

3oct

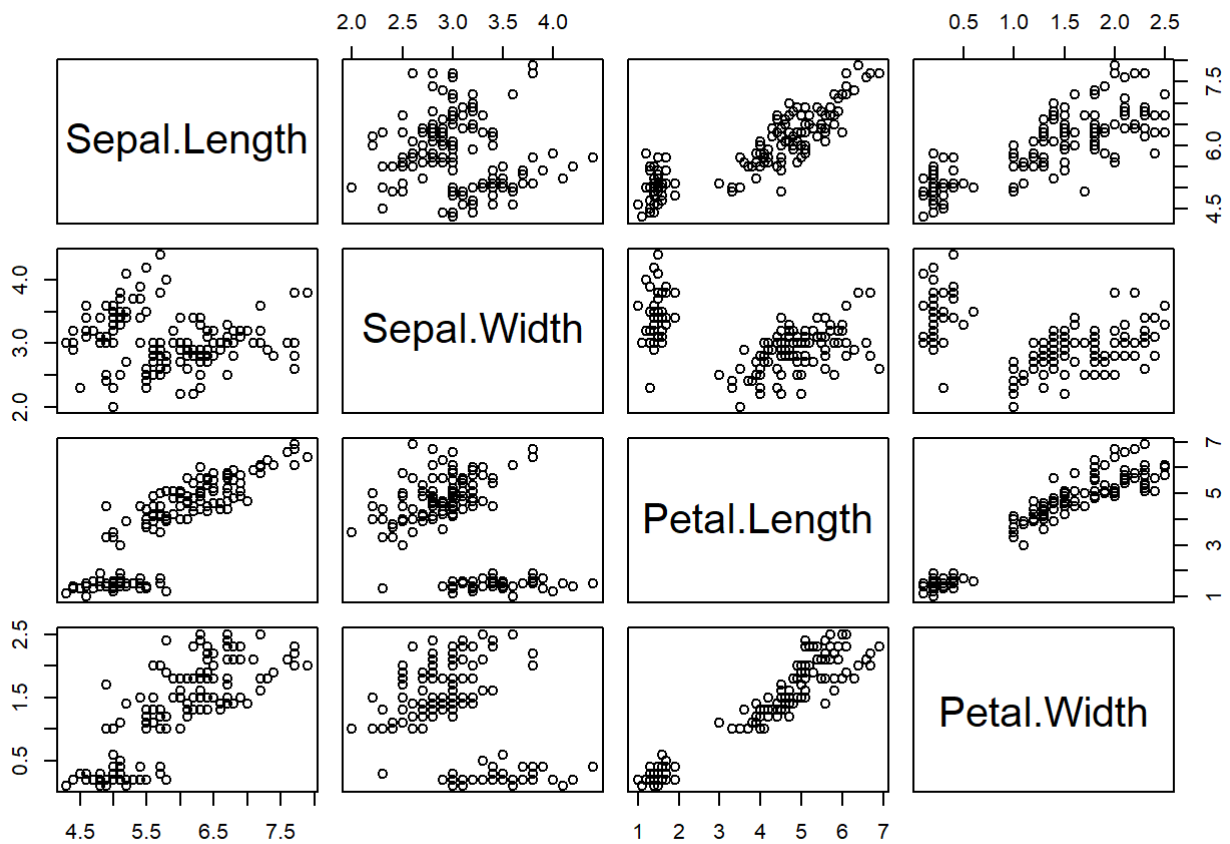
Devanshi

2023-10-03

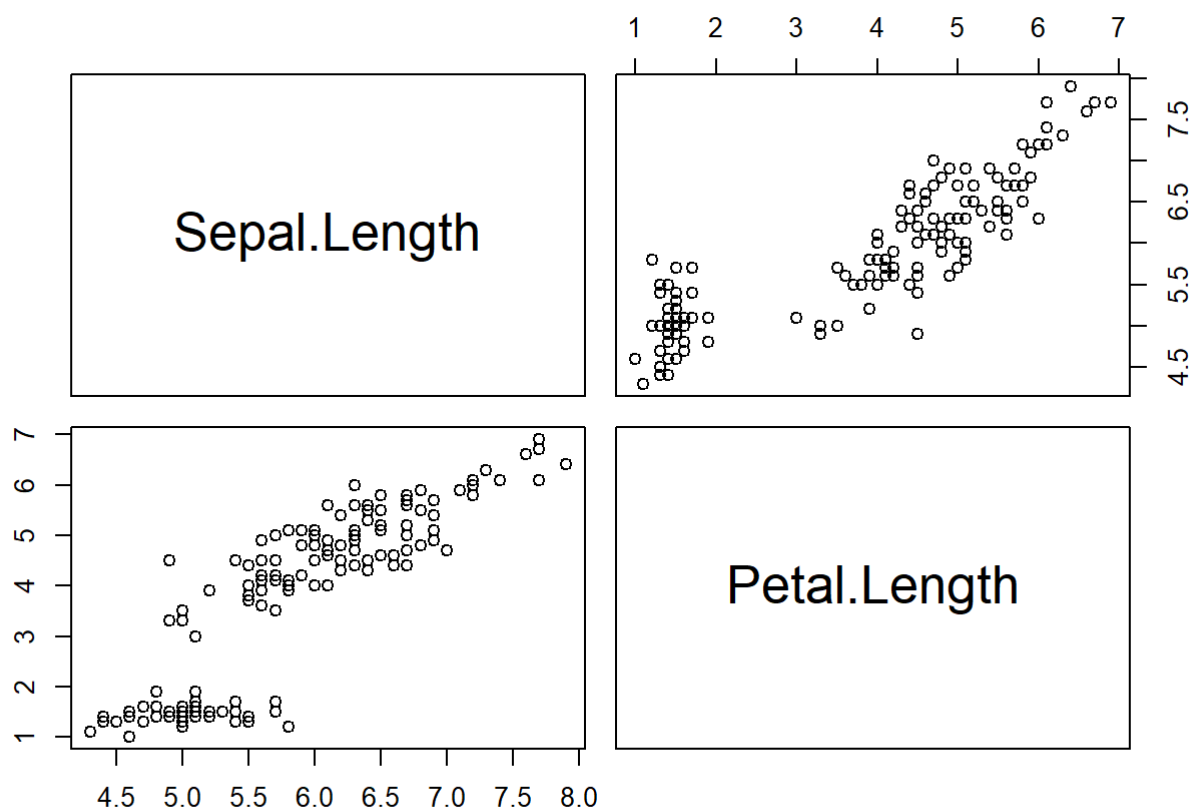
Pairs plotting (multiple correlation plot) Syntax: `pairs(~x+y+..., data=ourdata, additional arguments)`

Till now we were working with `plot()` command which gives us scatter plot for two variables. If we want to get the scatter plot for two variables. If we want to get the scatter plot for more than 2 variable and for each pair, we may use the `pairs()` command.

```
View(iris)
pairs(iris[-5]) #columns should be of numeric data type
```



```
attach(iris)
pairs(~Sepal.Length+Petal.Length, iris)
```



```
detach(iris)
```

Pairs plot provides us the matrix of scatter plots. It plots all the pairwise combinations for mentioned columns of data. Additional arguments: -main, sub, col - font.labels -labels -cex.labels -pch, cex

