Short Term Forecasting for advice using FLash o8 March, 2017

Short term forecasts help projecting the stock to forecast the implications of different management choices over a limited number of years. There are three necessary ingredients when projecting in FLR: a stock object - that you will forecast and about which you have made some assumptions about what will happen in the future, a stock-recruitment relationship, a projection control - specify what targets and when to hit them.

In FLR, the method for short term forecasts is called fwd(), which takes an FLStock, FLSR, and fwdControl.

Required packages

To follow this tutorial you should have installed the following packages:

- CRAN: ggplot2
- FLR: FLCore, FLash, FLAssess, FLBRP, ggplotFL

if you are using Windows, please use 32-bit R version You can do so as follows,

```
install.packages(c("FLAssess", "FLBRP", "ggplotFL"),
    repos = "http://flr-project.org/R")
```

The example used here is for the ple4 data that is available in FLR. This needs to be loaded first

```
library(FLAssess)
library(FLash)
library(ggplotFL)
library(FLBRP)
data(ple4)
summary(ple4)
An object of class "FLStock"
Name: Plaice in IV
Description: Imported from a VPA file. ( N:\Projecte [...]
Quant: age
Dims: age year
                           season area
                                           iter
                   unit
   10 52 1 1 1
                      1
Range: min max pgroup minyear maxyear minfbar maxfbar
       10 10 1957
                       2008
                               2
                                   6
```

```
catch
             : [1521111], units = t
             : [10 52 1 1 1 1], units = 10^3
catch.n
             : [ 10 52 1 1 1 1 ], units =
catch.wt
discards
             : [ 1 52 1 1 1 1 ], units = t
             : [ 10 52 1 1 1 1 ], units = 10^3
discards.n
             : [10 52 1 1 1 1], units = kq
discards.wt
landings
             : [ 1 52 1 1 1 1 ], units = t
landings.n
             : [10 52 1 1 1 1], units = 10^3
             : [ 10 52 1 1 1 1 ], units = kg
landings.wt
stock
             : [ 1 52 1 1 1 1 ], units = t
             : [10 52 1 1 1 1], units = 10^3
stock.n
             : [ 10 52 1 1 1 1 ], units =
stock.wt
             : [10 52 1 1 1 1], units = m
m
             : [ 10 52 1 1 1 1 ], units = NA
mat
harvest
             : [ 10 52 1 1 1 1 ], units = f
harvest.spwn : [ 10 52 1 1 1 1 ], units = NA
             : [ 10 52 1 1 1 1 ], units = NA
m.spwn
```

Extending the stock object for the projections

Our ple4 stock goes up to 2008, and in this example we want to make a 3 year projection, so we need to extend the stock by 3 years (nyears).

The projection will predict abundances in the future, but what will the future stock weights, maturity, natural mortality, etc. be like? The assumptions on those will affect the outcome of the projection. Rather than using window() or trim() that makes all future data NA, we use the stf() function (short term forecast) that has several options that allow to control the assumptions about the future stock parameters.

These assumptions specify how many years you want to average over to set future values For example, wts.nyears is the number of years over which to calculate the *.wt, *.spwn, mat and m slots. By default this is set to 3 years. This is a fairly standard assumption for STFs, i.e. the future mean weights at age will be the same as the mean of the last 3 years. Hence, What is going to happen in the next few years is probably similar to what happened in the last few years.

A simple 3 year forecast for the weights, natural mortality, etc. assuming these are equal to their averages over the last 3 years is done by

```
ple4_stf <- stf(ple4, nyears = 3, wts.nyears = 3,</pre>
    na.rm = TRUE)
```

