

# Homework1

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## Question 1

The Iowa data set `iowa.csv` is a toy example that summarises the yield of wheat (bushels per acre) for the state of Iowa between 1930-1962. In addition to yield, year, rainfall and temperature were recorded as the main predictors of yield.

a. First, we need to load the data set into R using the command `read.csv()`. Use the help function to learn what arguments this function takes. Once you have the necessary input, load the data set into R and make it a data frame called `iowa.df`.

```
#Use the help function  
?read.csv()
```

```
## starting httpd help server ... done
```

```
#load the data set  
iowadata<-read.csv("data/Iowa.csv",header = TRUE, sep = ";")  
iowa.df<-as.data.frame(iowadata)  
head(iowa.df)
```

```
##   Year Rain0 Temp1 Rain1 Temp2 Rain2 Temp3 Rain3 Temp4 Yield  
## 1 1930 17.75 60.2  5.83  69.0  1.49  77.9  2.42  74.4  34.0  
## 2 1931 14.76 57.5  3.83  75.0  2.72  77.2  3.30  72.6  32.9  
## 3 1932 27.99 62.3  5.17  72.0  3.12  75.8  7.10  72.2  43.0  
## 4 1933 16.76 60.5  1.64  77.8  3.45  76.4  3.01  70.5  40.0  
## 5 1934 11.36 69.5  3.49  77.2  3.85  79.7  2.84  73.4  23.0  
## 6 1935 22.71 55.0  7.00  65.9  3.35  79.4  2.42  73.6  38.4
```

b. How many rows and columns does `iowa.df` have?

```
#get the number of row and col separately  
nrow(iowa.df)
```

```
## [1] 33
```

```
ncol(iowa.df)
```

```
## [1] 10
```

```
#also we can get them directly  
dim(iowa.df)
```

```
## [1] 33 10
```

c. What are the names of the columns of iowa.df?

```
colnames(iowa.df)
```

```
## [1] "Year" "Rain0" "Temp1" "Rain1" "Temp2" "Rain2" "Temp3" "Rain3" "Temp4"  
## [10] "Yield"
```

d. What is the value of row 5, column 7 of iowa.df?

```
iowa.df[5,7]
```

```
## [1] 79.7
```

e. Display the second row of iowa.df in its entirety.

```
iowa.df[2,]
```

```
##   Year Rain0 Temp1 Rain1 Temp2 Rain2 Temp3 Rain3 Temp4 Yield  
## 2 1931 14.76 57.5  3.83    75  2.72  77.2   3.3  72.6  32.9
```

## Question 2: Syntax and class-typing.

a. For each of the following commands, either explain why they should be errors, or explain the non-erroneous result.

```
vector1 <- c("5", "12", "7", "32")  
max(vector1)  
sort(vector1)  
sum(vector1)  
  
#[1] "7"  
#[1] "12" "32" "5"  "7"  
#Error in sum(vector1) : 'type'(character)
```

The type of the elements is 'character'.

The way to find the max and to sort is according to ASCII code. Of course two characters can not be summed up.

b. For the next series of commands, either explain their results, or why they should produce errors.

```
vector2 <- c("5",7,12)
vector2[2] + vector2[3]
#Error in vector2[2] + vector2[3] :
```

The type of elements in the vector should be consistent, otherwise 7 and 12 will be automatically converted into the same character vector as "5", which cannot be summed.

```
dataframe3 <- data.frame(z1="5",z2=7,z3=12)
dataframe3[1,2] + dataframe3[1,3]
```

```
## [1] 19
```

Different types of variables can be stored in the data frame. The second and third elements in the first row are numeric vectors, which can be added together.

```
list4 <- list(z1="6", z2=42, z3="49", z4=126)
list4[[2]]+list4[[4]]
list4[2]+list4[4]
```

```
#[1] 168
#Error in list4[2] + list4[4] :
```

The elements of the first formula are in numerical form and therefore can be added; the elements of the second formula are in the form of a linked list and cannot be added.

