Concrete Compressive Strength Prediction

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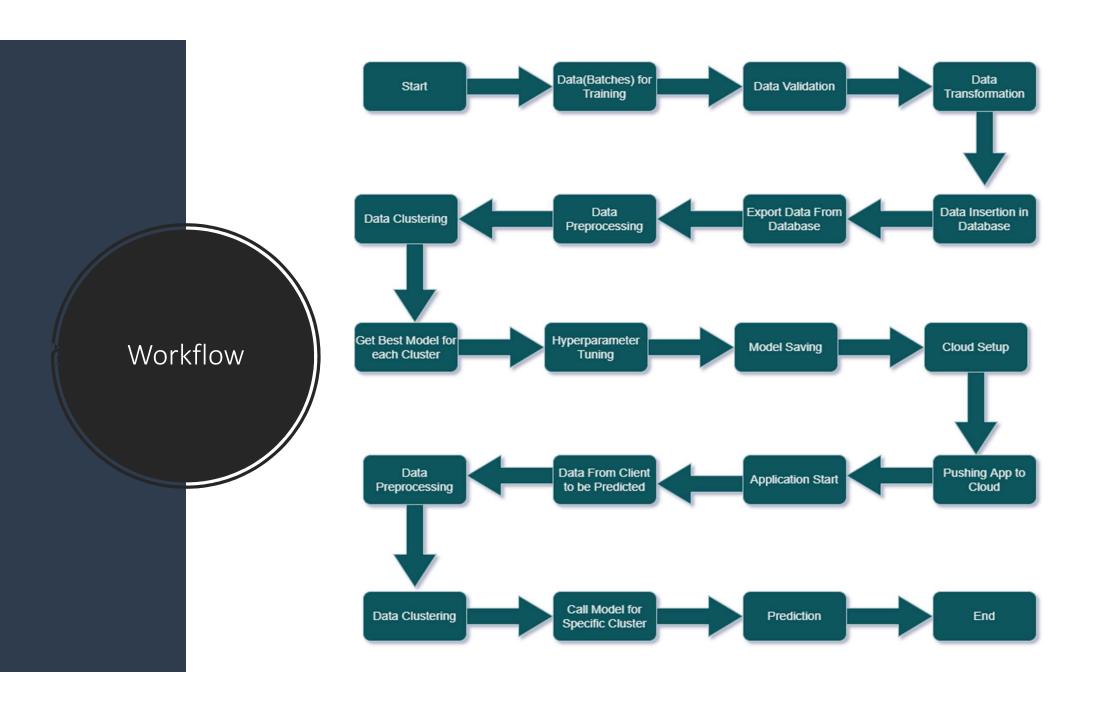
Introduction

 Concrete is most common and oldest material for construction worldwide. Infrastructure and construction are considered as extremely sensitive domains regarding the safety issue. So, since concrete being the main component or material for the construction it would be important to measure the strength or the power of the concrete using some reliable methods. Concrete is basically a composite material composed of various base materials like cement, water, Coarse Aggregate, Fine Aggregate, and some other components. Compressive strength of concrete is measured using a conventional crushing test on a concrete cylinder. Basically, it takes 28 days of time.

Objective

The quality of concrete is determined by its compressive strength, which is measured using a conventional crushing test on a concrete cylinder. The strength of the concrete is also a vital aspect in achieving the requisite longevity. It will take 28 days to test strength, which is a long period.

We can save a lot of time and effort by using Data Science to estimate how much quantity of which raw material we need for acceptable compressive strength.



