

Short Term Forecasting for advice using FLash

13 September, 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 regarding what will happen in the future,
- a stock-recruitment relationship,
- a projection control specifying what the targets are 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 the 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.

```
# Load the required packages
library(FLAssess)
library(FLash)
library(ggplotFL)
library(FLBRP)
# Load the data
data(ple4)
summary(ple4)
```

An object of class "FLStock"

Name: Plaice in IV

Description: Imported from a VPA file. (N:\Projecten\ICES WG\Demersale werkgroep [...]

Quant: age

Dims:	age	year	unit	season	area	iter
	10	52	1	1	1	1

Range:	min	max	pgroup	minyear	maxyear	minfbar	maxfbar
	1	10	10	1957	2008	2	6

catch	:	[1 52 1 1 1 1]	, units =	t
catch.n	:	[10 52 1 1 1 1]	, units =	10 ³
catch.wt	:	[10 52 1 1 1 1]	, units =	kg
discards	:	[1 52 1 1 1 1]	, units =	t
discards.n	:	[10 52 1 1 1 1]	, units =	10 ³
discards.wt	:	[10 52 1 1 1 1]	, units =	kg
landings	:	[1 52 1 1 1 1]	, units =	t

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

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 about these will affect the outcome of the projection. Rather than using `window()` or `trim()` that make all future data *NA*, we use the `stf()` function (short term forecast) that has several options that allow one 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, `wt.s.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 short term forecasts, i.e. the future weights at age will be the same as the average of the last 3 years. This assumes that what is going to happen in the next few years will be 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:

```
maxyr_stk <- range(ple4)[["maxyear"]]
ple4_stf <- stf(ple4,nyears=3,wt.s.nyears=3, na.rm=TRUE)
maxyr_stf <- range(ple4_stf)[["maxyear"]]
```

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

```
range(ple4_stf)
```

min	max	plusgroup	minyear	maxyear	minfbar	maxfbar
1	10	10	1957	2011	2	6

```
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	2010	2011
1	0.053000	0.048000	0.050000	0.050333	0.050333	0.050333
2	0.129000	0.093000	0.114000	0.112000	0.112000	0.112000
3	0.195000	0.239000	0.200000	0.211333	0.211333	0.211333
4	0.321000	0.241000	0.278000	0.280000	0.280000	0.280000
5	0.354000	0.337000	0.355000	0.348667	0.348667	0.348667

```

6  0.424000 0.394000 0.429000 0.415667 0.415667 0.415667
7  0.439000 0.458000 0.484000 0.460333 0.460333 0.460333
8  0.506000 0.412000 0.627000 0.515000 0.515000 0.515000
9  0.583000 0.526000 0.598000 0.569000 0.569000 0.569000
10 0.730673 0.548489 0.730672 0.669945 0.669945 0.669945

```

units: kg

For maturity the same assumption (of the future being the same as the average of 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