### Question 3: Working with functions and operators.

a. The colon operator will create a sequence of integers in order. It is a special case of the function `seq()` which you saw earlier in this assignment. Using the help command `?seq` to learn about the function, design an expression that will give you the sequence of numbers from 1 to 10000 in increments of 372. Design another that will give you a sequence between 1 and 10000 that is exactly 50 numbers in length.

```
?seq()
seq(1,10000,by=372)
```

```
## [1] 1 373 745 1117 1489 1861 2233 2605 2977 3349 3721 4093 4465 4837 5209
## [16] 5581 5953 6325 6697 7069 7441 7813 8185 8557 8929 9301 9673
```

```
seq(1,10000,length.out = 50)
```

```
## [1] 1.0000 205.0612 409.1224 613.1837 817.2449 1021.3061
## [7] 1225.3673 1429.4286 1633.4898 1837.5510 2041.6122 2245.6735
## [13] 2449.7347 2653.7959 2857.8571 3061.9184 3265.9796 3470.0408
## [19] 3674.1020 3878.1633 4082.2245 4286.2857 4490.3469 4694.4082
## [25] 4898.4694 5102.5306 5306.5918 5510.6531 5714.7143 5918.7755
## [31] 6122.8367 6326.8980 6530.9592 6735.0204 6939.0816 7143.1429
## [37] 7347.2041 7551.2653 7755.3265 7959.3878 8163.4490 8367.5102
## [43] 8571.5714 8775.6327 8979.6939 9183.7551 9387.8163 9591.8776
## [49] 9795.9388 10000.0000
```

b. The function `rep()` repeats a vector some number of times. Explain the difference between `rep(1:3, times=3)` and `rep(1:3, each=3)`.

```
rep(1:3, times=3)
```

```
## [1] 1 2 3 1 2 3 1 2 3
```

```
rep(1:3, each=3)
```

```
## [1] 1 1 1 2 2 2 3 3 3
```

The first three-digit number “1,2,3” is generated and repeated three times as a whole.

The second produces three digits “1,2,3”, where each digit is repeated three times.

## MB.Ch1.2.

The orings data frame gives data on the damage that had occurred in US space shuttle launches prior to the disastrous Challenger launch of 28 January 1986. The observations in rows 1, 2, 4, 11, 13, and 18 were included in the pre-launch charts used in deciding whether to proceed with the launch, while remaining rows were omitted. Create a new data frame by extracting these rows from orings, and plot total incidents against temperature for this new data frame. Obtain a similar plot for the full data set.

```
#install.packages("DAAG")  
library(DAAG)
```

```
## Warning: package 'DAAG' was built under R version 4.0.2
```

```
## Loading required package: lattice
```

