



# Data Visualization

<https://blog.modeanalytics.com/r-data-visualization-packages/>

1. Line and Scatter Plots
2. Bar Plots
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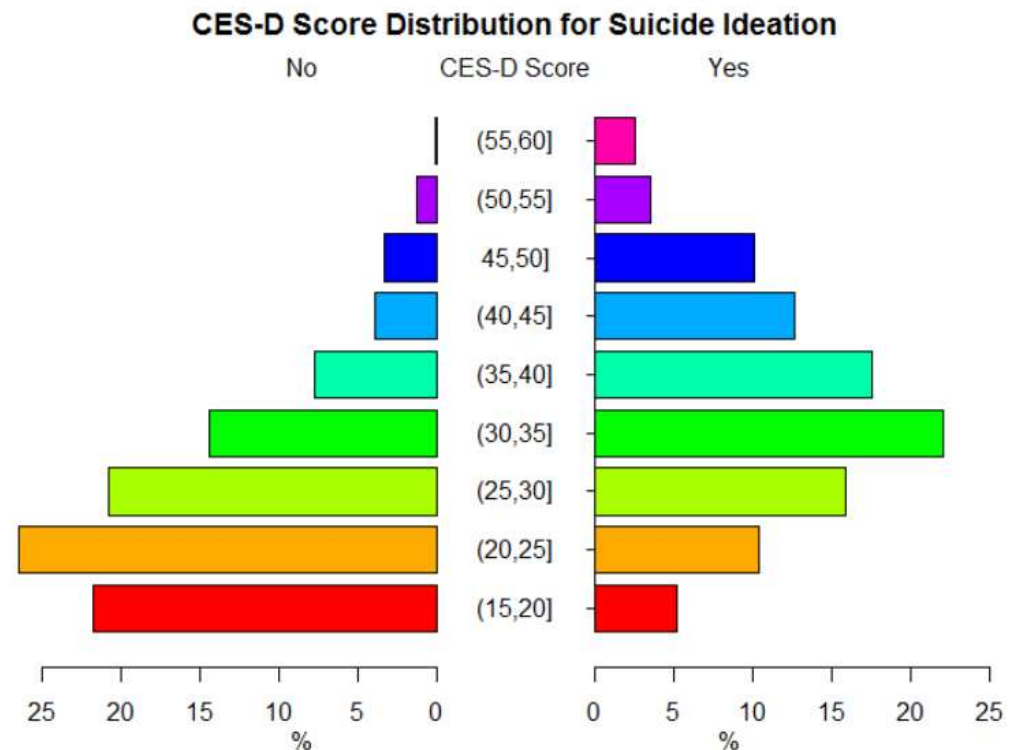
## References

[1] <https://www.r-graph-gallery.com/all-graphs/>

## Data Visualization: What it is and why it matters

[Source] [https://www.sas.com/en\\_us/insights/big-data/data-visualization.html](https://www.sas.com/en_us/insights/big-data/data-visualization.html)

- It enables decision makers to see analytics presented visually, so they can grasp difficult concepts or identify new patterns.
- It makes us to be able to dig for more insights
  - look at data differently, more imaginatively.



## 1. Line and Scatter Plots

**plot**(x, y, pch, type, lty, ...) {graphics}  
Generic function for plotting of R objects.

**points**(x, y, type = "p", ...) {graphics}  
Add Points to a Plot

pch= R plotting symbols.

○	△	+	×	◇	▽	▣	*	⊕	⊗	⊠	⊡	⊢	■	●	▲	◆	●	●	○	□	◇	△	▽	
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25

type = "p" for **p**oints, "l" for **l**ines, "b" for **b**oth,  
"c" for the lines part alone of "b", "o" for both '**o**verplotted',  
"h" for '**h**istogram' like vertical lines, "s" for stair **s**teps,  
"S" for other **s**teps, "n" for no plotting.



**faithful** {datasets} Old Faithful Geyser Data

Waiting time between eruptions and the duration of the eruption for the Old Faithful geyser in Yellowstone National Park, Wyoming, USA.

eruptions : Eruption time (in mins)

waiting : Waiting time to next eruption (in mins)

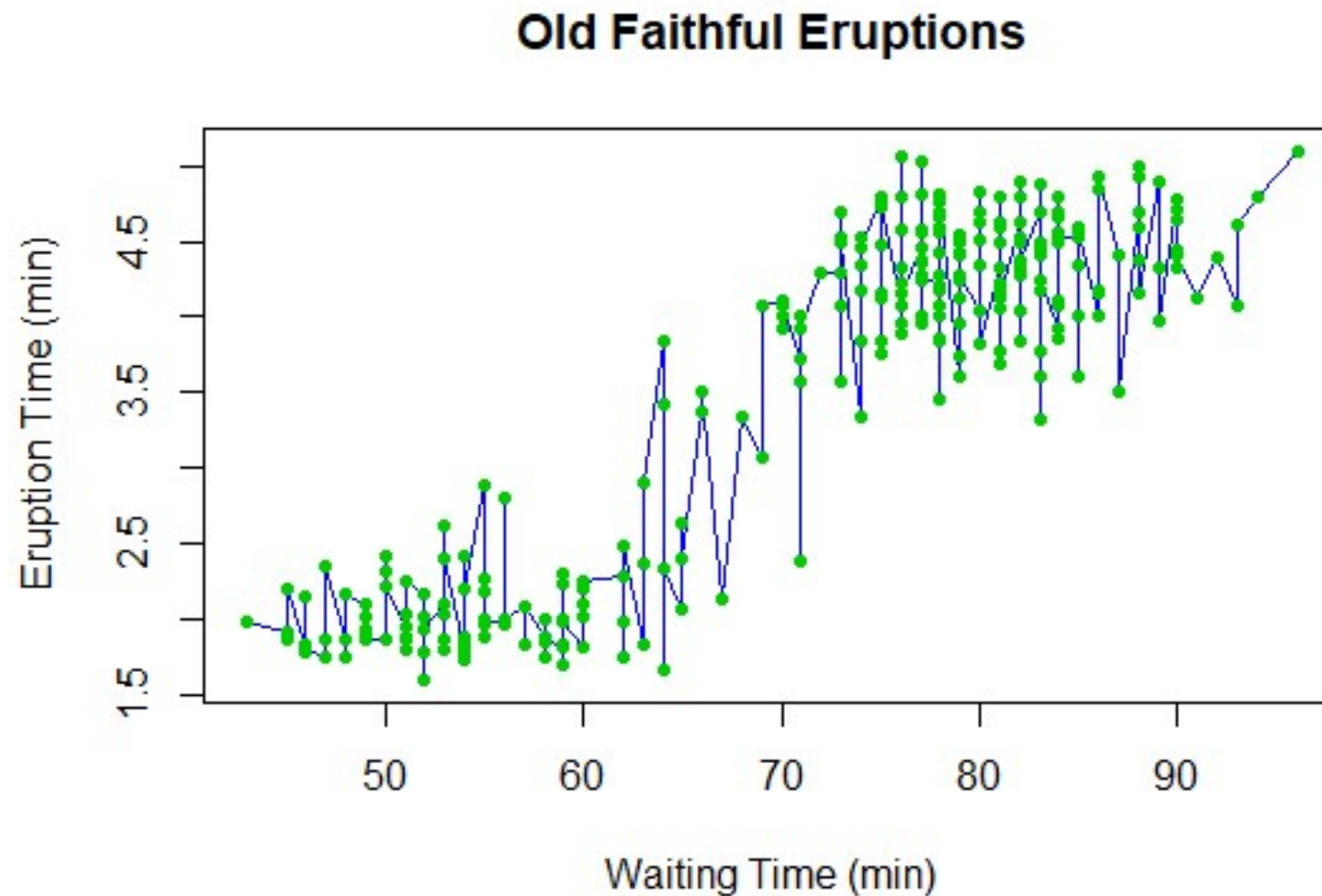
```
> head(faithful)
  eruptions waiting
1      3.600      79
2      1.800      54
3      3.333      74
4      2.283      62
5      4.533      85
6      2.883      55

> #sorting faithful data.frame by waiting
> fa <- faithful[order(faithful$waiting),]
> head(fa)
  eruptions waiting
265      1.983      43
127      1.917      45
131      1.867      45
161      2.200      45
135      1.833      46
188      1.833      46
```





```
x <- fa[,2]; y <- fa[,1]  
plot(x, y, type='l', col=4, xlab='Waiting Time (min)',  
      ylab='Eruption Time (min)', main='Old Faithful Eruptions')  
points(x,y,pch=20,col=3)
```



```
ggplot(data, mapping=aes(), ... ) {ggplot2}
```

Create a new ggplot

geom\_line() Connect the observations, ordered by x value.

geom\_point() Points, as for a scatterplot.

```
#(2)
```

```
library(ggplot2)
```

