

Linear Model Evaluation

1. Evaluating Our Regression Model

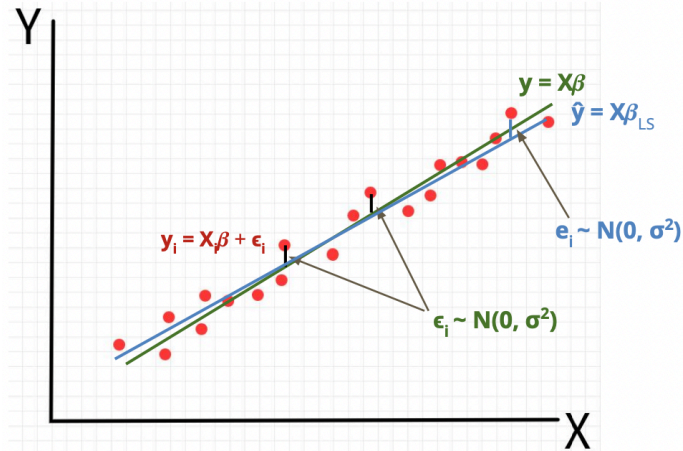
a. Some Notation

y_i is the "true" value from our data set (i.e. $\mathbf{x}_i\boldsymbol{\beta} + \epsilon_i$)

\hat{y}_i is the estimate of y_i from our model (i.e. $\mathbf{x}_i\boldsymbol{\beta}_{LS}$)

\bar{y} is the sample mean all y_i

i. $y_i - \hat{y}_i$ are the estimates of ϵ_i and are referred to as residuals



ii.

2. Metric for Evaluation for Fit of Our Model?

a. Is the value of the loss function sufficient? I.e.

$$\|y - X\beta\|_2^2 = \sum_i^n (y_i - \hat{y}_i)^2$$

i. Does not take into account the scales (income is higher and latitude is lower)

3. Evaluating Our Regression Model

a. $TSS = \sum_i^n (y_i - \bar{y})^2$ ← This is a measure of the spread of y_i around the mean of y

b. $ESS = \sum_i^n (\hat{y}_i - \bar{y})^2$ ← This is a measure of the spread of our model's estimates of y_i around the mean of y

c. $R^2 = ESS / TSS$

- i. It measures the refraction of variance that is explained by our model
(y-hat)

$$RSS = \sum_i^n (y_i - \hat{y}_i)^2$$

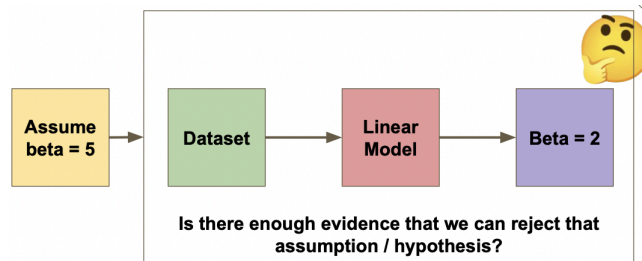
← This is what our linear model is minimizing

d.

$$R^2 = \frac{ESS}{TSS} = 1 - \frac{RSS}{TSS}$$

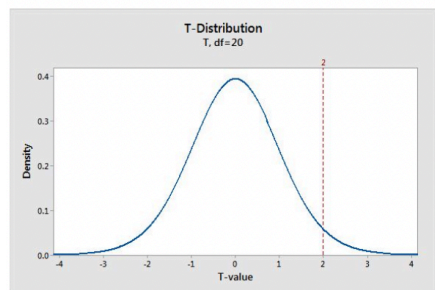
e.

4. Hypothesis Testing

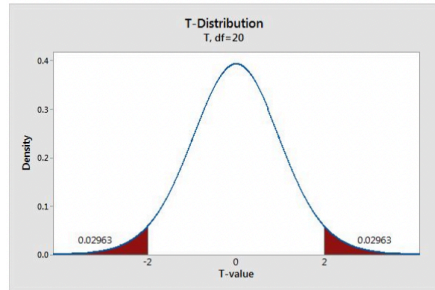


a.

- b. Each parameter of an independent variable x has an associated confidence interval and t-value + p-value
- c. If the parameter / coefficient is not significantly distinguishable from 0 then we cannot assume that there is a significant linear relationship between that independent variable and the observations y (i.e. if the interval includes 0 or if the p-value is too large)
- d. We want to know if there is evidence to reject the hypothesis $H_0: B = 0$ (i.e. that there is no linear relation between X and Y) using the information from B hat.
- e. We want to know the largest probability of obtaining the data observed, under the assumption that the null hypothesis is correct.
- f. How do we obtain that probability?
- i. A:
- g. Under the null hypothesis what should be the distribution of the normalized estimates? T-Distribution (parametrized by the sample size)



- h. We can then compute the t-value that corresponds to the sample we observed
- i. And then compute the probability of observing estimates of B at least as extreme as the one observed (i.e. trying to find evidence against H_0)
- j. The probability is called a p-value



- k. A p-value smaller than a given threshold would mean the data was unlikely to be observed under H_0 so we can reject the hypothesis H_0 . If not, then we lack the evidence to reject H_0 .