Designing an Experiment (part 1)



Probability and Statistics

COMS10011

Dr. Anne Roudaut csxar@bristol.ac.uk

let's do an experiment!

memorization game

group 1

memorize as much as you can

group 2

if you beat group 1 = chocolate!

take a piece of paper and a pen

I will tell a list of numbers "1,2,3,6,write"

only when "write" -> write the list on paper

I will show the list 1, 2, 3, 6

if you are correct continue the game

if you wrong stop the game, remember best score

practice trials

1, 4, 9 (size=3)

practice trials

8, 7, 3, 5, 6, 1, 2 (size=7)

let's start the real experiment!

3, 2, 8 (size=3)

4, 2, 5, 1 (size=4)

7, 2, 5, 3, 1 (size=5)

6, 2, 9, 8, 5, 1 (size=6)

7, 4, 1, 8, 6, 3, 2 (size=7)

2, 7, 4, 9, 3, 1, 5, 9 (size=8)

1, 6, 7, 8, 5, 3, 1, 4, 6 (size=9)

6, 4, 1, 9, 3, 8, 2, 1, 7, 9 (size=10)

2, 7, 4, 1, 5, 7, 3, 8, 6, 4, 7 (size=11)

what is your best score (size of the list)?

enter it at

https://tinyurl.com/COMS10011

let's first look at the results



research question / hypothesis?



in(dependant) variables?



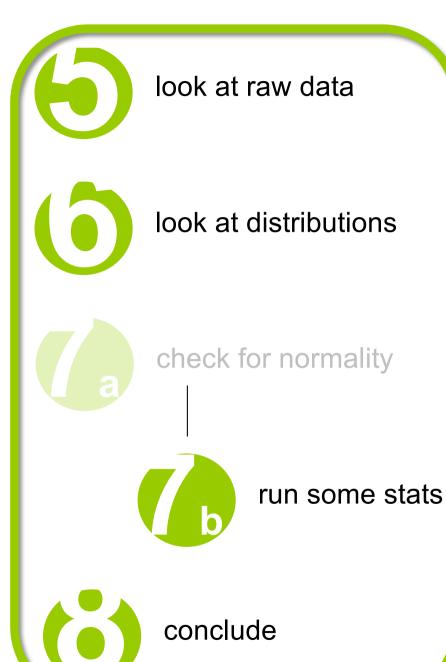
within or between subjects?



counterbalancing?



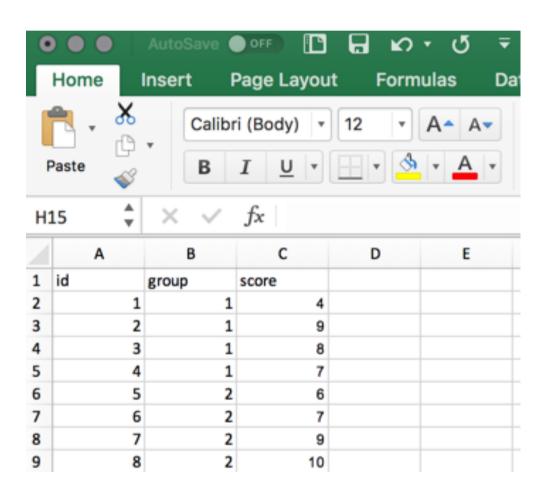
how many repetitions/trials?





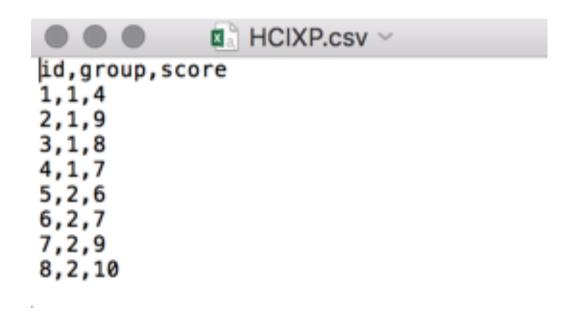
look at raw data

let's put everything in a table (excel is great for that)



save your file as a .csv (comma separated virgule is a format to store tables as text files)