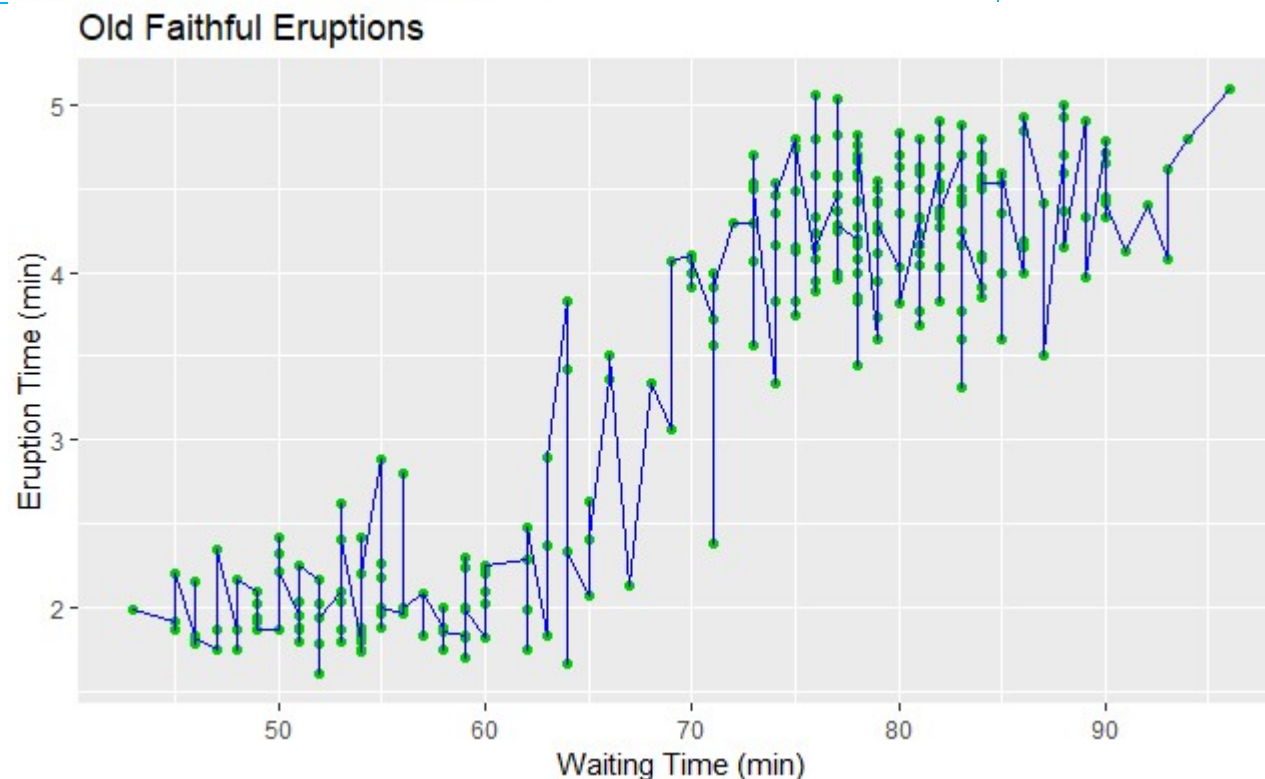
```
ggplot(fa, aes(x, y), xtitle='Waiting Time (min)') +
```

```
  geom_point(col=3) + geom_line(col=4) +
```

```
  xlab('Waiting Time (min)') +
```

```
  ylab('Eruption Time (min)') +
```

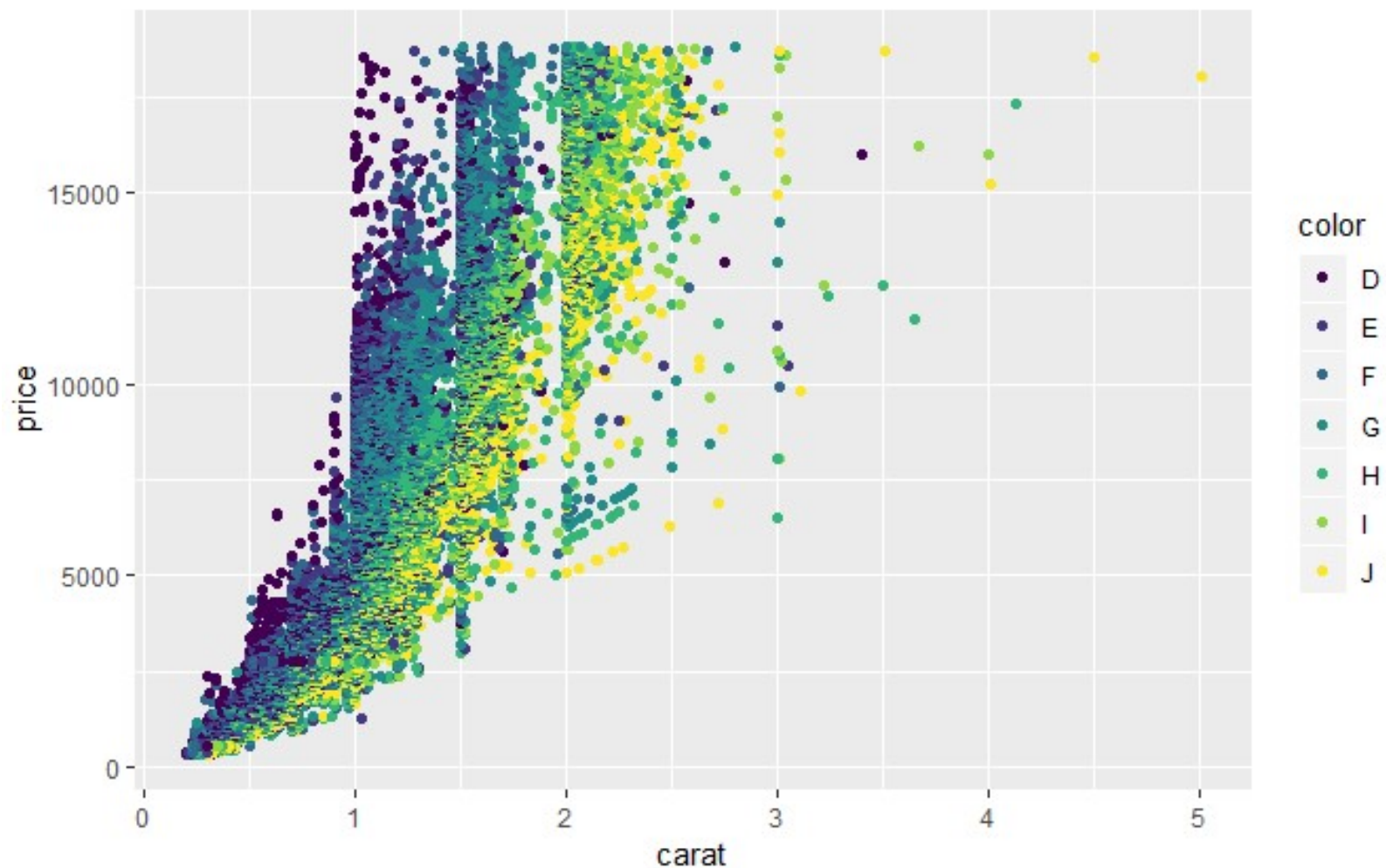
```
  ggtitle('Old Faithful Eruptions')
```



## **diamonds** {ggplot2}

A dataset containing the prices and other attributes of almost 54,000 diamonds.

```
#(3)  
head(diamonds)  
ggplot(data=diamonds, aes(x=carat, y=price)) +  
  geom_point(aes(color=color))
```



## 2. Bar Plots

**barplot**(height, ... ) {graphics}

Creates a bar plot with vertical or horizontal bars.

### (1) Simple Bar Plot

**cars** {datasets}

Speed and Stopping Distances of Cars

speed : speed (mph)

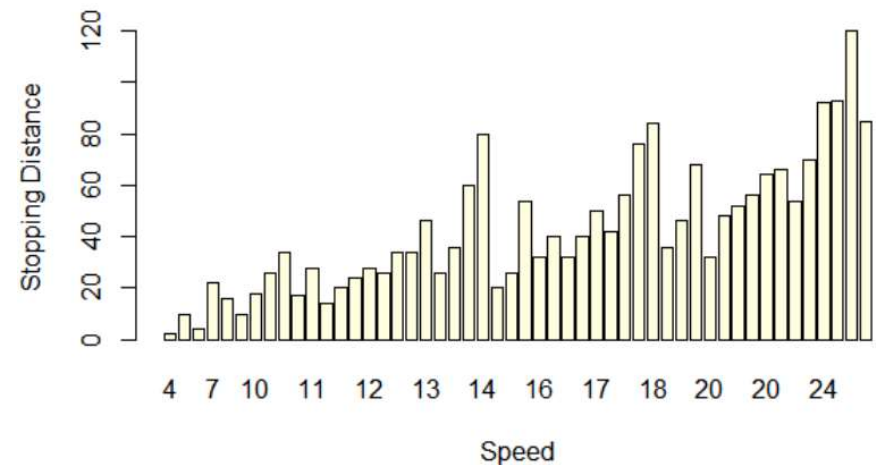
dist : stopping distance (ft)

[Source] Ezekiel, M. (1930) Methods of Correlation Analysis. Wiley.

```
> head(cars)
```

	speed	dist
1	4	2
2	4	10
3	7	4
4	7	22
5	8	16
6	9	10

```
barplot(cars[,2], col="cornsilk",  
        names.arg=cars[,1],  
        xlab="Speed",  
        ylab="Stopping Distance")
```





## (2) Grouped Bar plot

**sleep** {datasets}

Data which show the effect of two soporific drugs (increase in hours of sleep compared to control) on 10 patients.

