29/03/2024

Week-2 & Week-3

End-to-End Machine Learning Project

1. Problem formulation
   Formulate the problem by considering what are is the expected output and the features available.
2. Get the data
   * import os
     - provides functions for interacting with operating system
   * import urllib
     - provides functions for fetching URLs
   * import pandas as pd
     - provides functions to work with data sets
   * housing.head()
     - returns the top 5 rows of the file
   * housing.info()
     - gives the count of non-null values in each attribute
   * housing.describe()
     - gives the statistical measures such as count, mean, std, min, 25%, 50%, 75%, max for each attribute/features.
   * import matplotlib.pyplot as plt
     - a collection of command style functions which works like MATLAB.
   * housing.hist(bins=50, figsize=(20,15))
     - histogram is plotted which is used for univariate analysis and bins are used to specify the width of each group.
   * split_train_test()
     - it divides the dataset into train set and test set. test-ratio can also be specified to divide the dataset.
   * train_set.shape
     - gives the shape of the train set (i.e no. of rows and columns)
   * housing['id'].value_counts()
     - gives the count of values

3. Discover and visualize the data to gain insights
   * strain_test_set, shape()
   * We plot scatterplot by using
     housing.plot(kind='scatter', x='long', y='lat')
     plt.show()
   * Looking for correlation
     corr_matrix = housing.corr()
   * Using scattermatrix to plot correlations
     scatter_matrix(frame = housing[attributes], figsize=(12,8))

4. Prepare the data for machine learning algorithms.
   → Data cleaning
     * Get rid of the whole attribute
       housing.drop('total_bedrooms', axis=1)
     * Set missing values to some value
       zero, mean, median
   → Handling text & categorical attribute
     * Use 'OrdinalEncoder' to encode categorical features into ordinal integers.
   → Custom Transformers
     * We use fit, transform, fit_transform methods
   → Feature Scaling
     * Min-Max Scaling: for each value, we subtract by min & divide by max-min
     * Standardization: for each value, we subtract mean & divide by std

5. Select and train a model
   We use linear regression model
   DecisionTreeRegressor - It is a machine learning algo. used for regression tasks, where the goal is to predict a continuous target variable.

6. Fine-Tune your model
   → GridSearchCV is a model provided by scikit-learn library in Python for hyperparameter tuning of machine learning models.
   → RandomizedSearchCV can be used instead of above one
   → Evaluate your system on the test set by using mean_squared_error method.

7. Launch, Monitor & Maintain your System

We can automate this process by

- Collecting fresh data regularly & labeling it
- writing script to train models & fine tune the hyper-parameter
- writing script to evaluate the model