

# Chi-square test

Grzegorz Karas

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## 1 Definition [1, 2]

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The  $\chi^2$ -test is a statistical method to test hypothesis, where random variable follows multinomial distribution. It tests a null hypothesis stating that the frequency distribution of certain events observed in an observer sample is consistent with a particular theoretical distribution.

The most common  $\chi^2$ -test is a Pearson's chi-square test in which the test statistic is of following kind

$$\chi^2 = \sum_{i=1}^n \frac{(O_i - E_i)^2}{E_i} = N \sum_{i=1}^n \frac{(O_i/N - p_i)^2}{p_i} \quad (1)$$

where

- $\chi^2$  - Pearson's cumulative test statistic, which asymptotically approaches a  $\chi^2$  distribution.
- $O_i$  - number of observation of type i
- $N$  - total number of observations
- $E_i = Np_i$  - the expected (theoretical) count of type i
- $n$  - the number of cells in the table (*rows* · *columns*)

If the test fails, then the appropriate test statistic has approximately a noncentral  $\chi^2$  distribution with the same degrees of freedom (*df*) and a noncentrality parameter  $\lambda$ , which depends on alternative considered.

## 2 Test types [1]

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Pearson's chi-squared test is used to assess three types of hypothesis testing: goodness of fit, homogeneity, and independence.

- A test of **goodness of fit** establishes whether an observed frequency distribution differs from a theoretical distribution.  
The  $\chi^2$  test statistic follows the  $\chi^2$  distribution with  $df = n - 1$ .
- A test of **homogeneity** compares the distribution of counts for two or more groups using the same categorical variable.  
The  $\chi^2$  test statistic follows the  $\chi^2$  distribution with  $df = (rows - 1) \cdot (columns - 1)$ .
- A test of **independence** assesses whether observations consisting of measures on two variables, expressed in a contingency table, are independent of each other.  
The  $\chi^2$  test statistic follows the  $\chi^2$  distribution with  $df = (rows - 1) \cdot (columns - 1)$ .

### 2.1 Test of goodness of fit

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Let  $X = (X_1, \dots, X_k)$  be a multinomial random variable with parameters  $n, p_1, \dots, p_k$ . Suppose we wish to test

$$H_0 : p_i = p_i^0 \quad i = 1, 2, \dots, k \quad (2)$$

against

$$H_a : \text{not all } p\text{'s are as given by } H_0 \quad (3)$$

where the  $p'_i$  are given expected numbers. The value of the chi-square test-statistic is

$$\chi^2_{H_0} = \sum_{i=1}^k \frac{(x_i - np_i^0)^2}{np_i^0} = \sum_{i=1}^k \frac{(\hat{p}_i - p_i^0)^2}{p_i^0} \quad (4)$$

The chi-square test reject  $H_0$  if

$$\chi^2_{H_0} > \chi^2_{k-1, 1-\alpha}, \quad (5)$$

where  $\alpha$  is the significance level, and  $\chi^2_{k-1, 1-\alpha}$  is the quantile of order  $1 - \alpha$  of the central  $\chi^2$  distribution with  $k - 1$  degrees of freedom. The p-value of the test is

$$p - \text{value} = P(X > \chi^2_{H_0}) \quad (6)$$

To evaluate the power of the test let's precisely define the alternative  $H_a$  as follow.

$$H_a : \quad p_i = p_i^a \quad i = 1, 2, \dots, k \quad (7)$$

Thus the power of the test is

$$\text{Power} = P^\lambda(X > \chi^2_{k-1, 1-\alpha}) \quad (8)$$

where  $X$  is a random variable that follows the noncentral  $\chi^2$  distribution with the noncentrality parameter

$$\lambda = n \sum_{i=1}^k \frac{(p_i^a - p_i^0)^2}{p_i^0} \quad (9)$$

## 2.2 Test of independence

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Let  $X = (X_{ij}) \in \mathbb{R}^{r \times c}$  be a multinomial random variable with parameters  $n, p_{ij}$  where  $i = 1, 2, \dots, r$ ,  $j = 1, 2, \dots, c$  and  $\sum_{i=1}^r \sum_{j=1}^c p_{ij} = 1$ . Suppose we wish to test independence

$$H_0 : \quad p_{ij} = p_{i \cdot} p_{\cdot j} \quad i = 1, 2, \dots, r \quad j = 1, 2, \dots, c \quad (10)$$

against

$$H_a : \text{not all the equations given under } H_0 \text{ are satisfied} \quad (11)$$

where  $p_{i \cdot} = \sum_{j=1}^c p_{ij}$  and  $p_{\cdot j} = \sum_{i=1}^r p_{ij}$ . The value of the chi-square test-statistic is

$$\chi^2_{H_0} = \sum_{i=1}^r \sum_{j=1}^c \frac{(x_{ij} - x_{i \cdot} x_{\cdot j} / n)^2}{x_{i \cdot} x_{\cdot j} / n} \quad (12)$$