extra: increase in hours of sleep

group : drug given

ID : patient ID



```
> attach(sleep)
> sleep #list data
```

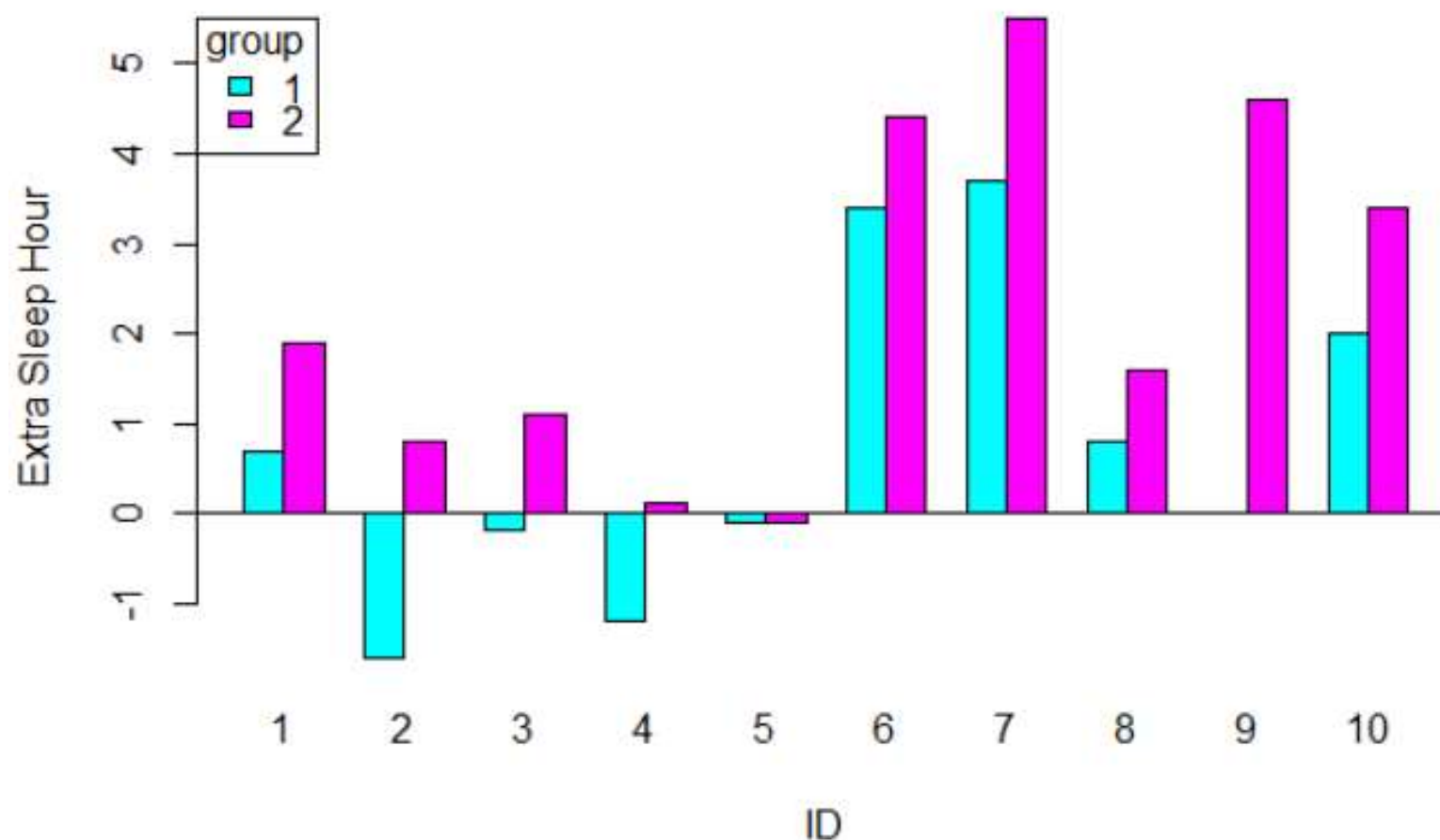
	extra	group	ID
1	0.7	1	1
2	-1.6	1	2
3	-0.2	1	3
4	-1.2	1	4
5	-0.1	1	5
6	3.4	1	6
7	3.7	1	7
8	0.8	1	8
9	0.0	1	9
10	2.0	1	10
11	1.9	2	1
12	0.8	2	2
13	1.1	2	3
14	0.1	2	4
15	-0.1	2	5
16	4.4	2	6
17	5.5	2	7
18	1.6	2	8
19	4.6	2	9
20	3.4	2	10

```
> y <- rbind(extra[1:10],extra[11:20])
```

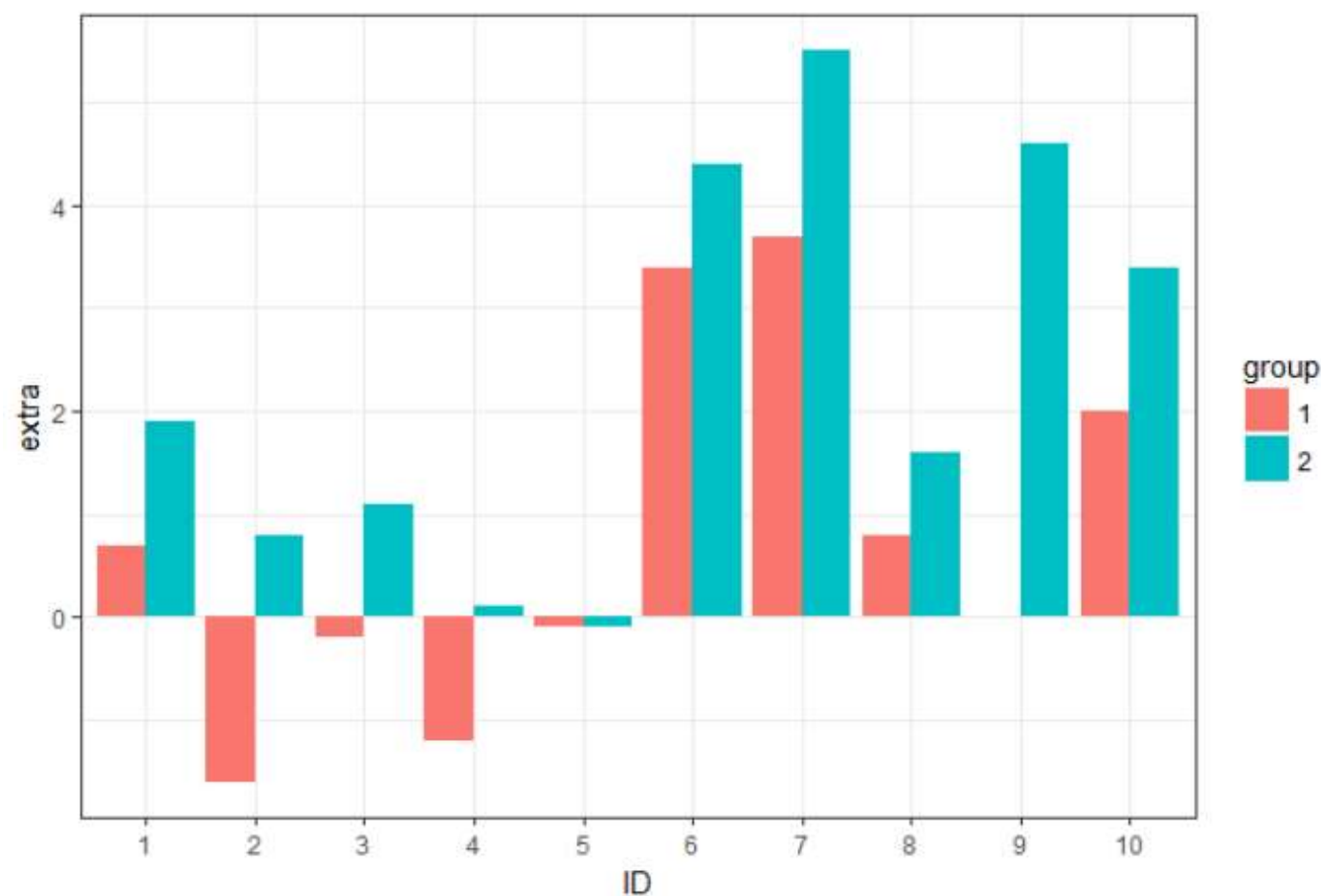
```
> y
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]
[1,]	0.7	-1.6	-0.2	-1.2	-0.1	3.4	3.7	0.8	0.0	2.0
[2,]	1.9	0.8	1.1	0.1	-0.1	4.4	5.5	1.6	4.6	3.4

```
barplot(y, beside=T, col=5:6, names.arg=ID[1:10],  
        xlab="ID",ylab="Extra Sleep Hour")  
abline(h=0)  
legend('topleft', title='group', legend=1:2, fill=5:6)
```



