SO, HOW PO YOU GO FROM PATA TO PECISIONS?

1. USE NUMBERS AND PLOTS TO DESCRIBE THE DATA

2. DRAW INFERENCES FROM THE DATA

3. ORGANIZE PATA USING ANALYTICAL TOOLS

4. QUANTIFY RELATIONSHIPS BETWEEN VARIABLES

5. VISUALLY COMMUNICATE WHAT YOU'VE LEARNT

SO, HOW PO YOU GO FROM PATA TO PECISIONS?

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5. VISUALLY COMMUNICATE WHAT YOU'VE LEARNT

OFTEN, A GOOD DECK CAN BE THE DIFFERENCE BETWEEN

"WOW, THAT WAS AN AMAZING JOB"

AND

"EH.. WHAT WAS IT THAT YOU DID FOR THE LAST 3 WEEKS?"

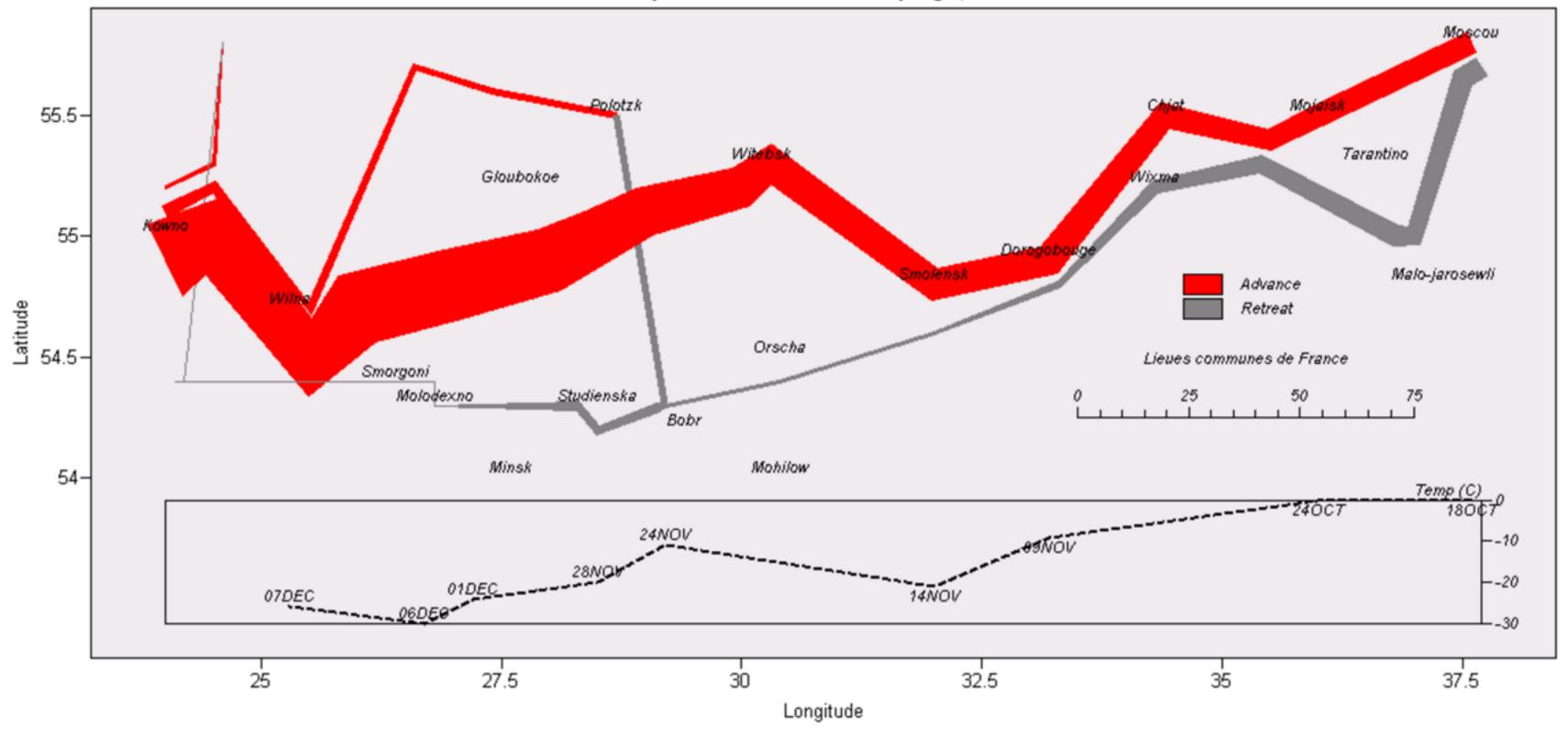
5. VISUALLY COMMUNICATE WHAT YOU'VE LEARNT

A GOOD DECK DOCUMENTS YOUR FINDINGS IN AN EXTREMELY EFFECTIVE WAY

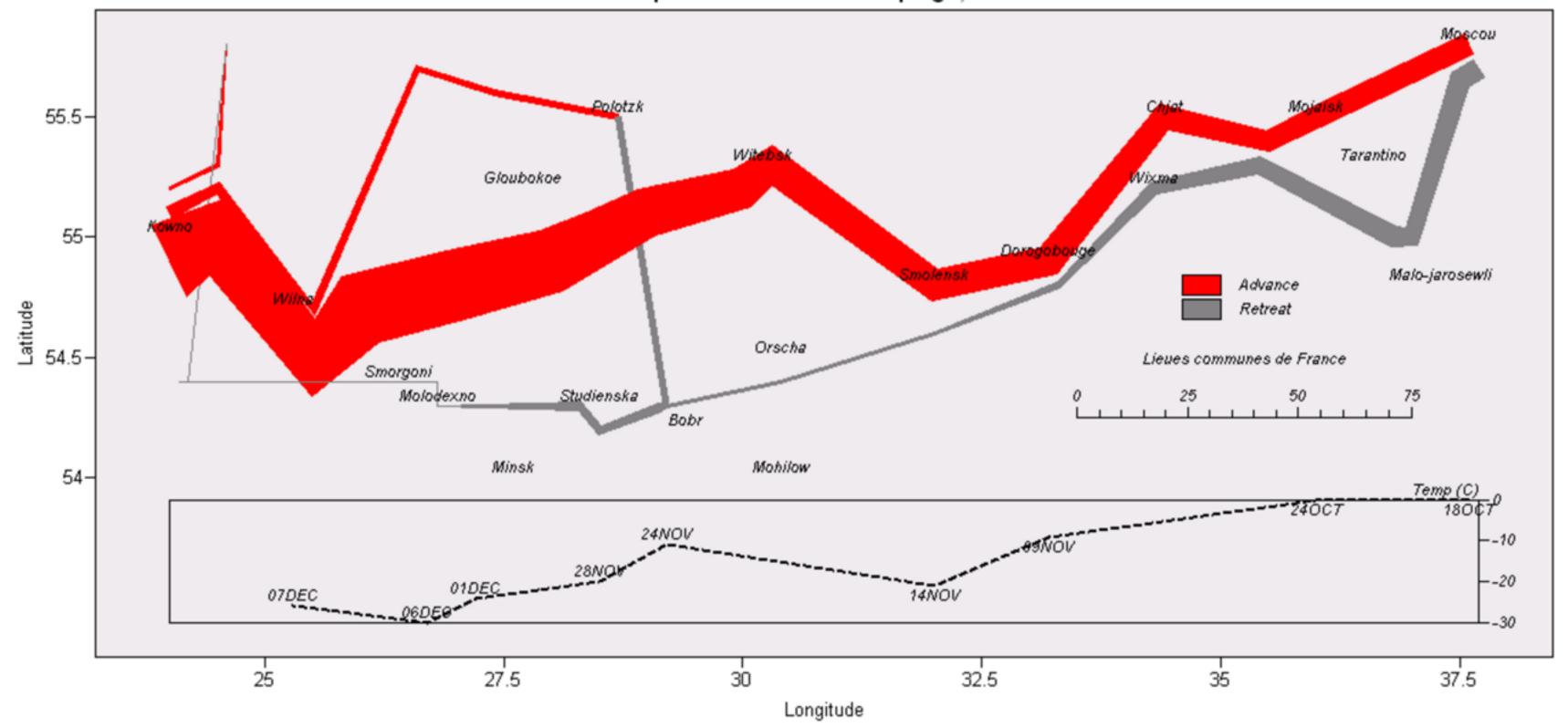
A WELL PLANNED GRAPH CAN BE MORE EFFECTIVE THAN A 100 TABLES AND SLIDES

5. VISUALLY COMMUNICATE WHAT YOU'VE LEARNT A WELL PLANNED GRAPH CAN BE USED TO TELL A STORY

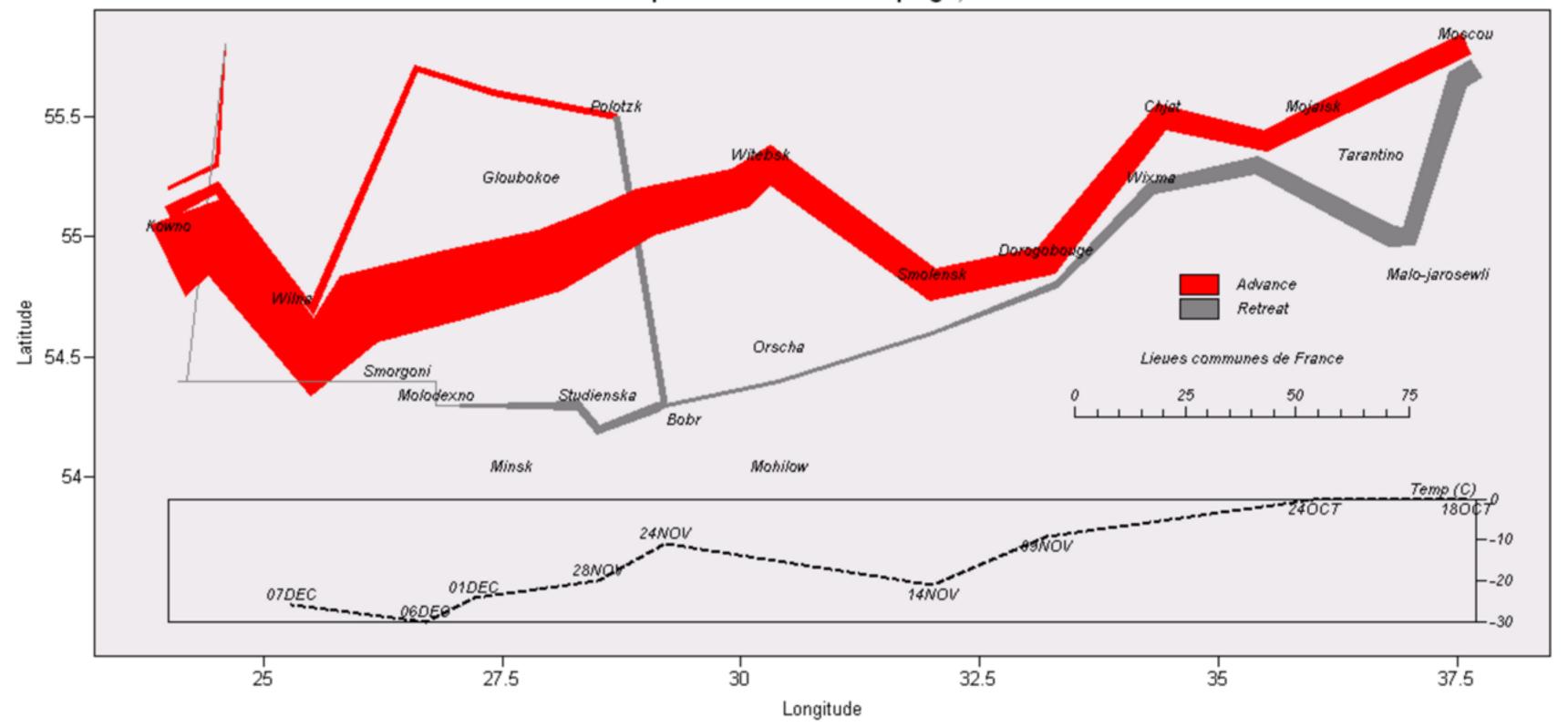




Napoleon's Russian Campaign, 1812

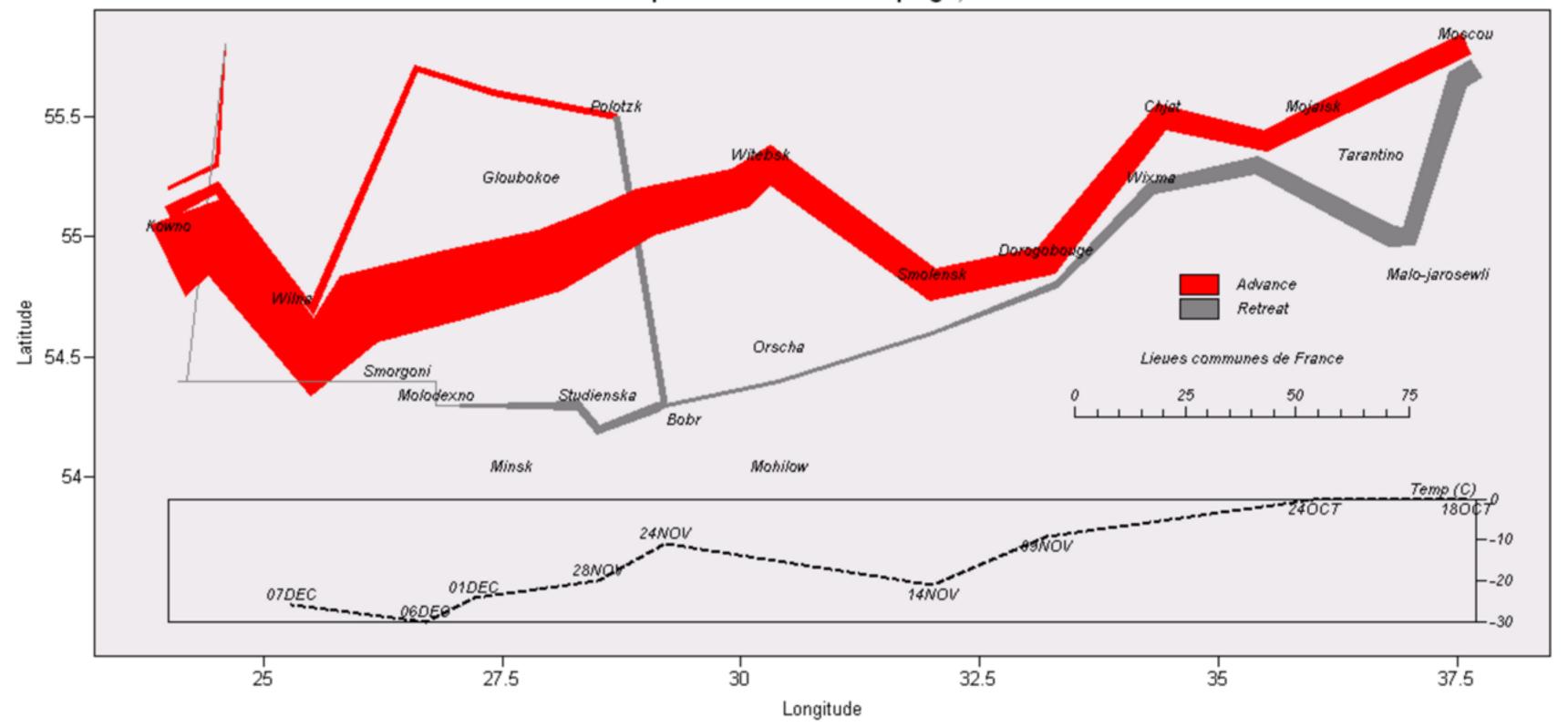


THIS IS A GRAPH "THAT MADE A NATION CRY"



