

Probabilistic Time to Collapse as a Risk Communication Tool or Adding the risk of stock collapse over time to stock assessments and harvest allocation decisions

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- Many fisheries continue to experience the collapse of their fish stocks and the loss of their fisheries
- Even though,
 - Many fisheries are managed based on scientific principles
 - Stock assessments and harvesting advice is provided with the precautionary principle in mind
 - Uncertainty is explicitly included in the development of the advice

(Bailey, 2011; Froese etal., 2021)



- Partly this can be explained as a failure to communicate
 - Providing information is not communicating
 - The message must be in a format useful to the receiver

• However, due to the inherent variability there is always a risk that a fish stock will collapse over a given amount of time.



• (Assuming) We know the current state of the stock, how much should we harvest?





 How much to harvest to avoid stock collapse (and ensure strong future growth)?





Year



- How much to harvest to avoid stock collapse (and ensure strong future growth)?
- So far so simple:
 - Choose harvest level with collapse in the future.





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- So far so simple:
 - Choose harvest level with collapse in the future.
 - 60kt: Collapse is expected far in the future, no worries





- How much to harvest to avoid stock collapse (and ensure strong future growth)?
 - 60kt seem safe?
- However, there is uncertainty about the dynamics!
 - There is a chance the stock will collapse far sooner!





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- How much to harvest to avoid stock collapse (and ensure strong future growth)?
 - 60kt seem safe?
- However, there is uncertainty about the dynamics!
 - There is a chance the stock will collapse far sooner!
 - Reduce the harvest level to eliminate the risk: 50kt





- How much to harvest to avoid stock collapse (and ensure strong future growth)?
 - □ 50kt seem safe?
- Fish recruitment and mortality is subject to random shocks!



Year Non exhaustive set of 1000 stochastic runs



- How much to harvest to avoid stock collapse (and ensure strong future growth)?
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Year Non exhaustive set of 1000 stochastic runs



- How much to harvest to avoid stock collapse (and ensure strong future growth)?
- Fish recruitment and mortality is subject to random shocks!
- Risk can not be eliminated, only reduced!
- How high a probability of collapse within a timeframe do we tolerate?
 - Policy question, not a science question!



Year Non exhaustive set of 1000 stochastic runs



- How much to harvest to avoid stock collapse (and ensure strong future growth)?
- Fish recruitment and mortality is subject to random shocks!
- Risk can not be eliminated, only reduced!
- How can the risk of collapse be communicated to decision makers?
 - These uncertainties are known, but their consequences for decisions are not communicated effectively.



Non exhaustive set of 1000 stochastic runs



- The biggest risk is that the stock collapses / that the fishery becomes unviable
- The risk can not be eliminated
- There is path dependence
 - Choices today restrict future choices
- The risk over time depending on choices needs to be communicated
 - Cumulative probability of collapse over time depending on harvests









Year (0 ≜ 1978)

Non exhaustive set of 1000 stochastic runs









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Year (0 ≜ 1978)

Non exhaustive set of 1000 stochastic runs



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Year (0 ≜ 1978)

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Biomass age 5+ history pred. h=112.5kt 400 95% CI stochastic runs biomass (1,000t) 300 200 100 0 1970 1975 1980 1985 1990 Year



Year (0 ≜ 1978)

Non exhaustive set of 1000 stochastic runs



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Year (0 ≜ 1978)

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Non exhaustive set of 1000 stochastic runs



- The Cumulative Distribution Function (CDF) of the time to collapse *collapses* the difficult abstract uncertainties into a single line.
- This allows an easy comparison of the future consequences of harvesting choices.



Year (0 ≜ 1978)



• We can go one step further to communicate in the language of risk managers.

Cumulative Probability of Stock Collapse



- The risk matrix is a product of the risk management literature.
- Part of the ISO 31000 risk management process
- Choices presented as combinations of outcome probability and severity
- Colours denote the range from tolerable to catastrophic



• We can go one step further to communicate in the language of risk managers.





- The standard risk matrix does not show • probabilistic outcomes for a single action.
- We extend the risk matrix to show the outcome probability of different actions over time.
- Each line represents the consequences of ٠ maintaining a harvest level through time.
- Risk preferences, the colouring, must come • from the decision makers!
- Colours and time horizon used here are for illustration.



Cumulative probability of collapse [%] Cumulative risk of collapse over 20k times 3125 SOW runs for each harvest level. Floating labels indicate harvests in 1,000t.



Risk Management Process

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Figure from Neuenhoff et al. 2019

Cod Biomass aged five plus. Data from Swain et al. 2019₂₉



Case Study: Cod Stocks in the Southern Gulf of St. Lawrence

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Figure from Neuenhoff et al. 2019

Biomass aged five plus, growth, fisheries, and mortality measured in biomass. Data from Swain et al. 2019

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Cumulative probability of collapse [%]

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Year

Cumulative probability of collapse [%]

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Cumulative probability of collapse [%]

- The shape of the cumulative distribution of collapse is responsive to the state of the stock.
 - As the collapse draws closer, the lines shift toward the top right corner of the matrix
 - As the uncertainty is reduced, the lines flatten/spread across less time

- The relative positioning of the CDFs for the different harvest levels gives a good indication of the effectiveness of the harvest (or other management measure)
 - E.g. in 1986 the decision maker could achieve almost any risk level by setting the harvest rate. In 1993 this ability is severely constrained.

