

# A Quick Introduction to using R for Data Analysis

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## Contents

1	Basic R Objects and Operations	1
2	Import a dataset into R environment and Simple Operation	3
3	Create your own function	7
4	Include Images Saved in An External File	8

## 1 Basic R Objects and Operations

```
# create a vector
x <- 1:10
x <- seq (30,3, by = -2)
a <- c(66.32, 69.87, 70.12, 90.37, 50.08, 61.20, 65.00, 57.65)
d <- a [1]
a [1] <- 85.34

mean (a)
```

```
## [1] 68.70375
```

```
ma <- mean (a)
# read a vector of numbers from a file
x <- scan("numbers.txt")
x2 <- scan("number2.txt")

# one can also read number without saving to a file
y <- scan(text = "7 8 9 10 11 12 13 13 14 17 17 45")
```

```
# create a matrix
A <- matrix (0, 4, 2)

A <- matrix (1:8, 4,2)
```

```
A
```

```
##      [,1] [,2]
## [1,]    1    5
## [2,]    2    6
## [3,]    3    7
## [4,]    4    8
```

```

D <- matrix (a, 4, 2, byrow=T)

D <- matrix(1:8, 2, 4)
D

##      [,1] [,2] [,3] [,4]
## [1,]    1    3    5    7
## [2,]    2    4    6    8

# create another matrix with all entry 0
B <- matrix (1:5000, 100, 50)

# assign a number to B
B[2,4] <- 45
B[1,]

## [1]    1  101  201  301  401  501  601  701  801  901 1001 1101 1201 1301 1401
## [16] 1501 1601 1701 1801 1901 2001 2101 2201 2301 2401 2501 2601 2701 2801 2901
## [31] 3001 3101 3201 3301 3401 3501 3601 3701 3801 3901 4001 4101 4201 4301 4401
## [46] 4501 4601 4701 4801 4901

B[,1]

## [1]    1    2    3    4    5    6    7    8    9   10   11   12   13   14   15   16   17   18
## [19] 19   20   21   22   23   24   25   26   27   28   29   30   31   32   33   34   35   36
## [37] 37   38   39   40   41   42   43   44   45   46   47   48   49   50   51   52   53   54
## [55] 55   56   57   58   59   60   61   62   63   64   65   66   67   68   69   70   71   72
## [73] 73   74   75   76   77   78   79   80   81   82   83   84   85   86   87   88   89   90
## [91] 91   92   93   94   95   96   97   98   99  100

B[1,] <- 1:50

# create a list
E <- list (newa = a, newA = A)
# list the names of components
names (E)

## [1] "newa" "newA"

# to look at the component of E
E$newA

##      [,1] [,2]
## [1,]    1    5
## [2,]    2    6
## [3,]    3    7
## [4,]    4    8

E$newa <- 10:17

# create a dataframe
scores <- c (30, 45, 50)
names <- c("Peter", "John", "Alice")
stat245_scores <- data.frame (names, scores)
stat245_scores

##   names scores

```

```
## 1 Peter      30
## 2  John      45
## 3 Alice      50

stat245_scores$names

## [1] "Peter" "John" "Alice"

stat245_scores$scores [1] <- 40
stat245_scores

##   names scores
## 1 Peter      40
## 2  John      45
## 3 Alice      50

stat245_scores$perc <- stat245_scores$scores/50 * 100
stat245_scores

##   names scores perc
## 1 Peter      40   80
## 2  John      45   90
## 3 Alice      50  100

stat245_scores$adj <- stat245_scores$perc + 10
stat245_scores

##   names scores perc adj
## 1 Peter      40   80  90
## 2  John      45   90 100
## 3 Alice      50  100 110
```

```
#####
```

