Computing Summary Statistics for Daily Data

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These examples demonstrate how to compute selected summary statistics for daily streamflow data. The examples can easily be extended to other statistics or data types.

```
> # Load the smwrBase and smwrData packages
> library(smwrBase)
> library(smwrData)
> # Retrieve streamflow data for the Choptank River near Greensboro, Maryland
> data(ChoptankFlow)
> # Print the first and last few rows of the data
> head(ChoptankFlow)
  agency_cd site_no
                      datetime Flow Flow_cd
1
       USGS 01491000 1990-01-01 385
       USGS 01491000 1990-01-02
                                649
3
      USGS 01491000 1990-01-03
                                353
4
      USGS 01491000 1990-01-04
                                246
5
      USGS 01491000 1990-01-05
                                225
      USGS 01491000 1990-01-06 214
```

> tail(ChoptankFlow)

```
agency_cd site_no
                         datetime Flow Flow_cd
8030
         USGS 01491000 2011-12-26 245
8031
         USGS 01491000 2011-12-27
                                   224
8032
         USGS 01491000 2011-12-28
                                   337
8033
         USGS 01491000 2011-12-29
                                   370
                                             Α
8034
         USGS 01491000 2011-12-30
                                   255
8035
         USGS 01491000 2011-12-31 221
                                             Α
```

- > # Check for missing values
- > with(ChoptankFlow, screenData(datetime, Flow, year = "calendar"))

No missing data between 1990-01-01 and 2011-12-31

1 Computing Daily Mean Values

The simplest and most straightforward way to compute summary statistics from arbitrarily grouped data is to use the tapply function. At its simplest, it requires only three arguments—X, the data to summarize; INDEX, the grouping data; and FUN, the summary statistic function.

The smwrBase package contains the baseDay function that can be used to group data by day, so that all data for each day, including February 29, can be summarized. The output can be arranged so that the sequence represents the calendar-, water- or climate-year; beginning January 1, October 1, or April 1, respectively.

The following script demonstrates how to use the tapply and baseDay functions to compute the daily mean streamflow for the previously retrieved data. It uses the with function to facilitate referring to columns in the dataset.

```
> # There are no missing values, so only need the basic
> # 3 arguments for tapply
> ChoptankFlow.daily <- with(ChoptankFlow, tapply(Flow,
      baseDay(datetime, numeric=FALSE, year="calendar"), mean))
> # Print the first and last few values of the output
> head(ChoptankFlow.daily)
  Jan 01
           Jan 02
                    Jan 03
                             Jan 04
                                      Jan 05
                                                Jan 06
144.8182 183.7727 174.6818 171.9091 159.4545 146.3182
> tail(ChoptankFlow.daily)
 Dec 26
           Dec 27
                    Dec 28
                             Dec 29
                                      Dec 30
                                               Dec 31
245.2727 276.3636 192.6818 161.4091 143.2727 135.9545
```

The output from tapply is an array. Because the output from this example is an array of one dimension, it is printed in the form of a named vector. Had the summary statistic function been quantile, for example, the output would have been a list.

The tapply function is very powerful and easy to use however, there are times when we want the output in the form of a dataset rather than a vector or array. In those cases, the aggregate function is a better alternative than the tapply function. The aggregate function has several usage options. The script below demonstrates how to build a formula to compute the same statistics that we computed in the previous script. Early versions of aggregate required the output of the summary statistic function to be a scalar, but that is no longer a limitation.

```
2 Jan 02 183.7727
3 Jan 03 174.6818
4 Jan 04 171.9091
5 Jan 05 159.4545
6 Jan 06 146.3182
```

> tail(ChoptankFlow.dailyDF)

```
baseDay(datetime, numeric = FALSE, year = "calendar") Flow
361 Dec 26 245.2727
362 Dec 27 276.3636
363 Dec 28 192.6818
364 Dec 29 161.4091
365 Dec 30 143.2727
366 Dec 31 135.9545
```

- > # Change the name of the grouping column
- > names(ChoptankFlow.dailyDF)[1] <- "Day"</pre>

Note that the grouping column, now called Day, is a factor. If character data are needed, executing the expression:

