

# Exercise 4: Linear Regression

## Setup of Working Environment

1. Create a folder on your Desktop and name it EE0005\_[LabGroup], where [LabGroup] is the name of your Group.
2. Download the .ipynb files and data files posted corresponding to this exercise and store in the aforesaid folder.
3. Open Jupyter Notebook and navigate to the aforesaid folder on Desktop.
4. Open and explore the .ipynb files (notebooks) that you downloaded, and go through “Preparation”, as follows.
5. The walk-through videos posted on NTU Learn may help you with this “Preparation” too.
6. Create a new Jupyter Notebook, name it Exercise4\_solution.ipynb, and save it in the same folder on the Desktop

## Preparation for the Exercises

M3 LinearRegression.ipynb

Check how to perform Linear Regression on the Pokemon data (pokemonData.csv)

## Objective

In the last Example Class, we have identified and analyzed some of the most relevant numeric variables in this dataset, which may affect the sale price of a house, and hence, will probably be most relevant in predicting “SalePrice”. In this Example Class, we will extract those numeric variables one-by-one and perform Linear Regression to predict “SalePrice”.

## Problems

### Problem 1 : Predicting SalePrice using GrLivArea

Download the Kaggle dataset “train.csv” from NTU Learn, posted corresponding to this Example Class. Extract the following Numeric variables from the dataset, and store as two new Pandas DataFrames.

```
houseGrLivArea = pd.DataFrame(houseData[ 'GrLivArea' ])      Above ground living area in SqFt
houseSalePrice = pd.DataFrame(houseData[ 'SalePrice' ])      Sale Price of house in US Dollars
```

- a) Plot houseSalePrice against houseGrLivArea using standard jointplot, and note the strong linear relationship. Remember the correlation coefficient between these two variables from the last Example Class? Check again.
- b) Import Linear Regression model from Scikit-Learn : `from sklearn.linear_model import LinearRegression`
- c) Partition both datasets houseGrLivArea and houseSalePrice into Train (1100 rows) and Test (360 rows) sets.  
Train datasets : houseGrLivArea\_train and houseSalePrice\_train (check both have 1100 rows)  
Test datasets : houseGrLivArea\_test and houseSalePrice\_test (check both have 360 rows)
- d) Training : Fit a Linear Regression model with `X = houseGrLivArea_train` and `y = houseSalePrice_train`
- e) Print the coefficients of the Linear Regression model you just fit, and plot the Regression line on a Scatterplot of houseGrLivArea\_train and houseSalePrice\_train using the standard slope-intercept form of straight line.
- f) Predict SalePrice for the test dataset houseGrLivArea\_test using the Linear Regression model, and plot the predictions on the Scatterplot of houseGrLivArea\_test and houseSalePrice\_test to visualize the accuracy.
- g) Find the Explained Variance ( $R^2$ ) of the model on the Train set and on the Test set to check Goodness of Fit.

### Problem 2 : Predicting SalePrice using Other Variables

Perform all the above steps on 'SalePrice' against each of the variables 'LotArea', 'TotalBsmtSF', 'GarageArea' one-by-one to perform individual Linear Regressions. Discuss with your Friends about the models, compare and contrast the Explained Variance ( $R^2$ ), check the predictions, and determine which model is the best to predict 'SalePrice'.