

# Statistical notes for clinical researchers: Chi-squared test and Fisher's exact test

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When we try to compare proportions of a categorical outcome according to different independent groups, we can consider several statistical tests such as chi-squared test, Fisher's exact test, or z-test. The chi-squared test and Fisher's exact test can assess for independence between two variables when the comparing groups are independent and not correlated. The chi-squared test applies an approximation assuming the sample is large, while the Fisher's exact test runs an exact procedure especially for small-sized samples.

## Chi-squared test

### 1. Independency test

The chi-squared test is used to compare the distribution of a categorical variable in a sample or a group with the distribution in another one. If the distribution of the categorical variable is not much different over different groups, we can conclude the distribution of the categorical variable is not related to the variable of groups. Or we can say the categorical variable and groups are independent. For example, if men have a specific condition more than women, there is bigger chance to find a person with the condition among men than among women. We don't think gender is independent from the condition. If there is equal chance of having the condition among men and women, we will find the chance of observing the condition is the same regardless of gender and can conclude their relationship as independent. Examples 1 and 2 in Table 1 show perfect independent relationship between condition (A and B) and gender (male and female), while example 3 represents a strong association between them. In example 3, women had a greater chance to have the condition A ( $p = 0.7$ ) compared to men ( $p = 0.3$ ).

Table 1. Contingency tables and procedure for Chi-squared test

		Observed frequency (O)			Expected frequency (E)			O - E			(O - E) <sup>2</sup> / E		
		A	B	Total	A	B	Total	A	B	Total	A	B	Total
Example 1	Male	50	50	100	50	50	100	0	0	0	0	0	0
	Female	50	50	100	50	50	100	0	0	0	0	0	0
	Total	100	100	200	100	100	200	0	0	0	0	0	0
Example 2	Male	30	70	100	30	70	100	0	0	0	0	0	0
	Female	30	70	100	30	70	100	0	0	0	0	0	0
	Total	60	140	200	60	140	200	0	0	0	0	0	0
Example 3	Male	30	70	100	50	50	100	-20	20	0	8	8	16
	Female	70	30	100	50	50	100	20	-20	0	8	8	16
	Total	100	100	200	100	100	200	0	0	0	16	16	32

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The chi-squared test performs an independency test under following null and alternative hypotheses,  $H_0$  and  $H_1$ , respectively.

$H_0$ : Independent (no association)

$H_1$ : Not independent (association)

The test statistic of chi-squared test:  $\chi^2 = \sum \frac{(O-E)^2}{E} \sim \chi^2$  with degrees of freedom  $(r-1)(c-1)$ ,

Where O and E represent observed and expected frequency, and r and c is the number of rows and columns of the contingency table.

The first step of the chi-squared test is calculation of expected frequencies (E). E is calculated under the assumption of independent relation or, in other words, no association. Under independent relationship, the cell frequencies are determined only by marginal proportions, i.e., proportion of A ( $60/200 = 0.3$ ) and B ( $1400/200 = 0.7$ ) in example 2. In example 2, the expected frequency of the male and A cell is calculated as 30 that is the proportion of 0.3 (proportion of A) in 100 Males. Similarly, the expected frequency of the male and A cell is 50 that is the proportion of 0.5 (proportion of A =  $100/200 = 0.5$ ) in 100 Males in example 3 (Table 1).

$$\text{Expected frequency (E) of Male \& A} = \frac{\text{Number of A} * \text{Number of Male}}{\text{Total number}} = p(A) * p(\text{male}) * \text{total number}$$

The second step is obtaining  $(O - E)^2/E$  for each cell and summing up the values over each cell. The final summed value follows chi-squared distribution. For the 'male and A' cell in example 3,  $(O - E)^2/E = (30 - 50)^2/50 = 8$ . Chi-squared statistic calculated =  $\sum \frac{(O-E)^2}{E} = 8 + 8 + 8 + 8 = 32$  in example 3. For examples 1 and 2, the chi-squared statistics equal zero. A big difference between observed value and expected value or a large chi-squared statistic implies that the assumption of independency applied in calculation of expected value is irrelevant to the observed data that is being tested. The degrees of freedom is one as the data has two rows and two columns:  $(r-1) * (c-1) = (2-1) * (2-1) = 1$ .

