Statistical Methods: Lecture 6

## Lecture Overview

Basics of Hypothesis Testing

Testing claims: proportions

Basics of Hypothesis Testing: additional information

# Testing a Claim

Aim: test a claim, i.e., "hypothesis". 2 sorts: null hypothesis / alternative hypothesis.

E.g., coin tossing. Test: "unbiased" vs. "biased". Toss 100 times.

Outcome not "fitting" claim (null hypothesis)? Then reject it.

- ► 47×H, 53×1\$  $\hat{p}_{100} = 0.47$  fair?
- ► 5×H, 95×1\$  $\hat{p}_{100} = 0.05$  biased
- ► 37×H, 63×1\$  $\hat{p}_{100} = 0.37$  —fair/biased??

To test hypotheses we need data (from experiments).

Claim: coin is fair (unbiased) (vs. claim: coin favors tails) Experiment: toss coin 100 times Outcome:  $37 \times \text{heads}$ ,  $63 \times \text{tails}$ 

Probability of 37 heads or less assuming claim is true?

If coin is fair, this probability is < 0.01.

- ⇒ Result "37 heads" unlikely with a fair coin.
- $\Rightarrow$  Deviation statistically significant at level 1.

Conclusion: based on outcome it is unlikely that coin is fair.

Now we are considering test for p, later we will consider tests for  $\mu$  and other tests.

2 hypotheses:

### Null hypothesis

 $H_0$ : population parameter = hypothetical value

### Alternative hypothesis

Depending on the claim we choose one out of three:

```
H_a: \begin{cases} \text{population parameter} < \text{hypothetical value} & \text{left-tailed test} \\ \text{population parameter} > \text{hypothetical value} & \text{right-tailed test} \\ \text{population parameter} \neq \text{hypothetical value} & \text{two-tailed test} \end{cases}
```

Left- and right-tailed tests: one-sided tests, two-tailed test: two-sided test.

### Measure "fitting"

Aim: find probability of

"if  $H_0$  were true, the experiment outcome is at least as extreme as observation."

- ls  $\hat{p}_{100} = 0.37$  likely if we assume that p = 0.5?
- ▶ Is  $\bar{x}_n = 145.85$  likely if we assume that  $\mu = 155$ ?

### Problem?

Sample statistic's (proportion/mean) distribution depends on parameter to be tested (see CLT).

### Solution?

Basics of Hypothesis Testing

000000

- Test statistic ... is a random variable used for deciding about  $H_0$ . Convert sample statistic by pretending  $H_0$  is true ((null) hypothetical value).

Aim: distribution of test statistic under  $H_0$  (null hypothesis). Must not depend on true p (next time:  $\mu$ ) but can depend on n.

Example: Test statistic for testing about p  $H_0$ : p = 0.4, i.e. hypothetical value  $p_0 = 0.4$ .

sample statistic 
$$\hat{P}_n$$
  $\Rightarrow$  test statistic  $\frac{\hat{P}_n - 0.4}{\sqrt{\frac{0.4 \cdot 0.6}{n}}}$ 

(i.e. insert hypothetical value  $p_0=0.4$  into  $\frac{P_n-p_0}{\sqrt{\frac{p_0(1-p_0)}{p_0(1-p_0)}}}$ ) What do we know about this test statistic?

### Five steps of hypothesis testing

- 0. identify population parameter
- 1. formulate  $H_0$  and  $H_a$  and choose significance level  $\alpha$
- 2. collect data
- 3a. choose test statistic and identify its distribution under  $H_0$
- 3b. calculate the value of test statistic based on data
- 3c. find P-value (depending on which alternative): probability under H<sub>0</sub> that test statistic has more extreme values than observed value
  - 4. formulate, based on P-value and  $\alpha$ , conclusion regarding  $H_0$ : reject (small P-value) or not (large P-value)

### Significance level

Denoted by  $\alpha$ , small, usually 0.05, but also 0.01 or 0.1.

### Be careful

Hypothesis testing is asymmetric: if  $H_0$  not rejected, we still do not "accept it"!

Example: Fair coin?

37 out of 100 coin tosses were heads. Claim: the coin is biased towards tails.

