

DataScience for Development and Social Change, 2015

The R Language

What it is, what you can do
with it

The R Programming Language

- ❖ Good at:
 - ❖ Rapid statistical analysis (4000+ R libraries)
 - ❖ Rapidly-created static graphics
- ❖ Bad at:
 - ❖ Web applications (although you can use Shiny)
 - ❖ Pretty interactive graphics (use D3)
 - ❖ Iteration (use almost anything else)

Download R from <http://www.r-project.org/>

Starting R

- ❖ 4 ways to run R:
 - ❖ Terminal window: type “r” (and “q()” to quit)
 - ❖ Rstudio: click on Rstudio tool
 - ❖ R script from the command line: “R <filename.r —no-save”
 - ❖ R script from inside R: source('myscript.r')
 - ❖ From Python: use the rpy2 library

Getting help

- ❖ Inside R or Rstudio:
 - ❖ `help(functionname)`
 - ❖ `example(functionname)`
 - ❖ e.g. `help(sum)`, `example(sum)`
- ❖ R community:
 - ❖ R bloggers
 - ❖ R local user groups
 - ❖ R mailing lists
 - ❖ Stack overflow

Variables

- ❖ data types:
 - ❖ 2.456
 - ❖ 'This is a string'
 - ❖ TRUE, FALSE
- ❖ Variables:
 - ❖ `x <- 15`
 - ❖ `x/4`
 - ❖ `y <- 'This is also a string!'`
 - ❖ `print(x)`

Vectors

- ❖ Vector = same as Python's list
 - ❖ `a <- c(1,5,2)`
 - ❖ note the 'c' function ('combine')
- ❖ NB if you try creating a mixed-variable vector, R will create a vector of strings. Try this:
 - ❖ `x <- c(1,'five',2,TRUE)`
- ❖ Indices:
 - ❖ `x[3]`
 - ❖ `x[2] <- 'four'`
 - ❖ `x[2:4]`
 - ❖ NB: R indices start at 1 (unlike Python, which start at 0)

Sequences

- ❖ 1:5
- ❖ seq(1,5)
- ❖ seq(1,5,0.3)
- ❖ x[2:4]

Functions

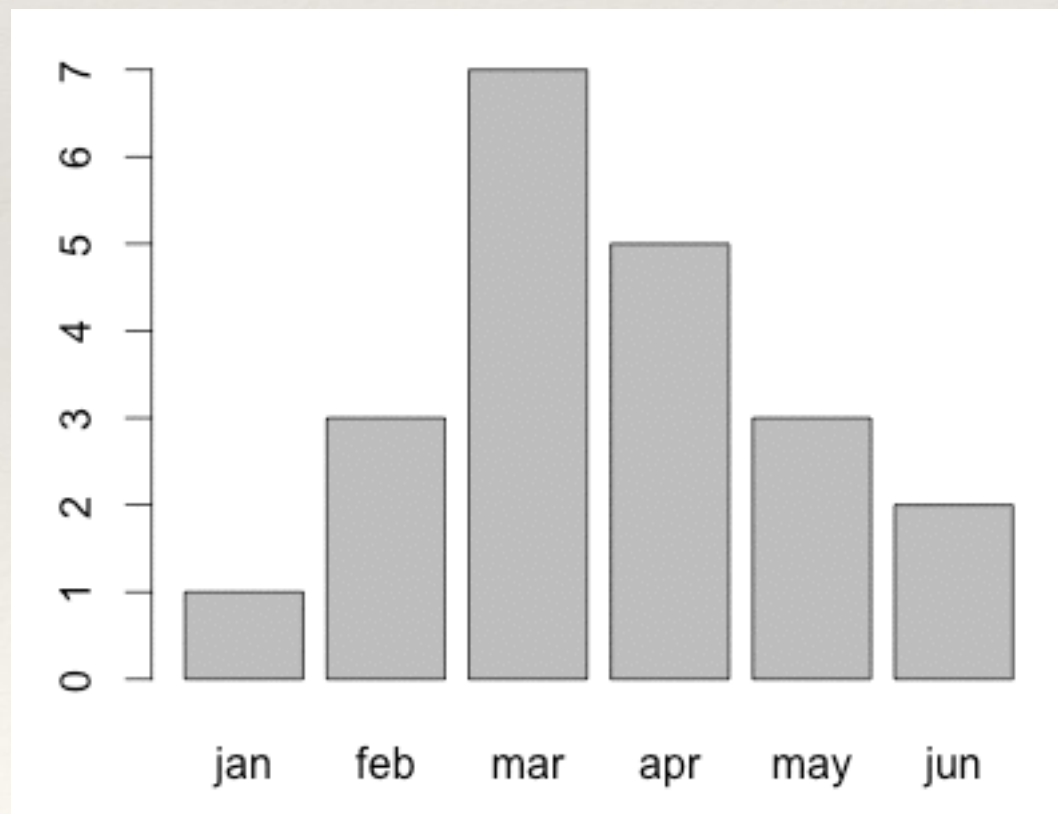
- ❖ `sum(1,2,3,7,3,4)`
- ❖ `a <- c(1,2,3,7,3,4)`
- ❖ `sum(a)`
- ❖ `max(a)`
- ❖ `sqrt(a)`

