Tips for Lab Sessions

- Get preparation: get familiar / try to tackle the problems before coming to lab.
- Do **ALL** the problems!
- Format you answer: use Markdown to organize your answer / conclusion.
- Watch the time: the DDL is at XX:20 PM, not XX:30 PM.
- Learn to Google for usage of basic functions.
- Do NOT mail your answer: submit your work **ONLY** via **NTU-Learn**.

Lab 4. Linear Regression

Linear Regression

• Goal: how to make quantitative prediction on Y given a good variable X (i.e. corr(X, Y) is high)?

Rationale: find coefficients (a, b) such that

$$Y \approx a + b X$$

holds on training data.

How: fit a linear regression model;
 evaluate the model performance on the testing data.

Linear Regression Workflow (Lab Mark Checkpoints)

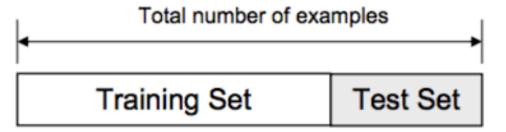
- 1. Split your Dataset: randomly split the dataset into training v.s. test dataset.
- 2. Fit & Evaluate Linear Model: **fit** a linear regression model with **different variable** *X* on the **training** set; **evaluate** the model on the **test** set.
- 3. <u>Model Selection</u>: **find the 'best'** model according to some performance metric.
- 4. Refine model: fit regression model again on outliers-free data.

Mark Checkpoint 1: Split the dataset

 Randomly split the train / test dataset.

Print the shape of train
 / test dataset

Import the required function from sklearn
from sklearn.model_selection import train_test_split

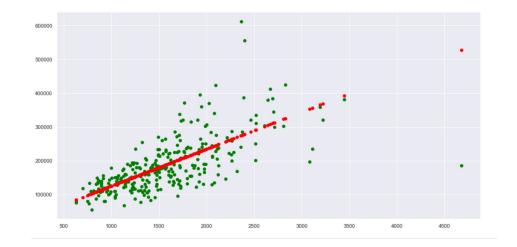


Mark Checkpoint 2: Fit Linear Models

Fit a linear model with training data.

Import LinearRegression model from Scikit-Learn from sklearn.linear_model import LinearRegression

- Google for basic attributes of LinearRegression().
- Plot the regression line with test data. Google for how to add plots on a plot.



Mark Checkpoint 3: Model Selection

- Compute goodness of fit: MSE, R^2 on training & test data.
- Explain in your word: Do you think this MSE is too large to be accurate?
- Compare multiple univariate models w.r.t both MSE and R^2 .

Mark Checkpoint 4: Refine your Model

- Identify the row indices & count the number of outliers.
- Visualize the outliers via boxplot.

- Remove the outliers using drop() function.
- Repeat the previous split + fit + evaluate process with the clean data.
- Compare this model with previous ones.