

ETC5242 Week 2

授课老师: Joe

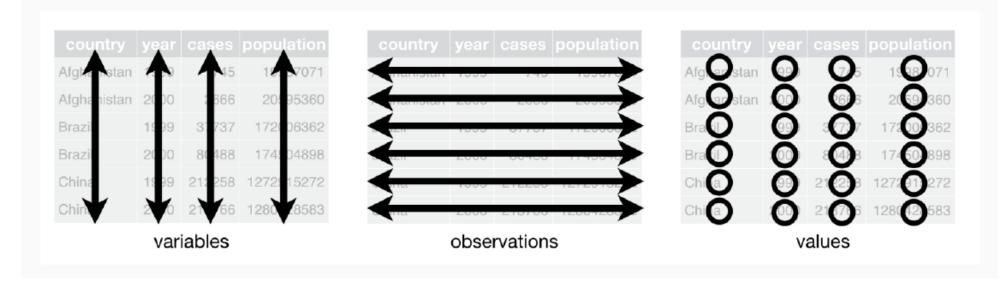


Recall the **tidyverse**: An **R** package, which is itself comprised of many other **R** packages

- ggplot2: data visualisation
- dplyr: data manipulation
- tidyr: data organisation
- readr: data import
- purrr: function iteration
- tibble: data storage
- stringr: string management
- forcats: categorial data functions

