The doBy package

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1 Introduction

The doBy package grew out of a need to calculate groupwise summary statistics in a simple way, much in the spirit of `PROC SUMMARY` of the SAS system. We have tried to keep the interface to the functions based on specifying formulas.

```
> library(doBy)
```

2 Data

The usage of the `doBy` package is based on the following datasets.

CO2 data The `CO2` data frame comes from an experiment on the cold tolerance of the grass species *Echinochloa crus-galli*. To limit the amount of output we modify names and levels of variables as follows

```
> data(CO2)
> CO2 <- transform(CO2, Treat = Treatment, Treatment = NULL)
> levels(CO2$Treat) <- c("nchil", "chil")
> levels(CO2$Type) <- c("Que", "Mis")
> CO2 <- subset(CO2, Plant %in% c("Qn1", "Qc1", "Mn1", "Mc1"))
```

Airquality data The `airquality` dataset contains air quality measurements in New York, May to September 1973. The months are coded as $5, \ldots, 9$. To limit the output we only consider data for two months:

```
> airquality <- subset(airquality, Month %in% c(5, 6))
```

Dietox data The `dietox` data are provided in the `doBy` package and result from a study of the effect of adding vitamin E and/or copper to the feed of slaughter pigs.

3 The `summaryBy` function

The `summaryBy` function is used for calculating quantities like "the mean and variance of x and y for each combination of two factors A and B". Examples are based on the `CO2` data.

3.1 Basic usage

For example, the mean and variance of `uptake` and `conc` for each value of `Plant` is obtained by:

```
> summaryBy(conc + uptake ~ Plant, data = CO2, FUN = function(x) {
+     c(m = mean(x), v = var(x))
+ })

  Plant conc.m conc.v uptake.m uptake.v
1   Qn1    435 100950    33.23    67.48
2   Qc1    435 100950    29.97    69.47
3   Mn1    435 100950    26.40    75.59
4   Mc1    435 100950    18.00    16.96
```

Defining the function to return named values as above is the recommended use of `summaryBy`. The function can also be defined outside the call to `summaryBy`:

```
> myfun1 <- function(x) {
+     c(m = mean(x), v = var(x))
+ }
> summaryBy(conc + uptake ~ Plant, data = CO2, FUN = myfun1)
```

Note that the values returned by the function has been named as `m` and `v`. If the result of the function(s) are not named, then the names in the output data in general become less intuitive:

```
> myfun2 <- function(x) {
+     c(mean(x), var(x))
+ }
> summaryBy(conc + uptake ~ Plant, data = CO2, FUN = myfun2)

  Plant conc.myfun21 conc.myfun22 uptake.myfun21 uptake.myfun22
1   Qn1          435       100950          33.23          67.48
2   Qc1          435       100950          29.97          69.47
3   Mn1          435       100950          26.40          75.59
4   Mc1          435       100950          18.00          16.96
```

3.2 Using a list of functions

It is possible use a list of functions. A typical usage will be by invoking a list of predefined functions:

```
> summaryBy(uptake ~ Plant, data = CO2, FUN = c(mean, var, median))

  Plant uptake.mean uptake.var uptake.median
1   Qn1       33.23      67.48          35.3
2   Qc1       29.97      69.47          32.5
3   Mn1       26.40      75.59          30.0
4   Mc1       18.00      16.96          18.9
```

Slightly more elaborate is

```
> mymed <- function(x) c(med = median(x))
> summaryBy(uptake ~ Plant, data = CO2, FUN = c(mean, var, mymed))

  Plant uptake.mean uptake.var uptake.med
1   Qn1       33.23      67.48       35.3
2   Qc1       29.97      69.47       32.5
3   Mn1       26.40      75.59       30.0
4   Mc1       18.00      16.96       18.9
```

The naming of the output variables determined from what the functions returns. The names of the last two columns above are imposed by `summaryBy` because `myfun2` does not return named values.

