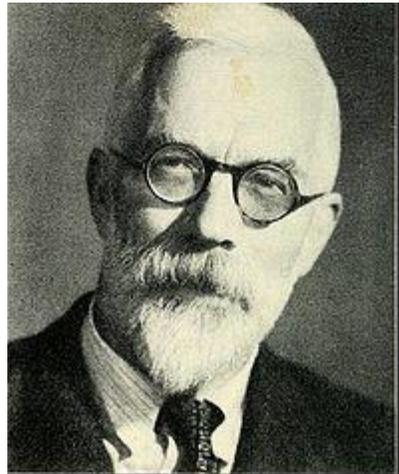


Hypothesis Testing

1. Illustrative example
2. Alternatives and testing errors
3. Mean and variance hypothesis tests
4. In-class exercise



Ronald Fisher
1935

Hypothesis Testing

Illustrative Example

Introduction

- Basic idea
 - » Perform statistical analysis of data to determine if a particular hypothesis should be accepted
 - » Use results of the hypothesis test for decision making
- General procedure
 - » Formulate the hypothesis
 - » Formulate an alternative to the hypothesis
 - » Choose a significance level α that represents the probability of rejecting a true hypothesis
 - » Perform statistical analysis on samples to test the validity of the hypothesis
 - » Either accept or reject the hypothesis based on the test

Illustrative Example

- A company intends to purchase a lot of 100 solar cells if the manufacturer's claim that the film thickness is 200 microns can be verified
- The null hypothesis (hypothesis) is that the film thickness $\mu = \mu_0 = 200$ microns
- The alternative hypothesis (alternative) is that $\mu = \mu_1 < \mu_0$
- The hypothesis will be accepted if it is satisfied with a probability α (significance level)
- Otherwise the hypothesis will be rejected and the solar cells will not be purchased
- The decision needs to be made with a small number of samples

Illustrative Example

- The thicknesses of $n = 25$ cells are measured to yield $\bar{x} = 197$ microns and $s = 6$ microns
- The question is if the sample mean \bar{x} is significantly different than the desired value μ_0
- If the thicknesses are assumed to be normally distributed, then T follows a t -distribution with $n = m - 1$ degrees of freedom:

$$T = \frac{\bar{X} - \mu_0}{S/\sqrt{n}}$$

- The sample mean \bar{x} and standard deviation s are observed values of \bar{X} and S
- Select a significance level $\alpha = 5\%$

Illustrative Example

- Need to find c such that: $P(T \leq c) = \alpha = 0.05$
- Because the t-distribution is symmetric, the c value can be obtained from Table A9:

$$P(T \leq \tilde{c}) = 1 - \alpha = 0.95 \quad \Rightarrow \quad \tilde{c} = 1.71 \quad \Rightarrow \quad c = -\tilde{c} = -1.71$$
$$m = n - 1 = 24$$

- If the hypothesis is true, then there is only a 5% chance that an observed value t of T will have a value $[-\infty, -1.71]$

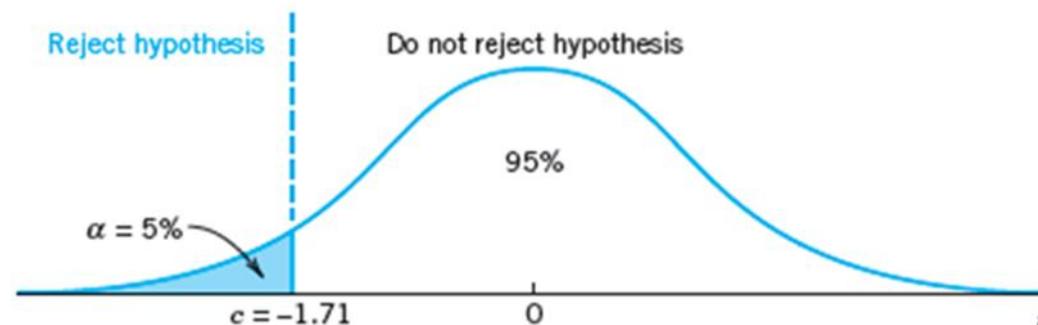


Fig. 532. t-distribution in Example 1

Illustrative Example

- Compute the t statistic from the samples:

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} = \frac{197 - 200}{6/\sqrt{25}} = -2.5$$