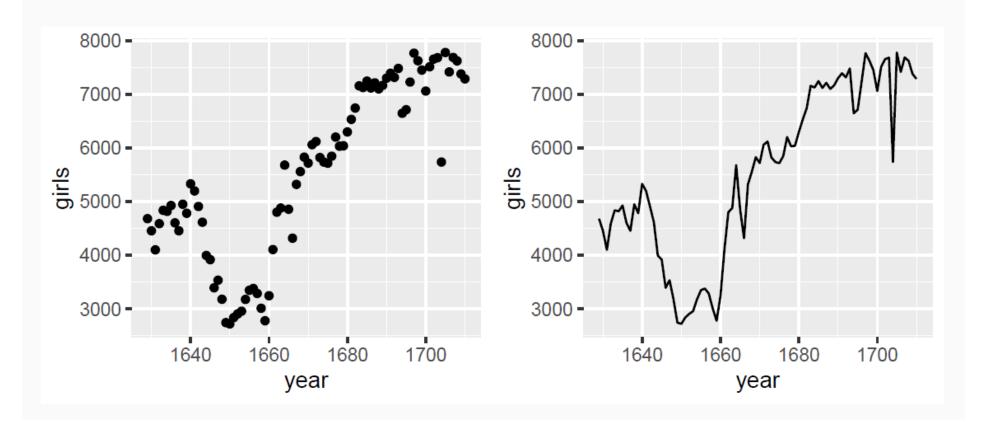


- Observations in rows
- Variables in columns
- Values in cells

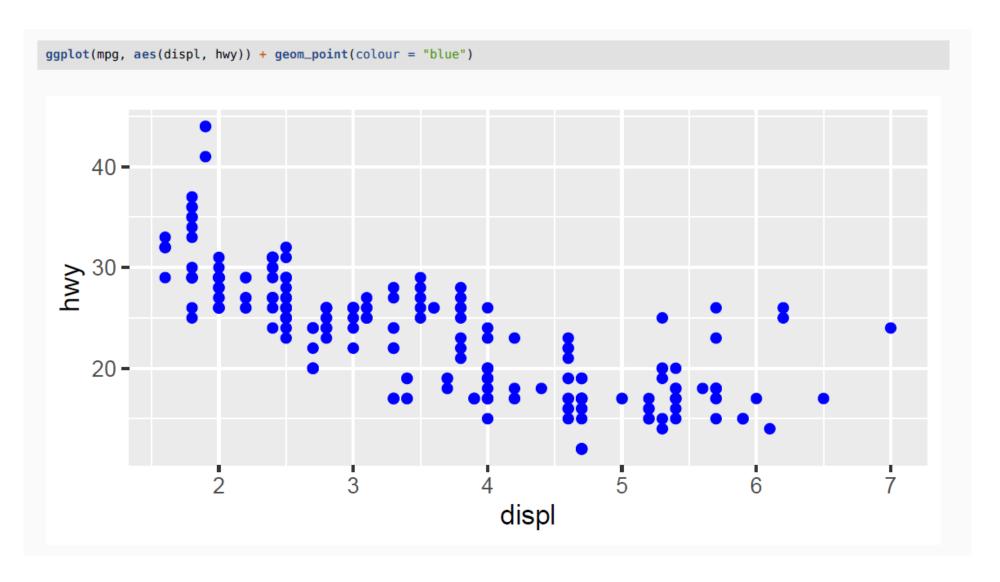




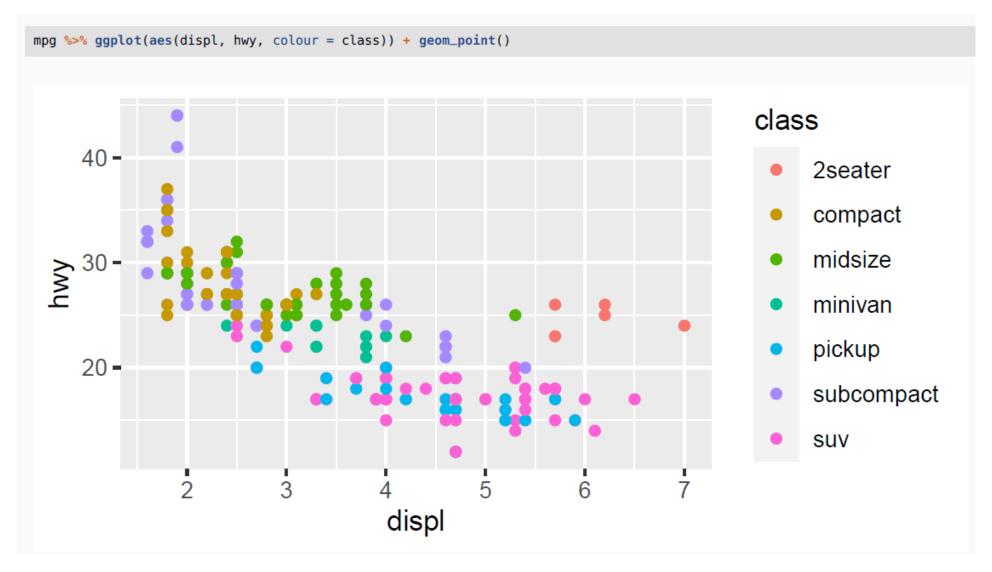
```
library(openintro)
library(gridExtra)
data(arbuthnot)
p1 <- ggplot(data = arbuthnot, aes(x = year, y = girls)) + geom_point()
p2 <- ggplot(data = arbuthnot, aes(x = year, y = girls)) + geom_line()
grid.arrange(p1, p2, ncol = 2)</pre>
```





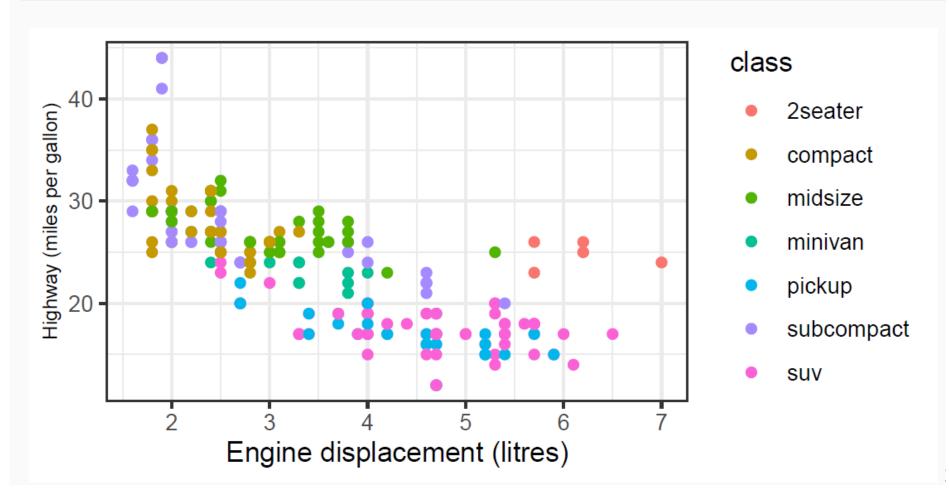








```
ggplot(mpg, aes(displ, hwy, colour = class)) + geom_point() +
    xlab("Engine displacement (litres)") + ylab("Highway (miles per gallon)") +
    theme_bw() + theme(axis.title.y = element_text(size = 8))
```





ETC5242 Week 3

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Hypothesis testing

Do males and females have the same chance of being promoted?

