

FIT1043 Introduction to Data Science

Week 3

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Week 2 Coverage Overview of Data Science Python for Data Science



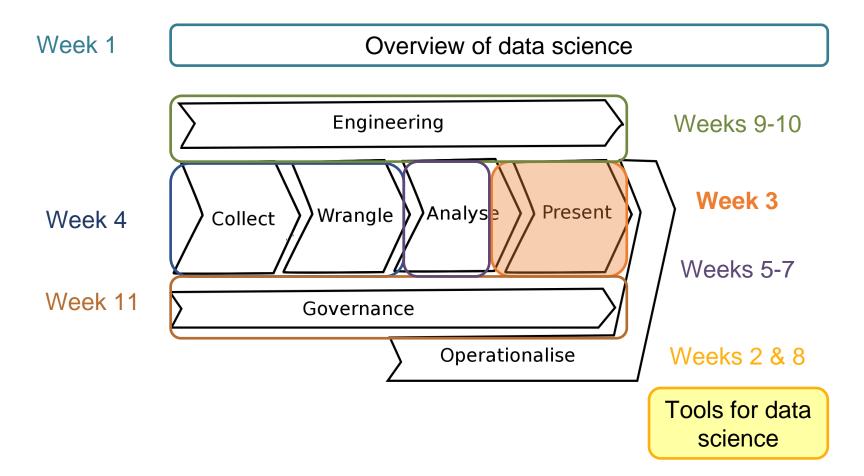


Week	Activities	Assignments
1	Overview of data science	
2	Introduction to Python for data science	
3	Data visualisation and descriptive statistics	
4	Data sources and data wrangling	
5	Data analysis theory	Assignment 1
6	Regression analysis	
7	Classification and clustering	
8	Introduction to R for data science	
9	Characterising data and "big" data	Assignment 2
10	Big data processing	
11	Issues in data management	
12	Industry guest lecture (tentative)	Assignment 3

Weekly Quiz From Week 2-11

- The quiz will open for 48 hours and you are allowed to have 2 attempts







Week 3 Outline

Introduction to Python for Data Science

Advanced Aggregation in Python

Data Visualisation

- Introduction and Basic Visualisation Plots
- Descriptive Statistics



Learning Outcomes

Week 3

By the end of this week you should be able to:

- Comprehend more sophisticated group-by operations and graphing in Python
- Comprehend the power/importance of data visualisation
- Differentiate between **approaches for data visualisation**, and explain where each approach is appropriate to be used
- Explain/differentiate different concepts in descriptive statistics



Turning Powerful Statistics into Art





https://www.visualcapitalist.com/infection-trajectory-flattening-the-covid19-curve/ **Visualization vs Pure Data** Country, Total Total Total Active Serious, Tot Cases/ Tot Deaths/ Other Cases Deaths Deaths Recovered 11 Cases Critical 1M pop 1M pop Cases 12 208 21 4 183 6,130 619 San Marino 0,000 Vatican City 4,994 132 38 94 2,701 Faeroe Islands 188 1 186 2,433 13 Andorra 737 56 Iceland 2 679 11 2,160 1,333 6 2,129 13 Luxembourg 1,319 1,000 1,338 51 51 Liechtenstein 153 1,259 18 10,897 131 10,613 141 Switzerland <u>ltaly</u> 74,386 7,503 9,362 57,521 3,489 1,230 124 49,515 3,647 5,367 40,501 3.166 1,059 78 Spain 30 790 31 Monaco https://www.worldometers.info/coronavirus/ 10 20 30 50

Days since the 100th confirmed case



Data Visualisation

From Introduction to Probability and Statistics for Engineers and Scientists, by S. M. Ross

From the previous slide, what do you think that was the purpose of the visualisation?

"... data visualisation is useful as a preliminary form of data analysis to get a "feel" for the data ..."



