

Manipulating FLR objects with apply and sweep

Laurence Kell

<Laurie.Kell@iccat.int>

&

Ernesto Jardim

<ernesto@ipimar.pt>

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1 Introduction

R has many functions and methods for manipulating data objects such as arrays, data.frames and lists. Once **FLR** is based on object oriented (OO) programming where data (i.e. objects) and actions (i.e. methods) are grouped together in S4 classes, these methods can be used with **FLR**. For example an `FLQuant` is derived from an array and so can be manipulated using functions written for arrays, while an **FLR** class has similarities to a list (in that it can contain a variety of data types) and functions that work for lists have been overloaded so that they work for **FLR** classes.

In **R** there are a set of methods that replace “for loops” like `apply` and `sweep`. To the new user this can appear to be confusing but using them speeds up code, helps conceptually by moving towards a "whole object" view and makes the code more readable. The `apply` family of functions allow functions to be applied on subsets of different types of **R** classes (i.e. `lapply`, `tapply`, `sapply`, `rapply`, `mapply`, etc). `Sweep` is useful when trying to perform an operation on two arrays that might have different dimensions.

We show how these and related functions can be used within **FLR**. Firstly showing how **R** functions used for arrays can be used with the `FLQuant` class, we then describe additional features and functions that have been added to **FLR**. Following this we describe functions that have been added for the **FLR** classes themselves. Let's start by loading **FLCore** and some data sets distributed with it.

```
> library(FLCore)
> data(ple4)
> data(ple4sex)
> data(nsher)
```

2 FLQuant

2.1 Apply

The `apply` function in base **R** applies a function to the margins of an array and returns a vector or array or list of the values obtained, i.e.

```
apply(X, INDEX, FUN)
```

X the array to be used.

MARGIN a vector giving the subscripts which the function will be applied over. 1 indicates rows, 2 indicates columns, c(1, 2) indicates rows and columns. For an FLQuant 1, 2, 3, 4, 5, 6 indicate age, year, unit, season, area and iter.

FUN the function to be applied. In the case of functions like +, %, etc., the function name must be quoted.

. . . optional arguments to FUN.

A simple example is the computation of means and sums over one or more dimensions of the FLQuant matrix.

```

> apply(catch.n(ple4), 1, mean)
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

      year
age  1
1  172759.0
2  286083.4
3  167406.7
4   89902.2
5   46690.4
6   22864.2
7   12323.7
8    6252.9
9    3824.0
10   8922.3

units:  thousands
> apply(catch.n(ple4), 2, sum)
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

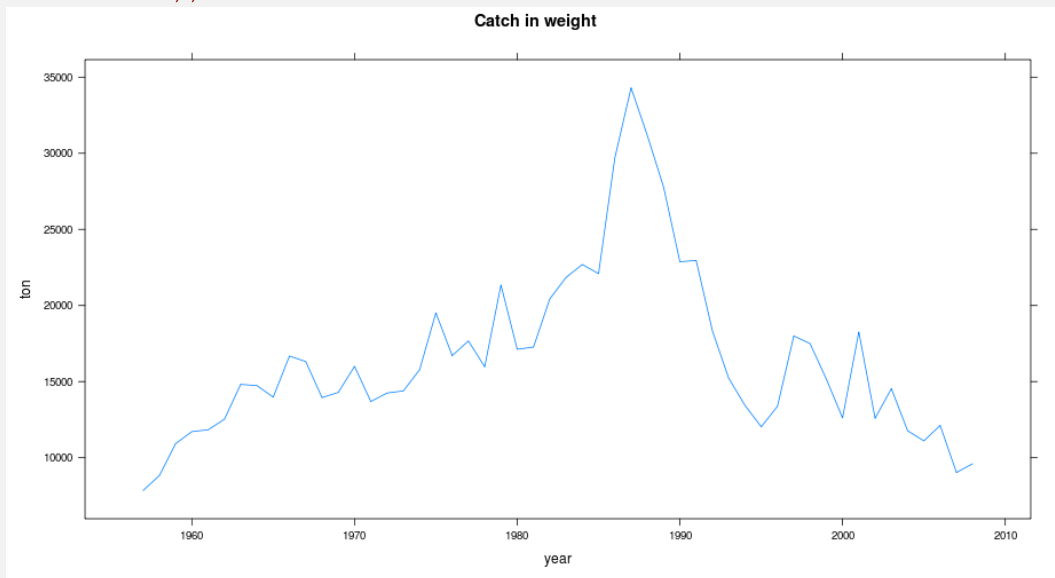
      year
age 1957  1958  1959  1960  1961  1962  1963  1964  1965
all 274463 348703 486073 500309 505585 554529 601358 578332 761373
      year
age 1966  1967  1968  1969  1970  1971  1972  1973  1974
all 800262 649376 508329 466118 614984 442137 419151 458236 749838
      year
age 1975  1976  1977  1978  1979  1980  1981  1982  1983
all 931012 679285 769163 704027 940585 700393 687764 1023162 1132551
      year
age 1984  1985  1986  1987  1988  1989  1990  1991  1992
all 1166447 1142931 2143204 2354937 1922494 1468754 1075597 1097352 850222
      year
age 1993  1994  1995  1996  1997  1998  1999  2000  2001
all 655108 515380 497708 697998 1061143 1187466 868346 648722 947864
      year
age 2002  2003  2004  2005  2006  2007  2008
all 789193 873227 733308 662652 741494 516023 580831

units:  thousands

