

# MATH 10

# INTRODUCTORY STATISTICS

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*Your friendly neighbourhood graduated student.*

# Homework and Stuff

- Deadline extended : hand in a hard copy during Thursday's class.
- I will go through some examples today to help with this homework.
- Office hours is getting crowded. ( ▪ `3` ▪ )
- Will try to set up a group tutoring/review session via Email.  
Weekend?
- Alternatively : I can work through HW<sub>4</sub> on Weds office hours.

# Homework and Stuff

- Next homework is lighter and has more practice questions on hypothesis testing.
- Homework will be posted by Weds noon.
- Answer key for midterm delayed but will also be posted Weds noon.

# Week 7

final exam in 3 weeks

- **Chapter 12 – Test of Means** ← today's lecture

More hypothesis testing. Difference between means.

- **Chapter 13 – Interestingly titled “Power”**

The idea of the “power” of a test.

- **Chapter 14 – (brief introduction to) Regression**

## Correction – Binomial Hypothesis Testing

- Emailed out a correction to the slides last week.
- I will tell the grader to give you credit if you used the wrong version of the standard error for homework. **Thankfully, I don't think anyone did.**
- Recall : sampling distribution of a sample proportion  $p$  is approximately Normal with mean  $\pi$ .
- Standard error depends on whether the population proportion  $\pi$  is known.

## Correction – Binomial Hypothesis Testing

- Recall : sampling distribution of a sample proportion  $p$  is approximately Normal with mean  $\pi$ .
- Standard error depends on whether the population proportion  $\pi$  is known.
- Formula sheet :  $\sqrt{\frac{\pi(1-\pi)}{n}}$  if  $\pi$  known. Otherwise,  $\sqrt{\frac{p(1-p)}{n}}$ .
- Hypothesis testing : we assume the null hypothesis is true, which means  $\pi$  is known and given by the null hypothesis.

# Hypothesis Testing – A Review of Main Ideas

- Recall what we learned about **Sampling Distributions**.
- Suppose we have a large population, and some variable  $X$  that we are interested in.
- We take a sample of size  $n$ , and calculate the sample mean  $\bar{X}$ .
- The sampling distribution of the mean tells us the probability of getting certain values of the sample mean. (**dist. of  $\bar{X}$** )

# Hypothesis Testing – A Review of Main Ideas

- Suppose we don't know what the true population mean is.
- But we have some kind of idea or hypothesis about its value.

$H_0$  : *population mean is  $\mu$*  ,       $H_A$  : *not  $\mu$*



# Hypothesis Testing – A Review of Main Ideas

- Suppose we don't know what the true population mean is.
- But we have some kind of idea or hypothesis about its value.

$$H_0 : \text{population mean is } \mu \quad , \quad H_A : \text{not } \mu$$

- Assuming that  $H_0$  is true, then we know what the sampling distribution of the mean is (*estimating the standard error in the t-dist case*).
- We can set a threshold for when we will find  $H_0$  too unlikely and reject it.

# Hypothesis Testing – A Review of Main Ideas

$H_0$  : population mean is  $\mu$  ,       $H_A$  : not  $\mu$

- We can set a threshold for when we will find  $H_0$  too unlikely and reject it.
- We build this threshold based on our data, which is the sample mean  $\bar{X}$ , calculated from a sample of size  $n$ .
- More precisely, threshold is based on  $P(\text{sample mean} \geq \bar{X})$  or  $P(\text{sample mean} \leq \bar{X})$ . Let's assume  $\bar{X} > \mu$ , so we focus on one.

# Hypothesis Testing – A Review of Main Ideas

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- Lingo / Convention : *Chapter 11, Section 6 – Significant Results*
- Textbook calls this a (one-tailed) probability value or p-value.
- The textbook defines two types: one-tailed and two-tailed p-value.
- Will talk about this later. Just focus on this one-tailed p-value for now.

