

Experimental Methods in Computer Science

Departamento de Engenharia Informática, FCTUC, 2023/2024

Experimental Methods in Computer Science

(Metodologias Experimentais em Informática)

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Hypothesis Testing

Hypothesis testing slides are mainly based on chapter 8 of the book "Essentials of Social Statistics for a Diverse Society"
Second Edition by Anna Leon-Guerrero, Chava Frankfort-Nachmias, SAGE Publications, Inc, 2010.

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Hypothesis testing steps: a more pragmatic approach

Approach already study:

1. State the hypothesis or claim to be tested
2. Select the criteria for a decision (e.g., $\alpha = 0.05$)
3. Compute the test statistic
4. Make a decision

Hypothesis testing steps: a more pragmatic approach

Approach already study:

1. State the hypothesis or claim to be tested
- ~~2. Select the criteria for a decision (e.g., $\alpha = 0.05$)~~
3. Compute the test statistic
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Hypothesis testing steps: a more pragmatic approach

Pragmatic approach:

1. State the hypothesis or claim to be tested
2. Compute the test statistic
3. Obtain p value
4. Make a decision

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Hypothesis testing scenario 1 (test for a mean)

Assume you are the database administrator of a big information system and you are unhappy with the execution time of a given SQL package.

From historical data (thousands of previous package executions), you know that the average execution time of the package is **83.54** seconds with a standard deviation of **16.36**.

You change the tuning of the database and run the package several times to check the effect.

Questions:

- Has the new tuning any effect?
- Is the new configuration better?

That is, is the execution time in the new configuration smaller than in the previous one?

Package exec. time	
74	} 32 times
66	
88	
68	
⋮	
87	
79	
78	
72	
86	
85	
86	

Avg = 78.15

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Example 2 - Step 1: State the hypothesis

(test for a **mean, directional**, known population; normal Z distribution)

- H_0 – The new configuration has no effect on the execution time of the SQL packaged. → The average execution time is 83.54
- H_1 – The execution time of the SQL packaged is **smaller** in the new configuration

We are testing whether the null hypothesis H_0 is true

Note that only the alternate hypothesis changed.

Directional or one-tailed tests are hypothesis tests where the alternative hypothesis is stated as greater than (>) or less than (<) the value stated in the null hypothesis

Example 2 - Step 2: Compute the test statistic

(test for a **mean, directional**, known population; normal Z distribution)

Test statistic:

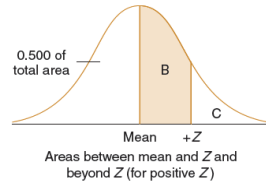
$$Z_c = \frac{M - \mu}{\sigma / \sqrt{n}} = \frac{78.15 - 83.54}{16.36 / \sqrt{32}} = -1.86$$

Example 2 - Step 3: Obtain the p value

(test for a **mean, directional**, known population; normal Z distribution)

Standard normal table example

$Z = 1.86$



To obtain **P value** look for 1.86 in the standard normal table. → the value is 0.0314 → **p = 3.14%**
(As it is a one-tailed we do not double the p)

A	B	C	A	B	C	A	B	C
Z	Area Between Mean and Z	Area Beyond Z	Z	Area Between Mean and Z	Area Beyond Z	Z	Area Between Mean and Z	Area Beyond Z
0.00	0.0000	0.5000	0.11	0.4562	0.4168	0.21	0.0832	0.4168
0.01	0.0040	0.4960	0.12	0.4522	0.4129	0.22	0.0871	0.4129
0.02	0.0080	0.4920	0.13	0.4483	0.4090	0.23	0.0910	0.4090
1.84	0.4671	0.0329	2.24	0.4875	0.0125	2.64	0.4959	0.0041
1.85	0.4678	0.0322	2.25	0.4878	0.0122	2.65	0.4960	0.0040
1.86	0.4686	0.0314	2.26	0.4881	0.0119	2.66	0.4961	0.0039
1.87	0.4693	0.0307	2.27	0.4884	0.0116	2.67	0.4962	0.0038

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Example 2 - Step 4: Make a decision

(test for a **mean, directional**, known population; normal Z distribution)

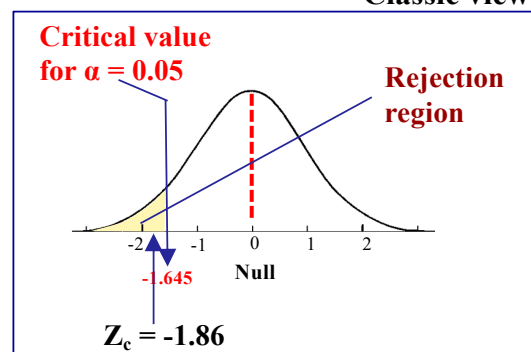
The probability of obtaining $Z_c = -1.86$ is given by the **p value**.

