

DATA, Exploratory Data Analysis, and R

Introduction to Data Visualization

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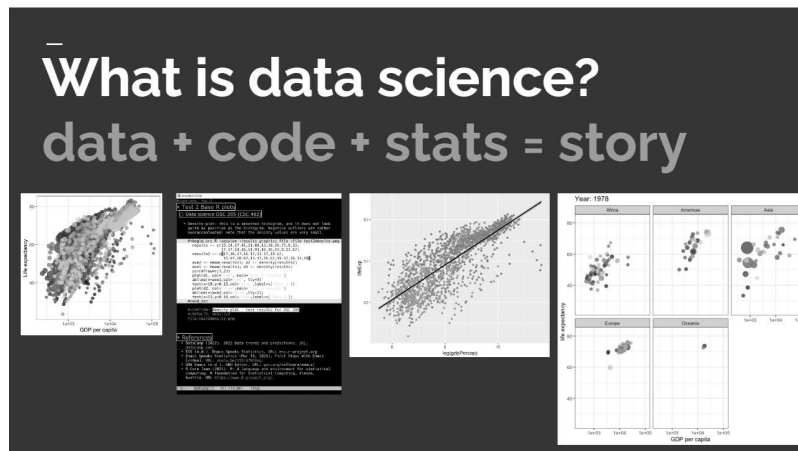


Figure 1: data science pipeline

- Why do we analyze data?
- Data as a concept and as a practice

- The importance of metadata
- Exploratory Data Analysis
- Numbers vs. graphs
- Data analysis workflow
- Why R?
- R package management
- Download R practice file

1 WHY DO WE ANALYZE DATA?

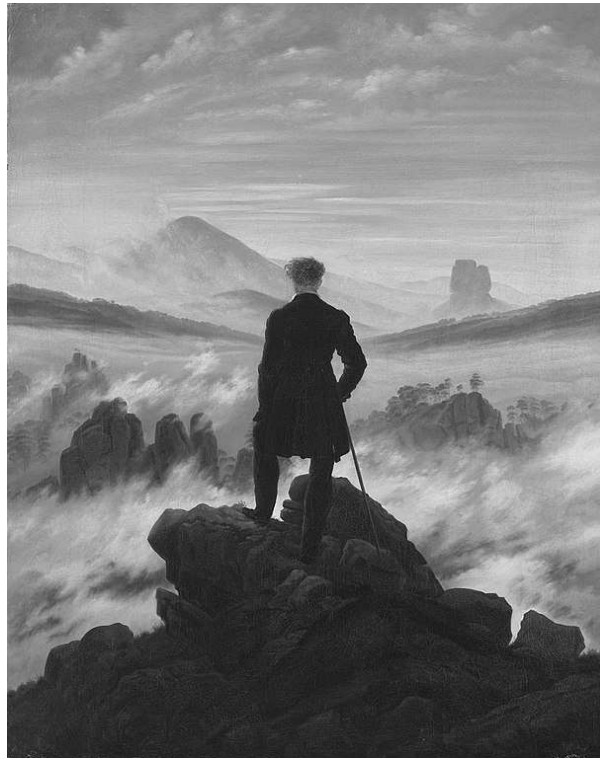


Figure 2: data science pipeline

- Well, what do you think?

Tip: reframe WHY questions as WHAT questions -

- What data?
- What analysis?
- What benefits?

Pearson:

1. to **understand** what has happened or what is happening;
2. to **predict** what is likely to happen, either in the future or in other circumstances we haven't seen;
3. to **guide** us in making decisions.

2 DATA - CONCEPT

- An **entity**, e.g.
 - family history of patient in medical study
 - competing company characteristics in marketing analysis
- An **event**, e.g.
 - demographic characteristics of voters
 - locations visited during a shopping visit
- A **process**, e.g.
 - manufacturing data from a production line
 - application information from a hiring process

3 DATA - PRACTICE

- Data structures = rectangular array of observed values
- Rows = observation of entity, event, or process
- Columns = recorded characteristic or attribute

```
## extract first six records from the mtcars data frame
head(mtcars)
```

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Mazda RX4	21.0	6	160	110	3.90	2.620	16.46	0	1	4	4
Mazda RX4 Wag	21.0	6	160	110	3.90	2.875	17.02	0	1	4	4
Datsun 710	22.8	4	108	93	3.85	2.320	18.61	1	1	4	1
Hornet 4 Drive	21.4	6	258	110	3.08	3.215	19.44	1	0	3	1
Hornet Sportabout	18.7	8	360	175	3.15	3.440	17.02	0	0	3	2
Valiant	18.1	6	225	105	2.76	3.460	20.22	1	0	3	1

- Data frame rows and columns have *names*
- Complete description with [help\(mtcars\)](#).
- Meta data supplement data frame content

4 META DATA



Figure 3: Greek goddess of peace and spring (Ειρήνη)

- "Data about data" (Greek μετά = 'after', 'beyond')
- Meta data in mtcars:
 - Original source of the data
 - Scientific paper analyzing the data
 - Description of the variables (columns)

What could be issues with metadata?