3.3 Naming output variables with the `postfix` argument

The `postfix` argument gives an altertive way of naming the output variables: For example, the functions `myfun1` and `myfun2` both returns two values. These can be named as:

```
> summaryBy(conc + uptake ~ Plant, data = CO2, postfix = list(c("mean1",
+     "var1"), c("mean2", "var2")), FUN = c(myfun1, myfun2))

  Plant conc.mean1 conc.var1 uptake.mean1 uptake.var1 conc.mean2 conc.var2
1   Qn1        435    100950        33.23       67.48        435    100950
2   Qc1        435    100950        29.97       69.47        435    100950
3   Mn1        435    100950        26.40       75.59        435    100950
4   Mc1        435    100950        18.00       16.96        435    100950
  uptake.mean2 uptake.var2
1        33.23       67.48
2        29.97       69.47
3        26.40       75.59
4        18.00       16.96
```

3.4 Copying variables out with the `id` argument

To get the value of the `Type` and `Treat` in the first row of the groups (defined by the values of `Plant`) copied to the output dataframe we use the `id` argument: as:

```
> summaryBy(conc + uptake ~ Plant, data = CO2, FUN = myfun1, id = ~Type +
+     Treat)

  Plant conc.m conc.v uptake.m uptake.v Type Treat
1   Qn1    435 100950    33.23    67.48  Que nchil
2   Qc1    435 100950    29.97    69.47  Que  chil
3   Mn1    435 100950    26.40    75.59  Mis nchil
4   Mc1    435 100950    18.00    16.96  Mis  chil
```

3.5 Statistics on functions of data

We may want to calculate the mean and variance for the logarithm of `uptake`, for `uptake+conc` (not likely to be a useful statistic) as well as for `uptake` and `conc`. This can be achieved as:

```
> summaryBy(log(uptake) + I(conc + uptake) + conc + uptake ~ Plant,
+     data = CO2, FUN = myfun1)

  Plant log(uptake).m log(uptake).v conc+uptake.m conc+uptake.v conc.m conc.v
1   Qn1         3.467       0.10168         468.2        104747    435 100950
2   Qc1         3.356       0.11873         465.0        105297    435 100950
3   Mn1         3.209       0.17928         461.4        105642    435 100950
4   Mc1         2.864       0.06874         453.0        103157    435 100950
  uptake.m uptake.v
1    33.23    67.48
2    29.97    69.47
3    26.40    75.59
4    18.00    16.96
```

If one does not want output variables to contain parentheses then setting `p2d=TRUE` causes the parentheses to be replaced by dots (“.”).

```
> summaryBy(log(uptake) + I(conc + uptake) ~ Plant, data = CO2,
+     p2d = TRUE, FUN = myfun1)

  Plant log.uptake..m log.uptake..v conc+uptake.m conc+uptake.v
1   Qn1         3.467       0.10168         468.2        104747
2   Qc1         3.356       0.11873         465.0        105297
3   Mn1         3.209       0.17928         461.4        105642
4   Mc1         2.864       0.06874         453.0        103157
```

3.6 Using "." on the left hand side of a formula

It is possible to use the dot (".") on the left hand side of the formula. The dot means "all numerical variables which do not appear elsewhere" (i.e. on the right hand side of the formula and in the `id` statement):

```
> summaryBy(log(uptake) + I(conc + uptake) + . ~ Plant, data = CO2,
+     FUN = myfun1)

  Plant log(uptake).m log(uptake).v conc+uptake.m conc+uptake.v conc.m conc.v
1   Qn1         3.467       0.10168         468.2        104747    435 100950
2   Qc1         3.356       0.11873         465.0        105297    435 100950
3   Mn1         3.209       0.17928         461.4        105642    435 100950
4   Mc1         2.864       0.06874         453.0        103157    435 100950
  uptake.m uptake.v
1    33.23    67.48
2    29.97    69.47
3    26.40    75.59
4    18.00    16.96
```