## 2 Import a dataset into R environment and Simple Operation

```
#####
```

```
# import myagpop.csv into an R data frame called 'myagpop'
agpop <- read.csv("agpop.csv")
```

```
# Now, we can use the data:
```

```
# preview agpop
```

```
head (agpop)
```

```
##               county state acres92 acres87 acres82 farms92 farms87 farms82
## 1 ALEUTIAN ISLANDS AREA   AK  683533  726596  764514      26      27      28
## 2     ANCHORAGE AREA     AK   47146   59297  256709     217     245     223
## 3     FAIRBANKS AREA     AK  141338  154913  204568     168     175     170
## 4           JUNEAU AREA   AK    210    214    127      8      8      12
## 5 KENAI PENINSULA AREA    AK   50810   85712   98035     93     119     137
## 6     AUTAUGA COUNTY     AL  107259  116050  145044     322     388     453
##   largef92 largef87 largef82 smallf92 smallf87 smallf82 region
## 1      14      16      20        6        4        1      W
## 2       9      10      11       41       52      38      W
## 3      25      28      21       12       18      25      W
```

```
## 4      0      0      0      5      4      8      W
## 5      9     18     17     12     18     19     W
## 6     25     32     32      8     19     17     S

# look at the variable name
colnames (agpop)

## [1] "county"  "state"    "acres92"  "acres87"  "acres82"  "farms92"
## [7] "farms87"  "farms82"  "largef92" "largef87" "largef82" "smallf92"
## [13] "smallf87" "smallf82" "region"

# find number of cols
ncol (agpop)

## [1] 15

# find number of rows
nrow (agpop)

## [1] 3078

# access a certain row
agpop [2, ]

##           county state acres92 acres87 acres82 farms92 farms87 farms82 largef92
## 2 ANCHORAGE AREA   AK   47146  59297  256709     217     245     223         9
##   largef87 largef82 smallf92 smallf87 smallf82 region
## 2        10        11        41        52        38      W

# access a certain column
agpop [1:20, "acres92"] ## equivalent to

## [1] 683533  47146 141338    210  50810 107259 167832 177189  48022 137426
## [11] 144799  96427  73841 109555 121504  99466  67950  61426  68478  47200
agpop$acres92[1:20]

## [1] 683533  47146 141338    210  50810 107259 167832 177189  48022 137426
## [11] 144799  96427  73841 109555 121504  99466  67950  61426  68478  47200
agpop$largef92[1:20]

## [1] 14  9 25  0  9 25 24 40  6  9 29 18  4 22 24  8  9 13  4  5

# find mean of acres92
mean (agpop $acres92)

## [1] 306677

# find sd of acres92
sd (agpop $acres92)

## [1] 424686.7

agpop_AK <- agpop [agpop$state == "AK", ]

agpop_AK <- subset (agpop, state == "AK")

agpop_W <- subset (agpop, region == "W")

agpop_largefarm <- subset (agpop, largef92 > 10)
```

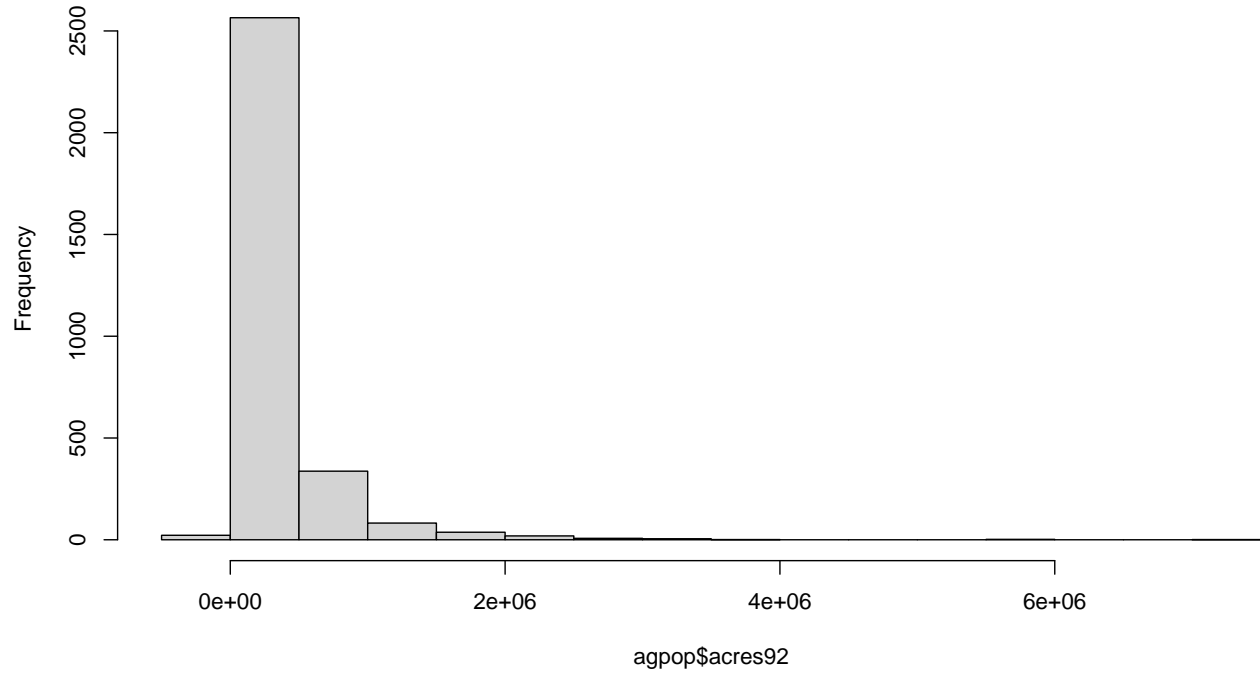
```
## simple analysis
```

```
summary (agpop)
```

```
##      county      state      acres92      acres87
## Length:3078      Length:3078      Min.   :    -99      Min.   :    -99
## Class :character      Class :character      1st Qu.:  80903      1st Qu.:  86236
## Mode  :character      Mode  :character      Median : 191648      Median : 199864
##                                         Mean  : 306677      Mean   : 313016
##                                         3rd Qu.: 366886      3rd Qu.: 372224
##                                         Max.   :7229585      Max.   :7687460
##      acres82      farms92      farms87      farms82
## Min.   :    -99      Min.   :    0.0      Min.   :    0.0      Min.   :    0.0
## 1st Qu.:  96397      1st Qu.: 295.0      1st Qu.: 318.5      1st Qu.: 345.0
## Median : 207292      Median : 521.0      Median : 572.0      Median : 616.0
## Mean   : 320194      Mean   : 625.5      Mean   : 678.3      Mean   : 728.1
## 3rd Qu.: 377065      3rd Qu.: 838.0      3rd Qu.: 921.0      3rd Qu.: 991.0
## Max.   :7313958      Max.   :7021.0      Max.   :7590.0      Max.   :7394.0
##      largef92      largef87      largef82      smallf92
## Min.   :    0.00      Min.   :    0.00      Min.   :    0.00      Min.   :    0.00
## 1st Qu.:    8.00      1st Qu.:    8.00      1st Qu.:    8.00      1st Qu.:   13.00
## Median :   30.00      Median :   27.00      Median :   25.00      Median :   29.00
## Mean   :   56.18      Mean   :   54.86      Mean   :   52.62      Mean   :   54.09
## 3rd Qu.:   75.00      3rd Qu.:   70.00      3rd Qu.:   65.00      3rd Qu.:   59.00
## Max.   :  579.00      Max.   :  596.00      Max.   :  546.00      Max.   : 4298.00
##      smallf87      smallf82      region
## Min.   :    0.00      Min.   :    0.00      Length:3078
## 1st Qu.:   17.00      1st Qu.:   16.00      Class :character
## Median :   35.00      Median :   34.00      Mode  :character
## Mean   :   59.54      Mean   :   60.97
## 3rd Qu.:   67.00      3rd Qu.:   67.00
## Max.   : 3654.00      Max.   : 3522.00
```

```
hist (agpop$acres92)
```

