



### **Deadlines are coming**

- Peer review of proposal and artefact on Monday
- Final deadline soon after! (check MyFalmouth)
- Make sure you go through the criteria in the Research Dissertation Handbook – don't forget to include anything that's required!



## **Ethics**



### Why is ethics important in research?

- Already discussed in detail earlier in the module (weeks 1 and 5)
- We must abide by Falmouth University's
   Research Integrity and Ethics Policy
- Participants in your research must give informed consent
- You must minimise potential risks arising from your research



### High risk categories

- Will your project involve clinical trials?
- Will your project involve the use of human blood or other human tissue?
- Will your project involve administering any drugs, placebos, food stuffs or drink to participants?
- Will your project involve the participation of NHS and/or Social Services staff, patients, equipment and/or facilities?
- Will your project involve participants who are particularly vulnerable? (e.g. refugees, prisoners, victims of violence)
- Will your project involve participants who are unable to give informed consent? (e.g. children, people with learning disabilities)
- Will your project risk causing psychological stress or anxiety or other harm or negative consequences beyond that normally encountered by the participants in their life outside research?

- Will your project involve actively deceiving the participants? (e.g., will participants be deliberately falsely informed, will information be withheld from them or will they be misled in such a way that they are likely to object or show unease when debriefed about the study)
- Will your project involve accessing and/or storing data that comes under the Official Secrets Act and/or poses a risk to National security?
- Is there potential for your project to have unintended harmful consequences (e.g. military use of technology / 'weaponisation' of artificial intelligence)?



### **Medium risk categories**

- Will your project involve participants?
- Will it be necessary for participants to take part in the study without their knowledge and consent at the time? (e.g. covert observation of people in non-public places)
- Will financial inducements (other than reasonable expenses and compensation for time) be offered to participants?
- Will your project involve collecting participant data (e.g. personal and/or sensitive data referring to a living individual)?

- Will your project involve accessing secondary data that is not in the public domain (e.g. personal data collected by another user)?
- Will your project involve accessing commercially sensitive information?
- Could your project have negative environmental impacts (e.g. disturbance of natural habitats; damage to, or contamination of, buildings/artefacts/wildlife)



#### Consent

- Participants in your research must give informed consent
- You must typically provide two documents
  - Information sheet
  - Consent form
- (In the case of an online study these could be electronic)
- See Appendices 5 and 6 of the Handbook for Research Integrity and Ethics (on LearningSpace)



#### **Information sheet**

- Should enable a potential participant to decide whether or not they want to take part
- Describe the project answer the question "why are you asking me to take part?"
- Explain what you are asking the participant to do
- Provide contact details for if participants have further questions



#### **Consent form**

- I confirm that I have read and understand the information provided above for the research study. I have had the opportunity to consider the information, ask questions and I have had these answered satisfactorily.
- I understand that my participation is voluntary and that I am free to withdraw at any time without giving any reason, without my legal rights being affected.
- I agree to take part in the above study.



#### **Personal data**

- Handling of personal data is covered by the General Data Protection Regulation (GDPR)
- Personal data = any data that relates to a specific living person
- If data is anonymous then it doesn't count as personal data – as long as it can't be de-anonymised



### Data handling best practices

- Only collect data that is necessary for the research
- Particularly, avoid collecting personal information (e.g. names, contact details) unless it's essential
- Ensure that data is stored and handled securely and GDPR-compliantly (e.g. on university network drives)



# **Testing**



### **Testing**

- One of the requirements for the dissertation is a plan to validate, verify and test your computing artefact
- Validation: are you building the right product?
- Verification: are you building the product right?



#### **Validation**

- Are you building the right product?
- What are the requirements?
- How will you determine whether the requirements have been met?



#### **Verification**

- Are you building the product right?
- Does your artefact work correctly?
- How will you verify this?
  - Unit tests?
  - Playtesting?
- If your artefact has a data collection component, how will you test this in particular?



# Data management



### Data management plan

- You must describe how you plan to collect, analyse and present data
- Basically demonstrate a pipeline from initial data collection to statistical analysis and presentation
- (In the next few slides I will refer to "R code" if using other tools e.g. Python then the principles still apply)



#### **Data collection**

- Advisable to collect data in a format that R can read, e.g. Comma Separated Values (CSV)
- Survey data can your survey tool export into an appropriate format?
- Paper surveys how will you transcribe them?
- Telemetry data how will the artefact output it and how will you collect it?