```

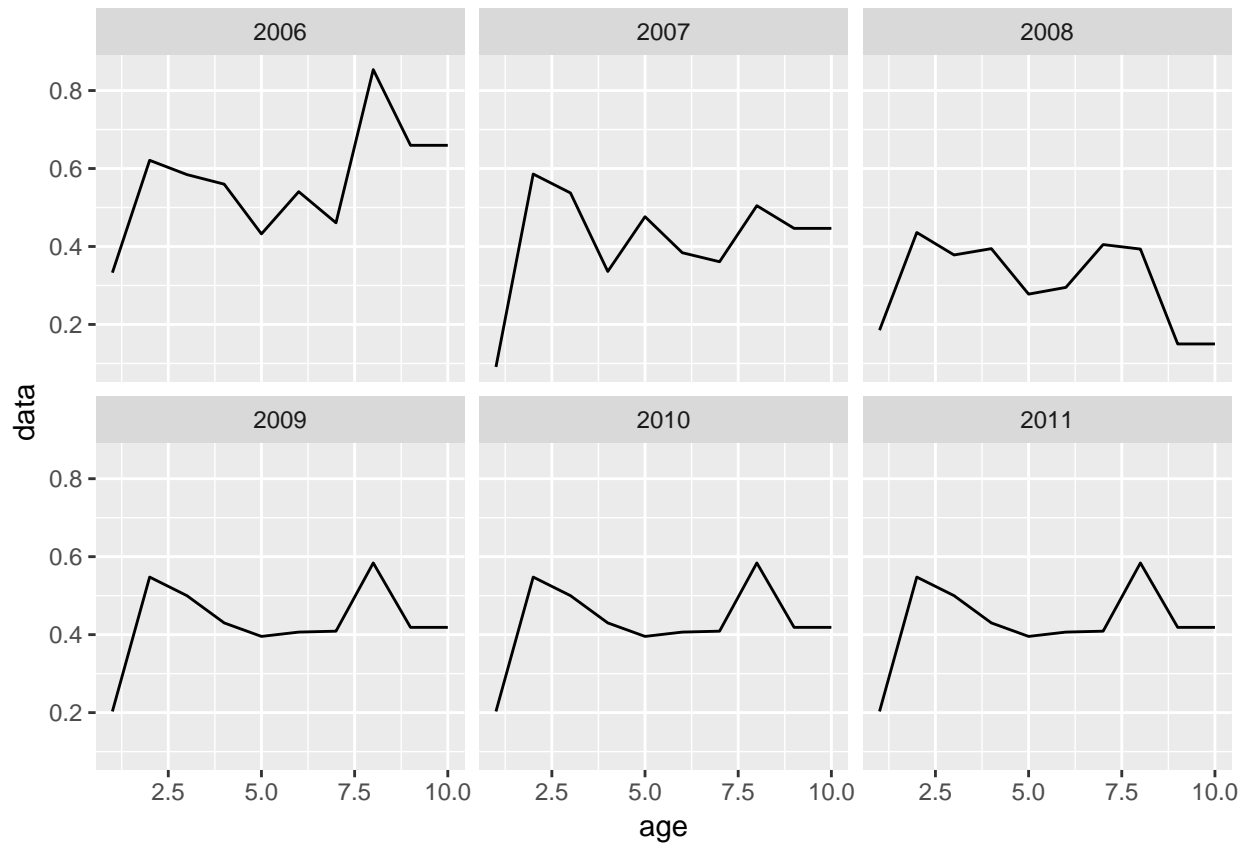
      year
age 2006 2007 2008 2009 2010 2011
1  0.0  0.0  0.0  0.0  0.0  0.0
2  0.5  0.5  0.5  0.5  0.5  0.5
3  0.5  0.5  0.5  0.5  0.5  0.5
4  1.0  1.0  1.0  1.0  1.0  1.0
5  1.0  1.0  1.0  1.0  1.0  1.0
6  1.0  1.0  1.0  1.0  1.0  1.0
7  1.0  1.0  1.0  1.0  1.0  1.0
8  1.0  1.0  1.0  1.0  1.0  1.0
9  1.0  1.0  1.0  1.0  1.0  1.0
10 1.0  1.0  1.0  1.0  1.0  1.0

```

units:

Notice that the future fishing mortality has also been set (average 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(~y
```



The stock numbers at age and the catch numbers at age are not forecast 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	2010	2011
1	820006.3	949341.2	844041.2	NA	NA	NA
2	554592.5	531914.3	784234.7	NA	NA	NA
3	422068.8	269655.1	267892.4	NA	NA	NA
4	88052.9	212889.2	142573.0	NA	NA	NA
5	136368.1	45514.9	137660.4	NA	NA	NA
6	17534.5	80086.2	25575.8	NA	NA	NA
7	16289.7	9240.6	49356.8	NA	NA	NA
8	4114.0	9297.5	5828.1	NA	NA	NA
9	2688.5	1585.1	5078.7	NA	NA	NA
10	2854.5	3660.1	6750.3	NA	NA	NA