```
#(2) visualization in ggplot2  
library(ggplot2)  
ggplot(sleep, aes(x=ID, y=extra, fill=group)) +  
  geom_bar(stat="identity", position="dodge") + theme_bw()
```



## (3) Bar Plot of Matrix/Table Data

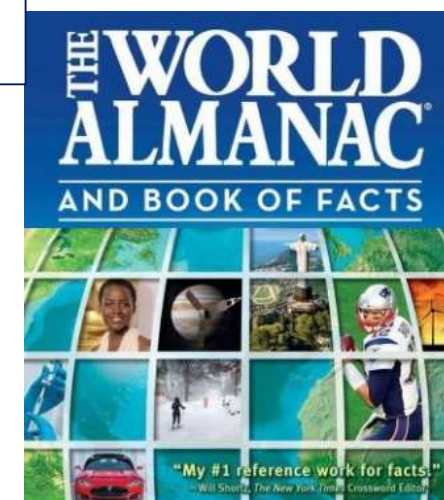
**USPersonalExpenditure** {datasets} Personal Expenditure Data

This data set consists of United States personal expenditures (in billions of dollars) in the categories; food and tobacco, household operation, medical and health, personal care, and private education for the years 1940, 1945, 1950, 1955 and 1960.

[Source] The World Almanac and Book of Facts, 1962, page 756.

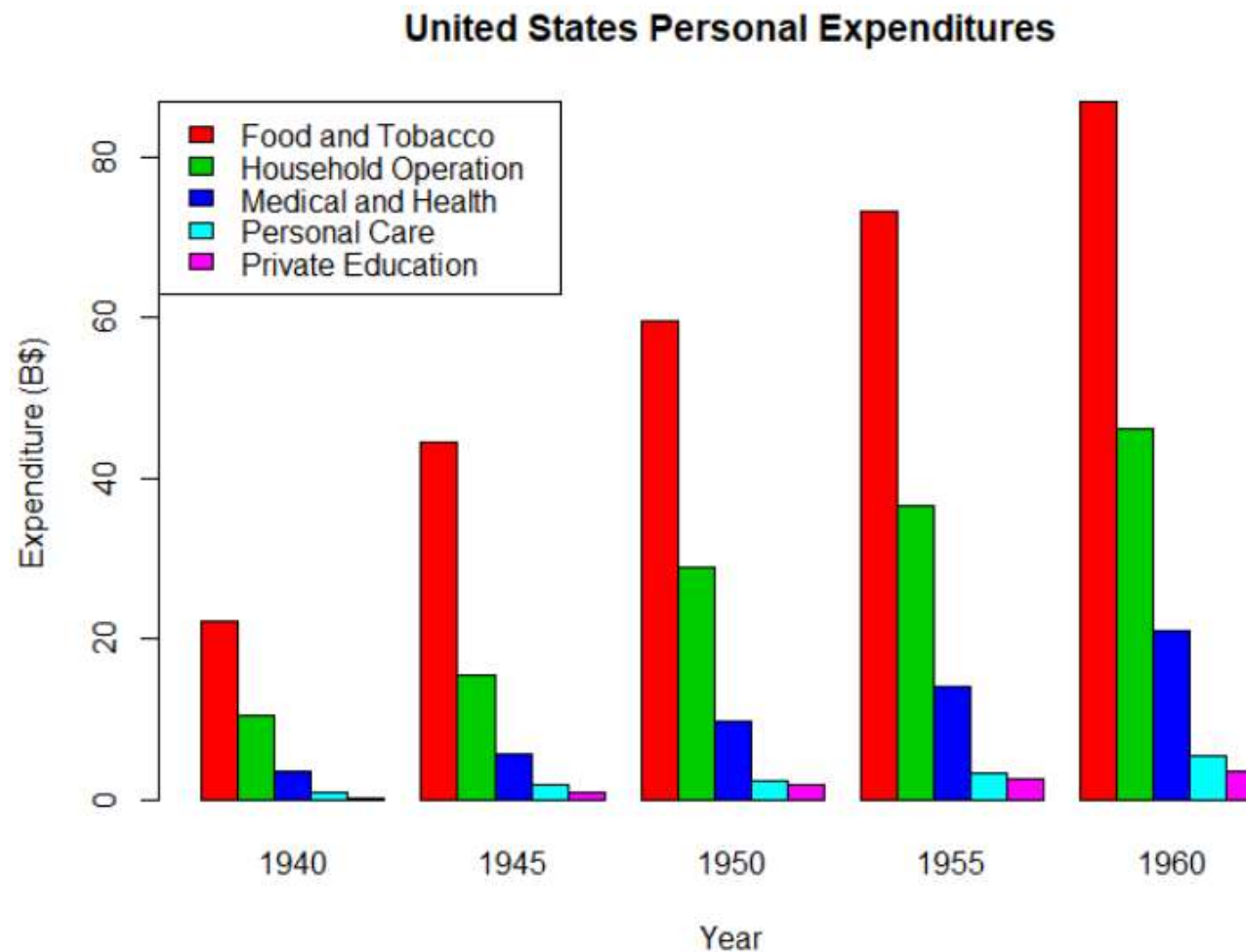
```
> data(USPersonalExpenditure)
> UPE <- USPersonalExpenditure
> UPE
```

	1940	1945	1950	1955	1960
Food and Tobacco	22.200	44.500	59.60	73.2	86.80
Household Operation	10.500	15.500	29.00	36.5	46.20
Medical and Health	3.530	5.760	9.71	14.0	21.10
Personal Care	1.040	1.980	2.45	3.4	5.40
Private Education	0.341	0.974	1.80	2.6	3.64





```
barplot(UPE, beside=T, col=2:6, ylab="Expenditure (B$)",  
        xlab='Year', main='United States Personal Expenditures')  
legend('topleft', legend=row.names(UPE), fill=2:6)
```



### 3. Pie Charts

**pie**(x, labels, ... ) {graphics}  
Draw a pie chart.

**Table 1** Description of the demographic factors (%) of individuals whose body mass index (BMI) was  $\geq 30$

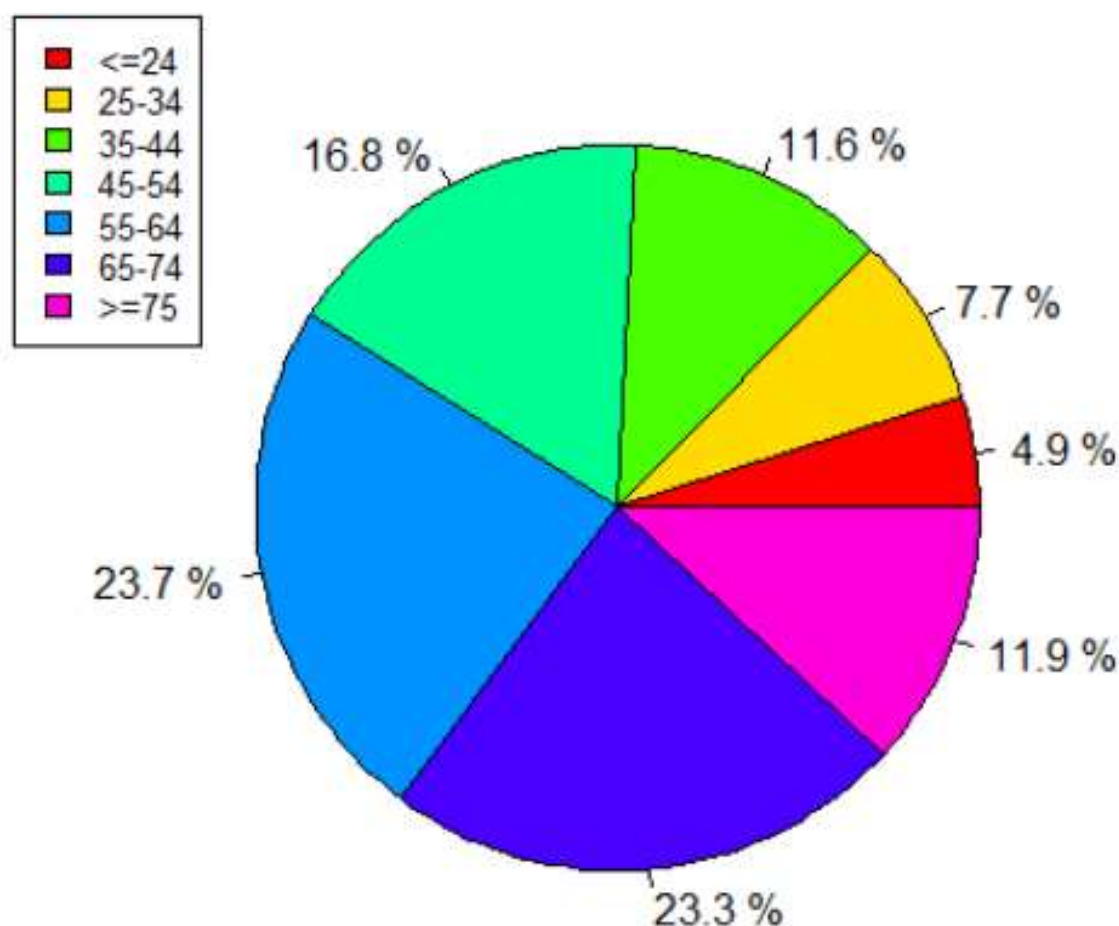
<i>Variable</i>	<i>Obese sample (BMI <math>\geq 30</math>)</i>
Age	
≤24	4.9
25–34	7.7
35–44	11.6
45–54	16.8
55–64	23.7
65–74	23.3
≥75	11.9

[Source] M. A. Green et al., *Journal of Public Health*, **38** (2016) pp. 258 –264

```
#(1) Simple pie chart
#sample data
Age <- c('≤24', '25-34', '35-44', '45-54', '55-64', '65-74', '≥75')
Obese <- c(4.9, 7.7, 11.6, 16.8, 23.7, 23.3, 11.9)
```