```

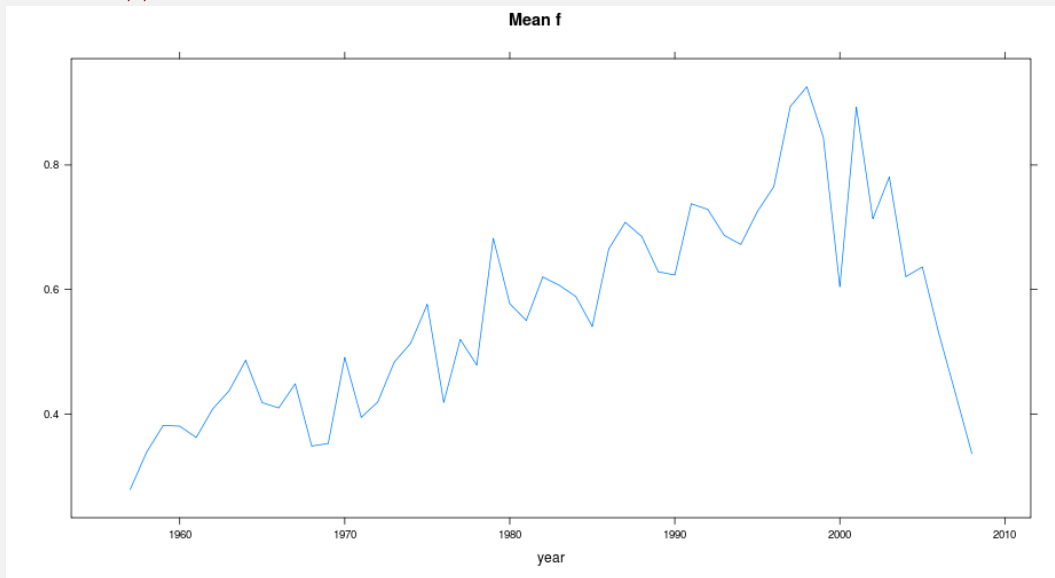
The function is particularly useful for operations that require combining the dims of an `FLQuant`, e.g. calculating trends over time or finding the average values at age. For example one can compute the catch in weight with

```
> flq <- apply(catch.wt(ple4) * catch.n(ple4), 2, mean)
> print(xyplot(data ~ year, data = flq, type = "l", main = "Catch in weight",
+           ylab = "ton"))
```



or mean fishing mortality with

```
> flq <- apply(harvest(ple4)[as.character(3:6)], 2, mean)
> print(xyplot(data ~ year, data = flq, type = "l", main = "Mean f",
+           ylab = ""))
```

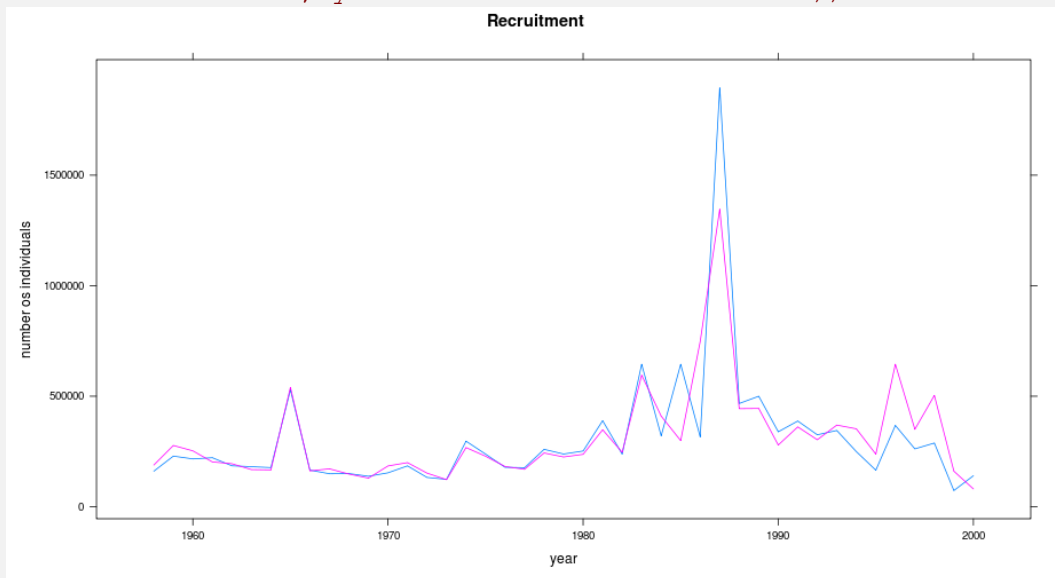


or recruitment by sex

```

> flq <- apply(rec(ple4sex), c(2, 3), sum)
> print(xyplot(data ~ year, group = unit, data = flq, type = "l",
+   main = "Recruitment", ylab = "number os individuals"))

```

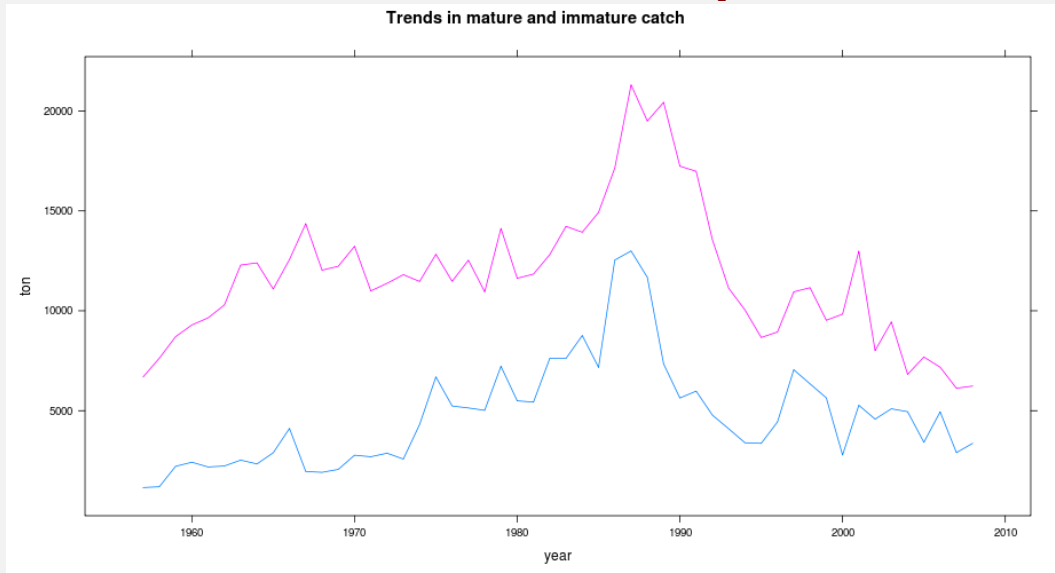


or look at trends in mature and immature catch

```

> matCatch <- apply(catch.wt(ple4) * catch.n(ple4) * mat(ple4),
+ 2, mean)
> immCatch <- apply(catch.wt(ple4) * catch.n(ple4) * (1 - mat(ple4)),
+ 2, mean)
> flqs <- FLQuants(imm = immCatch, mat = matCatch)
> print(xyplot(data ~ year, groups = qname, data = flqs, type = "l",
+ main = "Trends in mature and immature catch", ylab = "ton"))

