  | Environmental & ecosystem  | Fishery performance  |
| Level 3<br>Major Concern | Major problems with the stock assessment, very poor fits to data, high level of uncertainty, <u>strong retrospective bias.</u> | Stock trends are highly unusual; very rapid changes in stock abundance, or highly atypical recruitment patterns. | Multiple indicators showing consistent adverse signals a) across the same trophic level, and/or b) up or down trophic levels (i.e., predators and prey of stock) | Multiple indicators showing consistent adverse signals a) across different sectors, and/or b) different gear types |

# DEFINING STRONG RETROSPECTIVE PATTERN: CURRENT APPROACH

- Calculate the Mohn's rho statistic to measure the direction and magnitude of the retrospective pattern
- Hurtado-Ferro et al. (2015) rule of thumb used to determine the significance of the pattern
  - For most AFSC species:  
 $-0.15 < \rho < 0.2$
- Straightforward and simple rule



# UNCERTAINTY IN RETROSPECTIVE PATTERN

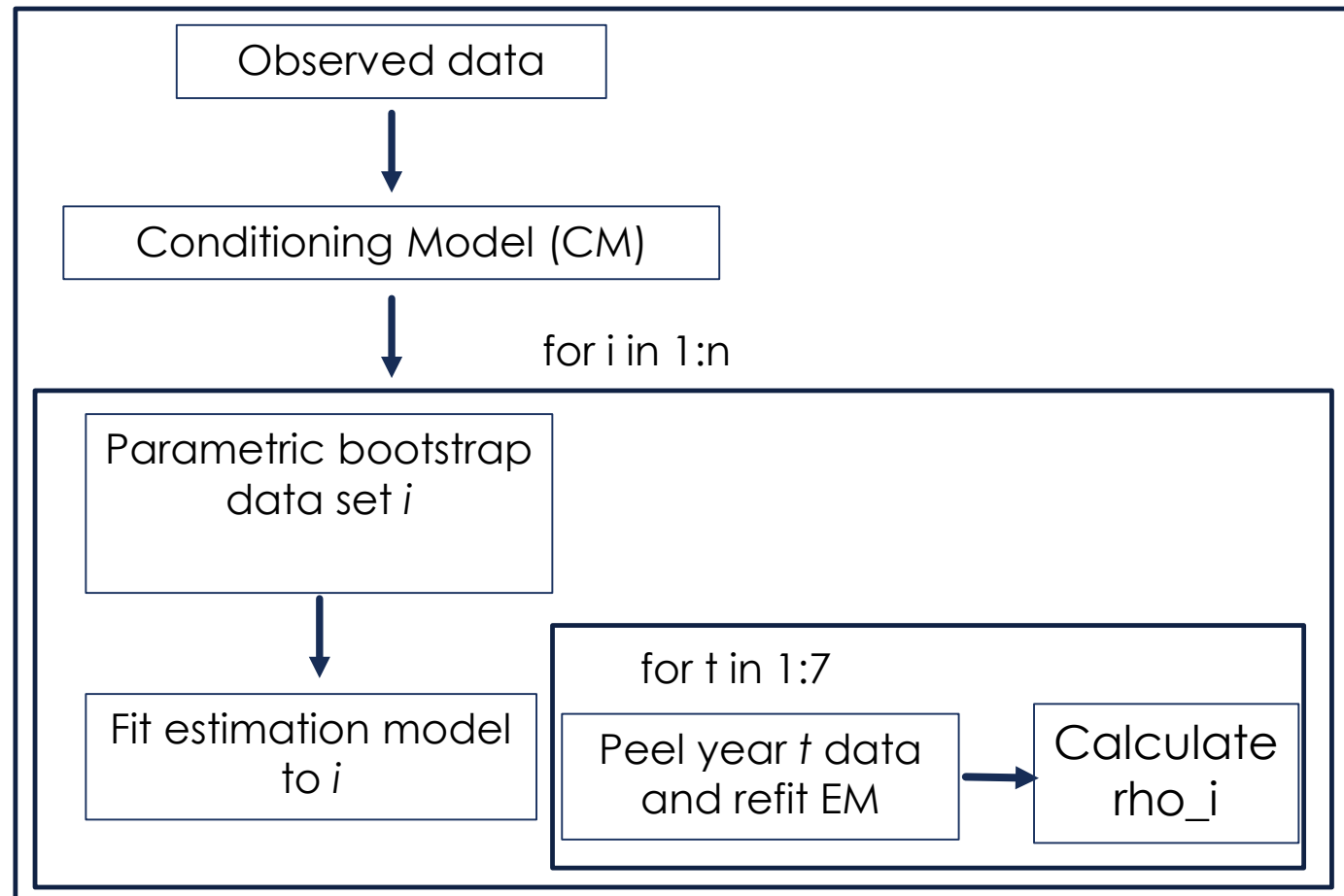
- Quantify uncertainty in  $\rho$  for individual assessments and use this as the metric to determine significance
- Parametric bootstrap procedure (this is built-in for Stock Synthesis 3 models)
  - Data generated from:
    - Assumed probability distribution of the observed data
    - Using expected values of model fit and weights given by input data
- Miller and Legault (2017) used a bootstrap approach to quantify uncertainty in  $\rho$ 
  - Data generated from:
    - Assumed distributions for each data source
    - Using mean of the observations and weights from observed data