- **Completeness** - origin
- **Consistency** - logic, values, (time) dependency
- **Accuracy** - origin and validity

"As potentially valuable as metadata is, we cannot afford to accept it uncritically: we should always cross-check the metadata with the actual data values, with our intuition and prior understanding of the subject matter, and with other sources of information that may be available." (Pearson, 2018)

5 TODO PRACTICE: META OR NOT META?

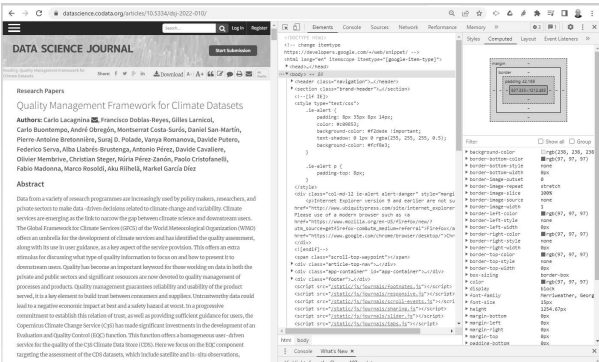


Figure 4: datascience.codata.org/articles/10.5334/dsj-2022-010/

Pair exercise: Identify the different types of data and metadata in the screenshot of an online journal article.

1. Article meta data: Journal title, "Research paper", title, authors
2. Layout meta data: HTML/CSS elements
3. Browser meta data: browser data (buttons for: download, font size, print, login, register, menu options; browser console; URL)
4. Article content data: abstract + paper text, tables and figures

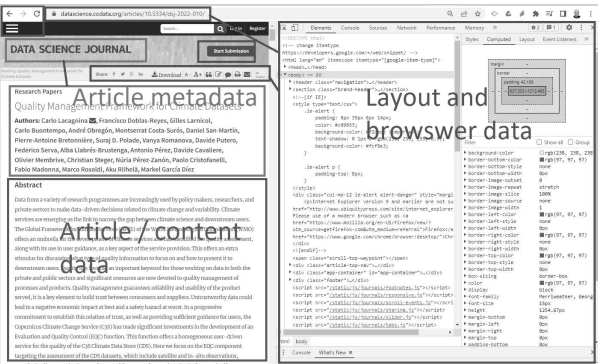


Figure 5: Solution

6 PROBLEM: MISSING VALUES

```
> library(MASS)
> data(package="MASS")
Data sets in package 'MASS':

Pima.te           Diabetes in Pima Indian Women
Pima.tr           Diabetes in Pima Indian Women
Pima.tr2          Diabetes in Pima Indian Women
```

Figure 6: Pima Indians data sets in the MASS package

Check out structure of Pima datasets:

```
str(Pima.te)
str(Pima.tr)
str(Pima.tr2)
```

- The MASS package contains three different versions of the Pima indians data set (diabetes in women of the Pima tribe)
- MASS metadata comments:

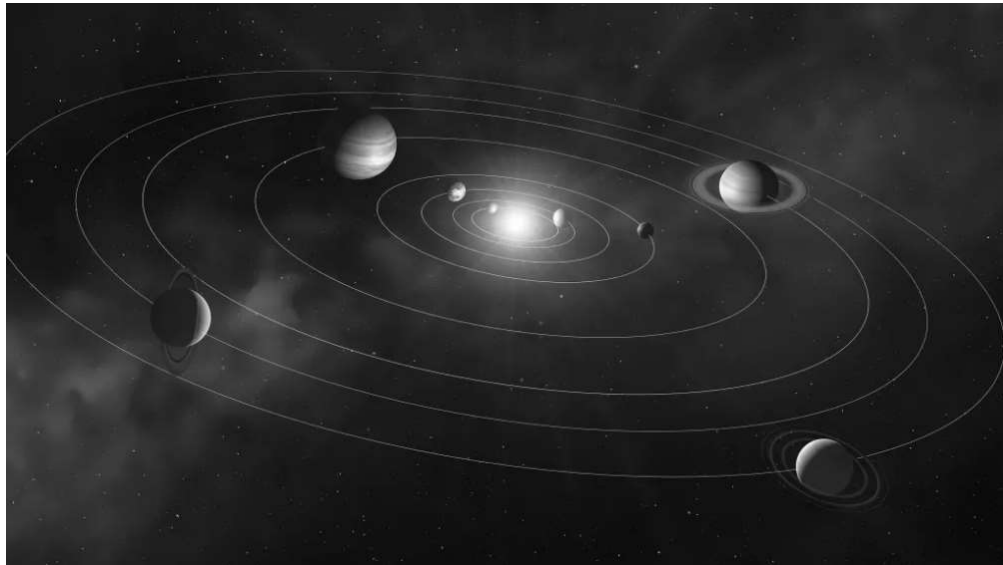
"The training set `Pima.tr` contains a randomly selected set of 200 subjects, and `Pima.te` contains the remaining 332 subjects. `Pima.tr2` contains `Pima.tr` plus 100 subjects with missing values in the explanatory variables."