Previously, the stock went up to 2008. Now the stock goes up to 2011 and the future weights are equal to the means of the last three observed years, as can be observed from the last six years of the stock weights of the ple4_stf object.

```
maxyr_stk <- range(ple4)[["maxyear"]]</pre>
range(ple4_stf)
      min
                max plusgroup
                                minyear
        1
                 10
                           10
                                   1957
 maxyear
            minfbar
                      maxfbar
                  2
     2011
                            6
maxyr_stf <- range(ple4_stf)[["maxyear"]]</pre>
stock.wt(ple4_stf)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age 2006
              2007
                       2008
                                2009
 1 0.053000 0.048000 0.050000 0.050333
  2 0.129000 0.093000 0.114000 0.112000
  3 0.195000 0.239000 0.200000 0.211333
  4 0.321000 0.241000 0.278000 0.280000
  5 0.354000 0.337000 0.355000 0.348667
  6 0.424000 0.394000 0.429000 0.415667
  7 0.439000 0.458000 0.484000 0.460333
 8 0.506000 0.412000 0.627000 0.515000
 9 0.583000 0.526000 0.598000 0.569000
  10 0.730673 0.548489 0.730672 0.669945
    year
age 2010
              2011
  1 0.050333 0.050333
 2 0.112000 0.112000
  3 0.211333 0.211333
  4 0.280000 0.280000
  5 0.348667 0.348667
  6 0.415667 0.415667
  7 0.460333 0.460333
  8 0.515000 0.515000
  9 0.569000 0.569000
  10 0.669945 0.669945
```

units: kg

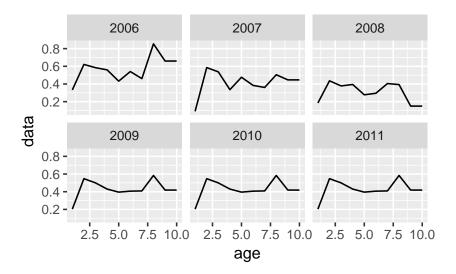
For maturity the same assumption (of the future being the same as the average as the last 3 years) holds, but those maturities were assumed constant already.

```
mat(ple4_stf)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age 2006 2007 2008 2009 2010 2011
    0.0
          0.0
               0.0
                    0.0
                         0.0
                              0.0
    0.5
          0.5
               0.5
                    0.5
                         0.5
                              0.5
     0.5
         0.5
               0.5
                    0.5
                         0.5
                              0.5
     1.0
         1.0
               1.0
                    1.0
                         1.0
                              1.0
     1.0
         1.0
               1.0
                    1.0
                         1.0
                              1.0
    1.0
         1.0
               1.0
                    1.0
                         1.0
                              1.0
         1.0
     1.0
               1.0
                    1.0
                         1.0
                              1.0
    1.0
         1.0
               1.0
                    1.0
                         1.0
                              1.0
    1.0
          1.0
               1.0
                    1.0
                         1.0
                              1.0
  10 1.0 1.0 1.0
                    1.0
                         1.0
                              1.0
```

units: NA

Notice that the future fishing mortality has also been set (mean of the last 3 years, by default).

```
ggplot(harvest(ple4_stf)[, ac((maxyr_stf - 5):maxyr_stf)]) +
    geom_line(aes(x = age, y = data)) + facet_wrap(~year)
```



The stock numbers at age and the catch numbers at age are not forecasted yet: this is what the fwd() function will perform later.

```
stock.n(ple4_stf)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age 2006
              2007
                       2008
                                 2009
 1 820006.3 949341.2 844041.2
                                       NA
  2 554592.5 531914.3 784234.7
                                       NA
  3 422068.8 269655.1 267892.4
                                       NA
     88052.9 212889.2 142573.0
  4
                                       NA
 5 136368.1 45514.9 137660.4
                                       NA
     17534.5 80086.2 25575.8
                                       NA
      16289.7
                9240.6 49356.8
                                       NA
 8
      4114.0
                9297.5
                                       NA
                         5828.1
 9
      2688.5
                1585.1
                         5078.7
                                       NA
  10
      2854.5
                3660.1
                         6750.3
                                       NA
    year
age 2010
              2011
 1
           NA
                    NA
 2
           NA
                    NA
  3
                    NA
           NA
  4
           NA
                    NA
 5
           NA
                    NA
  6
                    NA
           NA
 7
           NA
                    NA
 8
           NA
                    NA
  9
           NA
                    NA
  10
           NA
                    NA
```