# Hypothesis Testing – A Review of Main Ideas

- More precisely, threshold is based on  $P(\text{sample mean} \geq \bar{X})$  or  $P(\text{sample mean} \leq \bar{X})$ . Let's assume  $\bar{X} > \mu$ , so we focus on one.
- Using the sampling distribution, we can figure out what  $P(\text{sample mean} \geq \bar{X})$  is.
- Recall : probability is area under the curve.
- z- or t-tables give you the wrong area under the curve.
- Need to use  $P(A) = 1 - P(\text{not } A)$ .

# Hypothesis Testing – A Review of Main Ideas

- Using the sampling distribution, we can figure out what  $P(\text{sample mean} \geq \bar{X})$  is.
- We define a threshold for when  $H_0$  is “too unlikely for us”.
- We say that if  $P(\text{sample mean} \geq \bar{X})$  is “too small”, then we “don’t believe” in  $H_0$ . So, we reject  $H_0$ .
- Significant level  $\alpha$  determines when this probability is “too small”.

# Hypothesis Testing – A Review of Main Ideas

- Significant level  $\alpha$  determines when this probability is “too small”.
- Condition depends on which of the two types of test used.

## One-tailed test

$$P(\text{sample mean} \geq \bar{X}) < \alpha$$

## Two-tailed test

$$P(\text{sample mean} \geq \bar{X}) < \alpha/2$$

# Hypothesis Testing – A Review of Main Ideas

One-tailed test :  $P(\text{sample mean} \geq \bar{X}) < \alpha$

Two-tailed test :  $P(\text{sample mean} \geq \bar{X}) < \alpha/2$

- If this condition is met (write it down in the exam!) then we say :

We reject  $H_0$  at  $\alpha$  level of significance. “blah blah blah” is *probably* “blah blah blah”.

# Hypothesis Testing – A Review of Main Ideas

One-tailed test :  $P(\text{sample mean} \geq \bar{X}) < \alpha$

Two-tailed test :  $P(\text{sample mean} \geq \bar{X}) < \alpha/2$

- If this condition is **NOT** met (write it down in the exam!) then we say :

**We do not reject  $H_0$  at  $\alpha$  level of significance. Inconclusive.**



# Hypothesis Testing – A Review of Main Ideas

- If this condition is **NOT** met (write it down in the exam!) then we say :

**We do not reject  $H_0$  at  $\alpha$  level of significance. Inconclusive.**

- For exam, you can ask me what conclusion is the question looking for.
- E.g. what does it mean for the vote to be “too close to call”.
- There might be very precise questions asking you about the meaning of hypothesis testing in general. E.g. everything is probabilistic. Can always reject or not reject by picking the  $\alpha$ .
- Will go through later in this lecture!

## Example Related to HW<sub>4</sub> Qns 1

- $H_0: \pi = 0.50$  ,  $H_A: \pi \neq 0.50$  , is male proportion greater than 0.50?
- Sample proportion  $p = 0.60$ , sample size  $n = 25$ .

## Example Related to HW4 Qns 1

- $H_0: \pi = 0.50$  ,  $H_A: \pi \neq 0.50$  , is male proportion greater than 0.50?
- Sample proportion  $p = 0.60$ , sample size  $n = 25$ .
- Assuming null hypothesis is true, population proportion is  $\pi = 0.50$ .
- Applying Normal approximation to the binomial distribution, the sampling distribution is Normal with mean  $\pi = 0.50$  and variance  $\pi(1 - \pi)/n$ .

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- Applying Normal approximation to the binomial distribution, the sampling distribution is Normal with mean  $\pi = 0.50$  and variance  $\pi(1 - \pi)/n$ .
- Z-statistic/value is  $\frac{p - \text{mean}}{\text{standard error}} = \frac{p - \pi}{\sqrt{\frac{\pi(1-\pi)}{n}}} = \frac{0.1}{0.1} = 1$ .

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- $H_0: \pi = 0.50$  ,  $H_A: \pi \neq 0.50$  , is male proportion greater than 0.50?