### **Data analysis**

- Once the data is in R, how will you analyse it?
- What statistical test(s)?
- Provide sample code for how these will be run



### **Data presentation**

- How will data be presented?
- Again provide R code



Illustrating Your Findings



- There are various techniques for reformatting and reducing data to make the analysis more interpretable or to illustrate a key point
- Graphical representations will also assist in decision making and reinforce the justification for those decisions --- e.g., has a hypothesis been falsified? To what extent is it clearly falsified?
- An overall picture of the data can be gleaned and initial conclusions drawn



- It is important to select the **most** effective ways to illustrate your findings in the dissertation
- Your communication skills are under assessment --- keep all graphical depiction meaningful to justifying your analysis and/or your intellectual decisions
- Provides an overall picture of the data underlying your findings to reach and support your conclusions
- Be wary of delegating charts solely to important data:
  - Depictions can distort message of original data
  - Concise, but often lacks precision
  - Ensure adequate support in body of text
  - Leverage explicit references (e.g., "as shown in Figure 1")



- There are many ways of creating graphs in R and Rstudio!
- We will use a library called ggplot2, which you may need to install and load
- > install.packages("ggplot2", dependencies=TRUE)
- > library(ggplot2)
- Among its functions should be a qplot(), which covers most of the common charts.



Common formats for presenting information include:

- Bar Chart
- Histogram
- Frequency Polygon & Ogive
- Pie Chart
- Scatter Plot
- Box Plot



#### **Do You Know Your Charts and Graphs?**

Which chart or graph is associated with which description?

Frequency distribution	A circular chart with slices presenting percentage breakdown
Histogram	A summary of data presented as classes and frequencies
Ogive	A two-dimensional graph of data from two variables
Pie chart	A cumulative frequency polygon
Scatter plot	A vertical bar chart presenting a frequency distribution



- A bar chart or bar graph is a chart or graph that presents categorical data with rectangular bars with heights or lengths proportional to the values that they represent.
- The bars can be plotted vertically or horizontally.
- Bar charts are useful for displaying data that are classified into nominal or ordinal categories in order to make comparisons.



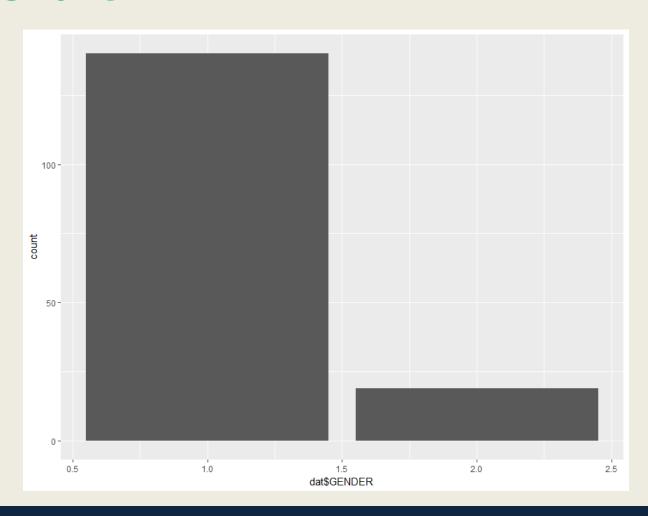
To a simple bar chart:

```
> qplot(dat$GENDER, geom="bar", stat="identity")
```

The **geom** argument refers to the type of chart that **qplot** will produce.

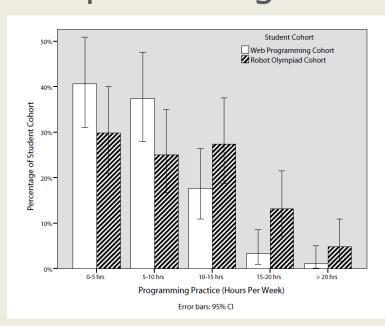
The **stat** argument is depreciated, but useful when no statistical analysis or summary statistic like a mean is needed. For a bar chart, "identify" defaults to a count.

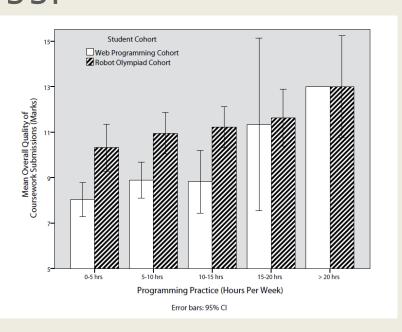






 Bar charts can be made much more complex using other ggplot functions:





http://www.cookbook-r.com/Graphs/Bar and line graphs (ggplot2)/



### Histogram

- A type of vertical bar chart used to depict a frequency distribution
- Construction steps:
  - Label the x axis with the class endpoints
  - Label the y axis with the frequencies
  - Label the chart with an appropriate title,
     i.e. not 'bar chart'
- A quick look at the histogram reveals which class intervals produce the highest frequency totals
  - E.g. which age group most often enrols in undergraduate computing courses?