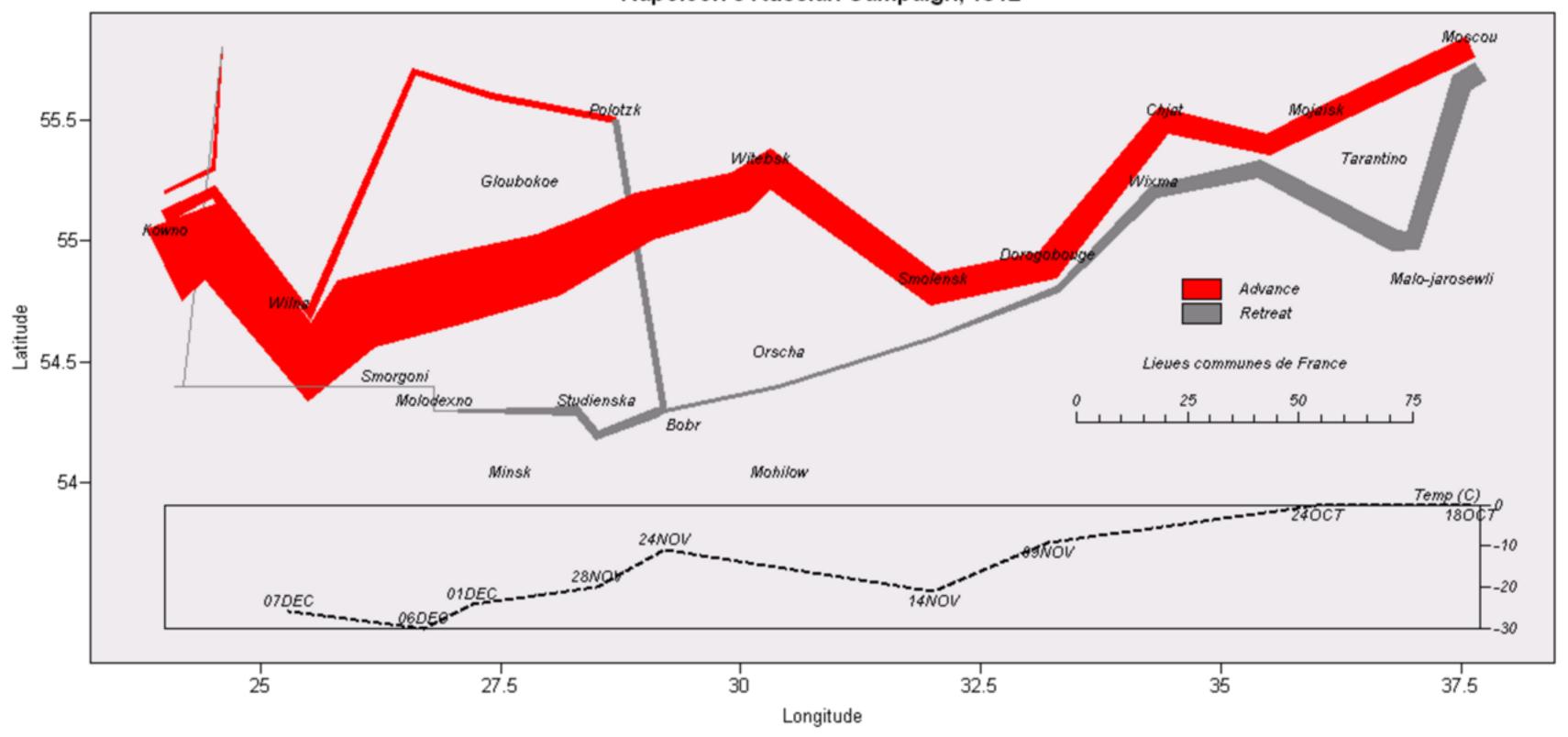
THIS WIDTH OF THE LINE SHOWS THE SIZE OF NAPOLEON'S ARMY



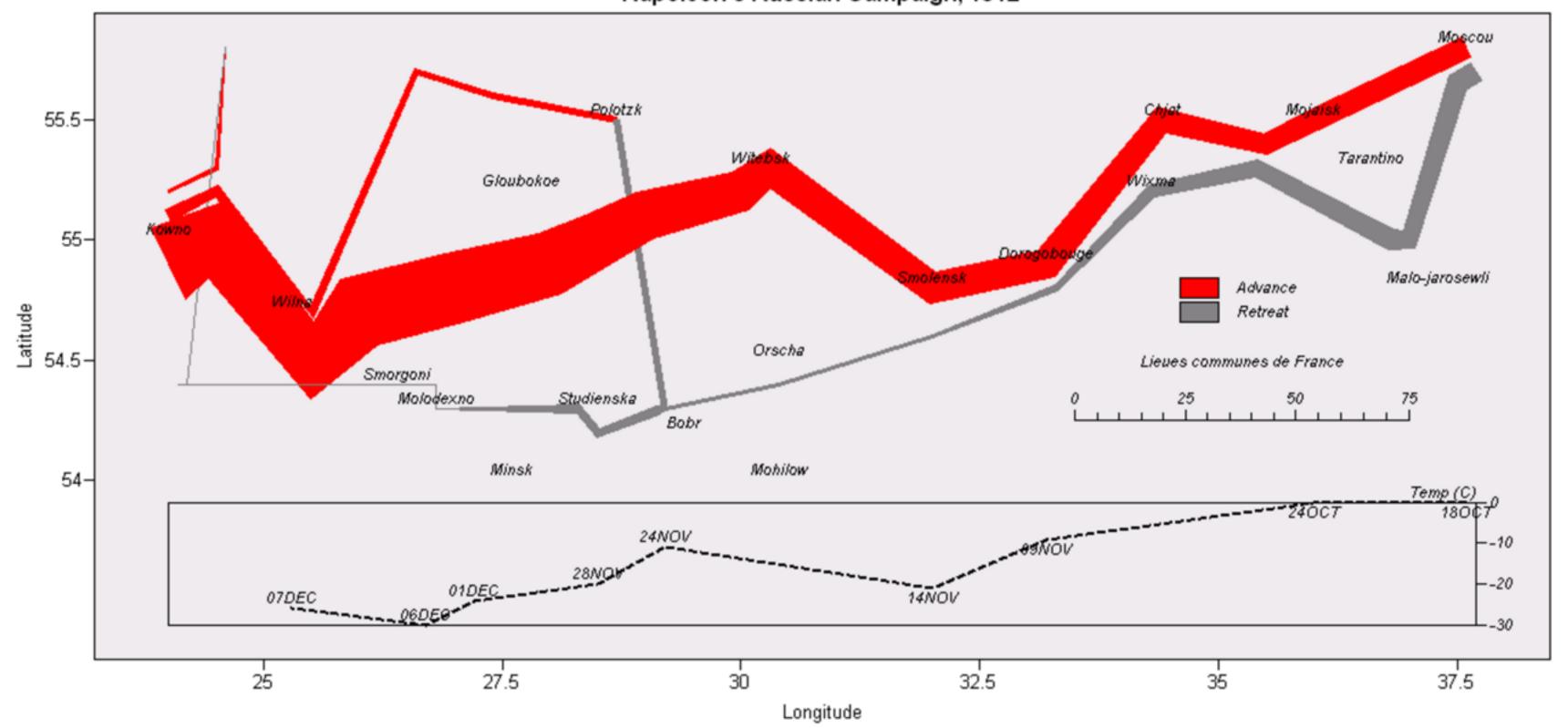


FROM LEFT TO RIGHT, THE RED PORTION SHOWS THE SIZE AS THE ARMY ADVANCED TO MOSCOW





FROM RIGHT TO LEFT, THE GREY PORTION SHOWS THE SIZE AS THE ARMY RETREATED FROM MOSCOW



THIS POIGNANT EXAMPLE IS ONE OF THE FIRST WELL-KNOWN DATA VISUALIZATIONS THAT "TELL A STORY" WITH DATA

"TELL A STORY" WITH DATA

THERE ARE MANY STANDARD WAYS TO VISUALIZE DATA

USING THEM TO TELL A STORY IS MORE AN ART THAN A SCIENCE

"TELL A STORY" WITH DATA

USING THEM TO TELL A STORY IS MORE AN ART THAN A SCIENCE

THE ARTISTRY COMES WITH EXPERIENCE

THE MECHANICS ARE EASY THOUGH AND IMPORTANT TO KNOW

R HAS BUILT-IN FUNCTIONALITY FOR LOTS OF DATA VISUALIZATIONS

HISTOGRAMS
BOXPLOTS
Q-Q PLOTS
SCATTER PLOTS

LINE CHARTS BAR HEAT MAPS PIE CHARTS HISTOGRAMS BOXPLOTS HEAT MAPS

Q-Q PLOTS SCATTER PLOTS PIE CHARTS LINE CHARTS

BAR

R HAS MANY PACKAGES THAT HELP US GO BEYOND THESE BASIC VISUALIZATIONS

AND ENHANCE THEM IN MANY WAYS

R HAS MANY PACKAGES THAT HELP US
GO BEYOND THESE BASIC
VISUALIZATIONS

RCOLORBREWER CONTROL COLOR PALETTES

GGPLOT2 COMPLEX 2D GRAPHS

```
x <- 1:10
y <- rnorm(10)
```

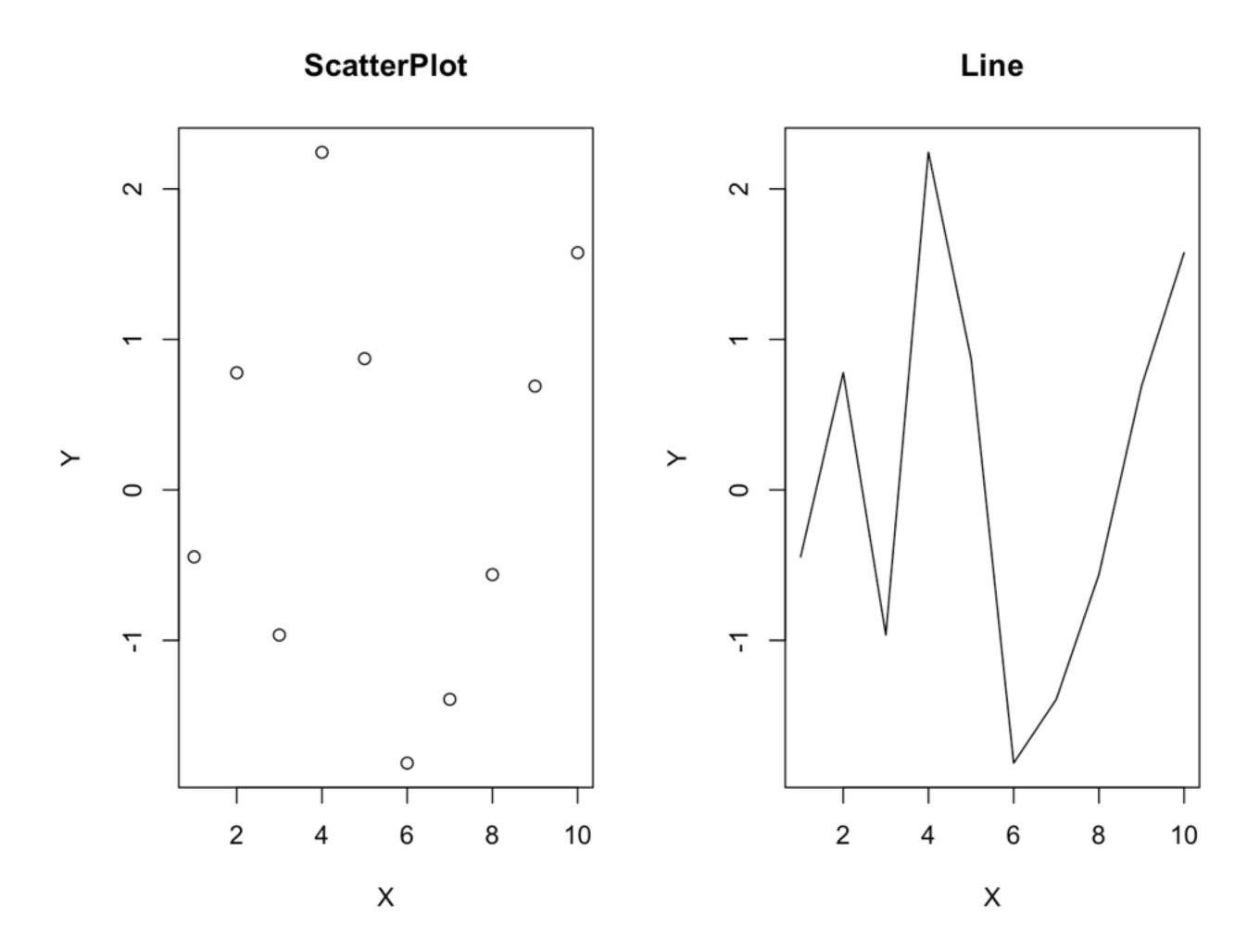
LET'S PLOT X VS Y IN 2 PIFFERENT WAYS

```
x <- 1:10
y <- rnorm(10)
par(mfrow=c(1,2))
```

THIS SET'S UP A PISPLAY FOR MULTIPLE PLOTS

THE PLOTS WILL BE ARRANGED IN 1 ROW AND 2 COLUMNS

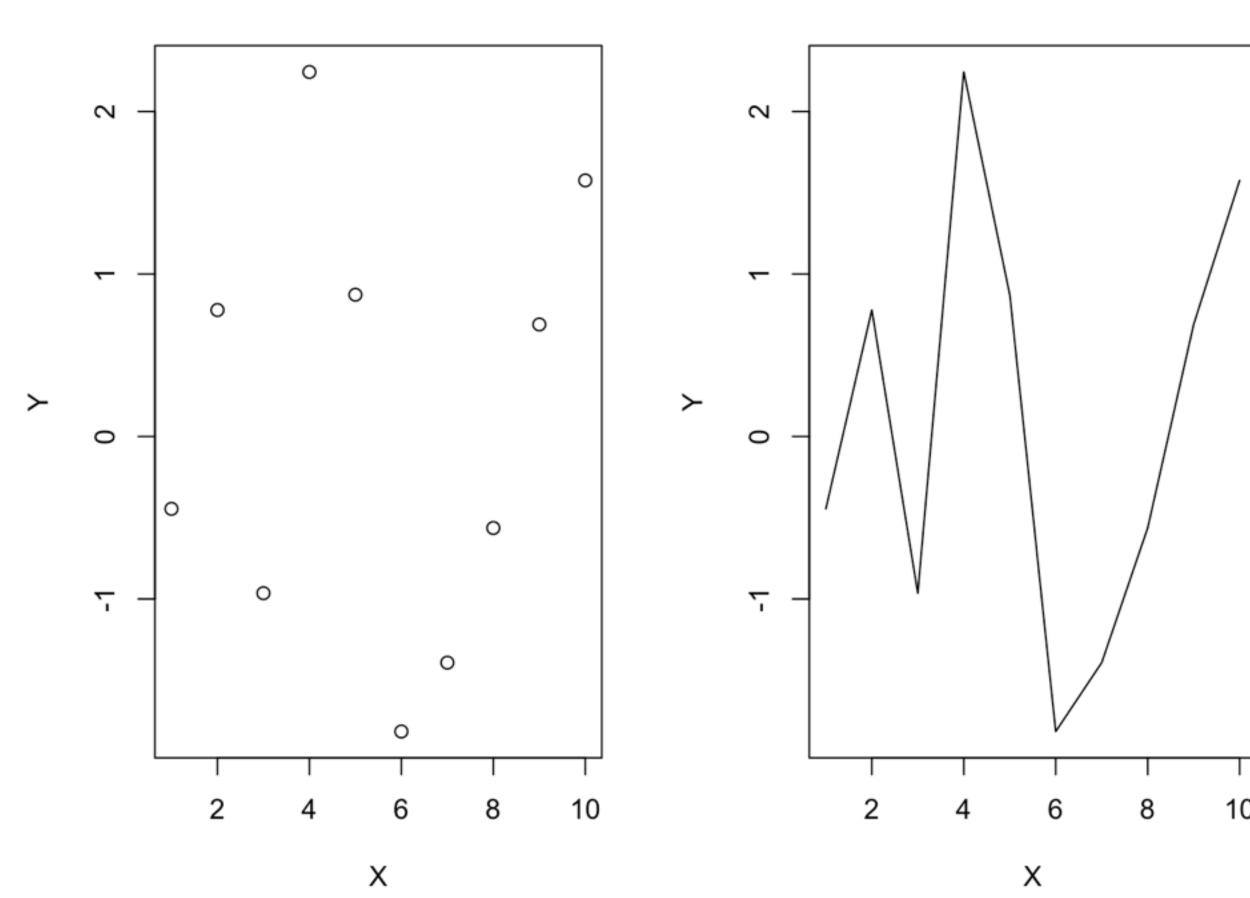
```
plot(x,y,type="p",main="ScatterPlot",xlab="X",ylab="Y")
plot(x,y,type="l",main="ScatterPlot",xlab="X",ylab="Y")
```



THE PLOT() FUNCTION IS THE MOST BASIC OF ALL VISUALIZATION FUNCTIONS

```
plot(x,y,type="p",main="ScatterPlot",xlab="X",ylab="Y")
plot(x,y,type="l",main="Line",xlab="X",ylab="Y")
```

X AND Y VARIABLES

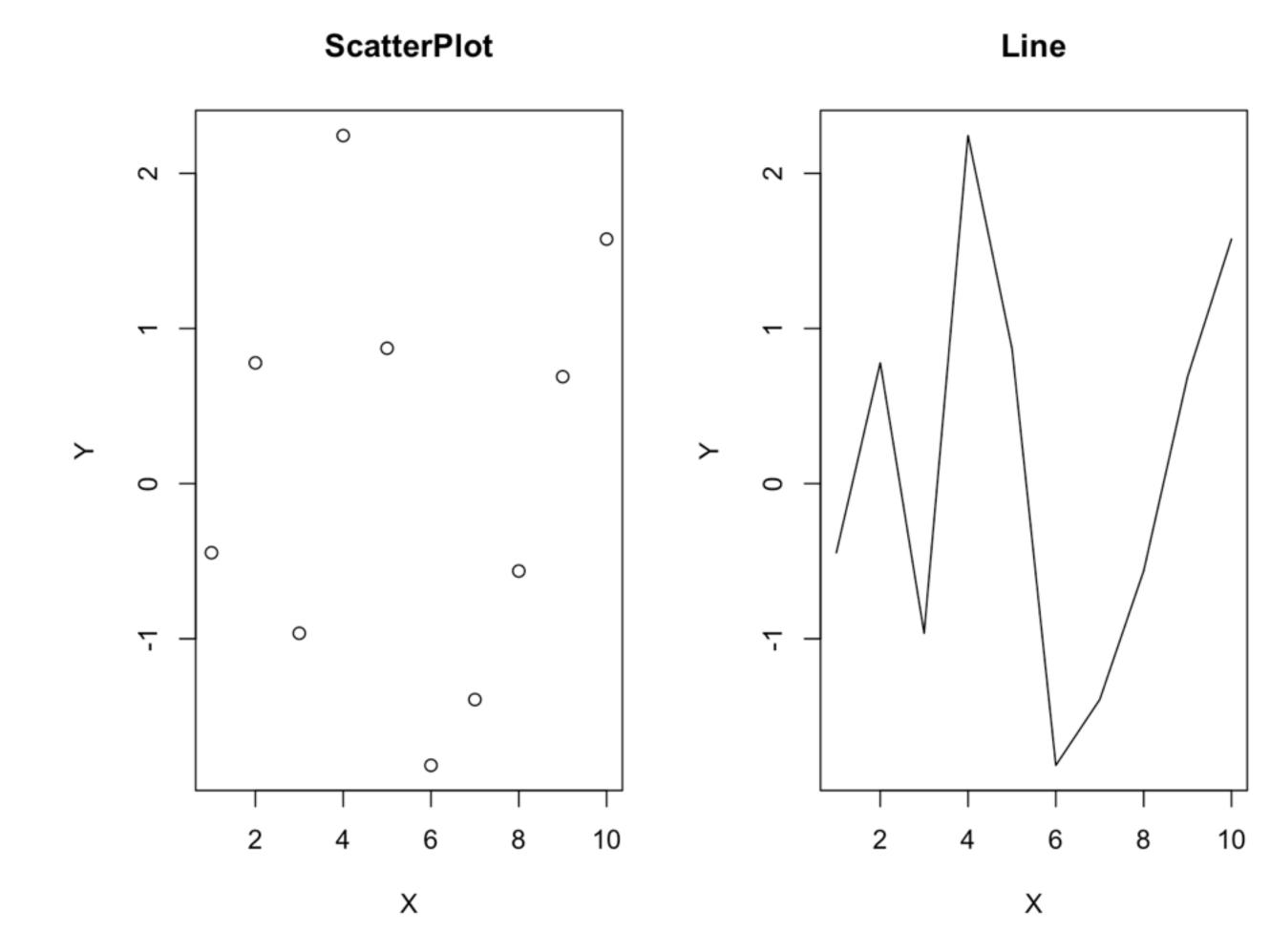


Line

ScatterPlot

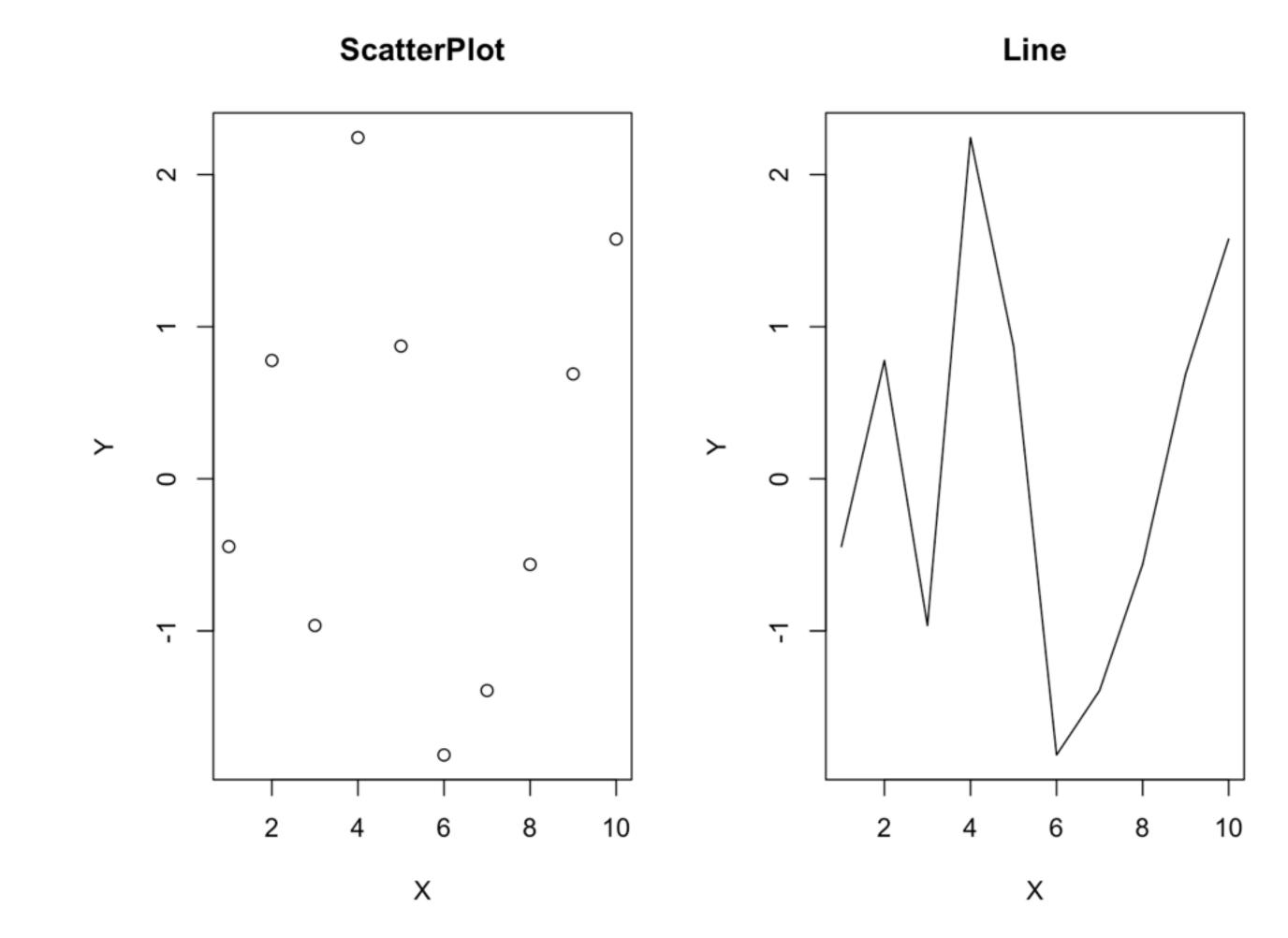
```
plot(x,y,type="p",main="ScatterPlot",xlab="X",ylab="Y")
plot(x,y,type="l",main="Line",xlab="X",ylab="Y")
```

