Intro to R Programming

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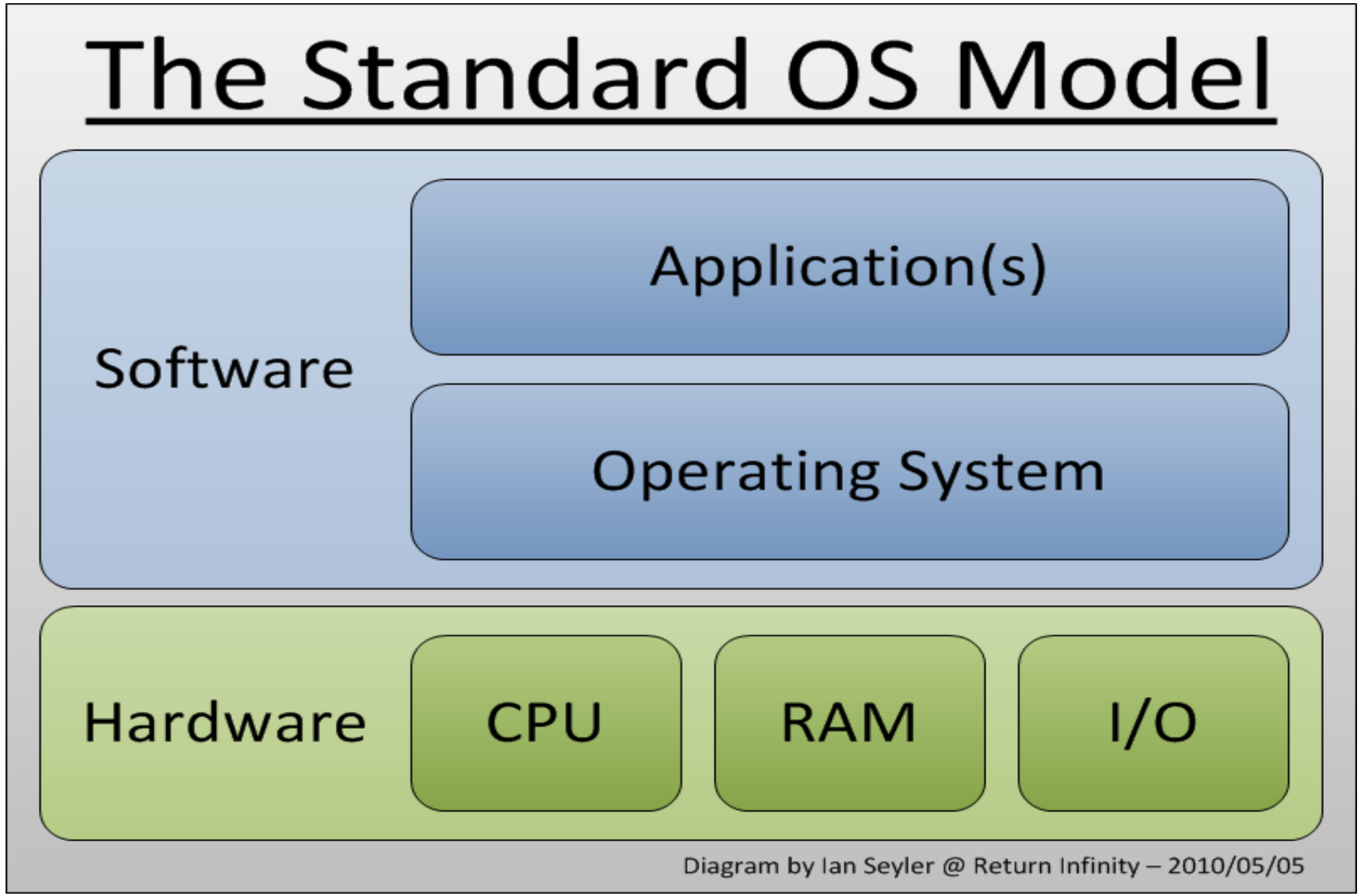
What this course is about?

- Basics of Computer Architecture and Programming
- Intro to Programming through R
- Popular R methods and their use in Data analysis
- Technical Documentation: Some tips for Word, Latex and R-markdown.

What's a computer look like?



What does it do?

- Perform Calculations!
  - Billions of them every second.
  - Cores, threads, clock speed
- Stores data
  - Cache vs RAM vs HDD
  - Speed vs storage cost
- Runs Software
  - System (OS): Linux, Windows and Mac-OS
  - Application: R, RStudio, Excel

What is a program?

- Translation of an algorithm into a language that computer understands
- An algorithm takes input, perform some operations and gives output
  - Executes in finite time
  - E.g. sorting, searching, reading, copying!

- Complexity of a Program
  - Time and space!
  - E.g. Fibonacci series!

- Programming Paradigms
  - Iterative vs Recursive
  - Procedural vs Object Oriented

- Good Program
  - Re-readable, organized and modular

Typical Programing Errors

- Syntactical (spelling mistake)
  - Will get caught very easily! Just run the program.

- Semantic Errors (meaningless operations)
  - For e.g. “nikhil”+32
  - Exceptions: like divide by 0.
  - May get caught. A warning will be thrown nonetheless.

- Logical Errors (Unintentional)
  - Program will crash, run forever or give a wrong answer!
  - Debugging requires some skill and experience.

What is R

- Implementation of S Programming language
  - Started as statistical environment
  - Explains the deep rootedness of R in statistics
  - Mostly written in C (earlier FORTRAN)
  - More info on Wikipedia!

- Philosophy behind R (or S, S+)
  - Interactive environment
  - Transition from users to Programmers as per need!
  - You don’t need to be a programmer to learn (and) use basic R
  - More info at http://ect.bell-labs.com/sl/S/history.html

What is R (cont.)

- Features
  - Very easy to follow and understand
    - Require understanding of vector and matrix indexing!
    - Interactive
  - Runs on all platforms.
    - Small software to download and load. Use packages as per need.
  - Free of cost. Open source software (GNU GPL). More info at www.fsf.org
  - Very active development
    - Frequent updates and releases
    - Very active and responsive user community – Stackoverflow!

- Drawbacks
  - Limited 3-D graphics capability
  - Everything must be in RAM – big data?
  - If a functionality is missing you got to code it yourself!

What if not R

- Closest cousin is MATLAB
  - Although used much more in engineering than in statistics
  - Syntax is similar to R (Read: http://mathesaurus.sourceforge.net/octave-r.html)
  - Python is also very popular although its more meaningful for data science

- Statistical Alternatives?
  - SAS and Stata
  - Both are paid software
  - Very different than R in syntax!
    - Non-interactive
    - Limited user community support
  - Despite the differences Stata is very popular in management research. And there are some die-hard SAS fans in Finance too.

Downloading and Installing R

- Download R: https://cran.r-project.org/
  - Choose base package for your OS
    - Windows: https://cran.r-project.org/bin/windows/base/R-3.5.0-win.exe
    - Linux: Use apt-get (Debian based) OR yum install (RPM based) from terminal.
    - Mac: https://cran.r-project.org/bin/macosx/R-3.5.0.pkg
  - Install R

- Download RStudio IDE
  - Choose the free RStudio Desktop edition
  - https://www.rstudio.com/products/rstudio/download/#download
  - Choose the appropriate one according to your OS
  - Install RStudio

Getting Help in R

- From Console
  - Just type: ? followed by function name without parenthesis
  - E.g. `?mean; ?sum; ?length;`
  - Clarify:
    - **?mean** - help for the function “mean”
    - **??mean** - will perform the search over the internet (CRAN database)
      - Look for base::mean!
    - **mean()** - call the function mean
    - **mean** - print the definition of the function “mean”

- From Web sources
  - Most reliable and easy to incorporate is www.stackoverflow.com.
  - www.r-bloggers.com is also quite helpful.
  - You can use https://cran.r-project.org for any resource on R
  - Even typing your question in google will get you good results!
    - 99% of your questions are already answered! You just need to find them!