```
#plot pie chart  
pie(Obese, labels=paste(Obese,'%'), main='Obesity Percents by Age Group',  
    col=rainbow(length(Age)))  
legend("topleft", Age, cex=0.8, fill=rainbow(length(Age)))
```

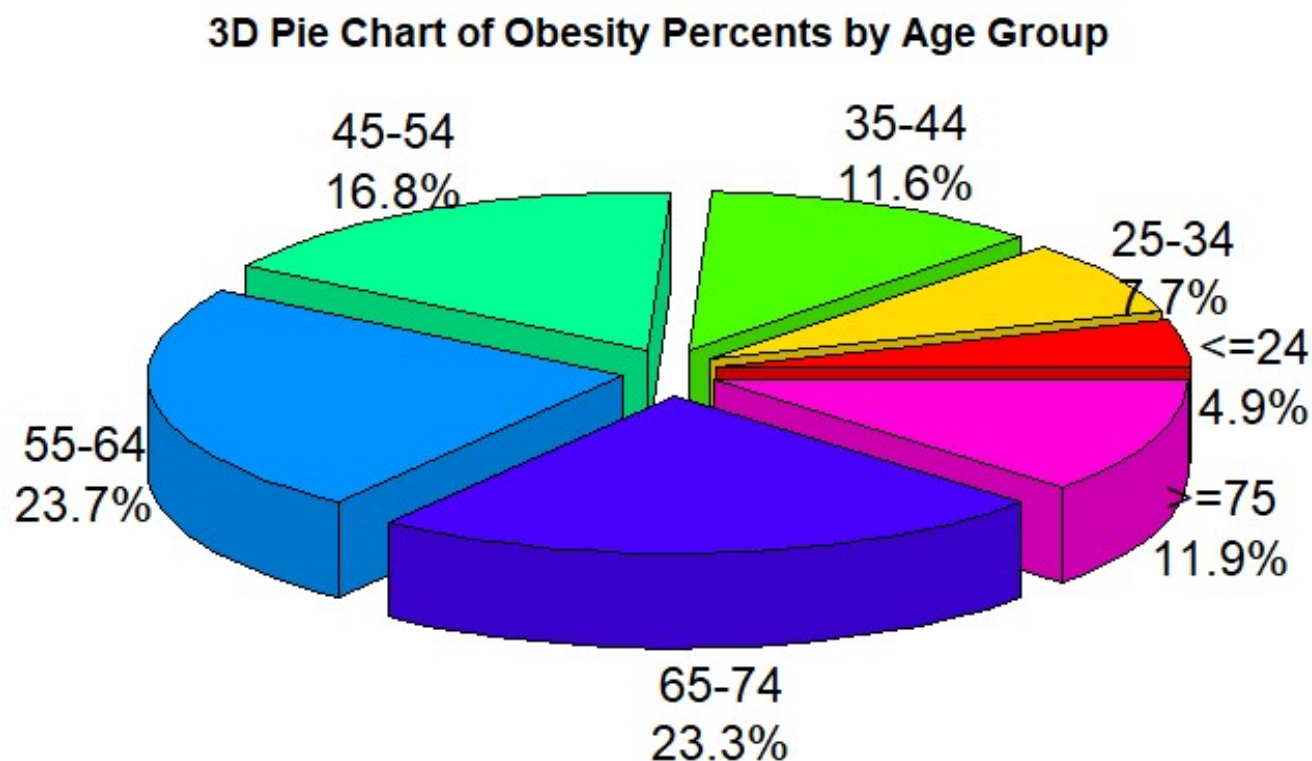
**Obesity Percents by Age Group**



**pie3D**(x, labels, explode, col, ... ) {plotrix}

Displays a 3D pie chart with optional labels.

```
library(plotrix)
xb <- paste(Age, "\n", Obese, '%', sep="")
pie3D(Obese, labels=xb, explode=0.1,
      col=rainbow(length(Age)),
      main="3D Pie Chart of Obesity Percents by Age Group")
```





## 4. Histograms

**hist**(x, breaks, freq, ... ) {graphics}

The hist( ) function computes a histogram of the given data values.

**cane** {boot} Sugar-cane Disease Data

n : The total number of shoots in each plot.

r : The number of diseased shoots.

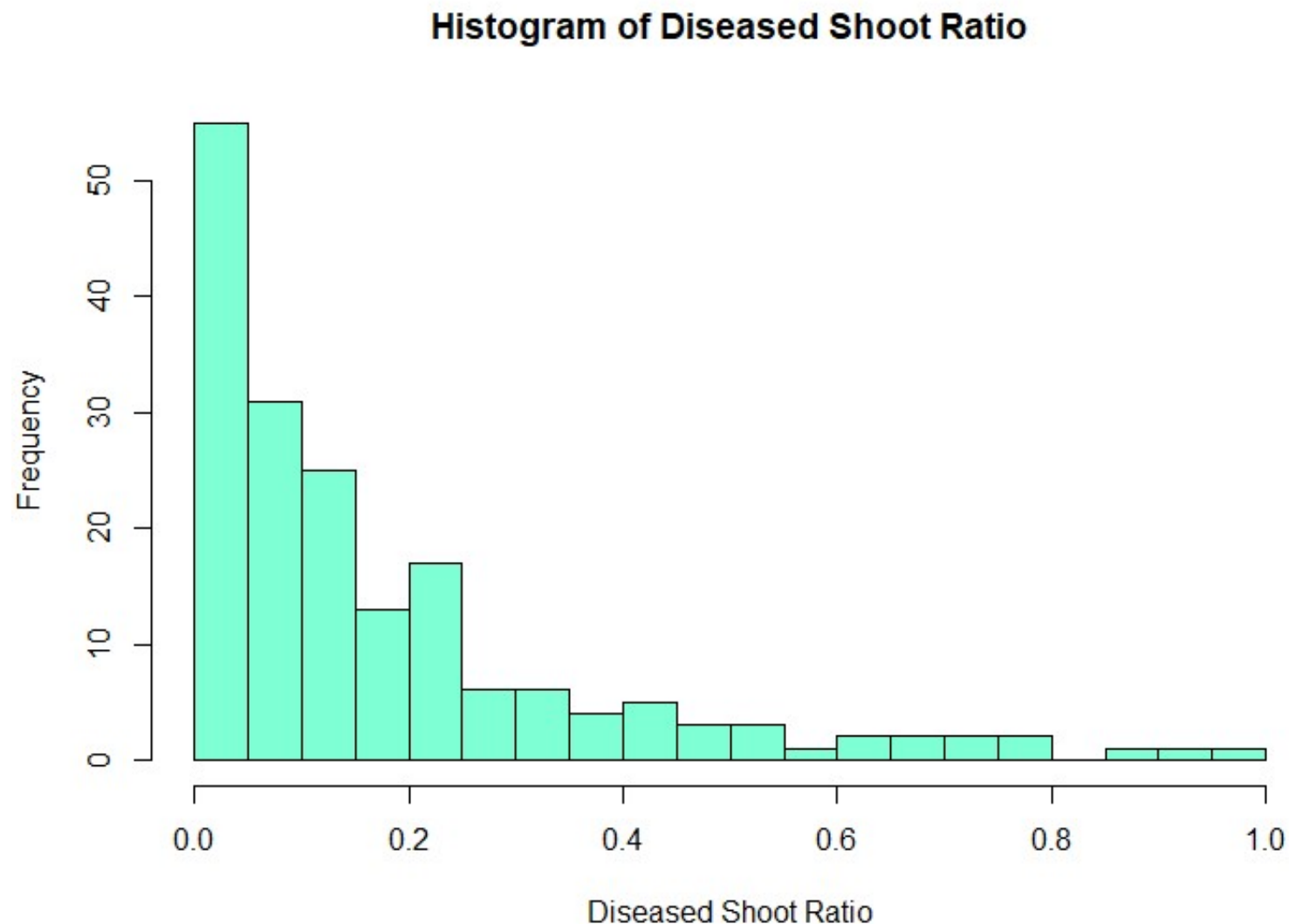
x : The number of pieces of the stems, out of 50

var : A factor indicating the variety of sugar-cane in each plot.

block : A factor for the blocks.

```
> #sample data: cane
> library(boot)
> dim(cane)
[1] 180 5
> str(cane)
'data.frame': 180 obs. of 5 variables:
 $ n : num 87 119 94 95 134 92 118 70 128 85 ...
 $ r : num 76 8 74 11 0 0 11 32 33 14 ...
 $ x : num 19 14 9 12 12 3 17 3 3 21 ...
 $ var : Factor w/ 45 levels "1","10","11",...: 1 12 23 34 41 42 43 44 45 2 ...
 $ block: Factor w/ 4 levels "A","B","C","D": 1 1 1 1 1 1 1 1 1 1 ...
> head(cane)
   n  r  x var block
1 87 76 19 1  A
2 119 8 14 2  A
3 94 74 9 3  A
4 95 11 12 4  A
5 134 0 12 5  A
6 92 0 3 6  A
```

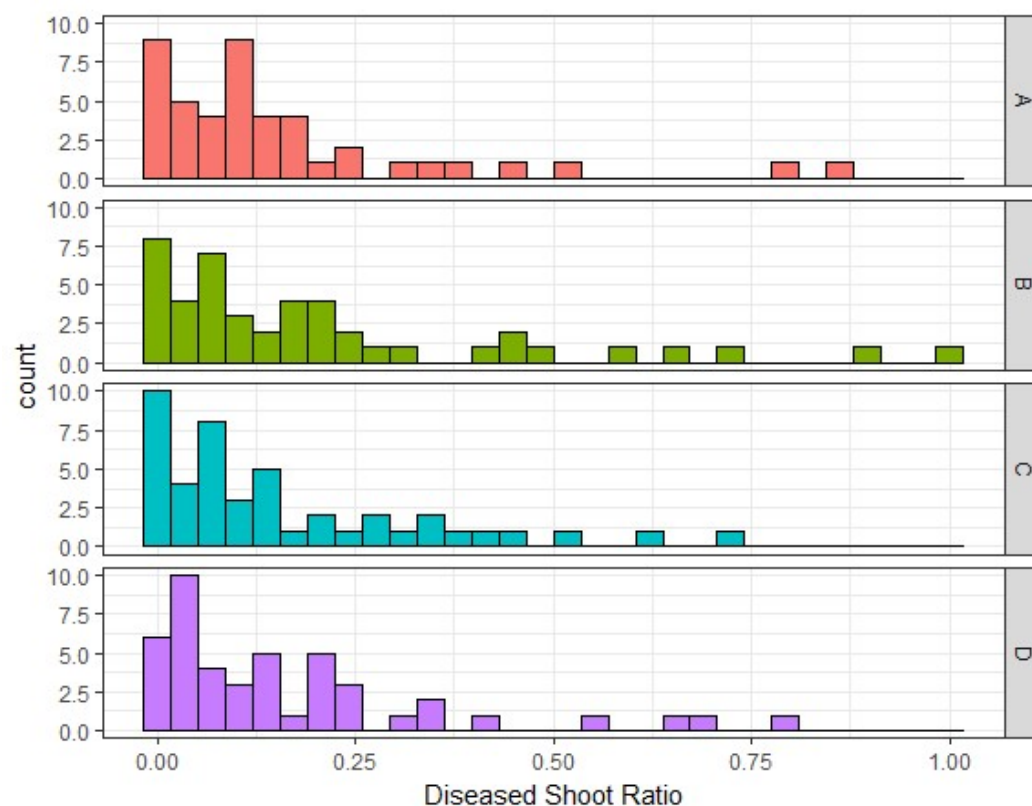
```
#(1) Simple Histogram  
ratio <- cane$r/cane$n  
hist(ratio, breaks=20, xlab='Diseased Shoot Ratio', col='aquamarine',  
     main='Histogram of Diseased Shoot Ratio')
```



**facet\_grid()** {ggplot2}

Lay out panels in a rectangular/tabular manner.