THE TYPE OF PLOT, P = POINT L=LINE



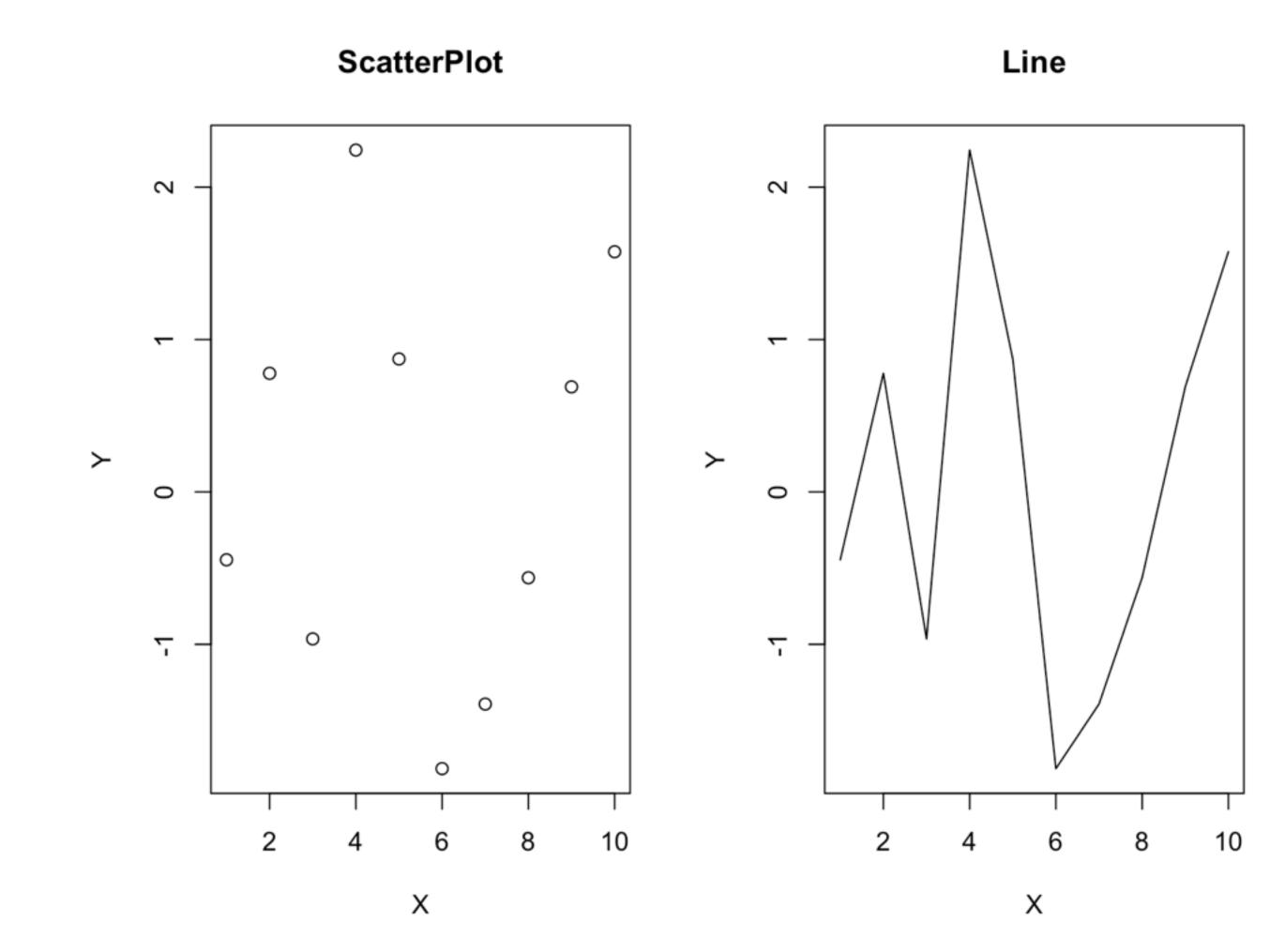
```
plot(x,y,type="p" main="ScatterPlot",xlab="X",ylab="Y")
plot(x,y,type="l",main="Line",xlab="X",ylab="Y")
```

THE TITLE OF THE PLOT



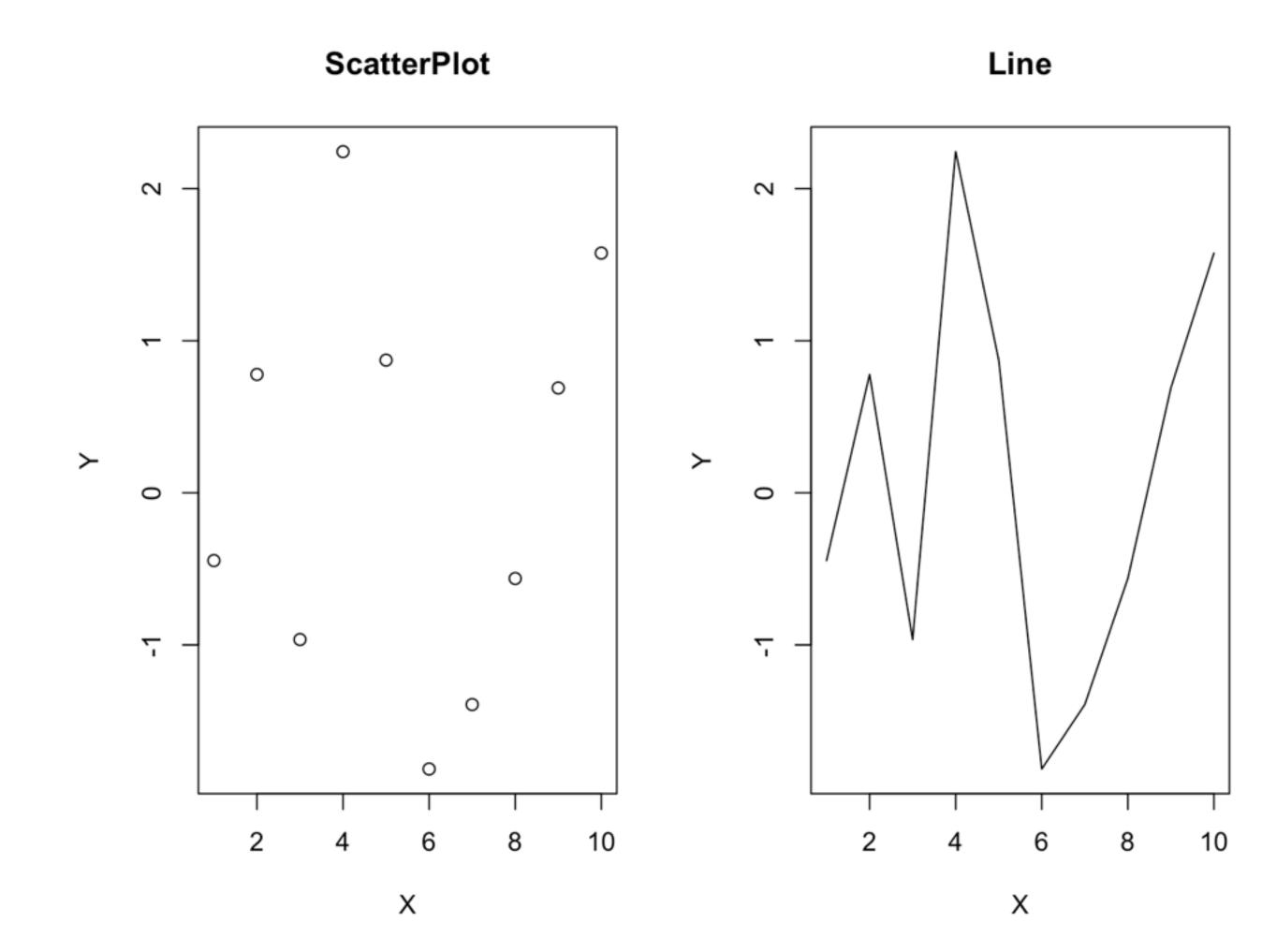
```
plot(x,y,type="p",main="ScatterPlot(,xlab="X"),ylab="Y")
plot(x,y,type="l",main="Line",xlab="X",ylab="Y")
```

THE X AXIS LABEL OF THE PLOT



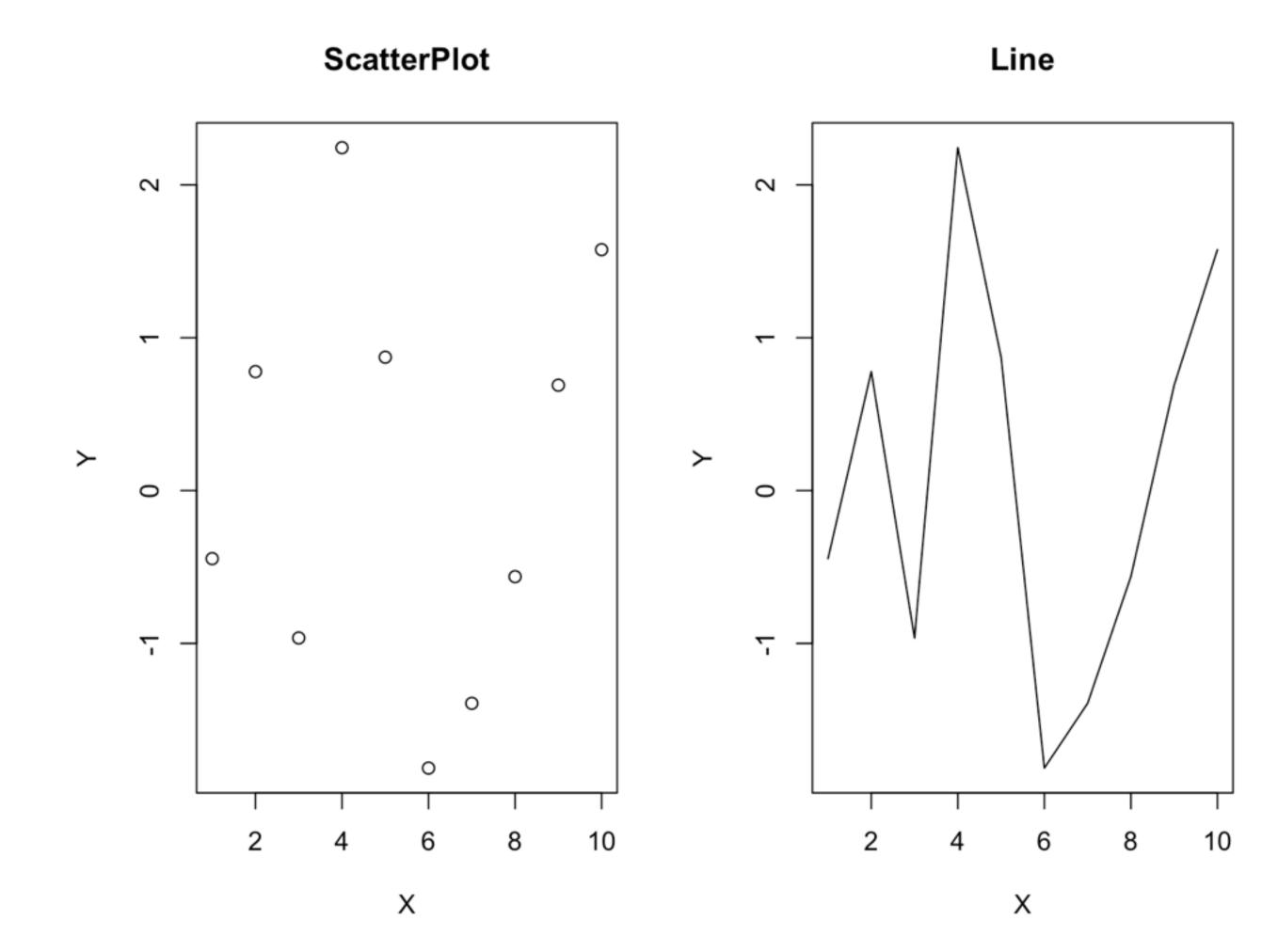
```
plot(x,y,type="p",main="ScatterPlot",xlab="X"(ylab="Y"))
plot(x,y,type="l",main="Line",xlab="X",ylab="Y")
```

THE Y AXIS LABEL OF THE PLOT



```
plot(x,y,type="p",main="ScatterPlot",xlab="X",ylab="Y")
plot(x,y,type="l",main="Line",xlab="X",ylab="Y")
```

THESE OPTIONS ARE THE SAME FOR OTHER FUNCTIONS LIKE HIST() ETC



WE'VE SEEN HOW TO DRAW A HISTOGRAM BEFORE

HERE IS SOME PATA

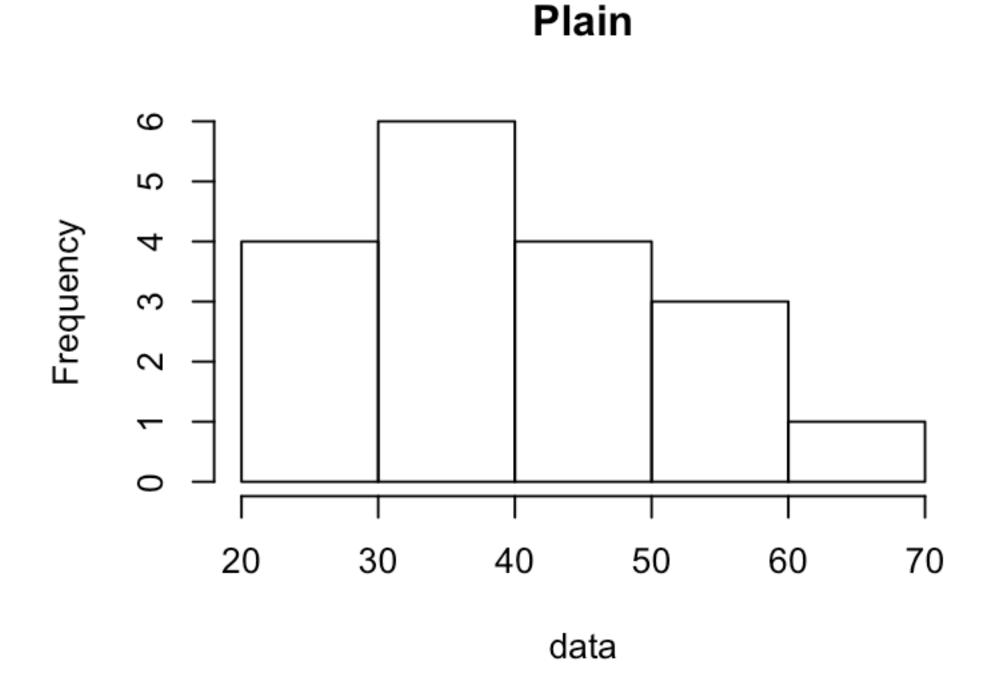
```
data \leftarrow c(40, 30, 20, 45, 60, 35, 40, 50, 55, 53, 22, 27, 48, 62, 33, 35, 40, 4
```

bins <- seq(20,70,by=10)

THESE ARE THE INTERVALS FOR THE HISTOGRAM

```
par(mfrow=c(2,2))
hist(data,breaks = bins,main="Plain")
```

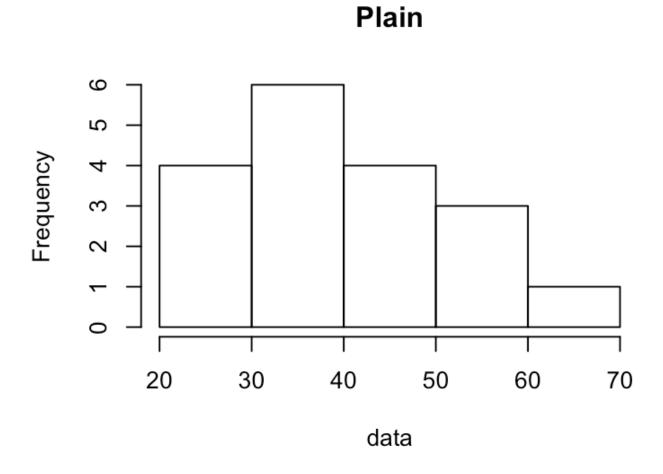
HIST() FUNCTION WILL DRAW A HISTOGRAM FOR THIS DATA USING THE SPECIFIED INTERVALS

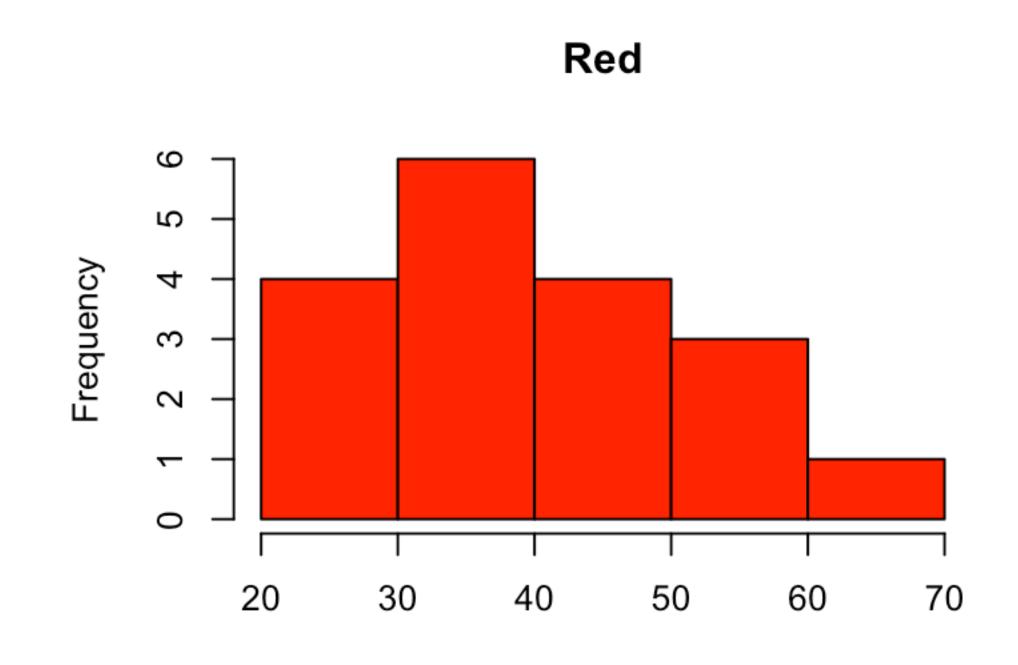


WE CAN ADD COLORS TO THIS HISTOGRAM USING COL OPTION

```
par(mfrow=c(2,2))
hist(data,breaks = bins,main="Plain")
```

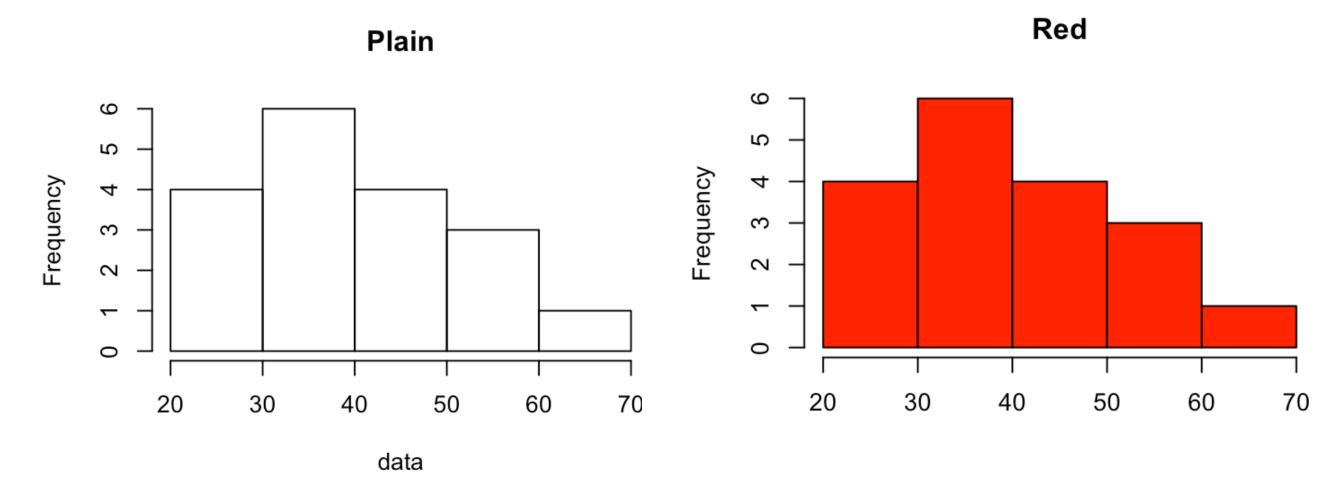
hist(data,breaks = bins,col=I("Red"),main="Red")





```
par(mfrow=c(2,2))
hist(data,breaks = bins,main="Plain")
```