you can open csv with excel, text file an many other software



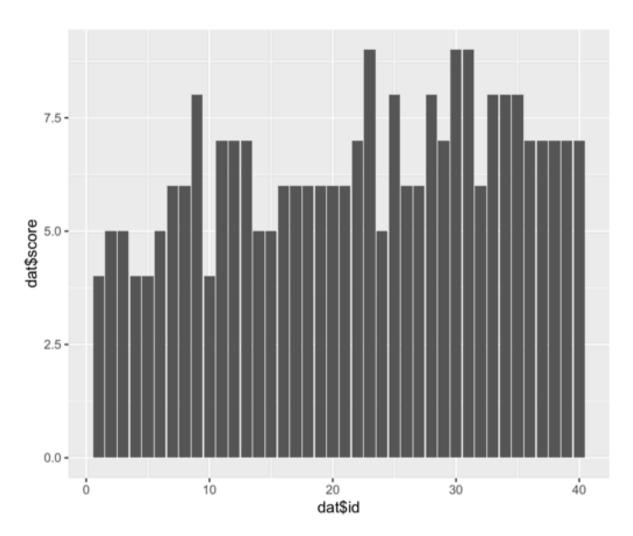


dat = read.csv("HCIXP.csv", header = TRUE)
print(dat) # look at the file in R



```
dat = read.csv("HCIXP.csv", header = TRUE)
print(dat) # look at the file in R
library(ggplot2)

ggplot(dat, aes(x = dat$id, y = dat$score)) +
geom_bar(stat = 'identity', position = 'dodge')
```

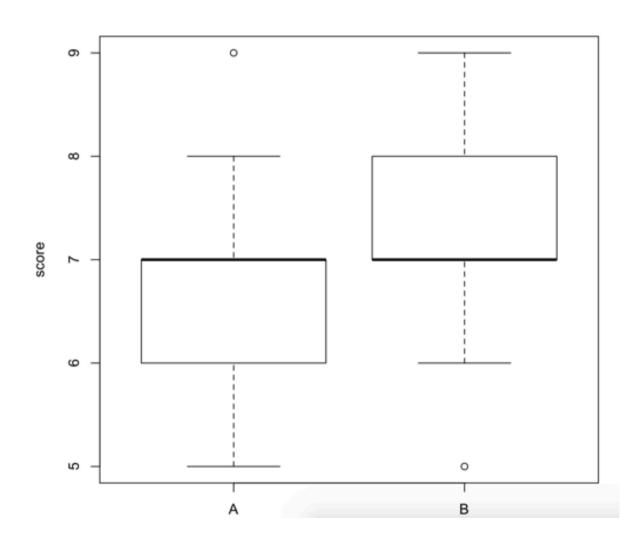


first: does the data look ok?

search for bugs, fatigue effect, learning effect or outliers (>3 times std) = remove / redo xp

plot(score ~ group, data = dat)



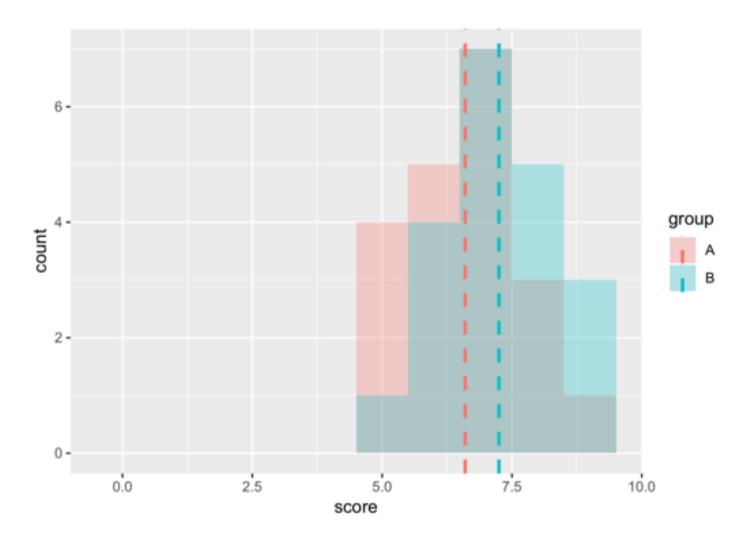




look at histograms

```
# Find the mean of each group
library(plyr)
cdat <- ddply(dat, "group", summarise,</pre>
score.mean=mean(score))
cdat
  group score.mean
   A 5.60
2 B 7.25
# Overlaid histograms with means
ggplot(dat, aes(x=score, fill=group)) +
geom histogram(binwidth=1, alpha=.3, position="identity")
+ geom vline(data=cdat, aes(xintercept=score.mean,
colour=group), linetype="dashed", size=1) +
```

expand limits(x = 0, y = 0)



your gut feeling: are these groups different?

