

# **To Do**

**Read Sections 5.1 - 5.2 (Hypothesis Testing)**

**Do End-of-Chapter Problems 1 - 8.**

**Assignment 3 due Friday November 11.**

**See detailed information posted on Learn regarding material covered by Midterm Test 2 (4:40 - 6:10 on Tuesday November 15).**

# **Last Class**

- (1) Null hypothesis, Alternative Hypothesis**
- (2) Test statistic or Discrepancy Measure**
- (3) Steps of a Test of Hypothesis**
- (4) Interpretation of a p-value**

# Steps of a Statistical Test of Hypothesis

(1) Assume that the null hypothesis  $H_0$  will be tested using data  $Y$ .

(2) Adopt a test statistic or discrepancy measure  $D(Y)$  for which, large values of  $D$  are less consistent with  $H_0$ . Let  $d = D(y)$  be the corresponding observed value of  $D$ .

(3) Calculate

$$\begin{aligned} p\text{-value} &= P(D \geq d \text{ assuming } H_0 \text{ is true}) \\ &= P(D \geq d; H_0) \end{aligned}$$

(4) Draw a conclusion based on the p-value.

# Guidelines for Interpreting the $p$ -value

These are only guidelines for this course.

The interpretation of a  $p$ -value must always be made in the context of a given study.

$p$ -value	Interpretation
$p > 0.1$	There is no evidence against $H_0$ based on the data.
$0.05 < p \leq 0.1$	There is some evidence against $H_0$ based on the data.
$0.01 < p \leq 0.05$	There is evidence against $H_0$ based on the data.
$0.001 < p \leq 0.01$	There is strong evidence against $H_0$ based on the data.
$p \leq 0.001$	There is very strong evidence against $H_0$ based on the data.

# Interpreting the $p$ -value

See the  $p$ -value bears:

<http://www.youtube.com/watch?v=ax0tDcFkPic&feature=related>

## **ESP Experiment: $n = 100$**

**Suppose we did the ESP experiment for  $n = 100$  trials and Student answered correctly 60 times.**

**The test statistic would now be  $D = |Y - 50|$  and the observed value is  $d = |60 - 50| = 10$ .**

# ESP Experiment: $n = 100$

$p$ -value =  $P(D \geq 10; \text{assuming } H_0 \text{ is true})$

=  $P(D \geq 10; H_0)$

=  $P(|Y - 50| \geq 10)$  where  $Y \sim \text{Binomial}(100, 0.5)$

$$= P\left(\frac{|Y - 50|}{\sqrt{100(0.5)(0.5)}} \leq \frac{10}{\sqrt{100(0.5)(0.5)}}\right) \text{ (no continuity correction used)}$$

$\approx P(|Z| \geq 2)$  where  $Z \sim N(0, 1)$

=  $2[1 - P(Z \leq 2) - 1] = 2(1 - 0.97725)$

= 0.04550

What would we conclude now about Student's ESP ability?

# Today's Lecture

- (1) Testing  $H_0: \mu = \mu_0$  when  $\sigma$  is unknown for  $G(\mu, \sigma)$  model.**
- (2) Statistical significance versus practical significance.**
- (3) Relationship between tests of hypothesis and confidence intervals**



# Tests of hypotheses for the parameters in a $G(\mu, \sigma)$ model

The  $G(\mu, \sigma)$  has two parameters  $\mu$  and  $\sigma$ .

Today we look at testing  $H_0: \mu = \mu_0$  when  $\sigma$  is unknown.

Next class we look at testing  $H_0: \sigma^2 = \sigma_0^2$  when  $\mu$  is unknown.

## Testing $H_0: \mu = \mu_0$ when $\sigma$ is unknown

Suppose  $Y_1, Y_2, \dots, Y_n$  is a random sample from a  $G(\mu, \sigma)$  distribution.

There is a close relationship between the pivotal quantities we used to find confidence intervals and test statistics for testing hypotheses.

Recall the pivotal quantity:

$$\frac{\bar{Y} - \mu}{S / \sqrt{n}} \sim t(n-1)$$

## Testing $H_0: \mu = \mu_0$ when $\sigma$ is unknown

To test  $H_0: \mu = \mu_0$  we use the test statistic

$$D = \frac{|\bar{Y} - \mu_0|}{S / \sqrt{n}}$$

Why does this test statistic make sense?

## Testing $H_0: \mu = \mu_0$ when $\sigma$ is unknown

To test  $H_0: \mu = \mu_0$  we use the test statistic

$$D = \frac{|\bar{Y} - \mu_0|}{S / \sqrt{n}}$$

Why does this test statistic make sense?

$E(\bar{Y}) = \mu_0$  if  $H_0: \mu = \mu_0$  is true.

# Testing $H_0: \mu = \mu_0$ when $\sigma$ is unknown

**Let**

$$d = \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}}$$

**be the observed value of  $D$  for an experiment which has been conducted.**

$$p\text{-value} = P(D \geq d; H_0 \text{ is true})$$

$$= P\left(\frac{|\bar{Y} - \mu_0|}{S / \sqrt{n}} \geq \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}}\right)$$

$$= P(|T| \geq d) \text{ where } T \sim t(n-1)$$

$$= 2[1 - P(T \leq d)]$$

## Example: Bias in a Measurement System

An inexpensive weight scale is tested by taking ten weighings of a known 1 kg weight.

The measurements were:

1.026	0.998	1.017	1.045	0.978
1.004	1.018	0.965	1.010	1.000

Assume  $Y_i \sim G(\mu, \sigma)$ ,  $i = 1, 2, \dots, 10$  where  $Y_i = i$ th measurement and  $\mu$  represents the mean measurement in repeated weighings of the 1 kg weight using this scale.