Vector names

- ❖ You can name each entry in your vector:
 - ❖ `x <- c(1,'five',2,TRUE)`
 - ❖ `names(x) = c('1st', '2nd', 'third', 'rainbow')`
 - ❖ `x['2nd']`
 - ❖ `x['rainbow'] <- FALSE`

Visualisations

- ❖ `b <- c(1,3,7,5,3,2)`
- ❖ `names(b) <- c('jan','feb','mar','apr','may','jun')`
- ❖ `barplot(b)`

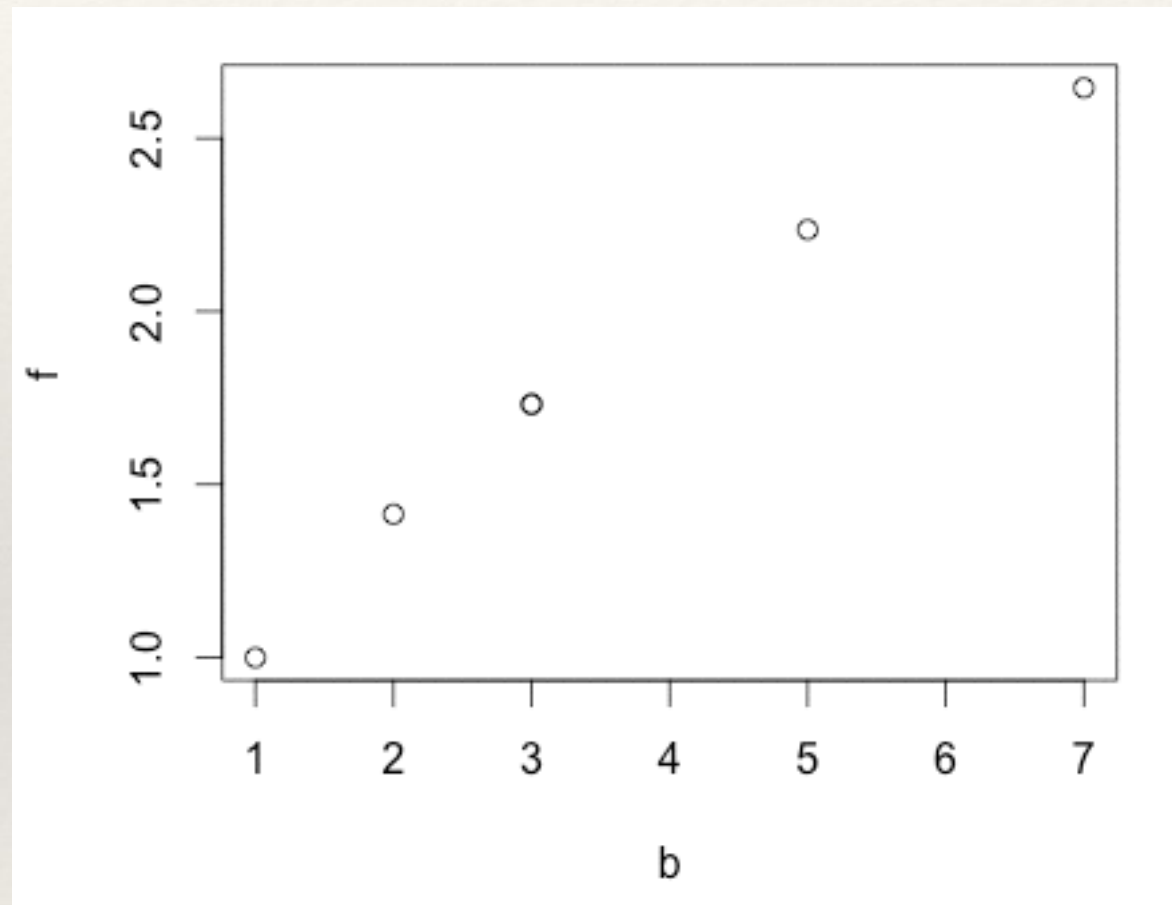


Vector Maths