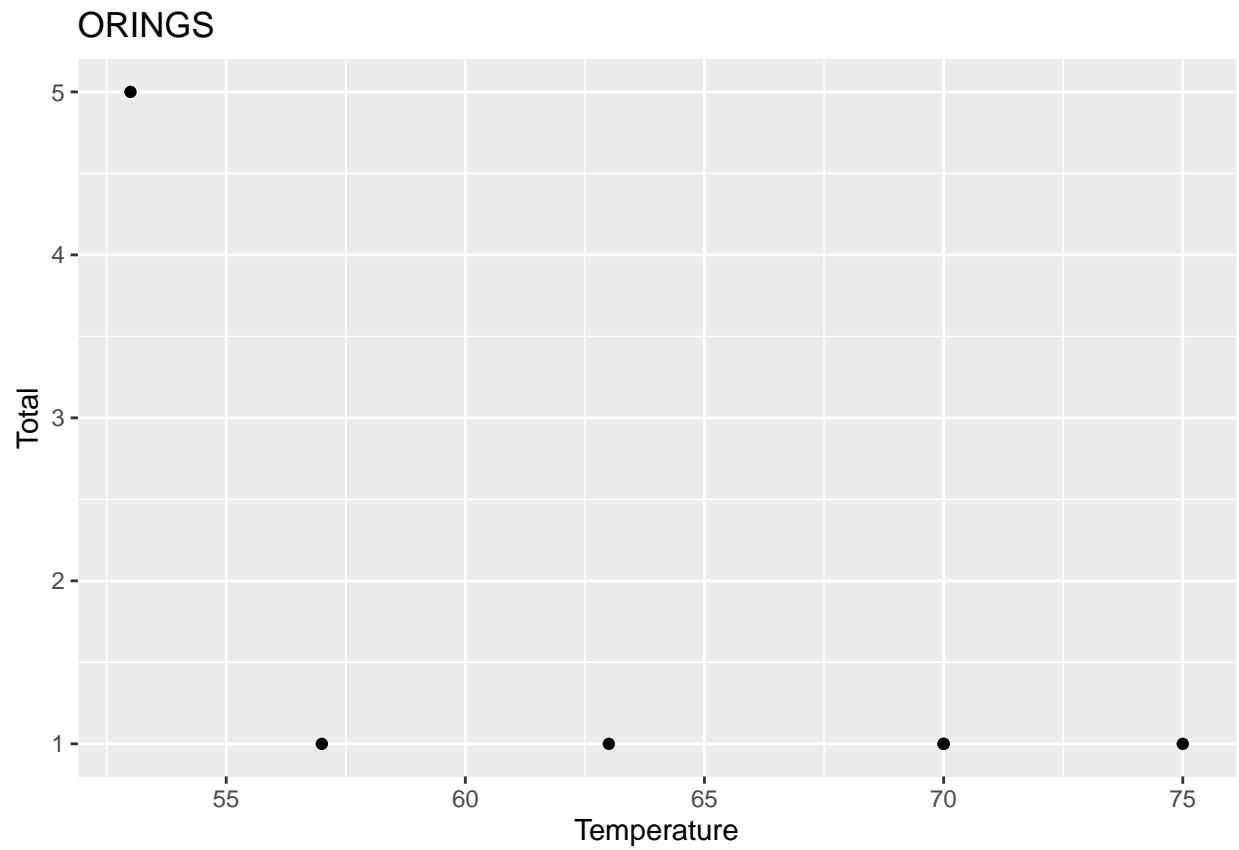
```
oringsdata<-data.frame(orings)  
head(oringsdata)
```

```
##   Temperature Erosion Blowby Total  
## 1           53        3      2     5  
## 2           57        1      0     1  
## 3           58        1      0     1  
## 4           63        1      0     1  
## 5           66        0      0     0  
## 6           67        0      0     0
```