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- (one-tailed) p-value is :

$$P(\text{sample prop} \geq 0.6) = P(Z \geq 1) = 1 - P(Z < 1) = 1 - 0.8413 = 0.1587.$$

- **Rejecting null or not depends on your significance level.**

## Example Related to HW4 Qns 1

- (one-tailed) p-value is :

$$P(\text{sample prop} \geq 0.6) = P(Z \geq 1) = 1 - P(Z < 1) = 1 - 0.8413 = 0.1587.$$

- Rejecting null or not depends on your significance level  $\alpha$ .
- We are comparing areas : is  $P(\text{sample prop} \geq 0.6)$  less than  $\alpha/2$ ?
- Suppose level of significance is  $\alpha = 0.10$ .
- Then,  $P(\text{sample prop} \geq 0.6) = 0.1587 > \frac{0.10}{2} = 0.05$ .

## Example Related to HW<sub>4</sub> Qns 1

- (one-tailed) p-value is :

$$P(\text{sample prop} \geq 0.6) = P(Z \geq 1) = 1 - P(Z < 1) = 1 - 0.8413 = 0.1587.$$

- Suppose level of significance is  $\alpha = 0.10$ .
- Then,  $P(\text{sample prop} \geq 0.6) = 0.1587 > \frac{0.10}{2} = 0.05$ . → write this in exam!!
- Conclusion : → variation of exactly what you should write in exam.

**We do not reject the null, at the  $\alpha$  level of significance. Results are inconclusive.**

## Chapter 12, Section 4 – Hypo. Test For Difference Between Means

- The general strategy :

$$z \text{ or } t = \frac{\textit{sample difference} - \textit{hypothesized difference}}{\textit{standard error}}$$

- Use  $z$  when population variances are given. Sampling distribution is normal.
- Use  $t$  when population variances are not given. Sampling distribution is the  $t$ -dist.
- $Df = (n-1) + (n-1) = 2(n-1)$ .



# Chapter 12, Section 4 – Hypo. Test For Difference Between Means

## Assumptions for the t-dist case

1. Both populations are **normally distributed** with the same **unknown variance**.
2. Both simple random samples are independent and have same size  $n$ .

$$MSE = \frac{S_1^2 + S_2^2}{2}$$

$$\text{Standard Error, SE} = \sqrt{\frac{2 MSE}{n}} = \sqrt{\frac{S_1^2 + S_2^2}{n}}$$

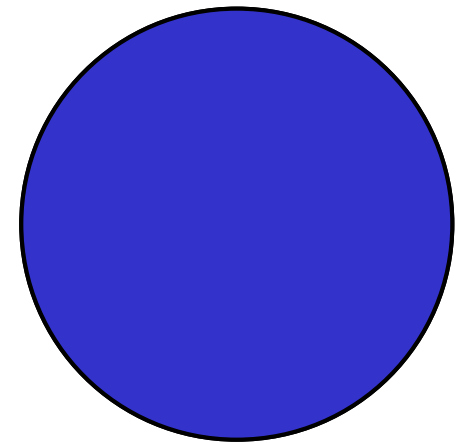
# Break time!! \o/

- No exercise today! Go enjoy your break. ^\_\_^

- Circle is a timer that becomes blue. O\_o  
*(please ignore if it glitches)*



**12 minutes**



# p-value and Bayes Theorem

- (one-tailed) p-value =  $P( D = \text{data or more extreme} \mid H = \text{null hypothesis is true} )$

- But  $P( H \mid D ) = \frac{P( D \mid H ) P( H )}{P( D )}$ .

- We can go further:  $P( H \mid D ) = \frac{P( D \mid H ) P( H )}{P( D \mid H ) P( H ) + P( H \mid D ) P( D )}$ .