R Input and Output

- Simple assignment
  - `X = 1;` (*or* `X <- 1;`)
  - Assignment is always right to left
    - Read 1 goes into X
    - We aren't comparing X with 1 here
  - The semi-colon isn't necessary in R, but it's a good practice to use it
  - `X = ;` is incomplete
  - # (prefix) is used as a comment. Use it for helpful comments.
  - Use Ctrl-Shift-C for multi-line comments

- Value of X can be seen by
  - `X;`

Vectors

- A sequence of numbers. Many ways to input!
  - `Y = c(1,7,-3,41); # concatenate arbitrary numbers`
  - `Y = 1:10; # natural numbers`
  - `Y = seq(1,100,9); # skip by 9`
  - `Y = rep(2, 3); # repeat 3 times`
  - `Y = rep(1:2, 3); # repeat the vector`
  - `Y = rep(1:2, each = 3); # repeat each element 3 times`
  - `Y = c(); # empty vector`
  - Execute this: `c(1:3, rep(c(5,7), each = 2), rep(9, 4), 7);`

- Length of vector: `length(Y);`

- Accessing i^{th} element of vector: `Y[i]; # square brackets`
  - `i` should be between `1` and `length(Y)`
  - Printing the entire vector is as before: `Y;`

Objects in R

- 5 basic (atomic) types of objects
  - character – strings
  - numeric – real numbers. Also called double.
  - integer – natural numbers. Default data type for numeric vectors.
    - `typeof(1:10)`
  - complex – complex numbers. We won't use them now!
  - logical – True/False (binary)

- Most basic collection of objects is a vector (also called an array)
  - Can only contain objects of same class (i.e. character or integer; not both)
  - "list" is a special type of object and can contain heterogeneous objects
    - Any Combination of vector, matrix, atomic types etc.
    - It can even contain another list as an object. E.g. linked-lists!
    - Due to its generality its very slow and hence rarely used with large datasets unless situation demands it

Numbers

- Default type of any number is numeric (i.e. real). `typeof(1)`
- R can differentiate between corner cases:
  - `1/0` is `Inf` -- `is.infinite();`
  - `0/0` is `NaN` -- `is.nan();`
  - Missing data is `NA` -- `is.na();`
  - Check what's `Inf-Inf` ?

- Arithmetic Operations
  - `*` multiplies
  - `/` divides
  - `^` takes exponent
  - `%%` is the modulo (remainder) operator. Try: `7 %% 2;`

Coercion

- Mixing Objects
  - Automatically coerced to the same class.
  - Try: `c(1:7, "a"); c(T, 2); c("a", FALSE);`
  - Implicit coercion!
  - Never use unless you know what you're doing!

- Explicit Coercion
  - `as.character(1:5);`
  - `as.numeric("iimb"); # warning!`
  - `as.logical(seq(-2,2,1));`

List

- Can carry different types of data together
  - `L = list(1, FALSE, 3.14, "iimb", "c", 4-3i);`
  - Print list: `L;`
  - L is in fact a list of lists. Check: `typeof(L); typeof(L[4]); typeof(L[[4]]);`
  - Single square brackets `[i]` access the i^{th} list embedded in the list `L`
  - Double square brackets `[[i]]` access the i^{th} element
  - Can append elements in list: `L = append(L, "7th");`
  - `unlist(L);` will coerce all elements into a single type and return a vector
  - Delete an element from a list:
    - I don't know how to do that!
    - Let's google: "delete element from list in R"
    - Open the answer on www.stackoverflow.com

Matrices

- Generalization of vectors
  - 2 dimensions instead on one!
  - N x K matrix means a matrix having N rows and K columns. Total of NK elements.
  - `M = matrix(nrow = 2, ncol = 3);`
  - Dimensions: `dim(M);`
  - Can think of M as
    - 3 columns vectors each of length 2, *or*
    - 2 row vectors each of length 3
  - Populate matrix: `M = rbind(1:3, 4:6);`
  - Alternatively populate as: `M = cbind(1:2, 3:4, 5:6);`

Matrices (cont.)

- Indexing a matrix
  - `M[i,j]` gives the element at i^{th} row and j^{th} column
  - `M[i,]` gives the entire i^{th} row (a vector)
  - `M[,j]` gives the entire j^{th} column (a vector)

- Matrix multiplication
  - `*` just does an element wise multiplication, i.e. $(M * M)_{ij} = M_{ij} * M_{ij}$
  - `%*%` performs the usual matrix multiplication. Try: `M %*% M`
    - Dimensions must match
    - Try `t(M) %*% M;`
    - `t(M)` takes transpose of a matrix!

Matrices (cont.)

- Identity matrix: `diag(3)`
- Diagonal Matrix: `diag(c(1,5,7)); diag(1:7);`
- Diagonal of a matrix: `diag(M)`
- Trace of a matrix: `sum(diag(M))`
- Inverse of a matrix:
  - Must be a square matrix: `M = matrix(1:9, nrow = 3, ncol = 3);`
    - Another way to create a matrix. Data is entered column-wise.
  - Determinant must be non-zero: `det(M); M[3,3] = 19; det(M);`
  - Inverse: `solve(M);`

Factors

- For categorical data.
  - Male, female
  - Cities in a dataset
  - Typically useful when the dataset is large but the no. of categories is small
  - Very useful in the regression framework using `lm();`
    - Automatically creates dummy for all but one categories.
  - Using factors is more descriptive than integer values
    - Rather than using 1 for PGP, 2 for FPM and 3 for Others; its more intuitive to use factors.
  - Example:
    - `sex = rep(c("male", "female"), 5);`
    - `sex_f = as.factor(sex);`
    - `Check: typeof(sex_f); as.integer(sex_f);`

Data Frame

- Probably the most important data type you'll use.
  - All external data (from excel, csv, tables, webpages etc) is read as data frame
  - It's a list where each element of list must have the same length.
  - Think of it like a matrix but with the flexibility that each column can have different data type. E.g. set of Names, weights and heights
  - Example:
    - `d = data.frame(name = c("a", "b"), weight = c(70, 75), height = c(1.78, 1.82));`
    - `d; d$name; d[1,]; d$weight; d[,3];`
    - `d$bmi = d$weight / (d$height)^2;`
    - `nrow(d); ncol(d); dim(d);`
    - `colnames(d)[1] = "names";`
    - `rownames(d) = c("I", "II");`

Reading Data

- Download some stock data from NSE
  - https://www.nseindia.com/products/content/equities/indices/historical_index_data.htm
  - Save the CSV file as data.csv
- From CSV (most common)
  - `setwd("D:/Opera Downloads/"); nifty = read.csv("data.csv");`
  - Alternatively: `nifty = read.csv("D:/Opera Downloads/data.csv");`
- From Excel
  - Search it yourself! It is not recommended btw.
- From clipboard
  - `read.table("clipboard");`
  - This is quick fix for small data transfer between R and excel. Use `read.csv()` as your primary method for data reading!

Reading Data (cont.)

