Groupwise computations and other utilities in the `doBy` package

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1 Introduction

The **doBy** package contains a variety of utility functions. This working document describes some of these functions. The package originally grew out of a need to calculate groupwise summary statistics (much in the spirit of `PROC SUMMARY` of the SAS system), but today the package contains many different utilities.

2 Data used for illustration

The description 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 Working with groupwise data

3.1 The `summaryBy` function

The `summaryBy` function is used for calculating quantities like "the mean and variance of numerical variables x and y for each combination of two factors A and B". Notice: A functionality similar to `summaryBy` is provided by `aggregate()` from base R.

```
myfun1 <- function(x){c(m=mean(x), s=sd(x))}
summaryBy(cbind(conc, uptake, lu=log(uptake)) ~ Plant,
          data=CO2, FUN=myfun1)
```

```
##   Plant conc.m conc.s uptake.m uptake.s  lu.m   lu.s
## 1   Qn1    435  317.7    33.23    8.215 3.467 0.3189
## 2   Qc1    435  317.7    29.97    8.335 3.356 0.3446
## 3   Mn1    435  317.7    26.40    8.694 3.209 0.4234
## 4   Mc1    435  317.7    18.00    4.119 2.864 0.2622
```

A simpler call is

```
summaryBy(conc ~ Plant, data=CO2, FUN=mean)
```

Instead of formula we may specify a list containing the left hand side and the right hand side of a formula[1] but that is possible only for variables already in the dataframe:

```
## Will fail because of log(uptake)
## summaryBy(list(c("conc", "uptake", "log(uptake)"), "Plant"),
##           data=CO2, FUN=myfun1)
## Works
summaryBy(list(c("conc", "uptake"), "Plant"),
          data=CO2, FUN=myfun1)
```

---

[1] This is a feature of `summaryBy` and it does not work with `aggregate`.

3.2 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:

```
x1 <- orderBy(~ Temp + Month, data=airquality)
head(x1)
```

```
##     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 can do

```
x2 <- orderBy(~ - Temp + Month, data=airquality)
```

An alternative form is:

```
x3 <- orderBy(c("Temp", "Month"), data=airquality)
x4 <- orderBy(c("-Temp", "Month"), data=airquality)
```

3.3 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)
lapply(x, head, 4)
```

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```
## $'5'
##    Ozone Solar.R Wind Temp Month Day
## 1    41      190  7.4   67      5   1
## 2    36      118  8.0   72      5   2
## 3    12      149 12.6   74      5   3
## 4    18      313 11.5   62      5   4
##
## $'6'
##     Ozone Solar.R Wind Temp Month Day
## 32    NA      286  8.6   78      6   1
## 33    NA      287  9.7   74      6   2
## 34    NA      242 16.1   67      6   3
## 35    NA      186  9.2   84      6   4
```

```r
attributes(x)
```

```
## $names
## [1] "5" "6"
##
## $groupid
##    Month
## 1      5
## 2      6
##
## $idxvec
## $idxvec$'5'
##  [1]  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## [26] 26 27 28 29 30 31
##
## $idxvec$'6'
##  [1] 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56
## [26] 57 58 59 60 61
##
##
## $grps
##  [1] "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5"
## [20] "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "5" "6" "6" "6" "6" "6" "6" "6"
```

```
## [39] "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6" "6
## [58] "6" "6" "6" "6"
##
## $class
## [1] "splitByData" "list"
```

An alternative call is

```
splitBy("Month", data=airquality)
```

3.4   The `subsetBy` function

Suppose we want to select 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:

```
x <- subsetBy(~Month, subset=Wind > mean(Wind), data=airquality)
head(x)
```

```
##        Ozone Solar.R Wind Temp Month Day
## 5.3      12     149 12.6   74     5   3
## 5.5      NA      NA 14.3   56     5   5
## 5.6      28      NA 14.9   66     5   6
## 5.8      19      99 13.8   59     5   8
## 5.9       8      19 20.1   61     5   9
## 5.15     18      65 13.2   58     5  15
```

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

3.5   The `transformBy` function

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

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```
x <- transformBy(~Month, data=airquality,
                 minW=min(Wind), maxW=max(Wind),
                 chg = diff(range(Wind)))
head(x)
```

```
##     Ozone Solar.R Wind Temp Month Day minW maxW  chg
## 5.1    41     190  7.4   67     5   1  5.7 20.1 14.4
## 5.2    36     118  8.0   72     5   2  5.7 20.1 14.4
## 5.3    12     149 12.6   74     5   3  5.7 20.1 14.4
## 5.4    18     313 11.5   62     5   4  5.7 20.1 14.4
## 5.5    NA      NA 14.3   56     5   5  5.7 20.1 14.4
## 5.6    28      NA 14.9   66     5   6  5.7 20.1 14.4
```