Step 0: Identify population parameter: population proportion of heads p

Step 1: Formulate  $H_0$  and  $H_1$  and choose significance level  $\alpha$ :

 $H_0: p = 0.5$  vs.  $H_1: p < 0.5$  with significance level  $\alpha = 0.05$ .

Other examples

More than 60% of students have a Snapchat account

 $H_0: p = 0.6$ 

 $H_1: p > 0.6$ 

20% of Dutch citizens support VVD

 $H_0: p = 0.2$ 

 $H_1: p \neq 0.2$ 

Less than 35% of women smoke

 $H_0: p = 0.35$ 

 $H_1: p < 0.35$ 

Step 2: Collect data:  $37 \times H$ ,  $63 \times T$ 

Step 3a: Choose test statistic and identify its distribution under  $H_0$ :

Use that the sampling distribution of the sample proportion approximately has a  $N(p, \frac{p(1-p)}{2})$  distribution. A good test statistic is

$$Z = \frac{\hat{P}_n - p}{\sqrt{\frac{p(1-p)}{n}}}$$

where p=0.5 is the claimed value from  $H_0$ . Under  $H_0$ ,  $Z \sim N(0,1)$  approximately (if n is large enough).

### Be careful

We use the claimed value (under  $H_0$ ), and not  $\hat{p}_n$ .

Step 3b: Calculate the value of the test statistic based on data:

It follows from the data that  $\hat{p}_n = 0.37$ , so the observed value of the test statistic is

$$z = \frac{0.37 - 0.5}{\sqrt{\frac{0.5 \cdot 0.5}{100}}} = -2.6$$

#### Step 3c: Find *P*-value:

We observed z = -2.6.

What is the probability of "test statistic takes even more extreme values"?

Our test is left-tailed:  $H_1: p < 0.5$ That means we have to compute

$$P(Z \le -2.6) = 0.0047.$$

#### Step 4: Conclude:

The *P*-value is smaller than the significance level  $\alpha=0.05$ , so  $H_0$  is rejected. We have enough evidence to confirm the coin is baised towards tails.

### Calculation of P-value

#### If the test is

- left-tailed: P-value = area to the left of z,
- right-tailed: P-value = area to the right of z,
- two-tailed & z < 0: P-value =  $2 \times$  area to the left of z,
- two-tailed & z > 0: P-value =  $2 \times$  area to the right of z.

### P-value and conclusion

- ▶ If P-value  $\leq \alpha$ : reject  $H_0$ .
- ▶ If P-value >  $\alpha$ : do not reject  $H_0$ .

### Another example

Claim(s): (Not) 10% of population is left-handed Data: n = 750 people surveyed, 92 are left-handed

- 0. identify population parameter
- 1. formulate  $H_0$  and  $H_a$  and choose significance level  $\alpha$
- 2. collect data
- 3a. choose test statistic and identify its distribution under  $H_0$
- 3b. calculate value of test statistic based on data
- 3c. find P-value (depending on  $H_a$ )
- 4. formulate conclusion regarding  $H_0$ : reject or do not reject  $H_0$

### Another example

Null hypothesis: 10% of population is left handed Data: n = 750 people surveyed, 92 are left-handed

0: population parameter: population proportion p

1:  $H_0: p = 0.1$   $H_a: p \neq 0.1$  significance level  $\alpha = 0.01$ 

2: data: check; (but  $H_0$ ,  $H_a$ , and  $\alpha$  need to be chosen before collecting data.)

3: Calculate score, where claimed value  $p_0 = 0.1$ 

$$z = \frac{\hat{p}_n - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}} = \frac{\frac{92}{750} - 0.1}{\sqrt{\frac{0.1 \cdot 0.9}{750}}} \approx 2.07$$

Under  $H_0$ : test statistic has approx. N(0,1)-distribution.

Test: two-tailed, and z > 0. Hence,

$$P$$
-value =  $2 \cdot P(Z \ge 2.07) = 2 \cdot (1 - P(Z \le 2.07)) = 2 \cdot 0.0192 = 0.0384$ .

4: conclusion?

#### Recall:

Claim: 20% of Dutch citizens support VVD —  $H_0: p=0.2$   $H_a: p\neq 0.2$  Claim: Less than 35% of women smoke —  $H_0: p=0.35$   $H_a: p<0.35$ 

## Claims and hypotheses

More examples of translating claims into hypotheses in the book. Careful:  $H_0$  AND  $H_a$  both consist of competing claims.