3.7 Using "." on the right hand side of a formula

The dot (".") can also be used on the right hand side of the formula where it refers to "all non–numerical variables which are not specified elsewhere":

```
> summaryBy(log(uptake) ~ Plant + ., data = CO2, FUN = myfun1)

  Plant Type Treat log(uptake).m log(uptake).v
1   Qn1  Que nchil         3.467       0.10168
2   Qc1  Que  chil         3.356       0.11873
3   Mn1  Mis nchil         3.209       0.17928
4   Mc1  Mis  chil         2.864       0.06874
```

3.8 Using "1" on the right hand side of the formula

Using 1 on the right hand side means no grouping:

```
> summaryBy(log(uptake) ~ 1, data = CO2, FUN = myfun1)

  log(uptake).m log(uptake).v
1         3.224        0.1577
```

3.9 Preserving names of variables using `keep.names`

If the function applied to data only returns one value, it is possible to force that the summary variables retain the original names by setting `keep.names=TRUE`. A typical use of this could be

```
> summaryBy(conc + uptake + log(uptake) ~ Plant, data = CO2, FUN = mean,
+     id = ~Type + Treat, keep.names = TRUE)

  Plant conc uptake log(uptake) Type Treat
1   Qn1  435  33.23       3.467  Que nchil
2   Qc1  435  29.97       3.356  Que  chil
3   Mn1  435  26.40       3.209  Mis nchil
4   Mc1  435  18.00       2.864  Mis  chil
```

4 The `orderBy` function

Ordering (or sorting) a data frame is possible with the `orderBy` function. Suppose we want to order the rows of the the `airquality` data by `Temp` and by `Month` (within `Temp`). This can be achieved by:

```
> x <- orderBy(~Temp + Month, data = airquality)
```

The first lines of the result are:

```
> head(x)

   Ozone Solar.R Wind Temp Month Day
5     NA      NA 14.3   56     5   5
18     6      78 18.4   57     5  18
25    NA      66 16.6   57     5  25
27    NA      NA  8.0   57     5  27
15    18      65 13.2   58     5  15
26    NA     266 14.9   58     5  26
```

If we want the ordering to be by decreasing values of one of the variables, we change the sign, e.g.

```
> x <- orderBy(~-Temp + Month, data = airquality)
> head(x)

   Ozone Solar.R Wind Temp Month Day
42    NA     259 10.9   93     6  11
43    NA     250  9.2   92     6  12
40    71     291 13.8   90     6   9
39    NA     273  6.9   87     6   8
41    39     323 11.5   87     6  10
36    NA     220  8.6   85     6   5
```

5 The `splitBy` function

Suppose we want to split the `airquality` data into a list of dataframes, e.g. one dataframe for each month. This can be achieved by:

```
> x <- splitBy(~Month, data = airquality)
```

Information about the grouping is stored as a dataframe in an attribute called `groupid` and can be retrieved with:

```
> attr(x, "groupid")

   Month
1      5
32     6
```

6 The `sampleBy` function

Suppose we want a random sample of 50 % of the observations from a dataframe. This can be achieved with:

```
> sampleBy(~1, frac = 0.5, data = airquality)
```

Suppose instead that we want a systematic sample of every fifth observation within each month. This is achieved with:

```
> sampleBy(~Month, frac = 0.2, data = airquality, systematic = T)
```

7 The subsetBy function

Suppose we want to take out those rows within each month for which the the wind speed is larger than the mean wind speed (within the month). This is achieved by:

```
> subsetBy(~Month, subset = "Wind>mean(Wind)", data = airquality)
```

Note that the statement `"Wind>mean(Wind)"` is evaluated within each month.

8 The transformBy function

The `transformBy` function is analogous to the `transform` function except that it works within groups. For example:

```
> transformBy(~Month, data = airquality, minW = min(Wind), maxW = max(Wind),
+     chg = sum(range(Wind) * c(-1, 1)))
```

9 The lapplyBy function

This `lapplyBy` function is a wrapper for calling lapply on the list resulting from first calling splitBy.