```



2.2 The apply family

To simplify some common operations with `apply` and make the code more readable, a set of wrappers are implemented in **FLR** to compute sums, totals, means and vars for each `FLQuant` dimension. These are named by the combination of the dimension and the computation, *e.g.*, to compute the sum over an `FLQuant` dimension use `xxxSums` (i.e. `quantSums`, `yearSums`, ...). `xxxSums` “condenses” the `xxx` dimension by summing up across it, i.e. for `FLQuant` objects with just age and year.

```

> flq1 <- apply(catch.n(ple4), 1, sum)
> flq2 <- yearSums(catch.n(ple4))
> all.equal(flq1, flq2)
[1] TRUE
> flq1 <- apply(catch.n(ple4), 2, sum)
> flq2 <- quantSums(catch.n(ple4))
> all.equal(flq1, flq2)
[1] TRUE

```

Where there are other dimensions using `apply` becomes more difficult


```

> flq1 <- apply(catch.n(ple4sex), c(1:2, 4:6), sum)
> flq2 <- unitSums(catch.n(ple4sex))
> all.equal(flq1, flq2)
[1] TRUE

```

xxxTotals is similar but rather than “condensing” a dimension it replicates the sum across it, making it easy to calculate proportions for example. These methods are only implemented for quant and year.

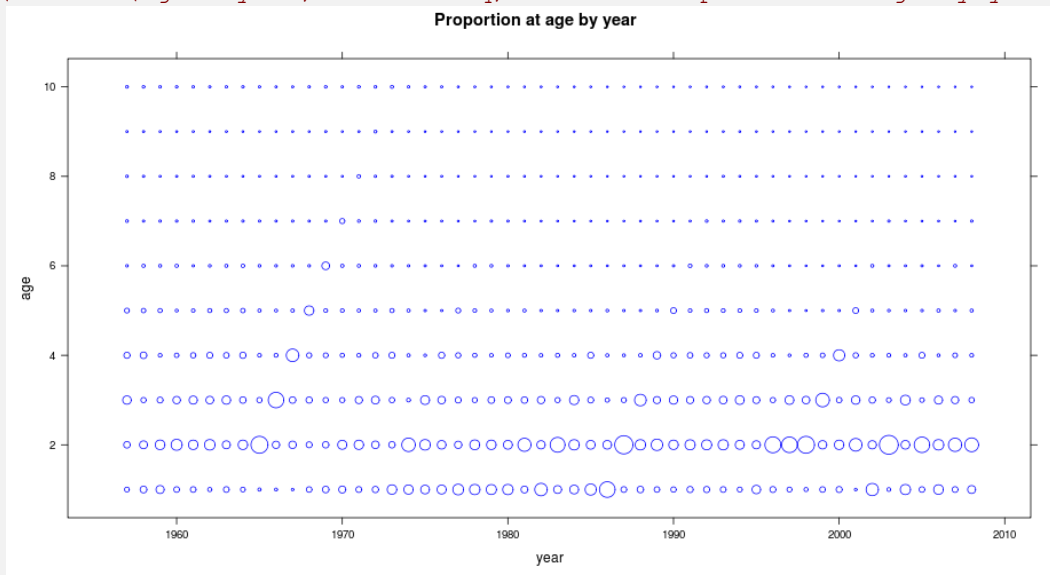
```

> yearTotals(catch.n(ple4))[, as.character(2004:2008)]
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

  year
age 2004  2005  2006  2007  2008
 1 733308 662652 741494 516023 580831
 2 733308 662652 741494 516023 580831
 3 733308 662652 741494 516023 580831
 4 733308 662652 741494 516023 580831
 5 733308 662652 741494 516023 580831
 6 733308 662652 741494 516023 580831
 7 733308 662652 741494 516023 580831
 8 733308 662652 741494 516023 580831
 9 733308 662652 741494 516023 580831
10 733308 662652 741494 516023 580831

units: thousands
> flq <- catch.n(ple4)/yearTotals(catch.n(ple4))
> print(bubbles(age ~ year, data = flq, main = "Proportion at age by year"))

```



xxxMeans and xxxVars are useful summary methods, although the names may be confusing since to calculate the mean by age you have to use yearMeans.

```

> yearVars(catch.wt(ple4sex))
An object of class "FLQuant"
, , unit = male, season = all, area = unique

      year
age 1
1  0.01401351
2  0.00098992
3  0.00108040
4  0.00172071
5  0.00262924
6  0.00304352
7  0.00325781
8  0.00526777
9  0.00407896
10 0.00648250
11 0.00673112
12 0.00754950
13 0.02284452
14 0.03602802
15 0.02287087

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

      year
age 1
1  0.01378140
2  0.00201949
3  0.00239446
4  0.00321595
5  0.00434408
6  0.00470233
7  0.00458162
8  0.00310044
9  0.00848206
10 0.00648272
11 0.00924828
12 0.00920520
13 0.01532536
14 0.02941595
15 0.06100897

units:  NA

```

Combining these methods make it simple to compute other statistics like the coefficient of variation

```

> sqrt(yearVars(catch.wt(ple4sex)))/yearMeans(catch.wt(ple4sex))
An object of class "FLQuant"
, , unit = male, season = all, area = unique

      year
age 1
1  1.524037
2  0.129218
3  0.120186
4  0.131541
5  0.140081
6  0.132601
7  0.125734
8  0.149398
9  0.125572
10 0.148333
11 0.147301
12 0.147094
13 0.245902
14 0.298204
15 0.224549

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

      year
age 1
1  1.515901
2  0.159042
3  0.151813
4  0.147431
5  0.136782
6  0.116272
7  0.100123
8  0.073513
9  0.110405
10 0.088979
11 0.096753
12 0.091153
13 0.110257
14 0.143566
15 0.191992

units:  NA NA

```

2.3 Sweep

`sweep` return an array/FLQuant derived from an input array/FLQuant by sweeping out a summary statistic

```
sweep(x, MARGIN, STATS, FUN="-", check.margin=TRUE, ...)
```

`x` an array of FLQuant

`MARGIN` a vector of indices giving the extents of `x` which correspond to `STATS`.

`STATS` the summary statistic which is to be swept out.