units: 10^3

Meanwhile, the landings and discards are average forecast ratios (proportion of total catch) of what is discarded and what is landed over the last 3 years of data.

```
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					
age	2006	2007	2008	2009	2010	2011
1	2.2047e+05	7.7312e+04	1.3541e+05	9.9308e-01	9.9308e-01	9.9308e-01
2	2.2508e+05	2.0514e+05	2.5263e+05	9.3166e-01	9.3166e-01	9.3166e-01
3	1.1022e+05	6.9312e+04	3.7393e+04	5.7876e-01	5.7876e-01	5.7876e-01
4	1.0656e+04	1.0735e+04	6.2370e+03	2.0786e-01	2.0786e-01	2.0786e-01
5	3.0000e+03	1.4370e+03	2.2820e+03	7.5111e-02	7.5111e-02	7.5111e-02
6	4.1000e+02	7.1510e+03	5.1700e+02	1.4546e-01	1.4546e-01	1.4546e-01
7	7.5400e+02	2.0400e+02	8.8820e+03	2.5884e-01	2.5884e-01	2.5884e-01
8	1.9400e+02	1.6490e+03	8.9100e+02	3.5033e-01	3.5033e-01	3.5033e-01
9	0.0000e+00	0.0000e+00	0.0000e+00	0.0000e+00	0.0000e+00	0.0000e+00
10	0.0000e+00	0.0000e+00	0.0000e+00	0.0000e+00	0.0000e+00	0.0000e+00

```
units: 10^3
```

```
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					
age	2006	2007	2008	2009	2010	2011
1	3.5500e+02	1.2860e+03	3.8000e+02	6.9226e-03	6.9226e-03	6.9226e-03
2	1.8987e+04	1.9205e+04	1.0970e+04	6.8339e-02	6.8339e-02	6.8339e-02
3	6.7465e+04	3.7309e+04	4.2865e+04	4.2124e-01	4.2124e-01	4.2124e-01
4	2.5254e+04	4.7053e+04	3.7970e+04	7.9214e-01	7.9214e-01	7.9214e-01
5	4.2525e+04	1.4971e+04	2.9476e+04	9.2489e-01	9.2489e-01	9.2489e-01
6	6.5550e+03	1.7142e+04	5.7000e+03	8.5454e-01	8.5454e-01	8.5454e-01
7	4.9670e+03	2.4590e+03	6.7520e+03	7.4116e-01	7.4116e-01	7.4116e-01
8	2.0530e+03	1.8560e+03	9.1200e+02	6.4967e-01	6.4967e-01	6.4967e-01
9	1.2350e+03	5.4300e+02	6.7300e+02	1.0000e+00	1.0000e+00	1.0000e+00
10	1.3190e+03	1.2590e+03	8.9600e+02	1.0000e+00	1.0000e+00	1.0000e+00

```
units: 10^3
```

```
# Compare above landings.n for the forecast years with
yearMeans((landings.n(ple4)/(landings.n(ple4)+discards.n(ple4)))[,ac((maxyr_stk-2):maxyr_stk)])
```

```
An object of class "FLQuant"
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
```

	year
age	1
1	0.0069226
2	0.0683387
3	0.4212371
4	0.7921356
5	0.9248890
6	0.8545368

```

7 0.7411596
8 0.6496718
9 1.0000000
10 1.0000000

units: NA
# Furthermore, landings and discards proportions sum to 1
landings.n(ple4_stf)[,ac((maxyr_stf-2):maxyr_stf)] + discards.n(ple4_stf)[,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     1     1
2   1     1     1
3   1     1     1
4   1     1     1
5   1     1     1
6   1     1     1
7   1     1     1
8   1     1     1
9   1     1     1
10  1     1     1

units: 1000

```

The stock-recruitment relationship (SRR)

A short term forecast does not use an SRR (in the traditional sense). Instead, it generally assumes that recruitment in the future is some mean (e.g. geometric mean) of the historic recruitments. However, we still need to have an SRR that contains this mean value, which is what we mimic for this example. First, we estimate the geometric mean recruitment, that we then add to an SRR object

```

mean_rec <- exp(mean(log(rec(ple4))))
ple4_sr <- as.FLSR(ple4, model="geomean")
params(ple4_sr)['a',] <- mean_rec
params(ple4_sr)