where  $x_{i \cdot} = \sum_{j=1}^c x_{ij}$  and  $x_{\cdot j} = \sum_{i=1}^r x_{ij}$ . We observe that

$$\chi^2_{H_0} = \frac{1}{n} \sum_{i=1}^r \sum_{j=1}^c \frac{\left( \frac{x_{ij}}{n} - \frac{x_{i \cdot}}{n} \frac{x_{\cdot j}}{n} \right)^2}{\frac{x_{i \cdot}}{n} \frac{x_{\cdot j}}{n}} = \frac{1}{n} \sum_{i=1}^r \sum_{j=1}^c \frac{(\hat{p}_{ij} - \hat{p}_{i \cdot} \hat{p}_{\cdot j})^2}{\hat{p}_{i \cdot} \hat{p}_{\cdot j}} \quad (13)$$

The chi-square test reject  $H_0$  if

$$\chi^2_{H_0} > \chi^2_{(r-1) \cdot (c-1), 1-\alpha}, \quad (14)$$

To evaluate the power of the test let's precisely define the alternative  $H_a$  as follow.

$$H_a : \quad p_{ij} = \underbrace{p_{i \cdot} p_{\cdot j}}_{p_{ij}^a} + \frac{c_{ij}}{\sqrt{n}}, \quad i = 1, 2, \dots, r \quad j = 1, 2, \dots, c, \quad \text{where} \quad \sum_{i=1}^r \sum_{j=1}^c c_{ij} = 0, \quad (15)$$

Thus the power of the test is

$$Power = P^\lambda \left( X > \chi_{(r-1) \cdot (c-1), 1-\alpha}^2 \right) \quad (16)$$

where  $X$  is a random variable that follows the noncentral  $\chi^2$  distribution with the noncentrality parameter

$$\lambda = \sum_{i=1}^r \sum_{j=1}^c \frac{c_{ij}^2}{p_{i \cdot} p_{\cdot j}} - \sum_{i=1}^r \frac{c_{i \cdot}^2}{p_{i \cdot}} - \sum_{j=1}^c \frac{c_{\cdot j}^2}{p_{\cdot j}}, \quad (17)$$

where  $c_{i \cdot} = \sum_{j=1}^c c_{ij}$  and  $c_{\cdot j} = \sum_{i=1}^r c_{ij}$ .

If  $\Delta_{ij} = c_{ij}/\sqrt{n}$ , then

$$\lambda = n \left[ \sum_{i=1}^r \sum_{j=1}^c \frac{\Delta_{ij}^2}{p_{i \cdot} p_{\cdot j}} - \sum_{i=1}^r \frac{\Delta_{i \cdot}^2}{p_{i \cdot}} - \sum_{j=1}^c \frac{\Delta_{\cdot j}^2}{p_{\cdot j}} \right], \quad (18)$$

## 2.3 Test of homogeneity

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Let  $X_i = (X_{ij}) \in \mathbb{R}^c$  be a multinomial random variable with parameters  $n_i, p_{ij}$  for  $i = 1, 2, \dots, r$  and  $\sum_{j=1}^c p_{ij} = 1$ . Suppose we wish to test homogeneity

$$H_0 : p_{1j} = p_{2j} = \dots = p_{rj} = p_{\cdot j} \quad j = 1, 2, \dots, c \quad (19)$$

against

$$H_a : \text{not all the equatons given under } H_0 \text{ are satisfied} \quad (20)$$

The value of a chi-square test-statistic is

$$\chi_{H_0}^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(x_{ij} - x_{i \cdot} x_{\cdot j} / n)^2}{x_{i \cdot} x_{\cdot j} / n} \quad (21)$$

where  $x_{i \cdot} = \sum_{j=1}^c x_{ij} = n_i$  and  $n = \sum_{i=1}^r n_i$ . The chi-square test reject  $H_0$  if

$$\chi_{H_0}^2 > \chi_{(r-1) \cdot (c-1), 1-\alpha}^2, \quad (22)$$

To evaluate the power of the test let's precisely define the alternative  $H_a$  as follow.

$$H_a : p_{ij} = p_{\cdot j} + \underbrace{\frac{c_{ij}}{\sqrt{n}}}_{p_{ij}^a}, \quad j = 1, 2, \dots, c \quad \text{where} \quad \sum_{j=1}^c c_{ij} = 0, \quad (23)$$

Thus the power of the test is

$$Power = P^\lambda \left( X_a > \chi_{(r-1) \cdot (c-1), 1-\alpha}^2 \right) \quad (24)$$

where  $X_a$  is a random variable that follows the noncentral  $\chi^2$  distribution with the noncentrality parameter

$$\lambda = \sum_{j=1}^c \frac{1}{p_{\cdot j}} \left[ \sum_{i=1}^r c_{ij}^2 \frac{n_i}{n} - \left( \sum_{i=1}^r c_{ij} \frac{n_i}{n} \right)^2 \right], \quad (25)$$