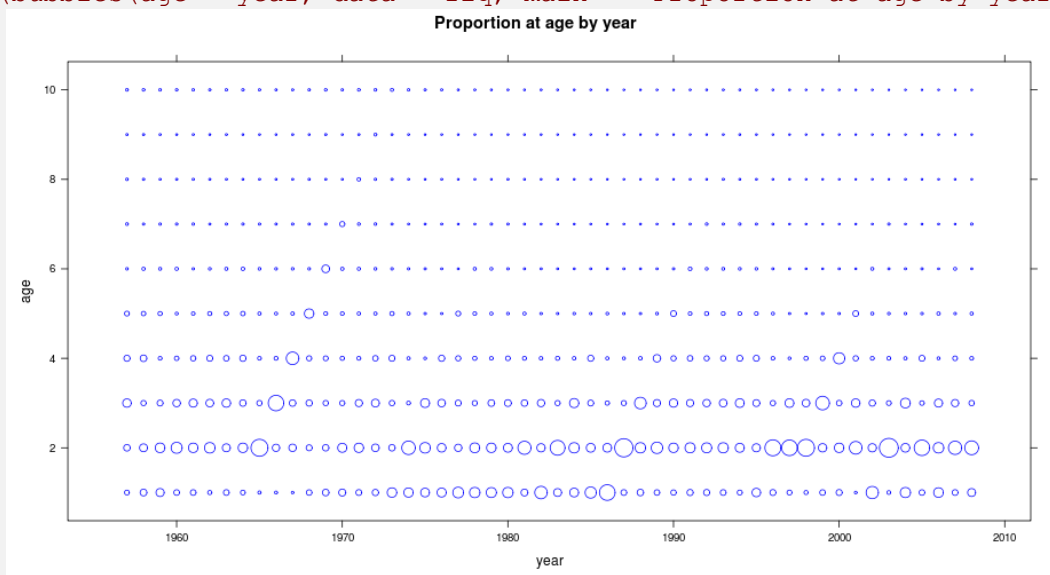
`FUN` the function to be used to carry out the sweep. In the case of binary operators such as `"/"` etc., the function name must be backquoted or quoted. (FUN is found by a call to `match.fun`.)

`check.margin` logical. If TRUE (the default), warn if the length or dimensions of `STATS` do not match the specified dimensions of `x`. Set to FALSE for a small speed gain when you *know* that dimensions match.

`...` optional arguments to FUN.

The function is useful when performing an operation on two arrays which don't have the same dimensions, for example when calculating catch proportions (see `yearTotals` example above), selection pattern by first scaling fishing mortality-at-age by the average value or looking at sex ratios where numbers by sex are divided by total numbers

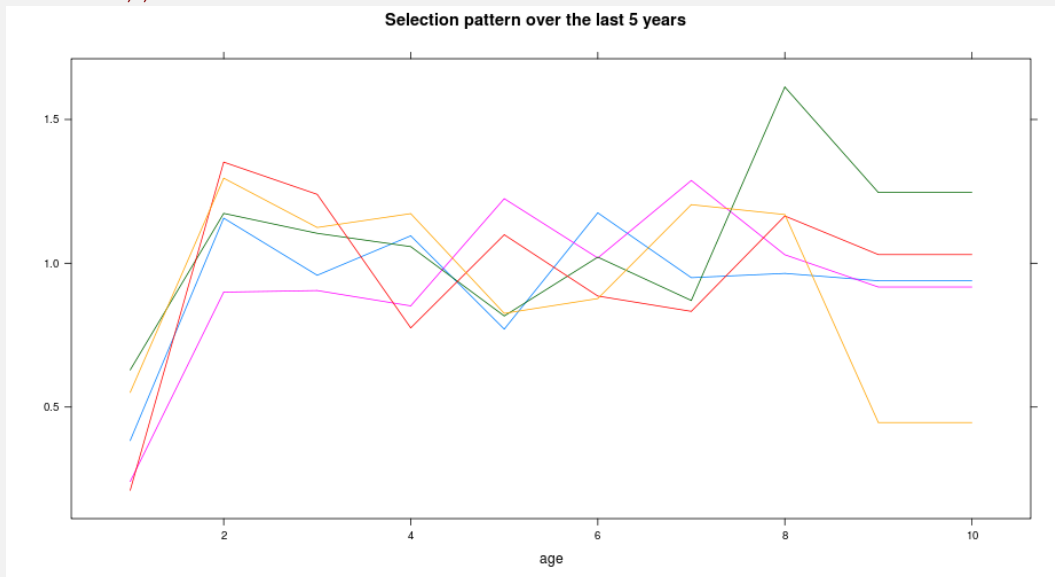
```
> flq <- sweep(catch.n(ple4), 2, apply(catch.n(ple4), 2, sum),
+             "/")
> print(bubbles(age ~ year, data = flq, main = "Proportion at age by year"))
```



```

> fbr <- apply(harvest(ple4)[ac(3:6)], 2, mean)
> selPattern <- sweep(harvest(ple4), 2, fbr, "/")
> print(xyplot(data ~ age, groups = year, data = trim(selPattern,
+   year = c(2004:2008)), type = "l", main = "Selection pattern over the last 5
years",
+   ylab = ""))

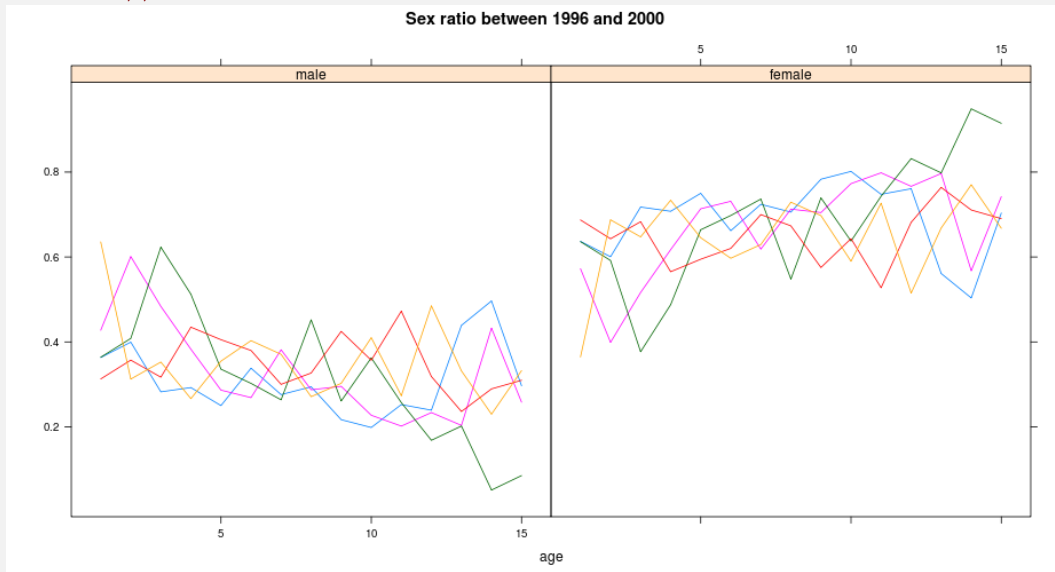
```



```

> num <- sweep(stock.n(ple4sex), c(1, 2), apply(stock.n(ple4sex),
+       c(1, 2), sum), "/")
> print(xyplot(data ~ age | unit, groups = year, data = trim(num,
+       year = c(1996:2000)), type = "l", main = "Sex ratio between 1996 and 2000",
+       ylab = ""))

