# Introduction to R 

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## Outline

## (1) Introduction to R

(2) Basic R Language

- Vectors, Matrices, Factors, Lists and Data Frames
- Indexing
- Conditional Selection
- Missing Values
(3) The R Environment
- R Session Management
- Useful Functions for Data Handling
- The Graphics Subsystem
- Data Input and Output


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## What is $R$ ?

- R is a GNU project (free-software, mass-collaboration project).
- To find out about R go to http://www.R-project.org/.
- Check also the NY Times article http://www.nytimes.com/2009/01/07/technology/ business-computing/07program.html?pagewanted=all.


## How to Install R?

- To download R go to http://cran.r-project.org/.
- You may also want to install Rstudio. To download Rstudio go to http://www.rstudio.com/. But I will not use Rstudio in class.


## Assignments

- R has symbolic variables, that can be used to represent values.
- $x$ <- 2 \# recommend
- $\mathrm{x}=2$
- Names of variables can be chosen quite freely in R. Letters, digits, underscore, and period (dot) can be used.
- height.1yr \# can be used to describe the height of a child at the age of 1 year
- Names are case-sensitive.
(e.g. Height and height do not refer the same variable)
- Try not to use variable names same as popular basic function names. (e.g. mean and diff)


## Vectorized Arithmetic

- R can handle entire data vectors as single objects.
- The construct $c()$ is used to define vectors.
- weight <- c (60, 72, 57, 90, 95, 72) \# (kg)
- height <- c ( $1.75,1.80,1.65,1.90,1.74,1.91$ ) \# (m)
- You can do calculations with vectors just like ordinary numbers, as long as they are of the same length.
- BMI <- weight/height^2
- Note that the operation is carried out element wise. (that is, the first value of BMI is $60 / 1.75^{\wedge} 2$ )
- Try calculate the mean of weight ( $\bar{x}=$ sum of all weights / number of observations) and compare the result with using mean() function. (Hint: use sum() and length())


## Graphics

- One of the most important aspects of the presentation and analysis of data is the generation of proper graphics.
- To check the relation between weight and height, the first idea is to plot one versus the other.
- plot(height, weight)
- plot(height, weight, pch = 2) \# pch (plotting character)
- To add horizontal line or vertical line to an existing plot use function abline()
- abline(h = 70, col = "blue") \# horizontal line
- abline(v = 1.85, col = "red") \# vertical line


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## Vectors

- A generic function of constructing vectors is c(), "concatenate".
- A numeric vector can take the numbers.
- c(1, 3, 5)
- A character vector can take the text strings, whose elements are specified and printed in quotes.
- c("Messi", "Ronaldo", "Hazard")
- A logical vector can take the value TRUE or FALSE.
- c(TRUE, FALSE, FALSE)
- c(T, F, F)
- All elements of a vector have the same type. If you create a vector of different types, they will converted to the least "restrictive" type.
- c(1, "Messi")


## Other Functions to Create Vectors

- seq(), "sequence", is used for equidistant series of numbers.
- $\operatorname{seq}(4,10)$
- seq(4, 10, 2) \# a sequence in jumps of 2
- $\operatorname{seq}(1.65,1.90,0.05)$ \# a sequence in jumps of 0.05
- $\operatorname{rep}()$, "replicate", is used to generate repeated values.
- \# Second argument is a number
- rep (2, 5)
- vec1 <- c(1, 2)
- rep(vec1, 5)
- \# Second argument is a vector
- rep(vec1, c(5, 10))
- rep(vec1, c(10, 10))
- rep(vec1, each = 10) \# special case where there are equally many replications of each value


