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Early Human Development xxx (xxxx) xxx-xxx



Contents lists available at ScienceDirect

Early Human Development



journal homepage: www.elsevier.com/locate/earlhumdev

WASP (Write a Scientific Paper) using Excel - 8: t-Tests

ARTICLE INFO

Keywords: Software

Computers

Biostatistics

Statistics

ABSTRACT

t-Testing is a common component of inferential statistics when comparing two means. This paper explains the central limit theorem and the concept of the null hypothesis as well as types of errors. On the practical side, this paper outlines how different *t*-tests may be performed in Microsoft Excel, for different purposes, both statically as well as dynamically, with Excel's functions.

1. Central limit theorem

A parametric statistical test is one that assumes that the data being analysed is normally distributed. The Central Limit theorem states that a substantial sample tends to be normally distributed, and that the mean of all such theoretical samples (were it possible to repeatedly sample the population) will be approximately equal to the population mean, even when the population from where it is taken is not very normally distributed. This allows a parametric test to be carried out in order to compare two means, using the t-distribution, as outlined in the previous paper in this series [1]. It is thanks to this fact that a panoply of parametric tests have evolved. Compared to non-parametric tests which do not invoke the Central Limit theorem, parametric tests are more accurate and precise, with higher statistical powers [2].

2. The null hypothesis

The null hypothesis is an a priori statement of impartiality, prior to analysis, which states that when testing two samples, it is assumed that these come from the same population and that they are not statistically significantly different. This is stated as:

H₀: $\mu 1 - \mu 2 = 0$, where $\mu 1$ represents one sample and $\mu 2$ represents the second sample.

The pre-established threshold or cut-off (known as α) to demonstrate a difference in the biological sciences is taken at 5% or significance testing for $p \le 0.05$, which is the 1.96 standard deviations that was previously discussed [3]. However, this implies that if testing were to be done a hundred times on random samples, 5% would be erroneously shown to be different despite there being any true difference, due to chance alone.

Statistical testing may thus lead to the rejection of the null hypothesis (the statistical test finds a difference, $p \le 0.05$, i.e. the test generates a 95% likelihood that results do not fit the null hypothesis) when the null hypothesis is true. Or the test may fail to reject the null hypothesis (p > 0.05) when it is false.

3. Errors

The statements above imply an inevitable lack of certitude and the possibility of errors, of which there are two types.

A Type 1 error (also known as an "error of the first kind", an α error) is a false positive result, a statistically significant result which does not represent reality. This thus causes the rejection of the null hypothesis when the null hypothesis is true since testing results in the claim that there is difference between two samples. At the accepted 5% level, this may naturally occur 5 times out of a hundred (tests) simple due to chance, the random nature of sampling. This is thus also referred to as an error of excessive credulity (Table 1). This typically occurs when the sample population is not representative of the population from which it was chosen.

A Type 2 error (also known as an "error of the second kind", a β error) is a false negative result, a statistically non-significant result which does not represent reality. This thus causes the acceptance of the null hypothesis, when the null hypothesis is false since testing results in the claim that there no difference between two samples (Table 1). Typically type 2 errors occur because the study is "underpowered" or had too few subjects to investigate a particular question.

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Table 1

Types of errors.

		Reality	Reality		
		H ₀ true	H ₀ false		
Test Result	Accept H _o Reject H _o	Accept H _o Type 1 error	Type 2 error Reject H ₀		

	A	В	C D	E	F	G	Н	1	J	K	L
1	HtFt	HtBk							t-Test: Two-Sample Assuming	Unequal Va	riances
2	192.9	185.3					-	×			
3	198.1	200.6	Data Analysis 🧳 🔜					×		HtFt	HtBk
4	198.1	190.5	Analysis Tools				OK	Mean	190.2773	195.8574	
5	190.5	200.6	Fourier Analysis						Variance	117.9018	99.86091
6	198.1	190.5	Histogram Cancel Moving Average Random Number Generation Rank and Percentile Regression Sampling					el	Observations	26	26
7	192.9	180.4						Hypothesized Mean Difference df t Stat	0		
8	190.5	213.4							df	50	
9	188.1	195.4							t Stat	-1.92816	
10	195.7	205.7	t-Test: Two-San	t-Test: Paired Two Sample for Means t-Test: Two-Sample Assuming Equal Variances					P(T<=t) one-tail	0.029762	
11	192.9	190.5	t-Test: Two-San	iple Assumin	g Unequal \	/ariances		t Critical one-tail	1.675905		
12	195.7	182.9	· · · · · · · · · · · · · · · · · · ·	\					P(T<=t) two-tail	0.059523	
13	200.6	210.9							t Critical two-tail	2.008559	
14	185.3	208.2									
15	200.6	200.6	t-Test: Two-Sa	mple Assu	iming Une	equal Varia	ances ?	×			
16	198.1	195.4	Input				_				
17	195.7	203.3	Variable 1 Range		SAS1:SAS27	1	C	ок			
18	190.5	203.3	Variable 2 Range		CPC1-CPC27		Ca	ncel			
19	203.3	175.3	Tunubic E hunge	L	3031.30327	HM		Hala			
20	180.1	190.5	Hypoth <u>e</u> sized Me	an Difference	8		eip				
21	182.9	190.5	✓ <u>L</u> abels								
22	177.7	198.1	<u>A</u> lpha: 0.05								
23	182.9	182.9	Output antions								
24	177.7	210.9	Output options		C1C1	EK:					
25	154.8	190.5		· [100	6.24					
26	205.7	195.7	O New Workshe	et <u>P</u> ly:							
27	177.7	200.6	O New Workboo)K							
28											

Fig. 1. t-Testing American football (HtFt) and basketball players (HtBk) using Excel's Data Analysis Toolpak.

4. Other fields of science

Other science fields are far more stringent. For example, in high-energy physics, the threshold of "evidence of a particle" detection is taken at p = 0.003 and "discovery" of a particle at p = 0.0000003, equivalent to five standard deviations, such that the test would have to be repeated 3.5 million times for a similar result to occur just once by chance alone [4].

5. t-Testing

Testing involves comparing means and since the *t*-test is a parametric test, it assumes that the data is approximately normally distributed. Three types of *t*-tests are available within Excel from the Data Tab-Data Analysis option, as shown in Fig. 1 which displays the analysis of a random selection of athletes from two groups: American football and basketball players. In this example, there are 26 players in each group and the null hypothesis states: H_0 = height of football players (HtFt) minus height of basketball players (HtBk) = 0.

In practice, the variance of the population being studied is very rarely known with any certainty. Most data is analysed using the option "*t*-test: Two-Sample Assuming Unequal Variances". Clicking on this option opens the dialog box at the bottom of Fig. 1 which should be filled in as shown in order to perform the test. Please note that Labels is ticked since each column contains a header row, and that alpha is preset at the conventional 0.05. The generated results are shown in the last three columns and are static – results will not change if the original data is changed.

The results additionally provide some descriptive statistics: the means of both samples, their variances, and the total number of observations. These are followed by the t statistic obtained for the test, the value for a one-tailed and a two-tailed test, and the critical values for both.

6. One- vs two-tailed tests

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In this case, p (two-tailed) = 0.0595, so the H<sub>0</sub> is accepted as the test has not reached statistical significance at the 0.05 (5%) level. The two-tailed
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