```



2.4 Re-shaping

Often the dimensions of an `FLQuant` have to be changed, e.g. to add or remove extra years or ages, this can be done by `trim` and `window`. The first allows the selection of different dimensions at once in a very flexible way, while the second deals only with `year` but allows it to be extended. When conducting Monte Carlo Simulations iterations can be added using `propagate`, or using the random variable methods (e.g. `rnorm`). Changing other dimensions can be a bit more tricky as this depends upon what the `FLQuant` represents, for example if you increase the number of seasons, sexes or areas how does weight-at-age change or how should mortality be partitioned? `expand` provides a method for increasing the size of an `FLQuant` without actually filling up the new object with values. In cases where there is an accepted algorithm, for example to create a plusgroup then a specific method such as `setPlusGroup` can be implemented.

```
trim(x, ...)
```

x

an FLQuant.

...

The dimensions of the FLQuant to be trimmed, which can any of year, unit, season, area or iter. In the case of the quant dimension it will depend on its name, usually age or all, but it can be something else that the user must know. In case of doubts run `quant` on your object. In any case the trim information is given in a numeric or character vector that must be in agreement with the dimensions of the quant. No point on selecting year 1 if it does not exist ...

Some examples below, note that the selection does not need to be continuous

```
> trim(catch.n(ple4), year = c(2000:2008), age = c(1:4))
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

      year
age 2000  2001  2002  2003  2004  2005  2006  2007  2008
  1 106056  34804 311003  68309 233397  96478 220828  78598 135786
  2 187647 388808 193978 560788 191141 342682 244064 224345 263599
  3  91904 240444 136274 101562 221343  72076 177680 106621  80258
  4 230201 123671  61739  68836  43238 103120  35910  57788  44207

units: thousands
> trim(catch.n(ple4sex), year = c(1995:2000), unit = "female",
+      age = 10:14)
An object of class "FLQuant"
, , unit = female, season = all, area = unique

      year
age 1995  1996  1997  1998  1999  2000
 10 2318.98 1848.49 1716.03 1317.92  657.66 1054.35
 11  727.62  977.43 1310.99  730.05  411.88  633.39
 12  599.75  448.37  634.74 1163.33  414.76  349.49
 13  560.61  210.95  298.14  592.06  611.70  224.43
 14  296.50  199.10  218.98  315.91  218.08  367.98

units: NA
```

Bare in mind that when changing the information about a stock by trimming some information, the other information must also be computed, e.g., if 'age' is trimmed in `ple4`, catch, landings and discards need to be recalculated. However, the effect of getting rid of younger ages is not necessarily the same as getting rid of older ages since often the oldest age is a plus group and represents the sum of all ages greater or equal to that age.

Another example to illustrate trim's flexibility

```
> trim(catch.n(ple4), year = c(1990:1995, 2000, 2008), age = c(4:1))
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

  year
age 1990  1991  1992  1993  1994  1995  2000  2008
  4 139099 156894 106842 84252 81274 79230 230201 44207
  3 266440 225170 179701 146463 137320 97151 91904 80258
  2 311831 344841 263573 208032 140851 119251 187647 263599
  1 146864 184587 142165 99832 63516 126614 106056 135786

units: thousands
```

As mentioned above, to select and extend continuous years on the object use

```
window(x, start, end)
```

x	an FLQuant.
start	First year in new FLQuant
end	Last year in new FLQuant

A simple example


```
> window(catch.n(ple4), 2005, 2010)
An object of class "FLQuant"
, , unit = unique, season = all, area = unique
```

	year					
age	2005	2006	2007	2008	2009	2010
1	96478	220828	78598	135786	NA	NA
2	342682	244064	224345	263599	NA	NA
3	72076	177680	106621	80258	NA	NA
4	103120	35910	57788	44207	NA	NA
5	21746	45525	16408	31758	NA	NA
6	15610	6965	24293	6217	NA	NA
7	5489	5721	2663	15634	NA	NA
8	2615	2247	3505	1803	NA	NA
9	2223	1235	543	673	NA	NA
10	613	1319	1259	896	NA	NA

```
units: thousands
```

While more technical manipulations like setting a new plus group can be done with

```
setPlusGroup(x, plusgroup, ...)
```

x an FLQuant.

plusgroups the age for the new plus group

... not implemented for the moment.

A simple example

```

> setPlusGroup(catch.n(ple4), 5)[, ac(2005:2008)]
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

      year
age 2005      2006      2007      2008
1  96478.0 220828.0  78598.0 135786.0
2  342682.0 244064.0 224345.0 263599.0
3   72076.0 177680.0 106621.0  80258.0
4 103120.0  35910.0  57788.0  44207.0
5   8049.3  10502.0   8111.8   9496.8

units: thousands
> quantMeans(catch.n(ple4)[ac(5:10), ac(2005:2008)])
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

      year
age 2005      2006      2007      2008
all  8049.3 10502.0   8111.8   9496.8

units: thousands

```

Note that when applied to a `FLQuant` the method simply computes the average of the age groups to be merged.

Expanding in general can be done with

```

expand(x, ...)

```

x an `FLQuant`.

... The dimensions of the `FLQuant` to be expanded, which can any of `year`, `unit`, `season`, `area` or `iter`. In the case of the `quant` dimension it will depend on its name, usually `age` or `all`, but it can be something else that the user must know. In case of doubts run `quant` on your object. In any case the `expand` information is given in a numeric or character vector that must be in agreement with the dimensions of the `quant`. In particular it must start on the initial value of the dimension. This is redundant and likely to be changed in the future.