### Histogram

To display a frequency table, examining birth years from 1975 to 2000 in 5-year intervals, use the following command:

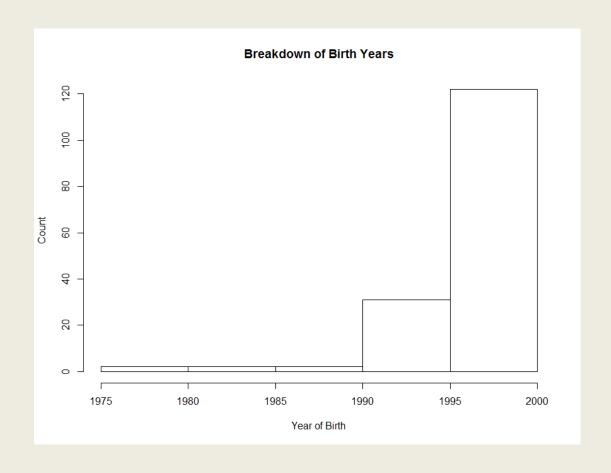
```
> summary(cut(dat$BIRTH_YEAR, c(1975,seq(1980,2000,5)),
include.lowest=T,right=FALSE))
```

To display a histogram based on this data:

```
> hist(dat$BIRTH_YEAR, xlab="Year of Birth",
ylab="Count", main="Breakdown of Birth Years", breaks=5)
```



### Histogram





### **Frequency Polygon**

- A graph in which line segments connecting the dots depict a frequency distribution
- Construction steps:
  - Label the x axis with the class endpoints
  - Label the y axis with the frequencies
  - Plot a dot for the frequency value at the midpoint of each class interval (different to Ogive)
  - Connect these dots with a line



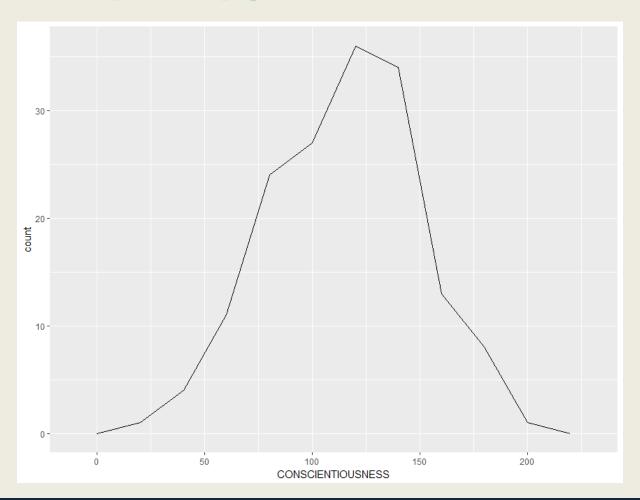
### **Frequency Polygon**

To display a histogram of students' conscientiousness, use the following command:

```
> ggplot(dat, aes(CONSCIENTIOUSNESS), stat="count") +
    geom freqpoly(binwidth = 20)
```



## **Frequency Polygon**





# **Ogive**

- A Cumulative Frequency (CF) polygon
- Construction steps:
  - Label the x axis with the class endpoints
  - Label the y axis with the cumulative frequencies
  - A dot of '0' is placed at the beginning of the first class
  - Mark a dot for the CF value at the end of each class interval
  - Connect these dots with a line



# **Ogive**

To construct an ogive, you need to format the data into cumulative frequencies:

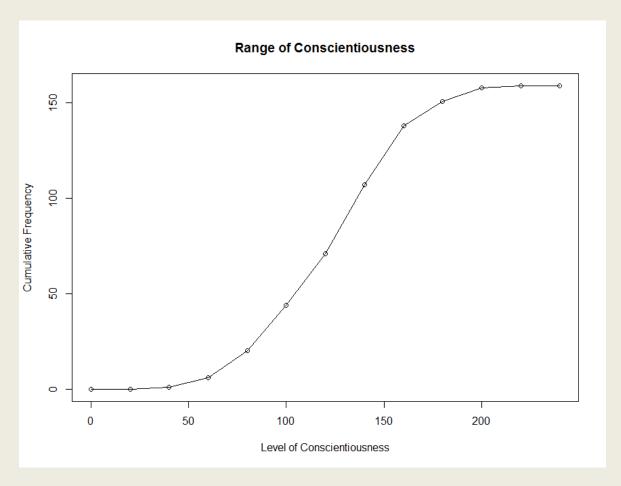
```
> cumfreq0 <- c(0, cumsum(table(cut(dat$CONSCIENTIOUSNESS,
seq(0, 240, by=20), right=FALSE))))
```