- Population proportion of **males** promoted: p_M
- Population proportion of **females** promoted: p_F
- Test Null hypothesis H_0 : $p_M = p_F$
 - (Gender has no effect on promotion decision)
- Against Alternative hypothesis H_1 : $p_M > p_F$
 - Gender has an effect on promotion decision, with a man more likely to be promoted than a woman

		decision		
		promoted	not promoted	Total
gender	$_{\mathrm{male}}$	21	3	24
gender	female	14	10	24
	Total	35	13	48

- Use the sample proportions as **point estimates** of the "true" p_M and p_F
 - ▶ **observed** $\hat{p}_M = \frac{\text{# males promoted}}{\text{# males considered for promotion}}$
 - **observed** $\hat{p}_F = \frac{\text{\# females promoted}}{\text{\# females considered for promotion}}$
- Do **observed** values satisfy $\hat{p}_M > \hat{p}_F$?
 - Equivalently, is $x_{obs} = \mathbf{observed} \, \hat{p}_M \hat{p}_F > 0$?
- We take x_{obs} is our **point estimate** for $p_M p_F$
- Could $x_{obs} > 0$ be due to "chance"?
- YES. Even when $p_M = p_F$ we can get $x_{obs} > 0$

- Restating the hypotheses of interest:
 - ► H_0 : $p_M p_F = 0$ (Null hypothesis)
 - ▶ H_1 : $p_M p_F > 0$ (Alternative hypothesis)
- Need the **decision rule** to decide whether to reject H_0
- We want to reject H_0 when x_{obs} is far from zero
 - ▶ zero is the value of the parameter (here $p_M p_F$) under H_0
- \Rightarrow Choose the **decision rule**:
 - ▶ Reject H_0 when $x_{obs} \ge x^{crit}$
 - Otherwise: Do not Reject H₀
- \blacksquare Here x^{crit} is the **critical value**
 - how is it set?



■ Critical value determined by desired to control **Type I error**

			Decision	
		Do not reject H₀		Reject H₀
Truth	H₀ true	no error		Type I Error
	H_1 true	Type II Error		no error

Table 1: Decision errors from an hypothesis test

- Fix $Pr(Type\ I\ error) = \alpha$, the significance level
 - choose α to be 'small' (e.g. $\alpha = 0.05$)

- "If H_0 is true and we repeated the experiment, what's the chance we would observed a value of $\hat{p}_M \hat{p}_F$ that is '**as or more extreme**' than we already have observed with our data?"
- The "chance" is a probability known as a **p-value**

$$p$$
-value = $\Pr(\hat{p}_M - \hat{p}_F \ge x_{obs} \mid H_0 \text{ is true})$

- ▶ A one-sided test: 'as or more extreme' implies $\geq x_{obs}$ values
- 'Probability' for a (hypothetical) repeated experiment (under H_0)
- \blacksquare \Rightarrow p-value approach yields same conclusion if use the same α
- ⇒ decision rule for p-value approach
 - ▶ If p-value $< \alpha$: Reject H_0
 - Otherwise: Do not reject H₀



- 1 A randomisation test: Use variability in observed data
 - NEW: A modern computational approach
- 2 A test based on the Central Limit Theorem (CLT)

Let $X = \hat{p}_M - \hat{p}_F$ denote the unobserved (random variable)

- Either before the data is collected
- Or from a hypothetical repeated experiment