Alternative forms:

```
x <- transformBy("Month", data=airquality,
                 minW=min(Wind), maxW=max(Wind),
                 chg = diff(range(Wind)))
```

4 Miscellaneous

4.1 Specialize

```
ff  <- function(a, b=2, c=4){a + b + c}
ff1 <- specialize(ff, arglist=list(a=1, b=7, yy=123))
ff1
```

```
## function (c = 4)
## {
##     1 + 7 + c
## }
## <environment: 0x55f7b6cd01a8>
```

```
gg  <- rnorm
gg1 <- specialize(gg, list(n=10))
gg1

## function (mean = 0, sd = 1)
## .Call(C_rnorm, 10, mean, sd)
## <environment: 0x55f7b6db9120>
```

Notice that this result is absurd:

```
f  <- function(a) {a <- a + 1; a}
f1 <- specialize(f, list(a = 10))
f1

## function ()
## {
##      10 <- 10 + 1
##      10
## }
## <environment: 0x55f7b14a3918>
```

4.2   The `firstobs()` / `lastobs()` function

To obtain the indices of the first/last occurences of an item in a vector do:

```
x <- c(1,1,1,2,2,2,1,1,1,3)
firstobs(x)

## [1]  1  4 10

lastobs(x)

## [1]  6  9 10
```

The same can be done on a data frame, e.g.

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```
firstobs(~Plant, data=CO2)
```

```
## [1]  1  8 15 22
```

```
lastobs(~Plant, data=CO2)
```

```
## [1]  7 14 21 28
```

4.3 The `which.maxn()` and `which.minn()` functions

The location of the n largest / smallest entries in a numeric vector can be obtained with

```
x <- c(1:4,0:5,11,NA,NA)
which.maxn(x,3)
```

```
## [1] 11 10  4
```

```
which.minn(x,5)
```

```
## [1] 5 1 6 2 7
```

4.4 Subsequences - `subSeq()`

Find (sub) sequences in a vector:

```
x <- c(1,1,2,2,2,1,1,3,3,3,3,1,1,1)
subSeq(x)
```

```
##   first last slength midpoint value
## 1     1    2       2        2     1
## 2     3    5       3        4     2
```

```
## 3     6    7       2        7      1
## 4     8   11       4       10      3
## 5    12   14       3       13      1
```

```
subSeq(x, item=1)
```

```
##   first last slength midpoint value
## 1     1    2       2        2      1
## 2     6    7       2        7      1
## 3    12   14       3       13      1
```

```
subSeq(letters[x])
```

```
##   first last slength midpoint value
## 1     1    2       2        2      a
## 2     3    5       3        4      b
## 3     6    7       2        7      a
## 4     8   11       4       10      c
## 5    12   14       3       13      a
```

```
subSeq(letters[x],item="a")
```

```
##   first last slength midpoint value
## 1     1    2       2        2      a
## 2     6    7       2        7      a
## 3    12   14       3       13      a
```

4.5 Recoding values of a vector - `recodeVar()`

```
x <- c("dec","jan","feb","mar","apr","may")
src1 <- list(c("dec","jan","feb"), c("mar","apr","may"))
tgt1 <- list("winter","spring")
recodeVar(x,src=src1,tgt=tgt1)
```

```
## [1] "winter" "winter" "winter" "spring" "spring" "spring"
```

4.6 Renaming columns of a dataframe or matrix – re-nameCol()

```
head(renameCol(CO2, 1:2, c("kk","ll")))
```

```
##     kk  ll conc uptake Treat
## 1 Qn1 Que   95   16.0 nchil
## 2 Qn1 Que  175   30.4 nchil
## 3 Qn1 Que  250   34.8 nchil
## 4 Qn1 Que  350   37.2 nchil
## 5 Qn1 Que  500   35.3 nchil
## 6 Qn1 Que  675   39.2 nchil
```