- The cumulative probability of collapse within a given time frame:
 - Incorporates and visualizes uncertainty present in the predictions
 - Shows how risk is affected by management choices
 - Does not require in depth knowledge from the user
 - It does require quite some knowledge from the modeler though
- In combination with pre-determined risk tolerances, it can show compliance or noncompliance of a proposed harvest level with the defined risk tolerances.
- We propose adding this visualization to existing stock advice procedures, based on projections made using existing state of the art models for the respective species.
 The most recent stock assessment of Cod stocks in the
- More details in our upcoming paper:
 - Accepted for publication in the ICES Journal of Marine Science

Thank you for your attention!

How long do you want (to be able) to fish?

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- To investigate the time to collapse as a communication tool in fisheries management we built a simple model of the sGSL cod stocks. Calibrated to Stock timeseries 1950 2017
- This model has the general properties of this fish stock
 - Increasing rate of natural mortality
 - Ricker recruitment
 - Random shocks to both natural mortality and growth
- We include two sources of uncertainty
 - Parametric, i.e. uncertainty about the calibration
 - Stochastic, i.e. irreducible uncertainty stemming from randon fluctuations
- This model is used only for illustration purposes. Application of the probabilistic time to collapse (or to recovery as was proposed in the session) in stock advice would be based on existing state-of-the-art models.
- Other sources of uncertainty, can be incorporated analogous to the parametric and stochastic uncertainty demonstrated here.

 To investigate the time to collapse as a communication tool in fisheries management we built a simple model of the sGSL cod stocks. Calibrated to Stock timeseries 1950 – 2017

Biomass in the next year: current biomass + growth

- natural mortality - harvests

Ricker growth function with random shocks

$$B_{t+1} = B_t + g(B_t) - m(B_t, t) - f_t$$

$$g(B_t) = \max(\alpha B_t e^{-\beta B_t} + \epsilon_t^g, 0)$$

Time varying rate of natural mortality with random shocks

$$m(B_t, t) = \max((\gamma_1 + \gamma_2 t + \gamma_3 t^2 + \epsilon_t^m)B_t, 0)$$

Random shocks, allowing for correlation.

$$(\epsilon_t^g \ \epsilon_t^m) \stackrel{\text{i.i.d.}}{\sim} \mathcal{N} \left(\mu = (0 \ 0), \ \Sigma = \begin{pmatrix} s_{g,g} & s_{g,m} \\ s_{g,m} & s_{m,m} \end{pmatrix} \right)$$

Parametric Uncertainty

- Parameter uncertainty is modelled by sampling the joint model likelihood, given the data available in a given year.
- Each of the parameters of the model where sampled five times equidistantly.
 - Their contribution to the overall distribution of collapse times is weighted by their likelihood of being correct given the data.

prob. 0.02 0.03

0.210

Fit Type: sTime3 n sample: 5

min.coefs

2 16080-01

Likelihood Cutoff Ratio: 10.00

0.220

0.230

m1

3125 possible parameter combinations •

Probabilities and Bivariate Probability Contours Fit Window 1950-2017

- The model for each year is calibrated only to the part of the time series available at that year.
- 25 years of data as the minimum for calibration means 1974 is the first year for analysis.
- In 1974 the increasing natural mortality rate is not yet reflected in the data, and hence also not in the calibrated model. Consequently, the predicted collapse time distributions do also not reflect that natural mortality will increase within the predicted time frame.
 - None the less, the produced risk assessment reflects the state of the knowledge at that time, including the uncertainty.
- With additional data points, the parametric uncertainty is reduced, but the uncertainty about future stock states from random shocks can not be reduced.

Data points, maximum likelihood fit, and prediction interval for the model calibrated to the years 1950 – 1974 (shown in saturated colours) the remaining data points are shown for context (in desaturated colours). The dashed lines show the 95% confidence interval of the model fit, the dotted lines additionally show the 95% confidence interval of the yearly shocks. The left panel shows the annual rate of biomass losses due to natural mortality. The right panel shows annual biomass gain due to natural growth.

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- The shocks of the 20,000 stochastic runs are generated with the same statistical properties as the model fit residuals.
- In the modelling framework chosen for our proof-of-concept model, the model fit residuals conceptually correspond to the shocks, the non-deterministic part of growth and mortality.
- Shock uncertainty in stock predictions is included in the model through 20,000 stochastic runs for each of the 3125 parameter combinations.
 - 62.5 million runs per harvest level for each of the 44 analysed years
 - 2.75 billion runs over all for the paper.

Natural mortality residuals