- The kaggle.com database is yet another version: more records, one more variable - the "Metadata" information is missing
- Missing data are often coded as 0 instead of NA leading to errors:

"A number of studies characterizing *binary classifiers* have been published using [the Pima] dataset as a benchmark where the authors were not aware that data values were missing."
(Pearson, 2018)

7 PROBLEM: VARIABLE DEFINITIONS

- How many planets are there orbiting the sun?



- Definitions count: e.g. *planethood* (Weintraub, 2007)
 1. the object is too small to generate nuclear fusion energy
 2. the object is big enough to be spherical
 3. the object must have a primary orbit around a star
- Unrecognized disagreements in the definition of a variable are possible between those who *measure and record* it, and those who use data in *analysis*.
- Prominent examples: when does a patient die of COVID-19? What is the cause of death? When do two patients have the same disease?

8 EXPLORATORY DATA ANALYSIS (EDA)



"We look at *numbers* or *graphs* and try to find *patterns*. We pursue leads suggested by background information, imagination, patterns perceived, and experience with other data analyses." (Diaconis, 1985)

- Analysis is always based on exploring numbers
- Non-numerical data are converted to numbers: e.g. *categorical* variables are converted from discrete named values ("political party", "city") into counts or relative frequencies
- Each discrete value or category is also called a "level"

9 TYPES OF CATEGORICAL VARIABLES

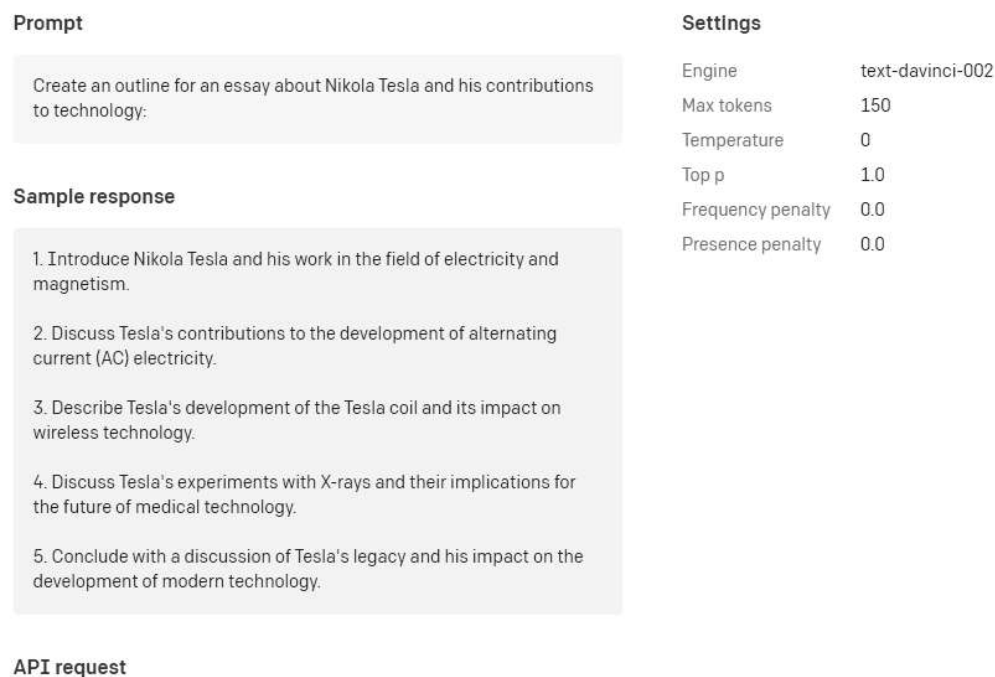


Figure 9: AI-generated outline for research topic (Source: OpenAI)

- Few levels (e.g. "Firm", "Party", "City")
- Many levels (e.g. US ZIP code with 40,000 levels)
- Exploitable sub-structure (e.g. text data¹)