Under the CLT:

$$X \stackrel{approx}{\sim} N\left(\mu_X, \ \sigma_X^2\right)$$

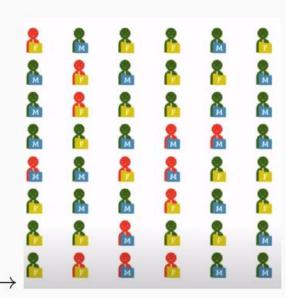
- $\mu_X = p_M p_F$ is the **mean** of X
- σ_X^2 is (an appropriate) **variance** of X
- $\Rightarrow \sigma_X = \sqrt{\sigma_X^2}$ is the standard error (SE)
 - ▶ but must be estimated since it will depend upon p_M and p_F



Idea: Use the variability in the data to approximate the p-value

- Shuffle = Randomly permute resumes (gender) assigned to supervisor (promotion outcome)
 - to simulate X under H₀
 - by breaking association (if present) between gender and promotion outcome







ETC5242 Week 4

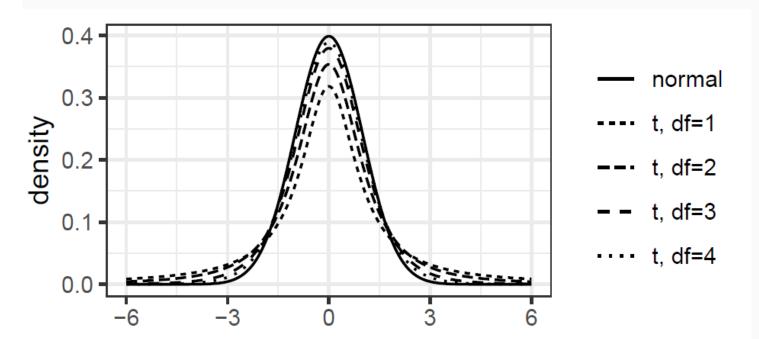
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Review: What purpose does a large sample serve?

- As long as observations are independent
- And the population distribution is not "extremely skewed"
- A "large" sample would ensure that...
 - the sampling distribution of the mean is nearly normal
 - ▶ the estimate of the standard error $(\frac{s}{\sqrt{n}})$ is reliable
 - (if skewed, even larger sample size is needed)
- $\blacksquare \Rightarrow s$ will be a good estimate of population standard deviation, σ

- Use when σ is unknown (almost always the case) to address the uncertainty of standard error estimate
- Is "bell shaped" but with thicker tails than the normal
 - centered at zero
 - one parameter: degrees of freedom (df) determine thckness of tails
 - compare with $N(\mu, \sigma^2)$, two parameters (mean= μ and SD = σ)





- for inference on a mean where
 - $ightharpoonup \sigma$ unknown, which is almost always
- calculated the same way

$$T = \frac{obs - null}{SF}$$

Here:

- obs refers to the value of an observed statistic
 - $ightharpoonup ar{x}$ from one sample, including \bar{x}_{Diff} for paired samples, where $d_i = x_{1i} x_{2i}$ is the "Diff" for pair i
 - $ightharpoonup \bar{x}_1 \bar{x}_2$ from two independent samples
- \blacksquare null refers to the corresponding value of the population quantity under H_0
- SE refers to the Standard Error, which is the standard deviation of the statistic
- p-value calculated using R
 - one or tail areas, based on H₁

- Find the following probabilities:
 - a. $Pr(|Z| > 2) = 0.0455 (\rightarrow reject)$
 - **b.** $Pr(|t_{df=50}| > 2) = 0.0509 (\rightarrow fail to reject?)$
 - c. $Pr(|t_{df=10}| > 2) = 0.0734 (\rightarrow fail to reject)$
- (And suppose you have a two sided hypothesis test, and your test statistic is
 2. Under which scenario would you be able to reject H₀ and the 5% significance level?)
- Generally degrees of freedom (df) is tied to the sample size

■ Biscuits after lunch study

biscuit intake	\bar{X}	S	n
solitaire	52.1 g	45.1 g	22
no distraction	27.1 g	26.4 g	22

 \blacksquare estimating the mean (single sample): point estimate \pm margin of error

$$ar{x} \pm t_{df}^{\star} SE_{ar{x}}$$
 $ar{x} \pm t_{df}^{\star} \frac{s}{\sqrt{n}}$
 $ar{x} \pm t_{n-1}^{\star} \frac{s}{\sqrt{n}}$