- ❖ `b <- c(1,3,7,5,3,2)`
- ❖ `b+1`
- ❖ `b/4`
- ❖ `sqrt(b)`
- ❖ `d <- c(3,6,5,4,3,7)`
- ❖ `b*d`
- ❖ `b == d`

More visualizations

- ❖ `b <- c(1,3,7,5,3,2)`
- ❖ `f <- sqrt(b)`
- ❖ `plot(b,f)`



Handling missing values

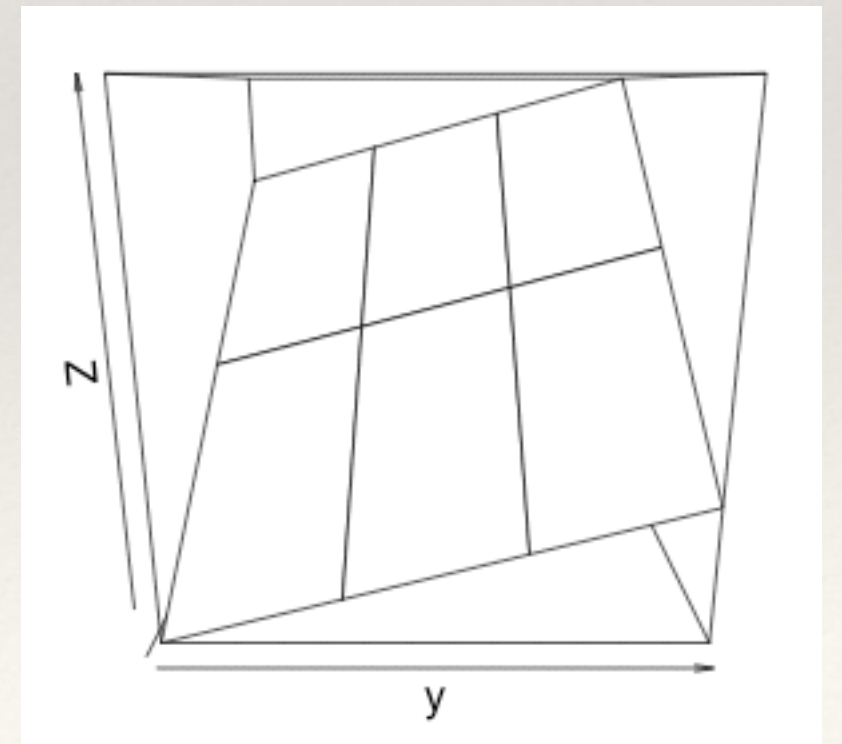
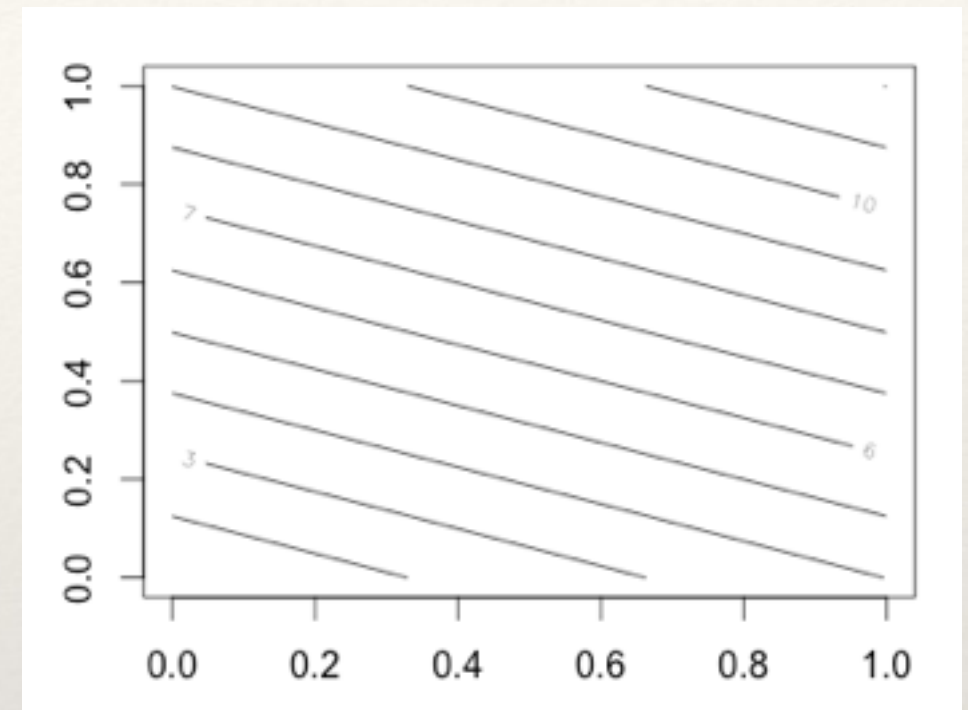
- ❖ `b <- c(1,3,7,NA,3,2)`
- ❖ `sum(b)`
- ❖ `sum(b, na.rm=TRUE)`

