

# Math 362: Mathematical Statistics II

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# Chapter 6. Hypothesis Testing

§ 6.1 Introduction

§ 6.2 The Decision Rule

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

§ 6.4 Type I and Type II Errors

§ 6.5 A Notion of Optimality: The Generalized Likelihood Ratio

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Suppose our friend Jory claims that he has some magic power to predict the side of a randomly tossed fair-coin.

Jory claims that he could do more than  $\frac{1}{2}$  of the time on average.

Let's test Jory to see if we believe his claim.

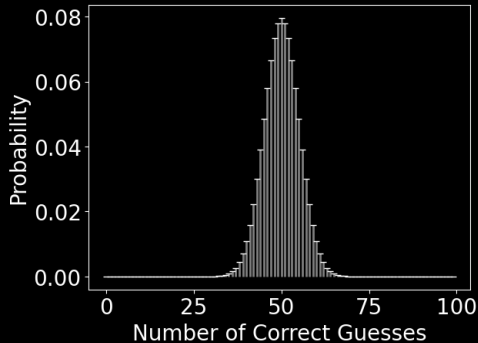
We made Jory guess a repeatedly tossed coin  
for 100 times.

He guesses correctly 54 times.

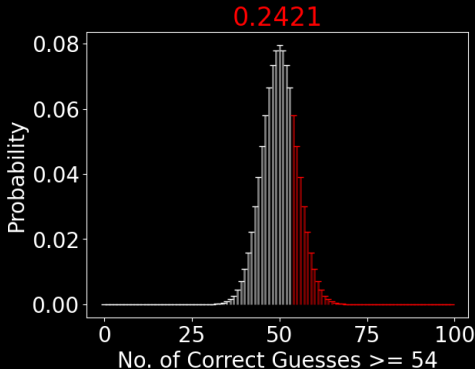
**Question:**

Does this provide strong evidence that Jory  
has the proclaimed magic power?

If Jory is guessing randomly, the number of correct guesses would follow a binomial distribution with parameters  $n = 100$  and  $p = 1/2$ .



What is probability that Jory gets 54 or more correct when guessing randomly?



$$\mathbb{P}(X \geq 54) = \sum_{n=54}^{100} \binom{100}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{100-n} = 0.2421.$$

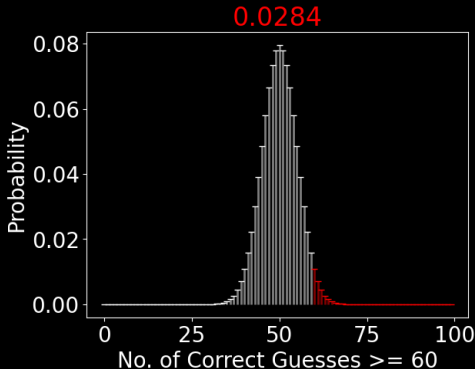


It is not unlikely to get this many correct guesses due to chance.

**Conclusion:**

There is No strong evidence that Jory has better than a  $1/2$  chance of correctly guessing the coin.

What is probability that Jory gets **60 or more** correct when guessing randomly?



$$\mathbb{P}(X \geq 60) = \sum_{n=60}^{100} \binom{100}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{100-n} = 0.0284.$$

Either

Jory is purely guessing with probability of success of  $\frac{1}{2}$ , and we witnessed a very unusual event due to chance.

Or

Jory is truly having the magic power to guess the coin.

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**Conclusion:**

We have strong evidence against  
**Red Hypothesis**

Or the test is in favor of  
**Green Hypothesis**

Before testing Jory, could you set up a threshold above which we will believe Jory's super power?

Find smallest  $m$  such that

$$\mathbb{P}(X \geq m) = \sum_{n=m}^{100} \binom{100}{n} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{100-n} \leq 0.05$$

↓

$$\boxed{m = 59}$$

b.c.  $\mathbb{P}(X \geq 58) = 0.067$  &  $\mathbb{P}(X \geq 59) = 0.044$

We have just informally conducted a hypothesis test with the null hypothesis

$$H_0 : p = \frac{1}{2}$$

against the alternative hypothesis

$$H_1 : p > \frac{1}{2}$$

under the significance level  $\alpha = 0.05$

which is equivalent to either

producing the critical region  
 $m \geq 59$

or

comparing with the p-value.

- ▶ **Test statistic:** Any function of the observed data whose numerical value dictates whether  $H_0$  is accepted or rejected.
- ▶ **Critical region  $\mathcal{C}$ :** The set of values for the test statistic that result in the null hypothesis being rejected.

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

- ▶ **Level of significance  $\alpha$ :** The probability that the test statistic lies in the critical region  $\mathcal{C}$  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  $|z| \geq z_{\alpha/2}$ .

- ▶ **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

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

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

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

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

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

$$\mathbb{P}(|Z| \geq 0.60)$$

$$= 2 \times 0.2743$$

$$= 0.5486.$$