

Hypothesis tests

with binomial example

STAT 587 (Engineering)
Iowa State University

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Statistical hypothesis testing

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What if you observed $y = 2$?

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What if you observed $y = 2$? Well, $P(Y = 2) \approx 0.01$.

Large data

Let $Y \sim \text{Bin}(n, \theta)$ with

$$H_0 : \theta = 0.5 \quad \text{versus} \quad H_A : \theta \neq 0.5.$$

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You collect data and observe $y = 6500$ out of $n = 13000$ attempts. Should you reject H_0 ?
Probably not since $6500 = E[Y]$ if H_0 is true. But $P(Y = 6500) \approx 0.007$.

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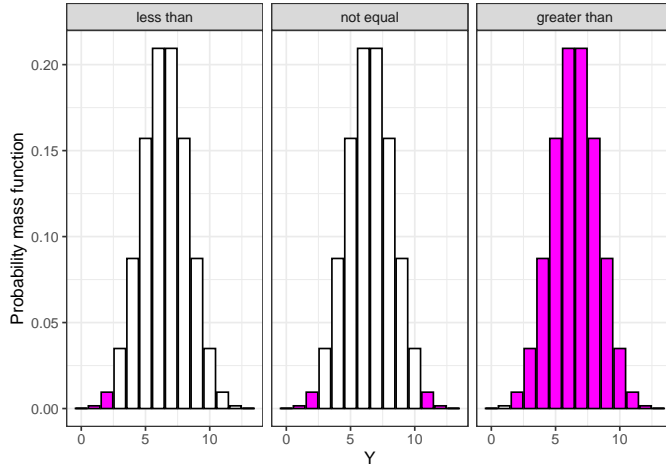
$$H_A : \theta > \theta_0 \implies Y \geq y$$

or

$$H_A : \theta \neq \theta_0 \implies |Y - n\theta_0| \geq |y - n\theta_0|.$$

as or more extreme regions

As or more extreme regions for $Y \sim \text{Bin}(13, 0.5)$ with $y = 2$



R “hand” calculation

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```
pbinom(y, size = n, prob = theta0)
```

```
[1] 0.01123047
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$$H_A : \theta > 0.5 \implies p\text{-value} = P(Y \geq y) = 1 - P(Y \leq y - 1)$$

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1-pbinom(y-1, size = n, prob = theta0)
```

```
[1] 0.998291
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$$H_A : \theta \neq 0.5 \implies p\text{-value} = P(|Y - n\theta_0| \leq |y - n\theta_0|)$$

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$$H_A : \theta \neq 0.5 \implies p\text{-value} = P(|Y - n\theta_0| \leq |y - n\theta_0|)$$

```
2*pbinom(y, size = n, prob = theta0)
```

```
[1] 0.02246094
```

R Calculation

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```
binom.test(y, n, p = theta0, alternative = "less")$p.value
```

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```
binom.test(y, n, p = theta0, alternative = "two.sided")$p.value
```

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- reject if $p\text{-value} < \alpha$ otherwise
- fail to reject.

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	H_0 true	H_0 not true
reject H_0	type I error	correct
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If we **fail to reject H_0** , insufficient evidence to say that the data are incompatible with this model.

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```
binom.test(3, 6, p = 1/6, alternative = "greater")$p.value
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[1] 0.06228567
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With a significance level of $\alpha = 0.05$, you fail to reject the null hypothesis.

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- Use p -values to determine whether to
 - reject the null hypothesis or
 - fail to reject the null hypothesis.
- More assessment is required to determine if other model assumptions hold.