Dataset

Description

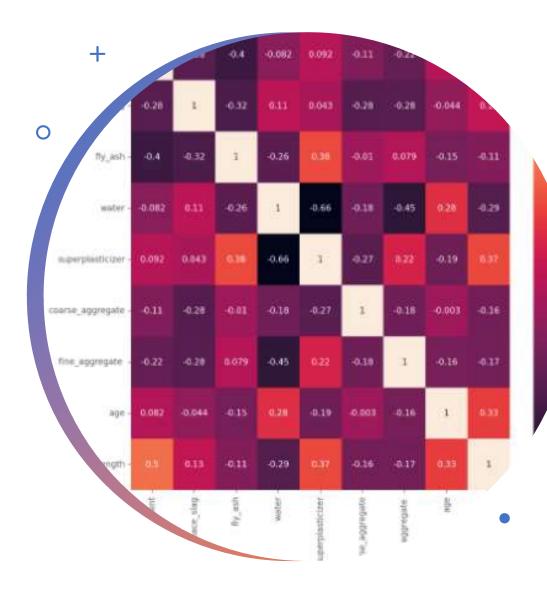
- The Following Dataset is used for the model training
- Which is Basically taken from UCI Public Repository:
- https://archive.ics.uci.edu/ml/ datasets/Concrete+Compressive +Strength

Name	Datatype	Measureme	Descriptio
		nt	n
Cement	Quantitativ	kg in a m3	Input
	e	mixture	Variable
Blast Furnace	Quantitativ	kg in a m3	Input
Slag	e	mixture	Variable
Fly Ash	Quantitativ	kg in a m3	Input
45.50	e	mixture	Variable
Water	Quantitativ	kg in a m3	Input
	e	mixture	Variable
Superplasticiz	Quantitativ	kg in a m3	Input
er	e	mixture	Variable
Coarse	Quantitativ	kg in a m3	Input
Aggregate	e	mixture	Variable
Fine	Quantitativ	kg in a m3	Input
Aggregate	e	mixture	Variable
Age	Quantitativ	kg in a m3	Input
VC 10.	e	mixture	Variable
Concrete	Quantitativ	Megapascal	Output
compressive	e		Variable
strength			

Data Preprocessing

- This data is having 9 columns:
- 8: Independent columns
- 1: Dependent column
- Here the datatypes of the columns are perfect, No need to update or typecast
- Here there are no Null values

```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1030 entries, 0 to 1029
Data columns (total 9 columns):
                                   Non-Null Count Dtype
     Column
                                   1030 non-null float64
    cement
    blast_furnace_slag
                                   1030 non-null
                                                 float64
    fly ash
                                  1030 non-null float64
    water
                                   1030 non-null float64
    superplasticizer
                                  1030 non-null float64
   coarse aggregate
                                   1030 non-null float64
    fine aggregate
                                   1030 non-null float64
                                   1030 non-null
                                                 int64
    concrete compressive strength 1030 non-null
                                                 float64
dtypes: float64(8), int64(1)
memory usage: 72.5 KB
```



EDA — Data Visualization

- Here Exploratory Data Analysis is done.
- Here no multicollinearity issue detected
- But by plotting the distribution plots of all the columns, it was detected that : not all columns are having normal or Gaussian distribution.
- So in the feature engineering step, feature scaling will be done.



Feature Engineering

- Data was clean but some of the feature engineering was needed.
- All the columns were scaled with Standardization to, which were required for clustering algorithm and other Machine Learning process.
- Data Imputation is done of there is any Null values in any new incoming data

Model Building

The pipeline in the program was set up that would:



Automatically scale and impute the data



Find optimum numbers of clusters with Silhoutte score



Perform Cross Validation and Hyperparameter tuning separately for each cluster



Create the clusters from the data which has optimum Silhoutte score

Model Building



AFTER PERFORMING
HYPERPARAMETER
TUNING GET BEST
R_SQUARED OF EACH
MACHINE LEARNING
MODEL.



SAVE ALL THE PERFORMANCE METRICES AND SCORES IN THE RESPECTING LOG FILE.



THEN SAVE THE
FINAL MODEL
WHICH HAS THE
MAXIMUM
R_SQUARED BY
REPLACING
EXISTING MODELS.



APPLICATION IS READY FOR THE PREDICTION.



Model deployment is done on Heroku server.

Link: https://ml-concrete-strength.herokuapp.com/



Thank You