

# Math 362: Mathematical Statistics II

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## Chapter 6. Hypothesis Testing

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§ 6.1 Introduction

§ 6.2 The Decision Rule

§ 6.3 Testing Binomial Data –  $H_0 : p = p_0$

## § 6.1 Introduction

Instead of numerical estimates of parameters, in the form of either single points or confidence intervals, we want to make a choice between two conflicting theories, or **hypothesis**:

1.  $H_0$ : the null hypothesis  
v.s.
2.  $H_1$ : the null hypothesis

Comments: Hypothesis testing and confidence intervals are dual concepts to each other: one can be obtained from the other. However, it is often difficult to specify  $\mu_0$  to the null hypothesis.

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## § 6.2 The Decision Rule

Go over the example first....

▶ **Test statistic:** Any function of the observed data whose numerical value dictates whether  $H_0$  is accepted or rejected.

▶ **Critical region  $C$ :** The set of values for the test statistic that result in the null hypothesis being rejected.

**Critical value:** The particular point in  $C$  that separates the rejection region from the acceptance region.

▶ **Level of significance  $\alpha$ :** The probability that the test statistic lies in the critical region  $C$  under  $H_0$ .

## Test Normal mean $H_0 : \mu = \mu_0$ ( $\sigma$ known)

### Setup:

1. Let  $Y_1 = y_1, \dots, Y_n = y_n$  be a random sample of size  $n$  from  $N(\mu, \sigma^2)$  with  $\sigma$  known.
2. Set  $\bar{y} = \frac{1}{n}(y_1 + \dots + y_n)$  and  $z = \frac{\bar{y} - \mu_0}{\sigma/\sqrt{n}}$ .
3. The level of significance is  $\alpha$ .

### Test:

$$\begin{cases} H_0 : \mu = \mu_0 \\ H_1 : \mu > \mu_0 \end{cases}$$

reject  $H_0$  if  $z \geq z_\alpha$ .

$$\begin{cases} H_0 : \mu = \mu_0 \\ H_1 : \mu < \mu_0 \end{cases}$$

reject  $H_0$  if  $z \leq -z_\alpha$ .

$$\begin{cases} H_0 : \mu = \mu_0 \\ H_1 : \mu \neq \mu_0 \end{cases}$$

reject  $H_0$  if  $|z| \leq z_{\alpha/2}$ .

- ▶ **Simple hypothesis:** Any hypothesis which specifies the population distribution completely.
- ▶ **Composite hypothesis:** Any hypothesis which does not specify the population distribution completely.

Conv. We always assume  $H_0$  is simple and  $H_1$  is composite.

**Definition.** The **P-value** associated with an observed test statistic is the probability of getting a value for that test statistic as extreme as or more extreme than what was actually observed (relative to  $H_1$ ) given that  $H_0$  is true.

Note: Test statistics that yield small P-values should be interpreted as evidence against  $H_0$ .

E.g. Suppose that test statistic  $z = 0.60$ . Find P-value for

$$\begin{cases} H_0 : \mu = \mu_0 \\ H_1 : \mu > \mu_0 \end{cases}$$

$$\begin{cases} H_0 : \mu = \mu_0 \\ H_1 : \mu < \mu_0 \end{cases}$$

$$\begin{cases} H_0 : \mu = \mu_0 \\ H_1 : \mu \neq \mu_0 \end{cases}$$

$$\mathbb{P}(Z \geq 0.60) = 0.2743. \quad \mathbb{P}(Z \leq 0.60) = 0.7257.$$

$$\begin{aligned} \mathbb{P}(|Z| \geq 0.60) \\ &= 2 \times 0.2743 \\ &= 0.5486. \end{aligned}$$

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## § 6.3 Testing Binomial Data – $H_0 : p = p_0$

**Setup:** Let  $X_1 = k_1, \dots, X_n = k_n$  be a random sample of size  $n$  from Bernoulli( $p$ ).  $X = \sum_{i=1}^n X_i \sim \text{Binomial}(n, p)$ . We want to test  $H_0 : p = p_0$ .

- |  |                   |
|--|-------------------|
| 1. When $n$ is large, use $Z$ score.               | Large-sample test |
| 2. Otherwise, use the exact binomial distribution. | Small-sample test |

$$\begin{aligned} & n \text{ is large} \\ & \Updownarrow \\ & 0 < np_0 - 3\sqrt{np_0(1-p_0)} < np_0 + 3\sqrt{np_0(1-p_0)} < n \\ & \Updownarrow \\ & n > 9 \times \max\left(\frac{1-p_0}{p_0}, \frac{p_0}{1-p_0}\right). \end{aligned}$$

# Large-sample test for $p$

## Setup:

1. Let  $X_1 = k_1, \dots, X_n = k_n$  be a random sample of size  $n$  from Bernoulli( $p$ ).
2. Suppose  $n > 9 \max\left(\frac{1-p_0}{p_0}, \frac{p_0}{1-p_0}\right)$ .
3. Set  $k = k_1 + \dots + k_n$  and  $z = \frac{k - np_0}{\sqrt{np_0(1-p_0)}}$ .
4. The level of significance is  $\alpha$ .

## Test:

$$\begin{cases} H_0 : p = p_0 \\ H_1 : p > p_0 \end{cases}$$

reject  $H_0$  if  $z \geq z_\alpha$ .

$$\begin{cases} H_0 : p = p_0 \\ H_1 : p < p_0 \end{cases}$$

reject  $H_0$  if  $z \leq -z_\alpha$ .

$$\begin{cases} H_0 : p = p_0 \\ H_1 : p \neq p_0 \end{cases}$$

reject  $H_0$  if  $|z| \leq z_{\alpha/2}$ .

## Small-sample test for $p$

E.g.  $n = 19, p_0 = 0.85, \alpha = 0.10$ . Find critical region for the two-sided test

$$\begin{cases} H_0 : p = p_0 \\ H_1 : p \neq p_0 \end{cases}$$

Sol.  $19 = n < 9 \times \max\left(\frac{0.85}{0.15}, \frac{0.15}{0.85}\right) = 51$ , so small sample test.

By checking the table, the critical region is

$$C = \{k : k \leq 13 \text{ or } k = 19\},$$

so that

$$\begin{aligned} \alpha &= \mathbb{P}(X \in C | H_0 \text{ is true}) \\ &= \mathbb{P}(X \leq 13 | p = 0.85) + \mathbb{P}(X = 19 | p = 0.85) \\ &= 0.099295 \approx 0.10. \end{aligned}$$

□

Binomial with  $n = 19$  and  $p = 0.85$

x	P(X = x)	
6	0.000000	} $\rightarrow P(X \leq 13) = 0.053696$
7	0.000002	
8	0.000018	
9	0.000123	
10	0.000699	
11	0.003242	
12	0.012246	
13	0.037366	
14	0.090746	
15	0.171409	
16	0.242829	
17	0.242829	
18	0.152892	
19	0.045599	$\rightarrow P(X = 19) = 0.045599$