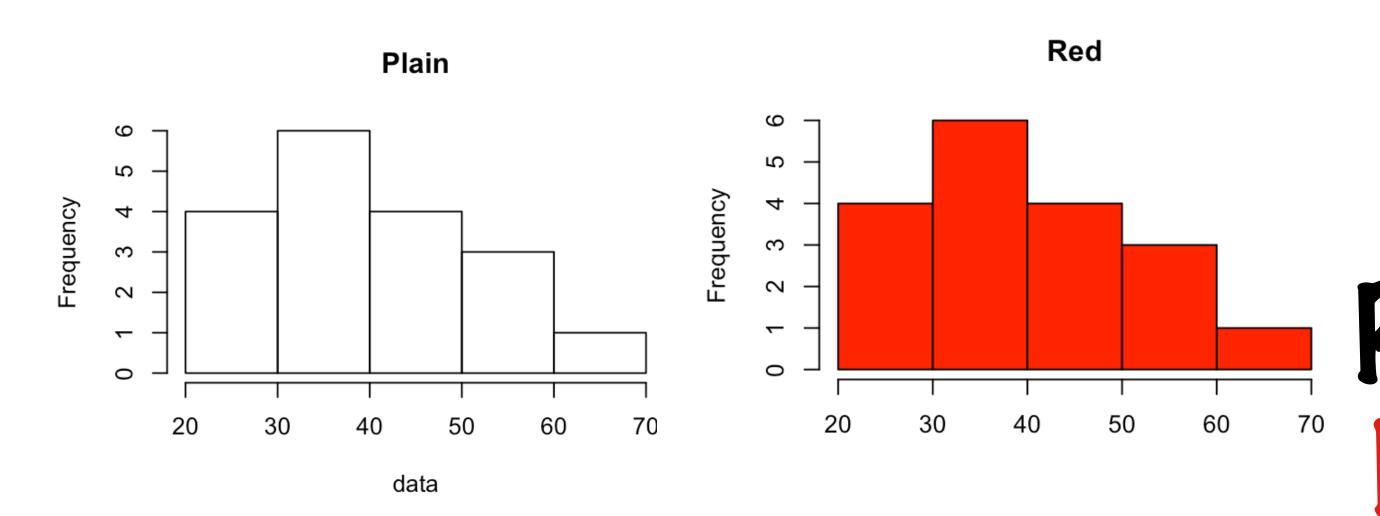
hist(data,breaks = bins,col=I("Red"),main="Red")



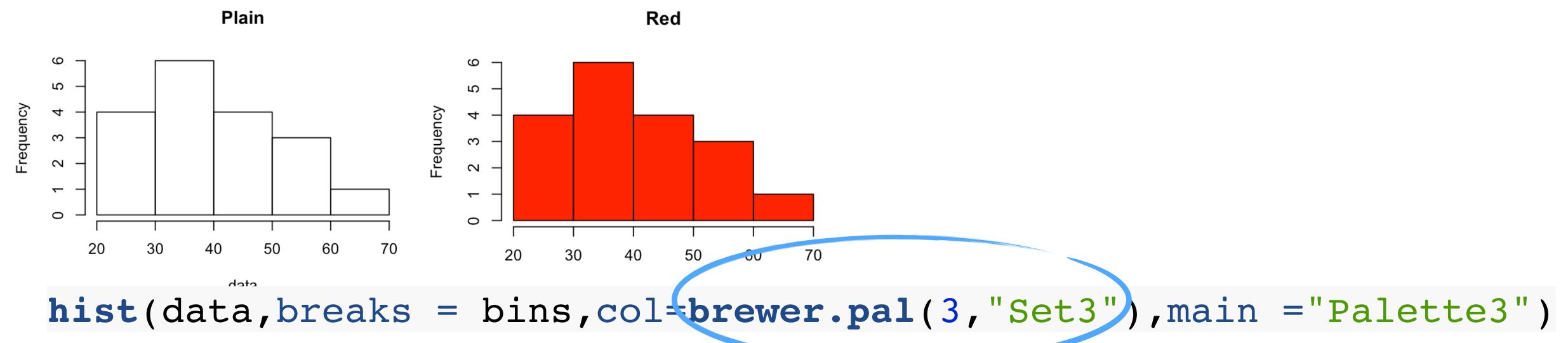
II) CAN BE USED TO ACCESS BUILT-IN COLORS

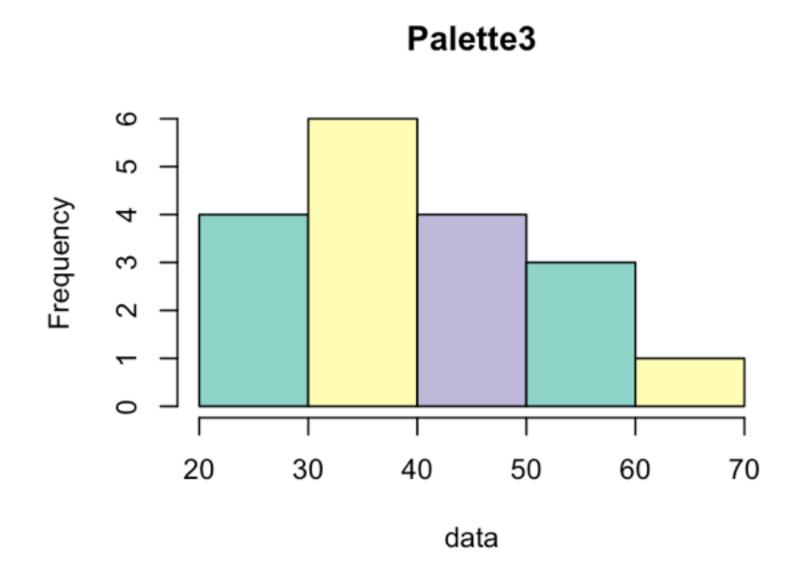
```
par(mfrow=c(2,2))
hist(data,breaks = bins,main="Plain")
hist(data,breaks = bins,col=I("Red"),main="Red")
```

library(RColorBrewer)

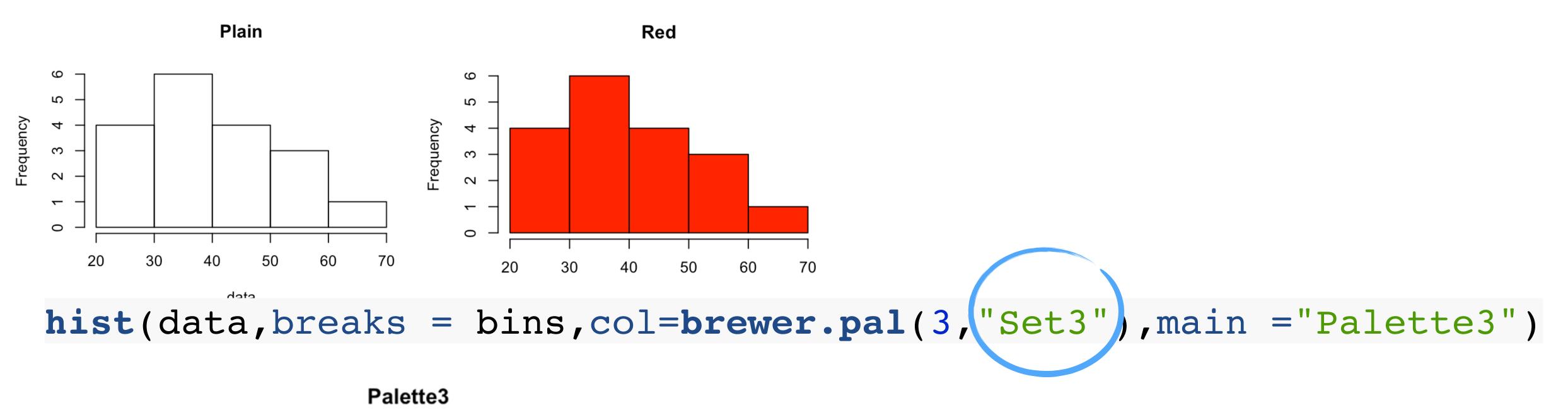


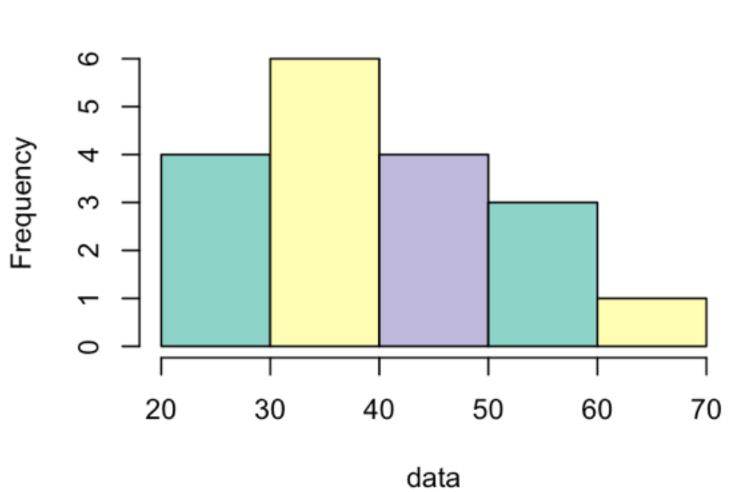
USE RCOLORBREWER PACKAGE TO ASSIGN A MULTICOLOR PALETTE, EACH BAR WILL HAVE A DIFFERENT COLOR



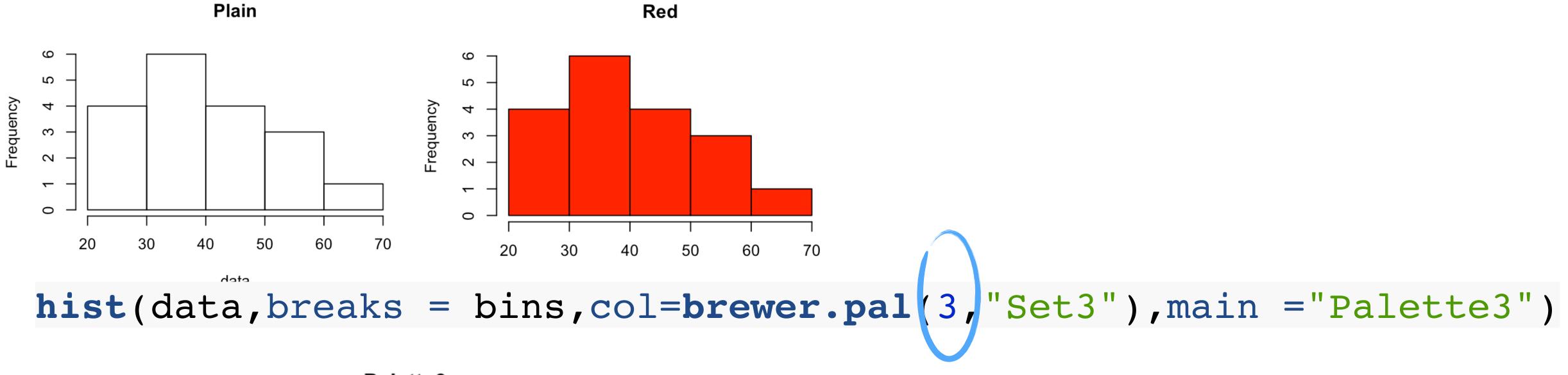


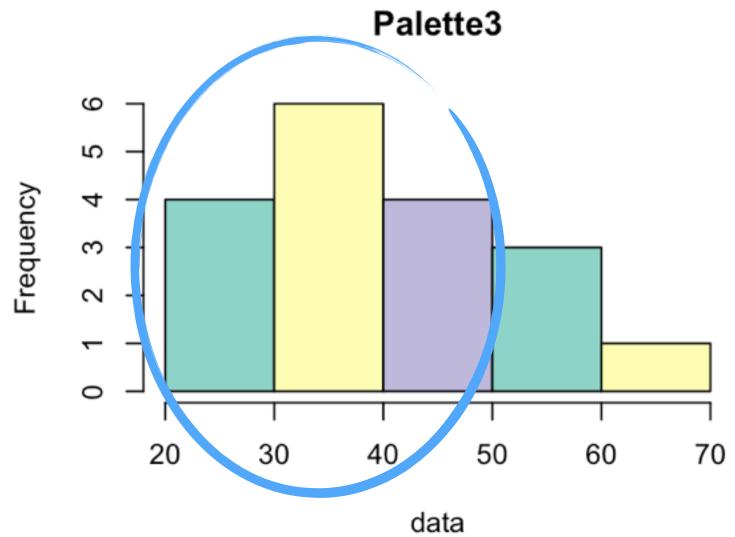
WE CAN SPECIFY THE PALETTE AND THE NUMBER OF COLORS TO PICK FROM THAT PALETTE



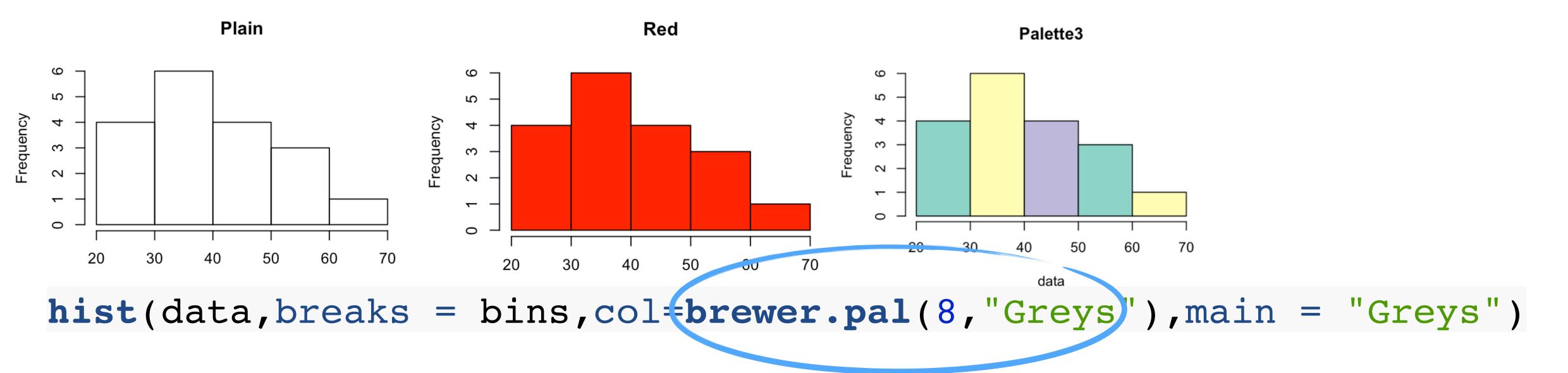


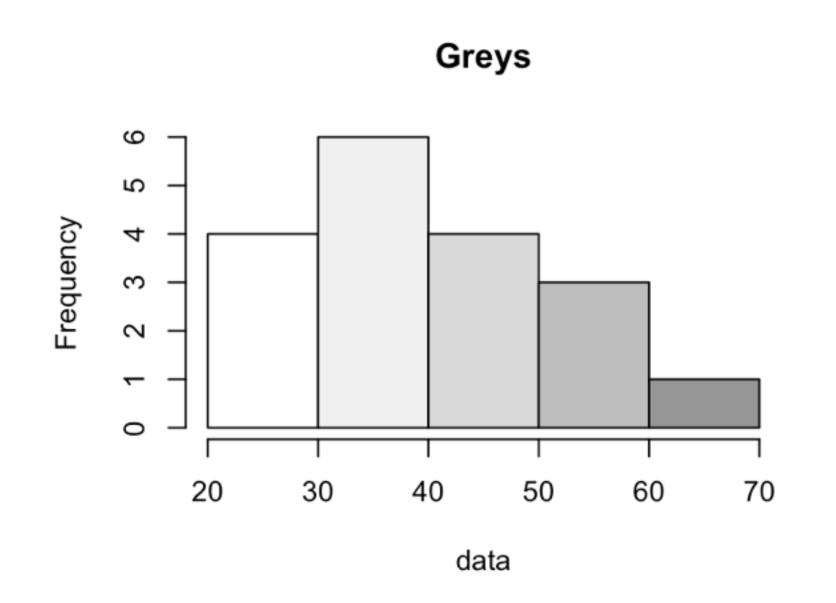
THIS IS ONE OF THE PALETES



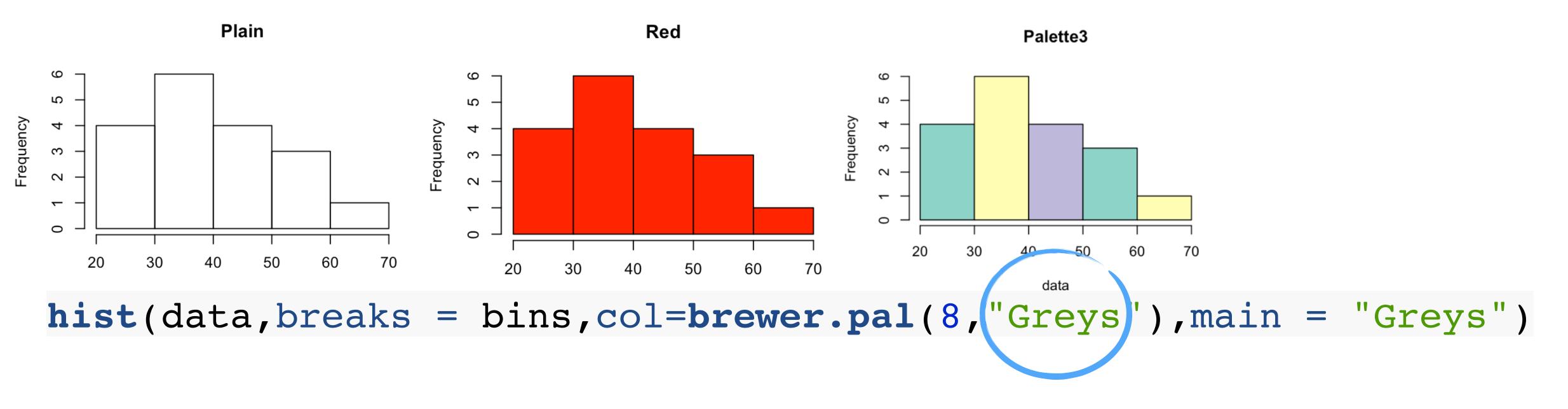


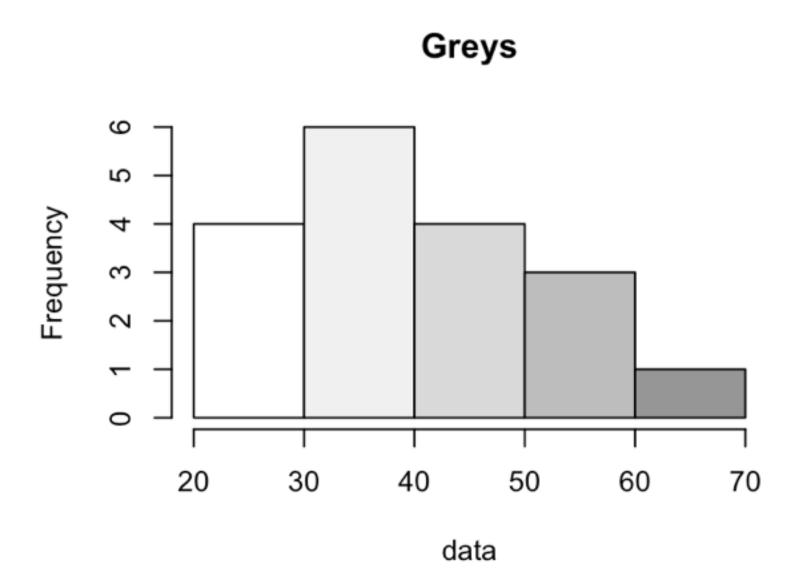
WE ARE USING 3 COLORS FROM THE PALETTE





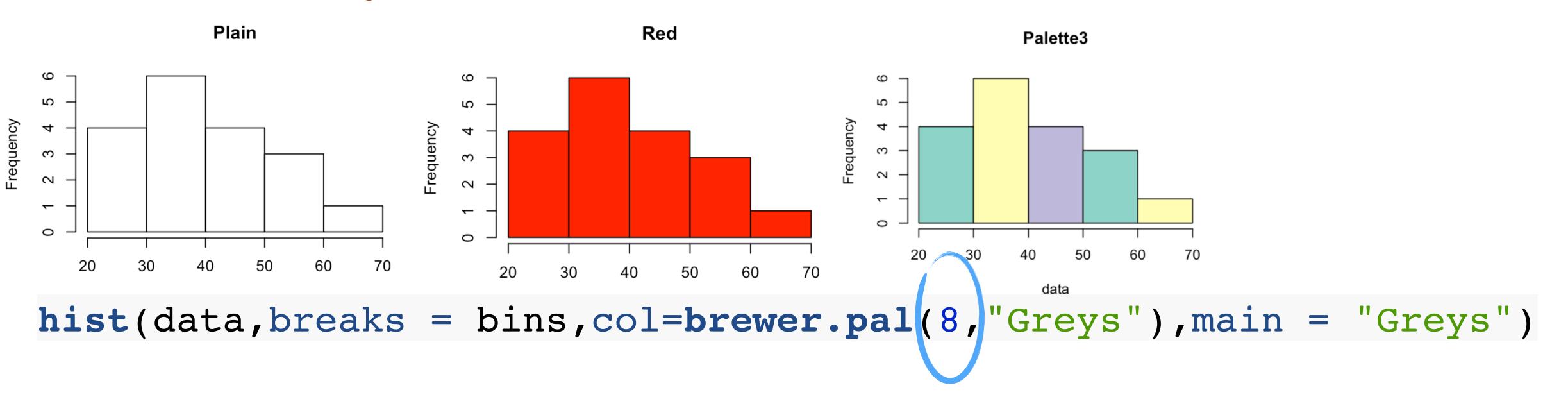
WE CAN SPECIFY THE PALETTE TO BE SHAPES OF A SINGLE COLOR