- Because $t = -2.5 < c = -1.71$:
 - » The null hypothesis that $\mu = \mu_0 = 200$ microns is rejected
 - » The alternative hypothesis that $\mu = \mu_1 < 200$ microns is accepted
 - » The solar cells are not purchased

Illustrative Example

- The likelihood of accepting the hypothesis will increase if:
 - » The sample mean \bar{x} is closer to the hypothesized mean μ_0
 - » The sample standard deviation s increases
 - » The number of samples n decreases
 - » The significance level α decreases

\bar{x}	s	n	α	m	c	t	Hypothesis
197	6	25	0.05	24	-1.71	-2.50	Reject
198	6	25	0.05	24	-1.71	-1.67	Accept
197	10	25	0.05	24	-1.71	-1.5	Accept
197	6	10	0.05	9	-1.83	-1.58	Accept
197	6	25	0.005	24	-2.80	-2.50	Accept

Hypothesis Testing

Alternatives and Testing Errors

Distribution for Mean Hypothesis Testing

- Let $\{X_1, \dots, X_n\}$ be independent normal random variables, each with the same mean μ and variance σ^2
- Then the random variable T follows a t-distribution with $m = n-1$:

$$T = \frac{\bar{X} - \mu}{S/\sqrt{n}} \quad \bar{X} = \frac{1}{n}(X_1 + \dots + X_n) \quad S^2 = \frac{1}{n-1} \sum_{j=1}^n (X_j - \bar{X})^2$$

- The t-distribution is tabulated in Table A9 where values of z are given as a function of values of:
 - » The cumulative distribution function $F(z)$
 - » The degrees of freedom m

Distribution for Variance Hypothesis Testing

- Let $\{X_1, \dots, X_n\}$ be independent normal random variables, each with the same mean μ and variance σ^2
- Then the random variable Y follows a chi-squared distribution with $m = n-1$:

$$Y = (n-1) \frac{S^2}{\sigma^2} \quad \bar{X} = \frac{1}{n} (X_1 + \dots + X_n) \quad S^2 = \frac{1}{n-1} \sum_{j=1}^n (X_j - \bar{X})^2$$

- The chi-squared distribution is tabulated in Table A10 where values of z are given as a function of values of:
 - » The cumulative distribution function $F(z)$
 - » The degrees of freedom m

One-Sided and Two-Sided Alternatives

- Let θ be an unknown parameter in a distribution for which it is hypothesized that $\theta = \theta_0$
- Alternatives:

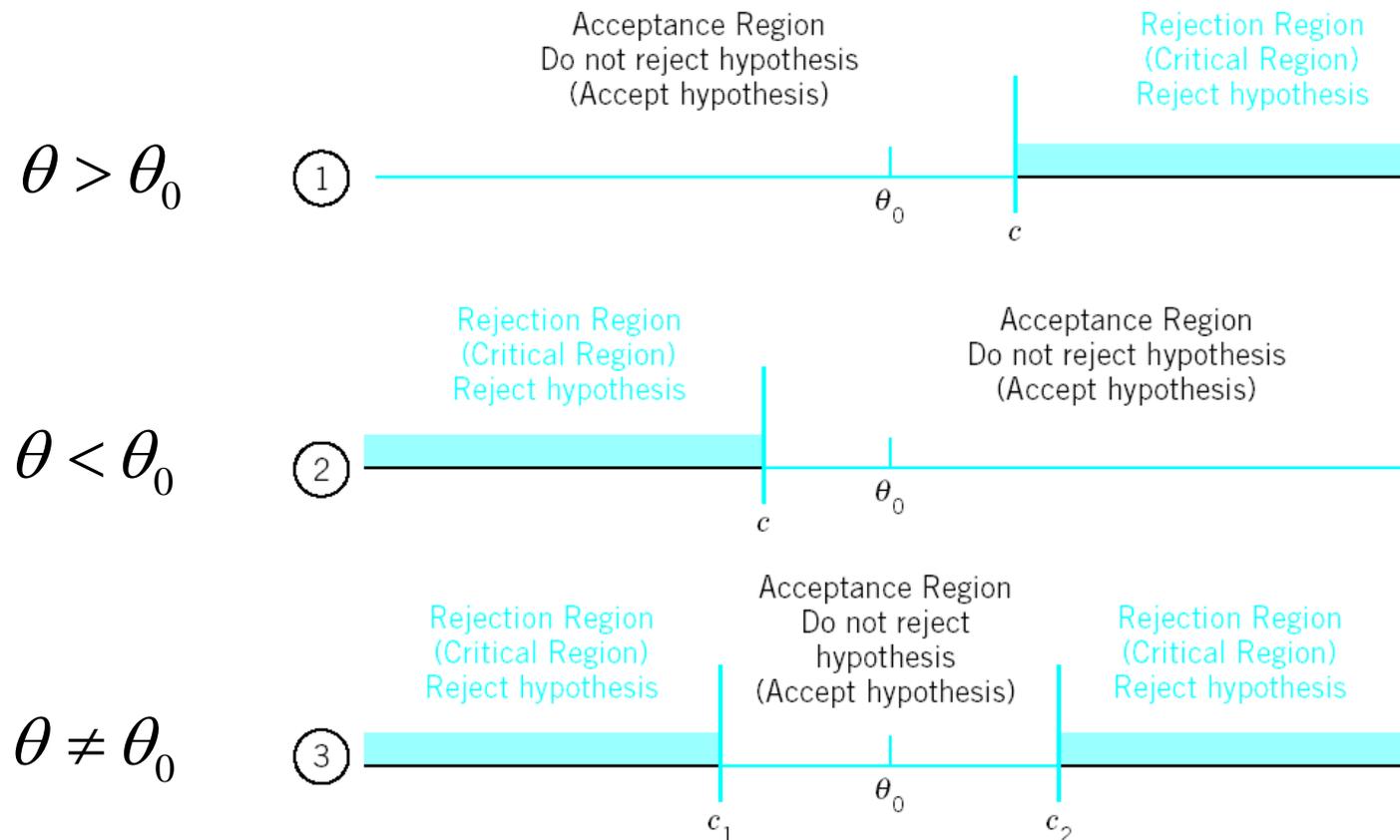


Fig. 532. Test in the case of alternative (1) (upper part of the figure), alternative (2) (middle part), and alternative (3)

Hypothesis Testing Errors

- Hypothesis testing involves a risk of making false decisions
- Type I error – reject a true hypothesis
 - » α = probability of making a Type I error
- Type II error – accept a false hypothesis
 - » β = probability of making a Type II error

Table 25.4 Type I and Type II Errors in Testing a Hypothesis $\theta = \theta_0$ Against an Alternative $\theta = \theta_1$

		Unknown Truth	
		$\theta = \theta_0$	$\theta = \theta_1$
Accepted	$\theta = \theta_0$	True decision $P = 1 - \alpha$	Type II error $P = \beta$
	$\theta = \theta_1$	Type I error $P = \alpha$	True decision $P = 1 - \beta$

Hypothesis Testing Errors

- Type I and II testing errors are conflicting requirements
- If the significance level α decreases, then β increases and the chance of accepting a false hypothesis increases
- The text shows how to calculate β from α and the critical value c