Data Visualisation

It is often useful to visualise data

- Can sometimes quickly reveal patterns
- However, going beyond two dimensions is problematic

Piece of paper or monitor now only limits us to 2 dimensional data (but with clever tricks, we can visualize up to 6 or 7 dimensions on a flat screen. We aren't going to cover that in this course but you get to see 3 dimensional data in your lab (by using colour).



Basic Types of Data

Numeric-Discrete

- Numeric, but the values are enumerable
- E.g., Number of live births, Age (in whole years)

Numeric-Continuous

- Numeric, not enumerable (i.e., real numbers)
- E.g., Weight, Height, Distance from CBD



Basic Types of Data

Categorical-Nominal

- Discrete numbers of values, no inherent ordering
- E.g., country, state, gender

Categorical-Ordinal

- Discrete number of states, but with an ordering
- E.g., Education status, State of disease progression



Data Visualisation

Visualise data to quickly reveal patterns

For numeric data (continuous and discrete), we can use:

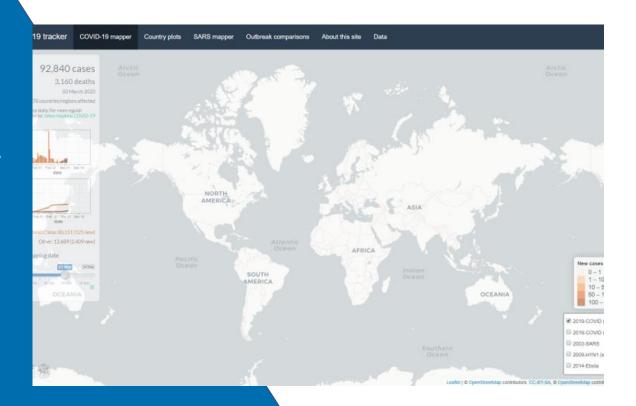
- Histograms
- Box plots (will revisit this in Descriptive Statistics)
- Motion charts

For categorical data, standard visualisations include:

- Frequency tables
- Bar graphs
- Pie charts



Visualizing Categorial and Numeric Data





Frequency Tables

(Not a Graph ... obviously)

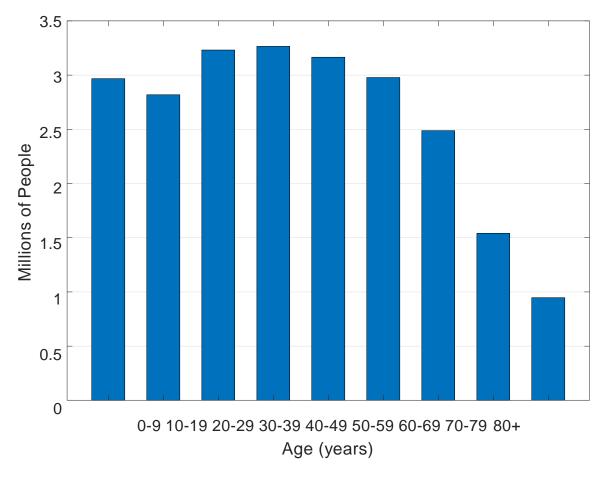
Age (years)	Number of People
0-9	2,967,425
10-19	2,818,778
20-29	3,231,395
30-39	3,265,526
40-49	3,164,712
50-59	2,977,883
60-69	2,488,396
70-79	1,540,373
80+	947,411

Australian Population by Age (2016 Census)

Frequency table is a **chart** that summarizes **values** and their **frequency**. It's a useful way to **organize** data if you have a list of numbers that represent the frequency of a certain outcome in a sample. Putting it into this form helps make it simpler to understand and further analyse.