```
head(renameCol(CO2, c("Plant","Type"), c("kk","ll")))
```

```
##     kk  ll conc uptake Treat
## 1 Qn1 Que   95   16.0 nchil
## 2 Qn1 Que  175   30.4 nchil
## 3 Qn1 Que  250   34.8 nchil
## 4 Qn1 Que  350   37.2 nchil
## 5 Qn1 Que  500   35.3 nchil
## 6 Qn1 Que  675   39.2 nchil
```

4.7 Time since an event - timeSinceEvent()

Consider the vector

```
yvar <- c(0,0,0,1,0,0,0,0,0,1,0,0,0,1,1,0,0,0,0,0)
```

Imagine that "1" indicates an event of some kind which takes place at a certain time point. By default time points are assumed equidistant but for illustration we define time time variable

```
tvar <- seq_along(yvar) + c(0.1,0.2)
```

Now we find time since event as

```
tse<- timeSinceEvent(yvar,tvar)
```

The output reads as follows:

- `abs.tse`: Absolute time since (nearest) event.

- `sign.tse`: Signed time since (nearest) event.

- `ewin`: Event window: Gives a symmetric window around each event.

- `run`: The value of `run` is set to 1 when the first event occurs and is increased by 1 at each subsequent event.
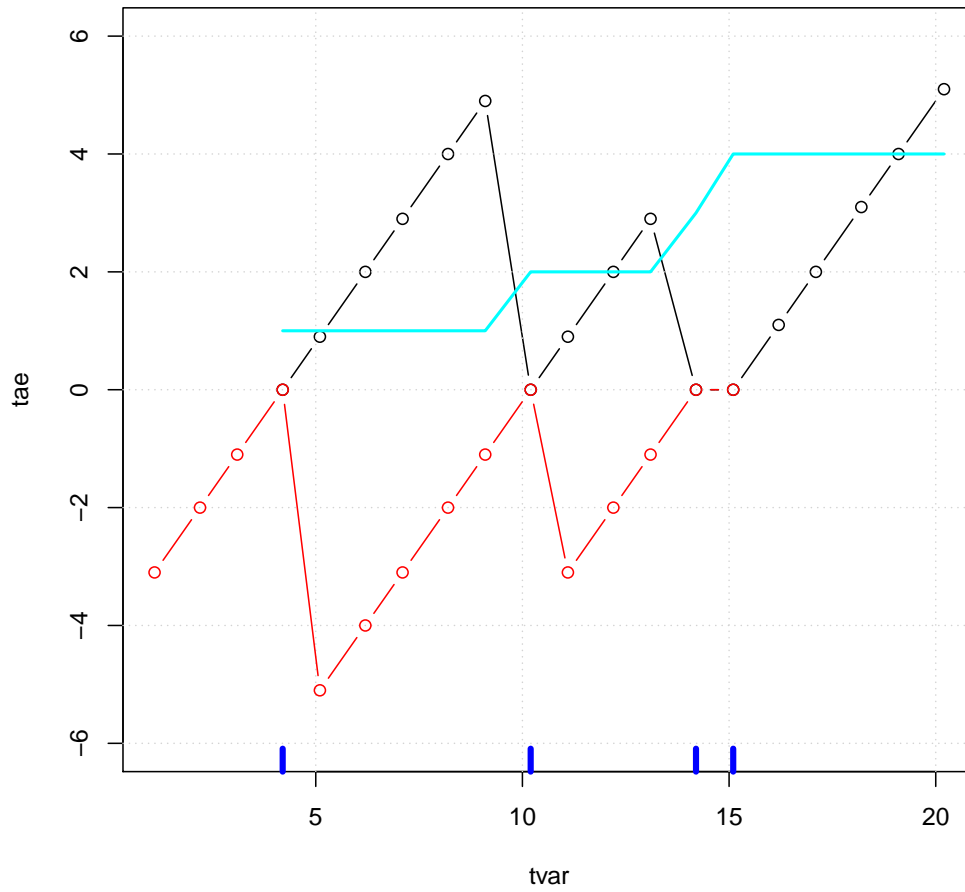
- `tae`: Time after event.

- `tbe`: Time before event.