```
#(2)Histograms of Diseased Shoot Ratio by block
library(ggplot2)
ggplot(cane, aes(x=r/n, fill=block)) +
  geom_histogram(colour="black") + theme_bw() +
  facet_grid(block ~ .) + xlab('Diseased Shoot Ratio') +
  theme(legend.position="none")
```



## 5. Bubble Plots

### Sample Data: USArrests

**USArrests** {datasets}      Violent Crime Rates by US State

This data set contains statistics, in arrests per 100,000 residents for assault, murder, and rape in each of the 50 US states in 1973.

Murder : Murder arrests (per 100,000)

Assault : Assault arrests (per 100,000)

UrbanPop : Percent urban population

Rape : Rape arrests (per 100,000)

```
> attach(USArrests)
```

```
> head(USArrests)
```

	Murder	Assault	UrbanPop	Rape
Alabama	13.2	236	58	21.2
Alaska	10.0	263	48	44.5
Arizona	8.1	294	80	31.0
Arkansas	8.8	190	50	19.5
California	9.0	276	91	40.6
Colorado	7.9	204	78	38.7

```
> summary(USArrests)
```

Murder		Assault		UrbanPop		Rape	
Min.	: 0.800	Min.	: 45.0	Min.	: 32.00	Min.	: 7.30
1st Qu.:	4.075	1st Qu.:	109.0	1st Qu.:	54.50	1st Qu.:	15.07
Median :	7.250	Median :	159.0	Median :	66.00	Median :	20.10
Mean :	7.788	Mean :	170.8	Mean :	65.54	Mean :	21.23
3rd Qu.:	11.250	3rd Qu.:	249.0	3rd Qu.:	77.75	3rd Qu.:	26.18
Max.	: 17.400	Max.	: 337.0	Max.	: 91.00	Max.	: 46.00



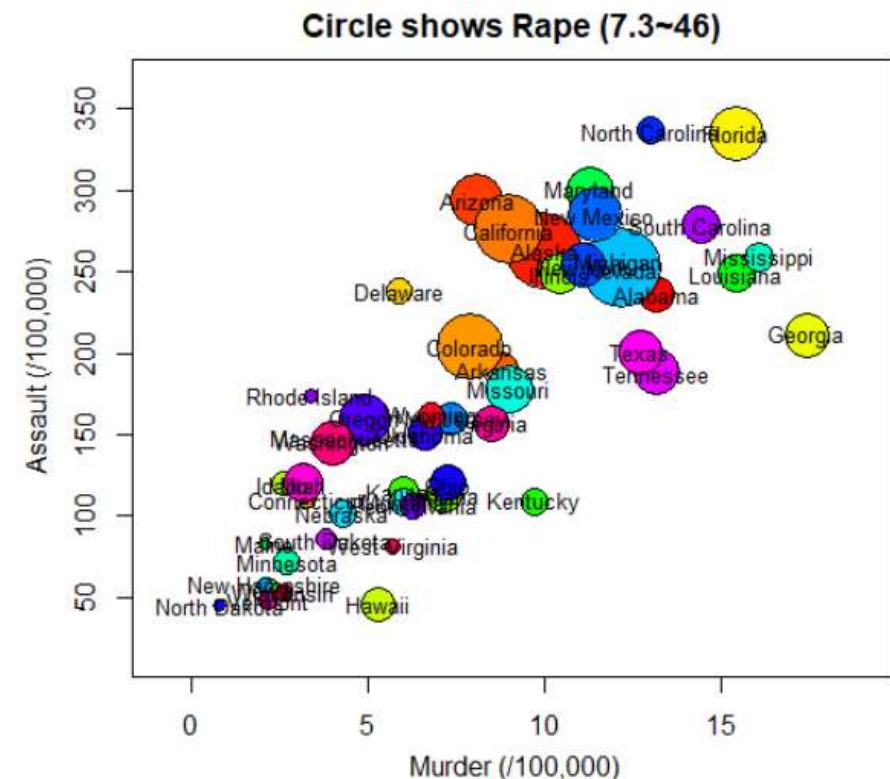
**symbols**(x, y, circles, ...) {graphics}

Draw Symbols (Circles, Squares, Stars, Thermometers, Boxplots)

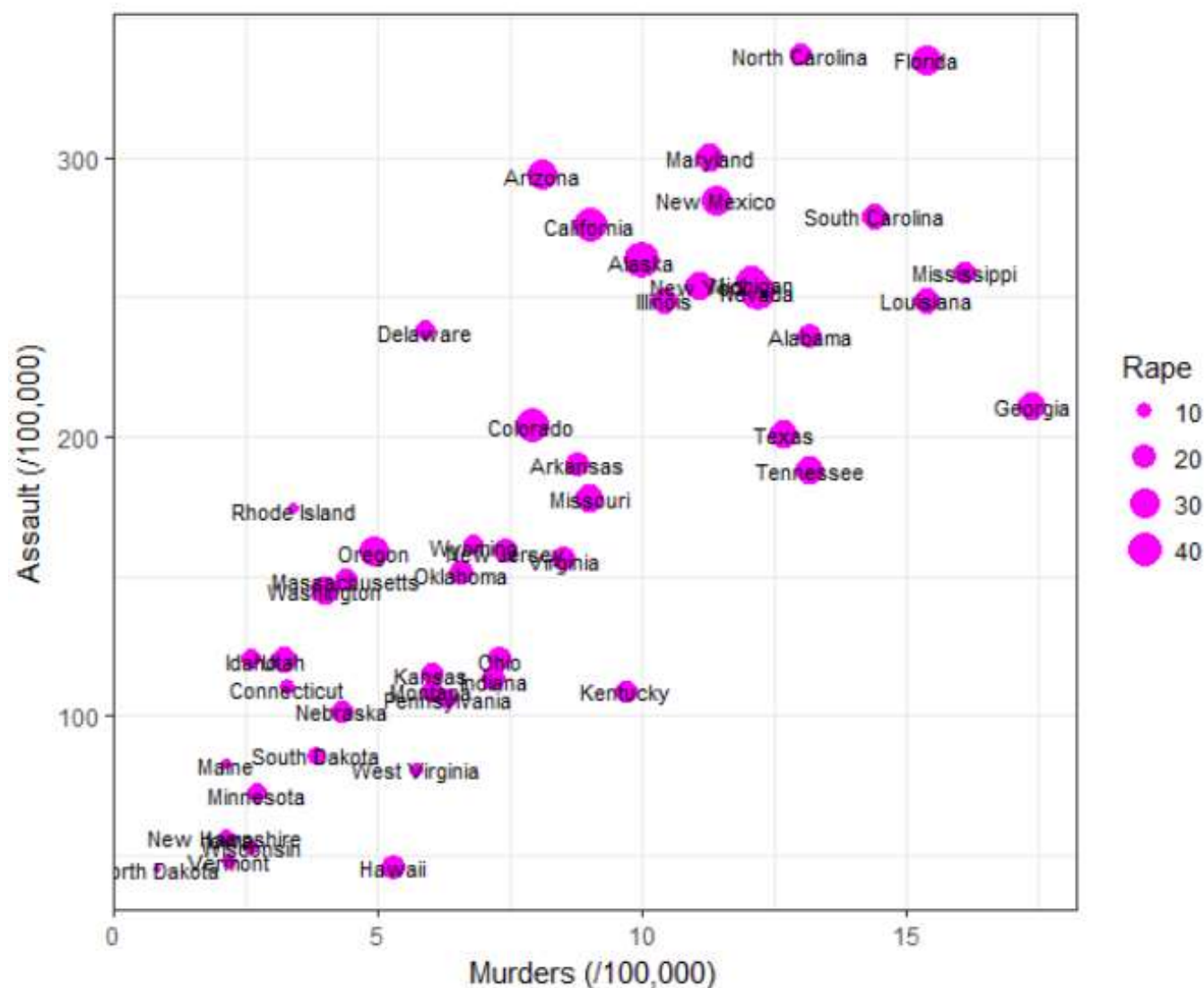
**palette**(value) {grDevices}

View or manipulate the color palette which is used when a col= has a numeric index.