Matrices (two-dimensional arrays)

- ❖ `x <- matrix(0, 4, 3)`
- ❖ `y <- matrix(1:12, 4, 3)`
- ❖ `y[2,3] <- 15`
- ❖ `y[,3]`
- ❖ `y[2,2:3]`
- ❖ `print(y[1,3])`
- ❖ `z <- 1:15`
- ❖ `dim(z) <- c(3,5)`
- ❖ `print(z)`

More visualisation

- ❖ `y <- matrix(1:12, 4, 3)`
 - ❖ `contour(y)`
 - ❖ `persp(y)`
-
- ❖ R comes with example datasets:
 - ❖ `contour(volcano)`
 - ❖ `persp(volcano)`

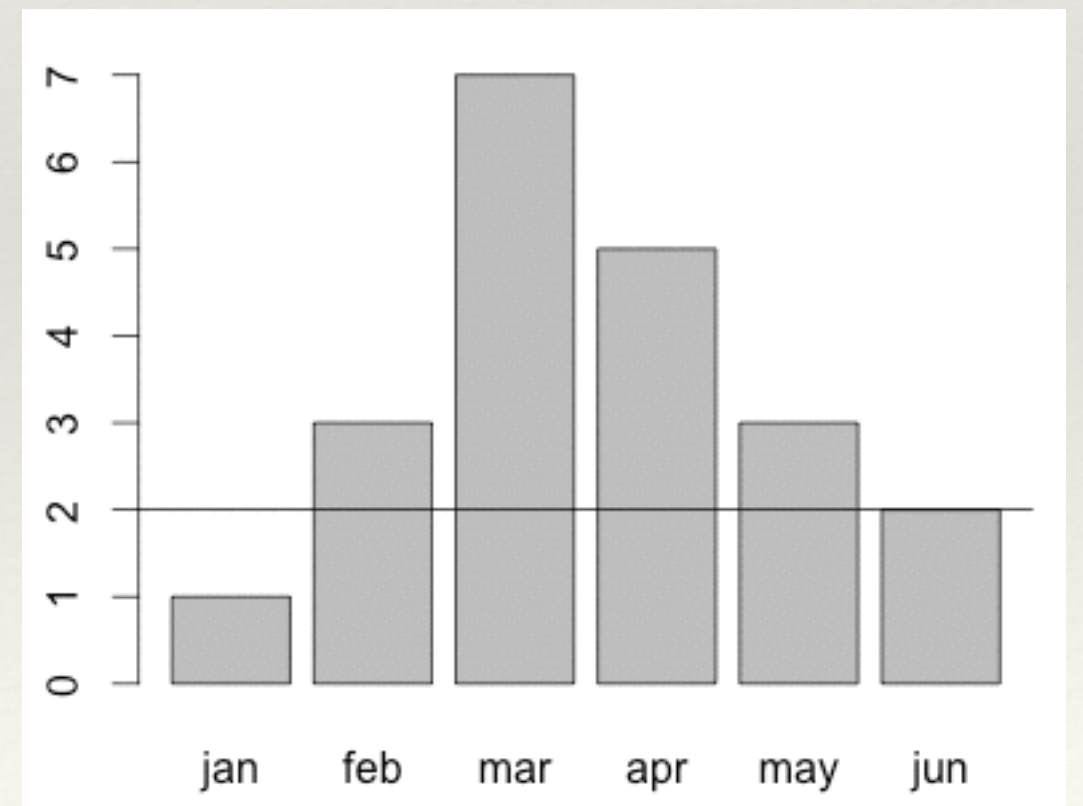


Getting Statistics

- ❖ `b <- c(1,3,7,5,3,2)`
- ❖ `mean(b)`
- ❖ `median(b)`
- ❖ `sd(b)`

Adding to visualizations

- ❖ `b <- c(1,3,7,5,3,2)`
- ❖ `names(b) <- c('jan','feb','mar','apr','may','jun')`
- ❖ `barplot(b)`
- ❖ `abline(h=2)`



Factors

- ❖ `a <- c('developed', 'transitioning', 'developed', 'developing', 'developed', 'developing', 'developed')`
- ❖ `classes <- factor(a)`
- ❖ `print(classes)`
- ❖ `print(levels(classes))`

DataFrames (tables)

- ❖ `refugees = c(23,175,15543,11,338796,1244,63,53)`
- ❖ `asylumseekers = c(140,77,2872,6,255,91,6,7)`
- ❖ `origins = c('Burundi','Dem. Rep. of the Congo', 'Eritrea',
