

# Test for normal variance, $H_0: \sigma^2 = \sigma_0^2$

Data:

$$X_1, \dots, X_n$$

$$X \sim N[\mu, \sigma^2]$$

Test statistic:

both unknown

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

$$T = \frac{(n-1)S^2}{\sigma_0^2} \sim \chi_{n-1}^2 \text{ when } H_0 \text{ is true.}$$

Critical point for the level  $\alpha$  test:

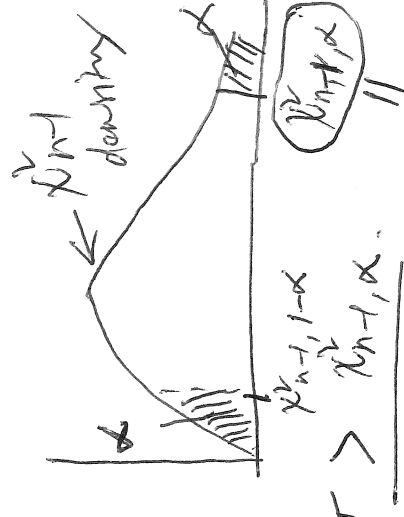
One-sided alternative:  $H_1: \sigma^2 > \sigma_0^2$  . Reject when  $T > \chi_{n-1, 1-\alpha}^2$

$H_1: \sigma^2 < \sigma_0^2$  - Reject when  $T < \chi_{n-1, 1-\alpha}^2$

Two-sided alternative:

$H_1: \sigma^2 \neq \sigma_0^2$  : Reject when  $T < \chi_{n-1, 1-\alpha/2}^2$  or  $T > \chi_{n-1, \alpha/2}^2$

• Chi-square test



$(1-\alpha)$ -th percentile of  $\chi_{n-1}^2$   
 $\chi_{n-1, 1-\alpha}^2$   
 $\chi_{n-1, \alpha}^2$  (circled)  
 going  $(1-\alpha, n-1)$ .

p-value:

already saw.



$H_1$	reject when	p-value	computing p-value
$\sigma^2 > \sigma_0^2$	$P[\chi^2_{n-1} \geq T_{obs}]$		$1 - \text{pchisq}(T_{obs}, n-1)$
$\sigma^2 < \sigma_0^2$	$P[\chi^2_{n-1} \leq T_{obs}]$		$\text{pchisq}(T_{obs}, n-1)$
$\sigma^2 \neq \sigma_0^2$	$\min \{ P[\chi^2_{n-1} \geq T_{obs}], P[\chi^2_{n-1} \leq T_{obs}] \}$		

Testing Hypothesis on ratio of two normal variances.

Set up:

$$X \sim N[\mu_X, \sigma_X^2]$$

$$Y \sim N[\mu_Y, \sigma_Y^2]$$

Data:  $y_1, \dots, y_m$

Data:  $x_1, \dots, x_m$

mutually indep.

$H_1$ : one of the three possibilities.

$$H_0: \sigma_X^2 = \sigma_Y^2$$

$H_1$ : one of the three possibilities.  
variance pivot for CF for  $\frac{\sigma_X^2}{\sigma_Y^2}$ .

Test statistic:

$$\frac{\frac{(m-1)S_X^2}{\sigma_X^2} \cdot \frac{1}{m}}{\frac{(n-1)S_Y^2}{\sigma_Y^2} \cdot \frac{1}{n}}$$

Note:

$$\sim F_{m-1, n-1}$$

F-test

$$T = \frac{S_X^2 / \sigma_X^2}{S_Y^2 / \sigma_Y^2} \xrightarrow{H_0} \frac{S_X^2}{S_Y^2}$$

$\sim F_{m-1, n-1}$  when  $H_0$  is true.

→ Proceed as in  $\chi^2$ -test to compute critical point and p-value.

# Chi-square tests [Chapter 10].

## Chi-square Goodness of Fit Test

Set up: Count data on a categorical variable. *Ex: Gender, Race, majors, etc.*

- There are  $k$  categories, labeled as  $i = 1, \dots, k$ .
- Observed data:  $O_i$ ,  $i = 1, \dots, k$ , where  $O_i = \#$  of observations in the  $i$ -th category. *(Assume/ignore the ordering of categories, the ordering observed frequencies, if it exists).*
- $n = \sum_{i=1}^k O_i$  is the total number of observations.

**Hypotheses:**  $H_0$  : The data follow a given model, versus,  $H_1$  : The data don't follow the given model.

- Let  $p_i$  = proportion of observations in the population that fall in the  $i$ -th category,  $i = 1, \dots, k$ . We can also think of  $p_i$  as the probability that a randomly selected observation from the population falls in the  $i$ -th category.
- $H_0 : p_i = \overset{\leftarrow \text{known}}{p_{i,0}}$ ,  $i = 1, \dots, k$ , where  $p_{i,0}$  are known proportions that add up to 1. *[i.e., all the cell probs. are known.]*
- Model is **completely known** under  $H_0$ .

Case 1: Two categories :  $(p_1, p_2)$ ,  $p_1 + p_2 = 1$ ,  $i=1$ ,  $p_2 = 1 - p_1$

$H_0: p_1 = p_{1,0}, p_2 = p_{2,0} = 1 - p_{1,0}$  , Nothing new here as we can label one category as "success" and the other as "failure", and use test for proportions.

Case 2: More than two categories

**Basic idea:** Compare  $O_i$ 's with  $E_i$ 's — counts expected assuming  $H_0$  is true — using a **chi-square statistic**

$$\chi^2 = \sum_{i=1}^k (O_i - E_i)^2 / E_i$$

- Large  $\chi^2$ : large diff. in obs. and exp. counts.
- Reject  $H_0$  when  $\chi^2$  is large.
- $E_i = n p_{i,0}$ ,  $i=1, 2, \dots, k$  ;  $\sum_{i=1}^k E_i = n$
- Null distribution: Approx.  $\chi^2_{k-1}$  dist. if  $n$  is large.
- Rule of thumb: All  $E_i \geq 5$ . Collapse adjacent categories if this is not the case.

**Ex:** Suppose 60 independent rolls of a die lead to the following data.

category	1	2	3	4	5	6	total
observed count ( $O_i$ )	4	6	17	16	8	9	60
expected count ( $E_i$ )	10	10	10	10	10	10	60

Is the die fair? Answer this question by performing an appropriate test of hypothesis at 5% level of significance.

$H_0$ : Die is fair,  $p_{i,0} = \frac{1}{6}$ ,  $i=1,2,\dots,6$ .

$H_1$ : Die is not fair, i.e., at least one  $p_{i,0}$  is  $\neq \frac{1}{6}$ .

$$E_i = n p_{i,0} = 60 \left(\frac{1}{6}\right) = 10.$$

$$\chi^2_{obs} = \frac{(4-10)^2}{10} + \frac{(6-10)^2}{10} + \dots + \frac{(9-10)^2}{10} = 14.2$$

$$p\text{-value} = P[\chi^2_{k-1} \geq \chi^2_{obs}] = 1 - P(\chi^2_{5, 14.2}) = 0.014$$

Conclusion: Reject  $H_0$  since  $p\text{-value} < \alpha = 0.05$ .  
Die is not fair.

R code:

```
> x <- c(4, 6, 17, 16, 8, 9)
> sum(x)
[1] 60
> sum( (x-10)^2/10)
[1] 14.2 →  $\chi^2_{df=5}$ 
> 1-pchisq(14.2, 5)
[1] 0.01438768 →  $p=1$ 
>
```