Recitation1

```
knitr::opts_chunk$set(echo = TRUE)
```

Marathon Dataset

The summary function provides min, max, mean, median, 1st quantile and 3rd quantile information for each column in the dataframe.

```
marathon <- read.csv('marathon.csv');</pre>
summary(marathon)
```

```
##
         Year
                          Time
##
    Min.
            :1971
                     Min.
                            :2.411
##
    1st Qu.:1978
                     1st Qu.:2.450
    Median:1985
                     Median :2.469
##
    Mean
            :1985
                     Mean
                            :2.566
##
    3rd Qu.:1992
                     3rd Qu.:2.542
##
    Max.
            :1999
                     Max.
                            :3.145
```

We see that the Marathon dataset contains 2 columns (Year and Time). When the dataset is large, we usually skim the dataset by examing the first few rows. This can be achieved with the function head, where the second input indicates the number of rows to be displayed. Similarly, tail function is used to examine the last few rows.

head (marathon, 10)

```
##
      Year
              Time
      1971 2.92278
## 1
## 2
      1972 3.14472
## 3
     1973 2.95194
## 4
      1974 3.12472
## 5
      1975 2.77056
      1976 2.65306
## 7
      1977 2.71944
## 8
     1978 2.54167
## 9 1979 2.45917
## 10 1980 2.42833
```

To access a particular column, say the Time column, we can use the command marathon\$Time or marathon ["Time"]. The former outputs a vector, and the latter outputs another dataframe.

Compare

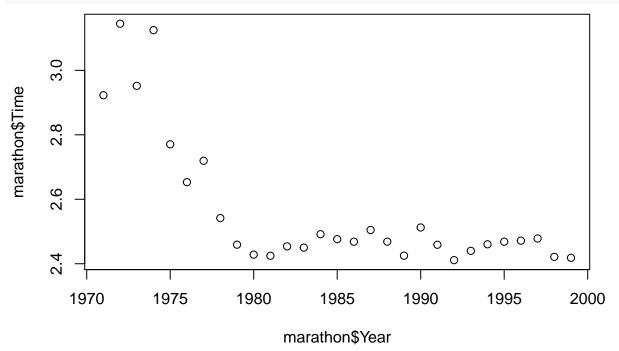
```
marathon$Time
    [1] 2.92278 3.14472 2.95194 3.12472 2.77056 2.65306 2.71944 2.54167
   [9] 2.45917 2.42833 2.42472 2.45389 2.45000 2.49167 2.47611 2.46833
## [17] 2.50472 2.46861 2.42500 2.51250 2.45889 2.41111 2.44000 2.46028
## [25] 2.46833 2.47167 2.47833 2.42139 2.41833
marathon['Time']
```

Time

```
2.92278
## 1
     3.14472
## 2
## 3
      2.95194
## 4
      3.12472
## 5
      2.77056
## 6
     2.65306
## 7
      2.71944
## 8
      2.54167
## 9
      2.45917
## 10 2.42833
## 11 2.42472
## 12 2.45389
## 13 2.45000
## 14 2.49167
## 15 2.47611
## 16 2.46833
## 17 2.50472
## 18 2.46861
## 19 2.42500
## 20 2.51250
## 21 2.45889
## 22 2.41111
## 23 2.44000
## 24 2.46028
## 25 2.46833
## 26 2.47167
## 27 2.47833
## 28 2.42139
## 29 2.41833
```

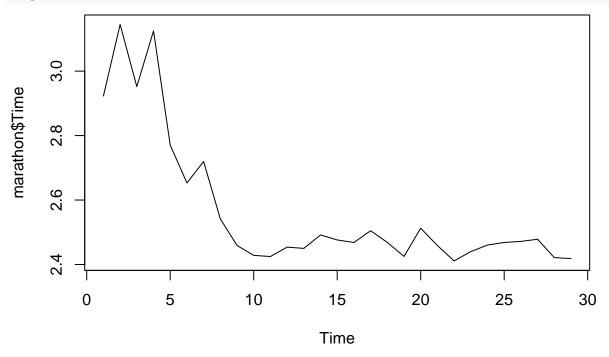
Finally, let us plot Time versus Year of the Marathon dataset. We can have either scatter plot

plot(marathon\$Year, marathon\$Time)



or the time series plot:

ts.plot(marathon\$Time)



Sleep Dataset

The Sleep dataset is a built-in dataset in R. Let us first gather some information about this dataset.

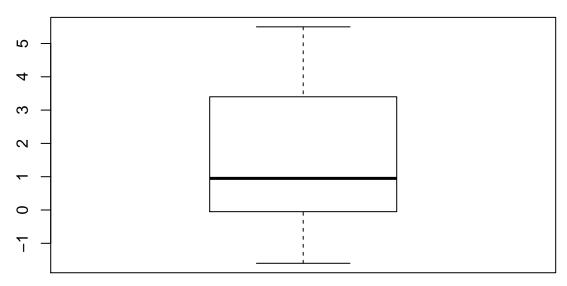
attributes(sleep)

```
## $names
## [1] "extra" "group" "ID"
##
## $class
## [1] "data.frame"
##
## $row.names
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
```

It has three columns, extra, group and ID.

Boxplot is a nice way to visualize the data, let us try for the extra column: the increase in the number of hours of sleep resulting from either drug.

```
boxplot(sleep['extra'])
```



In some data analysis, we need to focus on a subset of the data. For instance, we want to separate group=1 and group=2. This can be achieved using the following command:

```
group1 <- subset(sleep,group==1);
group2 <- subset(sleep,group==2)</pre>
```

We can check group1 and group 2:

group1

```
##
      extra group ID
## 1
        0.7
## 2
       -1.6
                 1
                     2
## 3
       -0.2
                 1
                     3
                     4
## 4
       -1.2
                 1
## 5
       -0.1
                 1
                    5
## 6
        3.4
                    6
                 1
##
        3.7
                 1
                    7
## 8
                    8
        0.8
                 1
## 9
        0.0
                 1
                    9
## 10
         2.0
                 1 10
```

group2

```
##
      extra group ID
## 11
        1.9
                 2
                    1
## 12
        0.8
                 2
                    2
## 13
                 2
                    3
        1.1
## 14
        0.1
                 2
                    4
## 15
       -0.1
                 2
                    5
## 16
        4.4
                 2
                    6
                 2
                    7
##
  17
        5.5
##
   18
        1.6
                 2
                    8
                 2 9
## 19
        4.6
## 20
        3.4
                 2 10
```

Now we can do a side by side comparison for the extra column between two groups in order to understand the effects of these two types of drugs.