The hypothesis of interest is  $H_0: \mu = 1$ . (Why?)

## Example: Bias in a Measurement System

**For these data**

$$\bar{y} = 1.0061, \mu_0 = 1, s = 0.0230, n = 10$$

$$d = \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}} = \frac{|1.0061 - 1|}{0.0230 / \sqrt{10}} = 0.839$$

$$\begin{aligned} p\text{-value} &= 2[1 - P(T \leq 0.839)] \quad T \sim t(9) \\ &= 2(1 - 0.7884) \approx 0.42 \end{aligned}$$

**Since the p-value  $\approx 0.42$  then based on the observed data there is no evidence against  $H_0: \mu = 1$ . There is no evidence that the scale is over or under weighing.**

## **Example: Bias in a Measurement System**

**For a different set of inexpensive weigh scales the observed data were:**

<b>1.011</b>	<b>0.966</b>	<b>0.965</b>	<b>0.999</b>	<b>0.988</b>
<b>0.987</b>	<b>0.956</b>	<b>0.969</b>	<b>0.980</b>	<b>0.988</b>



## Example: Bias in a Measurement System

**For these data**

$$\bar{y} = 0.981, \mu_0 = 1, s = 0.0170, n = 10$$

$$d = \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}} = \frac{|0.981 - 1|}{0.0170 / \sqrt{10}} = 3.534$$

$$p\text{-value} = 2[1 - P(T \leq 3.534)] \quad T \sim t(9) \\ = 0.0064$$

**Based on the observed data there is no evidence against  $H_0: \mu = 1$ , that is, there is strong evidence that the scale is over or under weighing. The observed data strongly suggest that the second scale is biased.**

## **Example: Bias in a Measurement System**

**Although there is strong evidence against  $H_0$  for the second scale, the degree of bias in its measurements is not necessarily large enough to be of practical concern.**

**In fact, a 95% confidence interval for the mean  $\mu$  is given by**

$$\bar{y} \pm 2.2622s / \sqrt{10} = 0.981 \pm 0.012$$
$$[0.969, 0.993]$$

**where  $P(T \leq 2.2622) = 0.975$  and  $T \sim t(9)$ .**

## **Example: Bias in a Measurement System**

**Evidently the second scale consistently understates the weight but the bias in measuring the 1 kg weight is likely fairly small (about 1% - 3%).**

**Is this bias of practical significance?**

# Statistical Significance versus Practical Significance

**Although we might be able to find evidence against a given hypothesis, this does not mean that the difference is of practical significance.**

**For example a person who is willing to toss a particular coin one million times can almost certainly find evidence against  $H_0: P(\text{heads}) = 0.5$ .**

# Statistical Significance versus Practical Significance

**Similarly, if we collect large amounts of financial data, it is quite easy to find evidence against  $H_0$ : stock index returns are Normally distributed.**

**Nevertheless for smaller amounts of data, the Normality assumption is usually made and considered useful.**

# ***p*-values and Confidence Intervals**

**If the evidence against  $H_0$  is statistically significant, the size of the *p*-value DOES NOT imply how “wrong”  $H_0$  is.**

**A confidence interval however does indicate the magnitude and direction of the departure from  $H_0$ .**

**If strong evidence against  $H_0$  is found in a particular direction then this might suggest conducting further experiments to investigate this evidence.**

# Relationship Between Tests of Hypothesis and Confidence Intervals

Suppose we test  $H_0: \mu = \mu_0$  for  $G(\mu, \sigma)$  data. Then

$$p\text{-value} \geq 0.05$$

$$\text{iff } P\left(\frac{|\bar{Y} - \mu_0|}{S / \sqrt{n}} \geq \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}}; H_0 \text{ is true}\right) \geq 0.05$$

$$\text{iff } P\left(|T| \geq \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}}\right) \geq 0.05 \quad \text{where } T \sim t(n-1)$$

$$\text{iff } P\left(|T| \leq \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}}\right) \leq 0.95$$

$$\text{iff } \frac{|\bar{y} - \mu_0|}{s / \sqrt{n}} \leq a \quad \text{where } P(|T| \leq a) = 0.95$$

$$\text{iff } \mu_0 \in \left[\bar{y} - as / \sqrt{n}, \bar{y} + as / \sqrt{n}\right]$$

**Which is a 95% confidence interval for  $\mu$ .**

# Relationship Between Tests of Hypothesis and Confidence Intervals

**In other words:**

**the p-value for testing  $H_0: \mu = \mu_0$  is greater than or equal to 0.05**

**if and only if**

**the value  $\mu = \mu_0$  is inside a 95% confidence interval for  $\mu$**

**(assuming we use the same pivotal quantity).**



# Relationship Between Tests of Hypothesis and Confidence Intervals

More generally, suppose we use the same pivotal quantity to construct a confidence interval for a parameter  $\theta$  and a test of the hypothesis  $H_0: \theta = \theta_0$ .

The parameter value  $\theta = \theta_0$  is inside a  $100q\%$  confidence interval for  $\theta$  if and only if the  $p$ -value for testing  $H_0: \theta = \theta_0$  is greater than  $1 - q$ .

# Relationship Between Tests of Hypothesis and Confidence Intervals

For the weigh scale example a 95% confidence interval for the mean  $\mu$  for the second scale was [0.969, 0.993].

Since  $\mu = 1$  is not in this interval we know that the  $p$ -value for testing  $H_0: \mu = 1$  would be less than 0.05.

In fact we showed the  $p$ -value equals 0.0064 which is indeed less than 0.05.