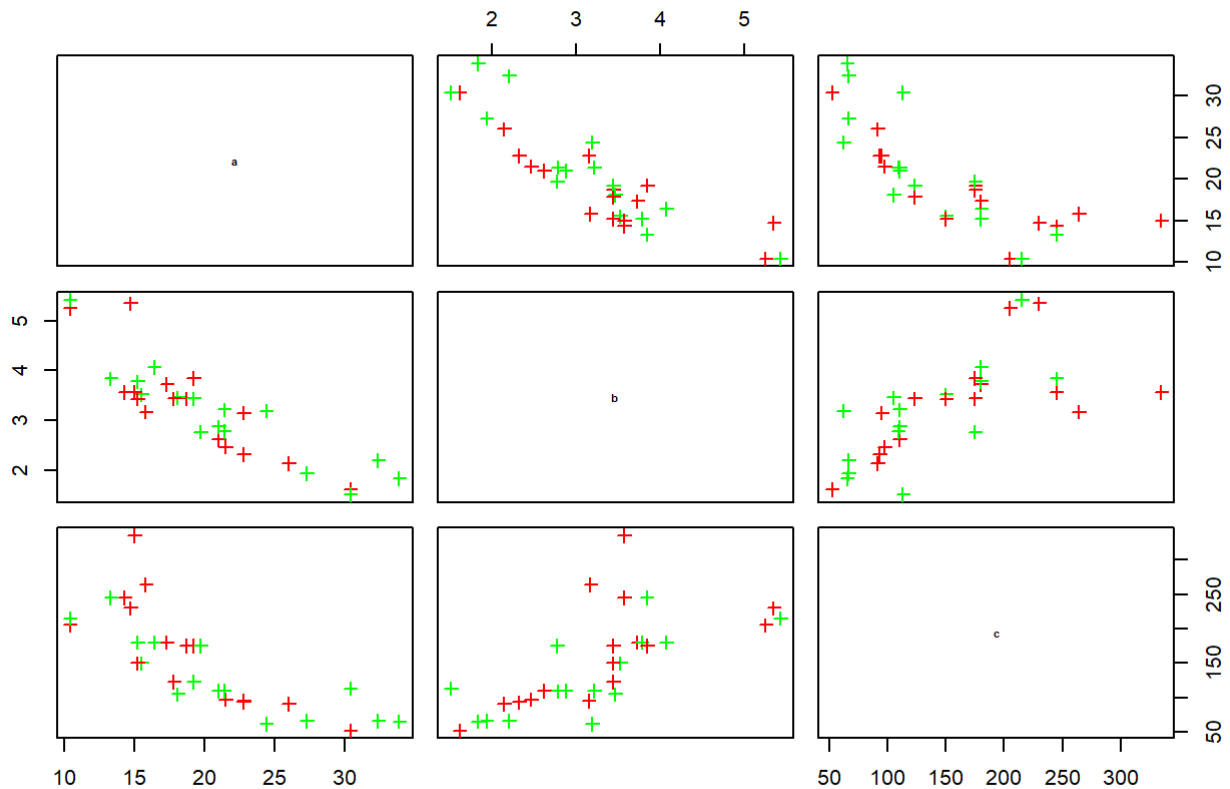
```
mtcars
```

##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

#Plot the scatter plot for each pair of mpg, wt, hp

```
attach(mtcars)
pairs(~mpg+wt+hp, mtcars,col=c("red", "green"),
      main="Pairs Plotting",
      labels=c("a","b","c"),
      font.labels=2,
      cex.labels=0.5,
      pch=3)
```

Pairs Plotting

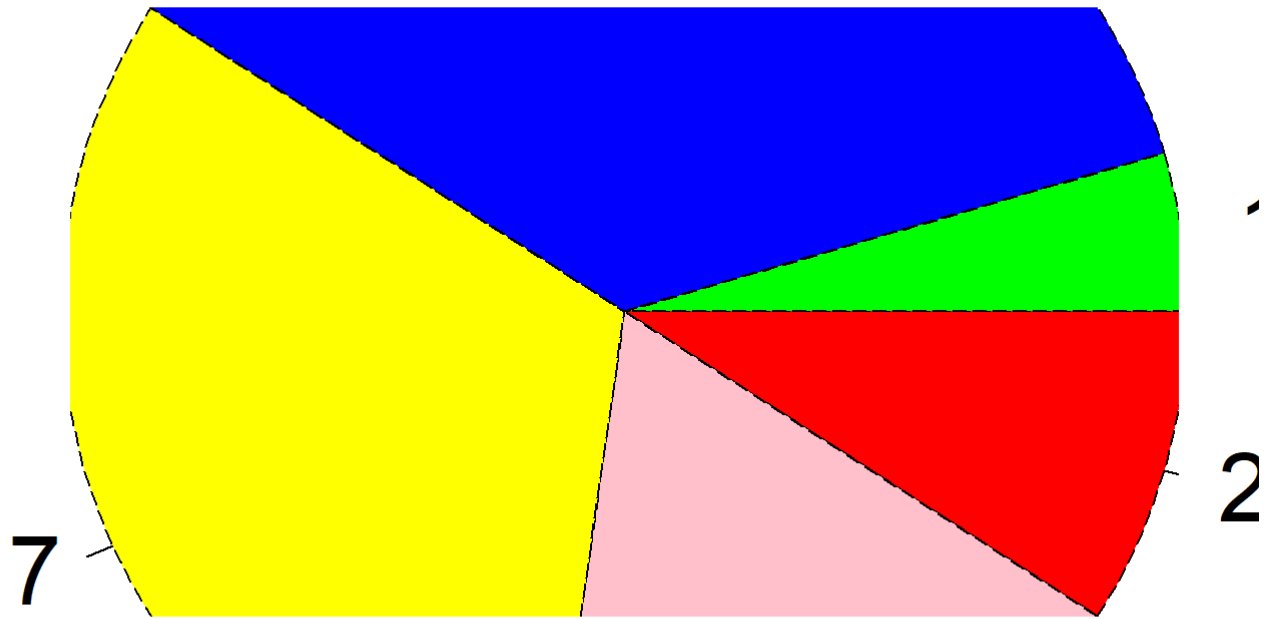


```
detach(mtcars)
```

#Pie Chart It displays proportional data using pie() command Syntax: pied(data, ...) Additional Arguments: - labels-name - col -main -clockwise= F(default)/ T Note that the initial angle is 0 deg for anticlockwise and 90 deg for clockwise -init.angle= -radius=

```
x=c(2,4,7,8,1)
pie(x, labels=x, col=c("red", "pink", "yellow", "blue", "green"),
    main= "Pie Chart",
    clockwise = T,
    init.angle = 0,
    cex=3,
    lty=5,
    radius=2)
```

Pie Chart



```
sum(x)
```

```
## [1] 22
```

```
x*100/sum(x)
```

```
## [1] 9.090909 18.181818 31.818182 36.363636 4.545455
```

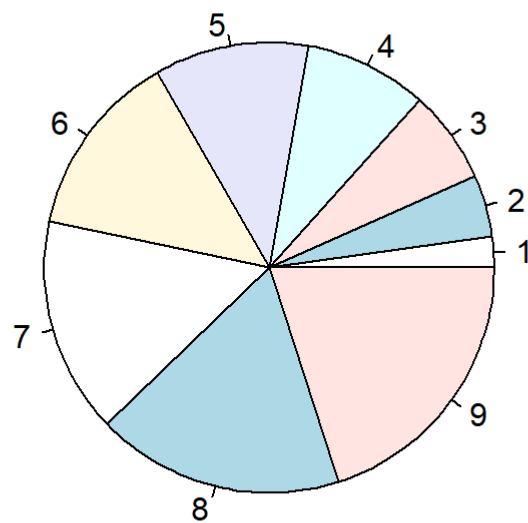
```
A= c(1,2,3,4,5,6,7,8,9)
```

```
B=matrix(A, 3)
```

```
B
```

```
##      [,1] [,2] [,3]  
## [1,]    1    4    7  
## [2,]    2    5    8  
## [3,]    3    6    9
```

```
pie(B)
```



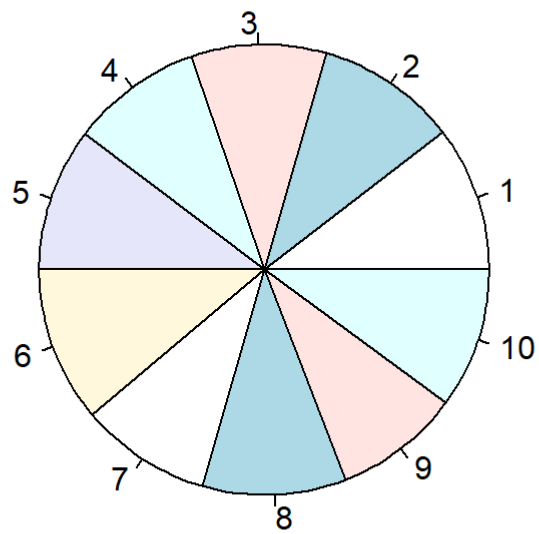
```
#data frame  
s=iris[1:10, 1:4]  
s
```

```
##      Sepal.Length Sepal.Width Petal.Length Petal.Width  
## 1          5.1         3.5         1.4         0.2  
## 2          4.9         3.0         1.4         0.2  
## 3          4.7         3.2         1.3         0.2  
## 4          4.6         3.1         1.5         0.2  
## 5          5.0         3.6         1.4         0.2  
## 6          5.4         3.9         1.7         0.4  
## 7          4.6         3.4         1.4         0.3  
## 8          5.0         3.4         1.5         0.2  
## 9          4.4         2.9         1.4         0.2  
## 10         4.9         3.1         1.5         0.1
```

```
class(s[1])
```

```
## [1] "data.frame"
```

