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- Basic cleaning of data.
- Using data on 'Single Family Housing'.

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- Use the lag of X variables.
- Y variable from 2013 and X variables from 2011.
- This will allow predictions about future market value of housing units.





Use the VALUE variable from 2013 data and X variables from 2011 data.

Merge the VALUE variable from 2013 data into the 2011 data.

(VLOOKUP command using the CONTROL variable to match the data files)



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- Data cleaning.
 - > Delete all rows for housing units that are not common across the two years.
 - Consider only 'Single Family Housing' TYPE = 1 and STRUCTURETYPE = 1
 - Delete all Housing units which have a market value of less than \$1000, that is, delete units with VALUE < \$1000</p>



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- □ R-square measure.
- 'Holdout Analysis'.
 - > Hold out some data and not include it in the regression.
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- □ From the data that you create, select 1000 Housing Units at random. This is your 'Holdout Sample'.
- □ To select a random sample...
 - > Use the =RAND() function in Excel.
 - Sort the data on this random value and select the top 1000 Housing Units for 'Holdout Sample'.
 - > Estimate regression model on the remaining set of Housing Units.
- Using 'β' coefficients from regression model and X variables from 'Holdout Sample', predict VALUE for each Housing Unit in the Holdout Sample.



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- Compare these thousand predictions with the actual market value of those housing units.
 - > Create a 'Mean Absolute Deviation' measure.
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Mean Absolute Deviation =
$$\left(\sum_{i=1}^{1000} |Actual_i - Predicted_i|\right) / 1000$$



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- 2. The calculation of 'Mean Absolute Deviation'.
- 3. You may extend analysis to multiple years.



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