 $\label{lower} {\tt ChoptankFlow.dailyDF\$Day} \ {\tt <-} \ {\tt as.character(ChoptankFlow.dailyDF\$Day)} \ {\tt will convert the column Day to character.}$

2 Computing Annual Mean Values

The previous example can easily be expanded to any grouping that the user desires. This example computes annual means by calendar year. The year function in lubridate is used to group the data by calendar year. The waterYear function in smwrBase can be used to group the data by water year.

```
> ChoptankFlow.yrDF <- aggregate(Flow ~
      year(datetime),
      data=ChoptankFlow, FUN=mean)
> # Change the name of the grouping column
> names(ChoptankFlow.yrDF)[1] <- "CalYear"</pre>
> # Print the first few values of the output
> head(ChoptankFlow.yrDF)
 CalYear
               Flow
1
     1990 114.67945
2
     1991 99.54521
3
     1992 85.62568
4
     1993 119.65726
5
     1994 203.11233
     1995 94.18000
```

> # There are no missing values

Other grouping functions include month (month) in lubridate, seasons (user-defined seasons) in smwrBase. Refer to the documentation for each of these functions for a description of the arguments.

3 Computing yearly and monthly Mean Values

Aggregation can also be done by multiple grouping variables. This example computes the mean streamflow for each month by year. This example uses the year and the month functions because the output is sorted by groups. The sequence of the groups in the call is important—the sorting is done in the order specified in the formula. For this example, the data are sorted by month and then by year, which in this case, keeps the order correct; grouping by water year would misplace October, November, and December. For a calendar year table, the months are in the correct order.

```
> # There are no missing values
> ChoptankFlow.my <- aggregate(Flow ~ month(datetime, label=TRUE) + year(datetime),
      data=ChoptankFlow, FUN=mean)
> # Rename columns 1 and 2
> names(ChoptankFlow.my)[1:2] <- c("Month", "Year")</pre>
> # Print the first few values of the output
> head(ChoptankFlow.my)
 Month Year
                 Flow
1
    Jan 1990 238.5161
   Feb 1990 152.0357
2
3
   Mar 1990 137.5806
4
    Apr 1990 223.6667
    May 1990 288.8710
    Jun 1990 108.4333
```

The output dataset may be used as is, or it could be restructured to a table of monthly values for each calendar year. To create a table by water year, the levels in the column Month must be reordered to begin in October and end in September.

```
> # Restructure the dataset
> ChoptankFlow.myTbl <- group2row(ChoptankFlow.my, "Year", "Month", "Flow")
> # Print the first few values of the output, set width for Vignette
> options(width=70)
> head(ChoptankFlow.myTbl)
 Year Jan. Flow Feb. Flow Mar. Flow Apr. Flow May. Flow Jun. Flow
1 1990 238.51613 152.03571 137.5806 223.66667 288.87097 108.43333
2 1991 252.35484 108.85714 182.6452 192.10000 88.74194
3 1992 80.93548 84.10345 197.8065 109.76667 115.00000
4 1993 174.29032 172.67857 512.1935 312.36667 107.77419
                                                         39.56667
5 1994 210.25806 372.57143 826.2903 331.33333 120.96774
6 1995 183.54839 122.60714 221.8065 84.03333 107.16129
            Aug.Flow Sep.Flow Oct.Flow
                                         Nov.Flow Dec.Flow
  Jul.Flow
1 58.83871 44.580645 23.800000 29.58065
                                                   47.48387
                                         23.83333
2 93.93548 48.032258 25.833333 26.58065
                                         24.63333
                                                   79.16129
3 24.19355 62.838710 40.366667 39.19355
                                         69.00000 123.00000
4 13.03871
            9.812903 9.526667 13.11935
                                         20.63333
                                                    52.90323
5 75.03226 157.387097 96.300000 55.51613
                                         59.23333
                                                   84.32258
6 23.87097 17.777419 15.520000 53.32258 123.90000 129.06452
```

Note that this example used the <code>group2row</code> function in <code>smwrBase</code>. The <code>reshape</code> function in <code>stats</code> and <code>stack</code> and <code>unstack</code> functions in <code>utils</code> are other functions that will restructure data.