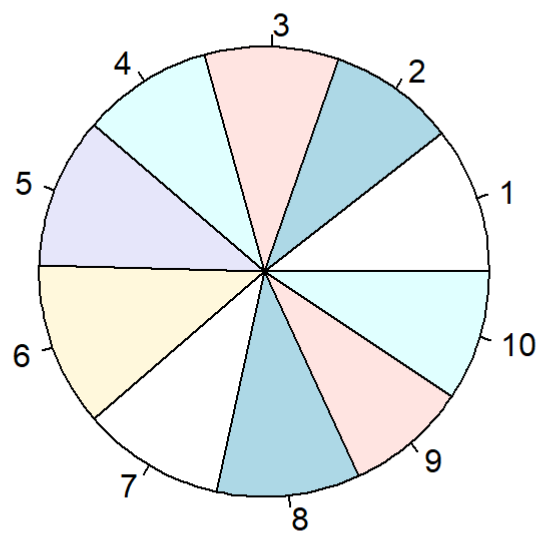
```
a=as.matrix(s[1]) #as.matrix converts dataframe to matrix  
pie(a)
```



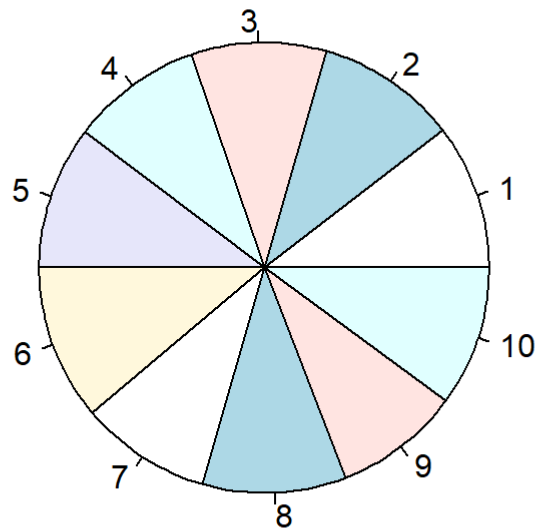
```
#plot a pie chart for any two columns of a data frame  
s=iris[1:10, 1:4]  
s
```

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width
## 1	5.1	3.5	1.4	0.2
## 2	4.9	3.0	1.4	0.2
## 3	4.7	3.2	1.3	0.2
## 4	4.6	3.1	1.5	0.2
## 5	5.0	3.6	1.4	0.2
## 6	5.4	3.9	1.7	0.4
## 7	4.6	3.4	1.4	0.3
## 8	5.0	3.4	1.5	0.2
## 9	4.4	2.9	1.4	0.2
## 10	4.9	3.1	1.5	0.1

```
b= as.matrix(s[1])  
c= as.matrix(s[2])  
pie(c)
```



`pie(b)`



#Box Plot, Box-Whisker Plot syntax: `boxplot(data, ...)` this type of plot is useful when you want to compare multiple type of samples. it shows the 5 number summary in a compact manner

Additional arguments for `boxplot` `range=0` is used to extend the whisker to minimum and maximum values

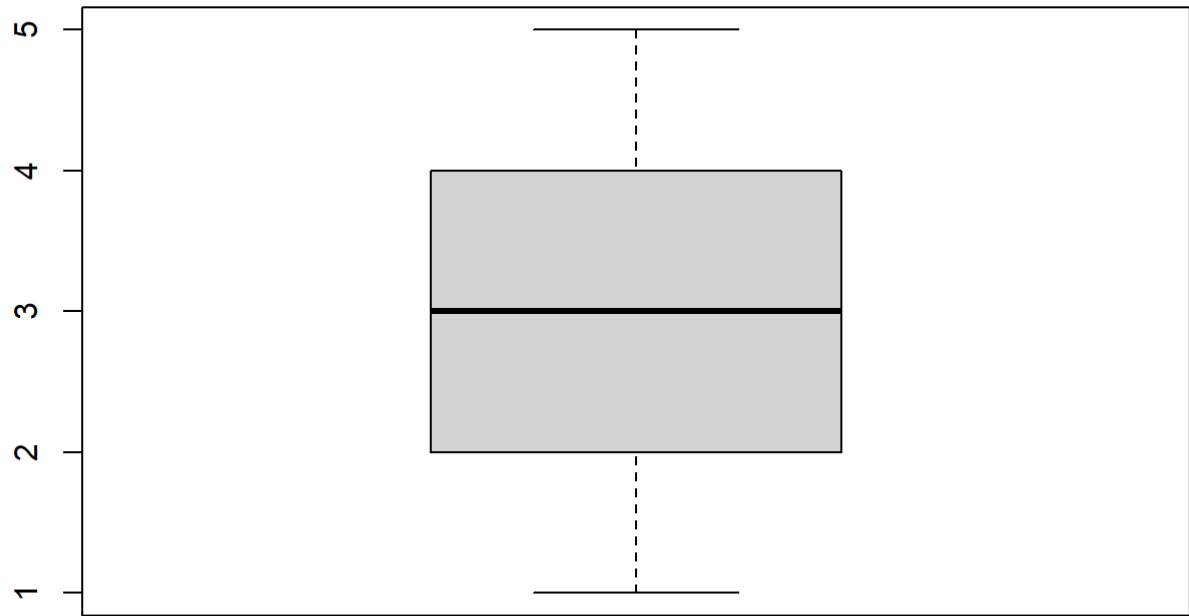
```
#vector
x=c(1,2,3,4,5)
summary(x) #6 number summary
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	1	2	3	3	4	5

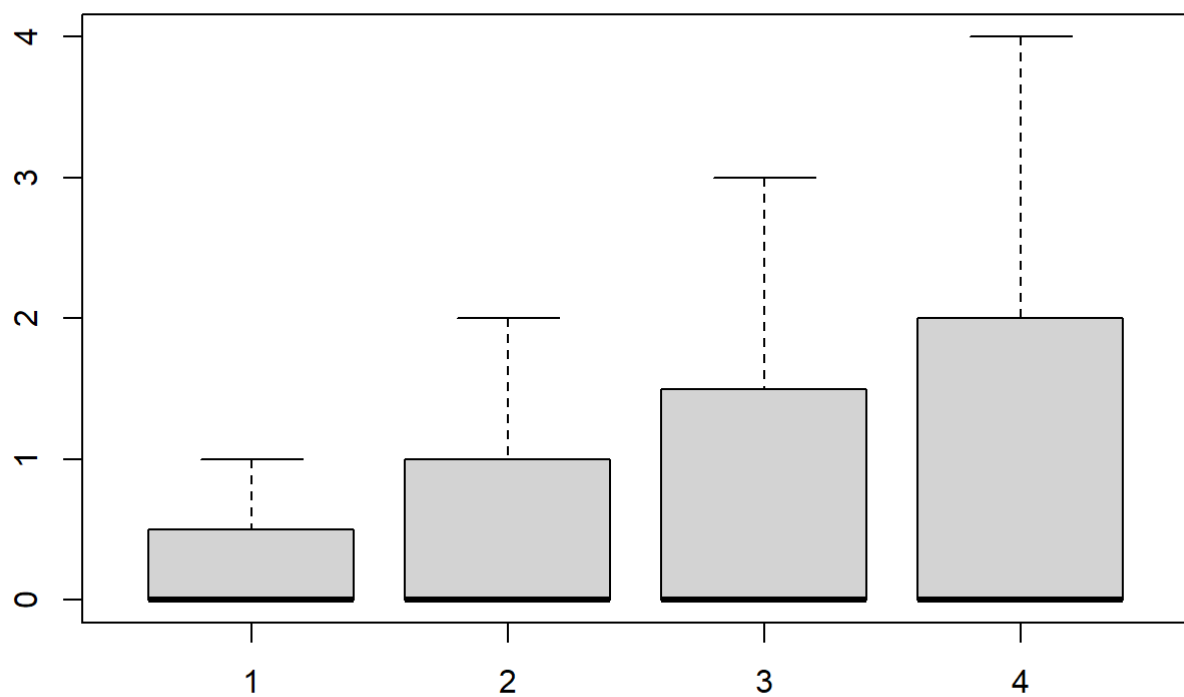
```
fivenum(x) #5 number summary
```