```
plot(sign.tse~tvar, data=tse, type="b")
grid()
rug(tse$tvar[tse$yvar==1], col="blue",lwd=4)
points(scale(tse$run), col=tse$run, lwd=2)
lines(abs.tse+.2~tvar, data=tse, type="b",col=3)
```
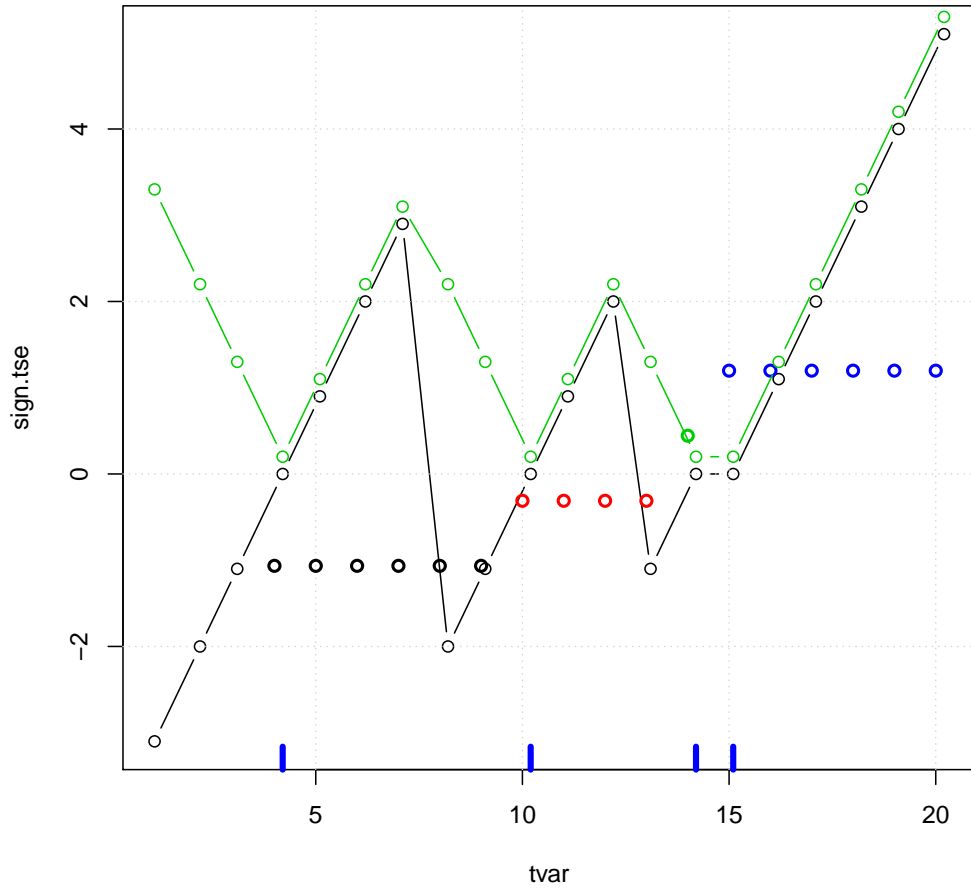
```
plot(tae~tvar, data=tse, ylim=c(-6,6),type="b")
grid()
lines(tbe~tvar, data=tse, type="b", col="red")
rug(tse$tvar[tse$yvar==1], col="blue",lwd=4)
lines(run~tvar, data=tse, col="cyan",lwd=2)