## Matrices

- A matrix in mathematics is just a two-dimensional array of numbers.
- Matrices are used a lot for numerical computation purpose.
- matrix() is used to construct a matrix.
- mat.elem <- seq(1,6)
- mat1 <- matrix(mat.elem, nrow = 2, ncol = 3)
- mat2 <- matrix(mat.elem, nrow = 2, ncol = 3, byrow = T) \# the matrix to be filled in a rowwise
- rownames() is used to name the rows of the matrix.
- colnames () is used to name the columns of the matrix.
- rownames(mat1) <- c("Male", "Female")
- colnames(mat1) <- c("A", "B", "C")
- $\operatorname{dim}()$ is mostly used to check the dimension of the matrix.
- dim(mat1)


## Matrices

- You can "glue" vectors together to make a matrix.
- rbind() is used to glue vectors rowwise.
- cbind() is used to glue vectors columnwise.
- vec1 <- c(1, 2, 3)
- vec2 <- c(4, 5, 6)
- rbind(vec1, vec2)
- cbind(vec1, vec2)


## Factors

- Factors are used to store categorical (or qualitative) variables. (e.g. gender, nationality, and eye color)
- factor () is used to construct a factor.
- gender <- factor(c("Male", "Female", "Male", "Male", "Female", "Female"))
- You may specify the levels by yourself by setting levels argument.
- nationality <- factor(c("uk", "us", "au", "uk", "us", "us"), levels = c("us", "fr", "au", "uk"))


## Lists

- Lists are used to combine collection of objects (e.g. vectors, factors, etc.) into a larger composite object.
- list() is used to construct a list.
- pre <- c(5260, 5470, 5640, 6180, 6390, 6515)
- post <- c(3910, 4220, 3885, 5160, 5645, 4680)
- my.list <- list(before = pre, after = post)
- The components of the list are named according to the argument names used in list(). To extract the named components:
- my.list\$before
- my.list\$after
- Many of R's built-in functions return their results in the form of list.


## Data Frames

- Almost every real data set is stored in the form of data frame.
- Each row of a data frame is an observation, each column of a data frame is a variable.
- data.frame() is used to construct data frames.
- my.df <- data.frame(pre, post)
- To extract the components of data frame:
- my.df\$pre
- my.df\$post


## Difference between Lists and Data Frames

- Lists are the most flexible data structure in R.
- Lists have no restriction on the class, length or structure of each element.
- Data frames are lists as well, but they have a few restrictions:
- All elements are vectors.
- All elements have an equal length.
- These restrictions are resulting two-dimensional structure.
- Data frames can follow some of the behavior of matrices. (e.g. select rows, columns, and elements)


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## Indexing of Vectors

- Brackets are used for selection of data, also known as indexing.
- pre[3] \# third element of vector
- It can be also used to modify elements of vector.
- pre[3] <- 6000
- pre[3] <- 5640
- If you want a subvector then you can index with a vector:
- pre $[c(1,3,5)]$
- pre[1,3,5] \# wrong!
- You can also use index vector stored in a variable.
- $v<-c(1,3,5)$
- pre[v]


## Indexing of Matrices

- Matrices may be indexed by giving two indices in the form,
- my.mat <- cbind(pre, post)
- my.mat[3,2] \# my.mat[row, column]
- my.mat [5,1] \# fifth row and first column of matrix
- It can be also used to modify elements of matrix.
- my.mat[3,2] <- 3575
- my.mat[3,2] <- 3885
- If you want a whole row or column from the matrix:
- my.mat[2, ] \# second row of matrix
- my.mat[, 1] \# first column of matrix
- If you want a submatrix then you can index with a vector:
- my.mat $[c(3,4)$,
- v <- c(3, 4)
- my.mat[v, ]


## Indexing of Data Frames

- Data frames may be indexed by giving two indices in the form similar to matrix,
- my.df <- data.frame(pre, post)
- my.df [1,2]
- It can be also used to modify elements of data frame.
- my.df[4,2] <- 4760
- my.df[4,2] <- 5160
- If you want a whole row or column from the matrix:
- my.df[2, ] \# second row of data frame
- my.df[, 2] \# second column of data frame
- my.df\$post \# second column of data frame


