

Monday, Mar 6

## Statistical Significance

A **statistically significant** result is one that is decidedly not due to “ordinary variation” in the data (i.e., not due to chance or not a coincidence). **Statistical tests** (aka *significance tests* or *statistical hypothesis tests* or *hypothesis tests*) are how we decide whether or not an observed result is statistically significant.

### Is a Coin Fair?

Suppose we flip a coin  $n$  times. We can consider the observation of each flip to be a random variable with the following distribution.

$x$	$P(x)$
Heads	$p$
Tails	$1 - p$

The value of  $p$  implies something about the coin.

1. If  $p = 0.5$  the coin is fair.
2. If  $p \neq 0.5$  the coin is not fair.

Assume we do not know the value of  $p$ . We flip the coin 30 times to produce a sample of  $n = 30$  observations. It comes up heads 20 times, so  $\hat{p} = 20/30 = 2/3 \approx 0.67$ . What might we decide about  $p$ ?

1. Conclude that  $p = 0.5$ . The result that  $\hat{p} = 2/3$  **is not** statistically significant.
2. Conclude that  $p \neq 0.5$ . The result that  $\hat{p} = 2/3$  **is** statistically significant.

How do we decide?

### Can Milena Read?

Suppose Milena plays  $n$  games of Pounce. We can consider the observation of her response to a single game to be a random variable with the following distribution.

$x$	$P(x)$
Correct	$p$
Incorrect	$1 - p$

The value of  $p$  implies something about Milena’s reading ability.

1. Milena cannot read. She is guessing so  $p = 1/3$ .
2. Milena can read (somewhat) so  $p > 1/3$ .

We do not know the value of  $p$ . Milena played Pounce 50 times to produce a sample of  $n = 50$  observations. She selected the correct word 25 times, so  $\hat{p} = 25/50 = 0.5$ . What would we decide about  $p$ ?

1. Conclude that  $p = 1/3$ . The result that  $\hat{p} = 0.5$  **is not** statistically significant.
2. Conclude that  $p > 1/3$ . The result that  $\hat{p} = 0.5$  **is** statistically significant.

What do we decide?

## The Sampling Distribution of $\hat{p}$

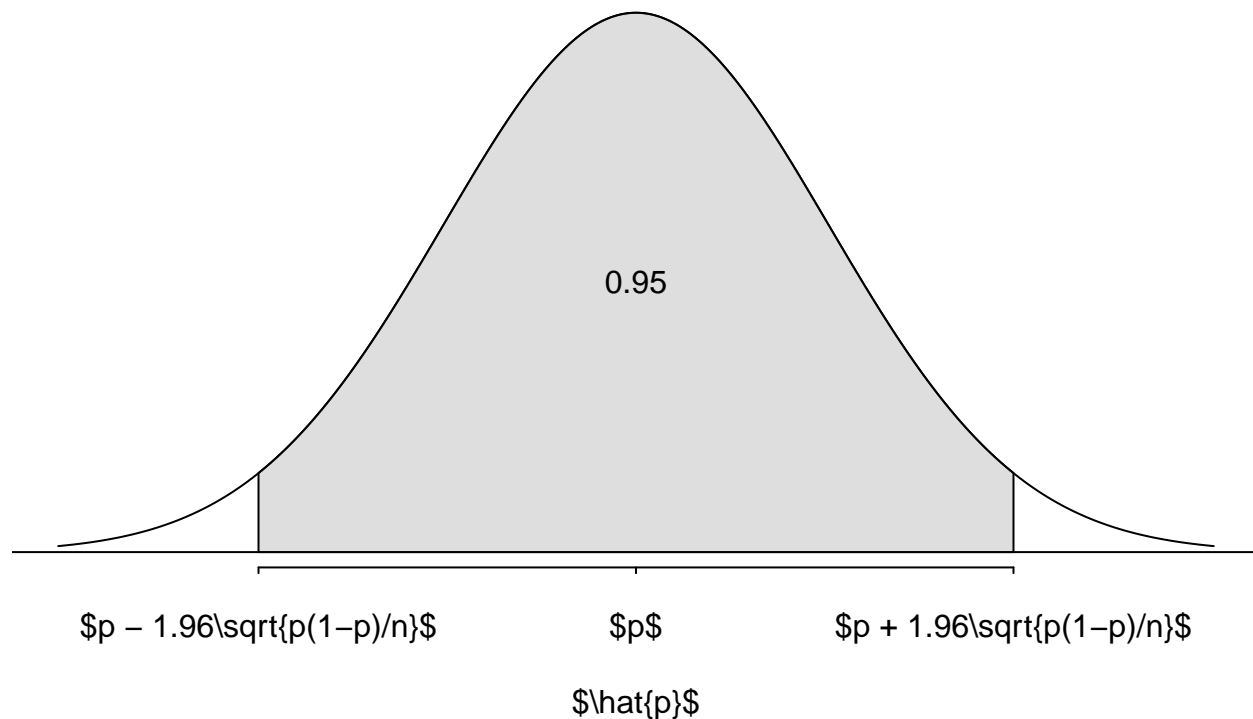
What do we know about the sampling distribution of  $\hat{p}$ ?

