

# Data

## Get Data

### ENTER RAW DATA:

1. Enter data in Excel (variables in columns, individuals in rows, first row has variable names, no spaces or special characters).
2. Save as "Comma Separated Values (\*.CSV)" file in your local directory/folder.

### DATA PROVIDED BY PROFESSOR:

1. Goto [Data Specific to MTH107 on Resources page](#).
2. Right-click on "data" link and save to your local directory/folder.

## Load CSV

1. Start script and save it in the same folder that contains the CSV file.
2. Select Session, Set Working Directory, To Source File Location menus.
3. Copy resulting `setwd()` code to script.
4. Use `read.csv()` to load data into `dfobj`.

```
dfobj <- read.csv("filename.csv")
```

5. Observe structure of data.frame.

```
str(dfobj)
```

## Filter Individuals

Individuals that meet a certain condition (or conditions) are filtered from the `dfobj` data.frame with `filterD()`.

```
newdf <- filterD(dfobj,cond)
```

where `cond` may be as follows (if `value` is text then it must be in quotes):

```
var == value      # equal to
var != value      # not equal to
var > value       # greater than
var >= value      # greater than or equal
var < value       # less than
var <= value      # less than or equal
var %in% c("val","val","val") # in the list
cond | cond       # either condition met
cond, cond        # both conditions met
```

Individual in row `rownum` is selected with:

```
dfobj[rownum,]
```

Individual in row `rownum` is excluded with:

```
dfobj[-rownum,]
```

# Exploratory Data Analysis

## Univariate

**QUANTITATIVE** – Summary statistics (mean, median, SD, IQR, etc.) and a histogram for the `qvar` variable.

```
hist(~qvar,data=dfobj,xlab="var label")
Summarize(~qvar,data=dfobj,digits=#)
```

**QUANTITATIVE BY GROUP** – Summary statistics and histograms for the `qvar` variable separated by groups in the `fvar` variable.

```
hist(qvar~fvar,data=dfobj,xlab="var label")
Summarize(qvar~fvar,data=dfobj,digits=#)
```

**CATEGORICAL** – Frequency and percentage tables and bar chart for the `fvar` variable.

```
(freq1 <- xtabs(~fvar,data=dfobj))
percTable(freq1,digits=#)
barplot(freq1,xlab="var label",
        ylab="Frequency")
```

## Bivariate

**QUANTITATIVE** – Correlation ( $r$ ) and scatterplot for the `qvarY` and `qvarX` variables.

```
plot(qvarY~qvarX,data=dfobj,
     ylab="yvar label",xlab="xvar label")
corr(~qvarY+qvarX,data=dfobj)
```

**CATEGORICAL** – Frequency and percentage tables for the `fvarRow` and `fvarCol` variables.

```
(freq2 <- xtabs(~fvarRow+fvarCol,
               data=dfobj))
percTable(freq2)      # total/table %
percTable(freq2,margin=1) # row %
percTable(freq2,margin=2) # column %
```

# R CHEATSHEET • MTH107

## Class R FAQ

by Derek H. Ogle, revised Oct-17

## Models

### Normal Distributions

```
distrib(val,mean=meanval,sd=sdval,
        type="q",lower.tail=FALSE)
```

where

- `val` is a value of the quantitative variable or area (i.e., percentage as a proportion).
- `meanval` is population mean ( $\mu$ )
- `sdval` is standard deviation ( $\sigma$ ) or error (SE)
- `type="q"` is included for reverse calculations
- `lower.tail=FALSE` is included for "right-of" calculations

For SE use (where `nval`=sample size):

```
sd=sdval/sqrt(nval)
```

### Linear Regression

The best-fit line between the `respvar` response and `expvar` explanatory variables.

```
(bfl <- lm(respvar~expvar,data=dfobj))
```

A visual of the best-fit line.

```
fitPlot(bfl,ylabel="yvar label",xlab="xvar label")
```

The  $r^2$  value.

```
rSquared(bfl)
```

Predict a value of `respvar` given the `expval` value of `expvar`.

```
predict(bfl,data.frame(expvar=expval))
```

## Hypothesis Testing

### Quantitative

#### ONE SAMPLE:

```
z.test(dfobj$qvar,mu=mu0,alt=HAtype,
       conf.level=confval,sd=sdval)
t.test(dfobj$qvar,mu=mu0,alt=HAtype,
       conf.level=confval)
```