Bar Charts



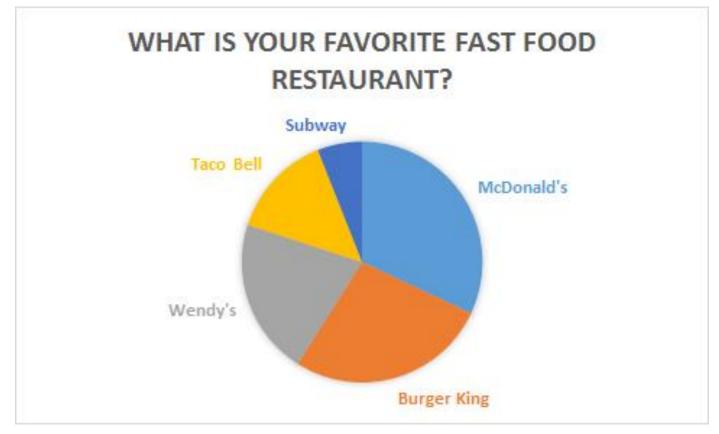
Australian Population by Age (2016 Census)

Compare between different groups, or Show changes over time

A bar diagram makes it easy to **compare** sets of **data between different groups at a glance**.



Pie Charts



Pie Chart is a type of graph in which a circle is divided into sectors that each represent a proportion of the whole

Note, usually, for 6 or less categories. When there are more categories, it is difficult for the eye to distinguish between the relative sizes of the different sectors and so the chart becomes difficult to interpret.



Histograms

Group **numeric** data into categories by putting into bins

If $y = (y_1, \dots, y_n)$ are our data points, we divide them into K equally spaced bins, i.e.,