- Viewing data
  - `View(nifty);`
- Date
  - `nifty$Date = as.Date(nifty$Date, format = "%d-%b-%Y");`
  - `n = nrow(nifty);`
  - `d = nifty$Date[1];`
  - `format(d, format = "%D");  # 04/02/18`
  - `format(d, format = "%d-%m-%y");  # 02-04-18`
  - `format(d, format = "%d.%b.%Y");  # 02.Apr.2018`
  - `format(d, format = "%d_%B_%Y");  # 02_April_2018`
- Alternatively,
  - `read.table("data.csv", header = T, sep = ",", nrows = 5);`

if-else

-
```
if(<COND_1>) {
  # do something!
}
```

-
```
if(<COND_1>) {
  # do something!
} else {
  # ...
}
```

-
```
if(<COND_1>) {
  # do something!
} else if(<COND_2>) {
  # ...
} else {
  # ...
}
```

```
if(nifty$Close[2] > nifty$Close[1]) {
  str = paste("Stock market closed green on", nifty$Date[2]);
} else if(nifty$Close[2] > nifty$Open[2]) {
  str = paste("Stock market closed above opening on", nifty$Date[2]);
} else {
  str = paste("Stock market was red and closed below opening on", nifty$Date[2]);
}
print(str);
```

for loop

- Looping is used to perform similar set of tasks repetitively
  - ```
    for(i in n:1) {
      print(nifty$Date[i]);
    }
    ```
  - `n:1;` is same as `seq(n,1,1);` i.e. backwards counting!
  - Alternatively, you can execute: `rev(nifty$Date);` or `nifty$Date[n:1];`
- Try avoiding loops if you can!
  - Increasing all dates by a week: `nifty$Date + 7`
  - Finding Daily growth: `nifty$Close[-1] / nifty$Close[-n]`
  - Daily diff. b/w high and low prices: `nifty$High - nifty$Low`
  - Question: find % growth in daily volatility
    - Volatility is defined as: $Vol_t = (High_t - Low_t)/Open_t$
    - Percentage Growth is defined as: $\%G = \frac{(Value_{t+1} - Value_t)}{Value_t} * 100$
    ```

Nested if-else and for loop

```
for(i in 2:n) {
  if(nifty$Close[i] > 1.01 * nifty$Close[i-1]) {
    # market gained more than 1%
    for(j in 1:ncol(nifty)) {
      print( paste("Gain", i, colnames(nifty)[j], nifty[i,j], sep =":") );
    } # end for(j)
  } else if(nifty$Close[i] < 0.99 * nifty$Close[i-1]) {
    # market lost more than 1%
    for(j in 1:ncol(nifty)) {
      print( paste("Loss", i, colnames(nifty)[j], nifty[i,j], sep =":") );
    }
  } else {
    print(paste("Market movement was within 1% for i =", i));
  } # end if()
} # end for(i)
```

Jumping

- Till now all our commands executed sequentially
- There may be circumstances when we need to jump
- Next and Break
  - `next` is used to skip an iteration, while `break` exits the loop entirely.
  - ```
    for(i in 1:10) {
      if(i <= 3) {
        next;
      }
      if(i > 6) {
        break;
      }
      print(i);
    }
    i;
    ```
- `return()` is used to exit a function with a value.
    ```

Function

- Organize often-used set of instructions separately in a “function”
- Calling a function will execute all the commands in the body of function
- We have used many functions till now
  - They end with parenthesis: `()`
  - E.g. `sum(); rbind(); vector(); format(); read.csv();` etc
  - Note that curly braces `{}` are used for if-else and for loops, square braces `[]` for vector/matrix indexing and parenthesis `()` for grouping, if-else & for conditions and functions.
- A function has
  - A name by which we call them, e.g. `sum`
  - A set of inputs to be put within parenthesis like numbers `1:10` in `sum()`
  - Return value which is the output of the function like the sum of numbers in `sum()`

Function Example

```
my_mean = function(x) {
  n = length(x);
  mean = sum(x) / n;
  return(mean);
}
```

- Name of the function is: `my_mean`
- Input is: `x`
- Output is: `mean`
  - Note that the mean here is just a name, we could well have used any other name without changing anything about our function

Function (Example) Cont.

- Alternate ways to write the same function
  - ```
    my_mean = function(x) {
      return( sum(x) / length(x) );
    }
    ```
    - No need to store sum and length. We can directly divide them!
  - ```
    my_mean = function(x) {
      sum(x) / length(x);
    }
    ```
    - No need for an explicit return. The last statement is returned by default.

- Try various value with `my_mean()` and the inbuilt `mean()`. See that the answers are exactly the same.

- Write a function for variance where $Var(x) = mean([x - mean(x)]^2)$
  - Compare it with the inbuilt `var()` function?

Multiple conditions & which() function

- The arguments to `if()` and `which()` and the output of `is.xx()` family of functions is a logical object, i.e. either `TRUE` or `FALSE`.

- A valid combination of logical objects is also a logical object. E.g.
  - Logical AND: `TRUE & FALSE` is `FALSE`
  - Logical OR: `TRUE | FALSE` is `TRUE`
  - Logical NOT: `! FALSE` is `TRUE`

- De Morgan's Law
  - $!(A \mid B) = (!A) \,\&\, (!B)$
  - $!(A \,\&\, B) = (!A) \mid (!B)$

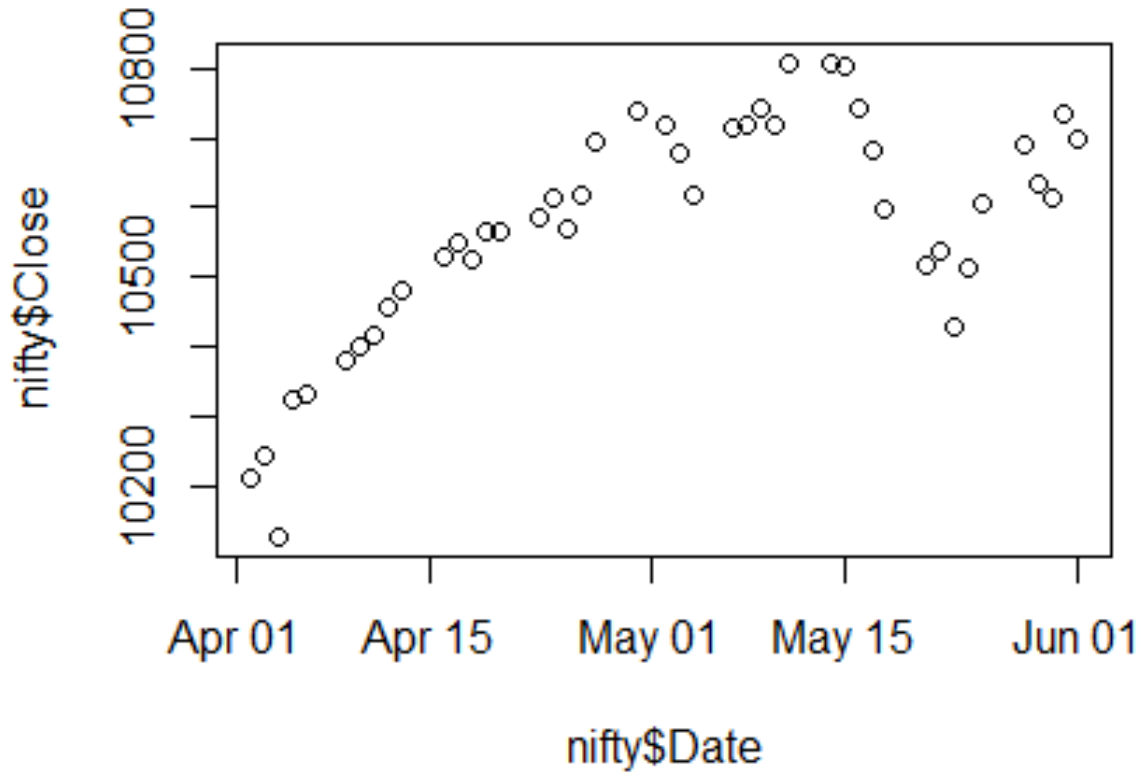
- The below two indexes are one and same (by De Morgan Law),
  - `day = as.numeric( substr(nifty$Date, 9, 10) );`
    - OR `day = as.numeric( format(df[,1], format = "%d") );`
  - `idx = which(nifty$Close > nifty$Open & (day < 5));`
  - `idx = which( !(nifty$Close < nifty$Open | (day >= 5)) );`

- `which()` gives the indexes matching the criterion. E.g. out of `101:200` which numbers are multiples of 2,3 and 5 ?
  - `count = 101:200;`
  - `which( count %% 2 == 0 & count %% 3 == 0 & count %% 5 == 0 );`
  - `count[count %% 2 == 0 & count %% 3 == 0 & count %% 5 == 0];`

- We can do multi-way match using `%in%`
  - `mult_17 = seq(17,300,17);`
  - `which(count %in% mult_17);`
  - `which(mult_17 %in% count);`
  - `which(!(count %in% mult_17));`
  - `which(!(mult_17 %in% count));`