10 SOME ISSUES WITH GRAPHS

- Humans are better at seeing patterns in graphs than numbers²

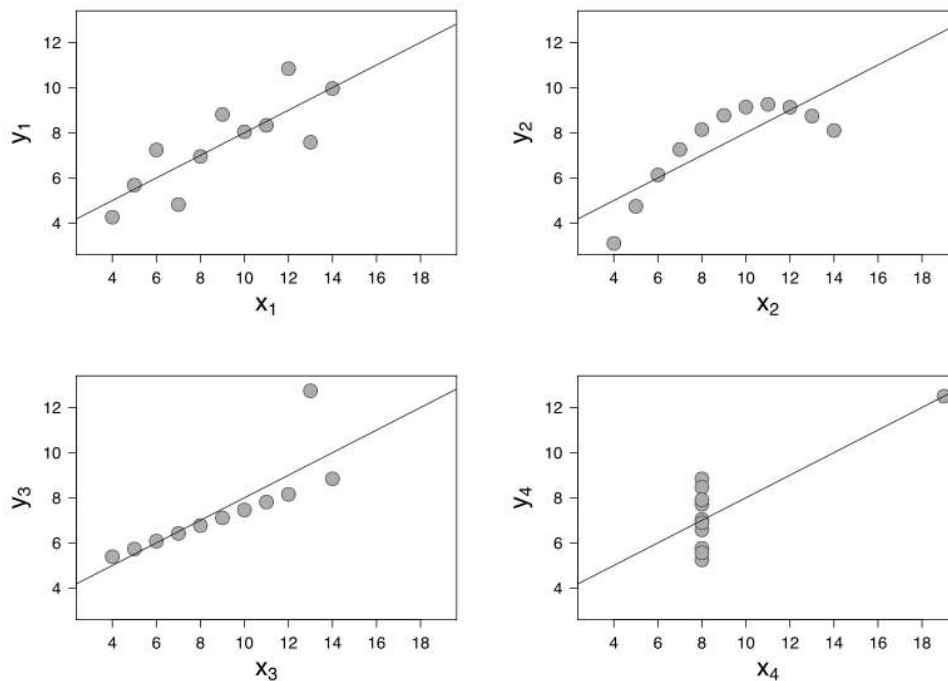


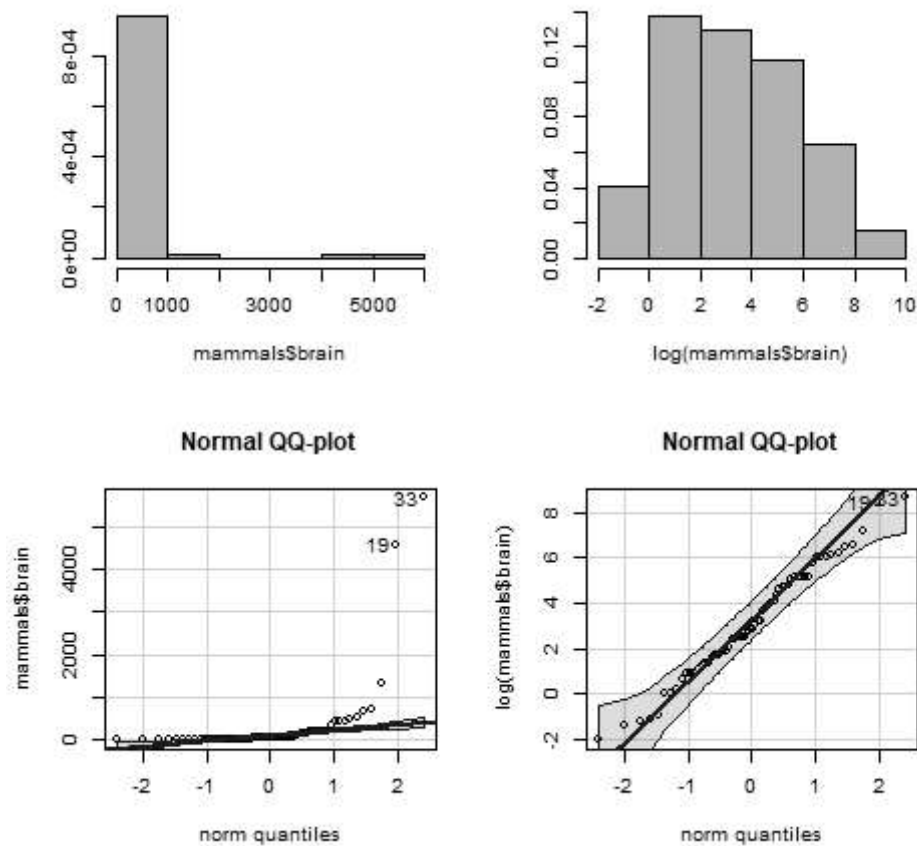
Figure 10: Anscombe dataset

- Use different graphs to explore and to explain - data mining is *exploratory*, data story telling is *explanatory*³
- Usefulness of a graph depends on **how data** are displayed, and strongly on **which data** are chosen to be displayed