'Rwanda', 'Somalia','South Sudan','Sudan','Uganda')`
- ❖ `countries <- factor(origins)`
- ❖ `popstats <- data.frame(refugees, asylumseekers, countries)`
- ❖ `print(popstats)`
- ❖ `print(popstats$asylumseekers)`

Reading CSV files into Dataframes

- ❖ List all files:
 - ❖ `list.files()`
- ❖ Read a CSV into a data frame:
 - ❖ `refugees <- read.csv('popstats_clean.csv')`
- ❖ Read a tab-separated file into a data frame:
 - ❖ `population <- read.table('sp_pop.tsv', sep='\t', header=TRUE)`

Exploring a Dataframe

- ❖ `refugees <- read.csv('popstats_clean.csv')`
- ❖ `dim(refugees)`
- ❖ `head(refugees)`
- ❖ `str(refugees)`

Merging dataframes

- ❖ `merge(x = refugees, y = population, by.x='Country.of.origin', by.y='Country.Name')`
- ❖ Spaces in headings got converted to dots
- ❖ Use `by.x`, `by.y` *unless* you have at least one column with the same name in both files

Science: correlation

- ❖ `plot(column1, column2)`
- ❖ `cor.test(column1, column2)`
- ❖ Is the p-value smaller than 0.05? Yes = strong evidence for correlation.