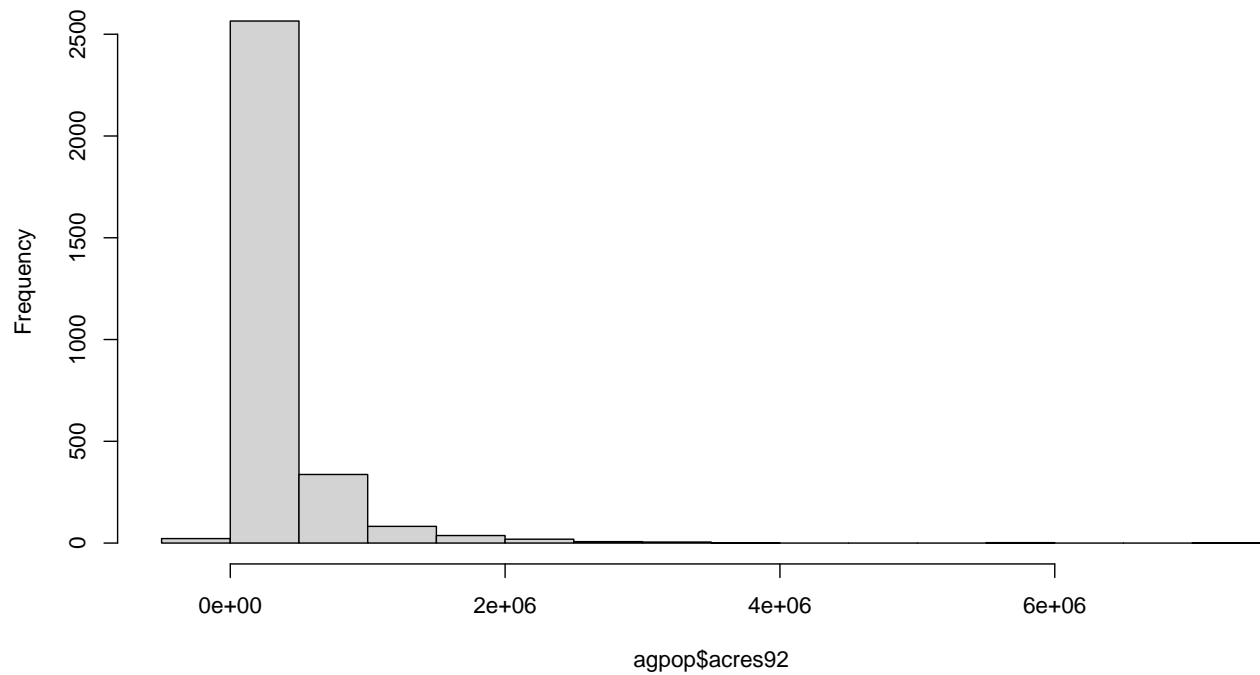
Histogram of agpop\$acres92



Produce Plots

```
#pdf ("hist_acres92.pdf") ## use this command and dev.off to save the output to a file  
hist (agpop$acres92)
```