```

```

An object of class "FLPar"
params
  a
9e+05
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 short term forecast 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 depending 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"

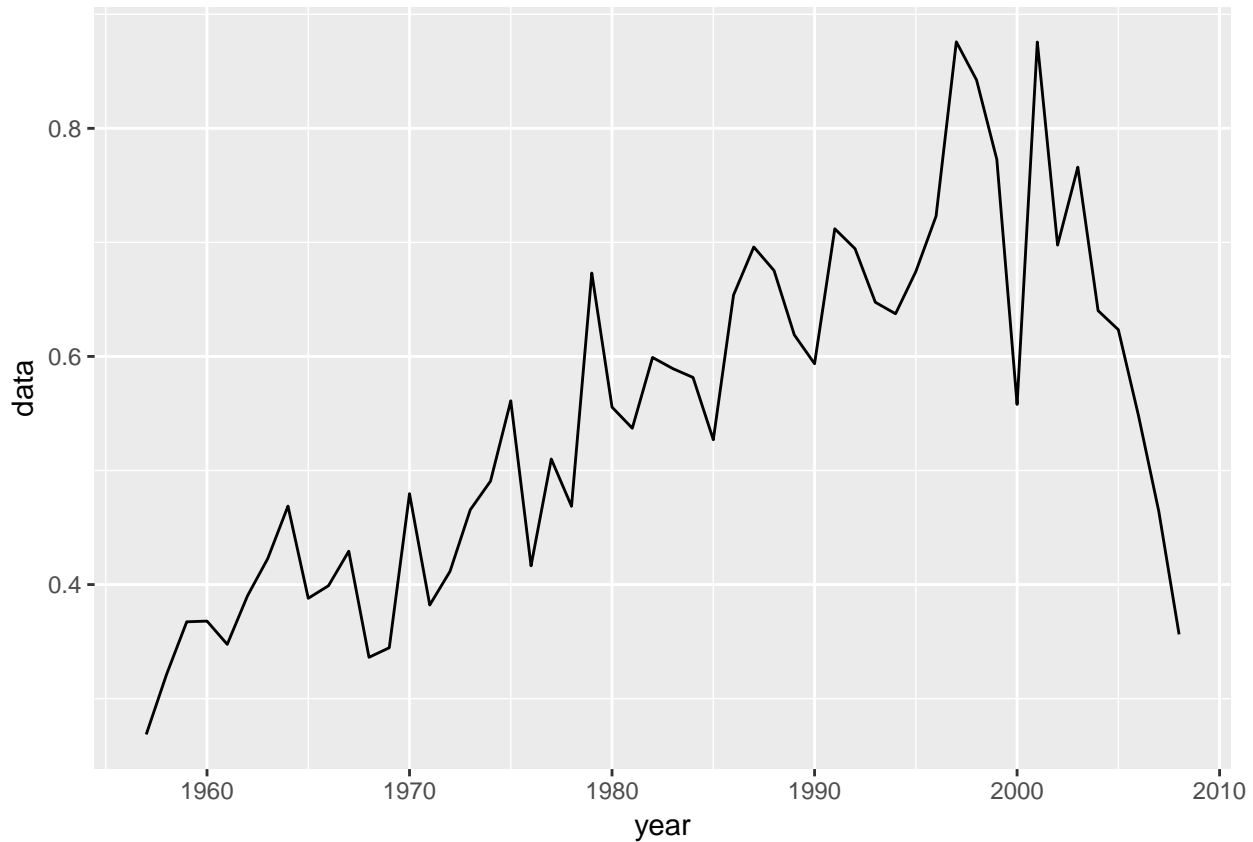
```
, , unit = unique, season = all, area = unique
```

```
      year
age 1957 1958 1959 1960 1961
all 0.269 0.321 0.367 0.368 0.348
```

```
[ ... 42 years]
```

```
      year
age 2004 2005 2006 2007 2008
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)])
```

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 medium term forecast tutorial)
- 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

```
# Set the control object - year, quantity and value for the moment
ctrl_target <- data.frame(year = 2009:2011, quantity = "f", val = fbar_SQ)
ctrl_f <- fwdControl(ctrl_target)
ctrl_f
```

Target

	year	quantity	min	val	max
1	2009	f	NA	0.3563	NA
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 here for the short term forecast examples.

Running the STF

Below we run a simple short term forecast (STF) with 'status quo' future fishing mortality. Remember we had to make assumptions about the future (weights, fishing mortality 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 update has forecast stock numbers, based on the recruitment, fishing mortality, and natural mortality assumptions. Because we now have stock numbers, we can calculate forecast ssb.

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

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

```
mean_rec
```

```
[1] 900370
```

```
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 2011
1	820006 949341 844041 900370 900370 900370

```
units: NA
```

Similarly, we check if the fbar in the forecast stock indeed corresponds to status quo fishing mortality.


```
round(fbar_SQ,3)
```

```
[1] 0.356
```

```
round(fbar(ple4_sq)[,ac((maxyr_stf-5):maxyr_stf)],3)
```

```
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 future mortality assumptions.