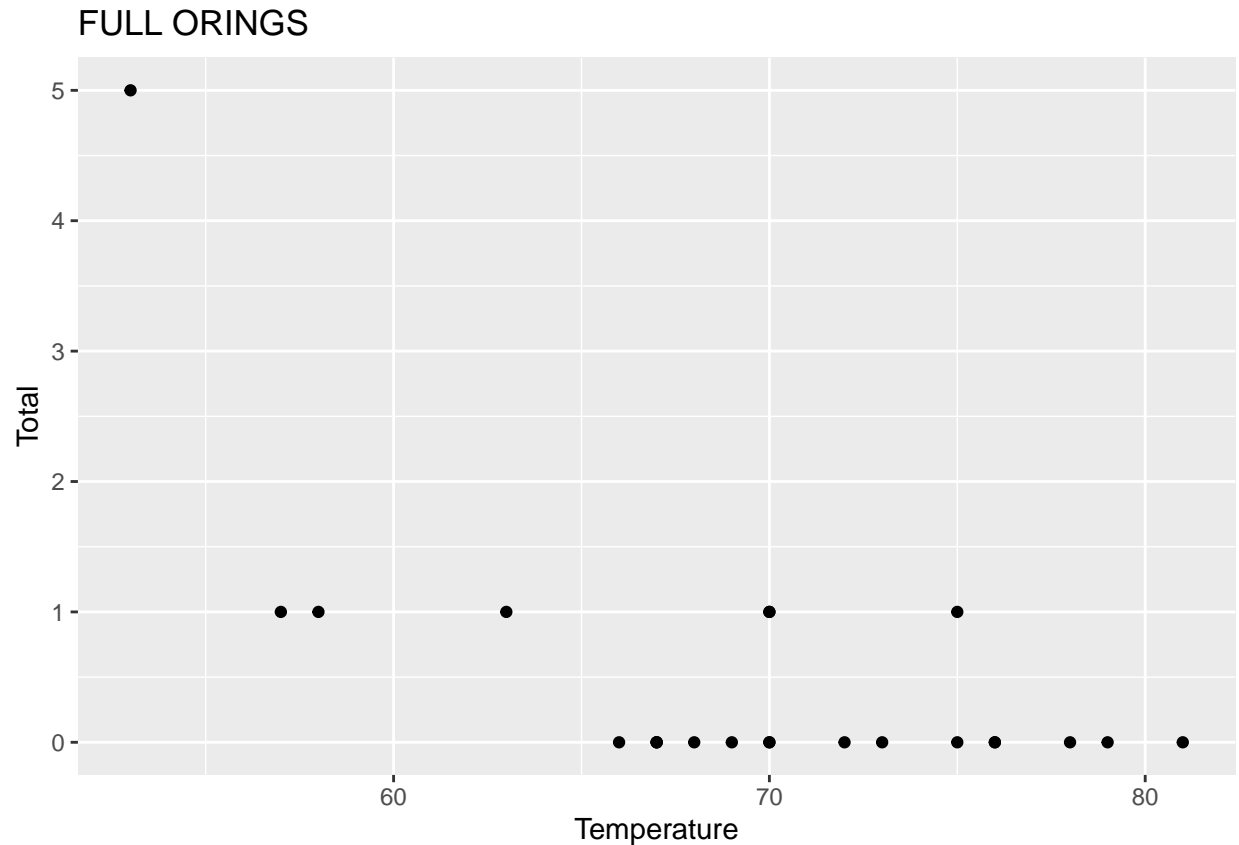
```
neworings<-oringsdata[c(1,2,4,11,13,18),]  
head(neworings)
```

```
##   Temperature Erosion Blowby Total  
## 1           53        3      2     5  
## 2           57        1      0     1  
## 4           63        1      0     1  
## 11          70        1      0     1  
## 13          70        1      0     1  
## 18          75        0      2     1
```

```
ggplot(data = neworings) +
  geom_point(aes(x = Temperature, y = Total)) +
  labs(x = "Temperature",
       y = "Total",
       title = "ORINGS")
```



```
ggplot(data = oringsdata) +
  geom_point(aes(x = Temperature, y = Total)) +
  labs(x = "Temperature",
       y = "Total",
       title = "FULL ORINGS")
```



MB.Ch1.4. For the data frame `ais` (DAAG package)

(a) Use the function `str()` to get information on each of the columns. Determine whether any of the columns hold missing values.

```
library(DAAG)
```

```
