

### **Linear Regression Models**

Segment 6 – Advanced Topics in Linear Regression

Topic 3 – F-test in Linear Regression

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# **Topics**



1. Recalling Hypothesis Tests

2. The F Statistic

3. The F-test Recipe







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- From a particular dataset, we have the estimate  $\hat{\beta}_j$  and the realized value of T denoted as t.
- We calculate  $P(T \ge |t|)$ , called the p-value, and reject the null hypothesis if the p-value is smaller than a threshold, typically 0.05.





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$$\frac{(\|\mathbf{r}_S\|^2 - \|\mathbf{r}\|^2)/(p - |S|)}{\|\mathbf{r}\|/(n - p)}$$

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- If this null hypothesis is true, and if the random error terms are normally distributed, then the F distribution can be used to determine how unlikely our observed value of the F statistic is given that the null hypothesis is true.
- If the F statistic much larger than 1, then we reject the null hypothesis;
- The F-test is a simple quantification our level of uncertainty about the linear relationship of multiple predictors at once with the response variable.

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• F statistic and the F-test in the context of linear regression.

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- F statistic and the F-test in the context of linear regression.
- t-test and F-test for analyzing predictors in linear regression.