1. The mean of  $\hat{p}$  is  $p$ .
2. The standard deviation (i.e., standard error) of  $\hat{p}$  is

$$\sqrt{\frac{p(1-p)}{n}}.$$

3. The shape of the sampling distribution is approximately that of a normal distribution.

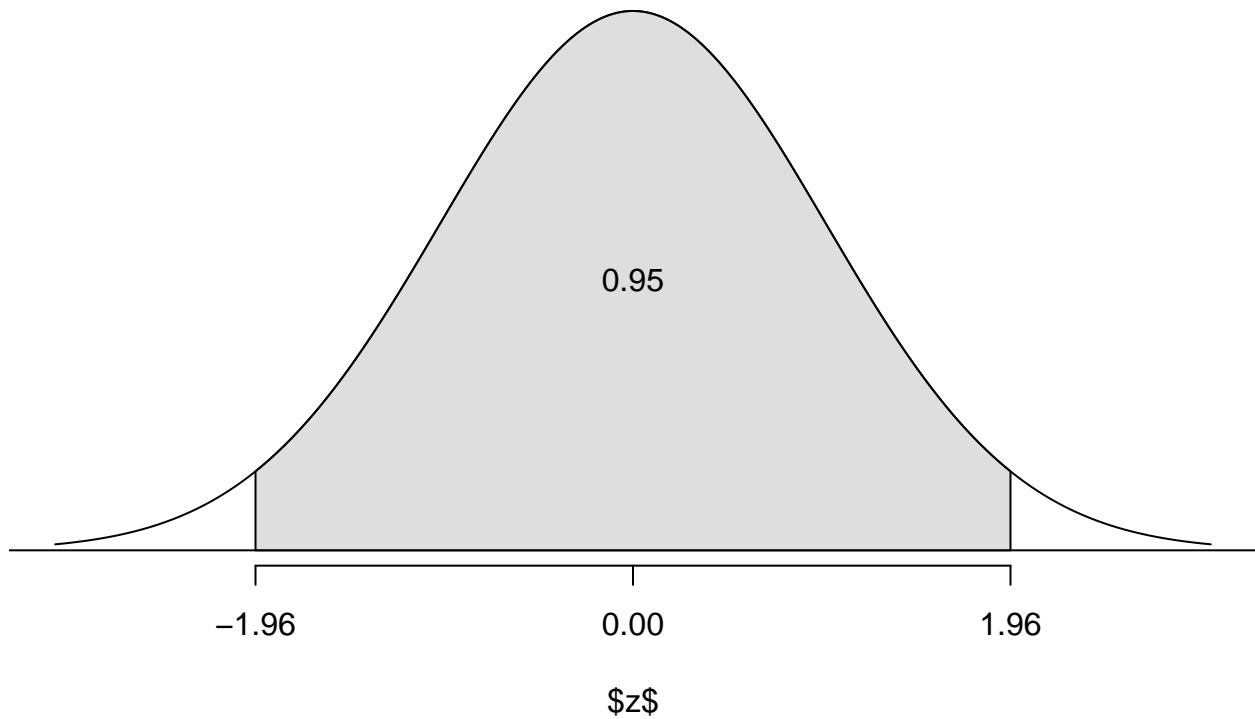
This is the sampling distribution of  $\hat{p}$ .