The final step is making conclusion referring to the chi-squared distribution. We reject the null hypothesis of independence if the calculated chi-squared statistic is larger than the critical value from the chi-squared distribution. In the chi-squared distribution, the critical values are 3.84, 5.99, 7.82, and 9.49, with corresponding degrees of freedom of 1, 2, 3, and 4, respectively, at an alpha level of 0.5. Larger chi-square statistics than these critical values of specific corresponding degrees of freedom lead to the rejection of null hypothesis of independence. In examples 1 and 2, the chi-squared statistic is zero which is smaller than the critical value of 3.84, concluding independent relationship between gender and condition. However, data in example 3 have a large chi-squared statistic of 32 which is larger than 3.84; it is large enough to reject the null hypothesis of independence, concluding a significant association between two variables.

The chi-squared test needs an adequate large sample size because it is based on an approximation approach. The result is relevant only when no more than 20% of cells with expected frequencies < 5 and no cell have expected frequency < 1.<sup>1</sup>

## 2. Effect size

As the significant test does not tell us the degree of effect, displaying effect size is helpful to show the magnitude of effect. There are three different measures of effect size for chi-squared test, Phi ( $\phi$ ), Cramer's V (V), and odds ratio (OR). Among them  $\phi$  and OR can be used as the effect size only in 2 x 2 contingency tables, but not for bigger tables.

$$\phi = \sqrt{\frac{\chi^2}{n}}$$

$$V = \sqrt{\frac{\chi^2}{n \cdot df}}, \text{ where } n \text{ is total number of observation, and } df \text{ is degrees of freedom calculated by } (r-1) * (c-1). \text{ Here, } r \text{ and } c \text{ are the numbers of rows and columns of the contingency table.}$$

In example 3, we can calculate them as  $\phi = \sqrt{\frac{\chi^2}{n}} = \sqrt{\frac{32}{200}} = 0.4$ ,  $V = \sqrt{\frac{\chi^2}{n \cdot df}} = \sqrt{\frac{32}{200 \cdot 1}} = 0.4$ , and  $OR = \frac{70 \cdot 70}{30 \cdot 30} = 5.44$ . Referring

to Table 2, the effect size  $V = 0.4$  is interpreted medium to large. If number of rows and/or columns are larger than 2, only Cramer's V is available.

**Table 2.** Effect size for chi-squared test, Cramer's V and its interpretation<sup>2</sup>

Degree of freedom	Small	Medium	Large
1	0.10	0.30	0.50
2	0.07	0.21	0.35
3	0.06	0.17	0.29
4	0.05	0.15	0.25
5	0.04	0.13	0.22

### 3. *Post-hoc* pairwise comparison of chi-squared test

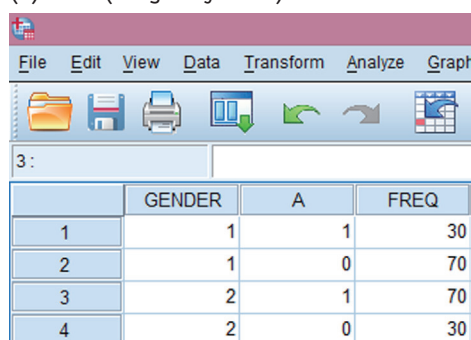
The chi-squared test assesses a global question whether relation between two variables is independent or associated. If there are three or more levels in either variable, a *post-hoc* pairwise comparison is required to compare the levels of each other. Let's say that there are three comparative groups like control, experiment 1, and experiment 2 and we try to compare the prevalence of a certain disease. If the chi-squared test concludes that there is significant association, we may want to know if there is any significant difference in three compared pairs, between control and experiment 1, between control and experiment 2, and between experiment 1 and experiment 2. We can reduce the table into multiple 2 x 2 contingency tables and perform the chi-squared test with applying the Bonferroni corrected alpha level (corrected  $\alpha = 0.05/3$  compared pairs = 0.017).