are these distributions likely to have happen by chance?
... is this the results of the factor (chocolate)?



use a statistic test

```
# Use a t-test (two-tails, unpaired)
t.test(dat$score[dat$group == "A"], dat$score[dat$group
=="B"], alternative = "two.sided")
      Welch Two Sample t-test
      data: dat$score[dat$group == "A"] and
      dat$score[dat$group == "B"]
      t = -1.8185, df = 37.982, p-value = 0.07688
      alternative hypothesis: true difference in means is
      not equal to 0
      95 percent confidence interval:-
      1.37361001 0.07361001
      sample estimates: mean of x mean of
      y 6.60 7.25
```

"We could not find any significance differences!"

p-value = 0.07

is is enough to say that the two groups are different?

-> nope, not under significant level of 0.05

can we say that the two groups are same then?

-> nope, can only prove things are different, but not that they are the same



conclude

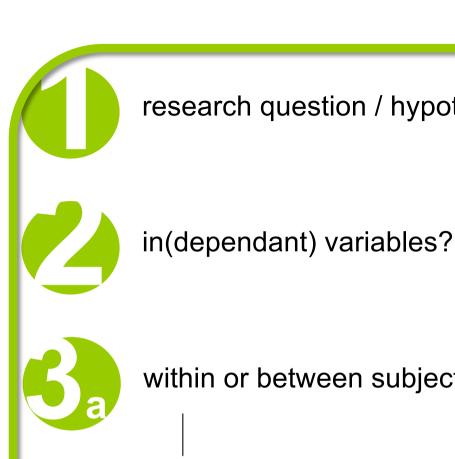
if p was lower than significance level we could say: "a student t-test showed significant difference between the two group (two-tailed t(46)=4.520, p < 0.005)"

otherwise:

"we did not find any significant results"

cannot conclude, no evidences to show that having chocolate rewards improve memorisation

let's go backward a little



research question / hypothesis?



look at raw data



look at distributions

within or between subjects?



counterbalancing?



check for normality



run some stats



how many repetitions/trials?



conclude



research question::

a statement that identifies a phenomenon to be studied

in our xp: I believe that rewards improve memorization skills

... suggested by <insert smart guess>

hypotheses::

statement of the predicted relationship between at least two experimental variables

provisional answer to a research question

in our xp: group chocolate will have a higher memorisation score than group with no reward



(in)dependent variable ::

the dependent variable is the event studied and expected to change whenever the independent variable is altered

vary A → make A an independent variable

so we want to show that A causes B

measure B → make B a dependent variable

in our xp?

independent variable = group type (nothing
vs. chocolate)

dependent variable = memorization score

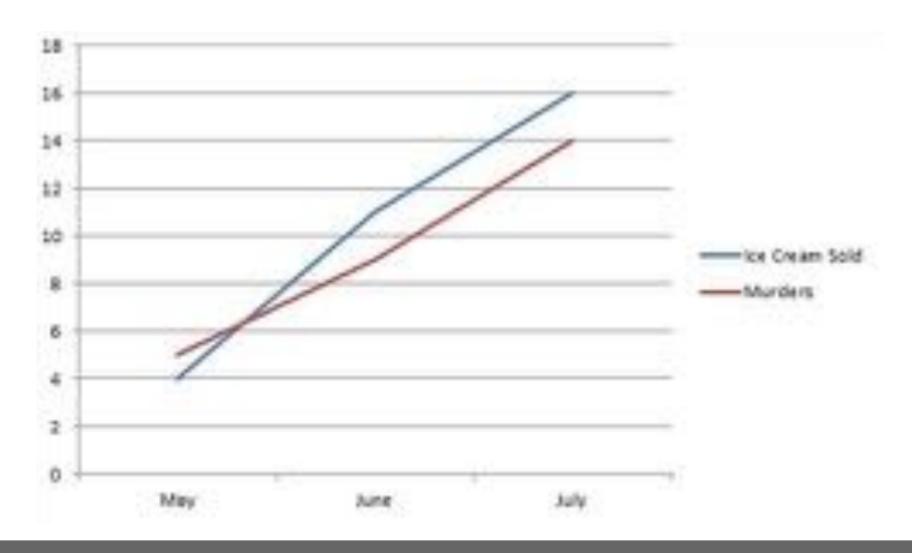
everything else should be a...