Science: linear models

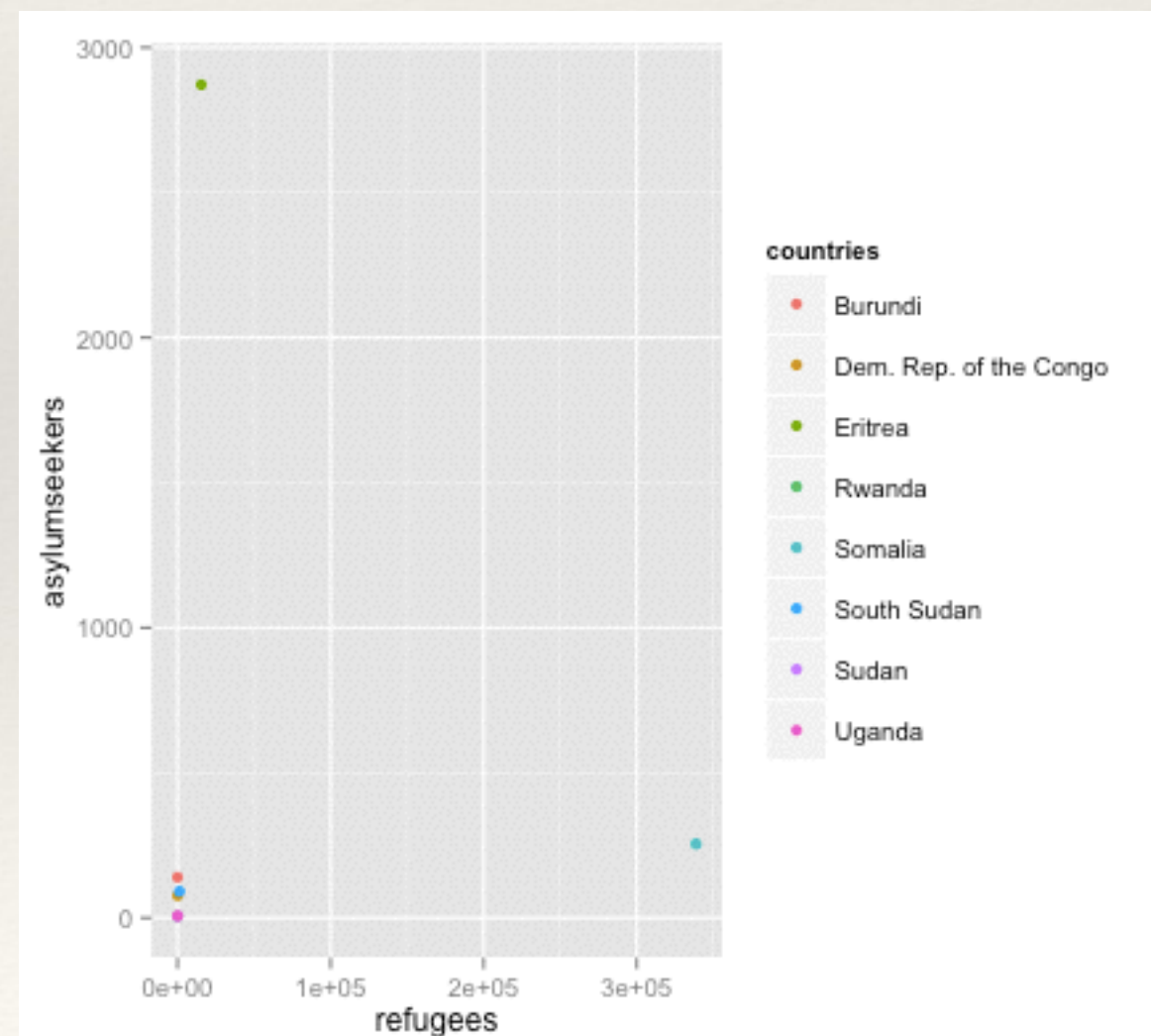
- ❖ `plot(column1, column2)`
- ❖ `line <- lm(column2 ~ column1)`
- ❖ `abline(line)`

R libraries

- ❖ R libraries are called “packages”. They live here: <http://cran.r-project.org/>
- ❖ To install package x, start R and type:
 - ❖ `install.packages('x')`
- ❖ To get information about a package, type:
 - ❖ `help(package = 'x')`
- ❖ You’ll probably want these packages:
 - ❖ `ggplot2`
 - ❖ `plyr`
 - ❖ `xlsx`

Using ggplot2

- ❖ `refugees = c(23,175,15543,11,338796,1244,63,53)`
- ❖ `asylumseekers = c(140,77,2872,6,255,91,6,7)`
- ❖ `origins = c('Burundi','Dem. Rep. of the Congo', 'Eritrea', 'Rwanda', 'Somalia','South Sudan','Sudan','Uganda')`
- ❖ `countries <- factor(origins)`
- ❖ `library(ggplot2)`
- ❖ `qplot(refugees, asylumseekers, color = countries)`



Continuing your R Journey

- ❖ Websites:

- ❖ <http://www.cookbook-r.com/>

- ❖ <http://www.statmethods.net/>

- ❖ The community sites from slide 3

- ❖ Books:

- ❖ Paul Teetor, “The R cookbook”

- ❖ Examples in O’Neil’s “Doing data science”