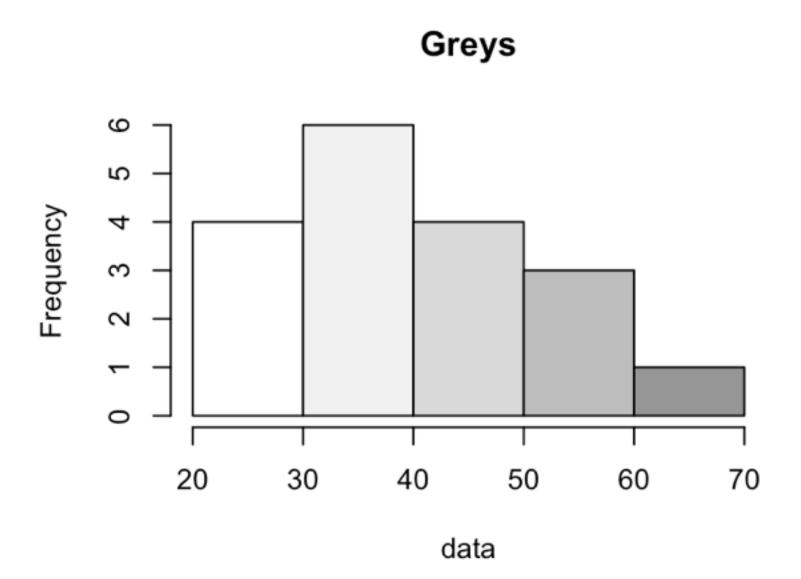




THE COLOR WHOSE SHAPES ARE CHOSEN

EXAMPLE 2: CONTROLLING COLOR PALETTES





THE NUMBER OF DISTINCT SHAPES

BARPLOTS ARE PRETTY SIMPLE:)

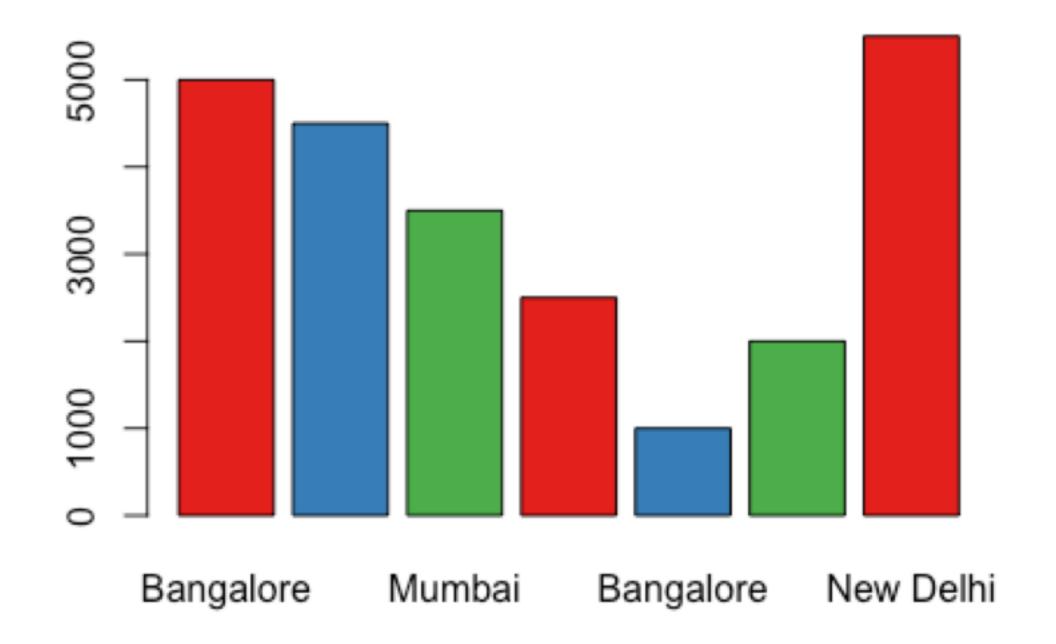
LET'S DRAW A BARPLOT TO REPRESENT SALES DATA

Bangalore	Clothing	INR 5000
New Delhi	Footwear	INR 4500
Mumbai	Cosmetics	INR 3500
Bangalore	Cosmetics	INR 2500
Bangalore	Footwear	INR 1000
Mumbai	Clothing	INR 2000
New Delhi	Clothing	INR 5500

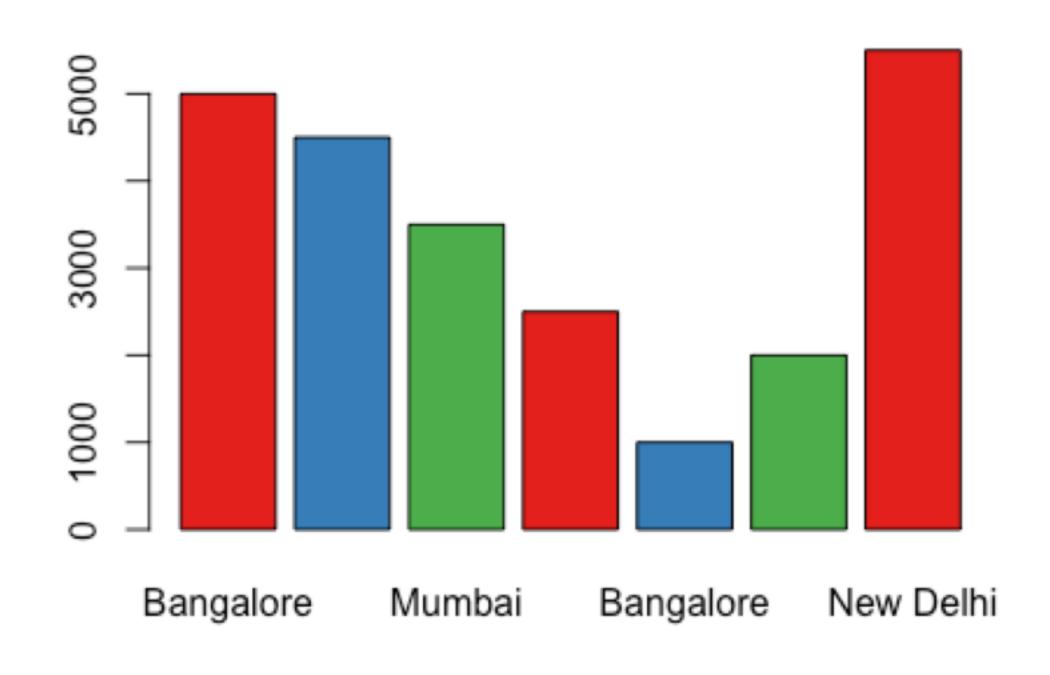
EXAMPLE 3: DRAWING A BARPLOT WE'VE PUT THIS DATA IN A DATA FRAME

sales <-data.frame(city,category,saleAmount)</pre>

barplot(sales\$saleAmount,names.arg = sales\$city,col=brewer.pal(3,"Set1"))

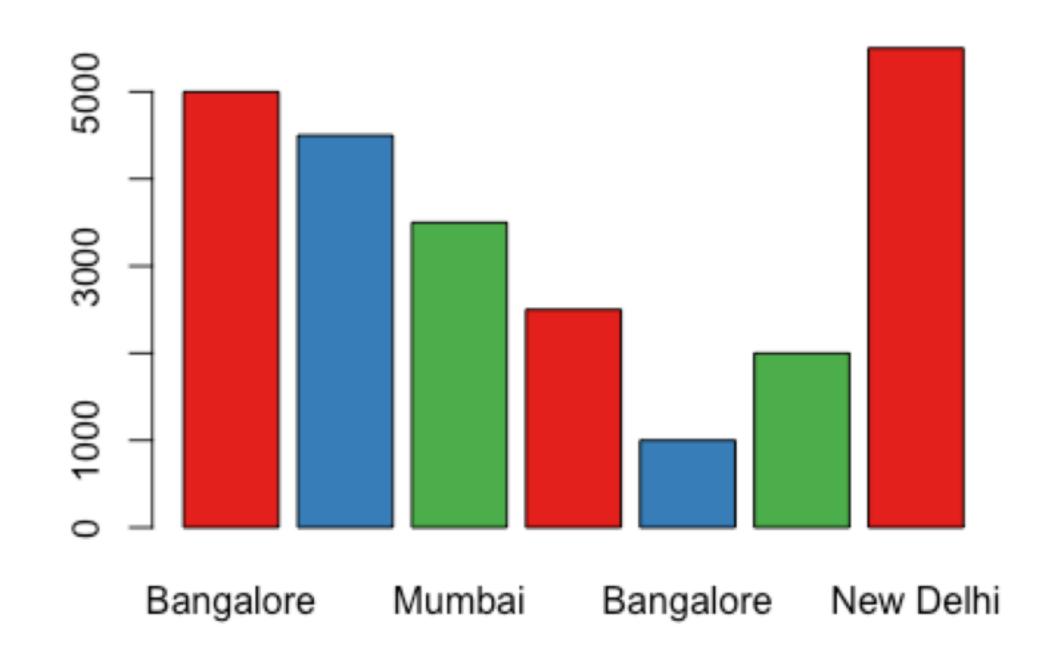


barplot(sales\$saleAmount,names.arg = sales\$city,col=brewer.pal(3,"Set1"))



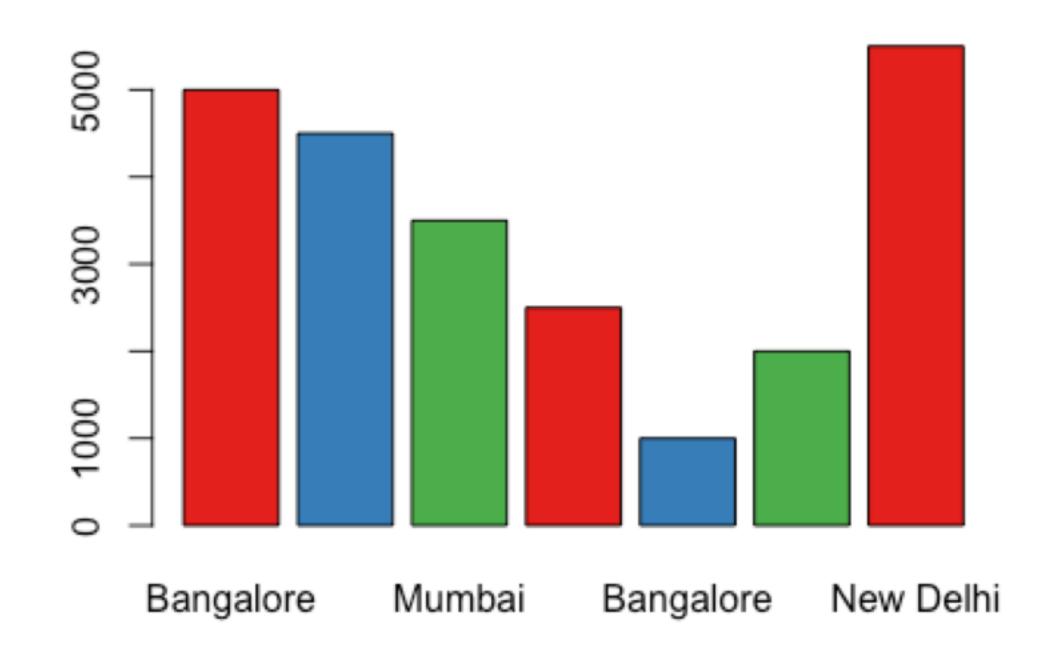
THE HEIGHTS OF THE BARS

barplot(sales\$saleAmount,names.arg = sales\$city/col=brewer.pal(3,"Set1"))



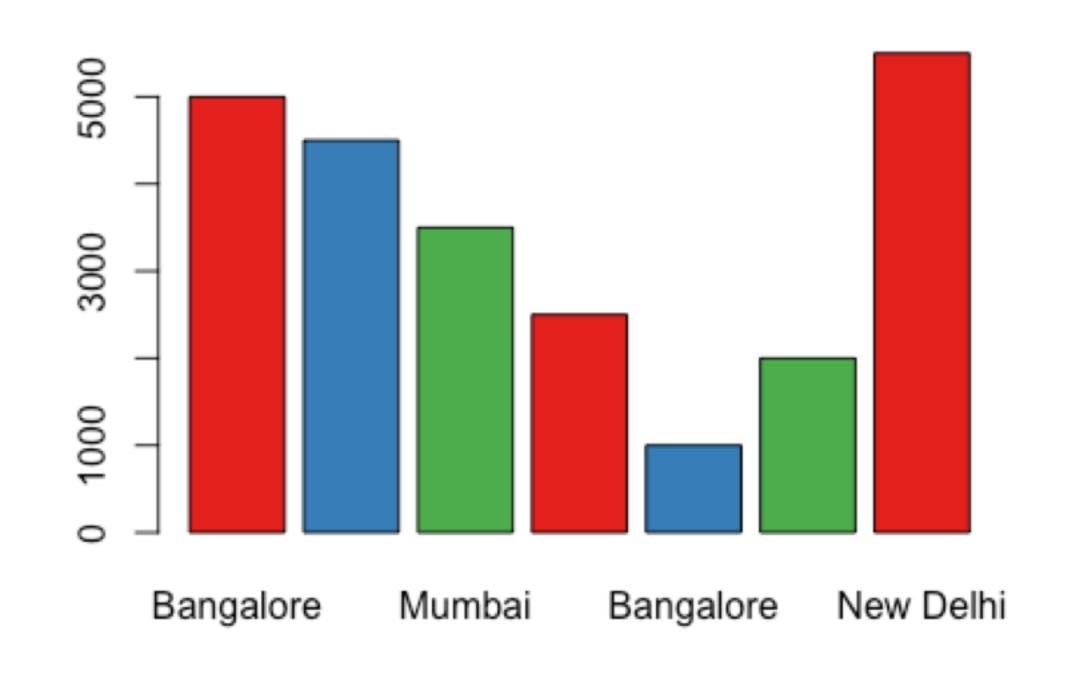
THE NAMES OF THE BARS

barplot(sales\$saleAmount,names.arg = sales\$city,col=brewer.pal(3,"Set1"))



THE COLOR PALETTE TO USE

barplot(sales\$saleAmount,names.arg = sales\$city,col=brewer.pal(3,"Set1"))

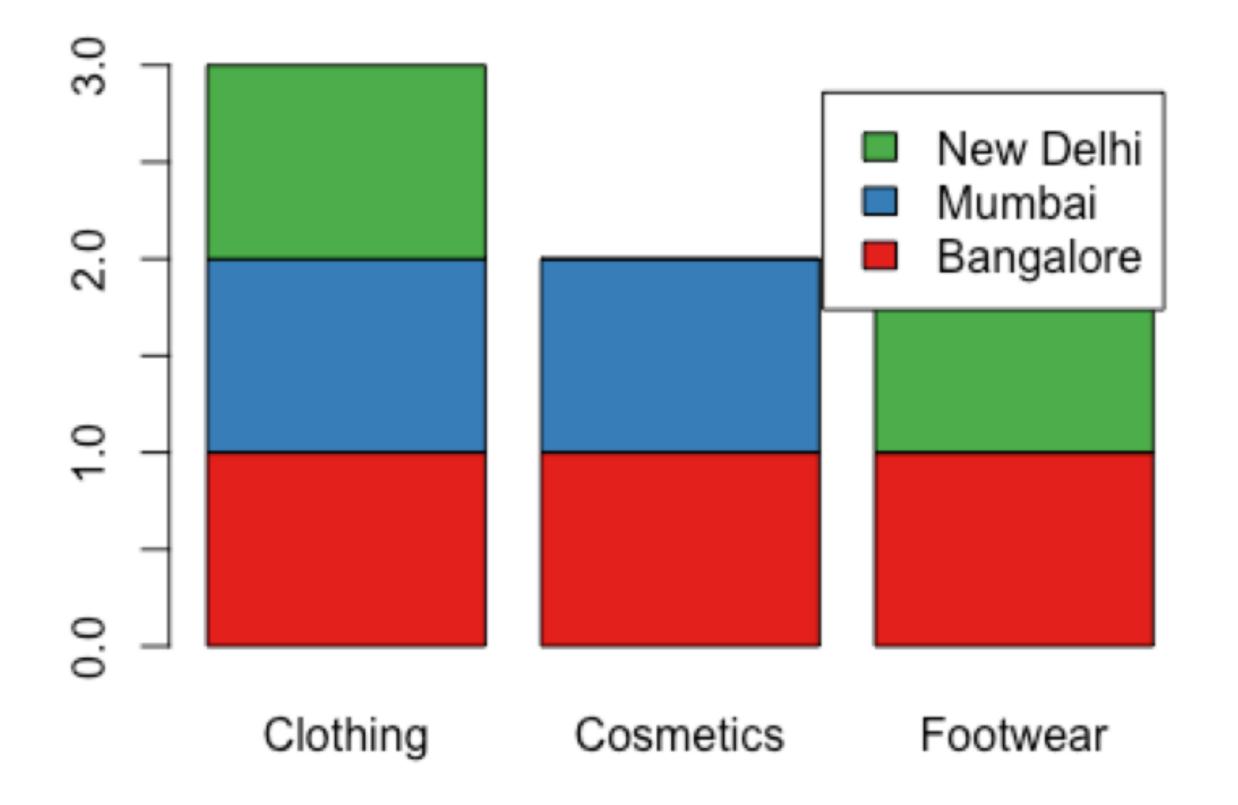


ONE COOL THING YOU CAN DO WITH BARPLOTS, IS PLOT STACKED BARS

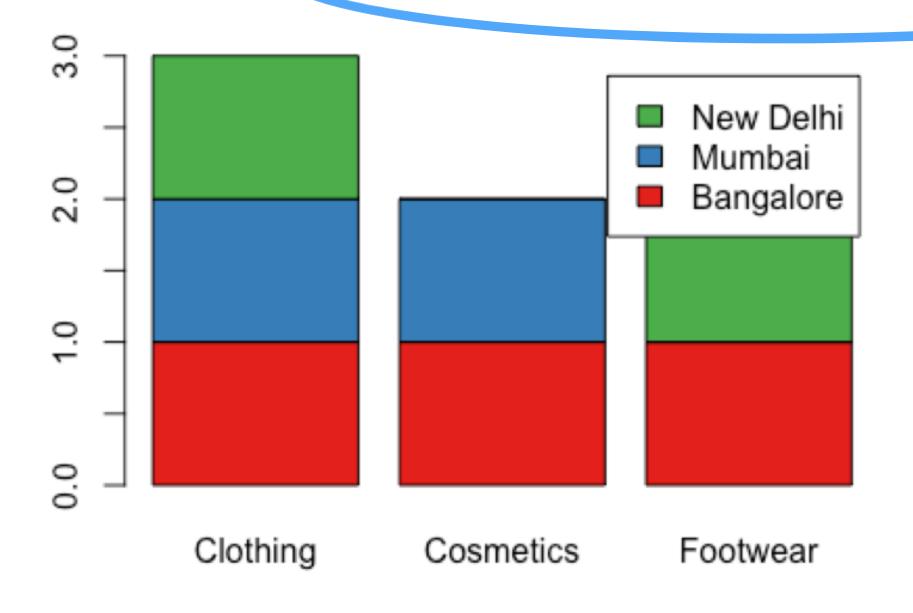
THESE ARE USEFUL TO VISUALIZE 2-DIMENSIONAL DATA

EXAMPLE 3: DRAWING A BARPLOT LET'S STACK THE SALES PER CITY BY CATEGORY

barplot(table(sales\$city,sales\$category),col=brewer.pal(3,"Set1"),legend.text=TRUE)