controlled variable ::

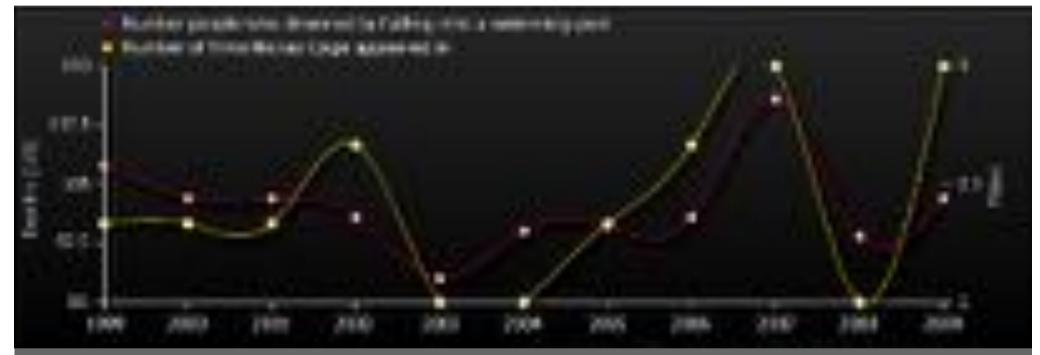
the variables that are kept constant to prevent their influence on the effect of the independent variable on the dependent avoid...

confounding variable ::

extraneous variables that correlates with both the dependent variable and the independent variable



ice cream consumption leads to murder counfounding: weather temperature



number of people drowned by falling into a swimming-pool correlates with number of films Nicolas Cage appeared in

this is not about correlation

this is about how to show causality, i.e., that some A causes some B

in our xp, do we have confounding variables?

yes, it is not greatly designed :s

gender, age, background, what you ate before, if you like chocolate or not, if you are competitive and want the others not to have chocolate, if some of the numbers are familiar to you etc.

what can we do about it?

- avoid them by controlling as much as you can in the environment
- if you cannot, make it an independent variable (e.g. gender)
- some are inherent *noise* (human individuality), use more participants to get *statistical power*

the goal of a quantitative study is to find a signal in a lot of noise

<u>experimental design:</u>

aims at maximizing your chances of finding the signal and not the noise

1. need to absolutely avoid systematic biases

(e.g., learning effect, fatigue). They give you false results!

2. avoid random noise. It makes your results nonsignificant. Clever experimental design is all about keeping the noise down e.g. in our xp, I made you practice before!



within vs. between?

within = all participants do same between = participants do only certain conditions



suffer less user variation

statistical power with less participants

no biases from other conditions (e.g. transfer of learning)

