

# **AE 102: Data Analysis and Interpretation**

## **Hypothesis testing**

January - April 2016

Department of Aerospace Engineering

## **Hypothesis?**

A supposition or proposed explanation made on the basis of limited evidence as a starting point for further investigation.

## **Hypothesis testing?**

- Use data to test a hypothesis
- Use probability theory to make this numerical

## **Examples**

- Tara asks BSA to manufacture the cycle seats
- A shipment of 100 seats arrives
- Should the shipment be accepted or not?

## **Example from the text**

- Company purchases cables which are supposed to take a load of 7000 psi.
- Large batch arrives, should they accept the shipment?

## **Examples**

- An experiment to determine if vitamin C helps heal a cold.
- Is there statistical evidence in the data?

## **Examples**

- A 4-sided die is tossed 1000 times and shows up 290 times
- Is it a fair die?

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- Note: example is from [stat414](#)

## Basic idea

- Make an initial assumption (hypothesis) about the parameter, called the **null hypothesis** or  $H_0$
- Collect data
- Decide if we **accept** the hypothesis or **reject** it

## Example

A 4-sided die is tossed 1000 times and shows up 290 times. Is it a fair die?

- Where do we start?

## Solution

- First formulate the hypothesis,  $H_0 : p = 0.25$
- Collect data
- Look at it, how?

## Solution

- Let  $Y$  be RV which is the number of 4's.
- $\hat{p} = Y/n = 290/1000$
- This is normally distributed
- As per assumption, mean =  $p_0 = 0.25$
- $\sigma = \sqrt{p_0(1 - p_0)/n}$
- Note that  $\hat{p}$  is an MLE

## Solution

- Convert to a standard normal  $Z = \frac{\hat{p} - p_0}{\sigma}$
- How many std-devs account for 0.29?  $Z = 2.92!$
- Is this reasonable or not?
- What if we obtained 0.255 instead of 0.29?

## Critical regions

1. Identify **critical region** where hypothesis is **rejected**
2. Use a **level of significance**,  $\alpha$

## Critical regions

- Could define that if  $Z > 1.645$ , we reject the null
  - That is if  $\hat{p} > 0.273$ , we reject the null
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- What is the probability of finding  $\hat{p} > 0.273$ ?
- Note that this is a one-tailed test (right-tailed test)

## The process

1. Define the null hypothesis,  $H_0$  and the alternative hypothesis  $H_A$
  2. Calculate the test statistic  $Z = \frac{\hat{p} - p_0}{\sigma}$
  3. Identify the critical region.
  4. Decide to reject/accept  $H_0$
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Note that some point estimator is used to find the  $\hat{p}$  value

## Example

1.  $H_0 : p_0 = 0.25, H_A : p_0 > 0.25$
2.  $Z = 2.92$
3. Set critical region as:  $Z_{critical} > 1.645$  (or  $\hat{p} > 0.27$ )
4. Reject null hypothesis

## Errors

- Type I error: rejecting  $H_0$  when it is true
- Type II error: accepting  $H_0$  when it is false

## Type I error

- If  $Z_{critical} = 1.645$ 
  - Type I error is 0.05 (this is  $\alpha$ )!
  - Why?

## Type I error

- Fix  $\alpha$  to set critical region!
- Typical  $\alpha$  values are 0.01, 0.05 etc.
- Called **significance level of the test**

## Type I error

- Given significance level  $\alpha$ , Type I error probability is  $\alpha$

## Type II error

- Say we got  $p = 0.273$  and accepted  $H_0$
- If actual  $p = 0.27$ , what would be the error?
- $\beta(p) = P_p\{\text{accept } H_0\}$
- $P\{\hat{p} < 0.273 \text{ given } p_0 = 0.27\} = 0.587$
- Clearly, this depends on  $p_0$  the true value
- More on this later

## The p-value

- Given the test-statistic value, its corresponding  $\alpha$  value is the **p-value**
- If p-value  $< \alpha$  accept  $H_0$ , else reject

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Note: this is not the  $p$  of the example problem!

## Example

- If  $\alpha = 0.05$
- When  $\hat{p} = 0.29$ ,  $P(p \geq 0.29) = 0.0018$
- p-value = 0.0018 (which is  $< \alpha$ )
- So reject the null

## Problem

If we obtained 265 4 values in the example, what is the corresponding "p-value"?

## Another Problem

- $q$  is proportion of two-wheeler drivers wearing helmets
- It was found in a study that  $q = 0.14$
- The traffic safety council: makes advertisements on TV to increase this
- After two months: 104/590 riders wear helmets
- Did the advertising help?

## Recap

- Statistical hypothesis: statement about some parameters of a population
- Null Hypothesis,  $H_0$
- Example: for a normal population with variance 1: -  $H_0 : \theta = 1$
- $\theta > 1$  is the alternate hypothesis,  $H_1$
- Critical region: values there reject  $H_0$
- Check that the hypothesis is consistent with the data

## Recap

- Type I error
- Type II error
- Choose a level of significance,  $\alpha$
- Test statistic should have probability of type I error  $< \alpha$  to reject the null

## Summary

1. Define the null hypothesis,  $H_0$  and the alternative hypothesis  $H_A$
2. Calculate the test statistic  $Z = \frac{\hat{p} - p_0}{\sigma}$
3. Set a level of significance,  $\alpha$
4. Find the  $p$  value of the test statistic
5. Decide to reject  $H_0$  if p-value  $< \alpha$

## One more example

A poll of 1000 people reveals that 47% of the population is very happy. The choices were, "very happy", "fairly happy" and "not happy". Can one say that the very happy people are a minority?