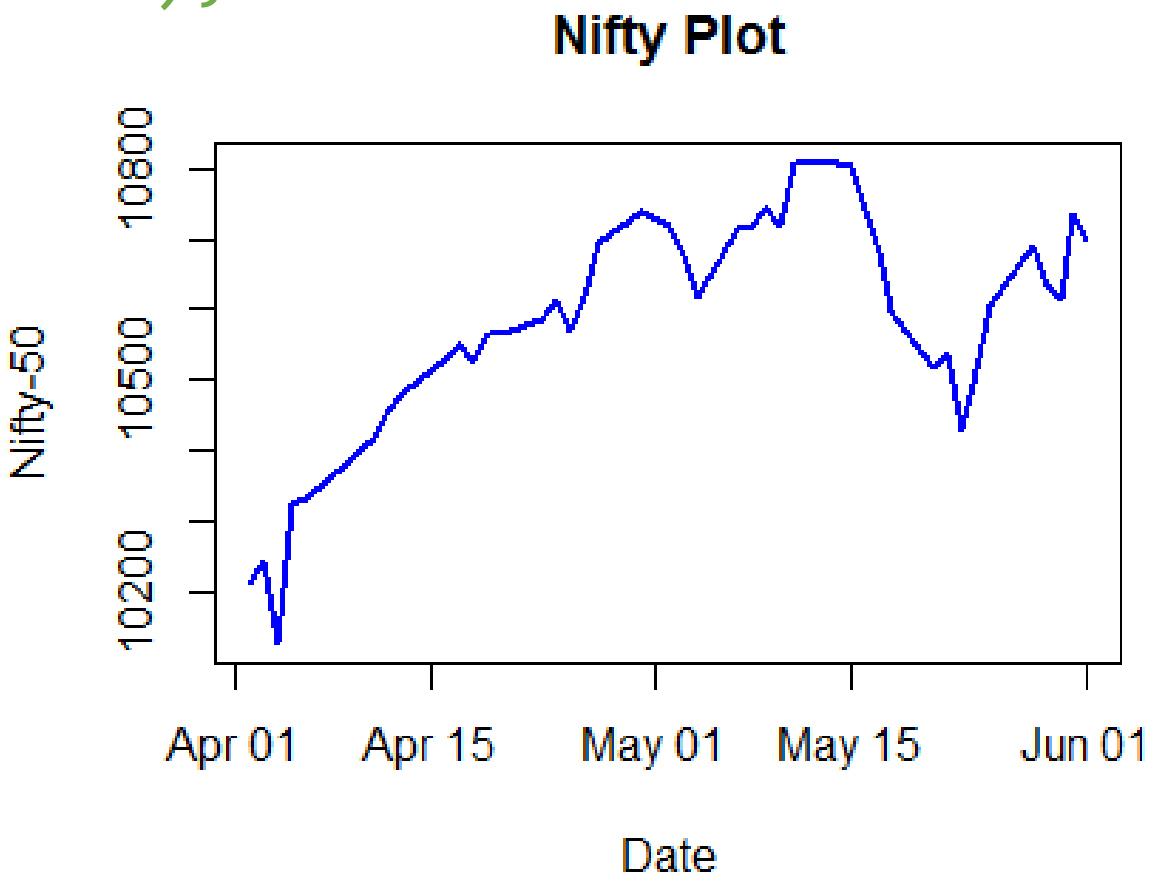
Plotting

- `plot(x, y, --options--);`
- `plot(nifty$Date, nifty$Close);`



Plotting (cont.)

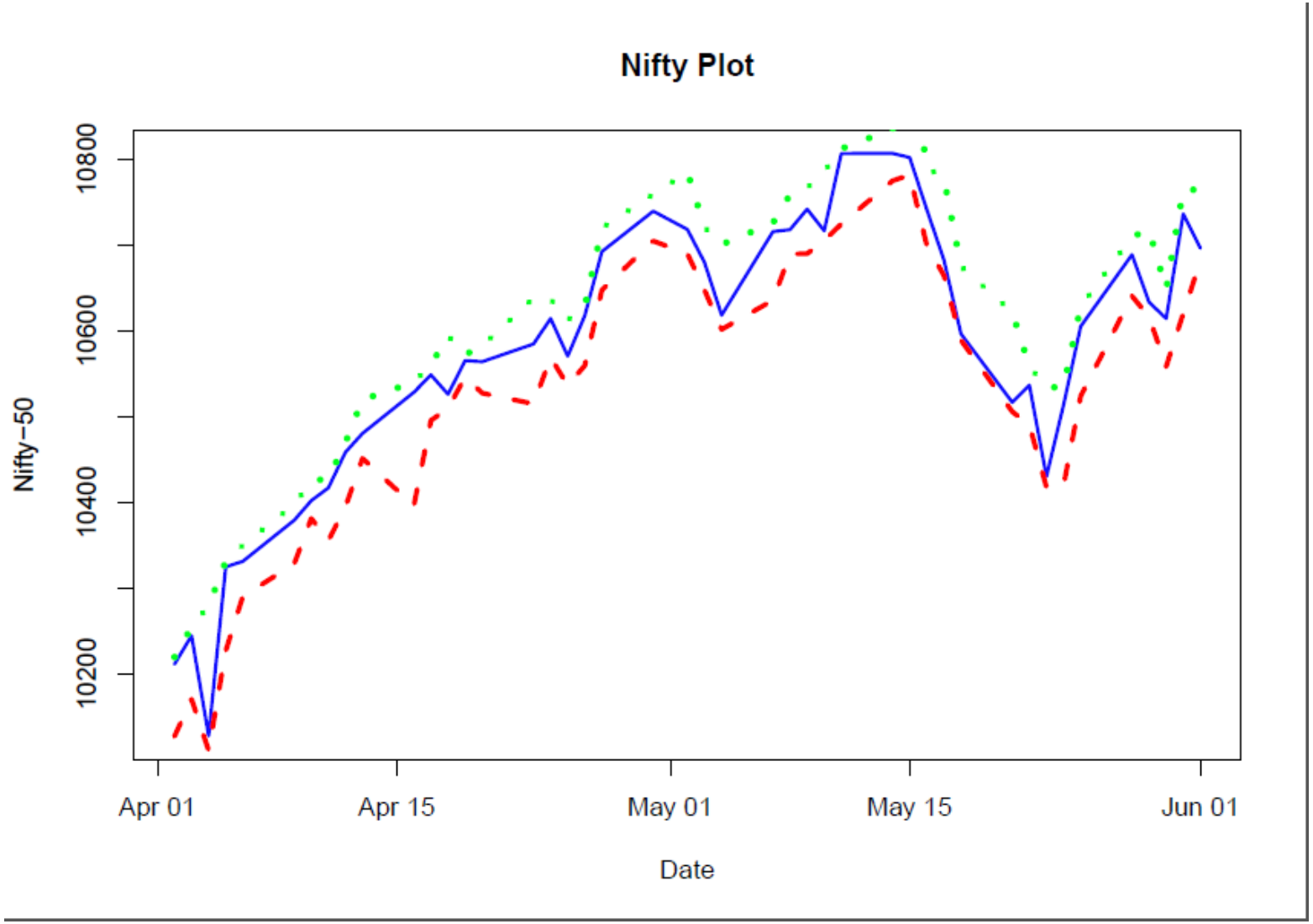
- `plot(nifty$Date, nifty$Close, type = 'l', col = "blue", lty = 1, lwd = 2, xlab = "Date", ylab = "Nifty-50", main = "Nifty Plot");`



Plotting (cont.)