```
#(1)
radius <- Rape/max(Rape) #circle size
N <- nrow(USArrests)
op <- palette(rainbow(N))
symbols(Murder, Assault, circles=radius,
        inches=0.25, fg='black', bg=1:N,
        xlab='Murder (/100,000)',
        ylab='Assault (/100,000)',
        main='Circle shows Rape (7.3~46)')
text(Murder, Assault, row.names(USArrests),
      cex=0.8)
palette(op)
```



```
#(2)
library(ggplot2)
ggplot(USArrests, aes(Murder, Assault, size=Rape, label=row.names(USArrests))) +
  geom_point(colour="magenta") + geom_text(size=3) + theme_bw() +
  xlab("Murders (/100,000)") + ylab("Assault (/100,000)")
```

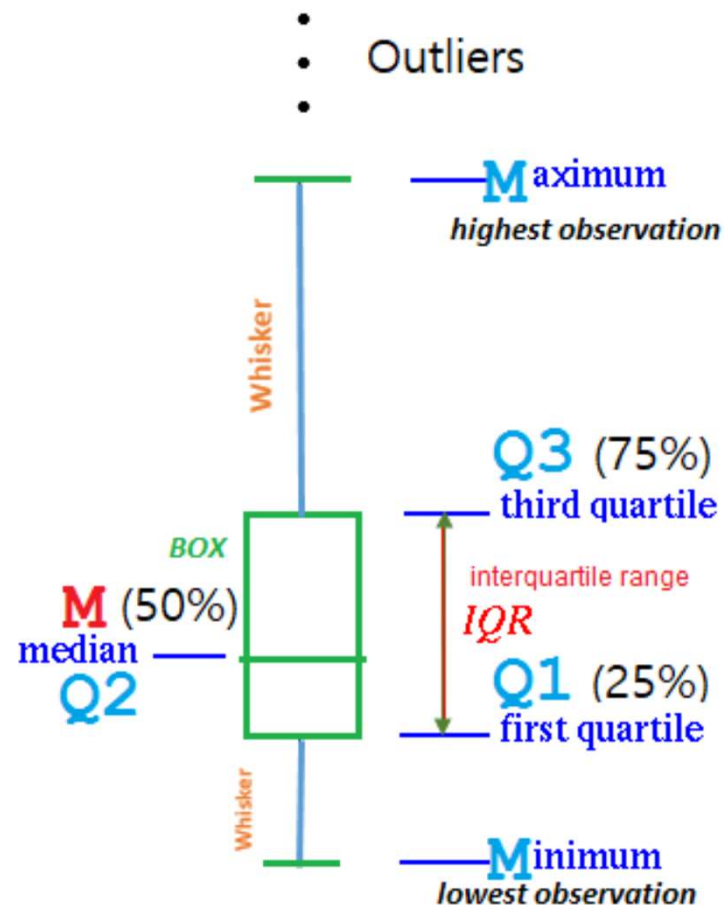


## 6. Box Plots

**boxplot**(x, data, ... ) {graphics}

Produce box-and-whisker plot(s) of the given (grouped) values.

A **box plot** is used to depict groups of numerical data through their quartiles.



**ToothGrowth** {datasets}

The Effect of Vitamin C on Tooth Growth in Guinea Pigs

len : Tooth length

supp : Supplement type (VC or OJ).

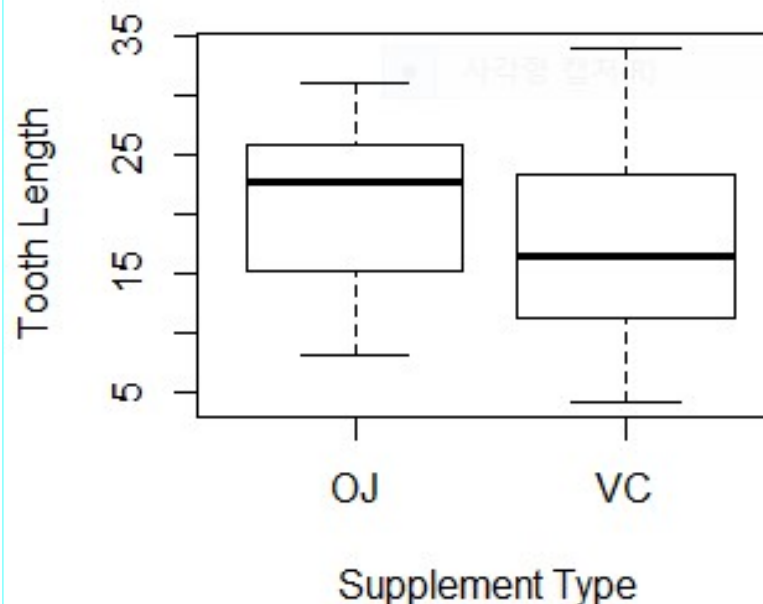
dose : Dose in milligrams/day

```
> head(ToothGrowth)
```

	len	supp	dose
1	4.2	VC	0.5
2	11.5	VC	0.5
3	7.3	VC	0.5
4	5.8	VC	0.5
5	6.4	VC	0.5
6	10.0	VC	0.5

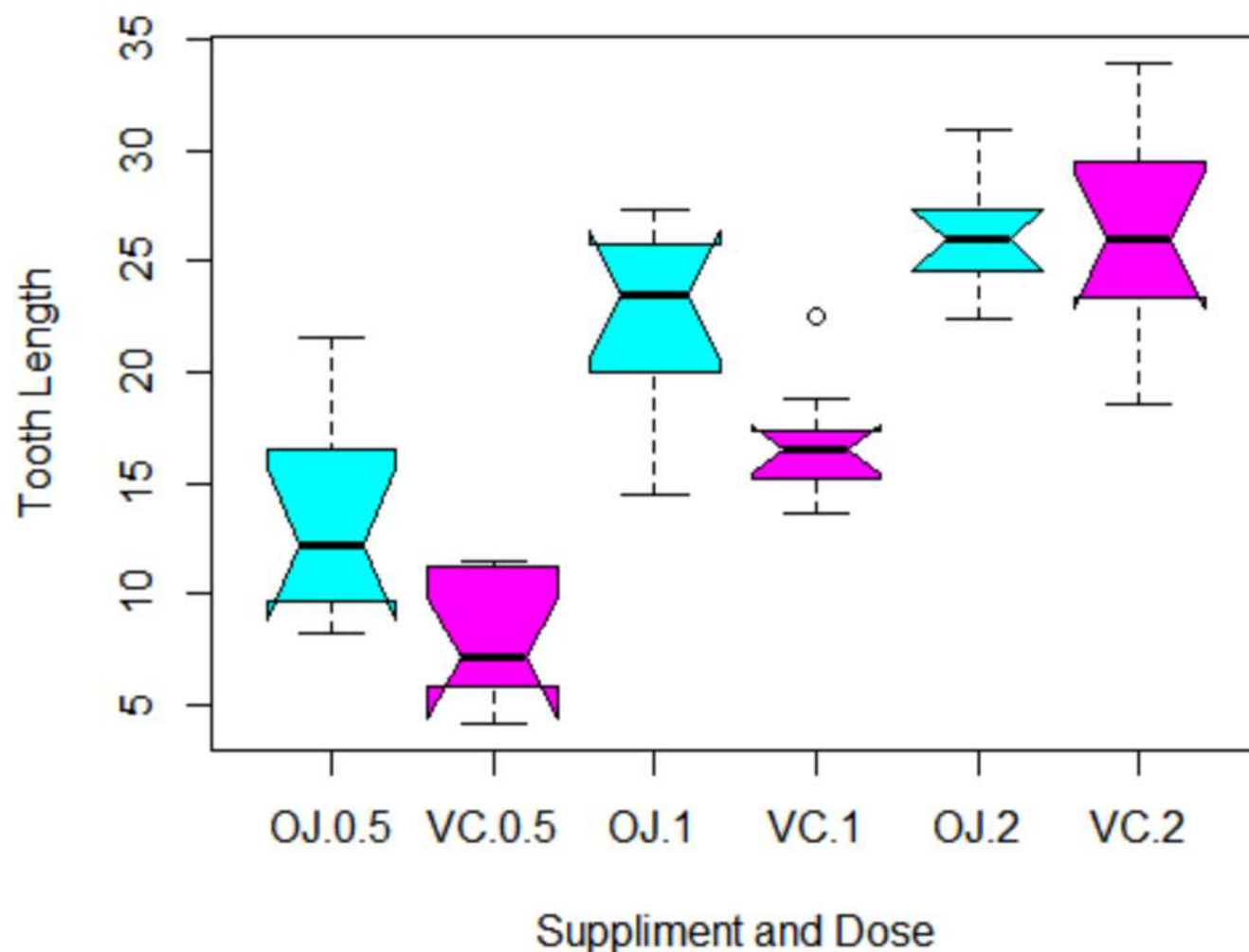
```
#(1) Simple boxplot
```