Meanwhile, the landings and discards are forecasted ratios of what is discarded and what is landed over the last 3 years of data.

```
landings.n(ple4_stf)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
```

units: 10³

```
age 2006
               2007
                          2008
 1 3.5500e+02 1.2860e+03 3.8000e+02
  2 1.8987e+04 1.9205e+04 1.0970e+04
  3 6.7465e+04 3.7309e+04 4.2865e+04
  4 2.5254e+04 4.7053e+04 3.7970e+04
  5 4.2525e+04 1.4971e+04 2.9476e+04
  6 6.5550e+03 1.7142e+04 5.7000e+03
  7 4.9670e+03 2.4590e+03 6.7520e+03
  8 2.0530e+03 1.8560e+03 9.1200e+02
 9 1.2350e+03 5.4300e+02 6.7300e+02
  10 1.3190e+03 1.2590e+03 8.9600e+02
    year
age 2009
               2010
                          2011
 1 6.9226e-03 6.9226e-03 6.9226e-03
  2 6.8339e-02 6.8339e-02 6.8339e-02
  3 4.2124e-01 4.2124e-01 4.2124e-01
  4 7.9214e-01 7.9214e-01 7.9214e-01
  5 9.2489e-01 9.2489e-01 9.2489e-01
  6 8.5454e-01 8.5454e-01 8.5454e-01
  7 7.4116e-01 7.4116e-01 7.4116e-01
 8 6.4967e-01 6.4967e-01 6.4967e-01
 9 1.0000e+00 1.0000e+00 1.0000e+00
  10 1.0000e+00 1.0000e+00 1.0000e+00
units: 10<sup>3</sup>
discards.n(ple4_stf)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
                2007
                          2008
age 2006
 1 2.2047e+05 7.7312e+04 1.3541e+05
 2 2.2508e+05 2.0514e+05 2.5263e+05
  3 1.1022e+05 6.9312e+04 3.7393e+04
  4 1.0656e+04 1.0735e+04 6.2370e+03
  5 3.0000e+03 1.4370e+03 2.2820e+03
  6 4.1000e+02 7.1510e+03 5.1700e+02
  7 7.5400e+02 2.0400e+02 8.8820e+03
  8 1.9400e+02 1.6490e+03 8.9100e+02
  9 0.0000e+00 0.0000e+00 0.0000e+00
  10 0.0000e+00 0.0000e+00 0.0000e+00
    year
```

```
2009
                2010
                           2011
age
 1 9.9308e-01 9.9308e-01 9.9308e-01
  2 9.3166e-01 9.3166e-01 9.3166e-01
  3 5.7876e-01 5.7876e-01 5.7876e-01
  4 2.0786e-01 2.0786e-01 2.0786e-01
  5 7.5111e-02 7.5111e-02 7.5111e-02
  6 1.4546e-01 1.4546e-01 1.4546e-01
  7 2.5884e-01 2.5884e-01 2.5884e-01
  8 3.5033e-01 3.5033e-01 3.5033e-01
  9 0.0000e+00 0.0000e+00 0.0000e+00
  10 0.0000e+00 0.0000e+00 0.0000e+00
units: 10<sup>3</sup>
```

The stock-recruitment relationship (SRR)

A short term forecast does not use a SRR, instead it generally assumes that recruitment in the future is the mean of the historic recruitments. However, we still need to have a SRR that contains the mean value, which is what we mimic for this example. First we estimate the geometic mean recruitment, that we then add to an SRR object

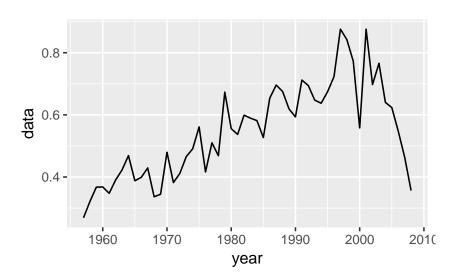
```
mean_rec <- mean(rec(ple4))</pre>
ple4_sr <- as.FLSR(ple4, model = "geomean")</pre>
params(ple4_sr)["a", ] <- mean_rec</pre>
params(ple4_sr)
An object of class "FLPar"
params
1039494
units: NA
```

The control object

The final thing we need to set up is the control object. This tells the projection what to do, i.e. what level of fishing mortality to use

A standard scenario for a STF is the 'status quo' scenario, assuming that the future mean fishing mortality will be the same as the mean of the last X years (X depends on the stock). We will set up this scenario as a simple example using the Fbar in the last year (2008), being 0.356 per year.

```
round(fbar(ple4), 3)
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age
     1957 1958 1959 1960 1961
 all 0.269 0.321 0.367 0.368 0.348
           42 years]
    year
     2004 2005 2006 2007 2008
age
 all 0.640 0.623 0.548 0.464 0.356
ggplot(fbar(ple4), aes(x = year, y = data)) +
   geom_line()
```