Note that the usage of `expand` for the non-numeric dimensions like `unit` and `area` can be tricky. For example if you add extra seasons or units what should be the new stock weights-at-age?

```

> expand(catch.n(ple4), age = 1:12)[, ac(2005:2008)]
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

  year
age 2005  2006  2007  2008
1   96478 220828  78598 135786
2   342682 244064 224345 263599
3    72076 177680 106621  80258
4   103120  35910  57788  44207
5    21746  45525  16408  31758
6    15610   6965  24293   6217
7     5489   5721   2663  15634
8     2615   2247   3505   1803
9     2223   1235    543    673
10     613   1319   1259    896
11      NA     NA     NA     NA
12      NA     NA     NA     NA

units:  thousands
> expand(catch.n(ple4), unit = c("unique", "female"))[, ac(2005:2008)]
An object of class "FLQuant"
, , unit = unique, season = all, area = unique

  year
age 2005  2006  2007  2008
1   96478 220828  78598 135786
2   342682 244064 224345 263599
3    72076 177680 106621  80258
4   103120  35910  57788  44207
5    21746  45525  16408  31758
6    15610   6965  24293   6217
7     5489   5721   2663  15634
8     2615   2247   3505   1803
9     2223   1235    543    673
10     613   1319   1259    896

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

  year
age 2005  2006  2007  2008
1      NA     NA     NA     NA
2      NA     NA     NA     NA
3      NA     NA     NA     NA
4      NA     NA     NA     NA
5      NA     NA     NA     NA
6      NA     NA     NA     NA
7      NA     NA     NA     NA
8      NA     NA     NA     NA
9      NA     NA     NA     NA
10     NA     NA     NA     NA

units:  thousands

```

And to expand and fill the empty information at once use

```
propagate(object, ...)
```

object an FLQuant.

... These can be `iter`, the number of iters to be added, and `fill.iter` a logical defining if the new iters should have the same information as the first (the default) or not.

Note the effect of the `fill.iter` argument.

```
> flq <- propagate(catch.n(ple4), 10)
> all.equal(iter(flq, 1), iter(flq, 10), check.attributes = FALSE)
[1] TRUE
> flq <- propagate(catch.n(ple4), 10, fill.iter = FALSE)
> all.equal(iter(flq, 1), iter(flq, 10), check.attributes = FALSE)
[1] "'is.NA' value mismatch: 520 in current 0 in target"
```

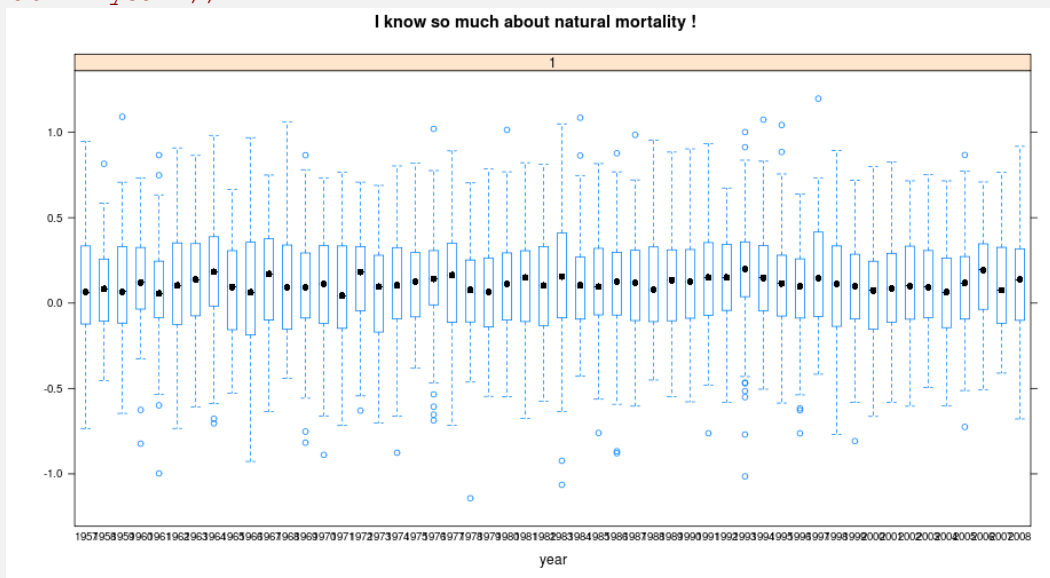
2.5 *Random Variables*

In **FLR** random variables are modelled in the 6th dimension of the FLQuant i.e. `iter`. Some distributions are already implemented in **FLR** making coding easier and more readable, which interact directly with FLQuants.

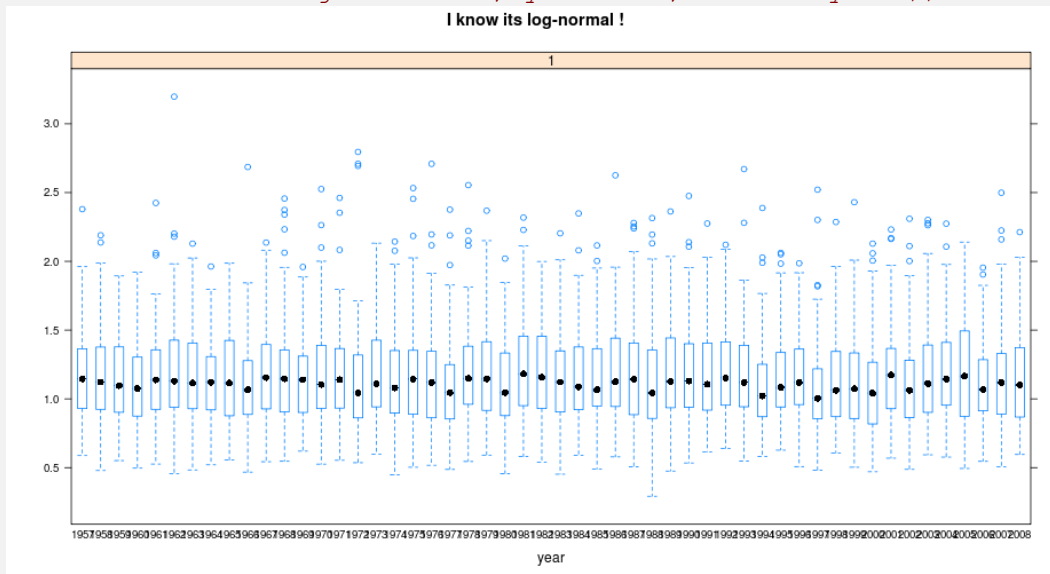
- ➔ `runif(n, min, max)`
- ➔ `rnorm(n, mean, sd)`
- ➔ `rlnorm(n, meanlog, sdlog)`

