

## Module 3 Lab:

# Outlier Detection with IQR (Inter-quartile Range)

## OBJECTIVE

Interquartile range (IQR) ([https://en.wikipedia.org/wiki/Interquartile\\_range](https://en.wikipedia.org/wiki/Interquartile_range)) is a measure of statistical dispersion, being equal to the difference between 75th and 25th percentiles, which is a popular method to detect outliers for preprocessing the data. The purpose of this lab is to implement IQR algorithm for outlier detection.

## PREREQUISITES

You have to complete all submodule 3.2 lecture slides and be familiar with IQR algorithm below.

### IQR Algorithm for Outlier Detection

1. Arrange the data in ascending order
2. Calculate Q1 (the first Quarter)
3. Calculate Q3 (the third Quartile)
4. Calculate IQR = (Q3 - Q1)
5. Calculate the lower Range  $T_{lower} = Q1 - (1.5 * IQR)$
6. Calculate the upper Range  $T_{upper} = Q3 + (1.5 * IQR)$
7. Detect outliers with the lower Range and the upper Range. If the data is not in the range  $[T_{lower}, T_{upper}]$ , then the data will be filtered out as outliers.

## INSTRUCTIONS

- Implement IQR algorithm with Python
- Test IQR on the attribute "LotArea" of training data of **House Price Prediction** from Kaggle Data to detect and remove its outliers. The data can be downloaded from Kaggle (<https://www.kaggle.com/c/house-prices-advanced-regression-techniques/overview>).
- Compare the original data and preprocessed data by plotting ([https://matplotlib.org/3.1.1/gallery/pyplots/boxplot\\_demo\\_pyplot.html#sphx-glr-gallery-pyplots-boxplot-demo-pyplot-py](https://matplotlib.org/3.1.1/gallery/pyplots/boxplot_demo_pyplot.html#sphx-glr-gallery-pyplots-boxplot-demo-pyplot-py))