## RESEARCH OBJECTIVES

- Compare parametric bootstrap approaches used to quantify uncertainty in Mohn's rho
- Demonstrate how uncertainty can be used to determine significance of retrospective pattern
- Contrast the results to the current rule of thumb

# PROCEDURE

- Generate  $n$  data sets using bootstrap procedure and assessment model
- For each bootstrap data set:
  - Fit estimation model (EM)
  - Run retrospective analysis
    - Seven peels
  - Calculate Mohn's rho



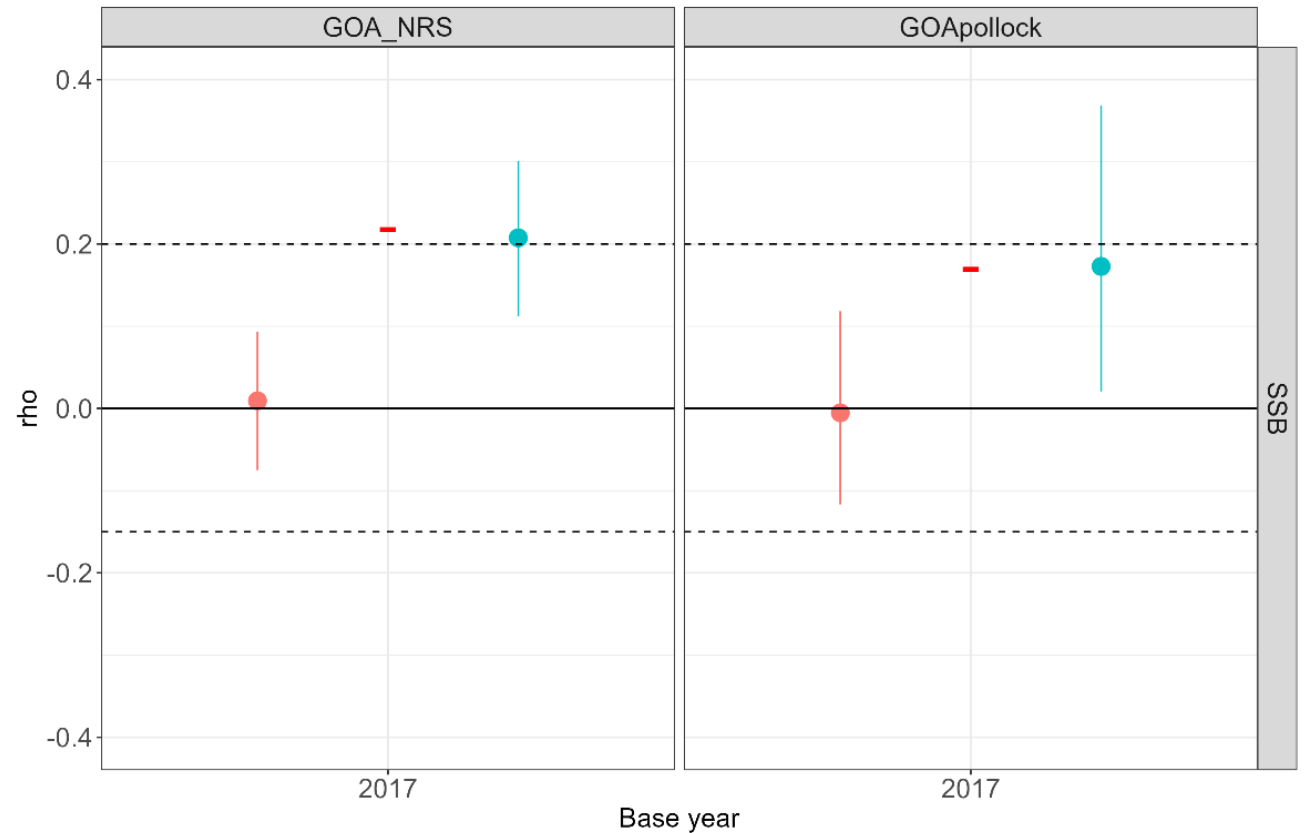


# INTERPRETATION DIFFERENCES

- Bryan and Monnahan
  - “Model” approach