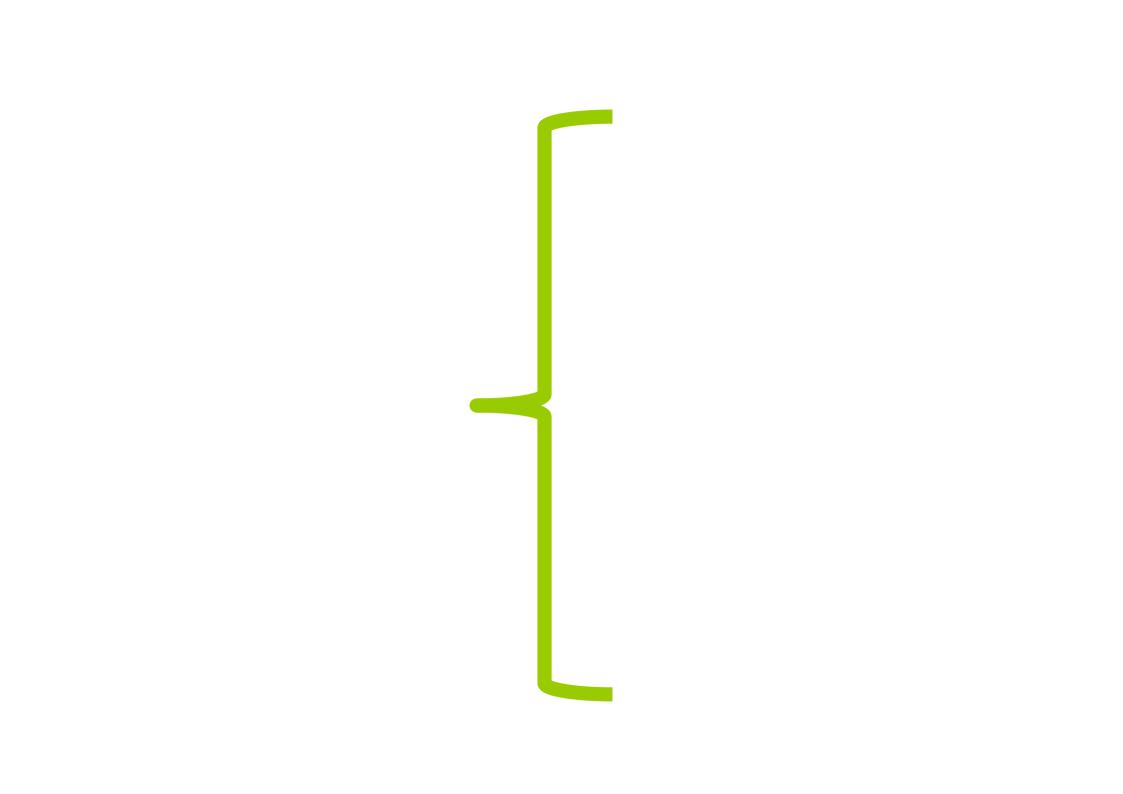
within vs. between?

within = all participants do same between = participants do only certain conditions in our xp, it had to be between subjects (because of the rewards)

participants did not do all conditions:

½ did the control condition

½ the reward condition







imagine a within subjects (test how fast we click an icon):

participants do all conditions: they start with the trackpad when finished they do the mouse

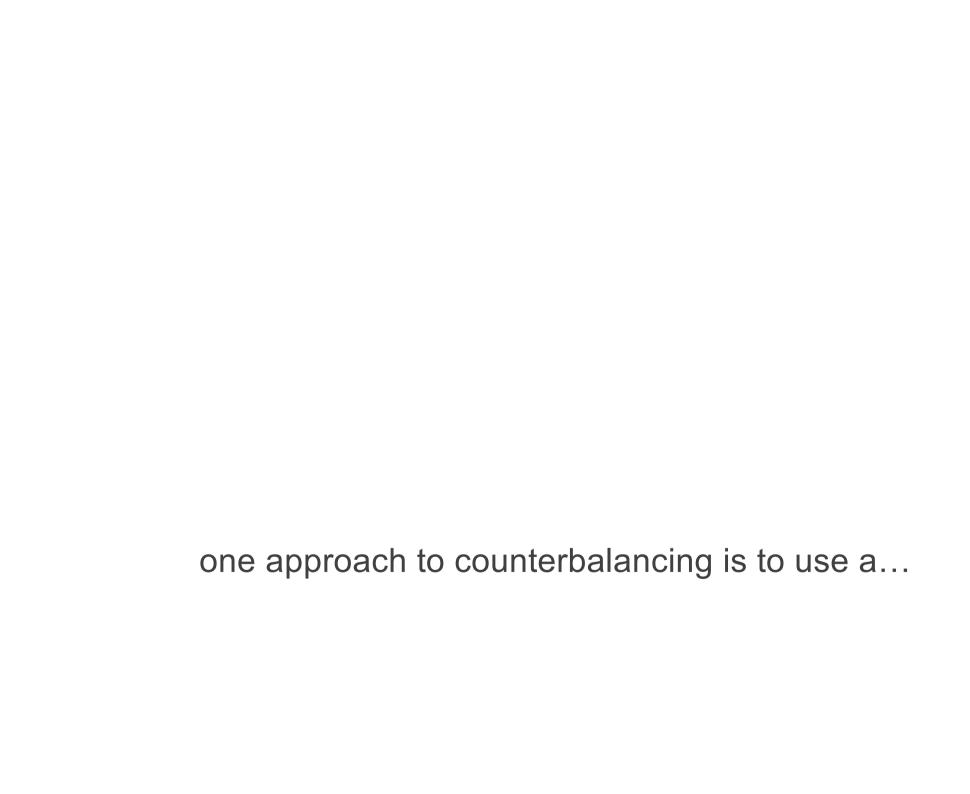
is it a good idea?

