

# Module 7: Sums of Squares and Multicollinearity

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# Welcome

- You can download this set of slides. Find with the materials for the live session in Module 7.7.
- Remind me to record the live session!
- Recommended: put yourself on mute unless you want to speak.
- There is a “raise hand” button for you. Click on “Reactions” in the panel at the bottom.

# Agenda

- Some comments on Module 7
- Q&A
- Small group discussion of guided question set
- Large group discussion of guided question set plus other questions that pop up

# Uses of Regression Models

Two main uses of regression models:

- 1 Prediction
  - 2 Explore relationship between response and multiple predictors simultaneously.
- Including more predictors or higher order terms typically improves model fit (in fact  $R^2$  never decreases), but also make the model more difficult to interpret.
  - Making a model more complicated than needed can result in **overfitting**, which leads to poor predictive performance on new data.

# Partial F Test

The partial  $F$  test allows us to assess if multiple predictors can be dropped simultaneously from the model. The partial  $F$  statistic measures the change in the  $SS_R$  (or  $SS_{res}$ ) with the removal of these predictors from the model.

- Essentially, is the improvement in  $SS_R$  (and hence  $R^2$ ) large enough that it is worth the extra complexity?

# Sum of Squares

- As long as we have the same response variable,  $SS_T$  is constant, regardless of the number and form of predictors used.
- $SS_T = SS_R + SS_{Res}$
- Each time predictors are added to the model, the  $SS_R$  increases and the  $SS_{Res}$  decreases by the same amount, since  $SS_T$  stays constant.

# Model Selection

- The partial F test only works when comparing two models, the parameters of one being a subset of the parameters of another.
- Other measures are used in other situations (module 9).

# Reading ANOVA Output from R

Let's look at the example from tutorial 7.



# Issues with Multicollinearity

When predictors are nearly linear dependent on each other. Issues:

- High variance with estimated coefficients: the estimated coefficient may be very different from the true value.
  - Caution with interpreting estimated coefficients in the usual manner.
  - Estimated coefficients tend to be large.
  - Algebraic sign of coefficients different than what is known theoretically.
  - Adding or removal of one or more data points results in large changes in the estimated regression coefficients.
- Predictions are fine but must be very careful with extrapolation.

## Detecting Multicollinearity

- Insignificant  $t$  tests for predictors that are known to be useful in predicting the response variable, and significant ANOVA  $F$  test.
- High VIFs (exceeds 10).
- High correlation between pairs of predictors.

# Solutions when Multicollinearity is Present

- Consider a subset of predictors (among those that are linearly dependent).
- Shrinkage methods.
- Dimension reduction methods.

# Small Group Discussion

# What is coming up...

Module 8: Categorical predictors.