



Data Collection and Preprocessing Phase

Date	9 July 2024
Team ID	SWTID1720162737
Project Title	Predicting Compressive Strength Of Concrete Using Machine Learning.
Maximum Marks	6 Marks

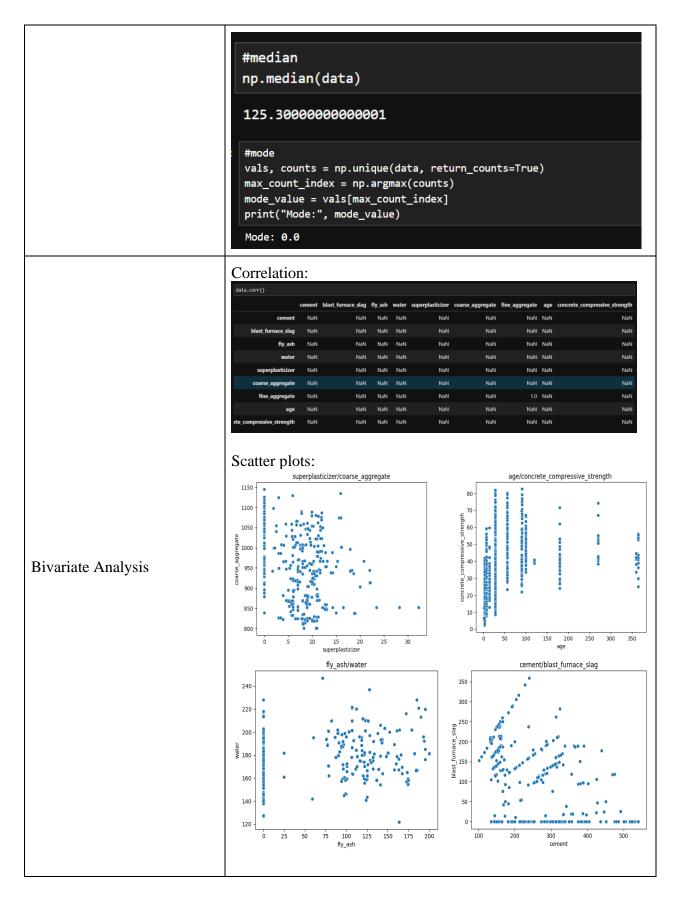
Data Exploration and Preprocessing Template

Identifies data sources, assesses quality issues like missing values and duplicates, and implements resolution plans to ensure accurate and reliable analysis.

Section	Description		
Data Overview	Dimensions:		
	[7]: data.shape		
	[7]: (1030, 9)		
	Descriptive statistics:		
Bata Overview	[9]: data_describe()		
	ount 1030,000000 1030,000000 1030,000000 1030,000000 1030,000000 1030,000000 1030,000000 1030,000000 1030,000000		
	nean 281.167864 73.895825 54.188350 181.567282 6.204660 972.918932 773.580485 45.662136 35.81794		
	std 104.506364 86.279342 63.997004 21.354219 5.973841 77.753954 80.175980 63.169912 16.70574		
	min 102.000000 0.000000 0.000000 121.800000 0.000000 801.000000 594.000000 1.000000 2.33000		
	25% 192.375000 0.000000 0.000000 164.900000 0.000000 932.000000 730.950000 7.000000 23.71000		
	50% 272,900000 22,000000 0.000000 185,000000 6400000 968,000000 779,500000 28,000000 34,44500		
	75% 350,000000 142,950000 118,300000 192,000000 10,200000 102,400000 824,000000 56,000000 46,13500		
	max 540,000000 359,400000 200,100000 247,000000 32,200000 1145,000000 992,600000 365,000000 82,60000		
Univariate Analysis	: #mean np.mean(data) : 269.444832793959		