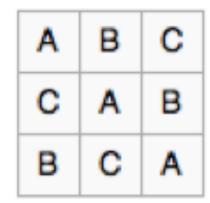
nope -> learning effect



counterbalancing ::

a method of avoiding confounding among variables presenting conditions in a different order

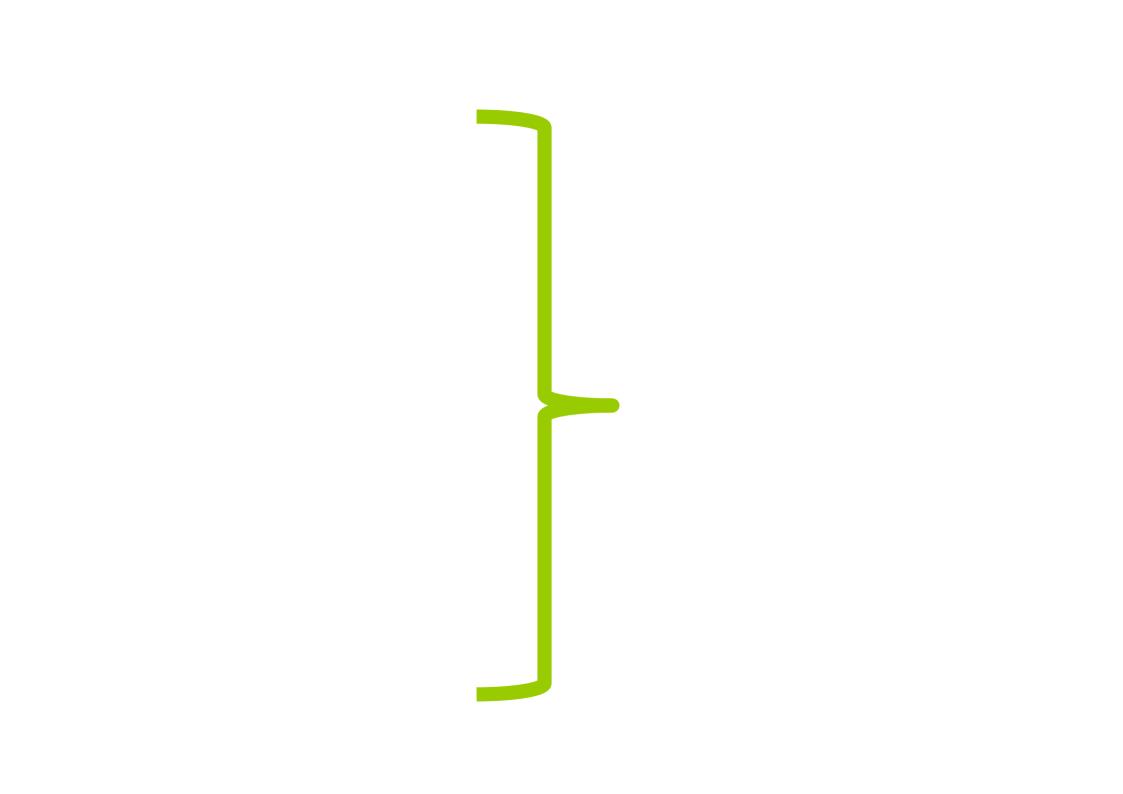




Latin square ::

an $n \times n$ array filled with n different Latin letters, each occurring exactly once in each row and exactly once in each column.







how many trials?

ideally make as much trials as you can to reduce noise but try to keep experiment around 30 min ... max 40 min

in our xp, we did only one trial because of time constraint, but should have done more to reduce noises

summary



research question / hypothesis?



look at raw data



in(dependant) variables?



look at distributions



within or between subjects?



we will see why check for normality



counterbalancing?



run some stats

so far we know t-test



how many repetitions/trials?



conclude

- 1. Explain the eight steps to design and analyze an experiment
- 2. Explain what is a within or between subject experiment
- 3. Explain what is a controlled variable or a confounding variable
- 4. Explain the difference between correlation and causality

take away

#