str(ais)
```

```
## 'data.frame':   202 obs. of  13 variables:
## $ rcc      : num  3.96 4.41 4.14 4.11 4.45 4.1 4.31 4.42 4.3 4.51 ...
## $ wcc      : num  7.5 8.3 5 5.3 6.8 4.4 5.3 5.7 8.9 4.4 ...
## $ hc       : num  37.5 38.2 36.4 37.3 41.5 37.4 39.6 39.9 41.1 41.6 ...
## $ hg       : num  12.3 12.7 11.6 12.6 14 12.5 12.8 13.2 13.5 12.7 ...
## $ ferr     : num  60 68 21 69 29 42 73 44 41 44 ...
## $ bmi      : num  20.6 20.7 21.9 21.9 19 ...
## $ ssf      : num  109.1 102.8 104.6 126.4 80.3 ...
## $ pcBfat   : num  19.8 21.3 19.9 23.7 17.6 ...
## $ lbm      : num  63.3 58.5 55.4 57.2 53.2 ...
## $ ht       : num  196 190 178 185 185 ...
## $ wt       : num  78.9 74.4 69.1 74.9 64.6 63.7 75.2 62.3 66.5 62.9 ...
## $ sex      : Factor w/ 2 levels "f","m": 1 1 1 1 1 1 1 1 1 ...
## $ sport    : Factor w/ 10 levels "B_Ball","Field",...: 1 1 1 1 1 1 1 1 1 ...
```

```
#is.na(ais)
which(is.na(ais))
```

```
## integer(0)
```