```
boxplot(len~supp, data=ToothGrowth,  
        xlab="Supplement Type", ylab="Tooth Length")
```



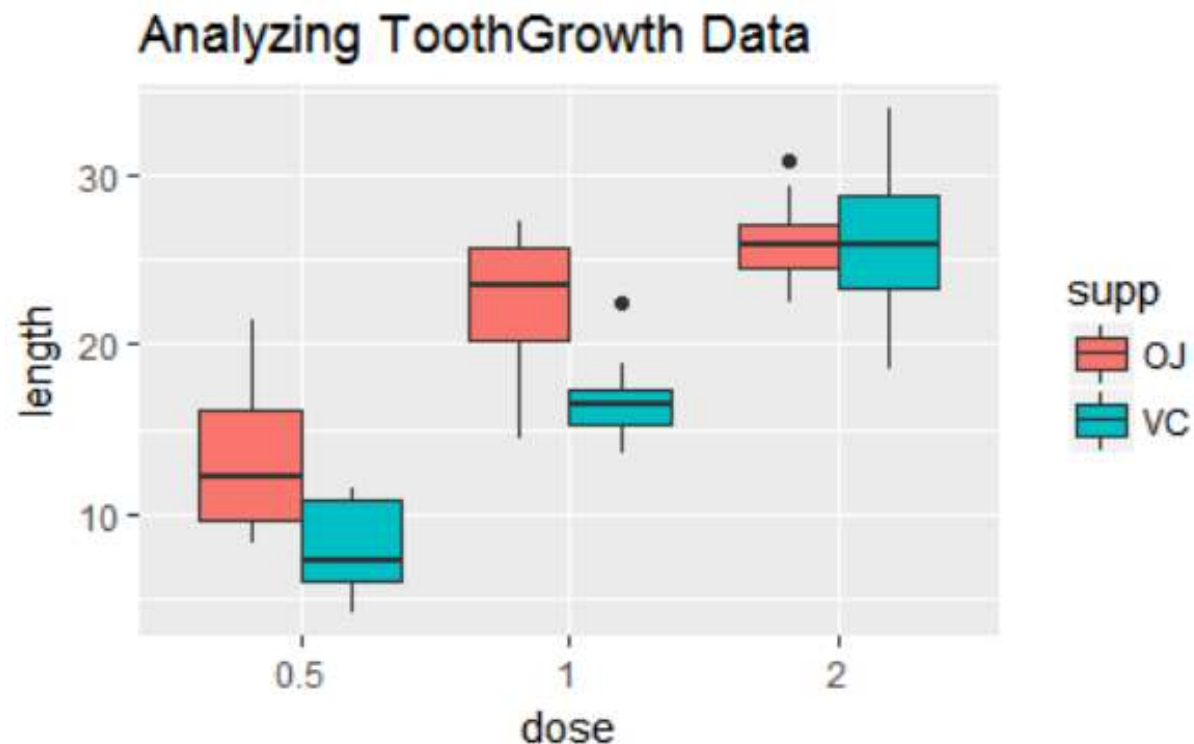


```
#(2) Boxplot of len against dose and supp factors  
boxplot(len~supp*dose, data=ToothGrowth, notch=T,  
        xlab="Suppliment and Dose", ylab="Tooth Length",  
        col=c("cyan", "magenta"))
```



**geom\_boxplot**(mapping, data, ... ) {ggplot2}  
A box and whiskers plot (in the style of Tukey)

```
#(3) boxplot in ggplot2  
library(ggplot2)  
ggplot(ToothGrowth, aes(x=factor(dose), y=len)) +  
  geom_boxplot(aes(fill=supp)) +  
  xlab("dose") + ylab("length") +  
  ggtitle("Analyzing ToothGrowth Data")
```

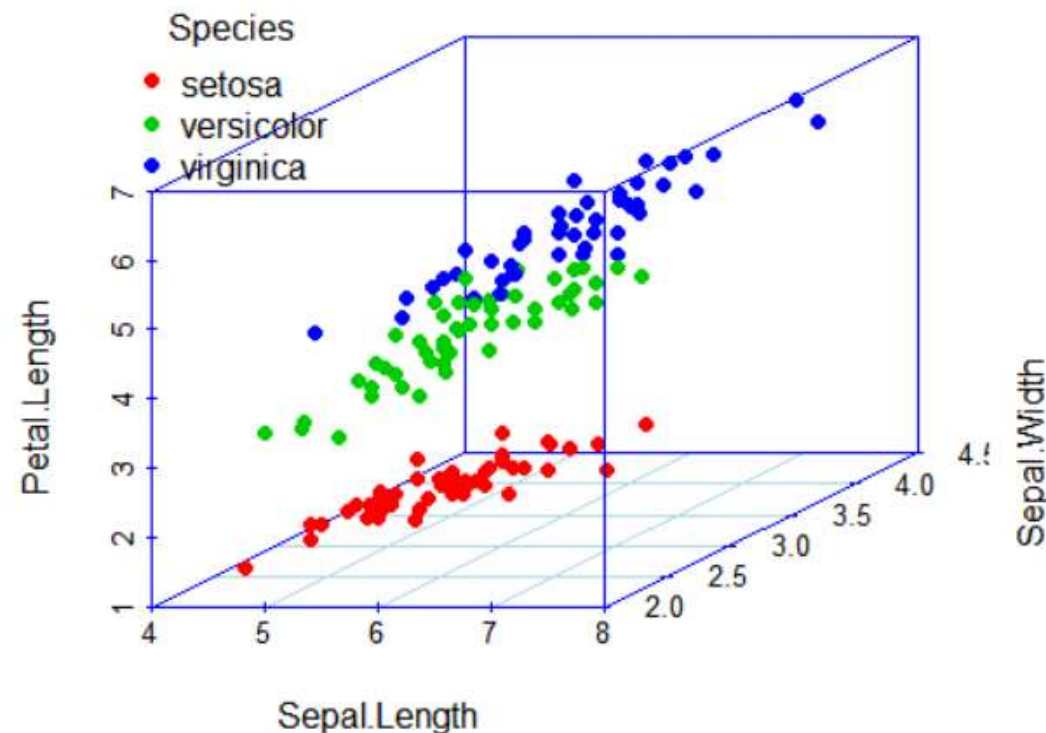


## 7. 3D Plots

**scatterplot3d**(x, y, z, ... ) {scatterplot3d}

Plots a three dimensional (3D) point cloud.

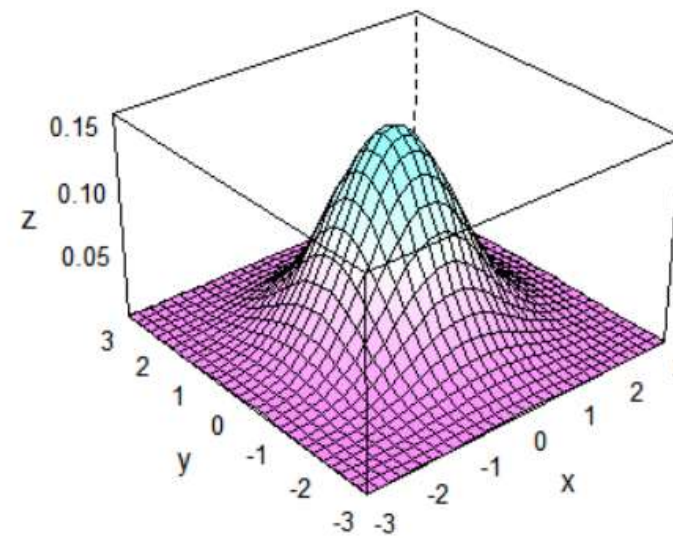
```
#(1) 3D Scatter Plot  
library(scatterplot3d)  
s3d <- scatterplot3d(iris[,1:3], color=c(2:4)[iris$Species],  
  col.axis="blue", col.grid="lightblue", pch=16, cex.symbols=1)  
legend(s3d$xyz.convert(2.5,3.1,8.8), legend=levels(iris$Species),  
  col=2:4, pch=16, bty="n", title="Species")
```



**wireframe**(x, data, ...) {lattice}

Generic functions to draw 3d scatter plots and surfaces.

```
#(2) 3D Wireframe Plot
library(lattice)
x <- seq(-3, 3, .2); y=x
dat <- expand.grid(x,y)
dat$z <- dnorm(dat[,1])*dnorm(dat[,2])
names(dat) <- c('x','y','z')
wireframe(z ~ x*y, data=dat, scales=list(arrows=FALSE),
          aspect=c(1,.6), drape=TRUE,
          par.settings=list(axis.line=list(col='transparent')))
```



0.16  
0.14  
0.12  
0.10  
0.08  
0.06  
0.04  
0.02  
0.00



## 8. Saving Plots

Since R runs on so many different operating systems, and supports so many different graphics formats, it's not surprising that there are a variety of ways of saving your plots.

**bmp, jpeg, png**(filename, width, height, pointsize, ... )

Graphics devices for BMP, JPEG and PNG format bitmap files.

**pdf**(file, width, height, pointsize, ... )

pdf starts the graphics device driver for producing PDF graphics.

- `jpeg("test.jpg"); plot(x,y); dev.off()`

# After the 'jpeg("test.jpg")' command all graphs are redirected to the file "test.jpg" in JPEG format.

The actual image data are not written to the file until the 'dev.off()' command is executed!

```
# Example: Saving Plots  
png('SampleSavePlot.png',width=580,height=640)  
x = rnorm(100, mean=0, sd=1)  
hist(x, freq=F, col='cyan')  
lines(density(x), col='red')  
dev.off()
```

SampleSavePlot.png →