11 TODO PRACTICE: RAW VS. TRANSFORMED GRAPH DATA

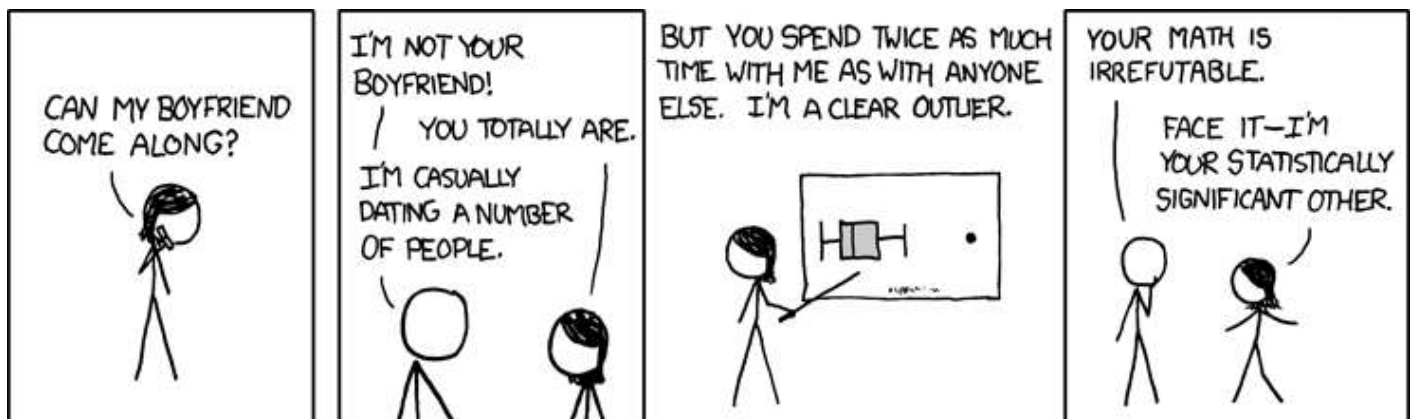
- The following two sets of plots are constructed from the `brain` element of the `mammals` dataset from the `MASS` package that lists body and brain weights for 62 different animals.
- **What do you think which graphs are more meaningful and why?**

```
library(MASS)
library(car)
par(mfrow=c(2,2))
truehist(mammals$brain)
truehist(log(mammals$brain))
qqPlot(mammals$brain)
title("Normal QQ-plot")
qqPlot(log(mammals$brain))
title("Normal QQ-plot")
```



- The plots are telling us something about the distribution of data values.
- The left-hand pair were generated from raw data values, the right-hand pair were generated from log-transformed data
- The right-hand pair suggests that the data exhibit a Gaussian (normal) distribution

12 R FOR EXPLORATORY ANALYSIS



- Exploratory analysis has more use for graphical tools
- R supports many different graphical displays and plot types
- Important focus: searching for anomalies and outliers in the data

13 DATA ANALYSIS WORKFLOW

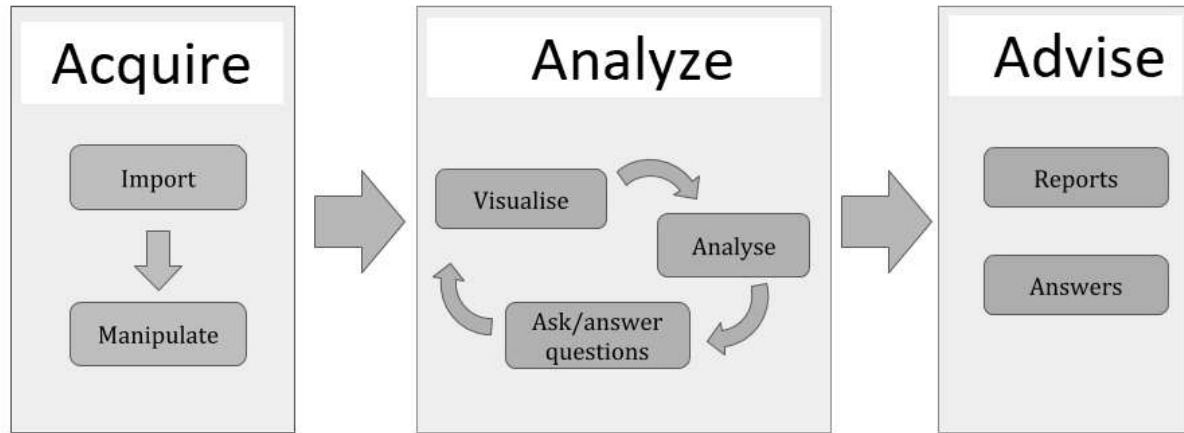


Figure 13: Data analysis workflow (emanuelaf.github.io - modified)

1. **Acquire**: make data available to the software
2. **Analyse**: perform the analysis
3. **Advise**: make analysis results available to those who need them
 - In training, the emphasis is often on (2) analysis, and pre-loaded, small, clean datasets and well-tested packages are used.
 - On the job, the emphasis is on (1) acquisition, and much time is spent importing and readying the data for analysis
 - In business, the main interest is (3) advice, hence the shift to storytelling and interpretation