(b) Make a table that shows the numbers of males and females for each different sport. In which sports is there a large imbalance (e.g., by a factor of more than 2:1) in the numbers of the two sexes?

```
ais<-data.frame(ais)
selectais=table(ais$sex,ais$sport)
selectais
```

```
##
##      B_Ball Field Gym Netball Row Swim T_400m T_Sprnt Tennis W_Polo
## f      13      7  4      23  22    9      11      4      7      0
## m      12     12  0       0  15   13     18     11     4     17
```

```
ais2<-t(selectais)
proportion_sport<-ais2[,1]/ais2[,2]
proportion_sport
```

```
##      B_Ball      Field      Gym      Netball      Row      Swim      T_400m      T_Sprnt
## 1.0833333 0.5833333      Inf      Inf 1.4666667 0.6923077 0.6111111 0.3636364
##      Tennis      W_Polo
## 1.7500000 0.0000000
```

```
ifimbanlance<-((ais2[,1]/ais2[,2]>2)|(ais2[,1]/ais2[,2]<0.5))
ifimbanlance
```

```
##      B_Ball      Field      Gym Netball      Row      Swim      T_400m T_Sprnt      Tennis      W_Polo
##      FALSE      FALSE      TRUE      TRUE      FALSE      FALSE      FALSE      TRUE      FALSE      TRUE
```

From the table `proportion_sport` above, we can find out the porportion of female to male.

The table `ifimbanlance` tells that if the proportion is imbalance. If yes, it turns out to be `True`, if no it turns out to be `False`.

## MB.Ch1.6.

Create a data frame called `Manitoba.lakes` that contains the lake's elevation (in meters above sea level) and area (in square kilometers) as listed below. Assign the names of the lakes using the `row.names()` function.

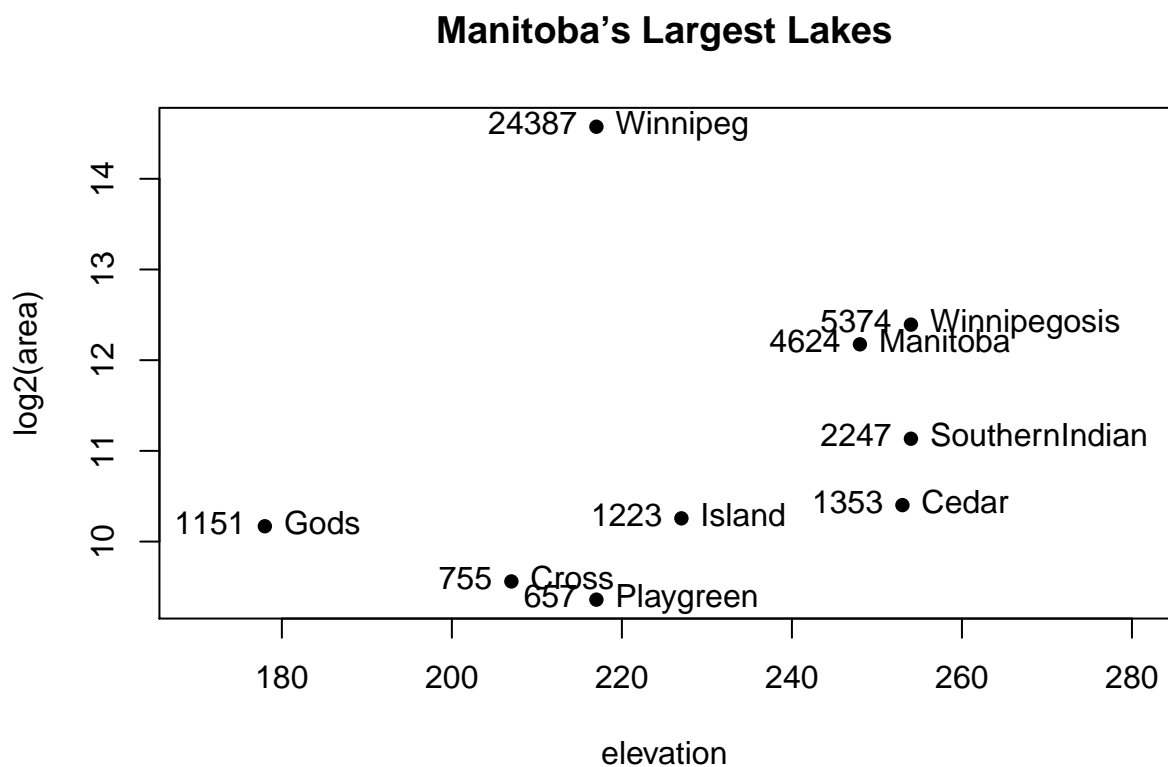
Lake Name	Elevation (m)	Area (km²)
Winnipeg	217	24387
Winnipegosis	254	5374
Manitoba	248	4624
SouthernIndian	254	2247
Cedar	253	1353
Island	227	1223
Gods	178	1151
Cross	207	755
Playgreen	217	657