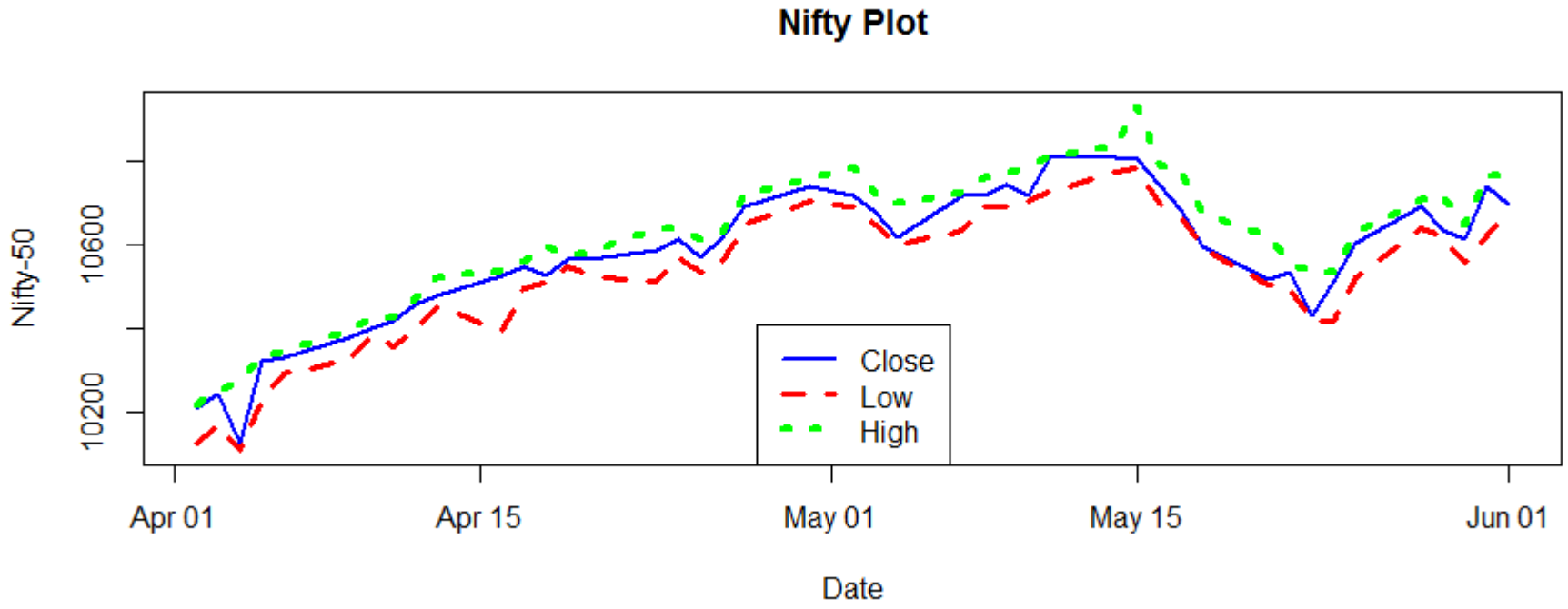
- `lines(nifty$Date, nifty$Low, type = 'l', lty = 2, col = "red", lwd = 3);`

- `lines(nifty$Date, nifty$High, type = 'l', lty = 3, col = "green", lwd = 4);`



Plotting (cont.)

- `y_lmt = c(min(nifty$Low), max(nifty$High));`
  - `Use plot(..., ylim = y_lmt, ...)`
- `legend("bottom", legend = c("Close", "Low", "High"), col = c("blue", "red", "green"), lty = 1:3, lwd = 2:4)`



aggregate()

- Aggregate values by subsets of data
  - Like mean air quality by month
  - Install the package: `install.packages("datasets");`
  - `df = datasets::airquality;`
  - `n = nrow(df);`
  - `head(df, n = 10); # first 10 rows, i.e. from 1 to 10`
  - `head(df, n = -10); # all but last 10 rows, i.e. from 1 to (n-10)`
  - `tail(df, n = 10); # last 10 rows, i.e. from (n-9) to n`
  - `tail(df, n = -10); # all but first 10 rows, i.e. from 11 to n`

- Avg. Ozone, Temp levels by month:
  - `aggregate(Ozone ~ Month, data = df, FUN = mean);`
  - `aggregate(Temp ~ Month, data = df, FUN = mean);`

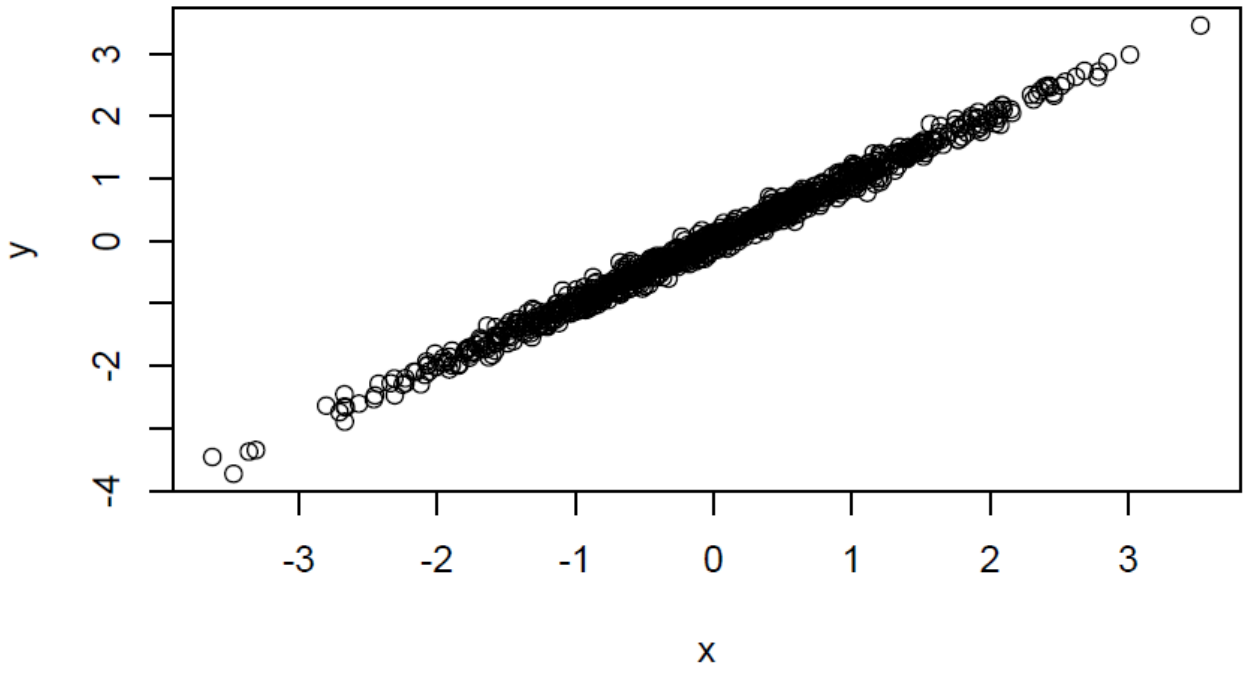
Correlation

- Difference between Independence and Correlation
  - $X \sim N(0,1), Y = X^2$. Are X and Y correlated?
  - Let's check in R
    - ```
      X = rnorm(1000, 0, 1);
      Y = X^2;
      cor(X,Y);
      ```
    - However, `cor(X,X^3)` != 0.
  - Let $Y \sim N(0,1)$ independent of X. Then is cor(X,Y) = 0? What about cor(X^q, Y)?
    - ```
      Y = rnorm(1000,0,1);
      ```
    - ```
      cor_q = rep(NA, 10);
      for(q in 1:10) {
          cor_q[q] = cor(X^q, Y);
      }
      cor_q;
      ```
  - Independence implies NO correlation of any functional form.
      ```

