

Class 3

- First look at the p-value to see if any values have any impact on the output.
- Normalization (scaling) has no effect on the quality of model. It just gives different numbers.
- R-Square doesn't account for overfitting. That's why we use Adjusted R-Square.
- the process is, include the squared version of vars in the model, and see if this improves Adjusted R and p-value
- Median residual better be zero. Compare median with the range
- Plot fitted versus residuals, if they are outside of zero range you can see where you make errors.
- That shows if we need to square, root

Model Selection and Feature Selection

Two reasons for not using all vars

- $p > n$
- Variance-bias trade-off which is the idea of overfitting *** Two types of errors: **** Bias: How well we work with training data ***** Variance: how well do we do with other training or test data? ** Focus on the big picture and weed out predictors that are irrelevant
- Best subset selection - (2^p so we can't test all subsets) select the best
- Cp estimate of MSE how good your model would do if you run the model on data
- Cp, AIC and BIC smaller is better, Higher Adjusted Square is better
- Cp and AIC are proportional so you can use just one. BIC penalizes the num of predictors for large n - $\log(n)$ — BIC grows faster than Cp
- Adjusted R-Square first goes up and then goes down. R-Square always goes up.