

Automated Model Search

• Given several (or many) predictor variables, linear and/or higher order terms plus interactions, some unneeded, automated search for a model can be very efficient

• Full Search: Run all model subsets

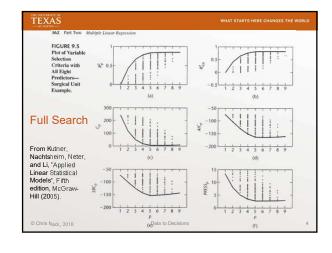
– For k parameters, there are 2<sup>k-1</sup> subsets

• Forward Stepwise Regression

– Add terms one at a time

• Backward Stepwise Regression

– Start with full model, then remove terms



Forward Stepwise Regression

• Write down full possible model

- Predictors, functions, interactions, etc.

• Pick two significance levels

-  $\alpha_{\rm add}$  (typically 0.05 or 0.10)

-  $\alpha_{\rm remove}$  (typically 0.01 or 0.05)

-  $\alpha_{\rm remove}$  must be >  $\alpha_{\rm add}$ • Step 1: regress y against each model term individually

- Calculate  $t = b_k SE(b_k)$ , and p-value for each term

- Pick smallest p-value, if it is less than  $\alpha_{\rm add}$ , then add it

Forward Stepwise Regression

• Step 2: repeat step 1, adding one new term to the existing model

- Calculate  $t = b_k / SE(b_k)$ , and p-value for each term

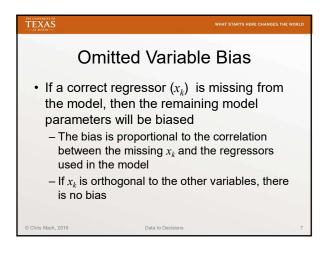
- Pick smallest p-value, if it is less than  $\alpha_{\rm add}$ , then add it to the model (now there are two terms in the model)

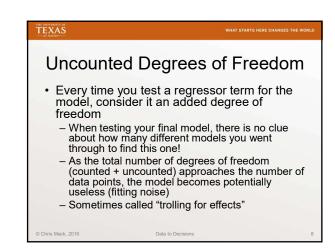
• Step 3: Check to see if a previously added term should be removed

- For all previously added terms, find the one with the lowest t-score, If p-value >  $\alpha_{\rm remove}$ , remove it

• Repeat steps 2&3 until the model no longer changes

• Backward Stepwise Regression – run in reverse





Model Validation

Just because a model can fit a data set does not mean it is good at making predictions

Model validation: estimating how good the model will be at making predictions

Three validation approaches

Model coherence – how well does this model fit within the framework of existing knowledge?

Compare model to new data

Data splitting: randomly split your data in two, use first group to calibrate (training sample), second group to validate (hold-out sample, test sample)

Comparing Model to New Data

• Regress model to new data – how have the coefficients changed?

- Use two-sample pooled t-test for each coefficient

• Compare existing model's predictions to new data

- Mean squared predicted residuals (MSPR)

- If MSPR is not too much greater than the original regression mean square error (MSE), the model is "valid" – we typically look for less than a factor of 2

Data Splitting

Typical data splitting uses about 50-80% of the data for training, 20-50% for validation

The data must be split randomly

E.g., break the data into five parts, then use five different 80/20 splits to train and test

Called 5-fold cross validation

Generalization: k-fold cross validation

Repeat many times with different random splits

Optimism principle: The validation data set will always have worse MSPR than the training MSE

Lecture 62: What have we learned?

• What are the three automated model search approaches?

• Explain omitted variable bias

• How can model building result in uncounted degrees of freedom?

• What are the three model validation approaches?