```
## [1] 1 2 3 4 5
```

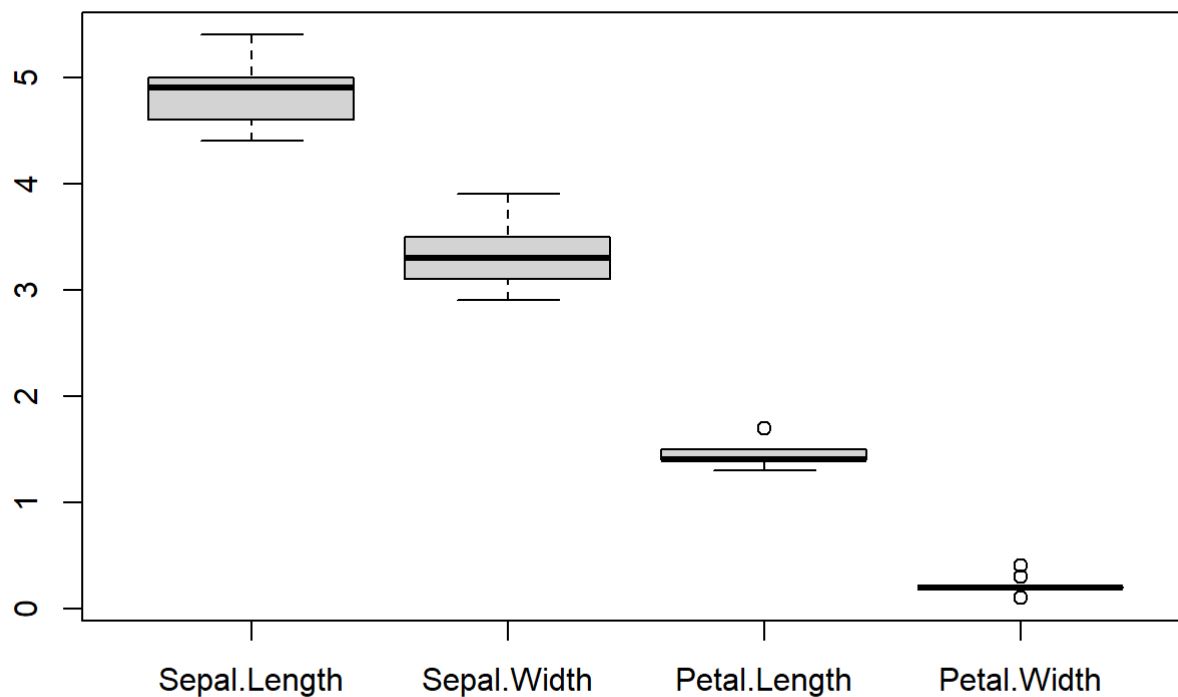
```
boxplot(x)
```



```
#matrix  
d=diag(1:4)  
boxplot(d)
```



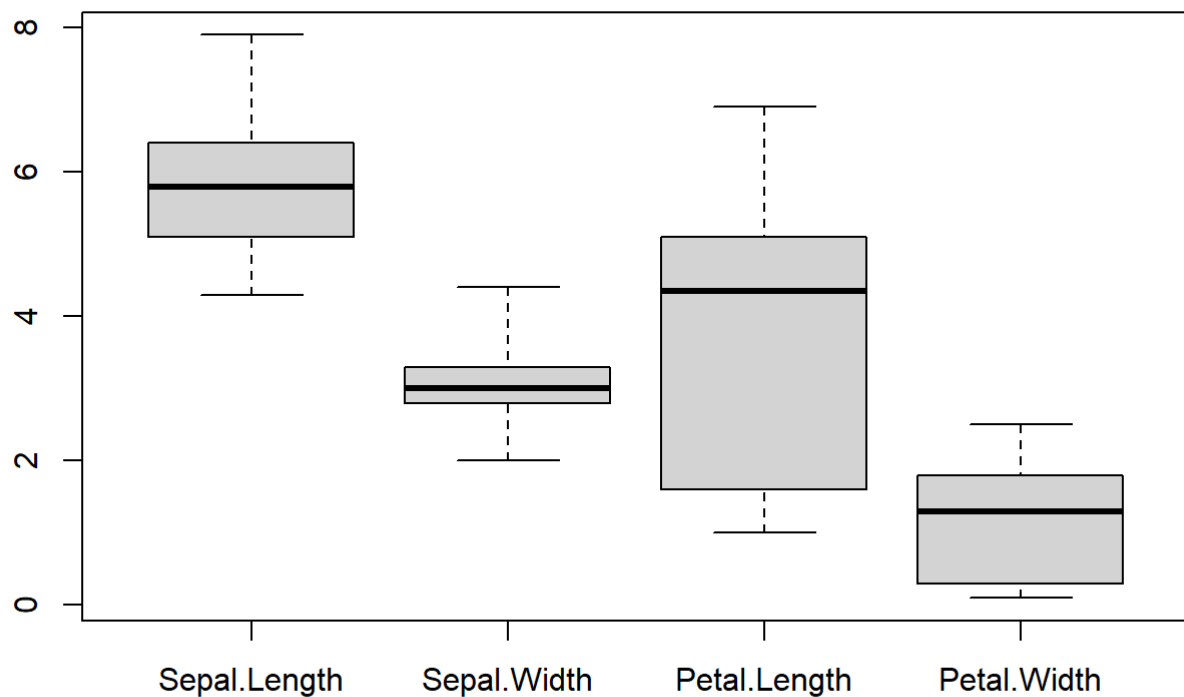
```
#dataframe  
s=iris[1:10, 1:4]  
boxplot(s)
```



```
#vertical boxplot
i1=iris[-5]
head(i1)
```

```
##   Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1         5.1         3.5         1.4         0.2
## 2         4.9         3.0         1.4         0.2
## 3         4.7         3.2         1.3         0.2
## 4         4.6         3.1         1.5         0.2
## 5         5.0         3.6         1.4         0.2
## 6         5.4         3.9         1.7         0.4
```

```
boxplot(i1,range=0)
```

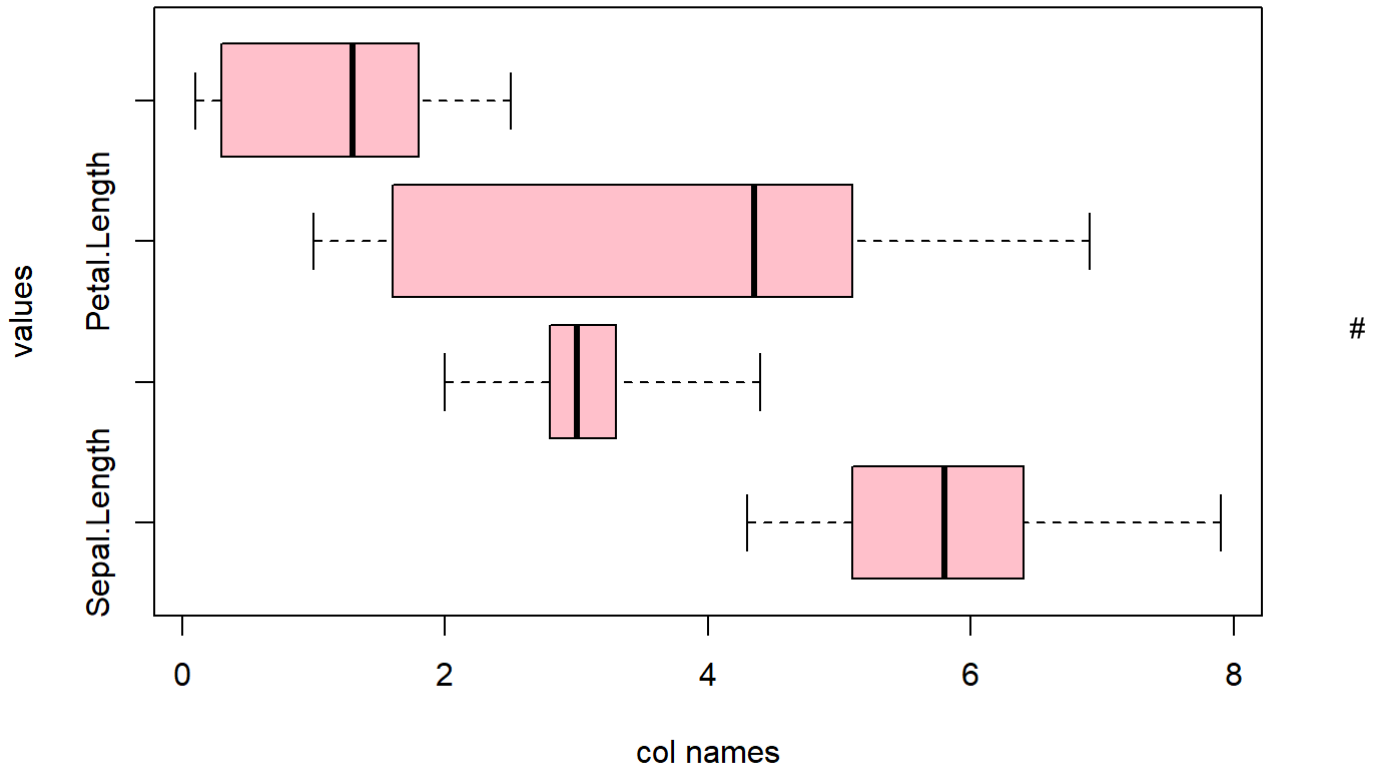


```
#horizontal boxplot
i1=iris[-5]
head(i1)
```

```
##   Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1         5.1         3.5         1.4         0.2
## 2         4.9         3.0         1.4         0.2
## 3         4.7         3.2         1.3         0.2
## 4         4.6         3.1         1.5         0.2
## 5         5.0         3.6         1.4         0.2
## 6         5.4         3.9         1.7         0.4
```

