SPiCT scenarios for the Greater silver smelt (Argentina silus) in Subareas 1, 2, and 4, and Division 3.a (Northeast Arctic, North Sea, Skagerrak and Kattegat)

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Introduction

This working document present a series of different assessments using the surplus production model in continous time (SPiCT; Pedersen and Berg (2017)) available as an R package (https://github.com/DTUAqua/spict).

Read in the data

```
## Read in the data
dat <- readxl::read_xlsx("../data/GSS_indices270120_AK.xlsx")
## Sum up the catches from each area to get the total catch
dat$catchTOT <- dat$catch1and2 + dat$catch3 + dat$catch4

## run retro or not
runretro <- FALSE</pre>
```

Scenario 1

Table 1: Input data for Scenario 1

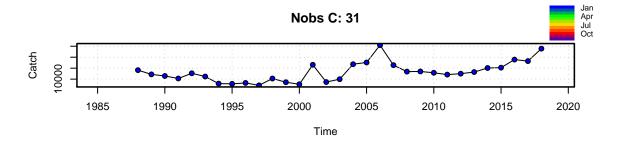
Input data	Name	Range	Notes
Catch Biomass indices	Total catch Shrimp survey	1988-2018 1984-2002 2005-2018	Split in two periods
			Default priors

```
## Choose only the years where the survey was in October
w <- !is.na(dat$northsea_month) & dat$northsea_month == 10
## Choose only the years where the survey was in January or February
v <- !is.na(dat$northsea_month) & dat$northsea_month %in% c(1, 2)</pre>
```

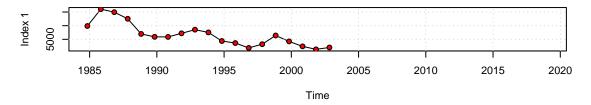
```
## Make the input list
inp_NS <- list(timeC = dat$year,</pre>
                                                                         ## Timing of catch
               obsC = dat$catchTOT,
                                                                         ## Observed catches
               timeI = list(dat$year[w] + dat$northsea_month[w] / 12, ## Timing of survey index
                             dat$year[v] + dat$northsea_month[v] / 12),
               obsI = list(dat$northsea_SA[w],
                                                                         ## Observed indices
                            dat$northsea_SA[v]),
               optimiser.control = list(iter.max = 1e5,
                                                                         ## Optimiser options
                                         eval.max = 1e5),
                                                                         ## sometimes help converge
               priors = list(
                                                                         ## List of priors (empty, i.e. de
                                                                         ## see possible priors with list.
                 ))
## Check input time series, remove missing and zero observations
inp_NS <- check.inp(inp_NS)</pre>
```

Removing zero, negative, and NAs in C series

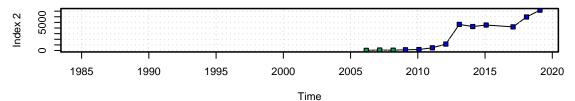
```
## Plot input data
plotspict.data(inp_NS)
```