If  $\Delta_{ij} = c_{ij}/\sqrt{n}$ , then

$$\lambda = n \sum_{j=1}^c \frac{1}{p_{\cdot j}} \left[ \sum_{i=1}^r \Delta_{ij}^2 \frac{n_i}{n} - \left( \sum_{i=1}^r \Delta_{ij} \frac{n_i}{n} \right)^2 \right] \quad (26)$$

It is worth to observe that  $\frac{n_i}{n}$  is the same as  $p_i$ . and the equation holds following form

$$\lambda = n \sum_{j=1}^c \frac{1}{p_{\cdot j}} \left[ \sum_{i=1}^r \Delta_{ij}^2 p_{i\cdot} - \left( \sum_{i=1}^r \Delta_{ij} p_{i\cdot} \right)^2 \right] \quad (27)$$

### 3 Sample size

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The sample size required for a test to reach predefined power can be calculated under following assumption

**Assumption 1.** *The alternative hypothesis is the one given by the observed sample.*

In other words observed estimates construct the alternative hypothesis. The procedure to retrieve the sample size is following:

1. Calculate contingency table in terms of probabilities ( $p_{ij}^a$ ).
2. For goodness-of-fit set the probabilities, for other test calculate expected values from the contingency table ( $\widehat{p}_{ij}$ ).
3. Calculate deltas  $\Delta_{ij}$  between contingency table and values from the previous step.
4. Calculate the  $\alpha$ -quantile  $\chi_{df,1-\alpha}^2$  of a central chi-square distribution.
5. Having defined target power ( $\beta$ ) of a test, find the noncentrality parameter ( $\lambda$ ) of a noncentral chi-square distribution. The parameter  $\lambda$  is found solving following equation

$$\beta = P^\lambda (X > \chi_{df,1-\alpha}^2). \quad (28)$$

6. Depending on the type of test calculate the sample size  $n$ . Details of the calculation is shown in the next sections.

#### 3.1 Sample size for test of goodness of fit

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The  $p_i^0$  from Equation 2 is defined a priori for every  $i$ . The  $p_i^a$  from Equation 7 is defined a posteriori and is equal to the estimate  $\widehat{p}_i$  derived from observed sample for every  $i$ . In this case the Equation 9 is following:

$$\lambda = n \sum_{i=1}^k \frac{(\widehat{p}_i - p_i^0)^2}{p_i^0} \quad (29)$$

So

$$n = \frac{\lambda}{\sum_{i=1}^k \frac{(\widehat{p}_i - p_i^0)^2}{p_i^0}} \quad (30)$$

#### 3.2 Sample size for test of independence

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Having performed steps until 5 the Equation 18 is following

$$\lambda = n \left[ \sum_{i=1}^r \sum_{j=1}^c \frac{\Delta_{ij}^2}{\widehat{p_{i\cdot} p_{\cdot j}}} - \sum_{i=1}^r \frac{\Delta_{i\cdot}^2}{\widehat{p_{i\cdot}}} - \sum_{j=1}^c \frac{\Delta_{\cdot j}^2}{\widehat{p_{\cdot j}}} \right], \quad (31)$$

So the sample size  $n$  is

$$n = \frac{\lambda}{\left[ \sum_{i=1}^r \sum_{j=1}^c \frac{\Delta_{ij}^2}{\widehat{p_{i\cdot} p_{\cdot j}}} - \sum_{i=1}^r \frac{\Delta_{i\cdot}^2}{\widehat{p_{i\cdot}}} - \sum_{j=1}^c \frac{\Delta_{\cdot j}^2}{\widehat{p_{\cdot j}}} \right]}, \quad (32)$$

### 3.3 Sample size for test of homogeneity

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Having performed steps until 5 the Equation 27 is following

$$\lambda = n \sum_{j=1}^c \frac{1}{\widehat{p_{\cdot j}}} \left[ \sum_{i=1}^r \Delta_{ij}^2 p_{i\cdot} - \left( \sum_{i=1}^r \Delta_{ij} p_{i\cdot} \right)^2 \right] \quad (33)$$

So the sample size  $n$  is

$$n = \frac{\lambda}{\sum_{j=1}^c \frac{1}{\widehat{p_{\cdot j}}} \left[ \sum_{i=1}^r \Delta_{ij}^2 p_{i\cdot} - \left( \sum_{i=1}^r \Delta_{ij} p_{i\cdot} \right)^2 \right]} \quad (34)$$

### References

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- [1] Chi-squared-test [https://en.wikipedia.org/wiki/Pearson%27s\\_chi-squared\\_test](https://en.wikipedia.org/wiki/Pearson%27s_chi-squared_test).
- [2] Guenther, W. (1977). *Power and Sample Size for Approximate Chi-Square Tests*. The American Statistician, 31(2), 83-85.