Below we define the last year of the stock and the status quo Fbar fbar_SQ.

```
fbar_SQ <- mean(fbar(ple4)[, as.character(maxyr_stk)])</pre>
```

Now we introduce the control object: fwdControl(). This takes 1 argument - a data.frame that sets:

- The quantity (or type) of the target (we are using Fbar but can also set catch etc. see MTF)
- The value of the target, here status quo F
- The year the target is to be hit
- Some other things that we will ignore for now

Let's make the data.frame

```
ctrl_target <- data.frame(year = 2009:2011, quantity = "f",</pre>
    val = fbar_SQ)
# Set the control object - year, quantity and
# value for the moment
ctrl_f <- fwdControl(ctrl_target)</pre>
ctrl_f
Target
  year quantity min
                       val max
              f NA 0.3563 NA
1 2009
2 2010
              f NA 0.3563 NA
3 2011
              f NA 0.3563 NA
    min
            val
                    max
  1
         NA 0.35631
                          NA
  2
         NA 0.35631
                          NA
  3
         NA 0.35631
                          NA
```

We see that we have what looks like our ctrl_target, but now it has two more columns (min and max). There is another table underneath which is for uncertainty but that we ignore in STF.

Running the STF

Below we run a simple STF with 'status quo' future fishing mortality. Remember we had to make assumptions about the future (weights, selection pattern, discard ratio etc.).

This is done using fwd(), which takes three objects: the stock, the control object, the SRR. It returns an updated FLStock object. This updated has forecasted stock numbers, based on the recruitment assumptions, the fishing mortality, and the natural mortality. Because we now have stock numbers, we can calculate forecasted ssb.

```
ple4_sq <- fwd(ple4_stf, ctrl = ctrl_f, sr = ple4_sr)</pre>
```

Below, we check if the recruitment in the forecasted stock indeed corresponds to the mean recruitment

```
mean_rec
[1] 1039494
rec(ple4_sq)[, ac((maxyr_stf - 5):maxyr_stf)]
```

```
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
  year
age 2006
            2007
                    2008
                            2009
                                    2010
 1 820006 949341 844041 1039494 1039494
age 2011
 1 1039494
units: NA
  Similarly, we check if the fbar in the forecasted stock indeed corre-
sponds to the status quo fishing mortality.
round(fbar_SQ, 3)
[1] 0.356
round(fbar(ple4_sq)[, ac((maxyr_stf - 5):maxyr_stf)],
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age 2006 2007 2008 2009 2010 2011
 all 0.548 0.464 0.356 0.356 0.356 0.356
units: f
  The stock numbers are calculated using the recruitment and the
future mortality
stock.n(ple4_sq)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
                                   2009
age 2006
               2007
                         2008
    820006.3 949341.2 844041.2 1039494.2
    554592.5 531914.3 784234.7 634556.4
 2
     422068.8 269655.1 267892.4 458861.8
  3
```

```
4
      88052.9 212889.2 142573.0
                                   166055.3
 5
     136368.1
               45514.9 137660.4
                                    86954.4
                                    94351.2
 6
      17534.5
                80086.2
                          25575.8
  7
      16289.7
                 9240.6
                          49356.8
                                    17228.2
  8
       4114.0
                 9297.5
                           5828.1
                                    29788.4
 9
       2688.5
                 1585.1
                           5078.7
                                     3558.4
  10
       2854.5
                 3660.1
                           6750.3
                                     9212.2
   year
age 2010
              2011
  1 1039494.2 1039494.2
     802562.9 802562.9
 2
     374263.6 473354.4
  4
     280909.2 229119.3
  5
     107361.5 181619.3
     57762.5
  6
               71318.6
 7
      62135.0
               38039.5
 8
      11325.1 40844.9
 9
      17078.0
                 6492.8
  10
       8330.8
               16575.1
```