No golden rule for choosing hypotheses; ... some guidelines:

- present/typical situation is in H<sub>0</sub>,
- ▶ hypothesis you wish to reject is in *H*<sub>0</sub>,
- ightharpoonup hypothesis you wish to confirm is in  $H_a$ .

### Three approaches to hypothesis testing

- confidence interval method (not used here)
- P-value method
- critical value method

### Methods: P-value vs critical value

Same conclusions. Critical value sometimes simpler.

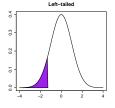
Critical region: all extreme scores for which  $H_0$  is rejected (at significance level  $\alpha$ ).

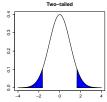
Critical region: depends on  $\alpha$ , distribution of test statistic and  $H_a$ .

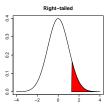
### Critical regions

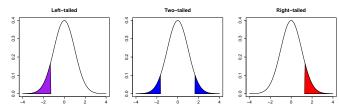
Fix  $\alpha$  and look at density of test statistic's distribution. If test is

- ▶ left-tailed: critical region = extreme left tail
- two-tailed: critical region = both extreme left AND right tails
- ▶ right-tailed: critical region = extreme right tail









- ightharpoonup area of purple region =  $\alpha$
- $\blacktriangleright$  total area of blue regions =  $\alpha$ , hence each part has area =  $\alpha/2$
- ightharpoonup area of red region =  $\alpha$

### Critical regions: test about proportion

Test statistic's distribution approx. N(0,1) under  $H_0$ .

For instance, for  $\alpha = 0.05$ , we reject  $H_0$  if

- ightharpoonup z < -1.645 (left-tailed test),
- ightharpoonup z < -1.96 OR  $z \ge 1.96$  (two-tailed test),
- ightharpoonup z > 1.645 (right-tailed test).

Example: Fair coin?

Only step 3 is different when the critical region is used instead of the P-value.

We test  $H_0$ : p = 0.5 vs.  $H_1$ : p < 0.5 with significance level  $\alpha = 0.05$ .

The test statistic is

$$Z=rac{\hat{P}_n-p}{\sqrt{rac{p(1-p)}{n}}}\sim N(0,1)$$
 under  $H_0$ 

The test is left-tailed, so the critical region is  $z \le -1.645$ .

We compute

$$z = \frac{0.37 - 0.5}{\sqrt{\frac{0.5 \cdot 0.5}{100}}} = -2.6.$$

The observed value lies in the critical region, therefore  $H_0$  is rejected. We have enough evidence to confirm the coin is biased towards tails.

Example: Left-handed people

Only step 3 is different when the critical region is used instead of the P-value.

We test  $H_0: p = 0.1$  vs.  $H_1: p \neq 0.1$  with significance level  $\alpha = 0.01$ .

The test statistic is

$$Z = \frac{P_n - p}{\sqrt{\frac{p(1-p)}{n}}} \sim N(0,1) \text{ under } H_0$$

The test is two-tailed, so the critical region is  $z \le -2.575$  or  $z \ge 2.575$ . (See Table 2; now  $\alpha = 0.01$ .)

We compute

$$z = \frac{92/750 - 0.1}{\sqrt{\frac{0.1 \cdot 0.9}{750}}} \approx 2.07.$$

The observed value does not lie in the critical region, therefore  $H_0$  is not rejected. We do not have enough evidence to confirm that the percentage of left-handed people is unequal to 10%.

Since hypothesis testing is based on data, randomness can cause errors.

	$H_0$ true	$H_0$ false
Reject H <sub>0</sub>	Type I error	correct
Do not reject $H_0$	correct	Type II error

- ▶ Type I error: mistake of rejecting  $H_0$  when it is true. Probability of a type I error  $\leq \alpha$  ( $\alpha$  is called the significance level).
- ▶ Type II error: mistake of not rejecting  $H_0$  when it is false. Its probability is denoted by  $\beta$ .

### Properties of two errors

- Probability of type I error is fixed.
- ▶ Prob. of type II error depends on  $(\alpha,)$  n and actual value of population parameter. Usually decreases with n.