## Kinds of alternate hypothesis

- Given some null hypothesis,  $H_0 : \mu = \mu_0$
- The possible alternate hypothesis are:
  1. Two-sided or two-tailed test:  $H_A : \mu \neq \mu_0$
  2. One-sided or one-sided test:
    - $H_A : \mu < \mu_0$
    - $H_A : \mu > \mu_0$

## Examples

- Your turn!

## A more formal discussion

### Some Definitions

- Simple hypothesis: when true completely specifies the population
- Composite hypothesis

## Tests for the mean

- n-sample from normal population, with known variance  $\sigma$
- $H_0 : \mu = \mu_0$ , some constant
- $H_1 : \mu \neq \mu_0$
- $Z = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$
- $P_{\mu_0}\{|\bar{X} - \mu_0| > c\} = \alpha$
- $P\{|Z| > \frac{c\sqrt{n}}{\sigma}\} = \alpha$

## Tests for the mean

- $P\{|Z| > \frac{c\sqrt{n}}{\sigma}\} = \alpha$
- $P\{Z > \frac{c\sqrt{n}}{\sigma}\} = \alpha/2$
- By definition  $P\{Z > z_{\alpha/2}\} = \alpha/2$
- Thus,  $\frac{c\sqrt{n}}{\sigma} = z_{\alpha/2}$
- Reject  $H_0$  if  $\frac{\sqrt{n}}{\sigma}|\bar{X} - \mu_0| > z_{\alpha/2}$

## One-sided tests

- $H_0 : \mu = \mu_0$ , some constant
- $H_1 : \mu > \mu_0$
- $P_{\mu_0}\{\bar{X} - \mu_0 > c\} = \alpha$
- In this case  $\frac{c\sqrt{n}}{\sigma} = z_\alpha$
- Reject  $H_0$  if  $\frac{\sqrt{n}}{\sigma}(\bar{X} - \mu_0) > z_\alpha$

## One-sided tests

- $H_0 : \mu = \mu_0$ , some constant
- $H_1 : \mu < \mu_0$
- $P_{\mu_0}\{\bar{X} - \mu_0 < c\} = \alpha$
- In this case  $\frac{c\sqrt{n}}{\sigma} = z_\alpha$
- Reject  $H_0$  if  $\frac{\sqrt{n}}{\sigma}(\bar{X} - \mu_0) < z_\alpha$

## Tests for unknown variance

- Use the sample variance
- Use t-distribution of  $n - 1$  d.o.f.
- Called a t-test

## Other cases

- Equality of means of normal populations
  1. Known variances
  2. Unknown variances which are same
  3. Unknown variances which are not same
- Paired t-test
  - Use when the samples are correlated.
- Test for variance of a normal population

## Paired tests: an example

- A new device supposedly increases mileage
- Measure mileage before and after and compare means
- *But* the two samples are correlated!
  - If car 1 is a Nano its mileage is good already
- Solution?

## What about Type II errors?

- Type II error: accepting  $H_0$  when it is false
- $\beta(\mu) = P_\mu\{\text{accept } H_0\}$
- Assumed  $\mu_0$ , but true mean is  $\mu$ , i.e.

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim \mathcal{N}(0, 1)$$

## Type II error: two-sided test

$$\begin{aligned}\beta(\mu) &= P_\mu \left\{ \frac{\sqrt{n}|\bar{X} - \mu_0|}{\sigma} \leq z_{\alpha/2} \right\} \\ &= P \left\{ -z_{\alpha/2} - \frac{\mu}{\sigma/\sqrt{n}} \leq \frac{\bar{X} - \mu_0 - \mu}{\sigma/\sqrt{n}} \leq z_{\alpha/2} - \frac{\mu}{\sigma/\sqrt{n}} \right\} \\ &= P \left\{ \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} - z_{\alpha/2} \leq \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \leq z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} \right\} \\ &= P \left\{ \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} - z_{\alpha/2} \leq Z \leq z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} \right\} \\ &= \Phi\left(\frac{\mu_0 - \mu}{\sigma/\sqrt{n}} + z_{\alpha/2}\right) - \Phi\left(\frac{\mu_0 - \mu}{\sigma/\sqrt{n}} - z_{\alpha/2}\right)\end{aligned}$$

## Type II errors

- This is called the **operating characteristic curve (OC)**
- One tries to minimize  $\beta$  for a given case
- $1 - \beta$  is called the **power of the test**
- One typically tries to minimize both  $\alpha, \beta$
- The  $\beta$  is useful to determine the sample size



## Type II errors: one-sided

$$\begin{aligned}\beta(\mu) &= P_{\mu} \left\{ \bar{X} \leq \mu_0 + z_{\alpha} \frac{\sigma}{\sqrt{n}} \right\} \\ &= P \left\{ Z \leq \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} + z_{\alpha} \right\} \\ &= \Phi \left( \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} + z_{\alpha} \right)\end{aligned}$$