It is convenient to convert  $\hat{p}$  into a  $z$ -score using the formula

$$z = \frac{\hat{p} - p}{\sqrt{p(1-p)/n}}.$$

This is the sampling distribution of  $z$ .



But we do not know the value of  $p$ !

### Null and Alternative Hypotheses

**Null Hypothesis ( $H_0$ ):** Usually the hypothesis of “no effect” (e.g., nothing is “happening”). In practice the null hypothesis is often that the parameter equals a *specific value* (although we will consider the case when it may be a range of values when we discuss *composite* null hypotheses).

**Alternative Hypothesis ( $H_a$ ):** Usually the hypothesis of an “effect” (e.g., something is “happening”). In practice the alternative hypothesis is usually that the parameter is in a *range of values*.

What would the null and alternative hypotheses be for the examples above?

## Test Statistics

A **test statistic** measures the discrepancy between the point estimate of the parameter and the hypothesized value of the parameter. A test statistic is computed *under the assumption that the null hypothesis is true*.

**Example:** The  $z$ -score

$$z = \frac{\hat{p} - p}{\sqrt{p(1-p)/n}}$$

is a test statistic. What would be the value of the test statistic for the examples above?

## Decision Making

**Modus Tollens:** If  $A$  then  $B$ . Not  $B$ . Therefore not  $A$ .

**Example:** If someone is a Hobbit ( $A$ ), then their feet will be hairy ( $B$ ). Your feet are not hairy (not  $B$ ). Therefore you are not a Hobbit (not  $A$ ).

**Example:** If it rains today ( $A$ ), then the ground will be wet ( $B$ ). The ground is not wet (not  $B$ ). Therefore it did not rain today (not  $A$ ).

**“Probabilistic” Modus Tollens:** If  $H_0$  is true ( $A$ ), then the test statistic *is likely* to be a “typical” value ( $B$ ). The test statistics is not a “typical” value (not  $B$ ). Therefore  $H_0$  is *decidedly* false (not  $A$ ).

**Example:** If  $H_0$  is true ( $A$ ), then it is likely that  $-1.96 < z < 1.96$  ( $B$ ). So if  $z > 1.96$  or  $z < -1.96$  (not  $B$ ), then we decide that  $H_0$  is not true (not  $A$ ).

What can we decide?

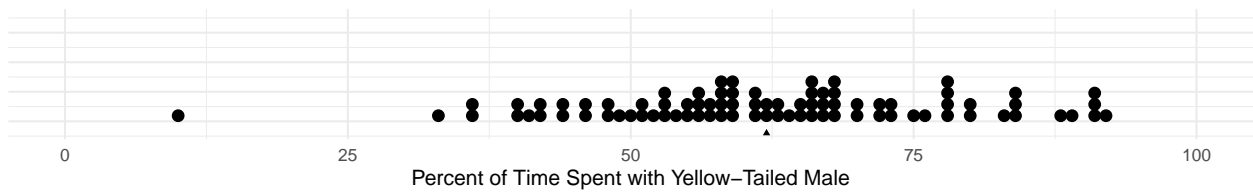
1. The test statistic *is* a “typical” value when  $H_0$  is true. *Do not reject*  $H_0$ . The result *is not* statistically significant.
2. The test statistic *is not* a “typical” value when  $H_0$  is true. *Reject*  $H_0$ . The result *is* statistically significant.

Note: This is not a true modus tollens argument. This argument can lead us to the wrong conclusion because it is still possible to observe an atypical value of the test statistic even if  $H_0$  is true.

**Example:** What might we decide for the previous examples?

## More Platies!

Do female platies have a preference for a yellow-tailed male?



In 67 out of 84 observations, the female platy spent a majority of her time with the yellow-tailed male. Is this statistically significant?