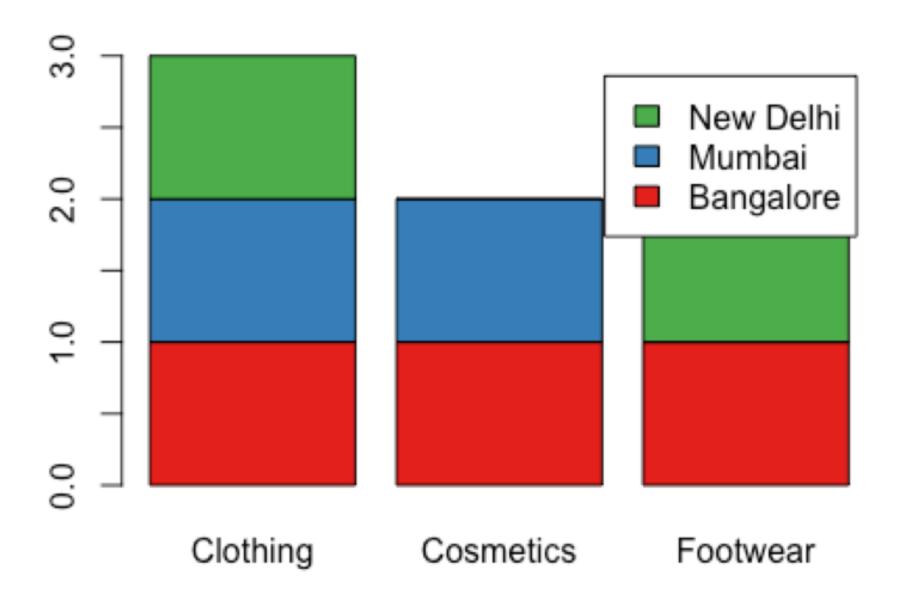
barplot(table(sales\$city,sales\$category), sol=brewer.pal(3, "Set1"),legend.text=TRUE)



THE HEIGHTS OF THE BARS AS A MATRIX

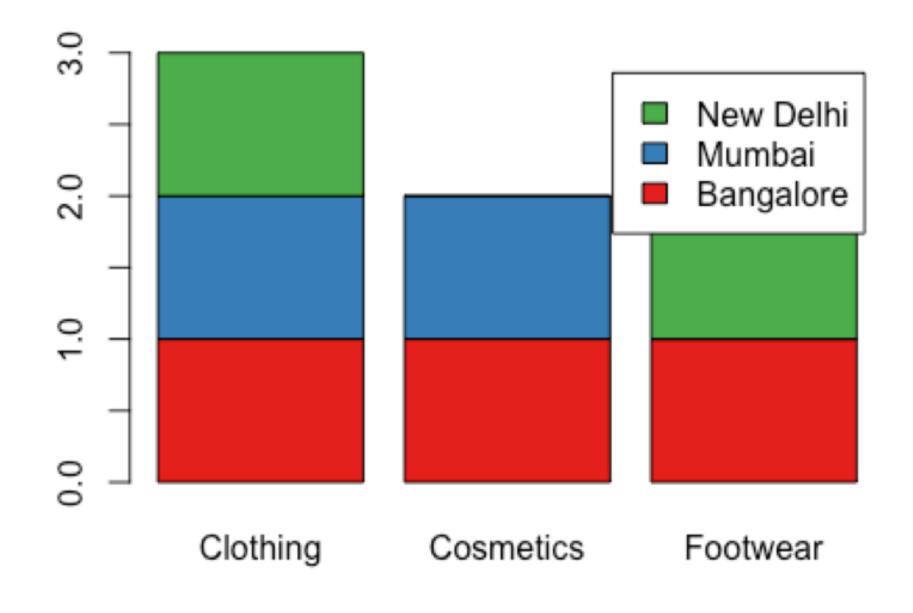
ROWS ARE CITIES COLUMNS ARE CATEGORIES VALUES ARE COUNTS

barplot(table(sales\$city,sales\$category),col=brewer.pal(3,"Set1"),legend.text=TRUE)



THE COLOR PALETTE TO BE USED

barplot(table(sales\$city,sales\$category),col=brewer.pal(3,"Set1"),legend.text=TRUE)



WE NEED A LEGEND TO DISTINGUISH THE CITIES

HEAT MAPS ARE PRETTY USEFUL TO VISUALIZE 2 DIMENSIONAL DATA

LET'S SAY WE HAVE 2 VARIABLES - HAIR COLOR AND EYE COLOR

HAIR COLOR AND EYE COLOR

WHAT COMBINATIONS OF HAIR COLOR AND EYE COLOR ARE VERY COMMON AND WHICH ARE UNCOMMON?

TO DRAW A HEAT MAP IN R, YOU NEED TO PUT YOUR DATA IN A MATRIX

THE VARIABLES ARE REPRESENTED BY THE ROW NAMES AND COLUMN NAMES

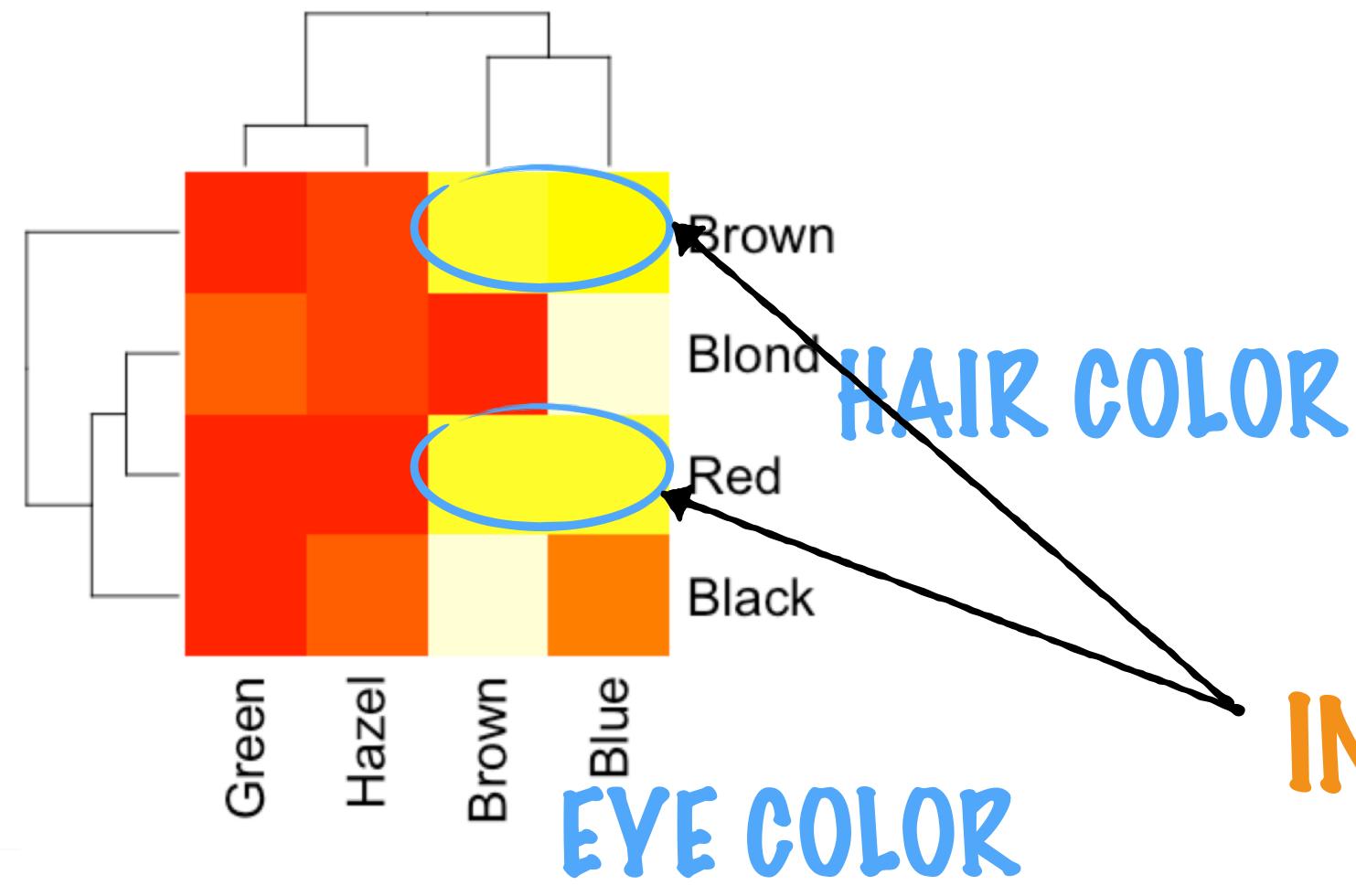
THE VALUES IN THE MATRIX REPRESENT THE INTENSITY OF COLOR

data (HairEyeColor)

THIS IS ANARRAY WITH 3 PIMENSIONS (HAIR COLOR, EYE COLOR AND GENDER)

INVOKE A BUILT-IN DATASET IN R WITH HAIR AND EYE COLOR OF SOME STUDENTS

heatmap(HairEyeColor[,,1])



LET'S LOOK AT A
HEAT MAP FOR
HAIR VS EYE
COLOR FOR
MALES

IMAX INTENSITY

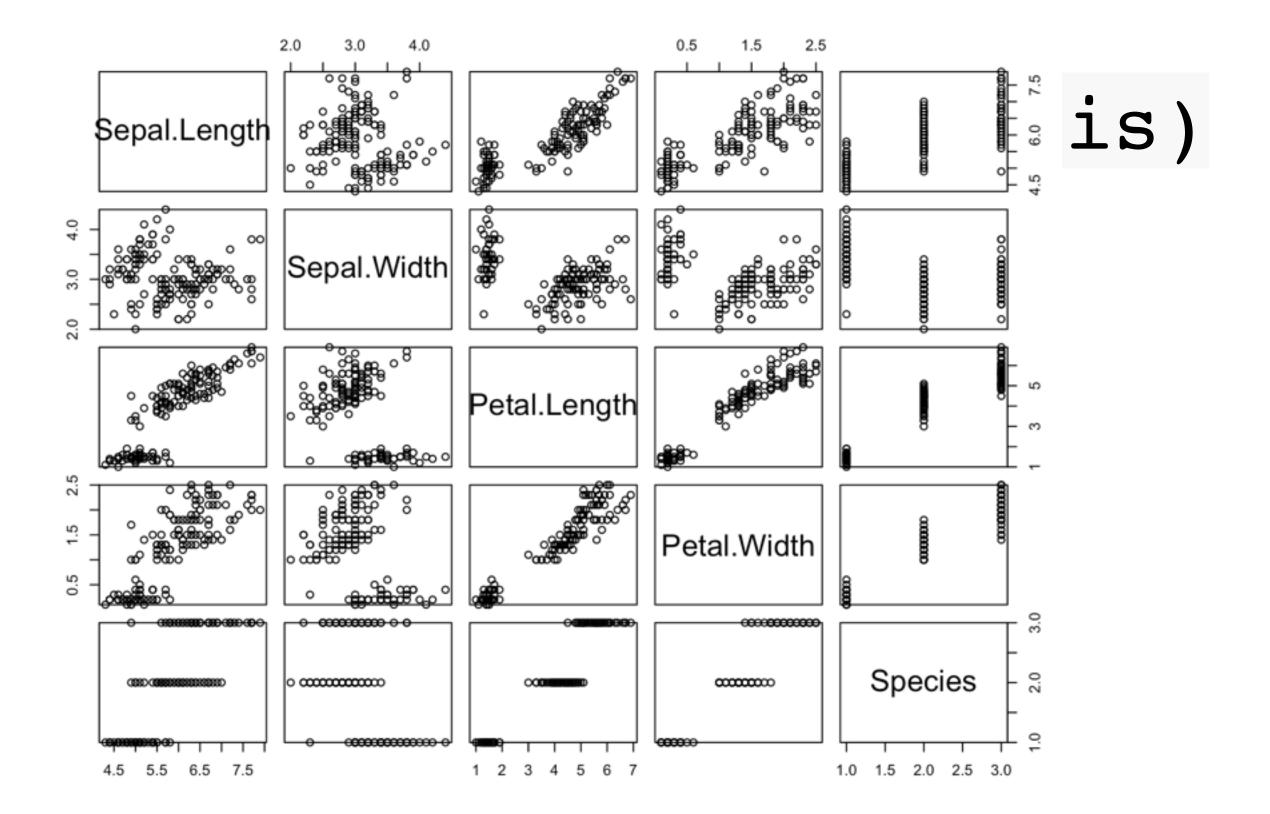
SCATTERPLOTS ARE VERY USEFUL TO UNDERSTAND RELATIONSHIPS BETWEEN VARIABLES

WHEN YOU HAVE LOTS OF VARIABLES, YOU CAN USE A SCATTERPLOT MATRIX TO VISUALLY EXPLORE THEIR RELATIONSHIPS

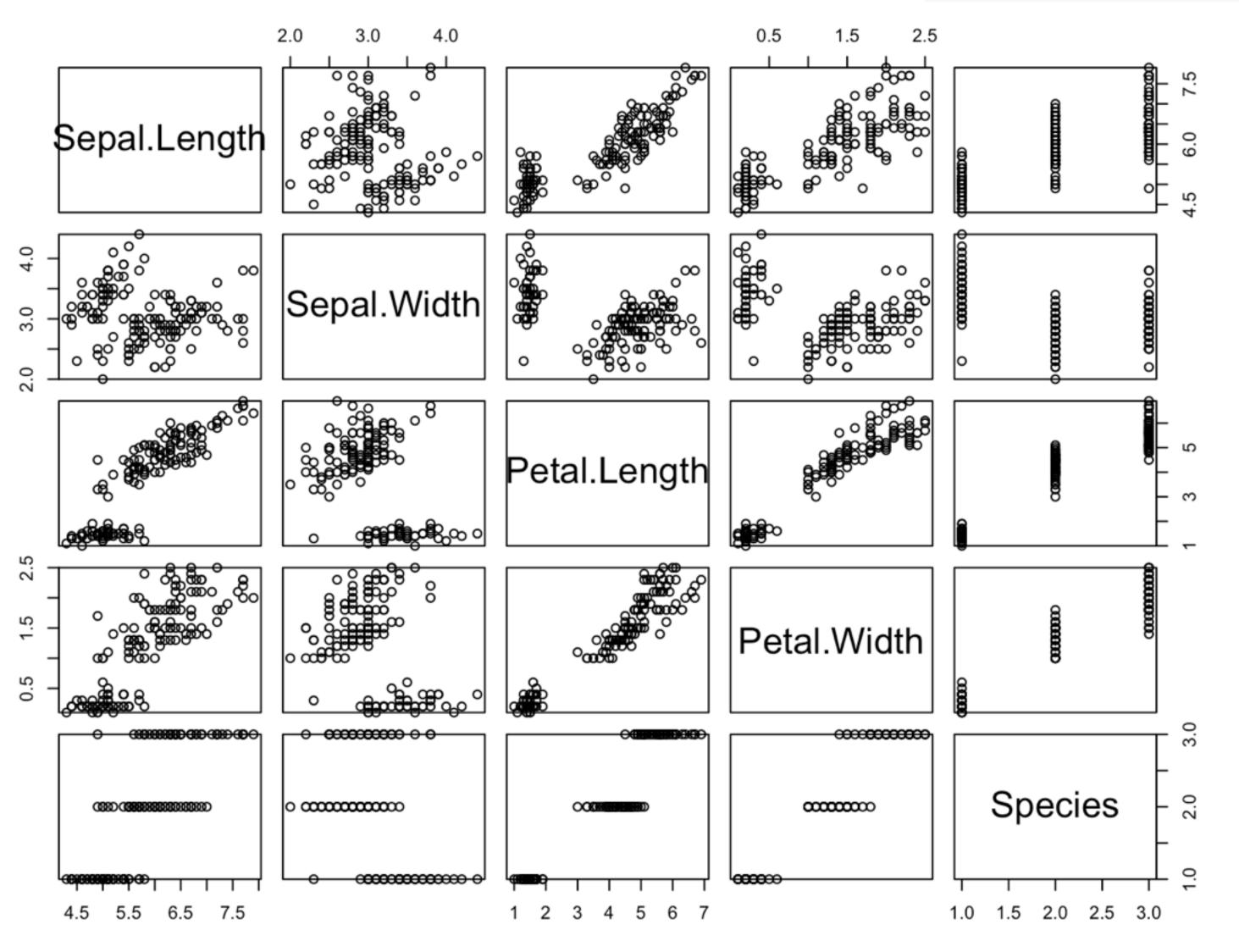
WHEN YOU USE PLOT() FUNCTION ON A DATA FRAME, IT WILL TREAT EACH COLUMN AS A VARIABLE AND PLOT A SCATTERPLOT MATRIX

IRIS IS A VERY FAMOUS DATASET WITH SOME MEASUREMENTS FROM DIFFERENT SPECIES OF IRIS FLOWERS

YOU CAN ACCESS THE IRIS DATASET IN R AS A DATA FRAME

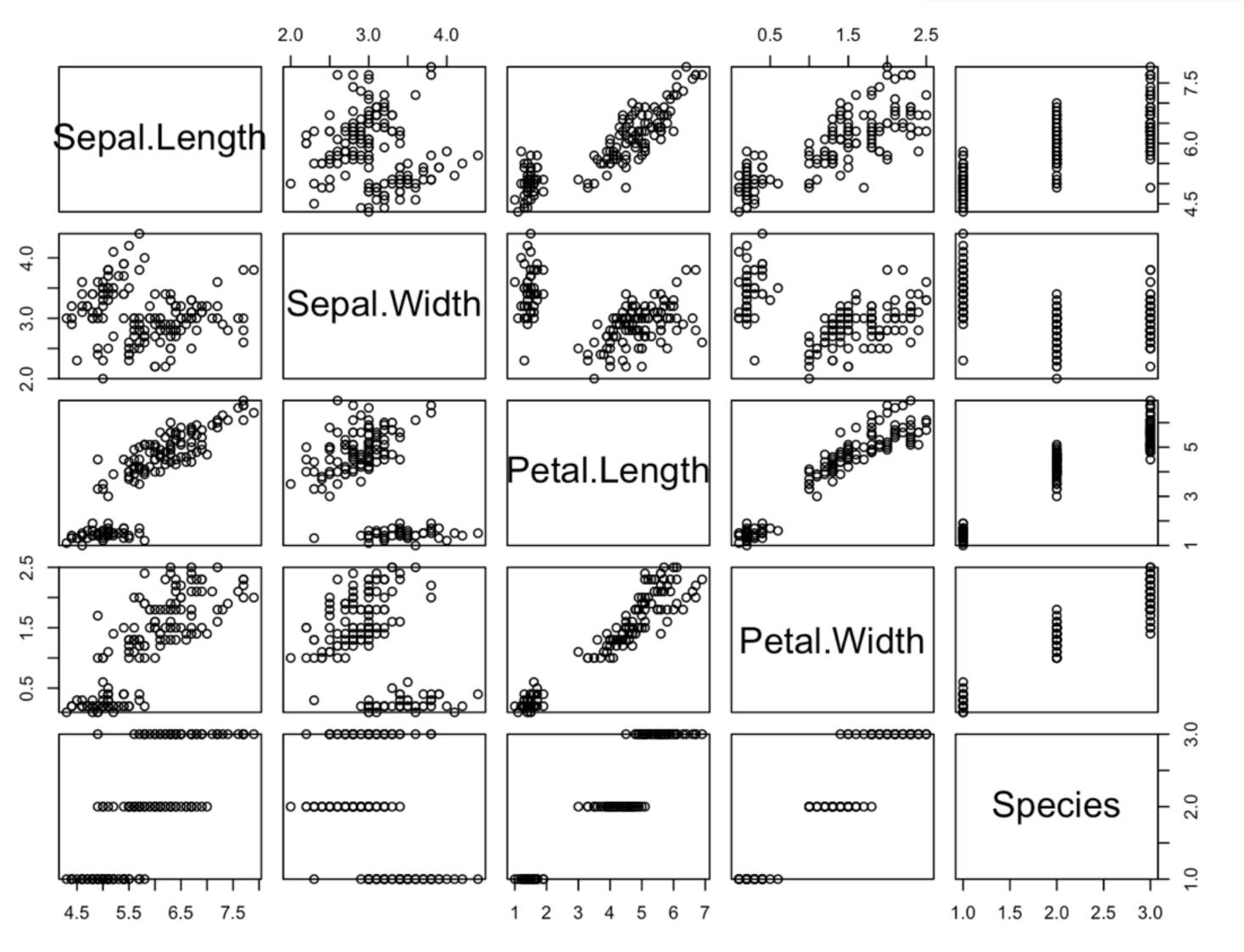


plot(iris)