```
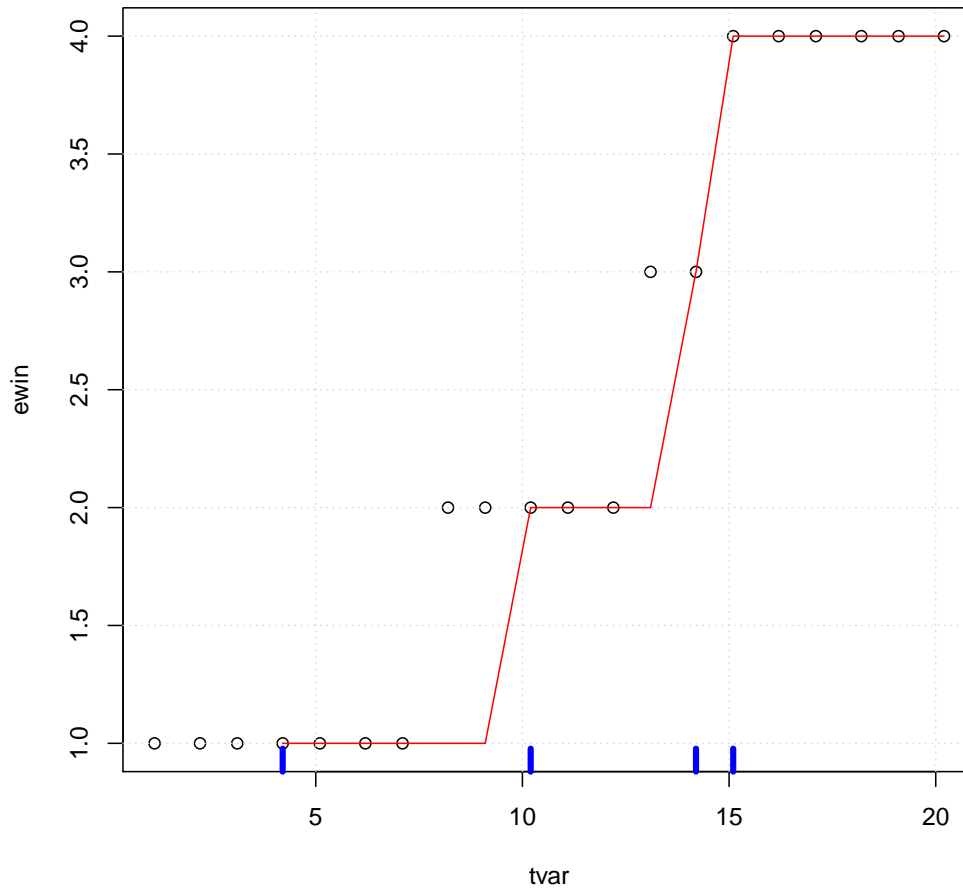
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```
plot(ewin~tvar, data=tse,ylim=c(1,4))
rug(tse$tvar[tse$yvar==1], col="blue",lwd=4)
grid()
lines(run~tvar, data=tse,col="red")
```
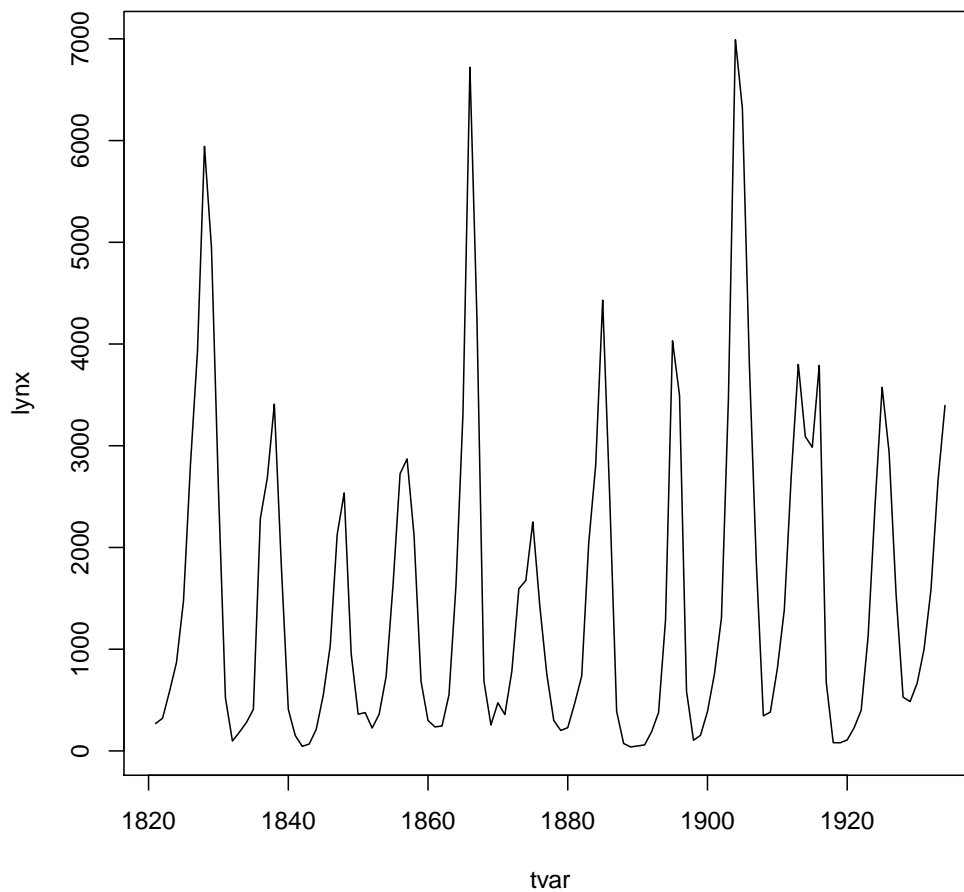
We may now find times for which time since an event is at most 1 as

```
tse$tvar[tse$abs<=1]
```

```
## [1]  4.2  5.1 10.2 11.1 14.2 15.1
```