# p-value and Bayes Theorem

- (one-tailed) p-value =  $P( D = \text{data or more extreme} \mid H = \text{null hypothesis is true} )$
- But  $P( H \mid D ) = \frac{P( D \mid H ) P( H )}{P( D )}$ .
- So p-value is NOT the probability that the null hypothesis is true, which is  $P(H|D)$ .
- Technically, we are using  $P( D \mid H )$  to “guess” whether  $P( H \mid D )$  would be small.
- $P( D \mid H )$  is probability of data given that null hypothesis is true.
- E.g. if two men committed the murder, what is prob. that DNA matches? → HW<sub>3</sub>

## Chapter 11, Section 10 – Misconceptions

Extremely important for the exams. Might be in final exam. (◡‿◡)

- Is the p-value the probability that the null hypothesis is false/true?
- Does a low p-value indicate a large effect?
- If an outcome is not statistically significant, does it mean that the null hypothesis is true?



# Psychology journal bans $P$ values

Test for reliability of results 'too easy to pass', say editors.

[Chris Woolston](#)

26 February 2015 | Clarified: 09 March 2015



PDF



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A controversial statistical test has finally met its end, at least in one journal. Earlier this month, the editors of *Basic and Applied Social Psychology* (BASP) announced that the journal would no longer publish papers containing  $P$  values because the statistics were too often used to support lower-quality research<sup>1</sup>.

Authors are still free to submit papers to BASP with  $P$  values and other statistical measures that form part of 'null hypothesis significance testing' (NHST), but the numbers will be removed before publication. [Nerisa Dozo](#), a PhD student in psychology at the University of Queensland in Brisbane, Australia, tweeted:

NATURE | NEWS



# Statisticians issue warning over misuse of $P$ values

Policy statement aims to halt missteps in the quest for certainty.

[Monya Baker](#)