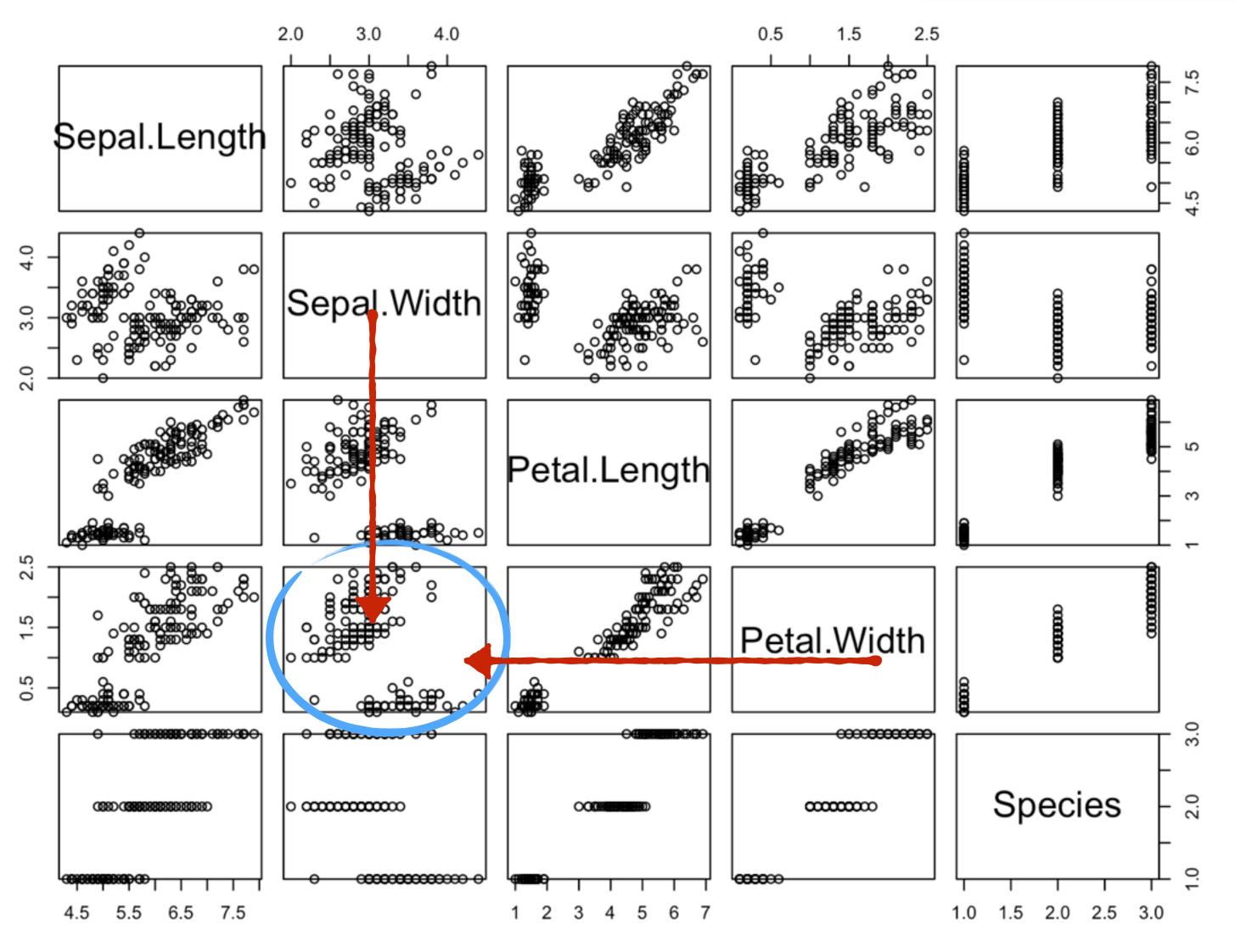
LET'S PARSE THIS GRAPH

plot(iris)



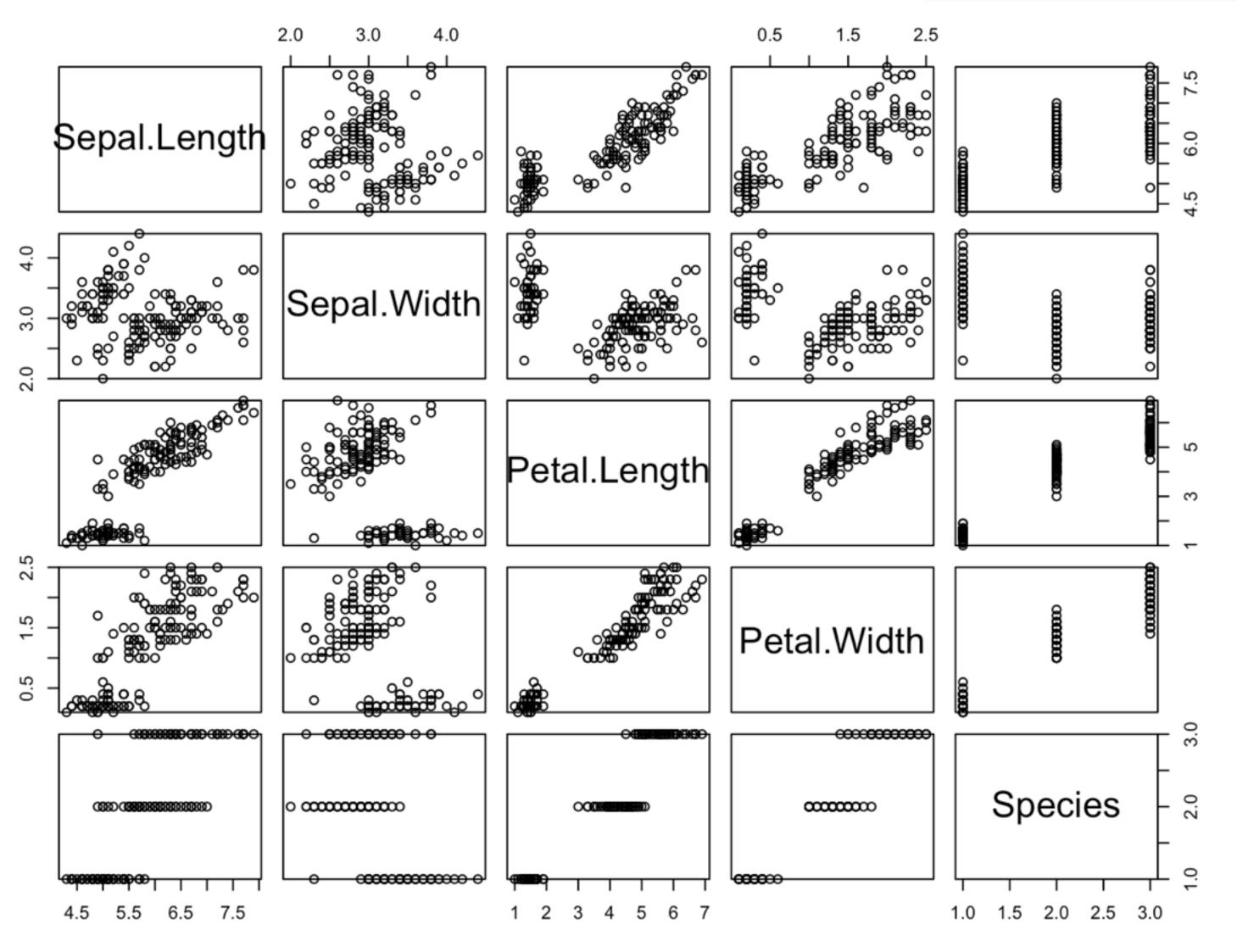
EACH GRAPH IN THIS MATRIX IS A SCATTER PLOT FOR 2 VARIABLES

plot(iris)



THIS IS A SCATTER PLOT OF PETAL.WIDTH(Y) VS SEPAL.WIDTH(X)

plot(iris)



WE ARE ABLE TO SEE ALL THE RELATIONSHIPS BETWEEN ALL PAIRS OF VARIABLES AT ONCE!

GGPLOT2 IS A PACKAGE FOR DATA VISUALIZATION IN R

IT CAN BE USED TO BUILD COMPLEX 2D GRAPHICS

WE HAVE SOME DATA OF STOCK MOVEMENTS FOR LAST 5 YEARS

LET'S PLOT ALL THE STOCK MOVEMENTS USING GGPLOT2

WE HAVE FILES WITH STOCK MOVEMENTS OF GOOGLE, NASDAQ, EXXON MOBIL AND S&P 500

```
googFile <- '/Users/swethakolalapudi/Desktop/Regression/goog.csv'
nasdaqFile <-'/Users/swethakolalapudi/Desktop/Regression/nasdaq.csv'
xomFile <- '/Users/swethakolalapudi/Desktop/Regression/xom.csv'
snpFile <-'/Users/swethakolalapudi/Desktop/Regression/snp.csv'</pre>
```

LET'S READ THESE INTO DATA FRAMES

```
goog <-read.table(googFile,header = TRUE, sep ="," [,c("Date","Adj.Close")]
goog["Series"] <- "Goog"</pre>
```

WE'LL READ ONLY 2 COLUMNS, THE DATE AND ADJUSTED CLOSING PRICE

```
goog <-read.table(googFile,header = TRUE, sep =",")[,c("Date","Adj.Close")]
goog["Series"] <- "Goog"</pre>
```

WE ADD AN ADDITIONAL COLUMN TO THE DATA FRAME TO SPECIFY THE SECURITY NAME

```
goog <-read.table(googFile, header = TRUE, sep =",")[,c("Date", "Adj.Close")]
goog["Series"] <- "Goog"

nasdaq <-read.table(nasdaqFile, header = TRUE, sep =",")[,c("Date", "Adj.Close
nasdaq["Series"] <- "Nasdaq"

xom <-read.table(xomFile, header = TRUE, sep =",")[,c("Date", "Adj.Close")]
xom["Series"] <- "Exxon Mobil"

snp <-read.table(snpFile, header = TRUE, sep =",")[,c("Date", "Adj.Close")]
snp["Series"] <- "S&P 500"</pre>
```

PO THIS FOR ALL THE FILES

EXAMPLE 6: PLOT MULTIPLE LINES USING GGPLOT2 AT THE END WE HAVE 4 DATA FRAMES

Date [‡]	Adj.Close [‡]	Series [‡]
2016-02-01	752.0000	Goog
2016-01-04	742.9500	Goog
2015-12-01	758.8800	Goog
2015-11-02	A742.000	Goog
2015-10-01	710.8100	Goog
2015-09-01	608.4200	Goog
2015-08-03	618.2500	Goog
2015-07-01	625.6100	Goog
2015-06-01	520.5100	Goog
2015-05-01	532.1100	Goog

Date [‡]	Adj.Close [‡]	Series [‡]
2016-02-01	4620.37	Nasdaq
2016-01-04	4613.95	Nasdaq
2015-12-01	5007.41	Nasdaq
2015-11 02		1 sdaq
2015-10-01	5053.75	Nasdaq
2015-09-01	4620.16	Nasdaq
2015-08-03	4776.51	Nasdaq
2015-07-01	5128.28	Nasdaq
2015-06-01	4986.87	Nasdaq
2015-05-01	5070.03	Nasdaq

Date [‡]	Adj.Close [‡]	Series [‡]
2016-02-01	75.60381	Exxon Mobil
2016-01-04	77.14977	Exxon Mobil
2015-12-01	77.24887	Exxon Mobil
2015-11-02	8) 2/ 1	Exxon Mobil
2015-10-01	81.28717	Exxon Mobil
2015-09-01	73.04449	Exxon Mobil
2015-08-03	73.91886	Exxon Mobil
2015-07-01	77.09778	Exxon Mobil
2015-06-01	80.98138	Exxon Mobil
2015-05-01	82.92805	Exxon Mobil

Date [‡]	Adj.Close [‡]	Series [‡]
2016-02-01	1939.38	S&P 500
2016-01-04	1940.24	S&P 500
2015-12-01	2043.94	S&P 500
2015-11-02	2)80.41	S&P 500
2015-10-01	2079.36	S&P 500
2015-09-01	1920.03	S&P 500
2015-08-03	1972.18	S&P 500
2015-07-01	2103.84	S&P 500
2015-06-01	2063.11	S&P 500
2015-05-01	2107.39	S&P 500

RBIND() CAN COMBINE THESE INTO A SINGLE DATA FRAME

RBIND() CAN COMBINE THESE INTO A SINGLE DATA FRAME

```
data <- rbind(goog,nasdaq,xom,snp)
data$Date <- as.Date(data$Date)</pre>
```

CONVERT THE DATE COLUMN TO DATE (FROM STRING)

WE ARE NOW REAPY TO PLOT THIS DATA require (ggplot2)

GGPLOT2 IS A DATA VISUALIZATION PACKAGE IT USES LAYERS TO PLOT THE DATA

EXAMPLE 6: PLOT MULTIPLE LINES USING GGPLOT2 require (ggplot2)

EACH LAYER IS APPED ON TOP OF THE OTHER

ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

THE FIRST LAYER IS AN EMPTY PLOT WHICH IS AWARE OF THE DATA FRAME IT WILL READ THE DATA FROM

EXAMPLE 6: PLOT MULTIPLE LINES USING GGPLOT2 require (ggplot2)

EACH LAYER IS APPED ON TOP OF THE OTHER

```
ggplot(data) aes(Date,Adj.Close)) + geom_line(aes(colour = Series))
```

THE FIRST LAYER IS AN EMPTY PLOT WHICH IS AWARE OF THE DATA FRAME IT WILL READ THE DATA FROM

ggplot(data aes(Date,Adj.Close) + geom_line(aes(colour = Series))

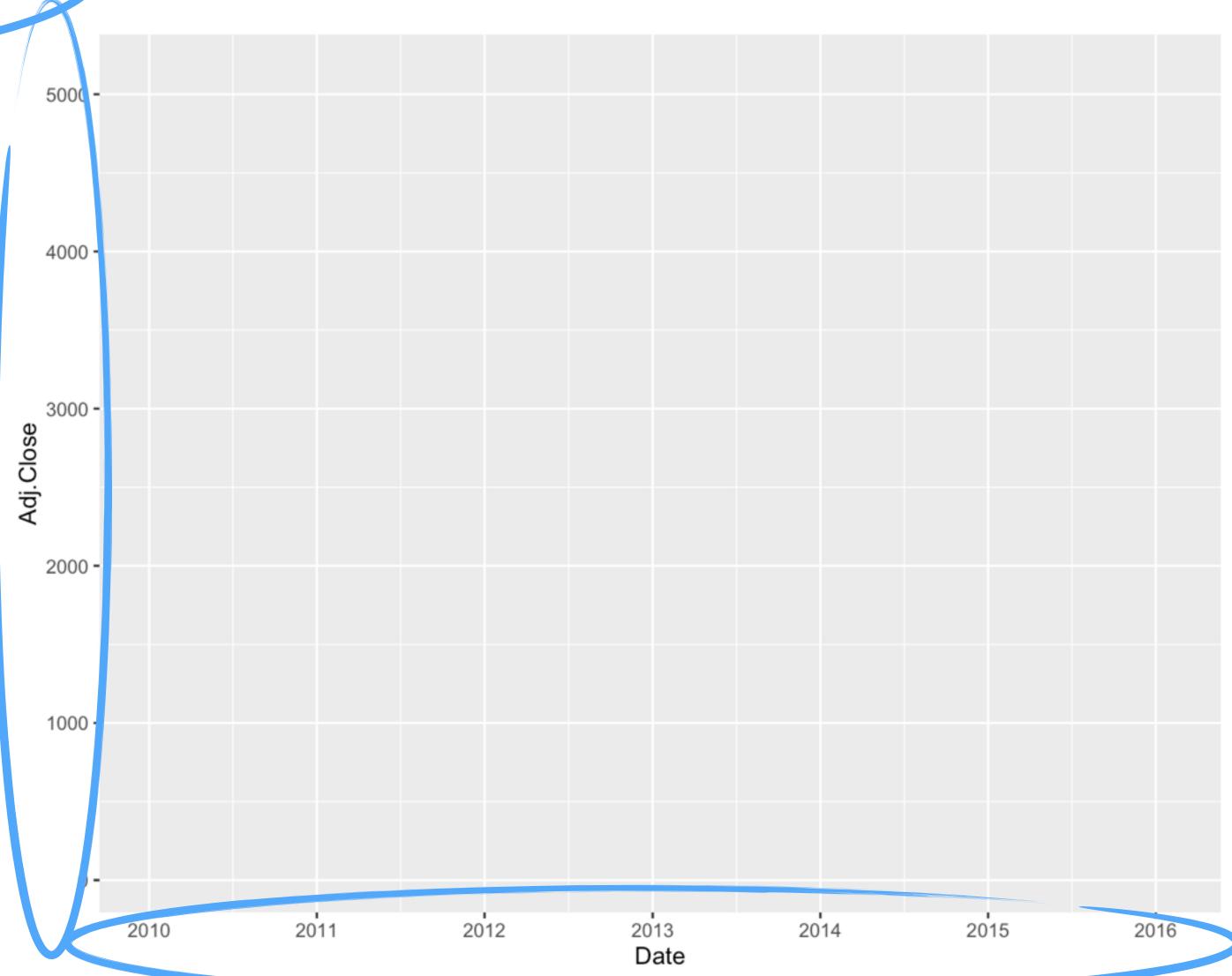
THE DATA TO BE USED FOR X-LABELS, Y-LABELS, ARE SPECIFIED USING A FUNCTION CALLED AES()

AES STANDS FOR AESTHETICS

IN EACH LAYER, YOU CAN SPECIFY THAT THE AESTHETICS NEED TO BE READ FROM SOME DATA FRAME COLUMN

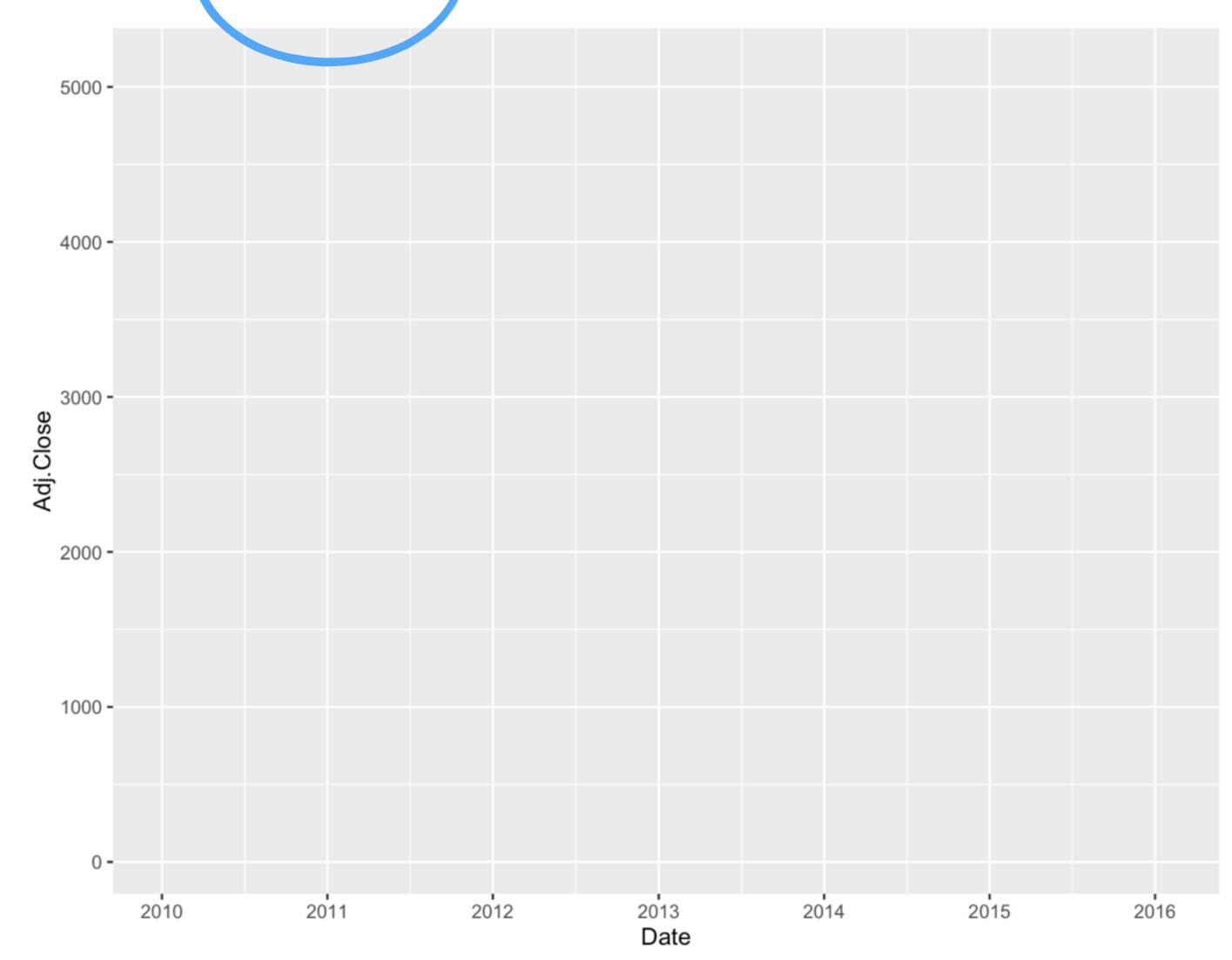
ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

