

## Machine Learning Concepts and Basics Notes



### Linear Regression:

- $y \approx B_0 * x_0 + B_1 * x_1 = y_p$ 
  - $B_i$  : unknown constants
  - $X_i$  : independent variable
  - $y$ : dependent variable

Not all  $x_i$  can be mapped to  $y_i$  perfectly there will almost always be an amount of deviation, hence the use of “ $\approx$ ”(almost equal to) symbol.

- The best fit trend line is an educated guess regarding what the linear relationship between  $x$  and  $y$  are.
  - $R^2$  value/correlation measures how related  $x$  and  $y$  are, closer to 1 the better.
- Sum of the squared errors (SSE) is the sum of the differences between  $y$  and predicted  $y_p$
- T-test is used to find the correlation between 2 dependent variables.

### Logistical Regression:

- Classification model, primarily binary
  - Find relation between predictor variables and categorical variables
- Subtypes:
  - Binary log: Yes/No decisions
  - Nominal log: 3+ categories with not natural ordering to levels i.e. search engines, colors, letters
  - Ordinal log: 3+ categories with a natural ordering to levels i.e. effectiveness of course, severity of medical condition, restaurant rating
- Metrics:
  - Wald Test: tests significance of individual coefficients (similar to t-test)
  - Goodness-of-fit-tests: used to find if observed values match expected value under model
    - Most common used: Chi-squared method

## Decision Trees:

- Decision trees are non-parametric supervised learning method used for classification and regression
  - Non-parametric: a statistical method in which the data are not assumed to come from prescribed models that are determined by a small number of parameters
- Random Forest Classifiers: each tree votes for a value and the result that has the most votes is chosen as the predicted value.
  - Primary concept of random forest and other forest type algorithms is:
    - A large number of uncorrelated models(trees) operating as a committee will outperform any individual constituent models
    - Low correlation between trees is important since each tree's error is covered by its neighbors
- Random Forest Regression: operates almost the same way as classifiers except all the results given by each tree are averaged together to generate a single value.

*\*For more workshops and information visit [go.tamui.edu/arc](http://go.tamui.edu/arc)*