(a) Use the following code to plot  $\log_2(\text{area})$  versus elevation, adding labeling information (there is an extreme value of area that makes a logarithmic scale pretty much essential):

```
Manitoba.lakes<-data.frame(elevation=c(217,254,248,254,253,227,178,207,217),area=c(24387,5374,4624,2247,1353,1223,1151,755,657),
row.names(Manitoba.lakes)<-c("Winnipeg","Winnipegosis","Manitoba","SouthernIndian","Cedar","Island","Gods","Cross","Playgreen"))
Manitoba.lakes
```

```
##           elevation  area
## Winnipeg          217 24387
## Winnipegosis       254  5374
## Manitoba           248  4624
## SouthernIndian     254  2247
## Cedar              253  1353
## Island             227  1223
## Gods               178  1151
## Cross              207   755
## Playgreen          217   657
```

```
attach(Manitoba.lakes)
plot(log2(area) ~ elevation, pch=16, xlim=c(170,280))
# NB: Doubling the area increases log2(area) by 1.0
text(log2(area) ~ elevation, labels=row.names(Manitoba.lakes), pos=4)
text(log2(area) ~ elevation, labels=area, pos=2)
title("Manitoba's Largest Lakes")
```

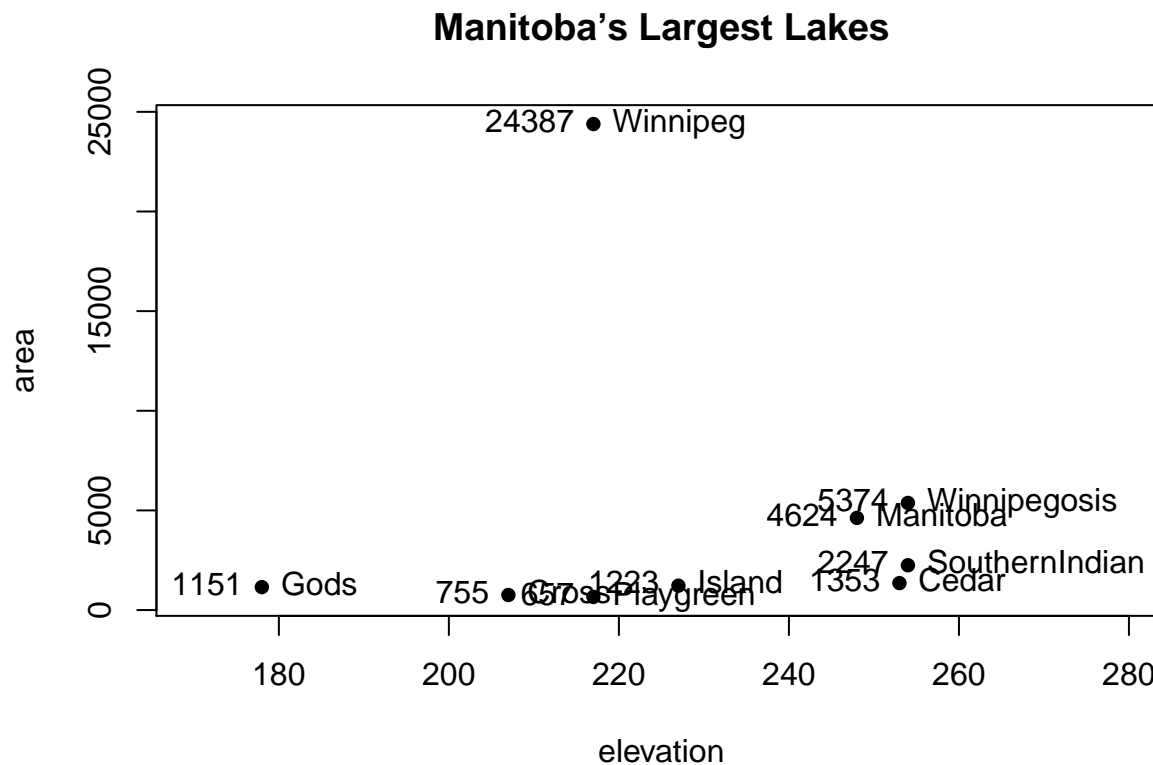


captions that explain the labeling on the points and on the y-axis. It will be necessary to explain how distances on the scale relate to changes in area.



(b) Repeat the plot and associated labeling, now plotting area versus elevation, but specifying `log="y"` in order to obtain a logarithmic y-scale.

```
plot(area ~ elevation, pch=16, xlim=c(170,280), ylog=T)
text(area ~ elevation, labels=row.names(Manitoba.lakes), pos=4,ylog=T)
text(area ~ elevation, labels=area, pos=2, ylog=T)
title("Manitoba's Largest Lakes")
```