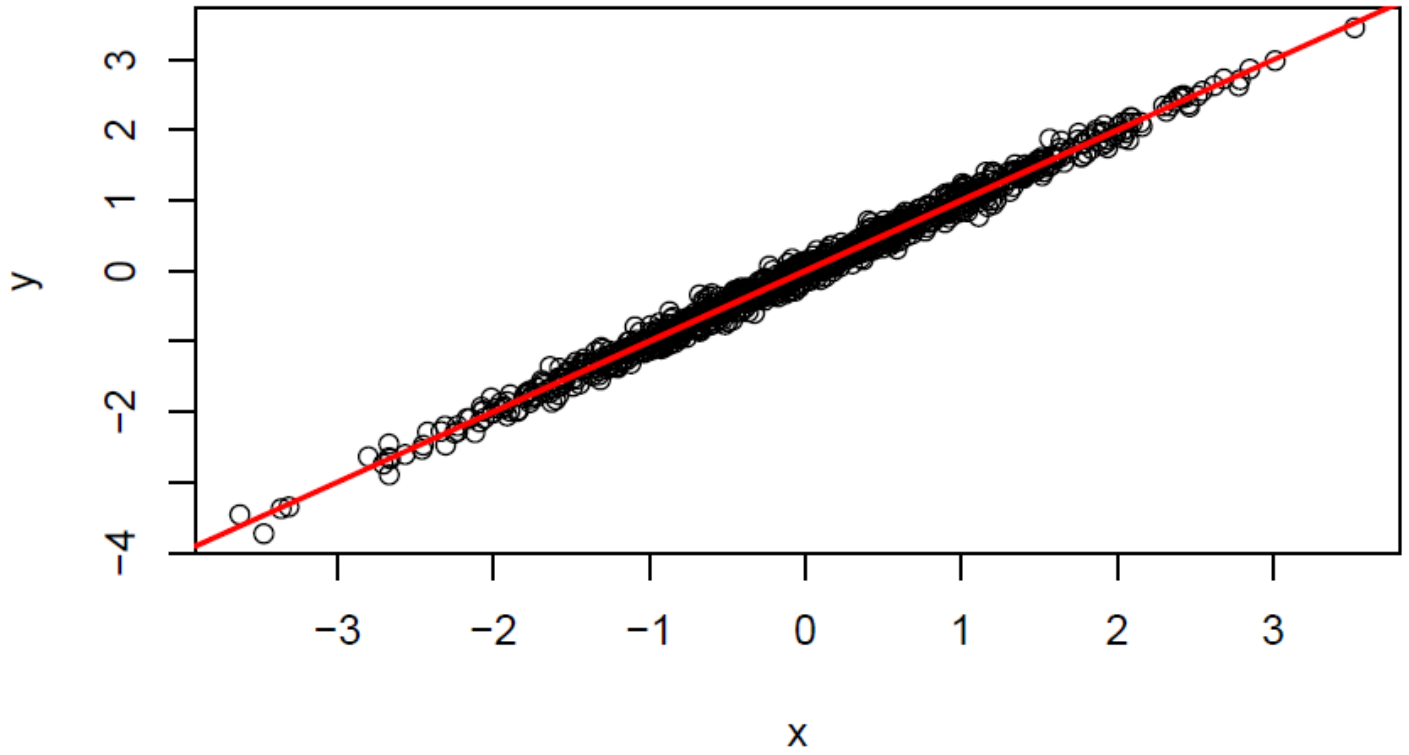
Regression

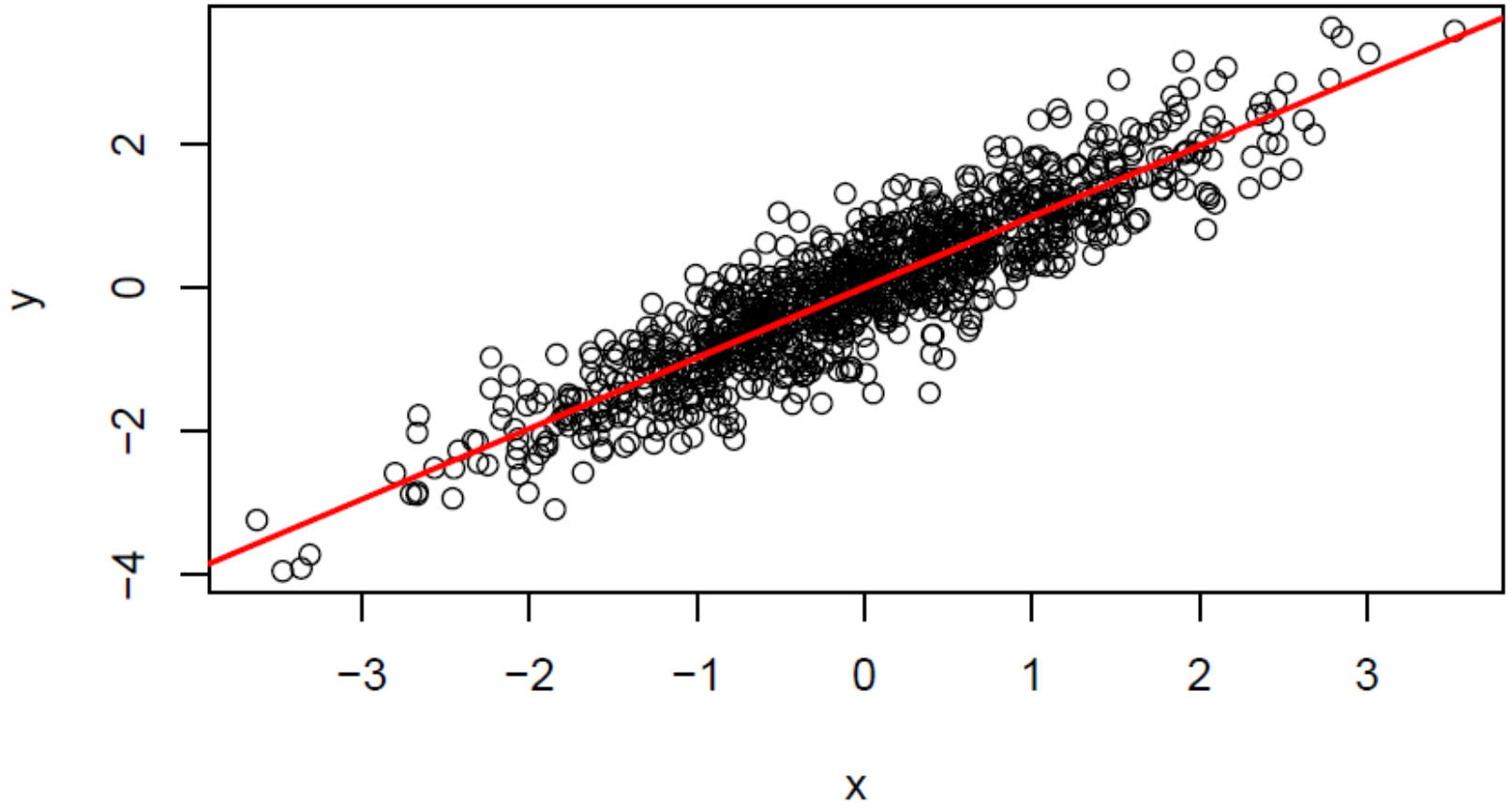
- Let, $Y = \beta_0 + \beta_1 \cdot X + u$ be the true model. By regressing Y on X, we hope to recover an unbiased estimate of β_1 and see how much of the variation in Y is explained by variation in X unrelated to variation in u.

```
n = 1000;
x = rnorm(n, 0, 1);
u = rnorm(n, 0, 0.1);
y = u + x;
plot(x,y);
```