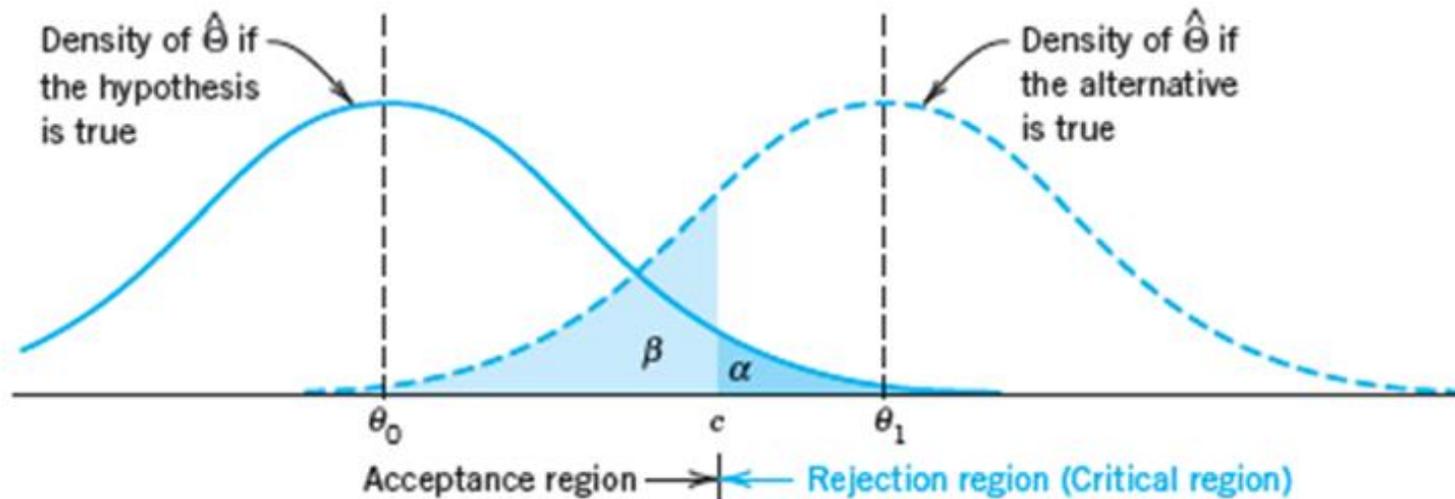


Fig. 534. Illustration of Type I and II errors in testing a hypothesis $\theta = \theta_0$ against an alternative $\theta = \theta_1$ ($> \theta_0$, right-sided test)

Hypothesis Testing

Mean and Variance Hypothesis Tests

Mean Test for Normal Distributions

- Data: n random samples $\{x_1, x_2, \dots, x_n\}$
- Method shown below is for left-handed test
- Hypothesis: mean is μ_0 instead of $\mu_1 < \mu_0$
- Select significance level α
- Compute observed value of T as:

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \quad \bar{x} = \frac{1}{n}(x_1 + \dots + x_n) \quad s^2 = \frac{1}{n-1} \sum_{j=1}^n (x_j - \bar{x})^2$$

- Determine c from Table A9 with $m = n-1$ as: $P(T < c) = \alpha$
- Accept hypothesis if $t > c$; otherwise reject hypothesis

Mean Hypothesis Test Example

- Measurements of polymer molecular weight (scaled by 10^{-5})
 $\bar{x} = 1.258$ $s^2 = 0.0049$
- Hypothesis: $\mu_0 = 1.3$ instead of the alternative $\mu_1 < \mu_0$
- Significance level: $\alpha = 0.10$
- Degrees of freedom: $m = 9$
- Critical value

$$P(T \leq \tilde{c}) = 1 - \alpha = 0.90 \quad \Rightarrow \quad \tilde{c} = 1.38$$

$$P(T \leq c) = \alpha = 0.10 \quad \Rightarrow \quad c = -\tilde{c} = -1.38$$

- Sample t

$$t = \frac{1.258 - 1.3}{\sqrt{0.0049} / \sqrt{10}} = -1.897 < c = -1.38$$

- Reject hypothesis

Variance Test for Normal Distributions

- Data: n random samples $\{x_1, x_2, \dots, x_n\}$
- Method shown below is for right-handed test
- Hypothesis: variance is σ_0^2 instead of $\sigma_1^2 > \sigma_0^2$
- Compute sample variance as: $s^2 = \frac{1}{n-1} \sum_{j=1}^n (x_j - \bar{x})^2$
- Select significance level α
- Determine c from Table A10 with $m = n-1$ as:
$$P(Y > c) = \alpha \quad \Rightarrow \quad P(Y \leq c) = 1 - \alpha$$
- Compute critical value of s^2 as: $c^* = \sigma_0^2 c / (n-1)$
- Accept hypothesis if $s^2 < c^*$; otherwise reject hypothesis

Variance Hypothesis Test Example

- Measurements of polymer molecular weight (scaled by 10^{-5})

$$\bar{x} = 1.258 \quad s^2 = 0.0049$$

- Hypothesis: $\sigma_0^2 = 0.005$ instead of the alternative $\sigma_1^2 > \sigma_0^2$
Significance level: $\alpha = 0.05$
- Degrees of freedom: $m = 9$
- Critical value

$$P(Y > c) = \alpha = 0.05 \quad \Rightarrow \quad P(Y \leq c) = 1 - \alpha = 0.95 \quad \Rightarrow \quad c = 16.92$$

$$c^* = \frac{\sigma_0^2 c}{n-1} = \frac{(0.005)(16.92)}{9} = 0.0094$$

- Since $s^2 < c^*$, the hypothesis is accepted

Hypothesis Testing

In-class Exercise