14 COMPUTERS

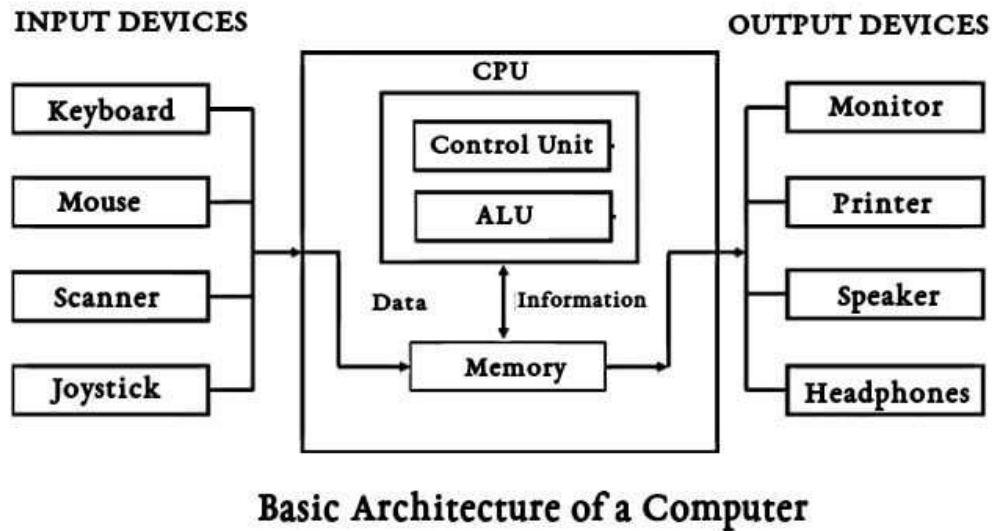


Figure 14: Von Neumann computer architecture (PSC Arivukal, 2020)

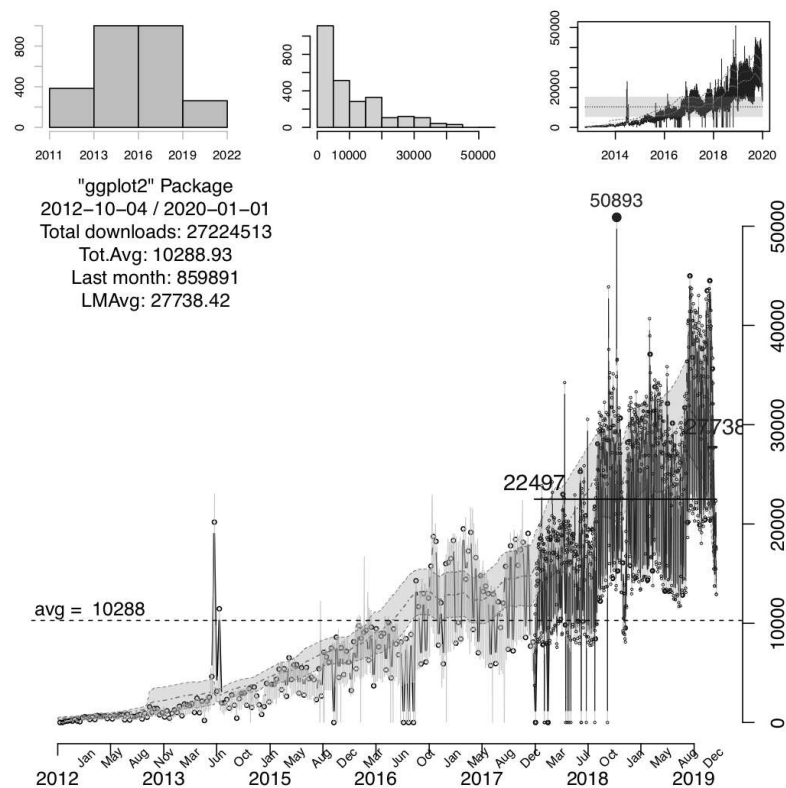
- RAM is several orders of magnitude faster than NVM
- Most R functions require raw data and results to fit in RAM
- OS and Internet impose infrastructure constraints⁴

15 WHY R?



- R is FOSS (Free Open Source Software) available for all OS
- Supported range of analysis methods ready for use
- Unix-style package and version control system
- Diverse, active community of users and developers

16 THE STRUCTURE OF R



1. Set of *base R packages* for basic statistics, data analysis, graphics
2. Set of *recommended packages* included in installations (like MASS)
3. Set of *optional add-on packages* for special purposes