IF YOU JUST USE THIS, AN EMPTY PLOT WITH X AND Y-AXIS WILL SHOW UP



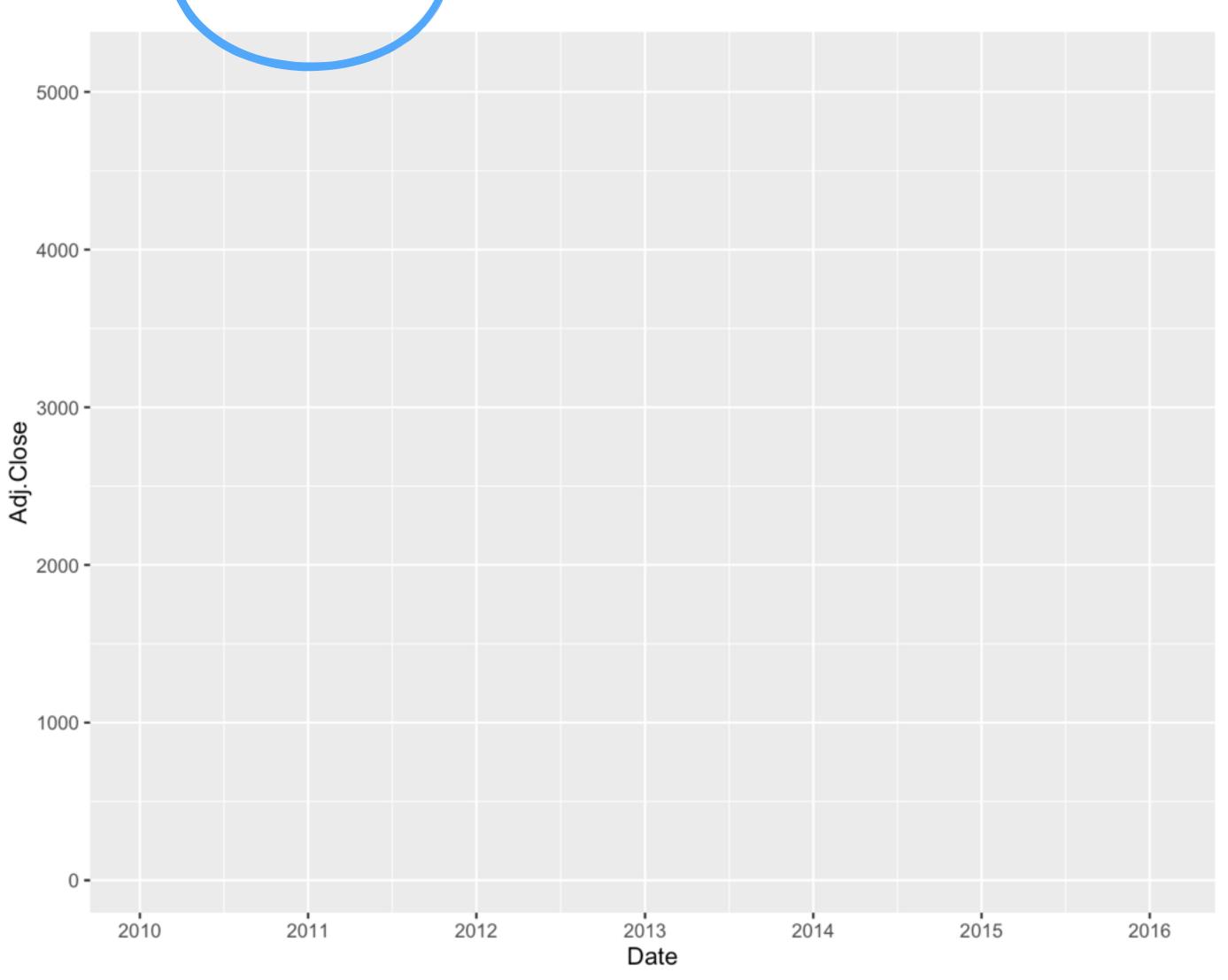
ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

NOW WE APP A SECOND LAYER WITH THE LINES TO BE PLOTTED



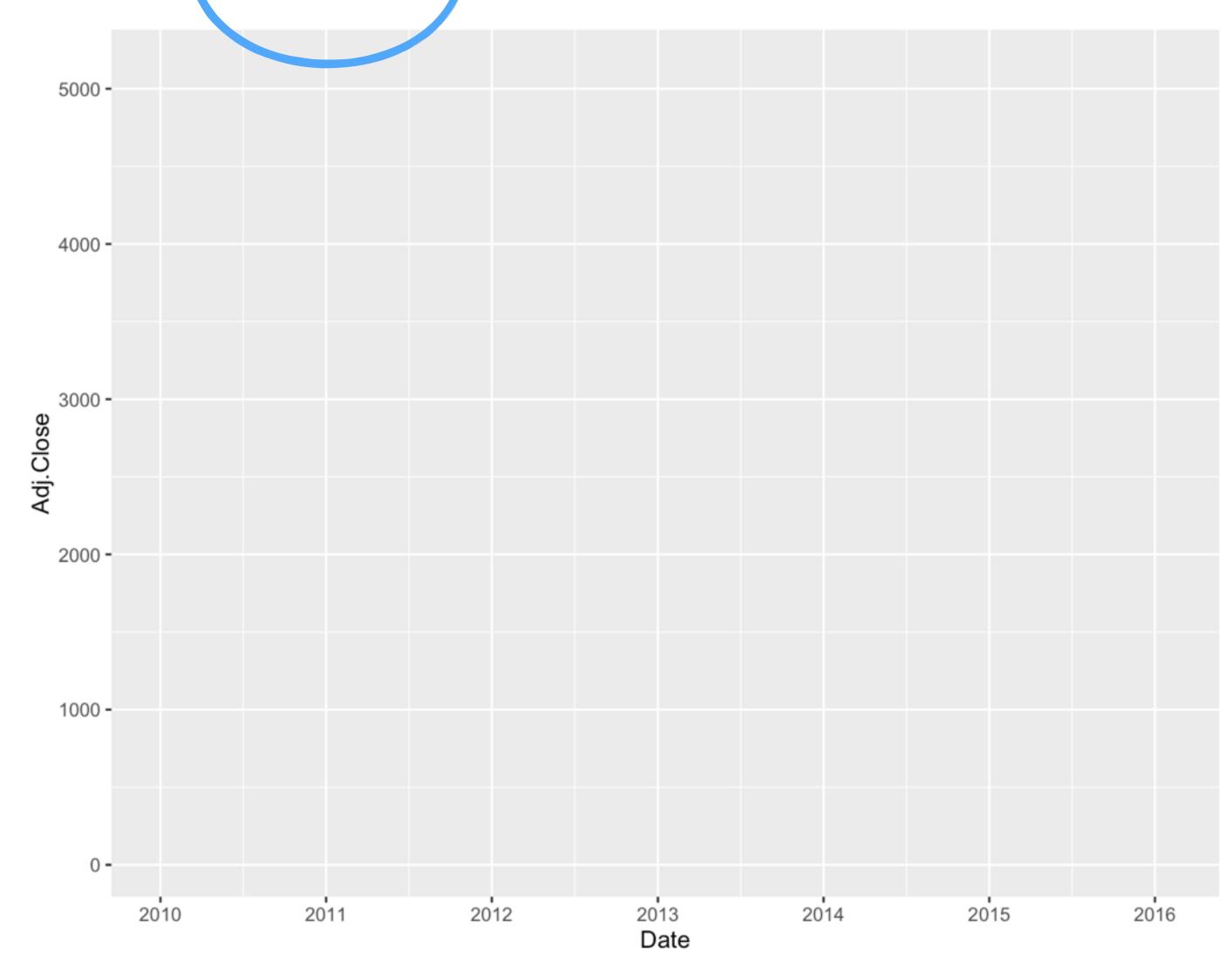
ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

GEOM_LINE() SPECIFIES THAT WE WANT A LINE GRAPH



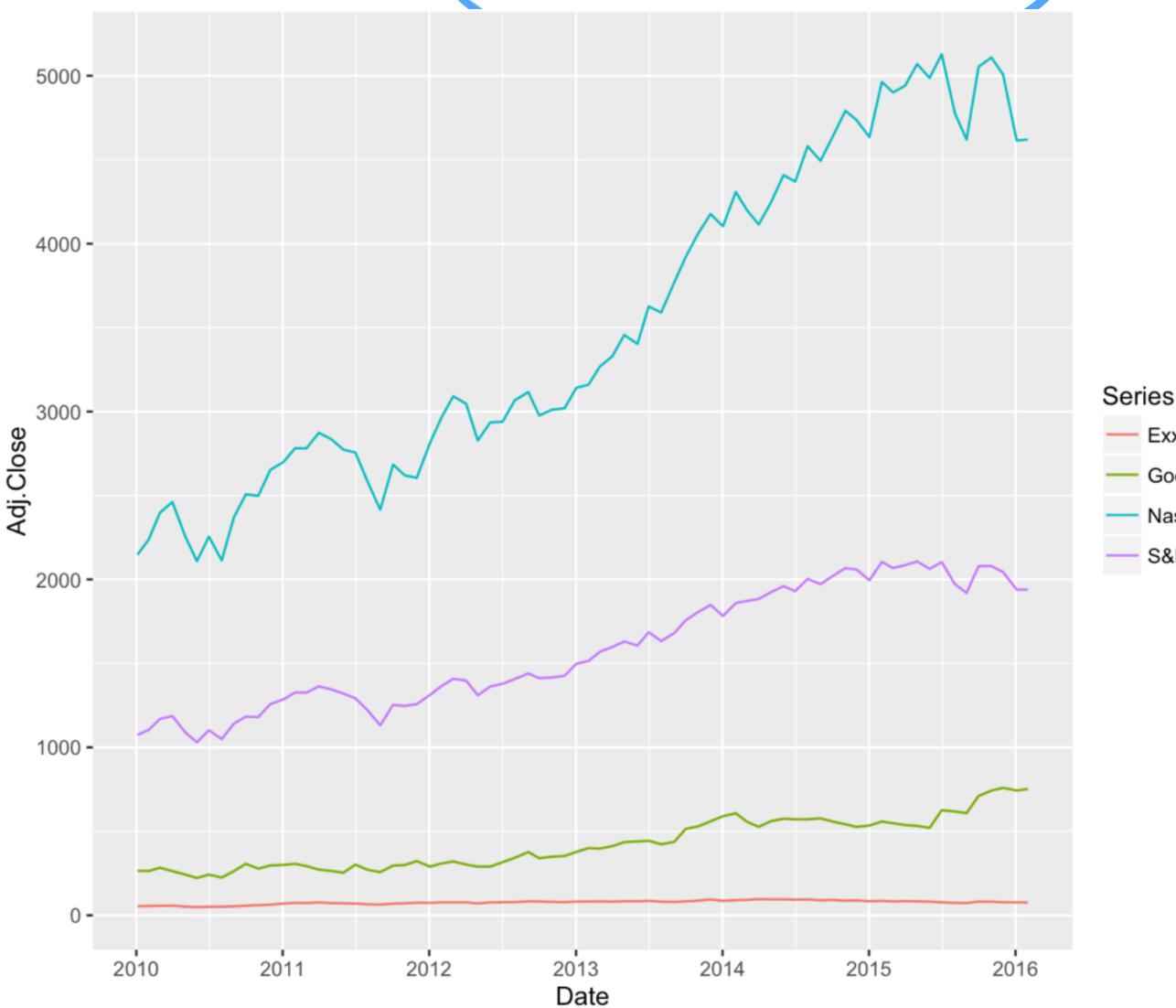
ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

THERE ARE DIFFERENT
FUNCTIONS FOR
DIFFERENT PLOTS
GEOM_HIST(),
GEOM_AREA(),
GEOM_BAR() ETC



ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

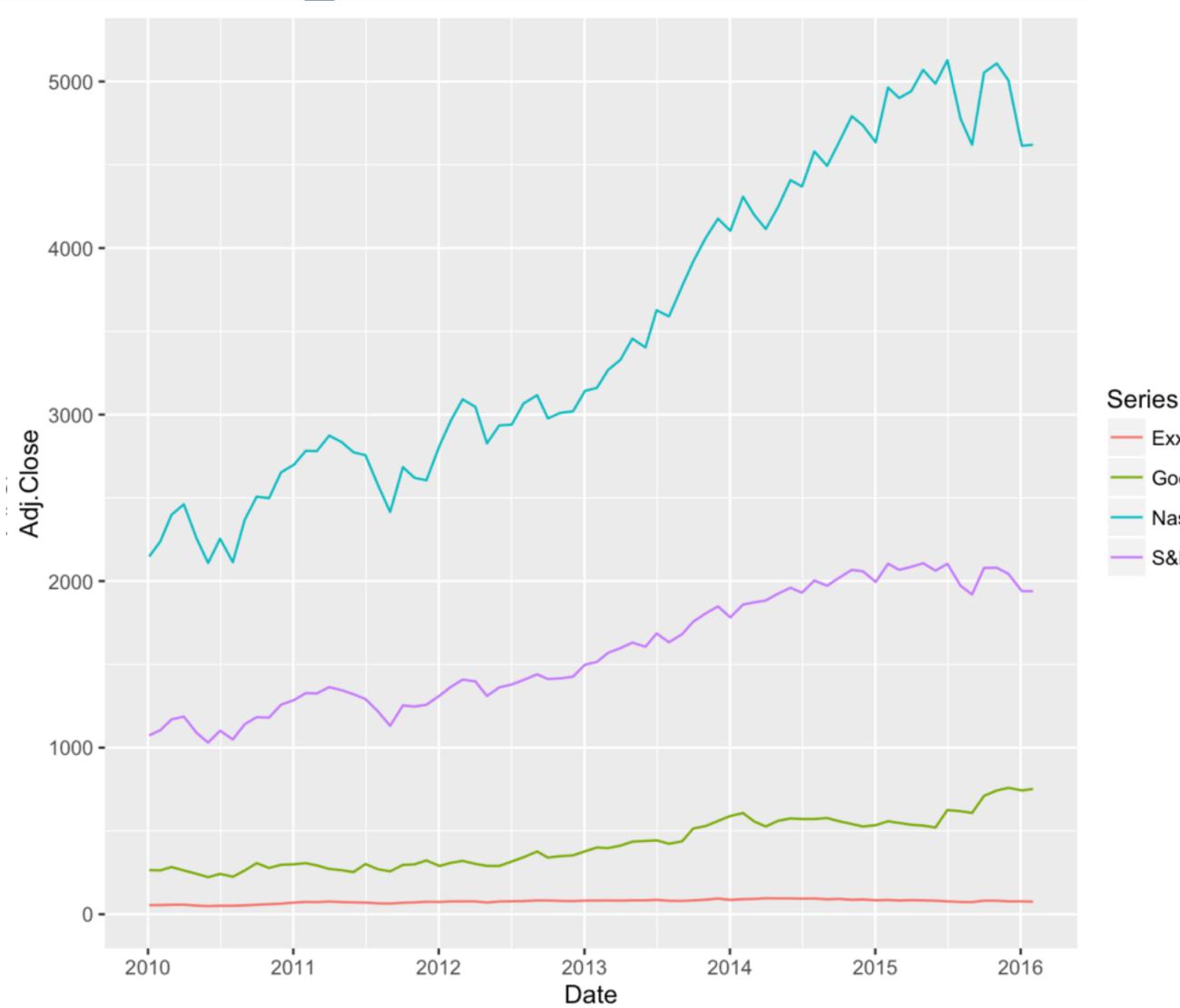
HERE WE SPECIFY
ANOTHER AESTHETIC,
THE LINE COLOR
SHOULD BE BASED ON
THE SERIES COLUMN
OF THE DATA FRAME



Exxon Mob

ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

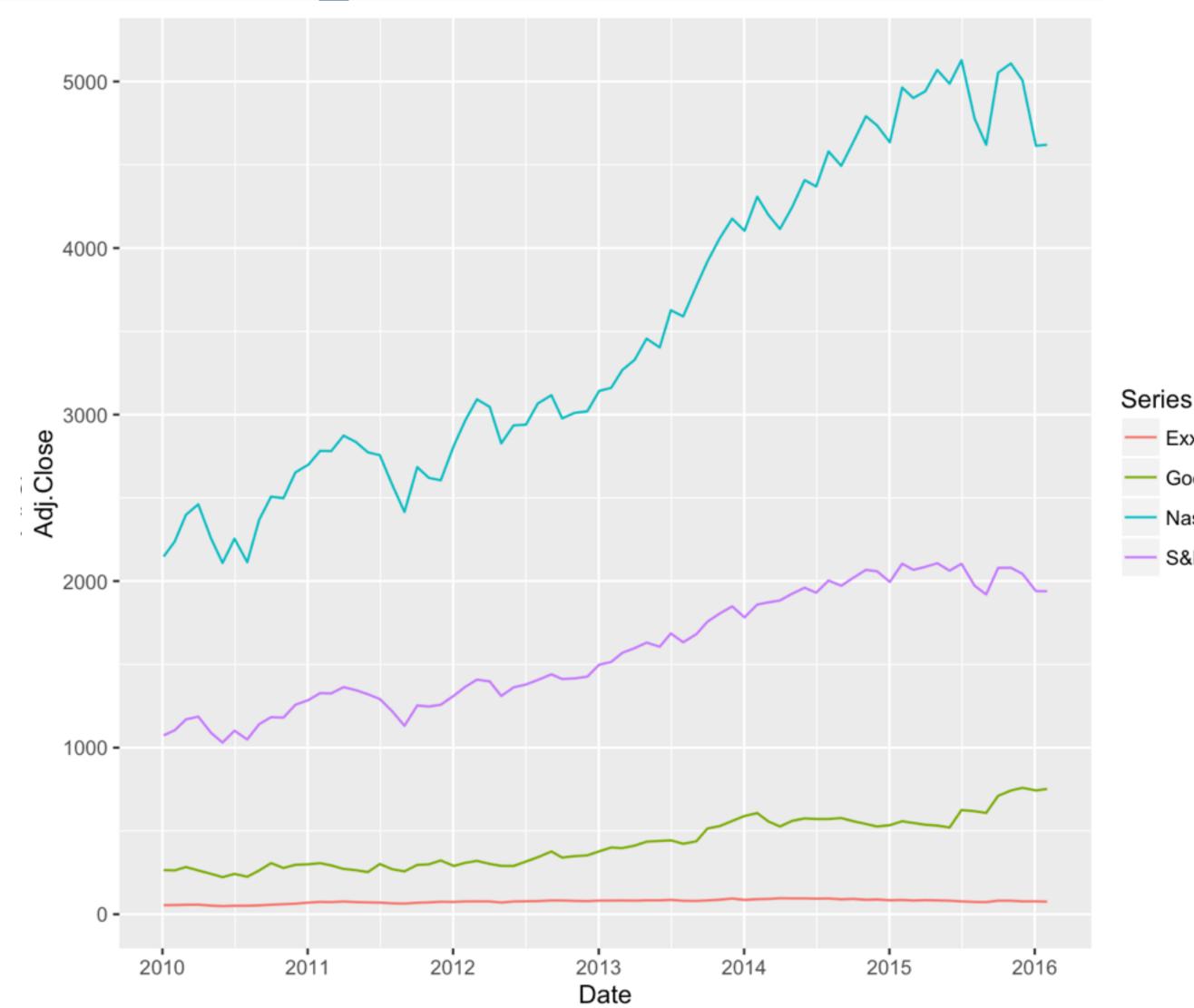
GGPLOT2 IS VERY POWERFUL AS IT GIVES YOU A LOT OF CONTROL OVER THE LAYERS



Exxon Mob

ggplot(data, aes(Date,Adj.Close)) + geom_line(aes(colour = Series))

YOU CAN DRAW A
COMBINATION LINEBAR CHART, LINEAREA CHART OR ANY
COMBINATION THAT
YOU LIKE



Exxon Mob

S&P 500

SO, HOW PO YOU GO FROM PATA TO PECISIONS?

- 1. USE NUMBERS AND PLOTS TO DESCRIBE THE DATA
- 2. DRAW INFERENCES FROM THE DATA
 - 3. ORGANIZE DATA USING ANALYTICAL TOOLS
 - 4. QUANTIFY RELATIONSHIPS BETWEEN VARIABLES

5. VISUALLY COMMUNICATE WHAT YOU'VE LEARNT

SO, HOW PO YOU GO FROM PATA TO PECISIONS?

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