For all of these check the relevant help page for a description of each argument. **FLR**'s implementation simply allows the usage of FLQuant objects for the arguments.

```
> flq <- rnorm(100, m(ple4), sd = 0.3)
> print(bwplot(data ~ year | factor(age), data = trim(flq, age = 1),
+   main = "I know so much about natural mortality !", ylab = "",
+   xlab = "year"))
```



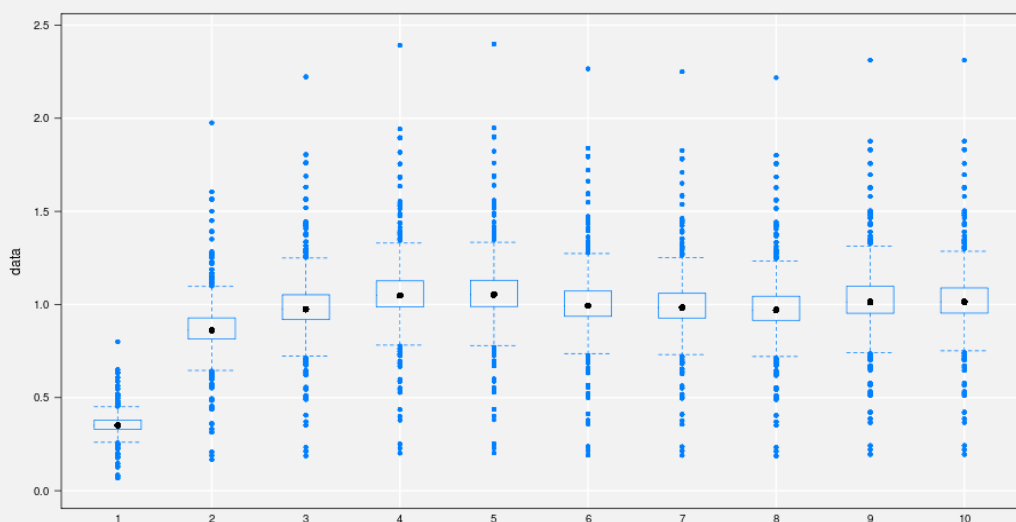
```
> flq <- rlnorm(100, m(ple4), sd = 0.3)
> print(bwplot(data ~ year | factor(age), data = trim(flq, age = 1),
+   main = "I know its log-normal !", ylab = "", xlab = "year"))
```



`runif` can be used to generate other distributions, since if you set `min=0` and `max=1`, then it returns the probability of a value and if you know the quantile function either empirical (e.g. `qnorm`) or non-parametric from a model fit (e.g. the residuals) then an appropriate pdf can be generated.

Creating an non-parametric random variable can be done using the basic **R** methods, sweep, apply and propagate.

```
> selPat <- sweep(harvest(ple4), 2, fbar(ple4), "/")
> mnSel <- apply(selPat, 1, mean)
> selRsd <- log(sweep(selPat, 1, mnSel, "/"))
> rvSel <- propagate(selRsd, 100)
> rvSel[] <- c(selRsd)[sample(1:length(selRsd), prod(dim(rvSel)),
+   replace = T)]
> rvSel <- sweep(exp(rvSel), 1, mnSel, "*")
> pfun <- function(x, y, ...) {
+   panel.grid(h = -1, v = -1, col = "white", lwd = 2)
+   panel.bwplot(x, y, ...)
+ }
> pset <- list(background = list(col = "gray95"), plot.symbol = list(pch = 19,
+   cex = 0.5))
> print(bwplot(data ~ factor(age), data = rvSel, par.settings = pset,
+   panel = pfun))
```



3 FLlst and FLQuants

The methods applied to FLQuant can also be used for the generic FLlst objects and more directly to FLQuants, by using lapply. Which will apply a function over all elements in a list. For more information on lapply look for the help pages.

```
lapply(X, FUN, ...)
```

X an FLlst object

FUN the function to be applied to each element of 'X'. In the case of functions like '+', '%*%', etc., the function name must be backquoted or quoted.

... optional arguments to 'FUN'

The examples below illustrate the functionality of `lapply`. Check that if the function applied does not return a number then the output of `lapply` will be a list instead of an `FLlst` object. The `as` method is used to coerce objects between classes, here it is used to coerce a `FLStock` object into a `FLQuants` list.

```
> flqs <- as(ple4, "FLQuants")
> flqs2 <- lapply(flqs, yearMeans)
> lst <- list()
> length(lst) <- length(flqs)
> names(lst) <- names(flqs)
> for (i in 1:length(flqs)) lst[[i]] <- yearMeans(flqs[[i]])
> all.equal(flqs2, lst, check.attributes = FALSE)
[1] TRUE
> is(flqs2)
[1] "FLQuants" "FLlst"       "list"       "vector"
> flqs3 <- lapply(flqs, units)
> is(flqs3)
[1] "list"       "vector"
```

Remember that `units` return a character with the units of the data, which is not possible to coerce into a `FLQuant`.

The next example is based on a function that calculates the average of the last `n` years.

```
> wts <- FLQuants(stock.wt = stock.wt(ple4), catch.wt = catch.wt(ple4),
+        landings.wt = landings.wt(ple4), discards.wt = discards.wt(ple4))
> mnWts <- lapply(wts, function(x, nyrs) apply(x[, dim(x)[2] -
+        0:nyrs], 1, mean), nyrs = 3)
```

More complicated procedures could be written in the same way, e.g. creating random variables from weights-at-age by turning the non parametric random variable example into a function and then using `lapply` with it.