units: NA

Note that the future harvest slot is different from the harvest from the one we set up, but that the selection pattern is the same: F at age has been multiplied by the same value to give us the target Fbar value

```
round(harvest(ple4_stf)[, ac((maxyr_stf - 2):maxyr_stf)],
   3)
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
   year
age 2009 2010 2011
 1 0.203 0.203 0.203
 2 0.548 0.548 0.548
 3 0.500 0.500 0.500
  4 0.430 0.430 0.430
 5 0.395 0.395 0.395
 6 0.407 0.407 0.407
 7 0.409 0.409 0.409
 8 0.584 0.584 0.584
 9 0.419 0.419 0.419
```

```
10 0.419 0.419 0.419
units: f
round(harvest(ple4_sq)[, ac((maxyr_stf - 2):maxyr_stf)],
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age 2009 2010 2011
 1 0.159 0.159 0.159
 2 0.428 0.428 0.428
 3 0.391 0.391 0.391
 4 0.336 0.336 0.336
  5 0.309 0.309 0.309
 6 0.318 0.318 0.318
 7 0.320 0.320 0.320
 8 0.456 0.456 0.456
 9 0.327 0.327 0.327
  10 0.327 0.327 0.327
units: f
harvest(ple4_stf)[, ac((maxyr_stf - 2):maxyr_stf)]/harvest(ple4_sq)[,
    ac((maxyr_stf - 2):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
    year
age 2009
          2010 2011
 1 1.2797 1.2797 1.2797
 2 1.2797 1.2797 1.2797
  3 1.2797 1.2797 1.2797
  4 1.2797 1.2797 1.2797
  5 1.2797 1.2797 1.2797
  6 1.2797 1.2797 1.2797
 7 1.2797 1.2797 1.2797
  8 1.2797 1.2797 1.2797
  9 1.2797 1.2797 1.2797
```

10 1.2797 1.2797 1.2797

units: f

The catch numbers come from the predicted abundance and harvest rates using the Baranov equation. The catches are then split into landings and discards using the ratios computed by stf().

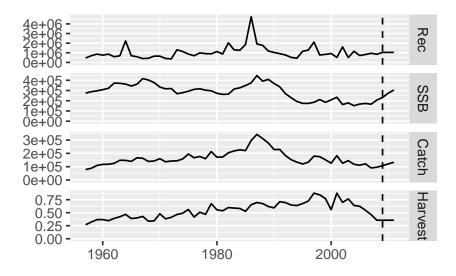
```
landings.n(ple4_sq)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
   year
age 2006
             2007
                      2008
                               2009
                        380.00 1006.12
 1
      355.00 1286.00
  2 18987.00 19205.00 10970.00 14418.91
  3 67465.00 37309.00 42865.00 59684.58
    25254.00 47053.00 37970.00 35832.72
  5 42525.00 14971.00 29476.00 20398.74
     6555.00 17142.00 5700.00 20939.33
     4967.00 2459.00 6752.00 3332.27
  8
     2053.00 1856.00
                       912.00 6773.26
     1235.00
              543.00
                        673.00
                                947.53
  10 1319.00 1259.00
                        896.00 2453.01
   year
age 2010
             2011
     1006.12 1006.12
 2 18236.48 18236.48
 3 48680.81 61569.64
  4 60616.80 49441.17
  5 25186.06 42606.28
 6 12819.21 15827.72
 7 12018.13 7357.58
     2575.09 9287.29
     4547.50 1728.89
  10 2218.31 4413.60
units: NA
discards.n(ple4_sq)[, ac((maxyr_stf - 5):maxyr_stf)]
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
   year
```

```
2007
                        2008
age
     2006
                                 2009
     220473.0 77312.0 135406.0 144332.6
  1
     225077.0 205140.0 252629.0 196572.8
              69312.0
  3
     110215.0
                       37393.0
                                  82004.2
  4
      10656.0
               10735.0
                          6237.0
                                   9402.9
  5
       3000.0
                1437.0
                          2282.0
                                   1656.6
                                   3564.4
  6
        410.0
                7151.0
                           517.0
  7
        754.0
                 204.0
                          8882.0
                                   1163.8
  8
        194.0
                1649.0
                           891.0
                                   3652.4
  9
          0.0
                    0.0
                             0.0
                                      0.0
  10
          0.0
                    0.0
                             0.0
                                      0.0
    year
age 2010
              2011
    144332.6 144332.6
  2
     248617.9 248617.9
  3
      66885.5 84594.2
  4
      15906.5
               12973.9
  5
       2045.4
                3460.1
  6
       2182.1
                2694.3
  7
       4197.2
                2569.5
  8
       1388.6
                5008.1
  9
                    0.0
          0.0
  10
          0.0
                    0.0
```

units: NA

We can see the projection using a plot here

```
plot(ple4_sq) + geom_vline(lty = 2, xintercept = an(ISOdate(maxyr_stk +
    1, 1, 1)))
```