```
boxplot(i1,range=0,horizontal =T, main="boxplot",col=c("pink"))
title(xlab="col names",ylab="values")
```

boxplot



make a boxplot for the first column of iris data according to species

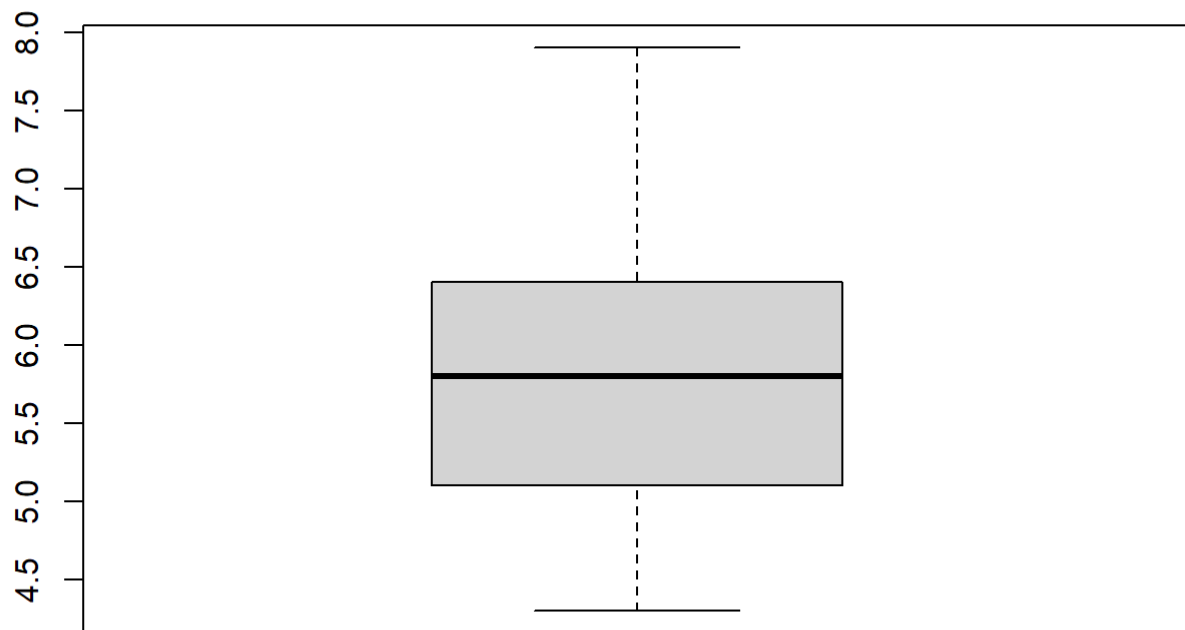
iris

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	5.1	3.5	1.4	0.2	setosa
## 2	4.9	3.0	1.4	0.2	setosa
## 3	4.7	3.2	1.3	0.2	setosa
## 4	4.6	3.1	1.5	0.2	setosa
## 5	5.0	3.6	1.4	0.2	setosa
## 6	5.4	3.9	1.7	0.4	setosa
## 7	4.6	3.4	1.4	0.3	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa
## 10	4.9	3.1	1.5	0.1	setosa
## 11	5.4	3.7	1.5	0.2	setosa
## 12	4.8	3.4	1.6	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 15	5.8	4.0	1.2	0.2	setosa
## 16	5.7	4.4	1.5	0.4	setosa
## 17	5.4	3.9	1.3	0.4	setosa
## 18	5.1	3.5	1.4	0.3	setosa
## 19	5.7	3.8	1.7	0.3	setosa
## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa
## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor

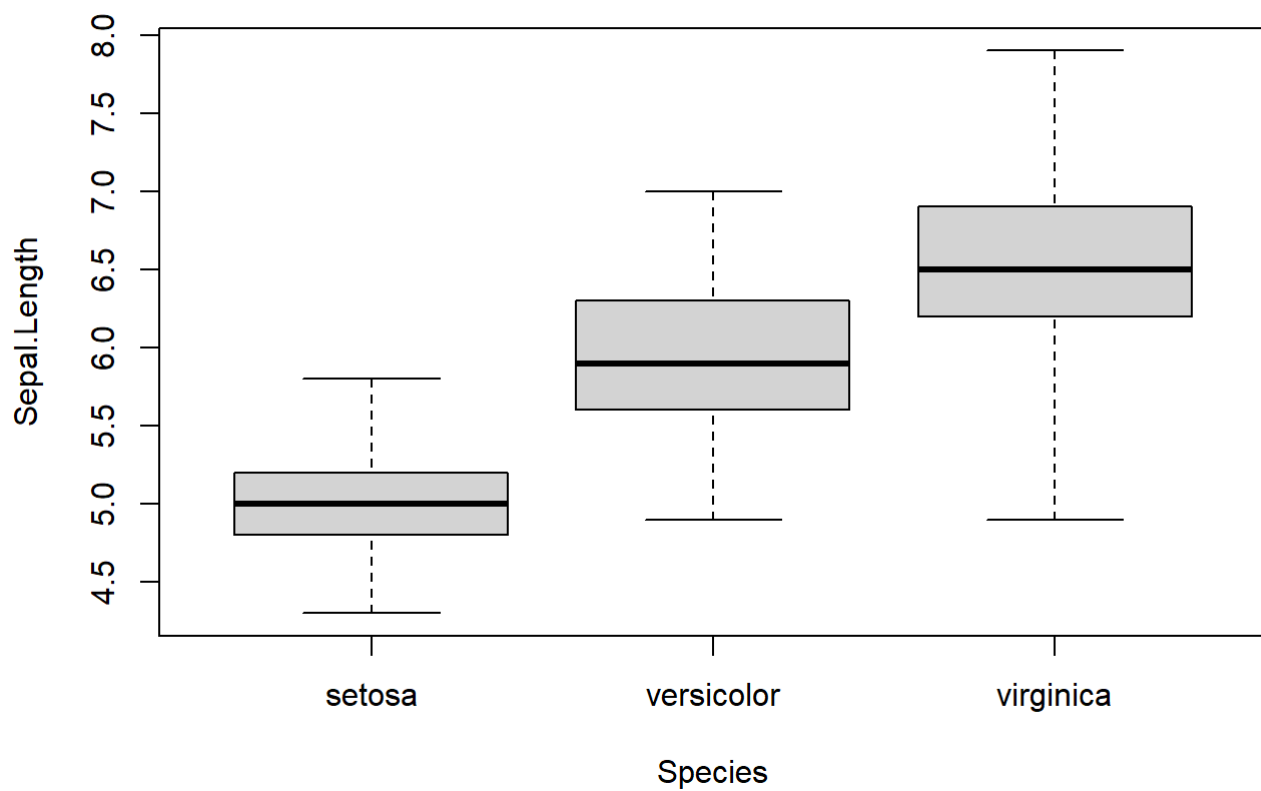
## 55	6.5	2.8	4.6	1.5 versicolor
## 56	5.7	2.8	4.5	1.3 versicolor
## 57	6.3	3.3	4.7	1.6 versicolor
## 58	4.9	2.4	3.3	1.0 versicolor
## 59	6.6	2.9	4.6	1.3 versicolor
## 60	5.2	2.7	3.9	1.4 versicolor
## 61	5.0	2.0	3.5	1.0 versicolor
## 62	5.9	3.0	4.2	1.5 versicolor
## 63	6.0	2.2	4.0	1.0 versicolor
## 64	6.1	2.9	4.7	1.4 versicolor
## 65	5.6	2.9	3.6	1.3 versicolor
## 66	6.7	3.1	4.4	1.4 versicolor
## 67	5.6	3.0	4.5	1.5 versicolor
## 68	5.8	2.7	4.1	1.0 versicolor
## 69	6.2	2.2	4.5	1.5 versicolor
## 70	5.6	2.5	3.9	1.1 versicolor
## 71	5.9	3.2	4.8	1.8 versicolor
## 72	6.1	2.8	4.0	1.3 versicolor
## 73	6.3	2.5	4.9	1.5 versicolor
## 74	6.1	2.8	4.7	1.2 versicolor
## 75	6.4	2.9	4.3	1.3 versicolor
## 76	6.6	3.0	4.4	1.4 versicolor
## 77	6.8	2.8	4.8	1.4 versicolor
## 78	6.7	3.0	5.0	1.7 versicolor
## 79	6.0	2.9	4.5	1.5 versicolor
## 80	5.7	2.6	3.5	1.0 versicolor
## 81	5.5	2.4	3.8	1.1 versicolor
## 82	5.5	2.4	3.7	1.0 versicolor
## 83	5.8	2.7	3.9	1.2 versicolor
## 84	6.0	2.7	5.1	1.6 versicolor
## 85	5.4	3.0	4.5	1.5 versicolor
## 86	6.0	3.4	4.5	1.6 versicolor
## 87	6.7	3.1	4.7	1.5 versicolor
## 88	6.3	2.3	4.4	1.3 versicolor
## 89	5.6	3.0	4.1	1.3 versicolor
## 90	5.5	2.5	4.0	1.3 versicolor
## 91	5.5	2.6	4.4	1.2 versicolor
## 92	6.1	3.0	4.6	1.4 versicolor
## 93	5.8	2.6	4.0	1.2 versicolor
## 94	5.0	2.3	3.3	1.0 versicolor
## 95	5.6	2.7	4.2	1.3 versicolor
## 96	5.7	3.0	4.2	1.2 versicolor
## 97	5.7	2.9	4.2	1.3 versicolor
## 98	6.2	2.9	4.3	1.3 versicolor
## 99	5.1	2.5	3.0	1.1 versicolor
## 100	5.7	2.8	4.1	1.3 versicolor
## 101	6.3	3.3	6.0	2.5 virginica
## 102	5.8	2.7	5.1	1.9 virginica
## 103	7.1	3.0	5.9	2.1 virginica
## 104	6.3	2.9	5.6	1.8 virginica
## 105	6.5	3.0	5.8	2.2 virginica
## 106	7.6	3.0	6.6	2.1 virginica
## 107	4.9	2.5	4.5	1.7 virginica
## 108	7.3	2.9	6.3	1.8 virginica
## 109	6.7	2.5	5.8	1.8 virginica
## 110	7.2	3.6	6.1	2.5 virginica

## 111	6.5	3.2	5.1	2.0	virginica
## 112	6.4	2.7	5.3	1.9	virginica
## 113	6.8	3.0	5.5	2.1	virginica
## 114	5.7	2.5	5.0	2.0	virginica
## 115	5.8	2.8	5.1	2.4	virginica
## 116	6.4	3.2	5.3	2.3	virginica
## 117	6.5	3.0	5.5	1.8	virginica
## 118	7.7	3.8	6.7	2.2	virginica
## 119	7.7	2.6	6.9	2.3	virginica
## 120	6.0	2.2	5.0	1.5	virginica
## 121	6.9	3.2	5.7	2.3	virginica
## 122	5.6	2.8	4.9	2.0	virginica
## 123	7.7	2.8	6.7	2.0	virginica
## 124	6.3	2.7	4.9	1.8	virginica
## 125	6.7	3.3	5.7	2.1	virginica
## 126	7.2	3.2	6.0	1.8	virginica
## 127	6.2	2.8	4.8	1.8	virginica
## 128	6.1	3.0	4.9	1.8	virginica
## 129	6.4	2.8	5.6	2.1	virginica
## 130	7.2	3.0	5.8	1.6	virginica
## 131	7.4	2.8	6.1	1.9	virginica
## 132	7.9	3.8	6.4	2.0	virginica
## 133	6.4	2.8	5.6	2.2	virginica
## 134	6.3	2.8	5.1	1.5	virginica
## 135	6.1	2.6	5.6	1.4	virginica
## 136	7.7	3.0	6.1	2.3	virginica
## 137	6.3	3.4	5.6	2.4	virginica
## 138	6.4	3.1	5.5	1.8	virginica
## 139	6.0	3.0	4.8	1.8	virginica
## 140	6.9	3.1	5.4	2.1	virginica
## 141	6.7	3.1	5.6	2.4	virginica
## 142	6.9	3.1	5.1	2.3	virginica
## 143	5.8	2.7	5.1	1.9	virginica
## 144	6.8	3.2	5.9	2.3	virginica
## 145	6.7	3.3	5.7	2.5	virginica
## 146	6.7	3.0	5.2	2.3	virginica
## 147	6.3	2.5	5.0	1.9	virginica
## 148	6.5	3.0	5.2	2.0	virginica
## 149	6.2	3.4	5.4	2.3	virginica
## 150	5.9	3.0	5.1	1.8	virginica