```
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
age 2006 2007 2008 2009 2010 2011
1 820006.3 949341.2 844041.2 900369.9 900369.9 900369.9
2 554592.5 531914.3 784234.7 634556.4 695149.1 695149.1
3 422068.8 269655.1 267892.4 458861.8 374263.6 410001.4
4 88052.9 212889.2 142573.0 166055.3 280909.2 229119.3
5 136368.1 45514.9 137660.4 86954.4 107361.5 181619.3
6 17534.5 80086.2 25575.8 94351.2 57762.5 71318.6
7 16289.7 9240.6 49356.8 17228.2 62135.0 38039.5
8 4114.0 9297.5 5828.1 29788.4 11325.1 40844.9
9 2688.5 1585.1 5078.7 3558.4 17078.0 6492.8
10 2854.5 3660.1 6750.3 9212.2 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 a constant 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)],3)

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:  NA

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 average 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	2010	2011
1	355.00	1286.00	380.00	871.47	871.47	871.47
2	18987.00	19205.00	10970.00	14418.91	15795.74	15795.74
3	67465.00	37309.00	42865.00	59684.58	48680.81	53329.26
4	25254.00	47053.00	37970.00	35832.72	60616.80	49441.17
5	42525.00	14971.00	29476.00	20398.74	25186.06	42606.28
6	6555.00	17142.00	5700.00	20939.33	12819.21	15827.72
7	4967.00	2459.00	6752.00	3332.27	12018.13	7357.58
8	2053.00	1856.00	912.00	6773.26	2575.09	9287.29
9	1235.00	543.00	673.00	947.53	4547.50	1728.89
10	1319.00	1259.00	896.00	2453.01	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"

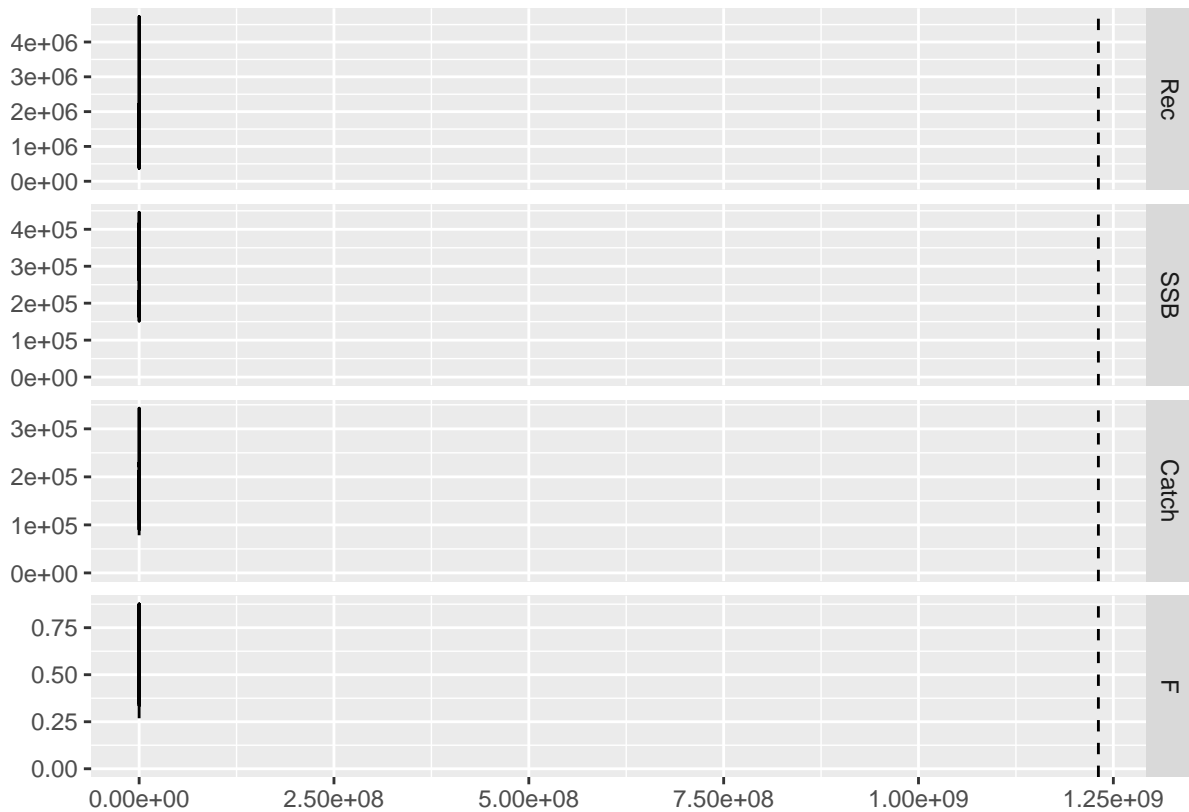
, , unit = unique, season = all, area = unique

