Preview on Testing and Inference: Assessing Accuracy of Coefficient Estimates

When you run a linear regression in R and look at the summary, you see more than just the coefficient estimates.

- ▶ Suppose we have a linear model $Y = \beta_0 + \beta_1 X + \epsilon$.
- \blacktriangleright H_0 (null hypothesis): There is no relationship between X and Y
- \blacktriangleright H_a (alternative hypothesis): There is some relationship between X and Y
- Mathematically, this corresponds to testing

$$H_0: \beta_1 = 0$$

versus

$$H_a$$
: $\beta_1 \neq 0$

- ▶ To test the null hypothesis, we need to determine whether $\hat{\beta}_1$, our estimate for β_1 is sufficiently far from zero so that we can be confident that β_1 is non-zero.
- We are using the concept of the *Standard Error*: When we use linear regression, we get an **estimate** of β , which we denote by $\hat{\beta}$. This **estimate** may be too high or too low. The standard error, roughly speaking, gives an indication about how much our estimate $\hat{\beta}$ differs on average from the true parameter β .
- ▶ If the standard error is small, then even small values of β_1 could provide enough evidence that $\beta_1 \neq 0$.
- ▶ If the standard error is large, then $\hat{\beta}_1$ must be large in absolute value in order for us to reject the null hypothesis.

▶ In practice, we compute the t-statistic

$$t = \frac{\hat{\beta}_1 - 0}{\mathsf{SE}(\hat{\beta}_1)}$$

For now: We can compute the probability of observing a value equal to or larger than |t| under the null hypothesis (i.e. assuming that $\beta_1 = 0$). This probability is called the **p** value.

- Broadly speaking: A small p-value indicates that it is unlikely to observe such a substantial association between the predictor and the response due to chance, in the absence of any real association between the predictor and the response.
- ▶ If we see a small p-value, we can reject the null hypothesis.
- ➤ Typical p-value cutoffs for rejecting the null hypothesis are 0.05 or 0.01.