

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: $n = 100$

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

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

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

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

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

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

$= 0.04550$

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

Today's Lecture

- (1) Testing $H_0: \mu = \mu_0$ when σ 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 μ and σ .

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

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

Testing $H_0: \mu = \mu_0$ when σ 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 σ 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 σ 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 σ 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 μ 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 ≈ 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:

1.011	0.966	0.965	0.999	0.988
0.987	0.956	0.969	0.980	0.988

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 μ 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 μ .

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 μ

(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 θ and a test of the hypothesis $H_0: \theta = \theta_0$.

The parameter value $\theta = \theta_0$ is inside a $100q\%$ confidence interval for θ 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 μ 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.