17 INSTALLATION AND LOADING R PACKAGES

18 OPTIONAL INSTALLATION IN THE RGUI

- Start the Rgui from the CMD line terminal
- The Rgui includes a command line and graphics
- The RTerm or R program is a console only
- In the R GUI, find the tab "Packages"
- Set CRAN mirror site (closest to you)
- Install or update package from list

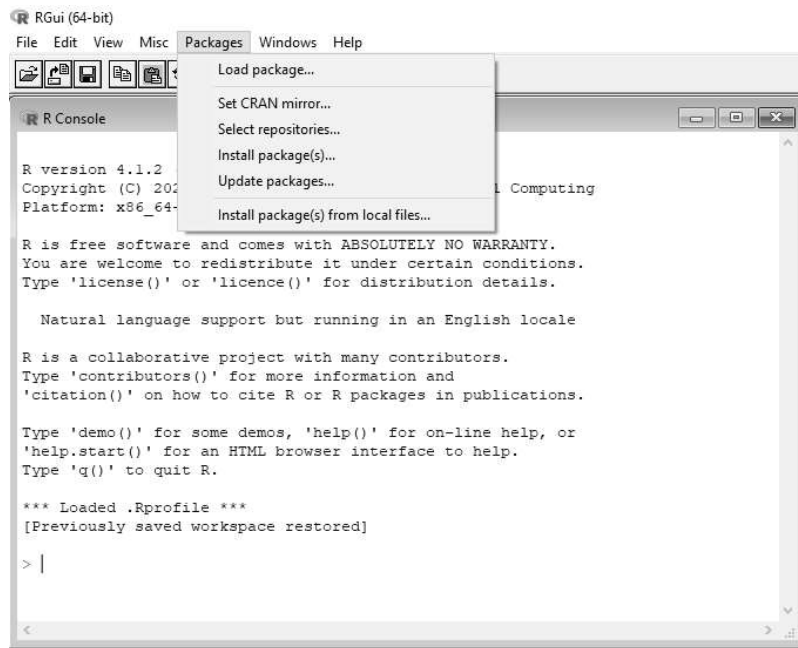


Figure 17: Package management in the Rgui program

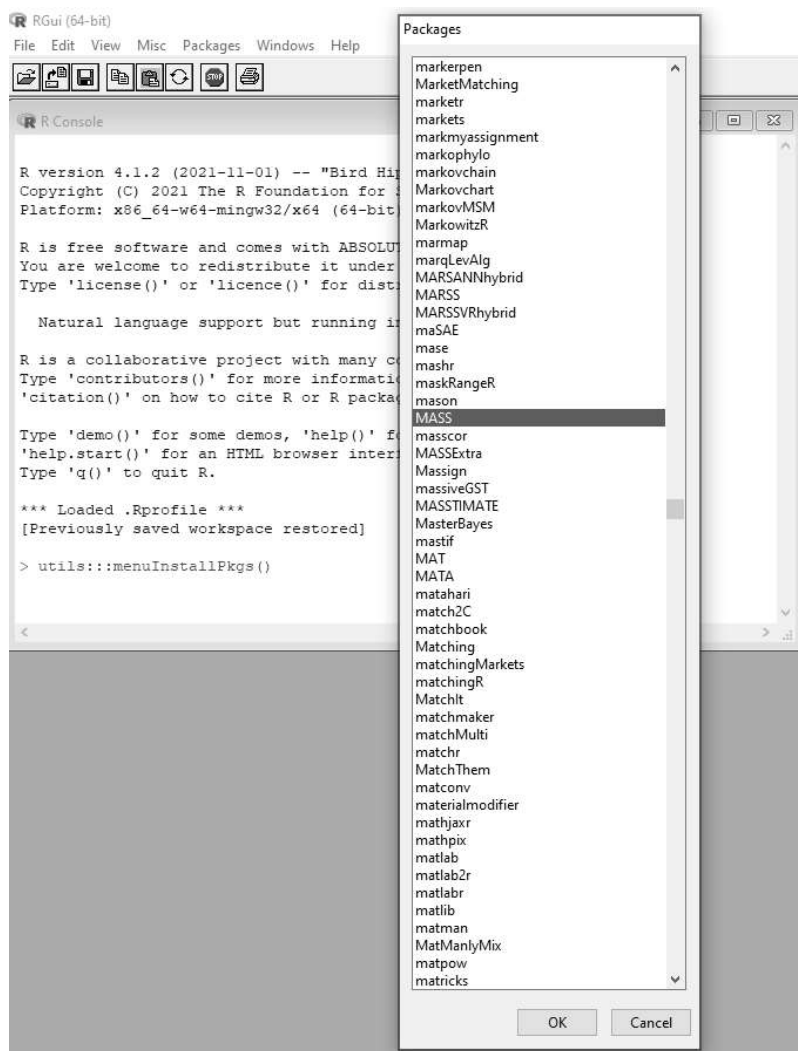


Figure 18: Package management in the Rgui program

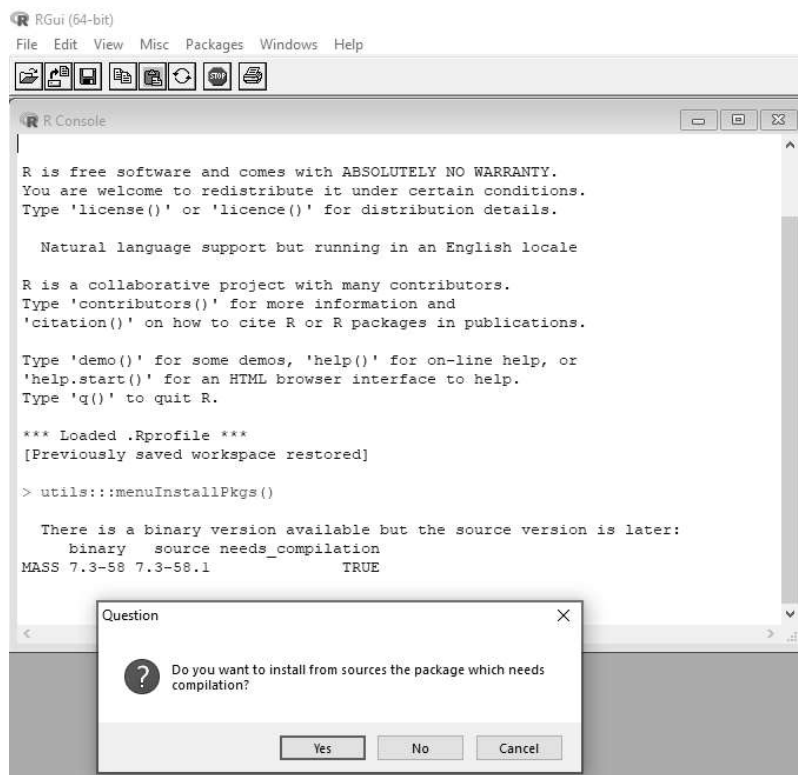


Figure 19: Package management in the Rgui program

19 QUESTIONS TO ASK FROM DATA

1. Where does the dataset come from, and how is it documented?
2. How many records (rows) does this dataset contain?
3. How many fields (variables, columns) are included in each record?
4. What kinds of variables are these (e.g. numerical, categorical)
5. Are there missing values?
6. If there are missing values: are these variables always observed?
7. If there are missing values: how are they represented?
8. Are the variables included in the dataset the ones we expect?
9. Are the variable values consistent with what we expect?
10. Do the variables exhibit the relationships we expect?

20 TODO PRACTICE: A REPRESENTATIVE R SESSION



1. Open the course directory in GitHub, <https://github.com/birkenkrahe/dviz>
2. Open /org/2_practice.org
3. Open the raw version of the file
4. Save file as 2_practice.org
5. Right click on the file in Explorer
6. Change Opens with: property to Emacs
7. Open file with Emacs from the Explorer

Summary:

21 CONCEPT SUMMARY

- Data are analysed to understand, predict, or guide decisions
- Data are entities, events or processes
- Meta data contain critical information for validation
- The data analysis workflow: acquire, analyze, advise