	year					
age	2006	2007	2008	2009	2010	2011
1	220473.0	77312.0	135406.0	125015.3	125015.3	125015.3
2	225077.0	205140.0	252629.0	196572.8	215343.2	215343.2
3	110215.0	69312.0	37393.0	82004.2	66885.5	73272.3
4	10656.0	10735.0	6237.0	9402.9	15906.5	12973.9
5	3000.0	1437.0	2282.0	1656.6	2045.4	3460.1
6	410.0	7151.0	517.0	3564.4	2182.1	2694.3
7	754.0	204.0	8882.0	1163.8	4197.2	2569.5
8	194.0	1649.0	891.0	3652.4	1388.6	5008.1
9	0.0	0.0	0.0	0.0	0.0	0.0
10	0.0	0.0	0.0	0.0	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 year 1 is typically called the ‘intermediate year’ (in ICES, this would be the year the Expert Group meets to run the assessment, and status quo F for the intermediate year is a common assumption), and 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 dataframe that creates 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)),
                        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"])
# Add the F0.1 scenario as a final scenario
fbar_scenarios <- rbind(fbar_scenarios, c(fbar_SQ,f01,f01))

# Add some names
colnames(fbar_scenarios) <- c("2009","2010","2011")
rownames(fbar_scenarios) <- c(fbar_multiplier, "f01")
fbar_scenarios
```

	2009	2010	2011
0	0.3563	0.0000	0.0000
0.2	0.3563	0.0713	0.0713
0.4	0.3563	0.1425	0.1425
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
f01	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),ncol = 8)
# Set some column names
colnames(stf_results) <- c('Fbar',
  paste0('Catch',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,'(%)'))

# Set up an FLStocks object to store the resulting FLStock each time
stk_stf <- FLStocks()
# Loop over the scenarios (each row in the fbar_scenarios table)
for (scenario in 1:nrow(fbar_scenarios)) {
  cat("Scenario: ", scenario, "\n")
  flush.console()
  # Make a target object with F values for that scenario
  # Set the control object - year, quantity and value for the moment
  ctrl_target <- data.frame(year = (maxyr_stf-2):maxyr_stf,
    quantity = "f",
    val = fbar_scenarios[scenario,])

  # ctrl_target
  ctrl_f <- fwdControl(ctrl_target)
