

# Hypothesis Testing – Distribution

Jiří Neubauer

Department of Econometrics FVL UO Brno  
office 69a, tel. 973 442029  
email: [Jiri.Neubauer@unob.cz](mailto:Jiri.Neubauer@unob.cz)

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- a distribution (normal, Poisson, ...).

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# Hypothesis Testing

reality	H is true		H is false	
decision about H		prob.		prob.
not reject	correct decision	$1 - \alpha$	type II error	$\beta$
reject	type I error	$\alpha$	correct decision	$1 - \beta$



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- If we accept the null hypothesis although is false, we call this **type II error**. The probability of this error is  $\beta$ . A number  $1 - \beta \Rightarrow$  a **power of the test** is the probability that we reject the null hypothesis  $H$ , if it is false.

# Hypothesis Testing

To test the null hypothesis we use a function of a random sample  $T = T(x_1, x_2, \dots, x_n)$ , so called test statistic, which has under the null hypothesis  $H$  known distribution (usually  $t, u, \chi^2, F$ ).  
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We divide the all possible values of the test statistic into

- $W_{1-\alpha}$  - **nonrejection region of  $H$**  – the set of values connected with the hypothesis  $H$ ,
- $W_{\alpha}$  - **rejection region of  $H$**  – the set of values connected with the hypothesis  $A$ .

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5. Calculate the value of the test statistic.

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## 6. Make a decision:

- If the value of the test statistic falls in the rejection region, we reject the null hypothesis  $H$  and say that we accept the alternative hypothesis  $A$  with the probability  $1 - \alpha$ .
- If the value of the test statistic falls in the nonrejection region, we do not reject the null hypothesis  $H$ .

# Chi-Square Goodness of Fit Test

We divide values of a random sample  $x_1, x_2, \dots, x_n$  into  $k$  disjunct classes, where  $n_j, j = 1, 2, \dots, k$ , is frequency of the class  $j$  and  $\pi_j$  is a probability that the random variable  $X$  has value from the class  $j$ , calculated on condition that  $X$  has an assumed distribution.

The main idea of the test is to compare relative frequencies  $n_j/n$  with theoretical probabilities  $\pi_j$ .

# Chi-Square Goodness of Fit Test

We state the null and alternative hypothesis:

$H$  : the random  $X$  has an assumed distribution  $\rightarrow$   $A$  : the random  $X$  has not an assumed distribution.

The test statistic is

$$\chi^2 = \sum_{j=1}^k \frac{(n_j - n\pi_j)^2}{n\pi_j},$$

which has under the null hypothesis  $H$  for large  $n$  (asymptotically) a Pearson  $\chi^2$ -distribution with  $\nu = k - c - 1$  degrees of freedom, where  $c$  is a number of estimated parameters of the assumed distribution.

A rejection region is

$$W_\alpha = \{\chi^2, \chi^2 \geq \chi_{1-\alpha}^2(\nu)\},$$

where  $\chi_{1-\alpha}^2(\nu)$  is a quantile of the Pearson  $\chi^2$ -distribution.

# Chi-Square Goodness of Fit Test

Recommendation:

$$n\pi_j > 5, \quad j = 1, 2, \dots, k.$$

If this condition is not satisfied, it is necessary to join the classes.



# Tests of Skewness and Kurtosis

The normal distribution has  $\alpha_3 = 0$  and  $\alpha_4 = 0$ . We can use these properties to test normality. We calculate a sample skewness and kurtosis (they are estimates of  $\alpha_3$  and  $\alpha_4$ )

$$\hat{\alpha}_3 = a_3 = \frac{1}{ns_n^3} \sum_{i=1}^n (x_i - \bar{x})^3, \quad \hat{\alpha}_4 = a_4 = \frac{1}{ns_n^4} \sum_{i=1}^n (x_i - \bar{x})^4 - 3.$$

We state hypothesis:

$$H_1 : \alpha_3 = 0 \rightarrow A_1 : \alpha_3 \neq 0$$

$$H_2 : \alpha_4 = 0 \rightarrow A_2 : \alpha_4 \neq 0$$

# Tests of Skewness and Kurtosis

1.  $H_1 : \alpha_3 = 0 \rightarrow A_1 : \alpha_3 \neq 0$

The test statistic is

$$u_3 = \frac{a_3}{\sqrt{D(a_3)}}, \quad \text{where} \quad D(a_3) = \frac{6(n-2)}{(n+1)(n+3)},$$

which has under the null hypothesis  $H_1$  asymptotically a normal distribution  $N(0, 1)$ . A rejection region is

$$W_\alpha = \left\{ u_3, |u_3| \geq u_{1-\frac{\alpha}{2}} \right\},$$

where  $u_{1-\frac{\alpha}{2}}$  is a quantile of  $N(0, 1)$ .

# Tests of Skewness and Kurtosis

2.  $H_2 : \alpha_4 = 0 \rightarrow A_2 : \alpha_4 \neq 0$

The test statistic is

$$u_4 = \frac{a_4 + \frac{6}{n+1}}{\sqrt{D(a_4)}}, \quad \text{where} \quad D(a_4) = \frac{24n(n-2)(n-3)}{(n+1)^2(n+3)(n+5)},$$

which has under the null hypothesis  $H_2$  asymptotically a normal distribution  $N(0, 1)$ . A rejection region is

$$W_\alpha = \left\{ u_4, |u_4| \geq u_{1-\frac{\alpha}{2}} \right\},$$

where  $u_{1-\frac{\alpha}{2}}$  is a quantile of  $N(0, 1)$ .

# Compound Tests of Skewness and Kurtosis

We state hypothesis:

$H$  : a random variable  $X$  has a normal distribution  $\rightarrow A$  : a random variable  $X$  has not a normal distribution.

Test statistic is

$$C = u_3^2 + u_4^2,$$

which has under the null hypothesis  $H$  approximately  $\chi^2$  distribution with two degrees of freedom.

$u_3$  and  $u_4$  are test statistics defined above. A rejection region is

$$W_\alpha = \{C, C \geq \chi_{1-\alpha}^2(2)\},$$

where  $\chi_{1-\alpha}^2(2)$  is a quantile of the Pearson  $\chi^2$ -distribution.