4.8   Example: Using `subSeq()` and `timeSinceEvent()`

Consider the `lynx` data:

```r
lynx <- as.numeric(lynx)
tvar <- 1821:1934
plot(tvar,lynx,type="l")
```


Suppose we want to estimate the cycle lengths. One way of doing this is as follows:

```r
yyy <- lynx>mean(lynx)
head(yyy)
```

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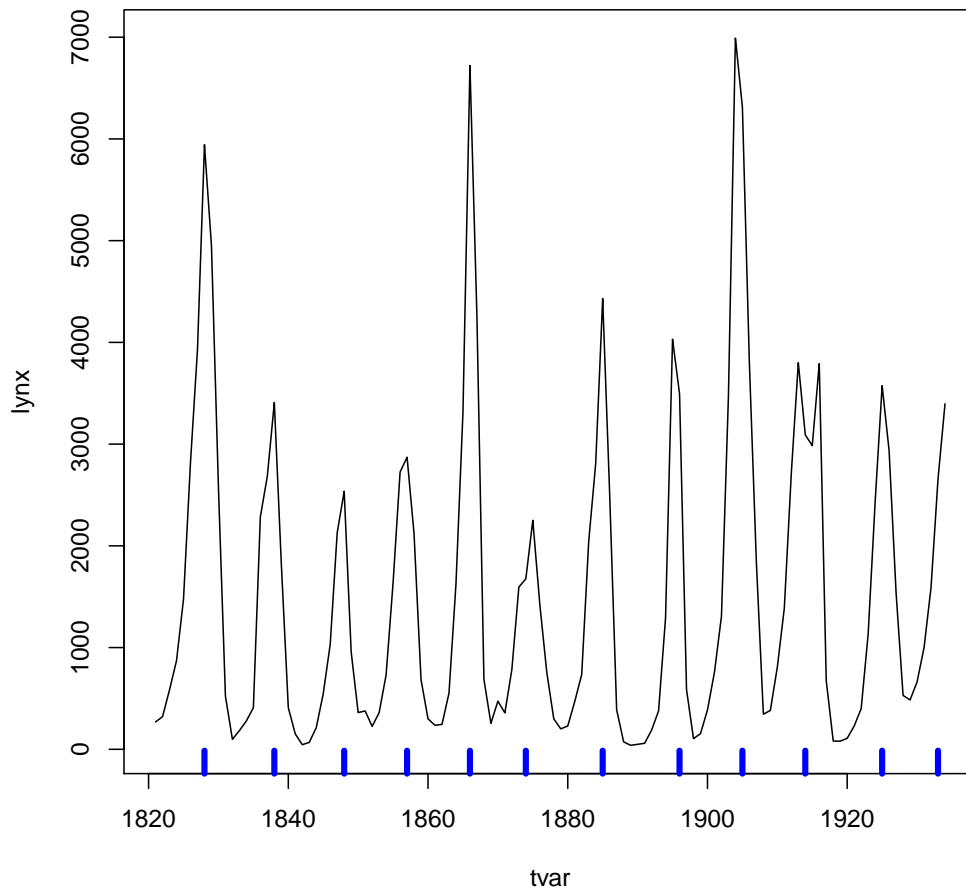
```
## [1] FALSE FALSE FALSE FALSE FALSE  TRUE
```

```
sss <- subSeq(yyy,TRUE)
sss
```

```
##     first last slength midpoint value
## 1       6   10       5        8  TRUE
## 2      16   19       4       18  TRUE
## 3      27   28       2       28  TRUE
## 4      35   38       4       37  TRUE
## 5      44   47       4       46  TRUE
## 6      53   55       3       54  TRUE
## 7      63   66       4       65  TRUE
## 8      75   76       2       76  TRUE
## 9      83   87       5       85  TRUE
## 10     92   96       5       94  TRUE
## 11    104  106       3      105  TRUE
## 12    112  114       3      113  TRUE
```

```
plot(tvar,lynx,type="l")
rug(tvar[sss$midpoint],col="blue",lwd=4)
```