  # Run the forward projection. We could include an additional argument, maxF.
  # By default the value of maxF is 2.0. It could be increased to 10.0, say,
  # so that F is less limited, and the bound is not hit (not a problem here).
```

```

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

# Fill results table
stf_results[scenario,1] <- round(fbar(ple4_fwd)[,ac(2011)],3) # final stf year
stf_results[scenario,2] <- catch(ple4_fwd)[,ac(maxyr_stk+1)] # 1st stf year
stf_results[scenario,3] <- catch(ple4_fwd)[,ac(maxyr_stk+2)] # 2nd stf year
stf_results[scenario,4] <- catch(ple4_fwd)[,ac(maxyr_stk+3)] # final stf year
stf_results[scenario,5] <- ssb(ple4_fwd)[,ac(maxyr_stk+2)] # 2nd stf year
stf_results[scenario,6] <- ssb(ple4_fwd)[,ac(maxyr_stk+3)] # final stf year

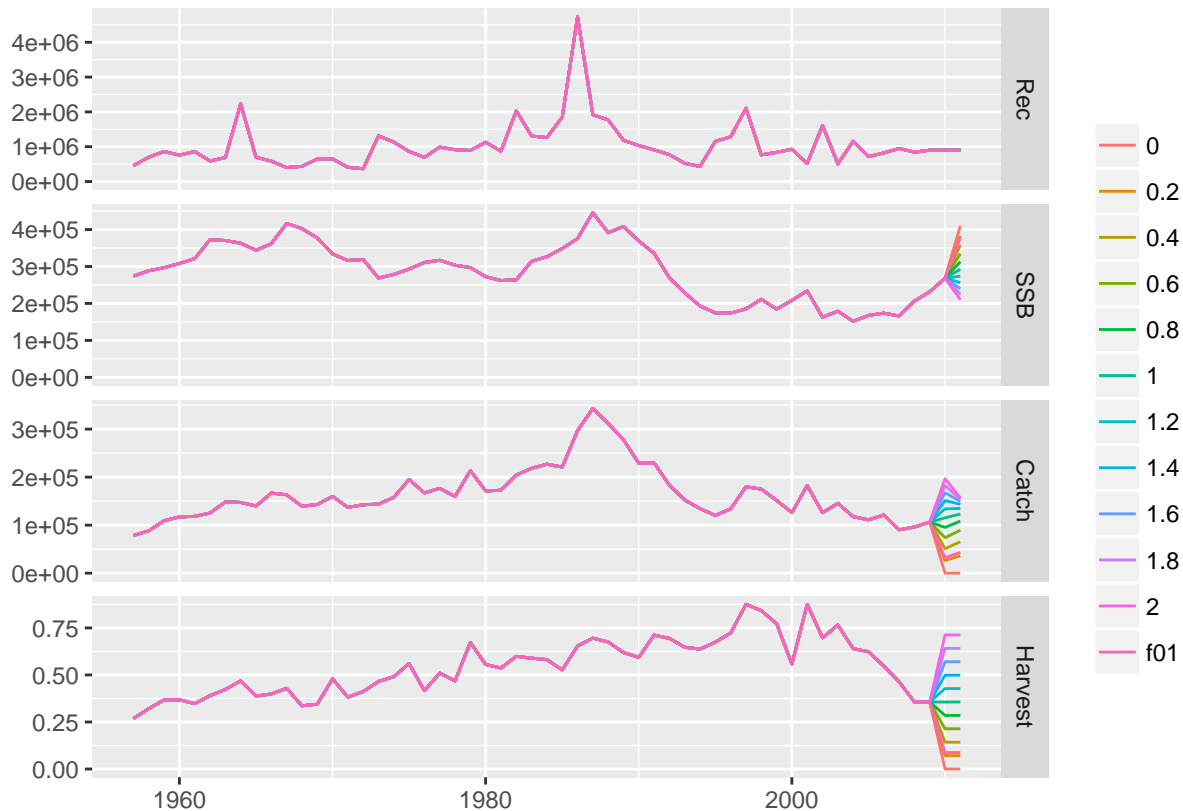
# change in ssb in last two stf years
stf_results[scenario,7] <- round((ssb(ple4_fwd)[,ac(maxyr_stk+3)]-ssb(ple4_fwd)[,ac(maxyr_stk+2)])/
                                ssb(ple4_fwd)[,ac(maxyr_stk+2)]*100,1)

# change in catch from true year, to 2nd to last stf year
stf_results[scenario,8] <- round((catch(ple4_fwd)[,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)

# 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_2008-2011(%)
0.000	106368	0	0	268307	410158	52.9	52.9
0.071	106368	26377	36153	268307	383213	42.8	42.8
0.143	106368	50963	65598	268307	358124	33.5	33.5
0.214	106368	73885	89367	268307	334758	24.8	24.8
0.285	106368	95259	108345	268307	312995	16.7	16.7
0.356	106368	115195	123285	268307	292719	9.1	9.1
0.428	106368	133794	134831	268307	273827	2.1	2.1
0.499	106368	151151	143535	268307	256221	-4.5	-4.5
0.570	106368	167352	149864	268307	239810	-10.6	-10.6
0.641	106368	182478	154218	268307	224511	-16.3	-16.3
0.713	106368	196604	156937	268307	210245	-21.6	-21.6
0.088	106368	32168	43457	268307	377302	40.6	40.6

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 3.4.1 (2017-06-30)
- FLCore: 2.6.5
- FLAssess: 2.6.1
- FLash: 2.5.8
- ggplotFL: 2.6.1
- ggplot2: 2.2.1
- FLBRP: 2.5.2
- **Compiled:** Wed Sep 13 16:40:11 2017

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