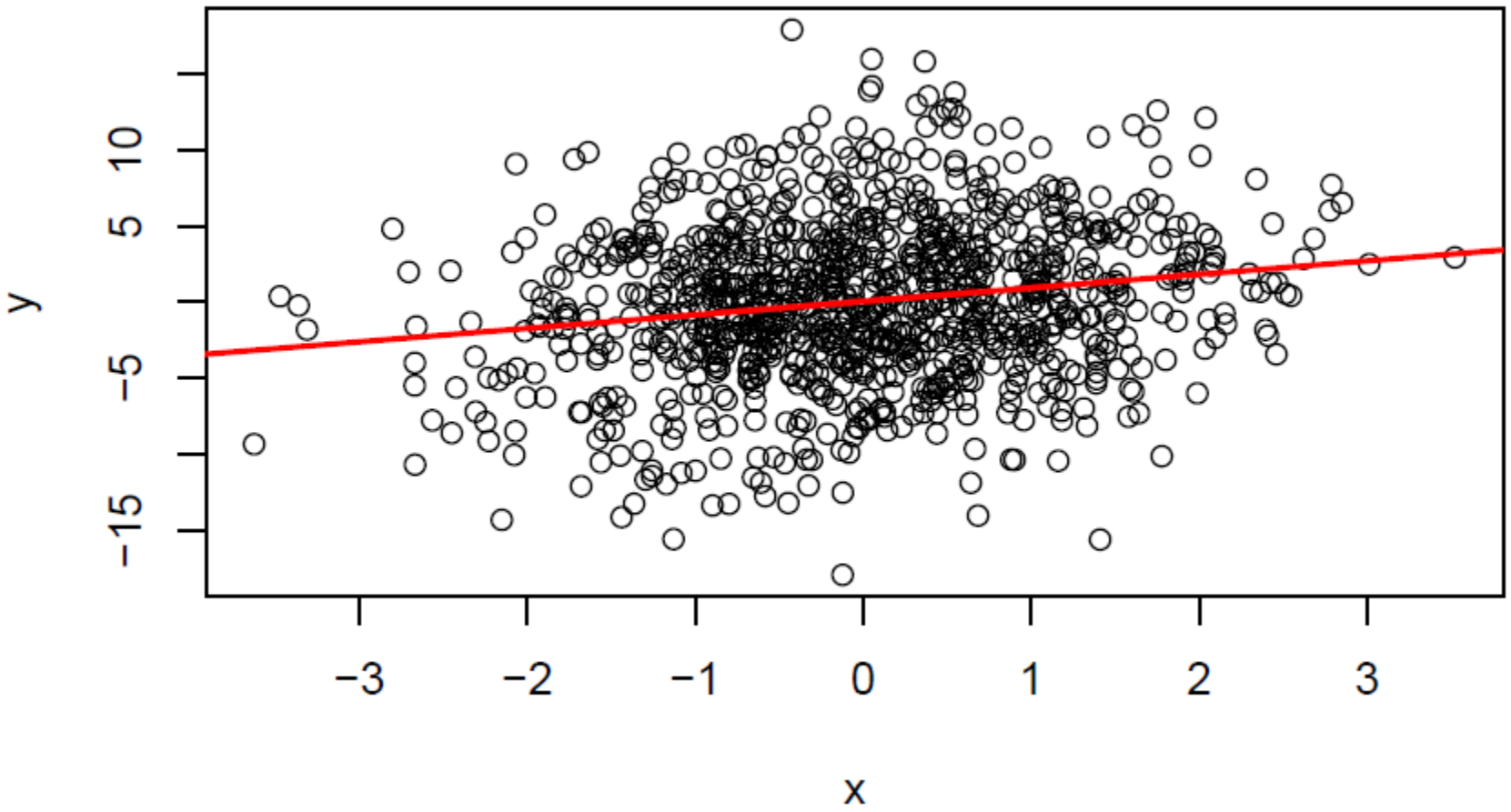


```
fit = lm(y ~ x);
summary(fit);
stargazer(fit, type = "html", out = "fit.html");
# Regression Line
abline(fit$coefficients, col = "red", lwd = 2);
```






```
u = rnorm(n, 0, 0.5);
y = u + x;
plot(x,y);
fit = lm(y ~ x);
summary(fit);
# Regression Line
abline(fit$coefficients, col = "red", lwd = 2);
```




```
u = rnorm(n, 0, 5);
y = u + x;
plot(x,y);
fit = lm(y ~ x);
summary(fit);
# Regression Line
abline(fit$coefficients, col = "red", lwd = 2);
```

Some Useful functions

- `unique(), duplicated()`
- `list.files()`
  - Pattern matching using regex
  - All files starting from “s”: `"^s"`
  - All files starting with “b” or “d”: `"^(b|d)“`
  - All CSV files: `".*.csv"`
- `order()`
  - Sort data/dataframes
  - It gives the sequence of ordered indexes NOT the ordered numbers
  - Can do 2-way and 3-way sorts
- `union(), intersect()`
- `cumsum(), cumprod()`
  - `cumprod()` using only `cumsum()` ?

Merging

- Combining two datasets is a very routine and important task
- Merging is akin to Joining (in relational database, SQL etc)

- Summary:

$r =$

attr1	attr2
a	r1
b	r2
c	r3

$s =$

attr1	attr3
b	s2
c	s3
d	s4

$r \bowtie s$

attr1	attr2	attr3
b	r2	s2
c	r3	s3

$r ⟕ s$

attr1	attr2	attr3
a	r1	*null*
b	r2	s2
c	r3	s3

$r ⟖ s$

attr1	attr2	attr3
b	r2	s2
c	r3	s3
d	*null*	s4

$r ⟗ s$

attr1	attr2	attr3
a	r1	*null*
b	r2	s2
c	r3	s3
d	*null*	s4

Merging Example

- Taken from [Stackoverflow webpage]
- Create two datasets:
  - `df1 = data.frame(cust_id = c(1:6), Product = c(rep("Toaster", 3), rep("Radio", 3)));`
  - `df2 = data.frame(cust_id = c(2, 4, 6), State = c(rep("Alabama", 2), rep("Ohio", 1)));`
  - The data looks like:
- We will need to use a new package `dplyr`
  - `install.packages("dplyr");`

df1		df2	
cust_id	Product	cust_id	State
1	Toaster	2	Alabama
2	Toaster	4	Alabama
3	Toaster	6	Ohio
4	Radio		
5	Radio		
6	Radio		

Inner Join

- `dplyr::inner_join(df1, df2);`

cust_id	Product	State
2	Toaster	Alabama
4	Radio	Alabama
6	Radio	Ohio

Full Join

- `dplyr::full_join(df1, df2);`

cust_id	Product	State
1	Toaster	NA
2	Toaster	Alabama
3	Toaster	NA
4	Radio	Alabama
5	Radio	NA
6	Radio	Ohio

Left Join

- `dplyr::left_join(df1, df2);`

cust_id	Product	State
1	Toaster	NA
2	Toaster	Alabama
3	Toaster	NA
4	Radio	Alabama
5	Radio	NA
6	Radio	Ohio

Right Join

- `dplyr::right_join(df1, df2);`
  - The above is identical to, `dplyr::left_join(df2, df1);`

cust_id	Product	State
2	Toaster	Alabama
4	Radio	Alabama
6	Radio	Ohio

Cartesian Product

- Every row of `df1` multiplied with every row of `df2`
- `cart_prd = merge(df1, df2, by = NULL);`

S.No.	cust_id.x	Product	cust_id.y	State
1	1	Toaster	2	Alabama
2	2	Toaster	2	Alabama
3	3	Toaster	2	Alabama
4	4	Radio	2	Alabama
5	5	Radio	2	Alabama
6	6	Radio	2	Alabama
7	1	Toaster	4	Alabama
8	2	Toaster	4	Alabama
9	3	Toaster	4	Alabama

S.No.	cust_id.x	Product	cust_id.y	State
10	4	Radio	4	Alabama
11	5	Radio	4	Alabama
12	6	Radio	4	Alabama
13	1	Toaster	6	Ohio
14	2	Toaster	6	Ohio
15	3	Toaster	6	Ohio
16	4	Radio	6	Ohio
17	5	Radio	6	Ohio
18	6	Radio	6	Ohio

- Cartesian products are extremely slow. Never do that even on a decent sized (> 1000 rows) dataset. Your computer will probably hang.
- Although there is no use of Cartesian products, it encapsulates all types of merges. Meaning we can extract any type of merge from a Cartesian product.
- Inner join can be extracted via
  - `idx = which(cart_prd$cust_id.x == cart_prd$cust_id.y);`
  - `cart_prd[idx,];`

