

## 9.4: P-Values

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# Definition

The **p-value** is the probability of obtaining a value of another test statistic at least as contradictory to  $H_0$  as the value you just calculated from the available sample; this probability is calculated under the assumption that  $H_0$  is true.

This should seem sort of complicated. We are comparing a hypothetical, unobserved test statistic,  $Z$ , and comparing it to our now nonrandom test statistic we just calculated,  $z$ .

# Definition

Say you had the setup:  $X_1, \dots, X_n \stackrel{iid}{\sim} \text{Normal}(\mu, \sigma^2)$ , with  $\sigma^2$  known. And  $H_0 : \mu = \mu_0$  versus  $H_a : \mu > \mu_0$ .

And say you calculated your test statistic  $z = \frac{(\bar{x} - \mu_0)}{\sqrt{\sigma^2/n}}$  (now lowercase).

Then

$$\text{p-value} = P(Z > z | \mu = \mu_0)$$

## Example 9.14 on page 456

We're doing a large-sample hypothesis test here for the population mean  $\mu$ .  $H_0 : \mu = 2.0$  versus  $H_a : \mu > 2$ . They tell us that  $z = \frac{\bar{x} - 2.0}{s/\sqrt{n}} = 3.04$ . If this was a question in one of the previous sections, we would find  $z_\alpha$  and reject if  $z > z_\alpha$ .

But let's do it with a p-value now. The values that are at least as contradictory to  $H_0$  as the test statistic is are all the numbers to the right of our test statistic. This is because  $H_a : \mu > 2$ .

$$\text{P-Value} = P(Z > 3.04 \text{ when } \mu = 2) = .0012$$

# Decision Rule

A small p-value (near 0) indicates rare-ness. Either your assumption of  $H_0$  is absurd, or you just observed some very rare data.

If  $\alpha$  is our pre-selected probability of type 1 error, the decision rule tells us to reject  $H_0$  if and only if our p-value is less than  $\alpha$ .

# Summary

Old decision rule: compare  $z$  with  $z_\alpha$ . Or compare  $t$  with  $t_{n-1,\alpha}$ .

New decision rule: compare p-value with  $\alpha$ .

The p-value is based on the percentile associated with your test statistic. All of this is just two sides of the same coin. The upshot of p-values is that they're more "report-able."

People looking at a statistical analysis can just trust you on your methodology and glance at the p-value; they don't need to know about statistical distributions.

Also, if  $\alpha = .1$ , then a p-value of .000000001 seems to say a bit more than .08.

How to calculate p-values:

- 1 For an upper-tailed test:  $P(Z \geq z)$  OR  $P(T_{n-1} \geq t)$
- 2 For a lower-tailed test:  $P(Z \leq z)$  OR  $P(T_{n-1} \leq t)$
- 3 For a two-tailed test:  $2 \times P(Z \leq -|z|)$  OR  $2 \times P(T_{n-1} \leq -|t|)$