Create the "event vector"

```
yvar <- rep(0,length(lynx))
yvar[sss$midpoint] <- 1
str(yvar)

##  num [1:114] 0 0 0 0 0 0 0 0 1 0 0 ...
```

```
tse <- timeSinceEvent(yvar,tvar)
head(tse,20)
```

```
##    yvar tvar abs.tse sign.tse ewin run tae tbe
## 1     0 1821       7       -7    1  NA  NA  -7
## 2     0 1822       6       -6    1  NA  NA  -6
## 3     0 1823       5       -5    1  NA  NA  -5
## 4     0 1824       4       -4    1  NA  NA  -4
## 5     0 1825       3       -3    1  NA  NA  -3
## 6     0 1826       2       -2    1  NA  NA  -2
## 7     0 1827       1       -1    1  NA  NA  -1
## 8     1 1828       0        0    1   1   0   0
## 9     0 1829       1        1    1   1   1  -9
## 10    0 1830       2        2    1   1   2  -8
## 11    0 1831       3        3    1   1   3  -7
## 12    0 1832       4        4    1   1   4  -6
## 13    0 1833       5        5    1   1   5  -5
## 14    0 1834       4       -4    2   1   6  -4
## 15    0 1835       3       -3    2   1   7  -3
## 16    0 1836       2       -2    2   1   8  -2
## 17    0 1837       1       -1    2   1   9  -1
## 18    1 1838       0        0    2   2   0   0
## 19    0 1839       1        1    2   2   1  -9
## 20    0 1840       2        2    2   2   2  -8
```