  - Simulated data matches the fitted model structure
  - Null distribution should be centered at 0
  - Rho values outside null distribution suggest significance
- Miller and Legault (2017)
  - “Data” approach
  - Simulated data matches the original data structure (including misspecification or data conflict)
  - Null distribution of rho centered at original rho
  - Rho distribution not containing 0 suggests significance

# CASE STUDIES

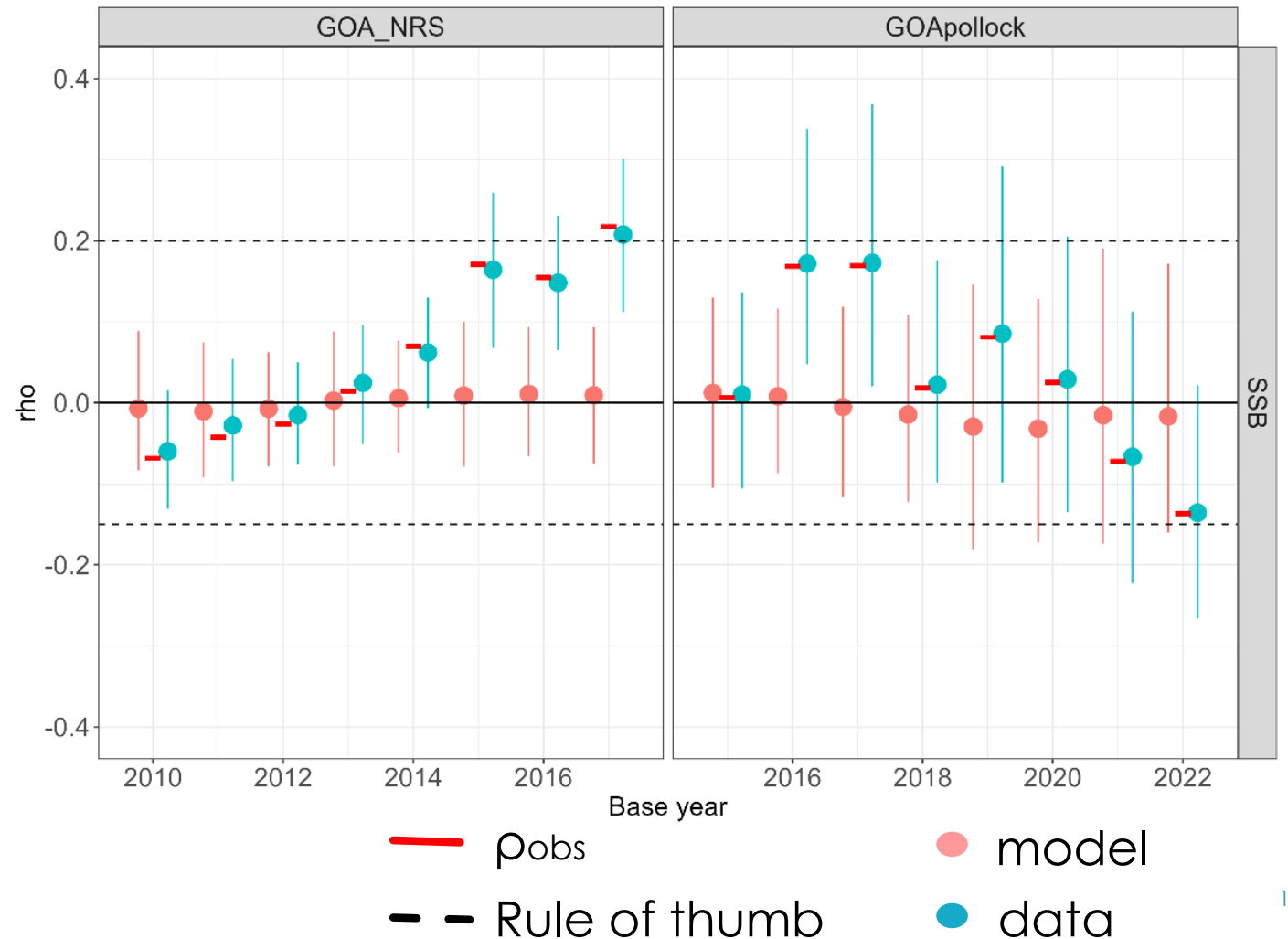
- Case studies
  - Gulf of Alaska walleye pollock (n=1000)
  - Gulf of Alaska northern rock sole (n=500)
- Results are shown for spawning stock biomass (SSB)