The number of samples that fall in bin (category) **K** are

$$v_k = \#\{y_j \in (\min\{y\} + (K - 1)w, \min\{y\} + Kw)\}$$

Where
$$w = \frac{\max\{\mathbf{y}\} - \min\{\mathbf{y}\}}{K}$$

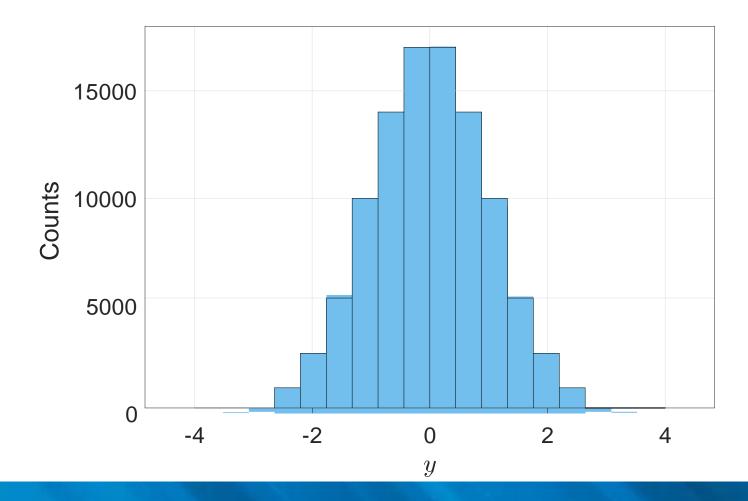
is the width of the bins

Note: In other words, it is a plot of (v_1, \ldots, v_K) using bar chart



Histogram

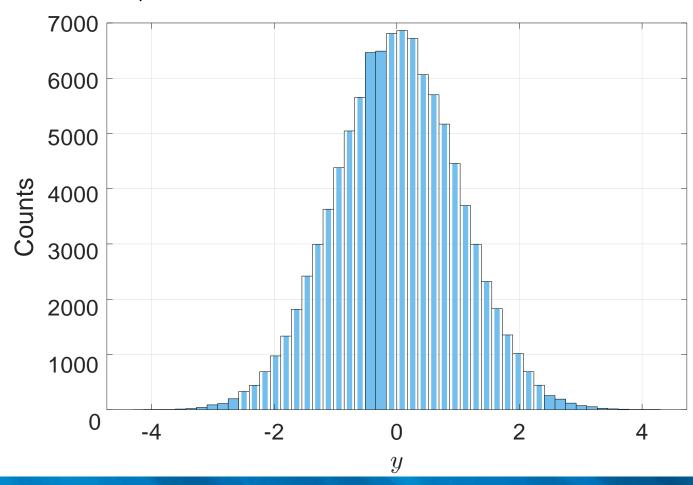
K = 20 bins





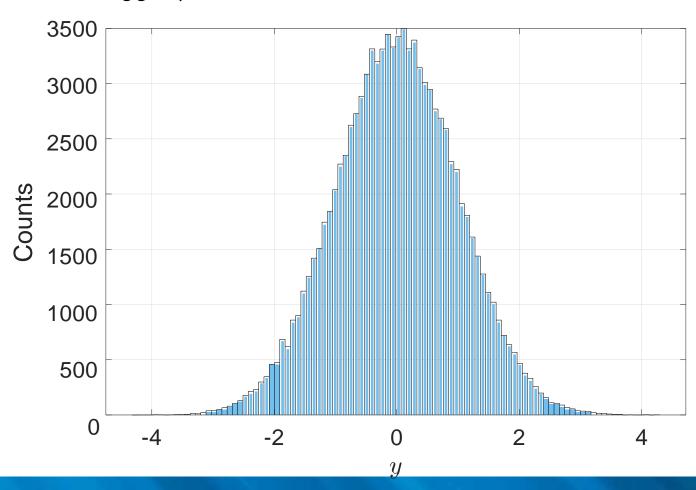
Histogram

K = 50 bins (looks smoother)



Histogram

K = 100 (starting to look ragged)



Motion Charts

Motivation

- Motion Charts are interactive multi-dimensional data visualisations
- Originally introduced to the world as GapMinder by Hans Rosling and made famous by his <u>TED talks</u>.

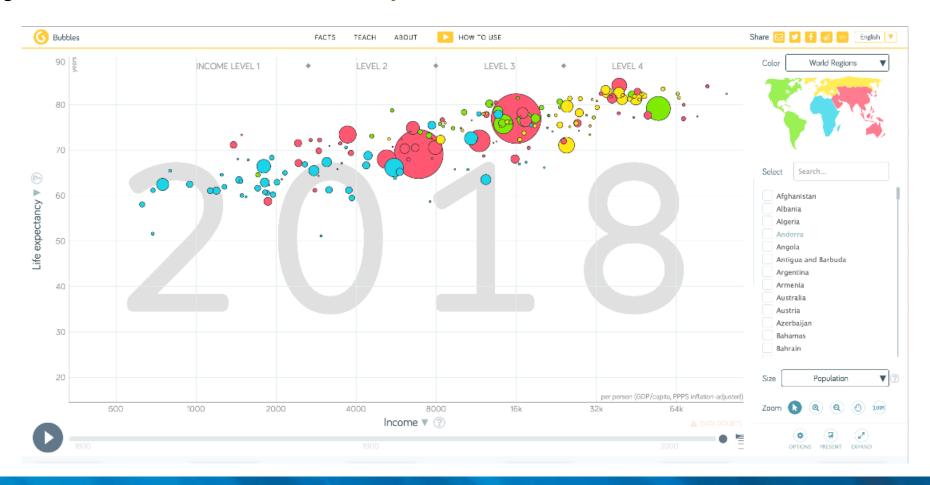
History

- The GapMinder technology was bought by Google and the name of motion charts changed to bubble charts
- But the <u>GapMinder website</u> is now up as a not-for-profit.



Motion Charts

Visualizing data in five dimensions: x-axis, y-axis, size of bubble, colour of bubble, and time.





Motion Charts

Advantages

- Time dimension allows deeper insights and observing trends
- Good for exploratory work
- Appeals to the brain at a more instinctual intuitive level

Disadvantages

- Not suited for static media
- Display can be overwhelming, and controls are complex
- Data scientists who branch into visualization must be aware of the limitations of uses