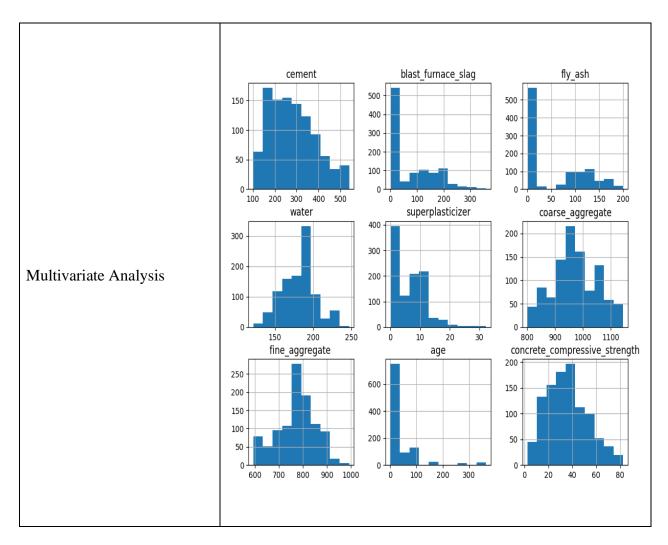






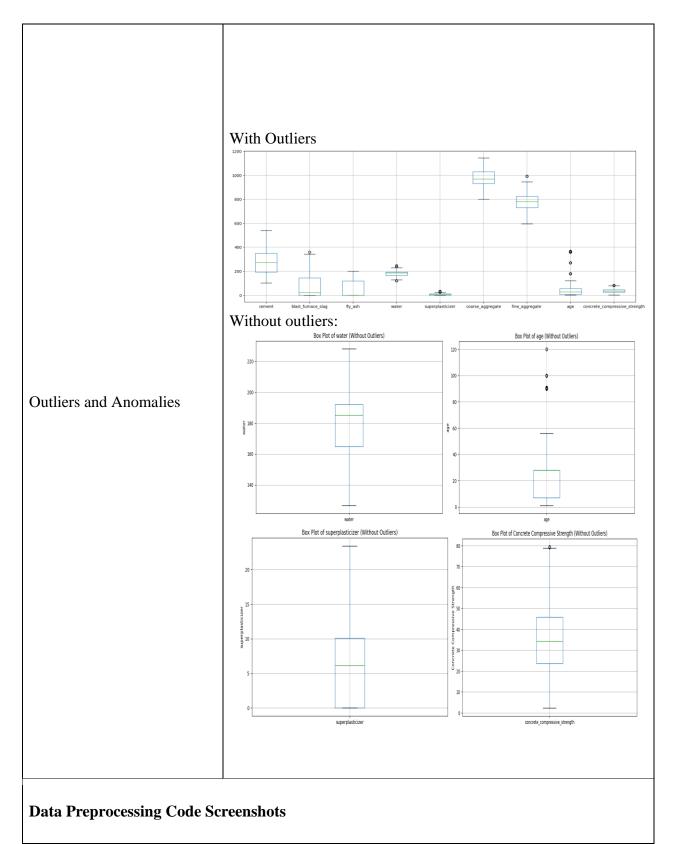
















Loading Data	data=pd.read_csv('concrete_data.csv') cement blast_furnace_slag fly_ash water superplasticizer coarse_aggregate fine_aggregate age concrete_compressive_strength 0 540.0 0.0 0.0 162.0 2.5 1050.0 676.0 28 79.99 1 540.0 0.0 0.0 162.0 2.5 1055.0 676.0 28 61.89 2 332.5 142.5 0.0 228.0 0.0 932.0 594.0 270 40.27 3 332.5 142.5 0.0 228.0 0.0 932.0 594.0 365 41.05 4 198.6 132.4 0.0 192.0 0.0 978.4 825.5 360 44.30
Handling Missing Data	data['cement']=data['cement'].fillna(data['cement'].mode()[0]) data['blast_furnace_slag']=data['blast_furnace_slag'].fillna(data['blast_furnace_slag'].mode()[0]) data['fly_ash']-data['fly_ash'].fillna(data['fly_ash'].mode()[0]) data['water']=data['water'].fillna(data['water'].mode()[0]) data['coarse_aggregate']=data['coarse_aggregate'].fillna(data['coarse_aggregate'].mode()[0]) data['superplasticizer']=data['ager'].mode()[0]) data['age']=data['age'].fillna(data['age'].mode()[0]) data['concrete_compressive_strength']=data['concrete_compressive_strength'].fillna(data['concrete_compressive_strength'].mode()[0]) But it is not necessary cause we don't have any missed values in the data set.
Data Transformation	<pre>#Scaling on Independent variables from sklearn.preprocessing import Standardscaler scale=StandardScaler() names=x.columns names Index(['cement', 'blast_furnace_slag', 'fly_ash', 'water', 'superplasticizer',</pre>
Feature Engineering	Attached the codes in final submission.
Save Processed Data	data=filtered_data data.shape (1021, 9)