Nobs I: 19



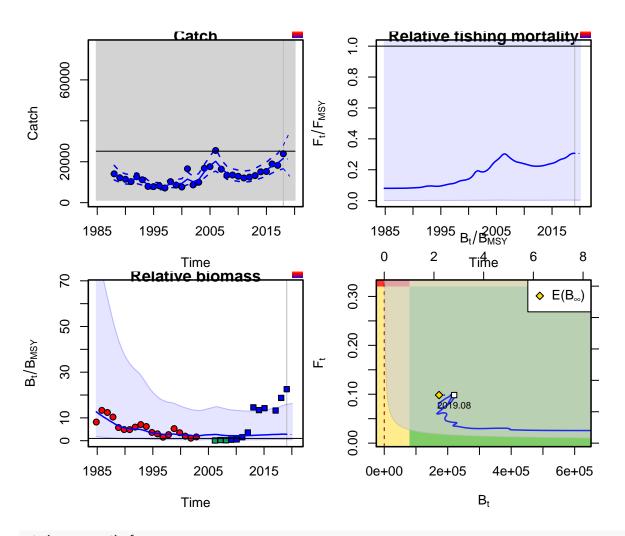
Nobs I: 13



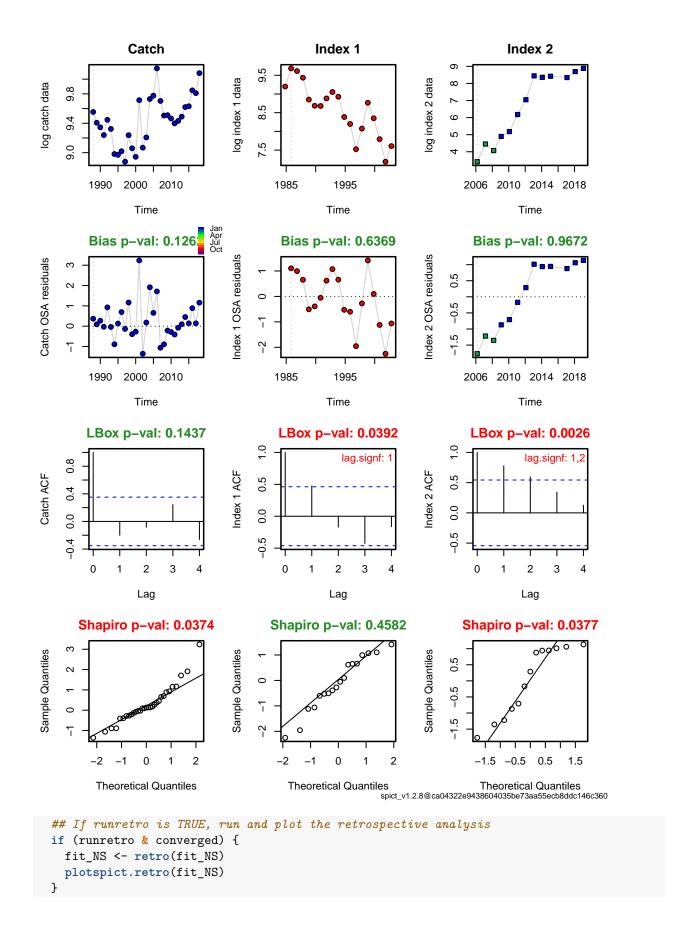
spict_v1.2.8@ca04322e9438604035be73aa55ecb8ddc146c360

```
## Fit spict
fit_NS <- fit.spict(inp_NS)</pre>
## Summary of the fit - in the vignette there is a line-by-line description of that summary
fit NS
## Convergence: 0 MSG: relative convergence (4)
## Objective function at optimum: 57.3810526
## Euler time step (years): 1/16 or 0.0625
## Nobs C: 31, Nobs I1: 19, Nobs I2: 13
##
## Priors
##
       logn \sim dnorm[log(2), 2^2]
   logalpha ~ dnorm[log(1), 2^2]
     logbeta ~ dnorm[log(1), 2^2]
##
##
## Model parameter estimates w 95% CI
##
                               cilow
               estimate
                                            ciupp
                                                     log.est
##
   alpha1 2.390221e+00
                           0.6561160 8.707540e+00 0.8713858
                           3.5324126 3.981958e+01 2.4731699
##
   alpha2 1.185998e+01
##
  beta
          9.416977e-01
                           0.3601728 2.462137e+00 -0.0600710
## r
          1.999363e-01
                           0.0614380 6.506482e-01 -1.6097566
##
   rc
          6.376559e-01
                           0.0245182 1.658384e+01 -0.4499565
   rold
##
          5.361626e-01
                           0.0090055 3.192149e+01 -0.6233178
##
          2.551782e+04 1202.3903144 5.415541e+05 10.1471325
##
          2.797479e+05 4660.6209528 1.679152e+07 12.5416442
          1.544270e-02
                           0.0000792 3.010343e+00 -4.1706220
##
   q1
##
          4.063500e-03
                           0.0000136 1.210055e+00 -5.5057111
  q2
          6.270977e-01
                           0.0459170 8.564407e+00 -0.4666529
##
  sdb
          1.576217e-01
                           0.0512043 4.852053e-01 -1.8475574
##
   sdf
          1.625876e-01
                           0.0768551 3.439552e-01 -1.8165384
##
   sdi1
          3.767507e-01
                           0.2468519 5.750051e-01 -0.9761716
  sdi2
          1.869391e+00
                           1.2648928 2.762781e+00 0.6256125
                           0.0978477 2.395782e-01 -1.8766094
##
   sdc
          1.531084e-01
##
## Deterministic reference points (Drp)
##
              estimate
                              cilow
                                           ciupp
                                                   log.est
## Bmsyd 8.003635e+04 544.6157987 1.176209e+07 11.290236
##
   Fmsyd 3.188279e-01
                         0.0122591 8.291918e+00 -1.143104
  MSYd 2.551782e+04 1202.3903144 5.415541e+05 10.147132