```
attach(iris)
boxplot(Sepal.Length)
```



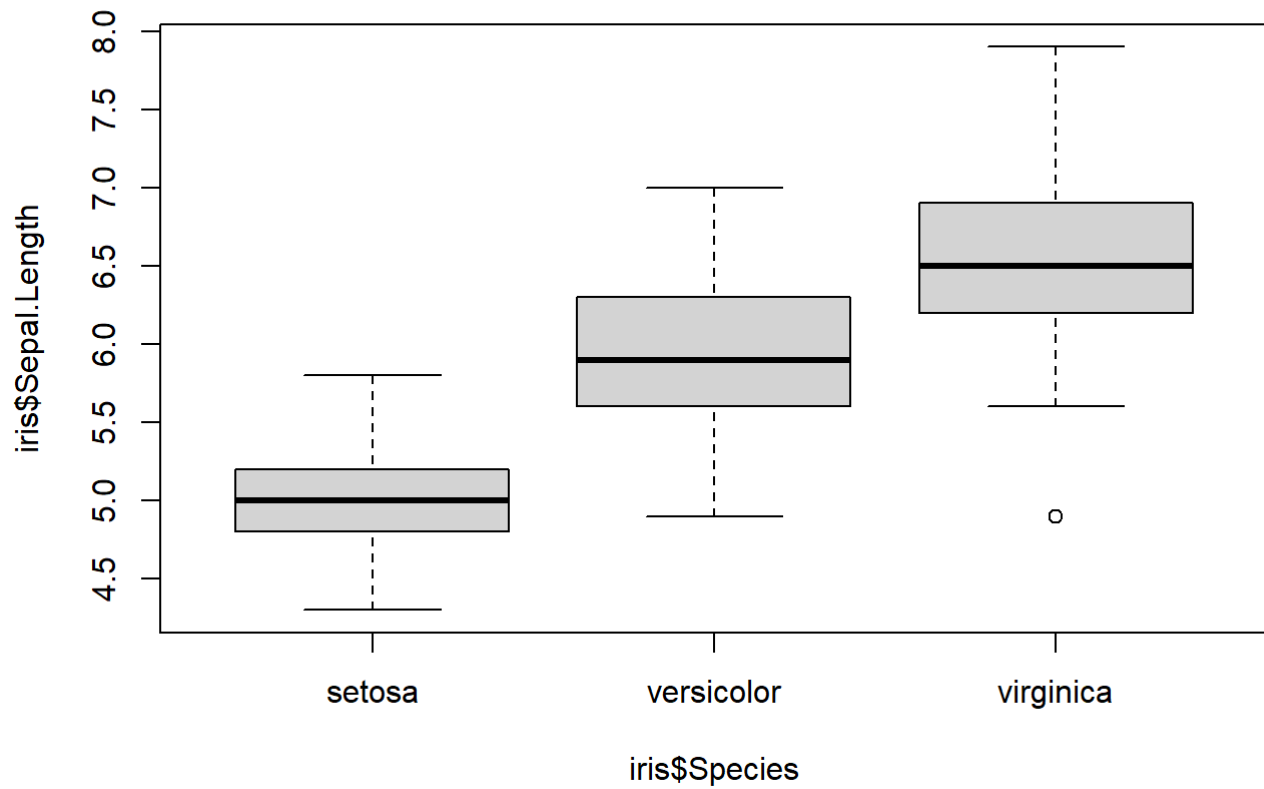
```
boxplot(Sepal.Length~Species, data=iris, range=0)
```