#### Fisher's exact test

Fisher's exact test is practically applied only in analysis of small samples but actually it is valid for all sample sizes. While the chi-squared test relies on an approximation, Fisher's exact test is one of exact tests. Especially when more than 20% of cells have expected frequencies < 5, we need to use Fisher's exact test because applying approximation method is inadequate. Fisher's exact test assesses the null hypothesis of independence applying hypergeometric distribution of the numbers in the cells of the table. Many packages provide the results of Fisher's exact test for 2 x 2 contingency tables but not for bigger contingency tables with more rows or columns. For example, the SPSS statistical package automatically provides an analytical result of Fisher's exact test as well as chi-squared test only for 2 x 2 contingency tables. For Fisher's exact test of bigger contingency tables, we can use web pages providing such analyses. For example, the web page 'Social Science Statistics' (<http://www.socscistatistics.com/tests/chisquare2/Default2.aspx>) permits performance of Fisher exact test for up to 5 x 5 contingency tables.

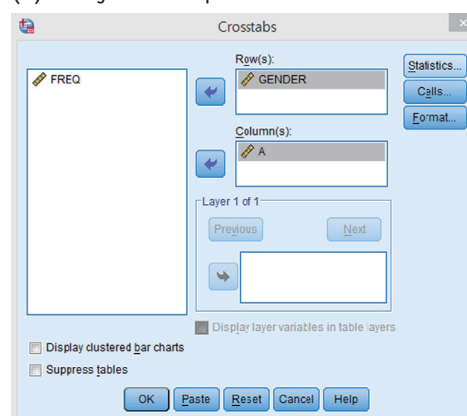
The procedure of chi-squared test and Fisher's exact test using IBM SPSS Statistics for Windows Version 23.0 (IBM Corp., Armonk, NY, USA) is as follows:

(a) Data (weight by FREQ)



	GENDER	A	FREQ
1	1	1	30
2	1	0	70
3	2	1	70
4	2	0	30

(b) Analyze-Descriptive Statistics -Crosstabs



(b-1) Statistics

(b-2) Cells

(c) Crosstable

GENDER * A Crosstabulation					
		A		Total	
		0	1		
GENDER	1	Count	70	30	100
		Expected Count	50.0	50.0	100.0
		% within A	70.0%	30.0%	50.0%
		% of Total	35.0%	15.0%	50.0%
	2	Count	30	70	100
		Expected Count	50.0	50.0	100.0
		% within A	30.0%	70.0%	50.0%
		% of Total	15.0%	35.0%	50.0%
Total		Count	100	100	200
		Expected Count	100.0	100.0	200.0
		% within A	100.0%	100.0%	100.0%
		% of Total	50.0%	50.0%	100.0%

(d) Chi-square test

**Chi-Square Tests**

	Value	df	Asymp. Sig. (2-sided)	Exact Sig. (2-sided)	Exact Sig. (1-sided)
Pearson Chi-Square	32.000 <sup>a</sup>	1	.000		
Continuity Correction <sup>b</sup>	30.420	1	.000		
Likelihood Ratio	32.913	1	.000		
Fisher's Exact Test				.000	.000
Linear-by-Linear Association	31.840	1	.000		
N of Valid Cases	200				

a. 0 cells (.0%) have expected count less than 5. The minimum expected count is 50.00.  
b. Computed only for a 2x2 table

(e) Effect size: Phi & Cramer's V

**Symmetric Measures**

		Value	Approx. Sig.
Nominal by Nominal	Phi	.400	.000
	Cramer's V	.400	.000
N of Valid Cases		200	

(f) Effect size: odds ratio

**Risk Estimate**

	Value	95% Confidence Interval	
		Lower	Upper
Odds Ratio for GENDER (1 / 2)	5.444	2.973	9.969
For cohort A = 0	2.333	1.685	3.232
For cohort A = 1	.429	.309	.594
N of Valid Cases	200		

## References

1. Bewick V, Cheek L, Ball J. Statistics review 8: qualitative data - tests of association. *Crit Care* 2004;8:46-53.
2. Cohen J. Statistical power and analysis for the behavioral sciences. 2nd ed. Hisdale, NJ: Lawrence Erlbaum Associates; 1988. p79-80.