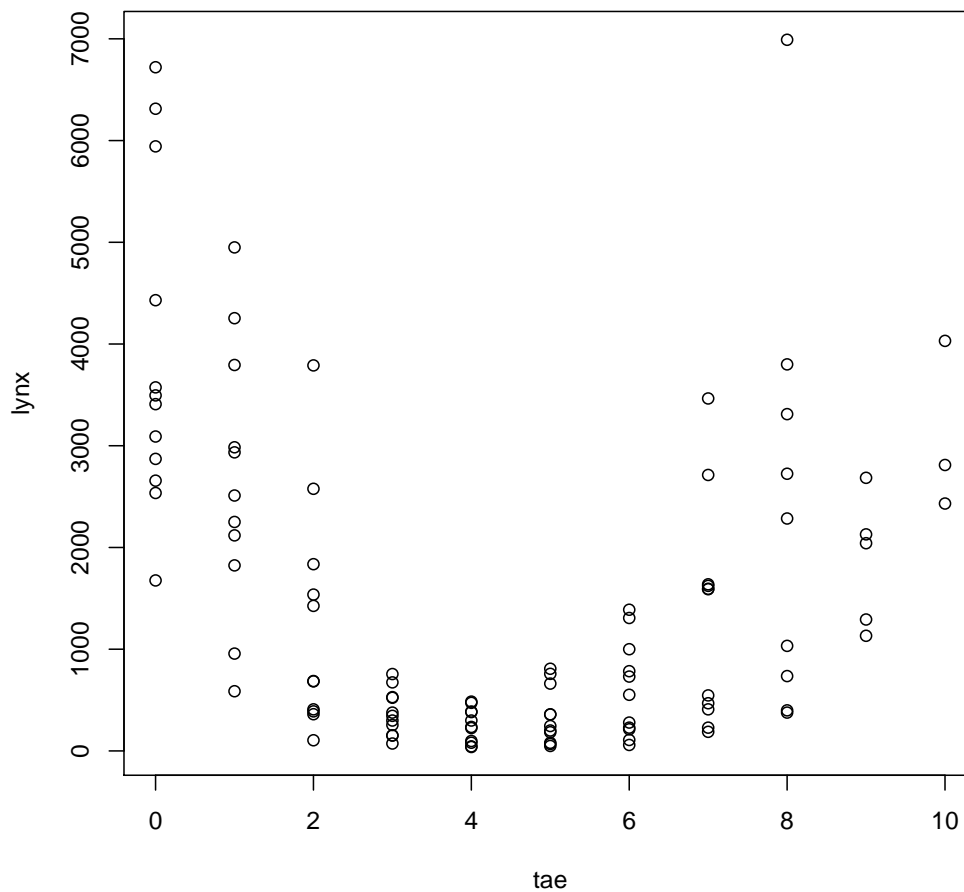We get two different (not that different) estimates of period lengths:

```
len1 <- tapply(tse$ewin, tse$ewin, length)
len2 <- tapply(tse$run, tse$run, length)
c(median(len1), median(len2), mean(len1), mean(len2))
```

```
## [1] 9.500 9.000 9.500 8.917
```
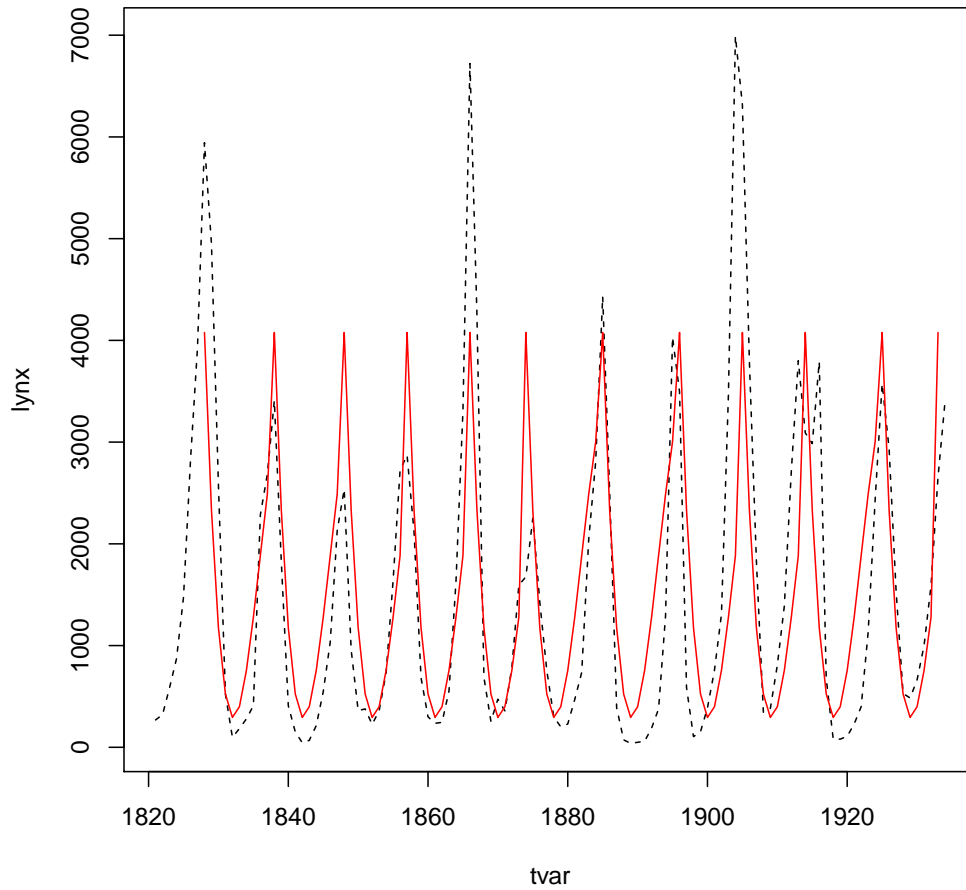
We can overlay the cycles as:
```

```
tse$lynx <- lynx
tse2 <- na.omit(tse)
plot(lynx ~ tae, data=tse2)
```



```
plot(tvar, lynx, type="l", lty=2)
mm <- lm(lynx ~ tae + I(tae^2) + I(tae^3), data=tse2)
lines(fitted(mm) ~ tvar, data=tse2, col="red")
```

5 Acknowledgements