## Negative Indexing

- Indexing is a special powerful ability of $R$.
- R can also do negative indexing.
- You can get all observations except numbers (or vectors) you specify.
- Negative indexing for vectors
- pre[-3]
- pre[-c(1, 3, 5)]
- Negative indexing for matrices
- my.mat [-3, ]
- Negative indexing for data frames
- my.df[-3, ]


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## Conditional Selection (Vectors)

- In practice, you often need to extract data that satisfy certain criterion.
- post[post < 4000]
- post[pre > 6000]
- which() function will return the position of the elements in a logical vector which are TRUE.
- post < 4000
- which(post < 4000)
- Comparison operators in R:
- < \# less than
- > \# greater than
- == \# equal to
- <= \# less than or equal to
- $>=$ \# greater than or equal to
- != \# not equal to


## Conditional Selection (Vectors)

- In practice, you also need to extract data satisfying several criteria.
- post[post < 4000 | post > 5500]
- post[pre > 6000 \& pre <= 6400]
- which () function usage
- post < 4000 | post > 5500
- which(post < 4000 | post > 5500)
- Logical operators in R:
- \& \# logical 'and"
- | \# logical "or"
- ! \# logical 'rnot"


## Conditional Selection (Matrices and Data Frames)

- In practice, it is very common to select just those rows of a matrix or data frame that meet some criteria.
- For matrix,
- my.mat[my.mat[, 1] > 6000, ]
- my.mat[my.mat[, 2] < 5000, ]
- my.mat[my.mat[, 1] > 6000 \& my.mat[, 2] < 5000, ]
- For data frame,
- my.df[my.df\$pre > 6000, ]
- my.df[my.df\$post < 5000, ]
- my.df[my.df\$pre > 6000 \& my.df\$post < 5000, ]
- Don't forget comma after the condition because we want to extract rows which are observations.


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## Missing Values

- In practical data analysis, a data is frequently unavailable. (e.g. the patient did not show up, an experiment failed, or the survey question left blank)
- R allows vectors to contain a special NA value.
- This NA is carried through in computations so that operations on NA yield NA result.
- weight <- c (60, 72, NA, 90, 95,72$)$
- weight - 75


## To Find Missing Values

- is.na() is used to find which elements of vector are recorded as missing, NA.
- midterm <- c (98, 72, NA, 89, 69, 92, 78, NA, 94, NA, 83)
- is.na(midterm)
- which(is.na(midterm))
- is.na() is important because "midterm == NA" gives NA as the result for any value of midterm.


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## Workspace

- All variables created in R are store in a common workspace.
- ls() is used to see which variables are defined in workspace.
- a <- seq $(1,10)$
- b <- 25:30
- c <- rep $(0,20)$
- d <- c(17, 73, 92, 102, 234)
- ls()
- rm() is used to delete some of the objects.
- rm(b)
- rm(a, c)
- rm(list=ls()) \# To delete all objects in workspace.


## Scripting

- Beyond a certain level of complexity, you don't want to work with R on a line-by-line basis.
- In this case you can work with $R$ scripts, collection of lines of $R$ code.
- You can also save your $R$ code when you work with $R$ scripts and load again later.


## Packages

- Some packages are part of basic installation.
- Others can be downloaded from CRAN, which hosts over 1000 packages for various purpose.
- To install the package,
- For Mac user go to "Packages \& Data" tab and click "Package Installer".
- For Window user go to ..
- install.packages("type package name")
- Let's try install ggplot2 package.


## Packages

- library() is used to load the package that you installed.
- library (ggplot2)
- The loaded packages are not considered part of the user workspace.
- If you terminate $R$ and start again, then you will have to load the package again.