— ρ<sub>obs</sub>      ● model  
- - Rule of thumb      ● data

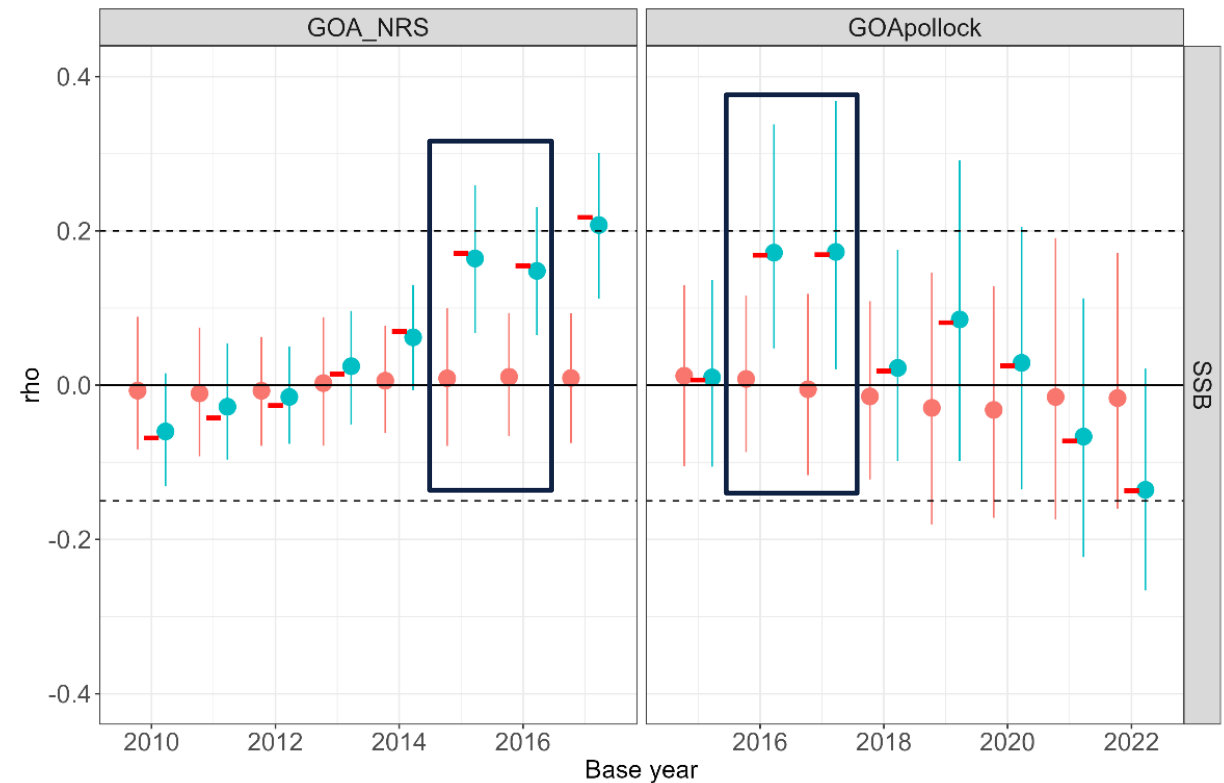
# RESULTS

- Evaluated the rho distribution for several terminal years
- In all cases, the model and data approaches agree



# RESULTS

- Uncertainty approaches and rule of thumb generally agree, but not always
  - Boxes show disagreement with the rule of thumb



- $\rho_{obs}$
- Rule of thumb
- model
- data

# WHEN AND HOW TO ACT

- Statistical significance from model or data approaches
- Scientific significance
  - Rho adjustment (Miller and Legault 2017):
    - $SSB_{\text{adjust}} = \frac{SSB_{\text{terminal}}}{1+\rho}$
    - Adjustment > 10% (AFSC threshold for major model change)
- Szuwalski et al. (2018) - modeling the wrong time-varying process can lead to true bias in reference points and catch advice

|                          | Scientific significance               |  |
|--------------------------|---------------------------------------|--|
| Statistical significance | not significant and small (No action) | Not significant and large (Modify model with care, Szuwalski et al.) |
|                          | Significant and small (Risk table)    | Significant and large (Modify model)                                 |

# CONCLUSIONS AND RECOMMENDATIONS FOR IMPLEMENTATION

- Model and data uncertainty approaches provide a new statistical basis to determine significance of retrospective patterns
- Model approach assumes the fitted model is correct
  - This approach evaluates the inherent retrospective pattern in the model and represents a case-specific rule of thumb