Short term forecast with many F scenarios

Typically when running STF you explore several different future F scenarios. The scenarios are based on 'F status quo', which we calculated above as the mean F of the last X years.

For a 3 year STF the F pattern is:

- year 1: fbar_status_quo
- year 2: fbar_status_quo * fbar_multiplier
- year 3: fbar_status_quo * fbar_multiplier

Note that the fbar_multiplier is the same for years 2 and 3 We are going to run several STFs with different values for the fbar_multiplier

```
fbar_multiplier <- seq(from = 0, to = 2, by = 0.2)
```

Next we are going to build a data.frame that builds these scenarios. Each column in the dataframe is a year, each row is a scenario. Note that if you project for more than 3 years you will need to add more columns / years to the matrix

```
fbar_scenarios <- cbind(rep(fbar_SQ, length(fbar_multiplier)),</pre>
    fbar_multiplier * fbar_SQ, fbar_multiplier *
        fbar_SQ)
```

Another scenario we are interested in is $F_{0.1}$. We can calculate this using FLBRP (or maybe you already have a value).

```
f01 <- c(refpts(brp(FLBRP(ple4)))["f0.1", "harvest"])</pre>
# Add the F0.1 scenario as a final scenario
fbar_scenarios <- rbind(fbar_scenarios, c(fbar_SQ,</pre>
    f01, f01))
# Add some names
colnames(fbar_scenarios) <- c("2009", "2010",</pre>
rownames(fbar_scenarios) <- c(fbar_multiplier,</pre>
    "f01")
fbar_scenarios
```

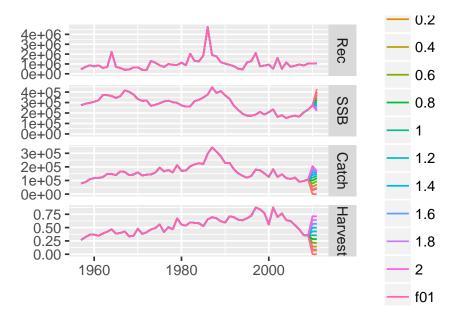
	2009	2010	2011
O	0.3563	0.0000	0.0000
0.2	0.3563	0.0713	0.0713
0.4	0.3563	0.1425	0.1425

	2009	2010	2011
0.6	0.3563	0.2138	0.2138
0.8	0.3563	0.2850	0.2850
1	0.3563	0.3563	0.3563
1.2	0.3563	0.4276	0.4276
1.4	0.3563	0.4988	0.4988
1.6	0.3563	0.5701	0.5701
1.8	0.3563	0.6414	0.6414
2	0.3563	0.7126	0.7126
fo1	0.3563	0.0876	0.0876