- R is FOSS, specialized on stats, and popular
- CRAN is the central hub for R package management

22 GLOSSARY

TERM	MEANING
Data frame	Rectangular array
Observation	Recorded event
Attribute	Characteristic
Meta data	Data about data
Data	Entity, event, process
Binary classifier	Attribute with 2 values
Missing value (NA)	Values that were not recorded
Categorical variable	Non-numerical, discrete
Level	Category, discrete value
Anomaly, outlier	Unusual data
CRAN	Comprehensive R Archive Network
Rgui	R console pgm with graphics
Rterm	R console (terminal) pgm only

23 REFERENCES

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Footnotes:

¹ Text data can be normalized (reduced - e.g. parsed into words, eliminating common words like "and", "of" and punctuation marks), and converted to numbers. The numbers are analyzed mathematically, and the result is transformed back to allow interpretation of the original text data. This technique leads to impressive NLP feats (so-called [transformer ML models](#) based on massive mined data sets, like [GPT-3](#).)

² The plots show Anscombe's quartet - four scatterplots which despite having different numerical values all have identical mean, variance, and standard correlation (Source: [revolutionanalytics.com](https://blog.revolutionanalytics.com)).

³ This difference goes deeper than data science: explanatory research is usually confirmatory (of some theory), while exploratory research is used to construct, or build, theory. Personal note: All of my own research has been exploratory.

⁴ Though they can also be enablers of education: e.g. Linux and the command line shell as a data science tool, and online REPL installations (usually Docker containers) as training grounds.

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Created: 2022-08-07 Sun 16:47