  - Improvement on current rule of thumb
    - Breivik, O.N. et al. (2023) – similar to our model approach for a state space assessment model

# CONCLUSIONS AND RECOMMENDATIONS FOR IMPLEMENTATION

- When should we implement this approach?
  - Introducing a new model
  - Stocks where observed rho changed dramatically between full assessments
  - Stocks that are changing rapidly or near overfished status
  - Stocks that have had historically large rho values

# ACKNOWLEDGMENTS

- Thanks to Tim Miller, Rick Methot, Kelli Johnson, and Ian Taylor



## REFERENCES

- Hurtado-Ferro, F., Szuwalski, C. S., Valero, J. L., Anderson, S. C., Cunningham, C. J., Johnson, K. F., Licandeo, R., McGilliard, C. R., Monnahan, C. C., Muradian, M. L., Ono, K., Vert-Pre, K. A., Whitten, A. R., and Punt, A. E. 2014. Looking in the rear-view mirror: bias and retrospective patterns in integrated, age-structured stock assessment models. *ICES Journal of Marine Science* 72 (1): 99–110.
- Miller, T.J., Legault, C.M. 2017. Statistical behavior of retrospective patterns and their effects on estimation of stock and harvest status. *Fisheries Research* 186: 109-120. <https://doi.org/10.1016/j.fishres.2016.08.002>
- Szuwalski, C. S. Ianelli, J. N. and Punt, Andre´ E. 2018. Reducing retrospective patterns in stock assessment and impacts on management performance. *ICES Journal of Marine Science*, 75: 596–609.
- Breivik, O.N., Aldrin, M., Fuglebakk, E., Nielsen, A. 2023. Detecting significant retrospective patterns in state space fish stock assessment. *CJFAS* 80: 1509-1518. [doi.org/10.1139/cjfas-2022-0250](https://doi.org/10.1139/cjfas-2022-0250)