

2x2 Contingency Tables and Test for 2 Independent Proportions

For a 2x2 contingency table, the χ^2 test of independence gives the same conclusion as the significance test for comparing two independent proportions. Why? How are these related?

	Y	N
Group 1		
Group 2		

Sig Test for $p_1 - p_2$:

$$H_0: p_1 - p_2 = 0$$

$$H_a: p_1 - p_2 \neq 0$$

χ^2 test :

$$H_0: \text{Independence}$$

$$H_a: \text{Association}$$

Equivalent

Example: Suppose we want to check if Gender is independent of Vegetarianism. We collect the following data on 679 randomly selected individuals.

Gender	Vegetarian?		Total
	Yes	No	
Male	5	154	159
Female	38	482	520
Total	43	636	679

Assumptions and Hypotheses for χ^2 test of independence:

- SRS - 679 ppl random + representative of population of interest
- Min 5 expected counts under H_0 in every cell

Assumptions and Hypotheses for 2 independent proportions test:

- SRS - same
- 10 succ + fail expected under H_0 for each group

Example: The Physician's Health Study I was a randomized, double-blind, placebo-controlled trial whose participants were all male doctors in the USA. It studied the relationship between heart attacks (Y/N) and taking aspirin (Y/N). One treatment group took Aspirin, while another took a placebo daily, for several years. <https://phs.bwh.harvard.edu/phs1.htm>

The results are summarized in the contingency table below:

	Heart Attack?		
	Yes	No	Total
Placebo	189	10,845	11,034
Aspirin	104	10,933	11,037
Total	293	21,778	22,071

- Have the assumptions been met to conduct a χ^2 test for independence?
 - SRS of participants (Not representative of all American adults, at best to males, doctors (or maybe well-educated + rich, higher interest in health))
 - min 5 expected per cell \rightarrow check obs.
- State the hypotheses for this problem.

H_0 : No association between Aspirin and Heart attack

H_a : Association

- Conduct the statistical analysis and state the conclusion.

Pearson's TS: $\chi^2 = 25.014$

p-value ≈ 0 from χ^2 distribution with 1 df.

Rej H_0 at all usual α 's \rightarrow Very strong evidence of association

- Compute Conditional Probabilities to describe the association. between Aspirin and Heart attack.

% of Aspirin had heart attacks: $104/11037 = .0094 = 0.94\%$

% of Placebo had heart attacks: $189/11034 = .0171 = 1.71\%$

- Compute the Relative Risk of heart attack for the placebo group vs aspirin.

$$RR = \frac{\hat{p}_1}{\hat{p}_2} = \frac{1.71\%}{0.94\%} = 1.82$$

(1.82 times)

Interpretation: Ppl who take placebo are almost twice as likely as ppl who take Aspirin of having a heart attack

Not all ppl though !!

Identify the test statistics, p-values, and draw conclusions based on the Minitab Output below:

Chi-Square Test: Veg, Not

Expected counts are printed below observed counts

Chi-Square contributions are printed below expected counts

M →

F →

	Veg	Not	Total
1	5 10.07 2.552	154 148.93 0.173	159
2	38 32.93 0.780	482 487.07 0.053	520
Total	43	636	679

Chi-Sq = 3.558, DF = 1, P-Value = 0.059

Test and CI for Two Proportions

M

F

Sample	X	N	Sample p
1	5	159	0.031447
2	38	520	0.073077

$\hat{p}_1 = 5/159$
 $\hat{p}_2 = 38/520$

Difference = p (1) - p (2)

Estimate for difference: -0.0416304

95% CI for difference: (-0.0767909, -0.00646989)

Test for difference = 0 (vs not = 0): Z = -1.89 P-Value = 0.059

Fisher's exact test: P-Value = 0.063

$$\hat{p}_1 - \hat{p}_2 = \frac{5}{159} - \frac{38}{520} = -0.0416304$$

95% CI for $p_1 - p_2$:

$$\hat{p}_1 - \hat{p}_2 \pm Z_{0.025} \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$$

$$= \frac{5}{159} - \frac{38}{520} \pm 1.96 \times \sqrt{\frac{5}{159} \times (1 - \frac{5}{159}) + \frac{38}{520} \times (1 - \frac{38}{520})}$$

$$= (-0.0767909, -0.00646989)$$

Test Statistic:

$$TS (Z) = \frac{\hat{p}_1 - \hat{p}_2 - 0}{\sqrt{\hat{p}(1-\hat{p})(\frac{1}{n_1} + \frac{1}{n_2})}} = \frac{\frac{5}{159} - \frac{38}{520}}{\sqrt{\frac{43}{679} \times (1 - \frac{43}{679}) \times (\frac{1}{159} + \frac{1}{520})}}$$

$$= -1.89$$

$$P\text{-value} = 2P(Z \leq -1.89) \xrightarrow{\text{z-table}} 0.059$$

$$\text{exp: } \frac{43 \times 159}{679} = 10.07 \quad \frac{636 \times 159}{679} = 148.93$$

$$\frac{43 \times 520}{679} = 32.93 \quad \frac{636 \times 520}{679} = 487.07$$

Contribution to TS:

$$\frac{(5-10.07)^2}{10.07} = 2.552$$

$$\frac{(154-148.93)^2}{148.93} = 0.173$$

$$\frac{(38-32.93)^2}{32.93} = 0.780$$

$$\frac{(482-487.07)^2}{487.07} = 0.053$$

$$\chi^2 = 2.552 + 0.173 + 0.780 + 0.053 = 3.558$$

$$DF = (2-1) \times (2-1) = 1$$

$$P\text{-value} = P(\chi^2 \geq 3.558) \xrightarrow{\text{use software}} 0.059$$

a random variable that follows χ^2 distribution with degrees of freedom 1.

$$(-1.89)^2 = 3.558$$

$$\star Z^2 = \chi^2$$

$$\star t_n^2 = F_{1,n}$$

$$\chi_n^2 = Z_1^2 + Z_2^2 + \dots + Z_n^2$$

$$Z_i \perp Z_j, (i \neq j)$$

$$Z_i \sim N(0,1)$$

$$t_n = \frac{X \sim N(0,1)}{\sqrt{Y/n}} \quad X \perp Y$$

$$\rightarrow \chi_n^2 \rightarrow \chi_m^2$$

$$F_{m,n} = \frac{X/m}{Y/n} \rightarrow \chi_m^2 \perp \chi_n^2$$

Extra Credit (0.5pts)

Example: Collect data in class, or use the General Social Survey to answer a question of interest involving two categorical variables.

Variables: Sexuality & Dating apps

		Dating apps		
		No	Yes	
Sexuality	Hetero	80	40	120
	Not hetero	7	17	24
		87	57	144

• Assumptions:

• H_0 :

H_a :

• TS:

• DF:

• p-value:

• conclusion:

- Conditional Prob: % who use Dating Apps for $\left\{ \begin{array}{l} \text{heterosexuals;} \\ \text{not;} \end{array} \right.$
- Marginal distributions for sexuality and Dating apps usage:
- RR of using dating apps for the heterosexuals v.s. not heterosexuals:
Interpretation of the RR:
- Joint distribution:

Deadline: March 9th 11:59pm ~~10:55am~~ (No email; No paper)

Upload a pdf file on Canvas "Extra Credit 2"

Nonparametric Statistics (Ch. 15)

Overview of Nonparametric Methods

- Normal-based procedures (Z and t tests) assume that the statistic has an approximately Normal distribution, either because the sample sizes are large enough or because the original distribution is Normal.
- When utilizing Nonparametric methods, we still have a parameter but we have no distribution assumptions for the response variable. In order to use these methods, our response variable Y must be quantitative.
- Nonparametric methods make inferences about: Populations
But not about means, instead inferences about median or entire distribution.
- Nonparametric methods are useful when:
 - small sample size (with outliers)
 - several groups with very different variances
 - data is quantitative but NOT continuous - particularly subjective ratings
- Nonparametric methods generally only need to assume:
 - SRS from population of interest

Remarks:

- Almost All the nonparametric methods we see in this class are based on RANKS.
- If assumptions of Normality and Equal Variances are met, then both parametric (normal-based) and nonparametric (distribution-free) methods can be used.
- However, parametric methods are **statistically more powerful** and are able to find true significant differences (assuming all assumptions are met), while nonparametric methods sometimes cannot.

$$\text{power} = 1 - \beta \quad , \quad \beta = P(\text{Not rej } H_0 \mid H_0 \text{ is true})$$

↓
Type II error