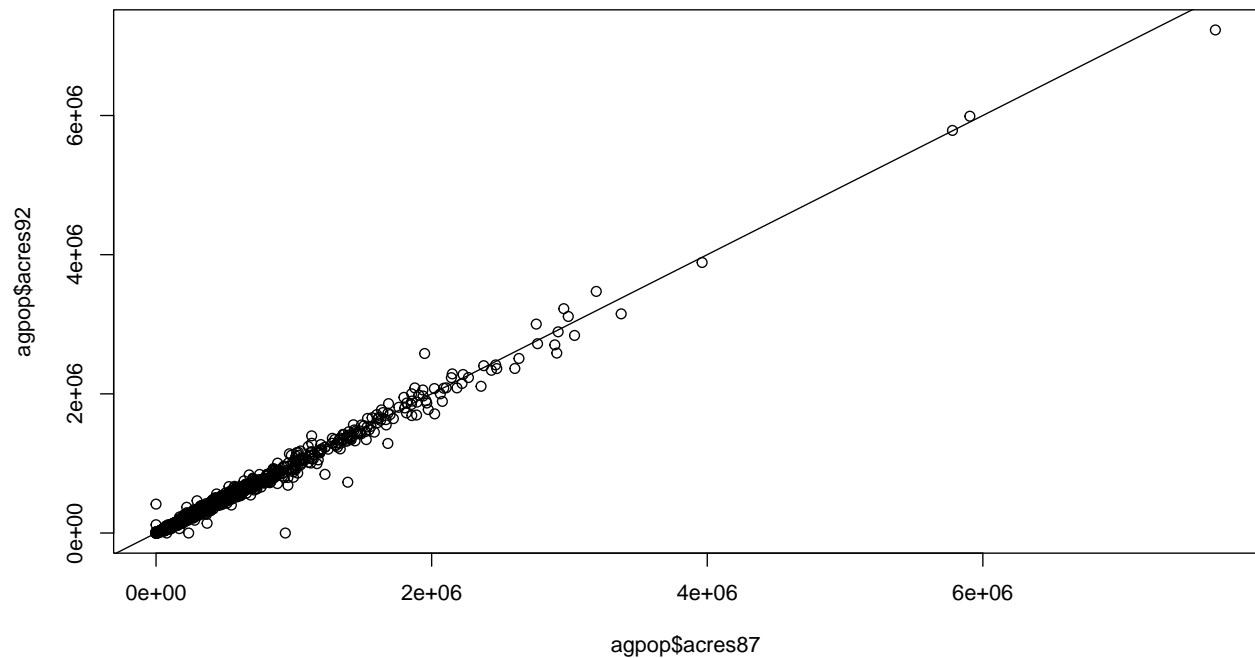
Histogram of agpop\$acres92



```
#dev.off()
```

```
#jpeg ("agpop_acres_87v92.jpg")
```

```
plot (agpop$acres87, agpop$acres92)
abline (a = 0, b = 1)
```



```
#dev.off()## this is used to close the jpeg file
```

### 3 Create your own function

```
## data is a matrix or data.frame
means_col <- function (data)
{
  n <- ncol (data)
  cmeans <- rep (NA, n)
  for (j in 1:n)
  {
    cmeans[j] <- mean (data[,j])
  }
  cmeans
}
```

```
## apply function
```

```
means_col (agpop[, 3:13])
```

```
## [1] 306676.97141 313016.37817 320193.69298 625.50357 678.28428
## [6] 728.06238 56.17674 54.86160 52.62248 54.09227
## [11] 59.53769
```

```
## R built-in function
```

```
colMeans (agpop[, 3:13])
```

```
## acres92 acres87 acres82 farms92 farms87 farms82
```

```
## 306676.97141 313016.37817 320193.69298 625.50357 678.28428 728.06238
## largef92 largef87 largef82 smallf92 smallf87
## 56.17674 54.86160 52.62248 54.09227 59.53769
```

## 4 Include Images Saved in An External File

Using the following R code to include your images saved in an external file.

```
knitr::include_graphics("handwriting.png")
```



Q1:

$$a + b = c$$

Q2 :

$$c = 1 + 2$$

You can hide the above R code by setting “echo=FALSE” for the r chunk. For example, I will include the image once again as follows:

Q1:

$$a + b = c$$

Q2 :

$$c = 1 + 2$$