



Data Collection and Preprocessing Phase

Date	15 July 2024
Team ID	739673
Project Title	Early Prediction Of Chronic Kidney Disease
Maximum Marks	6 Marks

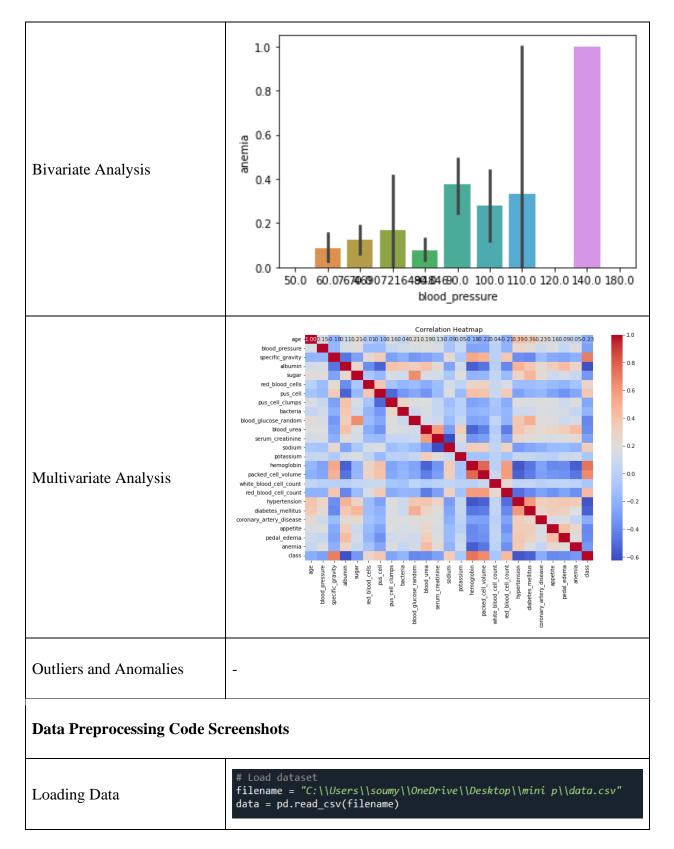
Data Exploration and Preprocessing Template

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 399rows x 26columns id age bp sg al su rbc pc pc ba pc wc rc htn dm cad appet pe ane classification 0 0 480 80.0 1.020 1.0 0.0 NaN normal notpresent notpresent 38 6000 NaN no no no good no no ckd 1 1 7.0 50.0 1.020 4.0 0.0 NaN normal notpresent notpresent 38 6000 NaN no no no good no no ckd 2 2 62.0 80.0 1.010 2.0 3.0 normal normal notpresent 31 7500 NaN no yes no poor no yes ckd 3 3 48.0 70.0 1.005 4.0 0.0 normal abnormal present notpresent 32 6700 3.9 yes no no poor yes yes ckd 4 4 51.0 80.0 1.010 2.0 0.0 normal normal notpresent 35 7300 4.6 no no no good no no ckd
Univariate Analysis	Distribution of age Distribution of blood gressure Distribution of sugar











Handling Missing Data	# Handle missing values data = data.apply(lambda x: x.fillna(x.mean()) if x.dtype.kind in 'biufc' else x.fillna(x.mode().iloc[0]))
Data Transformation	<pre># Clean categorical variables with incorrect values data['class'] = data['class'].replace('ckd\t', "ckd'') data['coronary_artery_disease'] = data['coronary_artery_disease'].replace('\tno', 'no') data['diabetes_mellitus'] = data['diabetes_mellitus'].replace(to_replace={'\tno': 'no', '\tyes': 'yes'; 'yes';)} # Encode categorical variables cat_cols = data.select_dtypes(include=['object']).columns label_encoders = {} for col in cat_cols: le = LabelEncoder() data[col] = le.fit_transform(data[col]) label_encoders[col] = le</pre>
Feature Engineering	Attached code in final submission.
Save Processed Data	-