Since **p = 0.0314** this means that the probability of getting an average of 78.15 if H_0 is true is 3.14%

Then I can **reject the H_0** with at least 95% of confidence

The execution time of the SQL packaged is smaller in the new configuration

Classic view



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Types of errors

The conclusion in Step 4 could be wrong, as we are looking at a sample with a limited number n of elements

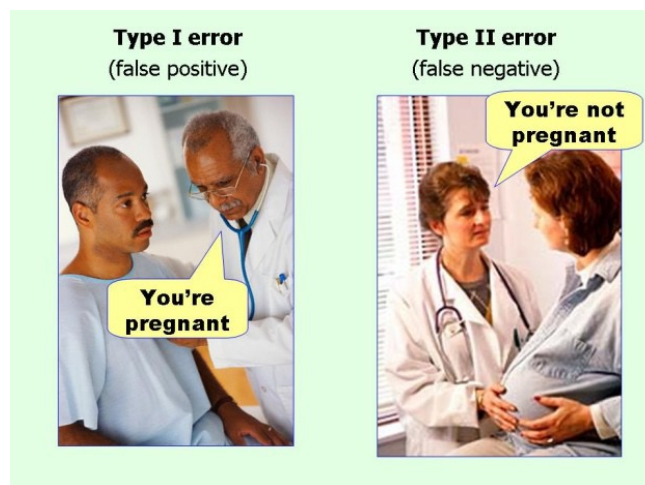
		Decision	
		Retain H_0	Reject H_0
Truth in the population	True	Correct $1 - \alpha$	Type I error α
	False	Type II error β	Correct $1 - \beta$ (Power)

False positive (arrow pointing to Type I error)

False negative (arrow pointing to Type II error)

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Types of errors



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Some errors are worse than others

The incorrect decision is to retain a false null hypothesis. This is equivalent of doing nothing. We can do more experiments and test again the null hypothesis. Not so bad...

		Decision	
		Retain H_0	Reject H_0
Truth in the population	True	Correct $1 - \alpha$	Type I error α
	False	Type II error β	Correct $1 - \beta$ (Power)

False negative

False positive

Some errors are worse than others

The incorrect decision is to reject a true null hypothesis. This means rejecting a previous notions of truth that are in fact true (this is equivalent to finding an innocent person guilty).

		Decision	
		Retain H_0	Reject H_0
Truth in the population	True	Correct $1 - \alpha$	Type I error α
	False	Type II error β	Correct $1 - \beta$ (Power)

False negative

False positive

Note that you can directly control the probability of a Type I error by stating an alpha level

The decision we are looking for

Strong conclusion

This is the decision we are looking for when we test the hypothesis. If we test it, it means we have doubts about such hypothesis.

		Decision	
		Retain H_0	Reject H_0
Truth in the population	True	Correct $1 - \alpha$	Type I error α
	False	Type II error β	Correct $1 - \beta$ (Power)

False positive (arrow to Type I error)

False negative (arrow to Type II error)

Decision we want (arrow to $1 - \beta$)

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The example again: some questions

Assume you are the database administrator of a big information system. You want to change the configuration of the database and run the package several times to see if the new configuration is better.

What should we do if we cannot have a relatively large number of samples?

From historical data (thousands of previous package executions), you know that the average execution time of the package is **83.54** seconds with a standard deviation of **16.36**.

You change the configuration of the database and run the package several times to see if the new configuration is better.

What should we do if we don't know the standard deviation of the population?

Questions.

- Has the new tuning any effect?
- Is the new configuration better?

In this cases we should use the **t Test**

Package exec. time
74
66
88
68
...
79
78
72
86
85
86

33 times

Avg = 78.15

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T-test

- The **t test** follows a Student's T-distribution (if the null hypothesis is true)
 - T-test should be applied when:
 - The **sample size is small** ($n < 30$)
 - The populations' **standard deviation is not known**
- (when the number of samples is large, t test and z test give similar results)

Hypothesis testing using T-test (one sample)

- Follows the same steps as for the Z test
- The critical value comes from the **T table** (considering $n - 1$ degrees of freedom)
- The **test statistics** is now the t-test (similar formula)

$$t_c = \frac{\bar{x} - \mu}{s / \sqrt{n}}$$


Diagram illustrating the components of the t-test formula:

- \bar{x} : Mean of the sample
- μ : Mean of the population (assumed, often stated as a target)
- s : Standard deviation of the sample
- n : Number of elements of the sample

Hypothesis testing steps: pragmatic approach

Steps:

1. State the hypothesis or claim to be tested
2. Compute the test statistic
3. Obtain p value
4. Make a decision

$$t_c = \frac{\bar{x} - \mu}{s / \sqrt{n}}$$


Example 3 - Hypothesis testing using T-test (one sample)

- A professor wants to know if their students are proficient in C programming. The professor wants the class to be able to score above 70 (0-100 scale) on the test (but doesn't want to examine all the students).
- The professor selects 6 students at random from the class and give them a C programming test.
- The six students get scores of 62, 92, 75, 68, 83, and 95.
- **Can the professor have 90% confidence that the mean score for the class on the test would be above 70?**

Example 3: **t** test (one sample)

Step 1- State the hypothesis

- **H₀: $\mu = 70$**

In words: the class knows how to program in C with a proficiency equivalent to 70 in the C programming test

- **H₁: $\mu_1 > 70$**

The class is better on C programming than the score of 70

Example 3: **T** test (one sample)

Step 2 - Compute the test statistic

- Average of the sample: 79.17
- Standard deviation of the sample: 13.17

Test statistic:

$$t_c = \frac{\bar{x} - \mu}{s / \sqrt{n}} = \frac{79.17 - 70}{\frac{13.17}{\sqrt{6}}} = 1.71$$

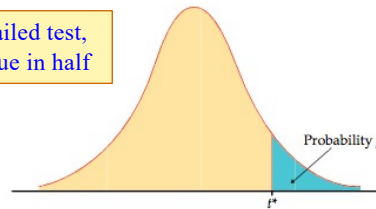
Example 3: T test (one sample)

Step 3 – Obtain p

- Co
- Lo
- tai
- As
- Lo

In a non-directional two-tailed test, we would divide the α value in half

Table entry for p and C is the critical value t^* with probability p lying to its right and probability C lying between $-t^*$ and t^* .



The probability of obtaining $t = 1.71$ is given by the p value. To obtain the P value look for 1.71 in the t table, for $df = 5$

$df = 5$

TABLE D
t distribution critical values

df	Upper-tail probability p										
	.25	.20	.15	.10	.05	.025	.02	.01	.005	.0025	.001
1	1.000	1.376	1.963	3.078	6.314	12.71	15.89	31.82	63.66	127.3	318.3
2	0.816	1.061	1.386	1.886	2.920	4.303	4.849	6.965	9.925	14.09	22.33
3	0.765	0.978	1.250	1.638	2.353	3.182	3.482	4.541	5.841	7.453	10.21
4	0.741	0.941	1.190	1.533	2.132	2.776	2.999	3.747	4.604	5.598	7.173
5	0.728	0.920	1.156	1.476	2.015	2.571	2.767	3.365	4.045	4.773	6.869
6	0.718	0.906	1.134	1.440	1.943	2.447	2.639	3.145	3.707	4.353	6.599
7	0.711	0.896	1.119	1.415	1.895	2.365	2.557	3.055	3.599	4.237	6.449
8	0.706	0.889	1.108	1.397	1.860	2.306	2.492	2.981	3.519	4.179	6.377
9	0.703	0.883	1.100	1.383	1.833	2.262	2.459	2.947	3.472	4.144	6.355
10	0.700	0.879	1.093	1.372	1.812	2.228	2.423	2.912	3.438	4.115	6.328
11	0.697	0.876	1.088	1.363	1.796	2.201	2.398	2.896	3.418	4.097	6.311
12	0.695	0.873	1.083	1.356	1.782	2.179	2.379	2.875	3.398	4.079	6.296
13	0.694	0.870	1.079	1.350	1.771	2.160	2.362	2.858	3.382	4.064	6.282
14	0.692	0.868	1.076	1.345	1.761	2.145	2.347	2.841	3.368	4.050	6.269

→ t score = 1.71

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Example 3: T test (one sample)

Step 4 - Make a decision

The probability of obtaining $t = 1.71$ is given by the p value. To obtain the P value look for 1.71 in the t table, for $df = 5$

→ the P value is between 5% and 10% ($P = 7.4\%$)

As $p < 10\%$ → Reject the null hypothesis(reach significance)

Means that the probability of getting an average score of 79.17 if H_0 is true is 7.4%

Conclusion: The class is better on C programming than the score of 70

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