#### Then plot the chart based on this data:

```
> plot(seq(0, 240, by=20), cumfreq0, main="Range of
Conscientiousness", xlab="Level of Conscientiousness",
ylab="Cumulative Frequency") + lines(seq(0, 240, by=20),
cumfreq0)
```



# **Ogive**





- A circular depiction of data where the area of the whole pie = 100% of the data being studied. Slices represent a % breakdown of each of the values
- Business uses: e.g. for depicting budget categories, market share, time and resource allocation
- Generally more difficult to interpret the size of the slices compared to the bars in a histogram. But- usage of '%' can clarify slice size



#### Construction steps:

1. Convert each toothpaste brand amount to a proportion by dividing each individual amount by the total

E.g. 
$$102 / 200 = 0.51$$

2. Convert each proportion to degrees by multiplying by 360°

E.g. 
$$0.51 * 360^{\circ} = 183.6^{\circ}$$



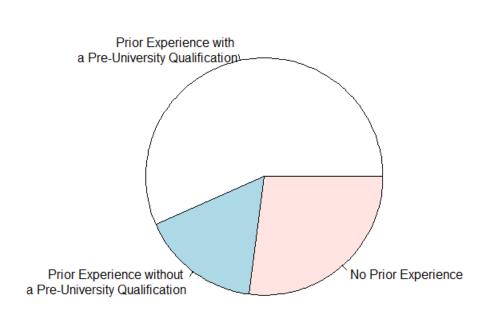
To construct a pie chart, you first need to label the categories:

```
> Lbls <- c("Prior Experience with \na Pre-University
Qualification", "Prior Experience without \na Pre-University
Qualification", "No Prior Experience")</pre>
```

Then plot the pie chart based on this data:

```
> pie(table(dat$PRIOR_EXP), labels = lbls)
```







#### Wrapping up commands into a single line:

> qplot(factor(dat\$PRIOR\_EXP, labels=c("Prior Experience with
\na Pre-University Qualification", "Prior Experience without \na
Pre-University Qualification", "No Prior Experience")),
dat\$ANXIETY, geom = "boxplot", main="Anxiety of Students from
Different Backgrounds", xlab="Background", ylab="Level of
Programming Anxiety")

Take note of the **factor()** command --- useful for distinguishing between different nominal characteristics or members of experimental and non-experimental groups!

The labels struct is setup in ascending order.



#### **Scatter Plot**

- Illustrates the relationship between two variables based on its underlying data points
- E.g. the link between neurotic personality traits and programming anxiety
- Scatter graph a two-dimensional graph plot of pairs of points from two variables
- Relationships will vary in strength, line of best fit used to indicate magnitude through slope



#### **Scatter Plot**

For the line of best fit, going to use the **rlm()** function for a robust linear model to mitigate the influence of dataset outliers:

> library(MASS)

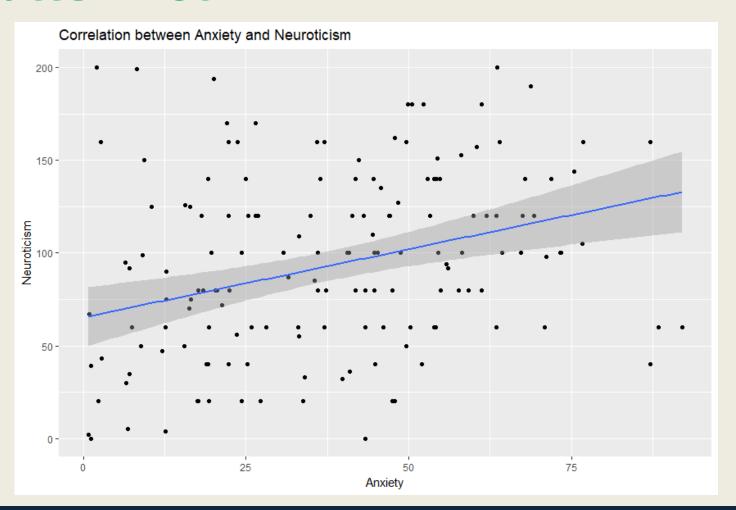
Then using the **qplot()** command with a combination of **geom** arguments including "point" (the scatter dots) and "smooth" (the line of best fit.

```
> qplot(dat$ANXIETY, dat$NEUROTICISM, geom = c("point",
"smooth"), method="rlm", main="Correlation between Anxiety and
Neuroticism", xlab="Anxiety", ylab="Neuroticism")
```

Try without the **method** argument. What happens?



## **Scatter Plot**





### **Box Plot**

- Illustrates proportions of a distribution, and useful for comparing groups
- Looking "down" on the distribution from the top, like viewing hills on a plane
- Lines outside the box represent range
- Box represents the lower and upper quartiles
- Line inside the box represents the median



### **Box Plot**