## ggplot2: Better Graphics for R

- Since we installed ggplot2 package let us try using it!
- p <- ggplot(mtcars, aes(wt, mpg))
- p + geom_point()
- Compare the ggplot2 plot with original plot() function.
- dev.new() \# opens new graphic window tab
- plot(mtcars\$wt, mtcars\$mpg)


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## attach()

- Let us use the famous data set iris which is available in R. Type iris directly to see what it looks like.
- The notation for accessing variables in the data set gets annoying if we repeatedly have to write longish commands like
- iris[iris\$Sepal.Length > 6.5 \& iris\$Sepal.Width < 3.0, ]
- attach() is used to make R to look for objects among the variables in a given data frame.
- Sepal.Length \# before using attach() it gives you an error
- iris\$Sepal.Length \# this was correct way
- attach(iris)
- Sepal.Length
- iris[Sepal.Length > 6.5 \& Sepal.Width < 3.0, ]


## subset ()

- subset () is used to select subsets of data frame.
- new.iris1 <- subset(iris, Sepal.Length > 6.5)
- new.iris2 <- subset(iris, Sepal.Length > 6.5 \& Sepal.Width < 3.0)
- new.iris3 <- subset(iris, Species == "setosa")
- new.iris4 <- subset(iris, select = c(Sepal.Length))
- new.iris5 <- subset(iris, select = c(Sepal.Length, Sepal.Width))


## transform()

- transform() is used to create new data frames with transformed variables.
- new.iris6 <- transform(iris, log.Sepal.Length = log(Sepal.Length))
- new.iris7 <- transform(iris, diff.Sepal = Sepal.Length Sepal.Width)


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## Plot Layout

- plot() is generic function for plotting R objects. By default, draw a scatter plot.
- plot(Sepal.Length, Sepal.Width)
- main = "put overall title" is used to put overall title of the plot.
- xlab = "put $x$-axis title" is used to change the $x$-axis title.
- ylab = "put y-axis title" is used to change the $y$-axis title.
- plot(Sepal.Length, Sepal.Width, main = "Sepal Iris", xlab = "Length", ylab = "Width")
- abline() is used to add straight line to the plot
- abline( $\mathrm{a}=0.3, \mathrm{~b}=0.5$ ) \# y $=\mathrm{a}+\mathrm{bx}$
- abline(h = 3.0) \# horizontal line
- abline(v = 6.0) \# vertical line


## Using par()

- par() is used to set many graphical parameters. Here, we focus on one main usage.
- $\operatorname{par}(\mathrm{mfrow}=c())$ is used to put multiple figures simultaneously.
- par(mfrow = c(1, 2))
- plot(Sepal.Length, Sepal.Width, main = "Sepal Iris", xlab = "Length", ylab = "Width")
- plot(Petal.Length, Petal.Width, main = "Petal Iris", xlab = "Length", ylab = "Width")


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## Data Input from a Text File (.txt)

- read.table() is used to import a text file to R .
- Please follow the steps below:
- Step 1: Prepare a new folder, say Rdemo. Download the data set (e.g. Auto.txt) to the folder.
- Step 2: Set the working directory of R to the folder where the data set is located.
- Step 3: Read in the data set by calling: autotxt <- read.table("Auto.txt")
- Assign the result of read.table() is always desirable.


## Data Input from an Excel File (.csv)

- read.csv() is used to import an Excel file to R.
- Please follow the steps below:
- Step 1: Prepare a new folder. Download the data set (e.g. Auto.csv) to the folder.
- Step 2: Set the working directory of R to the folder where the data set is located.
- Step 3: Read in the data set by calling:
autocsv <- read.csv("Auto.csv")
- Assign the result of read.csv() is always desirable.


## Data Output to a Text File / Excel File

- write.table() is used to export a data frame object to a text file.
- write. $\operatorname{csv}()$ is used to export a data frame object to a csv file.
- write.table(autotxt, file = "textauto.txt")
- write.csv(autocsv, file = "csvauto.csv")
- Check your directory folder to make sure you have a new file.

