

8 - Hypothesis Testing

In science, a *hypothesis* is a statement that you can test by experimentation.

Example Galileo Galilei (1564-1642) tested the hypothesis that *any pair of objects dropped from the same height fall to the Earth in equal time*.

To test this, Galileo dropped balls of various sizes off the Tower of Pisa. Each time, Galileo saw that the balls hit the ground at the same time.

In statistics, a *hypothesis* is:

e.g., The mean mass of a bag of Old Dutch Ketchup chips is 255g.

The evidence we use to test a statistical hypothesis is:

- To test a claim about a population mean μ , we use _____ from sample data.
- To test a claim about a population proportion p , we use _____ from sample data.
- To test a claim about a population variance σ^2 , we use _____ from sample data.

In statistics, a *hypothesis test* is a formal procedure that uses *sample statistics* to test a claim about the corresponding *population parameter*.

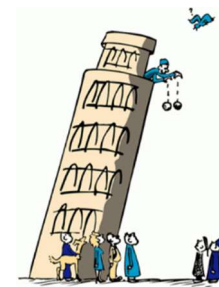
Hypothesis testing is based on the principle:

Sample data only provides adequate support for a hypothesis if that level of support is not likely to happen *by chance*.

Example Suppose I flip a coin 6 times and I get 5 “heads.” Can I conclude that the coin is biased towards heads?

Answer: No. The probability of getting 5 *or more* heads for a fair coin is: $\frac{6+1}{2^6} = 0.1094$

Since $0.1094 > 5\%$, we say that the evidence is *not statistically significant* at a 5% significance level. The coin is not biased towards heads, as far as we can tell.



Rejection Region Method for μ

Example (Two-Tailed) The network engineer claims that the mean time to transfer a 1 Gb file across a network is 12.44 seconds. You test this claim by sending a 1 Gb file across the network at 10 random times, recording the following transfer times. Test the engineers claim at a 5% significance level.

12.0	13.5	11.8	12.4	13.8
11.9	12.2	12.9	13.0	12.3

Step 1: (State the claim) – Which population parameter? What is the claim?

*Step 2: (Hypotheses) – Record the *null hypothesis* and the *alternative hypothesis**

Step 3: (Test Statistic) - Calculate the appropriate test statistic

Step 4: (Rejection Region) – What values of the test statistic would make us reject H_0 ?

Step 5: (Decision) – Reject H_0 or Fail to Reject H_0 ?

Step 6: (Conclusion) – State your conclusion using the wording of the original claim.

Example (Left-Tailed) One week later, the network engineer claims that the mean time to transfer a 1 Gb file across a network has been reduced *below* 10 seconds. You test this claim by sending a 1 Gb file across the network at 5 random times, recording the following transfer times. Test the engineers claim at a 5% significance level.

9.0	8.5	10.1	8.7	9.5
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p -value Method for μ

Definition The p -value of a given sample statistic is the probability of *randomly* obtaining a sample statistic as extreme as that given value, assuming the null hypothesis is true. (Here “extreme” means *in the direction specified in the alternative hypothesis*.)

We reject H_0 if the p -value is below α .

Example (Right-Tailed) A BCIT study aims to determine the average value of the variable:

X = amount of time a BCIT student works (at a job) each week

The study collects data for a random sample of students. Sample statistics are given:

$$n = 100; \bar{X} = 11.2; s = 4.5$$

Test the claim (at the 5% level) that the mean μ is *greater* than 10 hours per week.

Inferences about Population p

Inferences about p are based on the principle that:

Sample proportion \hat{p} is *normally distributed* as long as $np \geq 5$ and $nq \geq 5$.

$$\mu_{\hat{p}} = p$$

$$\sigma_{\hat{p}} = \sqrt{\frac{pq}{n}}$$

Example Seven out of the 46 past US presidents have been left-handed. In the general population, 10% of individuals are left-handed. Test the claim that US presidents are drawn from a population that is more likely to be left-handed than the general population.

Example (Wald Conf. Interval) Suppose a poll of $n = 1000$ randomly selected Canadian voters indicates that $X = 512$ will vote for the Blue political party.

- Determine a 95% confidence interval for the population proportion of Canadians who will vote for the Blue political party.
- Test the hypothesis that *half* of voters will vote for the Blue political party. Use the traditional method.

Notes on Steps 1 and 2 (Claim and Hypotheses)

- The *null* hypothesis always involves *equality*. It gives a *default* answer to the claim/question.
- The *alternative* hypothesis always involves *inequality*.
- Sample statistics (i.e., \hat{p} , \bar{X} , s^2) *never* appear in the hypotheses.

Example Translate each of the following claims into appropriate hypotheses.

- a) The mean age of a BCIT student is 24.2 years.
- b) This coin has at least a 50% chance of coming up heads.
- c) The average Canadian earns less than \$100 000.
- d) The average Canadian has at most 15.5 years of education.

Notes on Step 6 (Stating a Conclusion)

If the original claim was H_0 and we *reject* H_0 :

“The evidence is strong enough to reject the claim that ...”

If the original claim was H_0 and we *fail to reject* H_0 :

“The evidence is not strong enough to reject the claim that ...”

If the original claim was H_1 and we *reject* H_0 :

“The evidence is strong enough to accept the claim that ...”

If the original claim was H_1 and we *fail to reject* H_0 :

“The evidence is not strong enough to accept the claim that ...”

Errors in Hypothesis Testing

Type 1 Error - rejecting the null hypothesis when it is actually true.

Type 2 Error - failing to reject the null hypothesis when it is actually false.

Example In a Canadian court of law, you are presumed *innocent* until proven guilty.

H_0 :

H_1 :

Type 1 error would be

Type 2 error would be

		The Truth	
		H_0 is true	H_0 is false (H_1 is true)
Decision	Fail to Reject H_0		
	Reject H_0		

How can we reduce the probability of making a Type 1 error?

How can we reduce the probability of making a Type 2 error?