07 March 2016

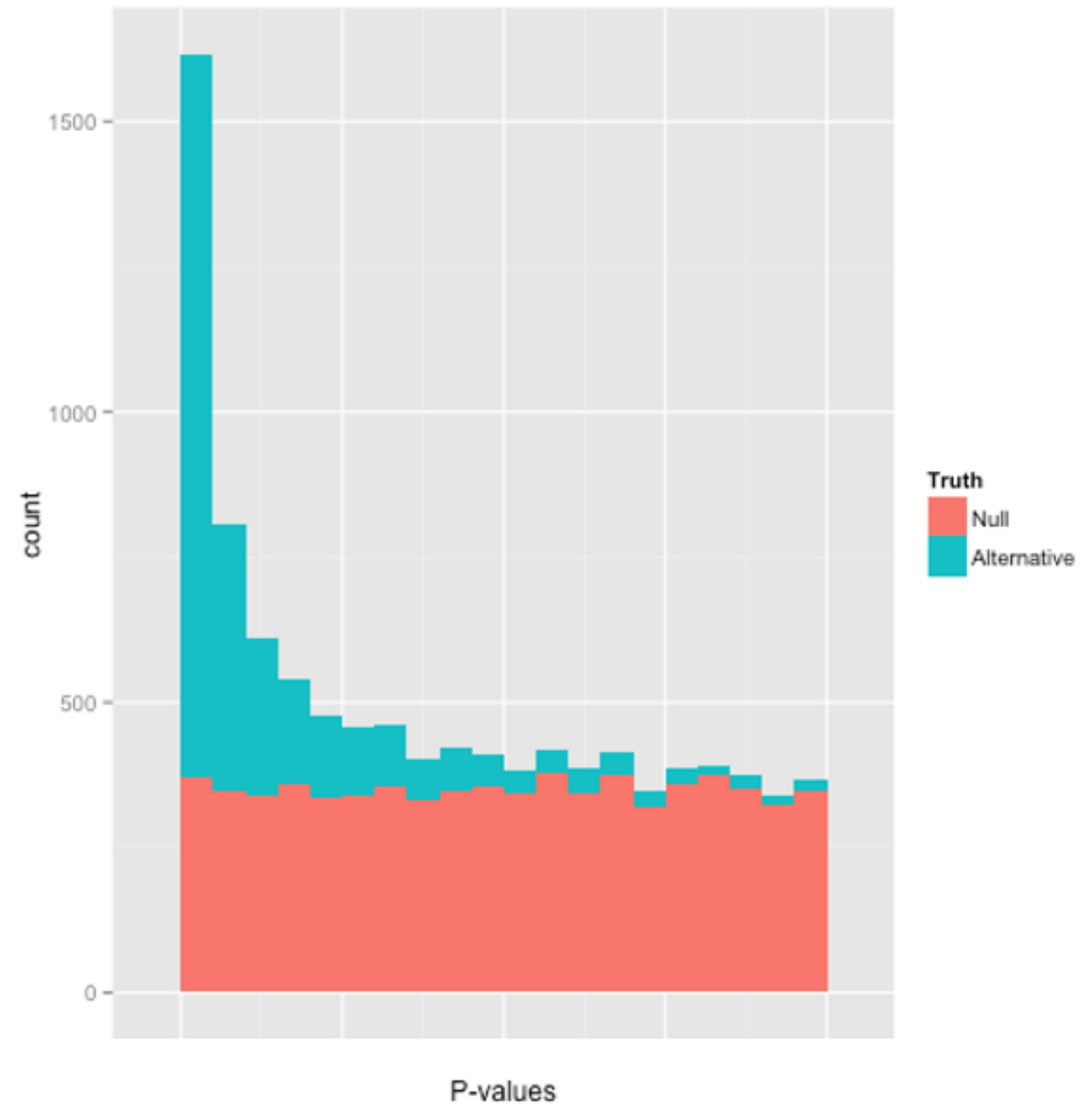


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Misuse of the  $P$  value — a common test for judging the strength of scientific evidence — is contributing to the number of research findings that [cannot be reproduced](#), the American Statistical Association (ASA) warns in a [statement](#) released today<sup>1</sup>. The group has taken the unusual step of issuing principles to guide use of the  $P$  value, which it says cannot determine whether a hypothesis is true or whether results are important.



# Disclaimer

- P-values from different experiments are *not exactly* compatible.
- The experimental conditions have to be identical.
- Mathematically : sampling distribution will change depending on sample size, and stratified vs simple random sampling etc!
- But if experimental conditions are identical. We should see those distributions.

*"The truth is never simple and rarely pure." – Oscar Wilde*

## p-value : lingo and conventions (not required for exam!)

- Officially, in mathematical statistics, the p-value is defined as the smallest significance level that will cause a rejection of the null.
- You might have also seen this general definition for a two-tailed p-value :

$$2 \cdot \min\{ P(\text{sample mean} \geq \bar{X}), P(\text{sample mean} \leq -\bar{X}) \}$$



## p-value : lingo and conventions (not required for exam!)

- You might have also seen this general definition for a two-tailed p-value :

$$2 \cdot \min\{ P(\text{sample mean} \geq \bar{X}), P(\text{sample mean} \leq -\bar{X}) \}$$

- Your textbook defines two versions of the p-value :

One-tailed p-value  $P(\text{sample mean} \geq \bar{X})$

Two-tailed p-value  $2 \cdot P(\text{sample mean} \geq \bar{X})$

- Your textbook's condition for a two-tailed test is :

$$2 \cdot P(\text{sample mean} \geq \bar{X}) < \alpha$$

## p-value : lingo and conventions (not required for exam!)

- Textbook : Two-tailed p-value  $2 \cdot P(\text{sample mean} \geq \bar{X})$
- Textbook's condition for a two-tailed test is :

$$2 \cdot P(\text{sample mean} \geq \bar{X}) < \alpha$$

- This is exactly the same as what we have been doing so far :

$$P(\text{sample mean} \geq \bar{X}) < \alpha/2$$

- Both full credit. If you already have  $P(\text{sample mean} \geq \bar{X})$ , no need to multiply it by 2.

# Skipped Chapters

- Chapter 12, Section 6, Pairwise Comparisons (Tukey HSD test),
- Chapter 12, Section 7, Specific Comparisons,
- Chapter 12, Section 8 , Correlated Pairs
- Chapter 12, Section 11, Pairwise (Correlated) (Bonferroni correction),
  
- are not required!

→ Syllabus on the website has been updated.