There are various results we want to extract from the STF, like predicted Catch, SSB and the relative change in these. First make an empty matrix in which to store the results

```
stf_results <- matrix(NA, nrow = nrow(fbar_scenarios),</pre>
    ncol = 8)
# Set some column names
colnames(stf_results) <- c("Fbar", paste0("Catch",</pre>
    maxyr_stk + 1), paste0("Catch", maxyr_stk +
    2), paste0("Catch", maxyr_stk + 3), paste0("SSB",
    maxyr_stk + 2), paste0("SSB", maxyr_stk +
    3), paste0("SSB_change_", maxyr_stk + 2, "-",
    maxyr_stk + 3, "(%)"), paste0("Catch_change_",
    maxyr_stk, "-", maxyr_stk + 2, "(%)"))
# Store the resulting FLStock each time
stk_stf <- FLStocks()</pre>
# Loop over the scenarios (each row in the
# fbar_scenarios table)
for (scenario in 1:nrow(fbar_scenarios)) {
    cat("Scenario: ", scenario, "\n")
    # Make a target object with F values for that
    # scenario
    ctrl_target <- data.frame(year = (maxyr_stf -</pre>
        2):maxyr_stf, quantity = "f", val = fbar_scenarios[scenario,
    # Set the control object - year, quantity and
    # value for the moment
    ctrl_f <- fwdControl(ctrl_target)</pre>
    # ctrl_target Run the forward projection. We
    # include an additional argument, maxF. By
    # default the value of maxF is 2.0 Here we
    # increase it to 10.0 so that F is not limited
```

```
ple4_fwd <- fwd(ple4_stf, ctrl = ctrl_f, sr = ple4_sr) #, maxF = 10.0)
    ## Check it has worked - uncomment out to check
    ## scenario by scenario
    ## plot(ple4_fwd[,ac(2001:2011)]) Store the
    ## result - if you want to, comment out if
    ## unnecessary
    stk_stf[[as.character(scenario)]] <- ple4_fwd</pre>
    # Fill results table
    stf_results[scenario, 1] <- round(fbar(ple4_fwd)[,</pre>
        ac(2011)], 3) # final stf year
    stf_results[scenario, 2] <- catch(ple4_fwd)[,</pre>
        ac(maxyr_stk + 1)] # 1st stf year
    stf_results[scenario, 3] <- catch(ple4_fwd)[,</pre>
        ac(maxyr_stk + 2)] # 2nd stf year
    stf_results[scenario, 4] <- catch(ple4_fwd)[,</pre>
        ac(maxyr_stk + 3)] # final stf year
    stf_results[scenario, 5] <- ssb(ple4_fwd)[,</pre>
        ac(maxyr_stk + 2)] # 2nd stf year
    stf_results[scenario, 6] <- ssb(ple4_fwd)[,</pre>
        ac(maxyr_stk + 3)] # final stf year
    # change in ssb in last two stf years
    stf_results[scenario, 7] <- round((ssb(ple4_fwd)[,</pre>
        ac(maxyr_stk + 3)] - ssb(ple4_fwd)[, ac(maxyr_stk +
        2)])/ssb(ple4_fwd)[, ac(maxyr_stk + 2)] *
        100, 1)
    # change in catch from true year, to 2nd to
    # last stf year
    stf_results[scenario, 8] <- round((catch(ple4_fwd)[,</pre>
        ac(maxyr_stk + 2)] - catch(ple4_fwd)[,
        ac(maxyr_stk)])/catch(ple4_fwd)[, ac(maxyr_stk)] *
        100, 1)
# Give the FLStocks object some names
names(stk_stf) <- rownames(fbar_scenarios)</pre>
# Plotting
plot(stk_stf)
```



The different scenarios are plotted above. The table of results is given below.

stf_results

Fbar	Catch2009	Catch2010	Catch2011	SSB2010	SSB2011	SSB_change_2010-2011(%)	Catch_change_2008-2010
0.000	107543	0	0	274322	427478	55.8	-10
0.071	107543	27670	39016	274322	399470	45.6	
0.143	107543	53457	70819	274322	373394	36.1	-,
0.214	107543	77495	96520	274322	349112	27.3	
0.285	107543	99907	117069	274322	326496	19.0	
0.356	107543	120809	133278	274322	305429	11.3	:
0.428	107543	140308	145840	274322	285800	4.2	
0.499	107543	158502	155345	274322	267507	-2.5	
0.570	107543	175484	162298	274322	250458	-8.7	
0.641	107543	191338	167127	274322	234562	-14.5	9
0.713	107543	206143	170197	274322	219741	-19.9	1
0.088	107543	33744	46903	274322	393326	43.4	-(

More information

- You can submit bug reports, questions or suggestions on this tutorial at https://github.com/flr/doc/issues.
- Or send a pull request to https://github.com/flr/doc/
- For more information on the FLR Project for Quantitative Fisheries Science in R, visit the FLR webpage, http://flr-project.org.

Software Versions

• R version 3.3.2 (2016-10-31)

• FLCore: 2.6.0.20170228 • FLAssess: 2.5.20150717

• FLash: 2.5.7

• ggplotFL: 2.5.9.9000

• ggplot2: 2.2.1

• FLBRP: 2.5.20170228

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