## Stochastic reference points (Srp)
                                                  log.est rel.diff.Drp
##
              estimate
                              cilow
                                           ciupp
   Bmsys 7.831263e+04 552.3441008 1.110335e+07 11.268464 -0.022010800
##
                          0.0122407 8.421108e+00 -1.136125 0.006954625
   Fmsys 3.210608e-01
   MSYs 2.514696e+04 1185.4485013 5.334435e+05 10.132492 -0.014747787
##
##
## States w 95% CI (inp$msytype: s)
##
                       estimate
                                      cilow
                                                   ciupp
                                                           log.est
  B 2019.08
##
                   2.203227e+05 776.4156248 6.252077e+07 12.302849
## F 2019.08
                  9.836480e-02
                                0.0003846 2.515559e+01 -2.319072
## B_2019.08/Bmsy 2.813374e+00
                                 0.4986872 1.587182e+01 1.034385
## F_2019.08/Fmsy 3.063744e-01
                                 0.0036177 2.594577e+01 -1.182947
##
```

```
## Predictions w 95% CI (inp$msytype: s)
                     prediction
##
                                                    ciupp
                                                            log.est
                                       cilow
## B 2019.08
                   2.203227e+05 7.764156e+02 6.252077e+07 12.302849
## F_2019.08
                   9.836480e-02 3.846000e-04 2.515559e+01 -2.319072
## B_2019.08/Bmsy 2.813374e+00 4.986872e-01 1.587182e+01 1.034385
## F_2019.08/Fmsy 3.063744e-01 3.617700e-03 2.594577e+01 -1.182947
## Catch 2019.08 2.110211e+04 1.295520e+04 3.437222e+04 9.957128
## E(B_inf)
                   1.727415e+05
                                          NA
                                                       NA 12.059552
## If the model converged, it reports convergence as 0
## Continue with plotting and diagnostics only if convergence was reached
converged <- fit_NS$opt$convergence == 0</pre>
if (converged) {
  ## Calculate the One Step Ahead (osa) residuals
  fit_NS <- calc.osa.resid((fit_NS))</pre>
  ## Make a plot showing relative F, relative B, Kobe plot catch
  par(mfrow = c(2,2), ## 2x2 subplots
      mar = c(4.1, 4.1, 0.5, 0.5)) ## Change default margins for the plots
  plotspict.catch(fit_NS)
  plotspict.ffmsy(fit_NS)
  plotspict.bbmsy(fit_NS)
  plotspict.fb(fit_NS)
```



```
if (converged) {
  plotspict.diagnostic(fit_NS)
}
```



Scenario 2

Scenario 3

Scenario 4

Referneces

Pedersen, Martin W., and Casper W. Berg. 2017. "A stochastic surplus production model in continuous time." Fish and Fisheries 18 (2): 226-43. https://doi.org/10.1111/faf.12174.