- `qvar` is a quantitative variable in `dfobj`
- `mu0` is the population mean in  $H_0$
- `HAtype` is "two.sided", "less", or "greater" for not equals, less than, and greater than  $H_A$
- `confval` is the confidence level (e.g., 0.95)
- `sdval` is the popn. standard deviation ( $\sigma$ )

#### TWO SAMPLE:

```
levenesTest(qvar~fvar,data=dfobj)
t.test(qvar~fvar,data=dfobj,alt=HAtype,
       conf.level=confval,var.equal=TRUE)
```

- `qvar` is a quantitative variable in `dfobj`
- `fvar` is a factor (categorical) variable in `dfobj`
- `var.equal=TRUE` if the population variances are thought to be equal

### Categorical

#### ONE SAMPLE:

Goodness-of-fit test for observed frequencies in `freq1` and expected values (or proportions) in `exp.p`.

```
(gof <- chisq.test(freq1,p=exp.p,
                  rescale.p=TRUE,correct=FALSE))
```

Extract the expected values.

```
gof$expected
```

Extract the residuals.

```
gof$residuals
```

Follow-up confidence intervals.

```
gofCI(gof,digits=3)
```

#### TWO SAMPLE:

Chi-square for `freq2` two-way observed frequency table.

```
(chi <- chisq.test(freq2,correct=FALSE))
```

Extract the expected values and residuals as for one-sample situation (but using `chi` instead of `gof`).

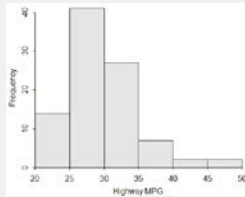
## Data

```
> library(NCStats)
> setwd("C:/aaaWork/Web/GitHub/NCMTH107")
> dfobj <- read.csv("93cars.csv")
> str(dfobj)
'data.frame':   93 obs. of  26 variables:
 $ Type      : Factor w/ 6 levels "Compact","Large",...: 4 ...
 $ HMPG      : int  31 25 26 26 30 31 28 25 27 25 ...
 $ Manual    : Factor w/ 2 levels "No","Yes": 2 2 2 2 1 ...
 $ Weight    : int  2705 3560 3375 3405 3640 2880 3470 ...
 $ Domestic  : Factor w/ 2 levels "No","Yes": 1 1 1 1 1 2 ...

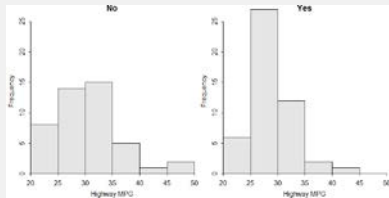
> newdf1 <- filterD(dfobj,Type=="Sporty")
> newdf2 <- filterD(dfobj,HMPG>30)
> newdf3 <- filterD(dfobj,Domestic!="Yes")
> newdf4 <- filterD(dfobj,Type %in% c("Sporty","Small"))
```

## Univariate EDA

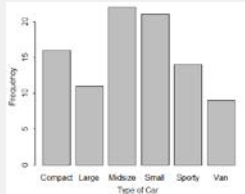
```
> Summarize(~HMPG,data=dfobj,digits=1)
      n mean sd min Q1 median Q3 max
93.0  29.1  5.3 20.0 26.0 28.0 31.0 50.0
> hist(~HMPG,data=dfobj,xlab="Highway MPG")
```



```
> Summarize(HMPG~Domestic,data=dfobj,digits=1)
Domestic n mean sd min Q1 median Q3 max
1 No 45 30.1 6.2 21 25 30 33 50
2 Yes 48 28.1 4.2 20 26 28 30 41
> hist(HMPG~Domestic,data=dfobj,xlab="Highway MPG")
```

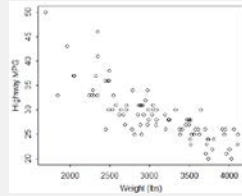


```
> ( freq1 <- xtabs(~Type,data=dfobj) )
Type
Compact Large Midsize Small Sporty Van
16      11      22      21      14      9
> percTable(freq1,digits=1)
Type
Compact Large Midsize Small Sporty Van Sum
17.2    11.8    23.7    22.6    15.1    9.7 100.1
> barplot(freq1,xlab="Type of Car",ylab="Frequency")
```