Merging with more than one variable

- Most of the merging usage is with two variables: date and company name. Generate data using below:

```
date = seq.Date(as.Date("2018-01-01"), by = 1, length.out = 5);
comp = c("A", "B", "C", "D", "E");
all_pairs = merge(comp, date, by = NULL);
# sales data - 15 points
idx = sample(1:nrow(all_pairs), 15, replace = F);
df1 = data.frame(comp = all_pairs$x[idx], date = all_pairs$y[idx],
                 sales = round(runif(15, min = 1e3, max = 1e5)));
# advertising data - 12 points
idx = sample(1:nrow(all_pairs), 12, replace = F);
df2 = data.frame(comp = all_pairs$x[idx], date = all_pairs$y[idx],
                 adv = round(runif(12, min = 1e2, max = 1e4)));
```

df1

comp	date	sales
B	05-04-2018	53429
A	03-04-2018	70750
B	03-04-2018	5426
B	06-04-2018	88504
E	06-04-2018	78449
A	04-04-2018	98954
D	04-04-2018	57618
E	04-04-2018	22011
C	02-04-2018	85976
D	02-04-2018	42842
E	02-04-2018	94057
A	05-04-2018	25063
E	05-04-2018	51320
D	06-04-2018	99720
E	03-04-2018	16845

df2

comp	date	adv
C	03-04-2018	980
C	06-04-2018	6183
B	06-04-2018	8077
B	02-04-2018	241
D	02-04-2018	3642
B	04-04-2018	3878
E	02-04-2018	1501
B	03-04-2018	6727
D	06-04-2018	4259
D	03-04-2018	4434
A	06-04-2018	6745
E	03-04-2018	2208

Inner Join

- `dplyr::inner_join(df1, df2);`

comp	date	sales	adv
B	03-04-2018	5426	6727
B	06-04-2018	88504	8077
D	02-04-2018	42842	3642
E	02-04-2018	94057	1501
D	06-04-2018	99720	4259
E	03-04-2018	16845	2208

Full Join

- `dplyr::full_join(df1, df2);`

comp	date	sales	adv
B	05-04-2018	53429	NA
A	03-04-2018	70750	NA
B	03-04-2018	5426	6727
B	06-04-2018	88504	8077
E	06-04-2018	78449	NA
A	04-04-2018	98954	NA
D	04-04-2018	57618	NA
E	04-04-2018	22011	NA
C	02-04-2018	85976	NA
D	02-04-2018	42842	3642
E	02-04-2018	94057	1501
A	05-04-2018	25063	NA
E	05-04-2018	51320	NA
D	06-04-2018	99720	4259
E	03-04-2018	16845	2208
C	03-04-2018	NA	980
C	06-04-2018	NA	6183
B	02-04-2018	NA	241
B	04-04-2018	NA	3878
D	03-04-2018	NA	4434
A	06-04-2018	NA	6745

Left Join

- `dplyr::left_join(df1, df2);`

comp	date	sales	adv
B	05-04-2018	53429	NA
A	03-04-2018	70750	NA
B	03-04-2018	5426	6727
B	06-04-2018	88504	8077
E	06-04-2018	78449	NA
A	04-04-2018	98954	NA
D	04-04-2018	57618	NA
E	04-04-2018	22011	NA
C	02-04-2018	85976	NA
D	02-04-2018	42842	3642
E	02-04-2018	94057	1501
A	05-04-2018	25063	NA
E	05-04-2018	51320	NA
D	06-04-2018	99720	4259
E	03-04-2018	16845	2208

Right Join

- `dplyr::right_join(df1, df2);`
  - The above is identical to, `dplyr::left_join(df2, df1);`

comp	date	sales	adv
C	03-04-2018	NA	980
C	06-04-2018	NA	6183
B	06-04-2018	88504	8077
B	02-04-2018	NA	241
D	02-04-2018	42842	3642
B	04-04-2018	NA	3878
E	02-04-2018	94057	1501
B	03-04-2018	5426	6727
D	06-04-2018	99720	4259
D	03-04-2018	NA	4434
A	06-04-2018	NA	6745
E	03-04-2018	16845	2208

Loop Functions

- Writing loops in a single command. Can come very handy and compact. The function name ends with “apply”.
- List of functions: `lapply(), apply(), sapply(), tapply(), mapply()`
- These functions work on a list of inputs, not just one input!
- E.g.
  - `X = list(a = 1:10, b = rnorm(100, 0, 1), c = runif(1e3, 9, 91));`
  - `lappy(X, mean); # returns a list`
  - `sapply(X, mean); # returns a vector`
- Let’s say we want to find `cor(X,X^i)` for `i in 1:10` w/o writing a loop?
  - `X = rnorm(1000, 0, 1);`
  - `sapply(1:10, function(i) cor(X,X^i));`
    - Here we have used anonymous function, i.e. a function w/o a name.

- `apply()` is mostly used for applying functions on rows or cols of a matrix
  - Like taking means by rows
  - `M = matrix(1:50, nrow = 10, ncol = 5);`
  - `apply(M, 1, mean); # mean of each row`
  - `apply(M, 2, mean); # mean of each col`
  - You can do the above using a loop also, but `apply()` is more compact.
  - The faster version of above are also available:
    - `rowSums(x)` is equivalent to `apply(x, 1, sum)`
    - `colMeans(x)` is equivalent to `apply(x, 2, mean)`

- `tapply()` is used to apply a function over a subset of vector. Lets say,
  - `x = c(rnorm(100), runif(100));`
  - `f = rep(1:2, each = 100);`
  - `tapply(x, f, mean);`

Pipe and the `dplyr` package

- Download the `tidyverse` package and load it: `library(tidyverse);`
  - Also load `dplyr: library(dplyr);`

- We will use the `gapminder` dataset.
  - Install `gapminder` package
  - `df = gapminder::gapminder; df = as.data.frame(df);`

- Selecting a particular column
  - `df %>% select(lifeExp);`
  - Performing operation on data
    - `df %>% select(lifeExp) %>% mean`

- Selecting some rows (based on a filtering criterion):
  - `df %>% filter(year > 2000)`
  - `df %>% filter(year > 2000 & country == "India")`

- Selecting some rows (based on a filtering criterion):
  - `df %>% filter(year > 2000)`
  - `df %>% filter(year > 2000 & country == "India")`

- Selecting and filtering together
  - `df %>% select(country, year, lifeExp) %>% filter(year > 2000)`
  - `df %>% select(country, year, lifeExp) %>% filter(country == "India")`

- Grouping data
  - `df %>% select(country, year, lifeExp) %>% group_by(lifeExp) %>% as.data.frame %>% head`
  - `df %>% group_by(country, gdpPercap) %>% as.data.frame %>% head`

- Create new columns (mutate)
  - `df %>% mutate(decade = 10*(as.integer(year/10))) %>% head`

- Ordering
  - `df %>% arrange(-pop) %>% head`

- Performing aggregation on grouped data is the most important usage of dplyr library
- Aggregation:
  - `df %>% select(-c(continent)) %>% group_by(country) %>% summarise_all(mean)`
  - `df %>% select(-c(continent)) %>% group_by(country) %>% summarise_all(mean)`
  - `df %>% group_by(continent) %>% summarise_all(mean)`
  - `df %>% group_by(continent, year) %>% summarise_all(mean) %>% filter(continent == "Asia")`
- Counting and ordering in Aggregation:
  - `df %>% group_by(continent, year) %>% summarise(n()) %>% filter(continent == "Asia")`
  - `df %>% group_by(continent) %>% summarise(n())`
  - `df %>% filter(year == 2007) %>% group_by(pop) %>% arrange(-pop)`
  - `df %>% filter(year == 2007) %>% group_by(pop) %>% arrange(-gdpPercap)`
  - `df %>% filter(year == 2007) %>% group_by(pop) %>% mutate(GDP = gdpPercap*pop) %>% arrange(-GDP) %>% as.data.frame %>% select(country, GDP) %>% head(15)`