■ Degrees of freedom for t statistic for inference on one sample mean:

$$df = n - 1$$

Find the critical t score using R

■ Estimate the average after-lunch snack compumption (in grams) of people who eat lunch **distracted** using a 95% confidence interval

$$\bar{x} \pm t^* SE = 52.1 \pm 2.08 \times \frac{45.1}{\sqrt{22}}$$

= $52.1 \pm 2.08 \times 9.62$
= 52.1 ± 20
 $\Rightarrow (32.1, 72.1)$

Hypothesis test application

■ Suppose the suggested service size of these biscuits is 30g. Do these data provide convincing evidence that the amount of snacks consumed by distracted eaters post-lunch is different than the suggested serving size?

$$H_0: \mu = 30$$
 vs. $H_1: \mu \neq 30$

$$T = \frac{52.1 - 30}{9.62} = 2.3$$

p-value: use $2 \times probability greater than T under t distribution with df=21:$

[1] 0.0318023

 \Rightarrow Reject H_0 at level α = 0.05 since p-value = 0.0318 < 0.05

Video 3: Analysing paired data

- When two sets of observations have this special correspondence (not independent), they are said to be **paired**
- To analyse paired data, it is often useful to look at the difference in outcomes of each pair of observations:

$$DIff_i = read_i - write_i$$
, for each $i = 1, 2, ..., n = 200$

parameter of interest: Average difference between the reading and writing scores of all high school students

xbar_DIff	s_DIff	n_DIff	tstat
-0.545	8.88667	200	-0.867

■ point estimate? Average difference between the reading and writing scores of **sampled** high school students: \bar{x}_{Dlff}

- $H_0: \mu_{Diff} = 0$ vs. $H_1: \mu_{Diff} \neq 0$
- Same structure as one-sample mean test
- Test statistic:

$$T = \frac{-0.545 - 0}{8.887 / \sqrt{200}} = -0.867$$

- degrees of freedom: 200 1 = 199
- \Rightarrow p-value: 2*pt(-0.867, df=199) = 0.387
- Interpretation?: p-value of obtaining a random sample of 200 students where the average difference between the reading and writing scores is at least 0.545 (in either direction), if in fact the true average difference between the scores is zero.

Pr(observed or more extreme outcome $\mid H_0$ is true)

- Refer back to earlier **Biscuits after lunch study**
- \blacksquare point estimate \pm margin of error

$$(\bar{x}_1 - \bar{x}_2) \pm t^* SE_{(\bar{x}_1 - \bar{x}_2)}$$

- Standard error of difference between two independent means: $SE_{(\bar{X}_1 \bar{X}_2)} = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$
- We add the two variances (inside the square root) even though we are looking for the Standard error of the difference $(SE_{(\bar{x}_1-\bar{x}_2)})$
- Video advocates:

 DF for t statistic for inference on difference of two independent means: $df = min(n_1 1, n_2 1)$



Video 1: Independence:

- Within groups: sampled observations must be independent
 - random sample/assignment
 - if sampling without replacement, n < 10% of population
- Between groups: the two groups must be independent of each other (non-paired)
- 2 Sample size/skew: The more skewed the populations, the large the sample size we need from those distributions

- \blacksquare \Rightarrow Confidence interval for $\mu_{wd} \mu_{wod}$: (1.83g, 48.17g)
 - obtained from:

$$(\bar{x}_{wd} - \bar{x}_{wod}) \pm t_{df}^{\star} SE = (52.1 - 27.1) \pm 2.08 \times \sqrt{\frac{45.1^2}{22} + \frac{26.4^2}{22}}$$

= $25 \pm 2.08 \times 11.14$
= $25g \pm 23.17g$

 \blacksquare \Rightarrow Hypothesis test:

$$H_0: \mu_{wd} - \mu_{wod} = 0$$
 vs. $H_1: \mu_{wd} - \mu_{wod} \neq 0$

- $T_{21} = \frac{25-0}{11.14} = 2.24 \Rightarrow \text{p-value: } 2^*\text{pt}(2.24, \text{df=}21, \text{lower.tail=FALSE}) = 0.036$
 - ► Reject $H_0: \mu_{wd} \mu_{wod} = 0$ at the 5% level