Now suppose one is dealing with a large simulation study and using lists of `FLStock` objects, the so called `FLStocks`, which is an extended class of `FLlst`. From this object one is interesting in extracting all the

catch matrices,

```
> stks <- FLStocks(stk1 = ple4, stk2 = ple4, stk3 = ple4)
> is(stks)
[1] "FLStocks" "FLlst"    "list"      "vector"
> flqs <- lapply(stks, catch.n)
> is(flqs)
[1] "FLQuants" "FLlst"    "list"      "vector"
```

because `catch.n` is a method one can apply it with `lapply` and get those slots, with the benefit of generating a `FLQuants` object, instead of a simple list. One way of introducing variability in catch-at-age in all `FLStocks` is the following:

```
> stkS <- lapply(stks, function(x) {
+   catch.n(x) <- rlnorm(1000, catch.n(x), sd = 0.3)
+   x
+ })
```

Note that one can embed `lapply` within other `lapply` calls but be aware that the sequence of methods can have a large impact on the speed and memory used.

```
> system.time(lst1 <- lapply(lapply(stkS, catch.n), trim, age = 1:3))
  user  system elapsed 
0.900   0.040   0.947 
> system.time(lst2 <- lapply(lapply(stkS, trim, age = 1:3), catch.n))
  user  system elapsed 
0.350   0.000   0.361 
> all.equal(lst1, lst2)
[1] TRUE
```

An interesting method, with lots of applications for plotting is `mcf`, which makes `FLQuants` compatible with respect to their dimensionality. Hence, the `FLQuants` in the returned object all have the same dimensions, padded with NAs if necessary.

```
mcf(object, ...)
```

`object` A list of `FLQuant`, not necessarily a `FLQuants`.

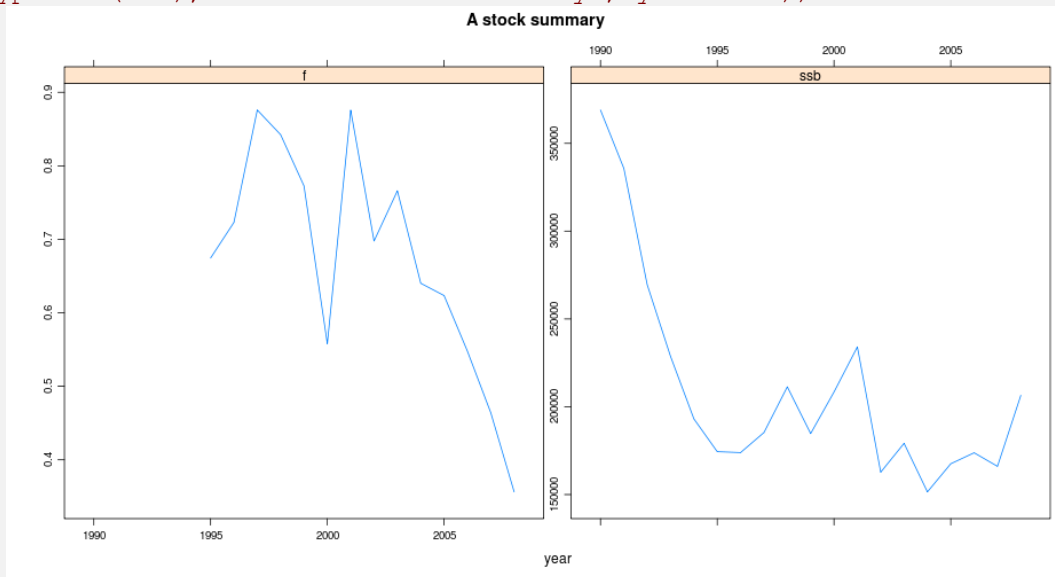
`...` Not implemented for the moment.

A simple example is


```

> flqs <- mcf(list(f = fbar(ple4)[, ac(1995:2008)], ssb = ssb(ple4)[,
+   ac(1990:2008)]))
> is(flqs)
[1] "FLQuants" "FLlst"      "list"      "vector"
> print(xyplot(data ~ year | qname, data = flqs, scales = list(y = "free"),
+   type = c("a"), main = "A stock summary", ylab = ""))

```



4 FLR Complex objects

Examples of [FLR](#) complex objects are `FLBiol`, `FLStock` and `FLIndex`, often you may wish to process all of them using the same function, this can be done using `qapply`.

```
qapply(X, FUN, ...)
```

X	an FLR object, <code>FLStock</code> , <code>FLBiol</code> , <code>FLFleet</code> , <code>FLIndex</code> , etc
FUN	the function to be applied to each element of 'X'. In the case of functions like '+', '%*%', etc., the function name must be backquoted or quoted.
...	optional arguments to 'FUN'

The examples below illustrate the functionality of `qapply`.

```

> dm <- qapply(ple4, dim)
> do.call("rbind", dm)
      [,1] [,2] [,3] [,4] [,5] [,6]
catch      1  52    1    1    1    1
catch.n    10  52    1    1    1    1
catch.wt    10  52    1    1    1    1
discards     1  52    1    1    1    1
discards.n  10  52    1    1    1    1
discards.wt 10  52    1    1    1    1
landings     1  52    1    1    1    1
landings.n  10  52    1    1    1    1
landings.wt 10  52    1    1    1    1
stock        1  52    1    1    1    1
stock.n     10  52    1    1    1    1
stock.wt    10  52    1    1    1    1
m           10  52    1    1    1    1
mat         10  52    1    1    1    1
harvest     10  52    1    1    1    1
harvest.spwn 10  52    1    1    1    1
m.spwn      10  52    1    1    1    1
> ple4new <- qapply(ple4, apply, 1, mean, na.rm = TRUE)
> all.equal(ple4new, qapply(ple4, yearMeans))
[1] TRUE
> ple4new <- qapply(ple4, function(x) {
+   names(dimnames(x))[1] <- "what"
+   return(x)
+ })
> unlist(qapply(ple4new, quant))
      catch      catch.n      catch.wt      discards      discards.n      discards.wt
"what"      "what"      "what"      "what"      "what"      "what"
landings    landings.n    landings.wt      stock      stock.n      stock.wt
"what"      "what"      "what"      "what"      "what"      "what"
      m      mat      harvest harvest.spwn      m.spwn
"what"      "what"      "what"      "what"      "what"

```

5 Final thoughts

As with all computing languages there's more than one way of doing the same thing and the “best” way will depend on the objectives of the analysis. Writing one off functions can often help in analyses, however if one is interested in speed it's a good idea to test your code and try to compute everything on the lowest level of the language. Using functions can also obscure what is being done and clarity is important for transparency.

FLR provides a few methods to deal with its objects which in our view cover all the important manipulations one is required to do. The biggest value is on the mixing and combination of these methods, but off course if something is missing **R** is a great language for quick development. Another major gain is in using other packages which may have implemented better methods than ours, that's the case of plyr (<http://had.co.nz/plyr/>).