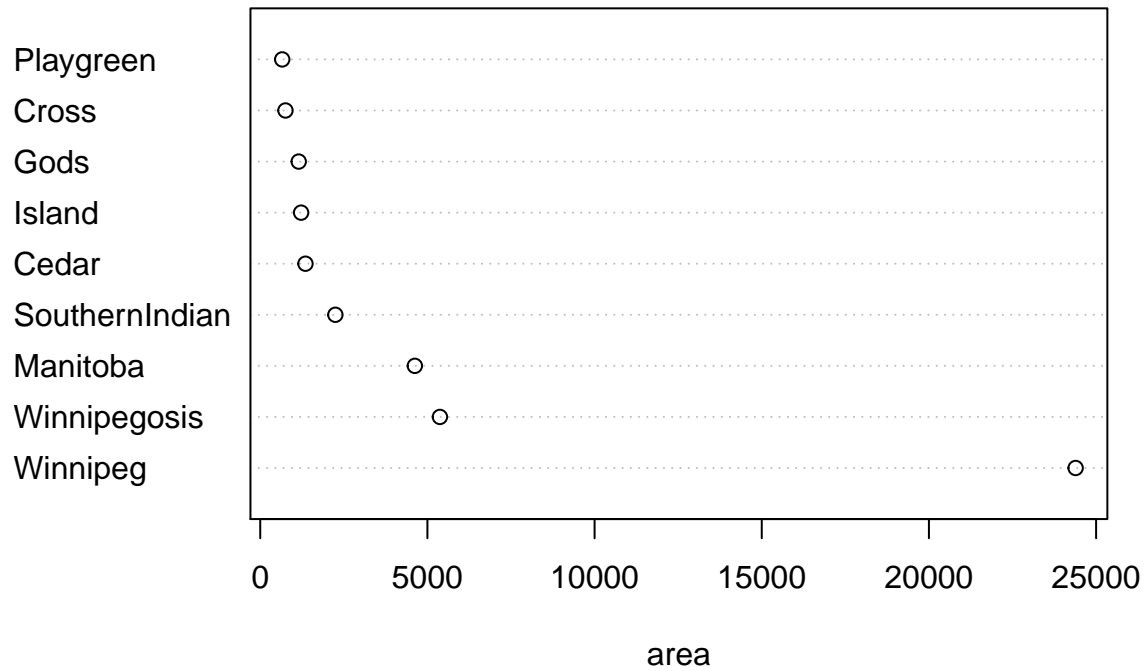


## MB.Ch1.7.

Look up the help page for the R function `dotchart()`. Use this function to display the areas of the Manitoba lakes (a) on a linear scale, and (b) on a logarithmic scale. Add, in each case, suitable labeling information.

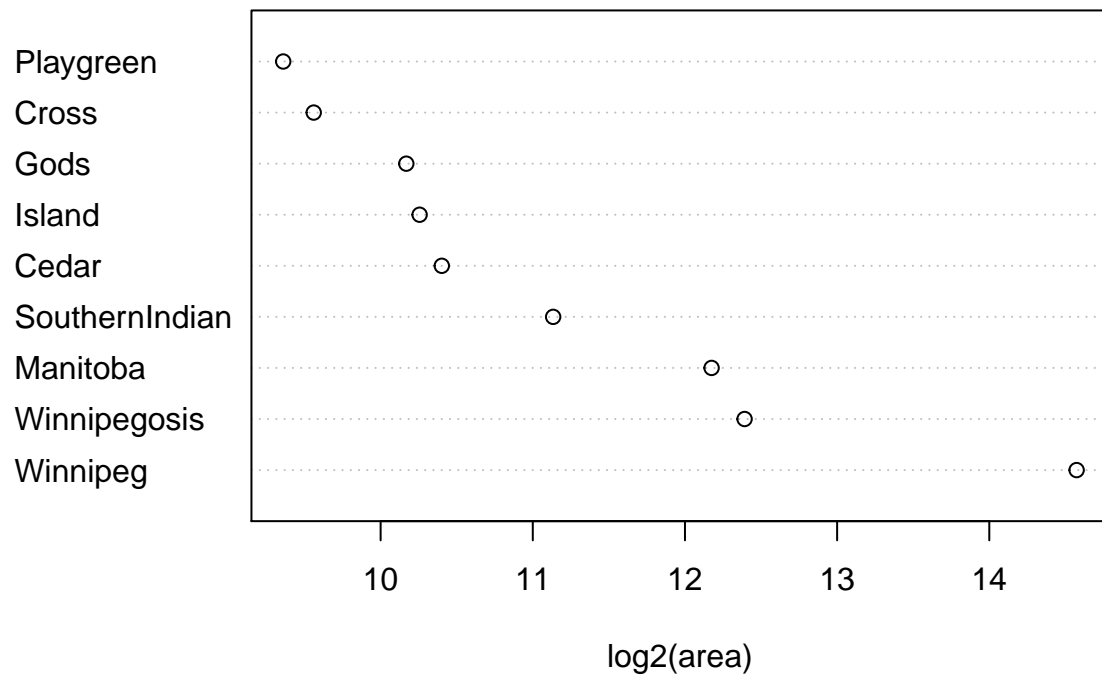
The areas of the Manitoba lakes (a) on a linear scale

```
dotchart(area, labels = row.names(Manitoba.lakes), xlab = "area", ylab = NULL)
```



The areas of the Manitoba lakes (a) on a logarithmic scale

```
dotchart(log2(area), labels = row.names(Manitoba.lakes), xlab = "log2(area)", ylab = NULL)
```



MB.Ch1.8. Using the `sum()` function, obtain a lower bound for the area of Manitoba covered by water.

```
sum(area)
```

```
## [1] 41771
```