## Bivariate EDA

```
> plot(HMPG~Weight,data=dfobj,xlab="Weight (lbs)",
      ylab="Highway MPG")
```



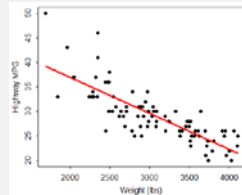
```
> corr(HMPG~Weight,data=dfobj)
[1] -0.8106581

> ( freq2 <- xtabs(~Domestic+Manual,data=dfobj) )
Domestic No Yes
No        6 39
Yes       26 22
> percTable(freq2,digits=1)
Domestic
Manual
Domestic No Yes Sum
No        6.5 41.9 48.4
Yes      28.0 23.7 51.7
Sum      34.5 65.6 100.1
> percTable(freq2,margin=1,digits=1)
Manual
Domestic No Yes Sum
No       13.3 86.7 100.0
Yes      54.2 45.8 100.0
> percTable(freq2,margin=2,digits=1)
Manual
Domestic No Yes
No       18.8 63.9
Yes      81.2 36.1
Sum     100.0 100.0
```

## Linear Regression

```
> ( bfl <- lm(HMPG~Weight,data=dfobj) )
Coefficients:
(Intercept)      Weight
51.601365      -0.007327

> fitPlot(bfl,xlab="Weight (lbs)",ylab="Highway MPG")
```



```
> rSquared(bfl)
[1] 0.6571665
> predict(bfl,data.frame(Weight=3000))
29.62019
```

## Hypothesis Tests

```
> z.test(dfobj$HMPG,mu=26,alt="greater",conf.level=0.95,sd=6)
z = 4.9601, n = 93, Std. Dev = 6.000, Std. Dev of the sample
mean = 0.622, p-value = 3.523e-07
alternative hypothesis: true mean is greater than 26
95 percent confidence interval:
28.06264      Inf
sample estimates:
mean of dfobj$HMPG
29.08602
```

```
> t.test(dfobj$HMPG,mu=26,alt="two.sided",conf.level=0.99)
t = 5.5818, df = 92, p-value = 2.387e-07
alternative hypothesis: true mean is not equal to 26
99 percent confidence interval:
27.63178 30.54026
sample estimates:
mean of x
29.08602
```

```
> levenesTest(HMPG~Domestic,data=dfobj)
Df F value Pr(>F)
group 1 5.3595 0.02286 *
91
```

```
> t.test(HMPG~Manual,data=dfobj,alt="less",conf.level=0.99,
      var.equal=TRUE)
t = -4.2183, df = 91, p-value = 2.904e-05
alt. hypothesis: true difference in means is less than 0
99 percent confidence interval:
-Inf -1.980103
sample estimates:
mean in group No mean in group Yes
26.12500      30.63934
```

```
> exp <- c(1,1,1,1,1,1)/6
> (gof<-chisq.test(freq1,p=exp,rescale.p=TRUE,correct=FALSE))
X-squared = 8.871, df = 5, p-value = 0.1143
```

```
> gof$expected
Compact Large Midsize Small Sporty Van
15.5    15.5    15.5    15.5    15.5    15.5
```

```
> gof$residuals
Compact Large Midsize Small Sporty Van
0.12700 -1.14300 1.65100 1.39700 -0.38100 -1.65100
```

```
> gofCI(gof,digits=3)
p.obs p.LCI p.UCI p.exp
Compact 0.172 0.109 0.261 0.167
Large 0.118 0.067 0.199 0.167
Midsize 0.237 0.162 0.332 0.167
Small 0.226 0.153 0.321 0.167
Sporty 0.151 0.092 0.237 0.167
Van 0.097 0.052 0.174 0.167
```

```
> ( chi <- chisq.test(freq2,correct=FALSE) )
Pearson's Chi-squared test with freq2
X-squared = 17.1588, df = 1, p-value = 3.438e-05
```

```
> chi$expected
Manual
Domestic No Yes
No 15.48387 29.51613
Yes 16.51613 31.48387
```

```
> chi$residuals
Manual
Domestic No Yes
No -2.410160 1.745645
Yes 2.333627 -1.690214
```