



## **Data Collection and Preprocessing Phase**

Date	15 July 2024
Team ID	
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	Descrip	tion				
		age	blood_pressure	specified_gravity	albumin	sugar ı
	0	48.0	80.0	3	1	0
	1	7.0	50.0	3	4	0
	2	62.0	80.0	1	2	3
	3	48.0	70.0	0	4	0
	4	51.0	80.0	1	2	0
Data Overview						
	395	55.0	80.0	3	0	0
	396	42.0	70.0	4	0	0
	397	12.0	80.0	3	0	0
	398	17.0	60.0	4	0	0
	399	58.0	80.0	4	0	0
	400 rc	ws × 2	5 columns			

red_blood_cells	pus_cell	pus_cell_clumps	bacteria
1	1	0	0
1	1	0	0
1	1	0	0
1	0	1	0
1	1	0	0
1	1	0	0
1	1	0	0
1	1	0	0
1	1	0	0
1	1	0	0

blood_glucose_random	 packed_cell_volume	white_blood_cell_count
121.000000	44.0	7800.0
148.036517	38.0	6000.0
423.000000	31.0	7500.0
117.000000	32.0	6700.0
106.000000	35.0	7300.0
140.000000	47.0	6700.0
75.000000	54.0	7800.0
100.000000	49.0	6600.0
114.000000	51.0	7200.0
131.000000	53.0	6800.0

red_blood_cell_count	hypertension	diabetesmellitus	coronary_artery disease
5.200000	1	2	0
4.707435	0	1	0
4.707435	0	2	0
3.900000	1	1	0
4.600000	0	1	0
4.900000	0	1	0
6.200000	0	1	0
5.400000	0	1	0
5.900000	0	1	0
6.100000	0	1	0

appetite	pedal_edema	anemia	class
арреше	pedal_edeiiid	ancina	Class
0	0	0	0
0	0	0	0
1	0	1	0
1	1	1	0
0	0	0	0
0	0	0	1
0	0	0	1
0	0	0	1
0	0	0	1
0	0	0	1

```
numerical_columns = ['age', 'blood_pressure', 'blood_glucose_random', 'blood_urea',
                                     'serum_creatinine', 'sodium', 'potassium', 'hemoglobin',
                                     'packed_cell_volume', 'white_blood_cell_count', 'red_blood_cell_count
Univariate
               for col in numerical_columns:
Analysis
                   plt.figure(figsize=(10, 6))
                   sns.histplot(data[col], kde=True)
                   plt.title(f'Distribution of {col}')
                   plt.xlabel(col)
                   plt.ylabel('Frequency')
                   plt.show()
               # Bivariate analysis between two numerical variables
               plt.figure(figsize=(10, 6))
               sns.scatterplot(x='age', y='blood_pressure', data=data)
               plt.title('Scatter Plot between Age and Blood Pressure')
Bivariate
               plt.xlabel('Age')
Analysis
               plt.ylabel('Blood Pressure')
               plt.show()
               plt.figure(figsize=(12, 10))
               sns.pairplot(data[numerical_columns])
               plt.suptitle('Pair Plot for Numerical Columns', y=1.02)
               plt.show()
Multivariate
               17.5
Analysis
               10.0
               7.5
Outliers and
               Identification and treatment of outliers.
Anomalies
Data Preprocessing Code Screenshots
```

Code to load the dataset into the preferred environment (e.g., Python, R).

Loading Data





Handling Missing Data	<pre>data['blood_glucose_random'].fillna(data['blood_glucose_random'].mean(),inplace=True) data['blood_pressure'].fillna(data['blood_pressure'].mean(),inplace=True) data['blood_urea'].fillna(data['blood_urea'].mean(),inplace=True) data['packed_cell_volume'].fillna(data['packed_cell_volume'].mean(),inplace=True) data['potassium'].fillna(data['potassium'].mean(),inplace=True) data['red_blood_cell_count'].fillna(data['red_blood_cell_count'].mean(),inplace=True) data['senum_creatinine'].fillna(data['senum_creatinine'].mean(),inplace=True) data['sodium'].fillna(data['sodium'].mean(),inplace=True) data['white_blood_cell_count'].fillna(data['white_blood_cell_count'].mean(),inplace=True) data['age'].fillna(data['age'].mode()[0],inplace=True) data['pus_cell_clumps'].fillna(data['pus_cell_clumps'].mode()[0],inplace=True) data['appetite'].fillna(data['appetite'].mode()[0],inplace=True) data['abbumin'].fillna(data['appetite'].mode()[0],inplace=True) data['pus_cell'].fillna(data['pus_cell'].mode()[0],inplace=True) data['red_blood_cells'].fillna(data['red_blood_cells'].mode()[0],inplace=True) data['coronary_artery disease'].fillna(data['coronary_artery disease'].mode()[0],inplace=True) data['anemia'].fillna(data['bacteria'].mode()[0],inplace=True) data['anemia'].fillna(data['sugar'].mode()[0],inplace=True) data['diabetesmellitus'].fillna(data['diabetesmellitus'].mode()[0],inplace=True) data['diabetesmellitus'].fillna(data['diabetesmellitus'].mode()[0],inplace=True) data['diabetesmellitus'].fillna(data['docada edema'].mode()[0],inplace=True)</pre>
	<pre>data['diabetesmellitus'].fillna(data['diabetesmellitus'].mode()[0],inplace=True) data['pedal_edema'].fillna(data['pedal_edema'].mode()[0],inplace=True) data['specified_gravity'].fillna(data['specified_gravity'].mode()[0],inplace=True)</pre>
Data Transformation	<pre>data.packed_cell_volume=pd.to_numeric(data.packed_cell_volume,errors='coerce') data.white_blood_cell_count=pd.to_numeric(data.white_blood_cell_count,errors='coerce') data.red_blood_cell_count=pd.to_numeric(data.red_blood_cell_count,errors='coerce')</pre>

Feature Engineering	
Save Processed Data	