Suppose we want to calculate the weekwise feed efficiency of the pigs in the `dietox` data, i.e. weight gain divided by feed intake.

```
> data(dietox)
> dietox <- orderBy(~Pig + Time, data = dietox)
> v <- lapplyBy(~Pig, data = dietox, function(d) c(NA, diff(d$Weight)/diff(d$Feed)))
> dietox$FE <- unlist(v)
```

Technically, the above is the same as

```
> dietox <- orderBy(~Pig + Time, data = dietox)
> wdata <- splitBy(~Pig, data = dietox)
> v <- lapply(wdata, function(d) c(NA, diff(d$Weight)/diff(d$Feed)))
> dietox$FE <- unlist(v)
```

10 Miscellaneous

10.1 The esticon function

Consider a linear model which explains `Ozone` as a linear function of `Month` and `Wind`:

```
> data(airquality)
> airquality <- transform(airquality, Month = factor(Month))
> m <- lm(Ozone ~ Month * Wind, data = airquality)
> coefficients(m)

(Intercept)      Month6      Month7      Month8      Month9        Wind
     50.748     -41.793      68.296      82.211      23.439      -2.368
Month6:Wind Month7:Wind Month8:Wind Month9:Wind
      4.051      -4.663      -6.154      -1.874
```

When a parameter vector β of (systematic) effects have been estimated, interest is often in a particular estimable function, i.e. linear combination $\lambda^\top \beta$ and/or testing the hypothesis $H_0 : \lambda^\top \beta = \beta_0$ where λ is a specific vector defined by the user.

Suppose for example we want to calculate the expected difference in ozone between consequtive months at wind speed 10 mph (which is about the average wind speed over the whole period).

The `esticon` function provides a way of doing so. We can specify several λ vectors at the same time. For example

```
> Lambda

     [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
[1,]    0   -1    0    0    0    0  -10    0    0     0
[2,]    0    1   -1    0    0    0   10  -10    0     0
[3,]    0    0    1   -1    0    0    0   10  -10     0
[4,]    0    0    0    1   -1    0    0    0   10   -10

> esticon(m, Lambda)

Confidence interval ( WALD ) level = 0.95
  beta0 Estimate Std.Error t.value  DF Pr(>|t|) Lower.CI Upper.CI
1     0   1.2871    10.238  0.1257 106  0.90019  -19.010   21.585
2     0 -22.9503    10.310 -2.2259 106  0.02814  -43.392   -2.509
3     0   0.9954     7.094  0.1403 106  0.88867  -13.069   15.060
4     0  15.9651     6.560  2.4337 106  0.01662    2.959   28.971
```

In other cases, interest is in testing a hypothesis of a contrast $H_0 : \Lambda\beta = \beta_0$ where Λ is a matrix. For example a test of no interaction between `Month` and `Wind` can be made by testing jointly that the last four parameters in `m` are zero (observe that the test is a Wald test):

```
> Lambda

     [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10]
[1,]    0    0    0    0    0    0    1    0    0     0
[2,]    0    0    0    0    0    0    0    1    0     0
[3,]    0    0    0    0    0    0    0    0    1     0
[4,]    0    0    0    0    0    0    0    0    0     1

> esticon(m, Lambda, joint.test = T)

  X2.stat DF Pr(>|X^2|)
1   22.11  4  0.0001906
```

For a linear normal model, one would typically prefer to do a likelihood ratio test instead. However, for generalized estimating equations of glm–type (as dealt with in the packages geepack and gee) there is no likelihood. In this case `esticon` function provides an operational alternative.

Observe that another function for calculating contrasts as above is the `contrast` function in the Design package but it applies to a narrower range of models than `esticon` does.

11 Final remarks

The `esticon` functions and other smaller functions are likely to be removed from the doBy package in the future. Credit is due to Dennis Chabot, Gabor Grothendieck, Paul Murrell and Erik Jørgensen for reporting various bugs and making various suggestions to the functionality in the doBy package.