```
detach(iris)
```

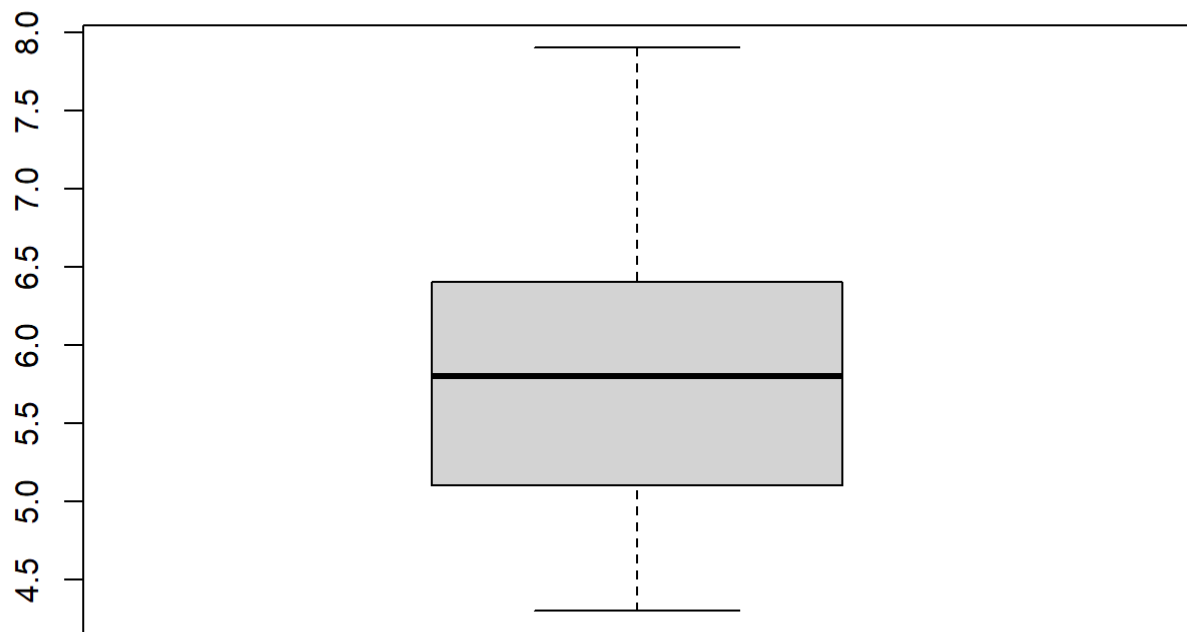
```
#second method
```

```
boxplot(iris$Sepal.Length~iris$Species)
```

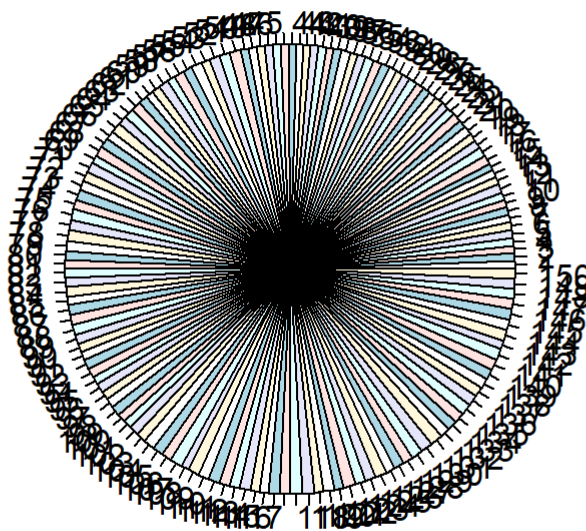


```
#third method
```

```
with(boxplot(Sepal.Length),data=iris)
```



```
with(pie(Sepal.Length),data=iris)
```



```
a=(c(9,25,15,2,14,25,24,47,2,3,5,9,14,24,29,34))
mat1=matrix(a,8,2);mat1
```

```
##      [,1] [,2]
## [1,]    9    2
## [2,]   25    3
## [3,]   15    5
## [4,]    2    9
## [5,]   14   14
## [6,]   25   24
## [7,]   24   29
## [8,]   47   34
```

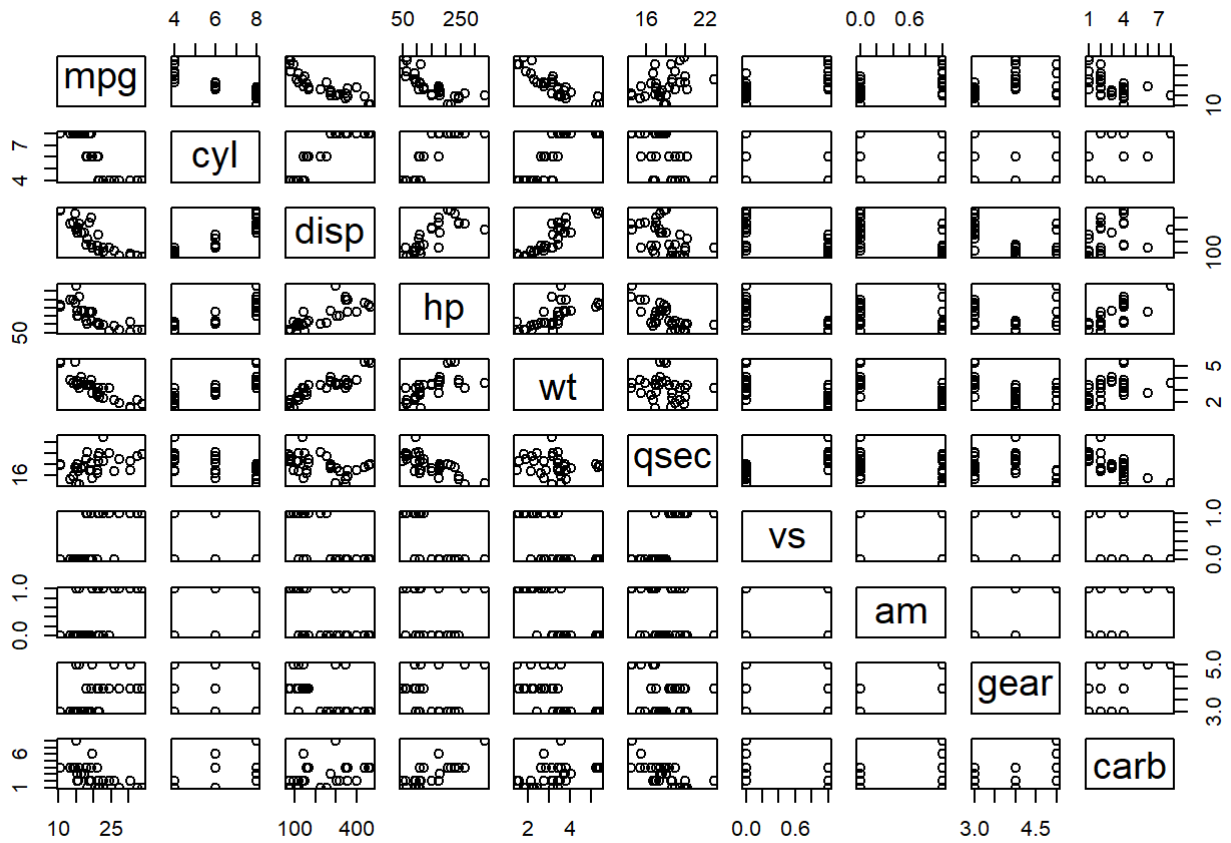
```
row.names(mat1)=(c("Taw","Torridge","Ouse","Exe","Lyn","Brook","Ditch","Fal"))
colnames(mat1)=(c("count","speed"))
mat1
```

```
##      count speed
## Taw         9    2
## Torridge    25    3
## Ouse        15    5
## Exe          2    9
## Lyn         14   14
## Brook       25   24
## Ditch       24   29
## Fal        47   34
```

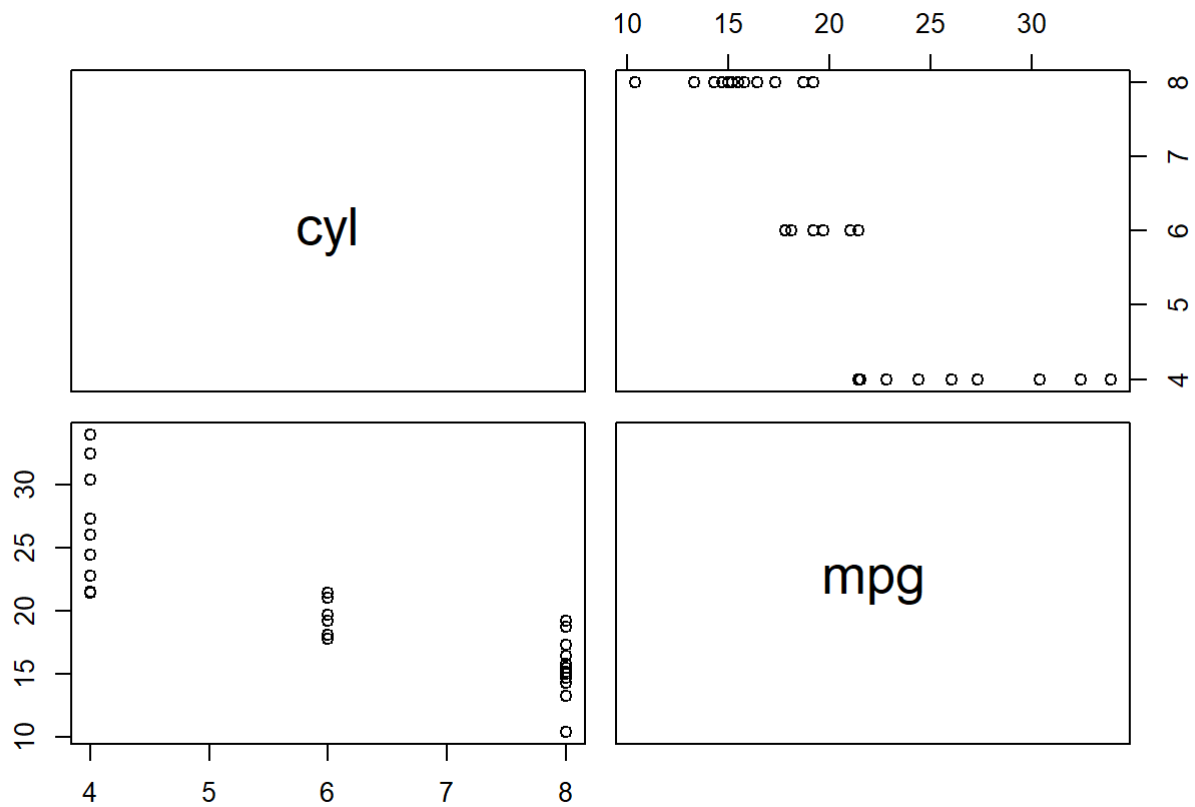
```
#pairsplot
mtcars
```

##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

```
pairs(mtcars[-5])
```



```
attach(mtcars)
pairs(~cyl+mpg,mtcars)
```

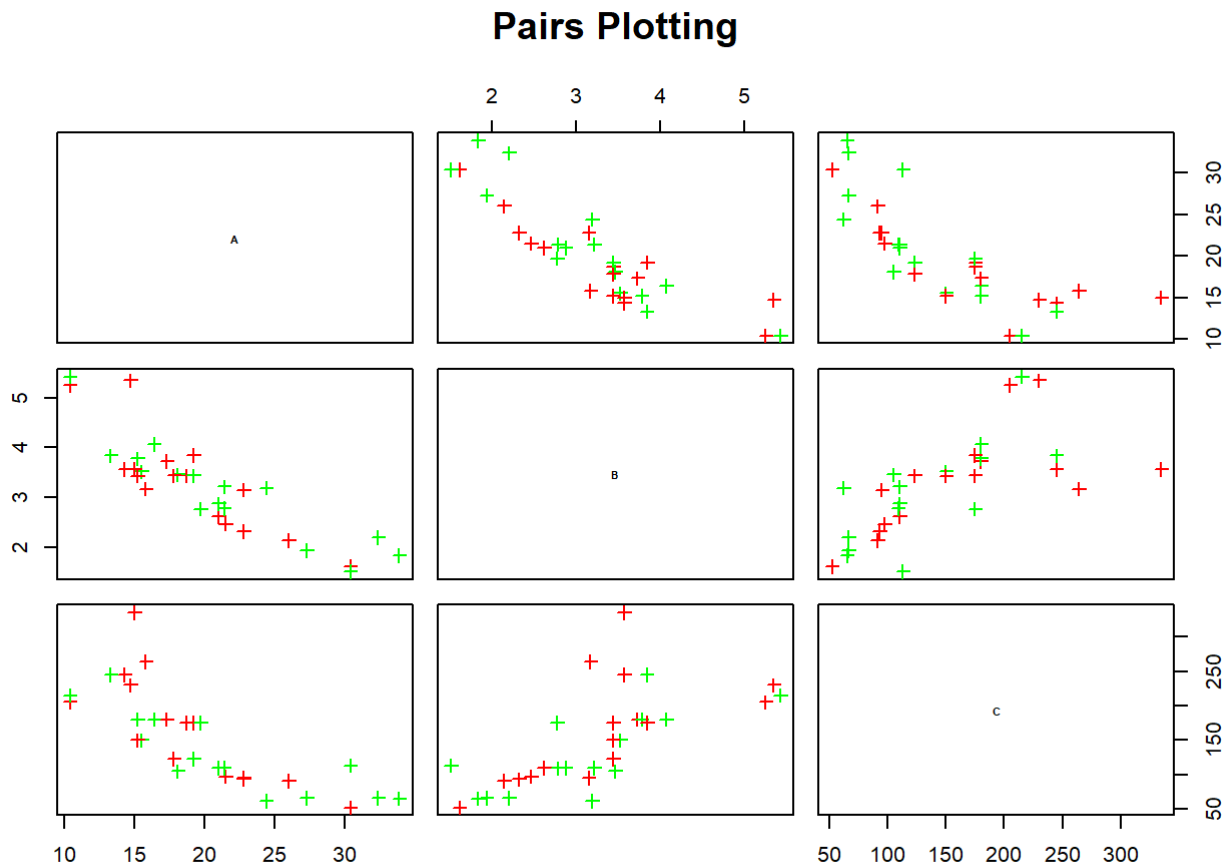


```
detach(mtcars)
```

```
#scatterplot
```

```
attach(mtcars)
```

```
pairs(~mpg+wt+hp,mtcars,col=c("red","green"),main="Pairs Plotting",labels=c("A","B","C"),font.labels=2,cex.labels=0.5,pch=3)
```



```
detach(mtcars)
```

```
#pie chart
```

```
s=mtcars[1:10,1:4];s
```

##	mpg	cyl	disp	hp
## Mazda RX4	21.0	6	160.0	110
## Mazda RX4 Wag	21.0	6	160.0	110
## Datsun 710	22.8	4	108.0	93
## Hornet 4 Drive	21.4	6	258.0	110
## Hornet Sportabout	18.7	8	360.0	175
## Valiant	18.1	6	225.0	105
## Duster 360	14.3	8	360.0	245
## Merc 240D	24.4	4	146.7	62
## Merc 230	22.8	4	140.8	95
## Merc 280	19.2	6	167.6	123

```
class(s[1])
```

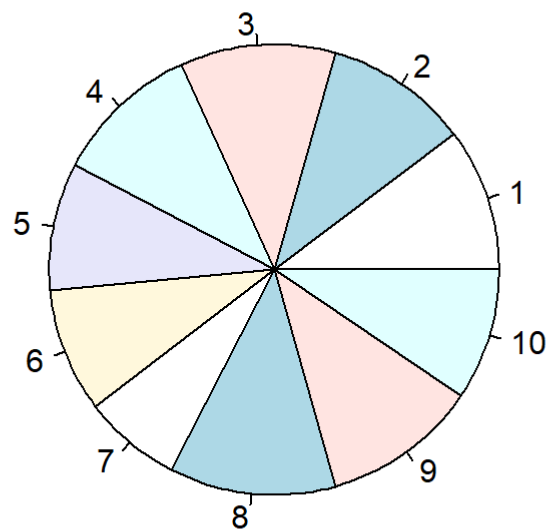


```
## [1] "data.frame"
```

```
d=as.matrix(s[1]);d
```

```
##           mpg
## Mazda RX4    21.0
## Mazda RX4 Wag 21.0
## Datsun 710    22.8
## Hornet 4 Drive 21.4
## Hornet Sportabout 18.7
## Valiant      18.1
## Duster 360   14.3
## Merc 240D    24.4
## Merc 230     22.8
## Merc 280     19.2
```

```
pie(d)
```



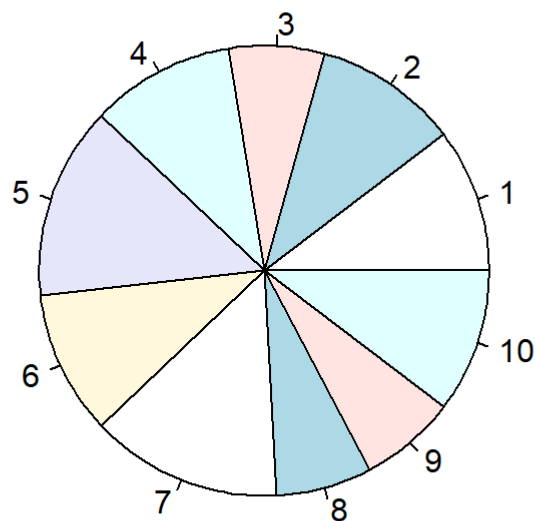
```
#pie chart for two columns
s=mtcars[1:10, 1:4]
s
```

##		mpg	cyl	disp	hp
##	Mazda RX4	21.0	6	160.0	110
##	Mazda RX4 Wag	21.0	6	160.0	110
##	Datsun 710	22.8	4	108.0	93
##	Hornet 4 Drive	21.4	6	258.0	110
##	Hornet Sportabout	18.7	8	360.0	175
##	Valiant	18.1	6	225.0	105
##	Duster 360	14.3	8	360.0	245
##	Merc 240D	24.4	4	146.7	62
##	Merc 230	22.8	4	140.8	95
##	Merc 280	19.2	6	167.6	123

```

b= as.matrix(s[1])
c= as.matrix(s[2])
pie(c)

```



```
pie(b)
```

