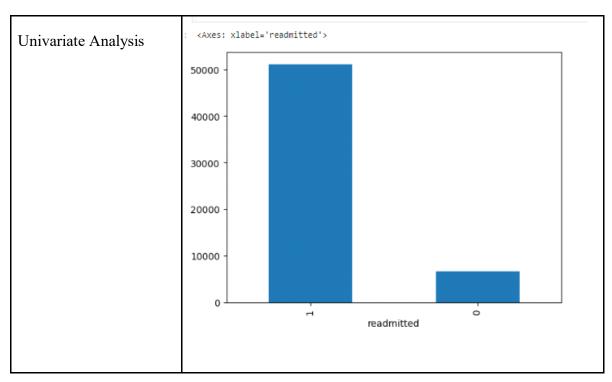
Data Collection and Preprocessing Phase

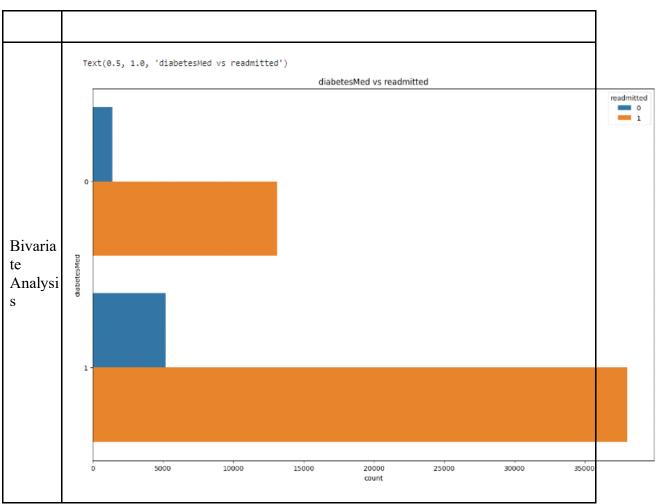
Date	20 JUNE 2024
Team ID	team-739679
Project Title	HOSPITAL READMISSION PREDICTOR
Maximum Marks	6 Marks

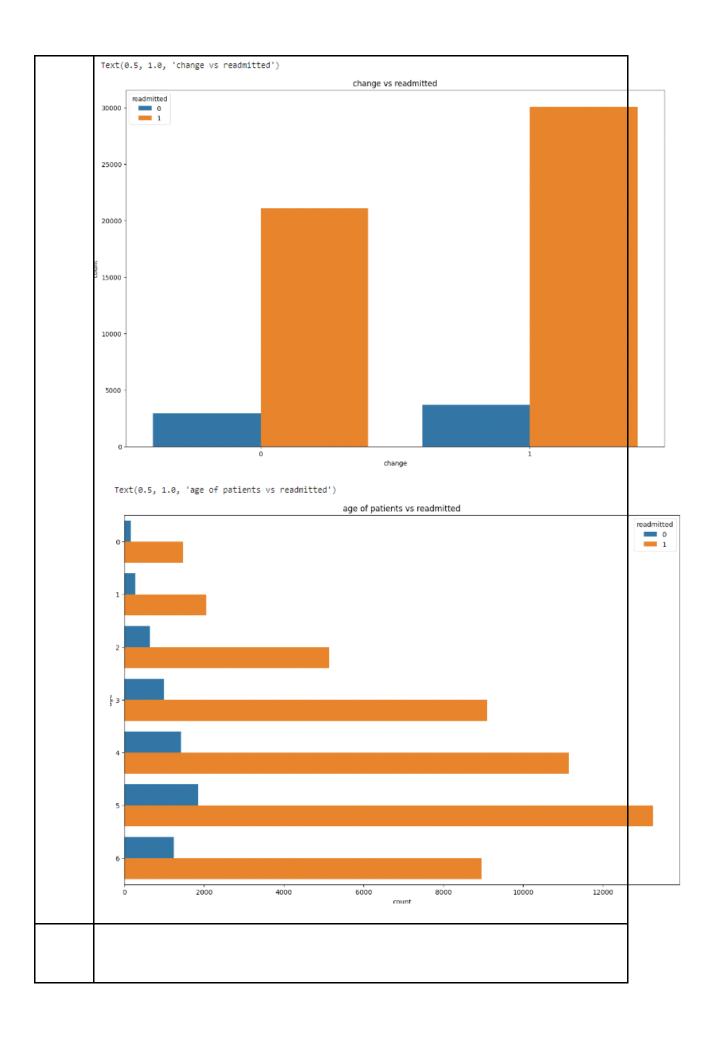
Data Exploration and Preprocessing Report

Dataset variables will be statistically analyzed to identify patterns and outliers, with Python employed for preprocessing tasks like normalization and feature engineering. Data cleaning will address missing values and outliers, ensuring quality for subsequent analysis and modeling, and forming a strong foundation for insights and predictions.

Section	Description								
	Dime 5 row statis	$r_{\rm S} \times 2$		lur	nns <u>Desc</u>	<u>riptive</u>			
		race	gender	age	admission_type_id	discharge_disposition_id	admission_source_id	time_in_hospital	num_lab_procedure
	0	Caucasian	Female	[0- 30)	Others	Others	Referral	-1.632460	-0.10847
Data Overview	1	Caucasian	Female	[0- 30)	Emergency	Discharged to home	Emergency Room	-0.271144	0.80852
	2 Africa	nAmerican	Female	[0- 30)	Emergency	Discharged to home	Emergency Room	-0.825587	-1.62961
	3	Caucasian	Male	[30- 40)	Emergency	Discharged to home	Emergency Room	-0.825587	0.04419
	4	Caucasian	Male	[40- 50)	Emergency	Discharged to home	Emergency Room	-1.632460	0.40069
	5 rows × 2	29 columns							







	-		
N 4 11:			
Multi			
variate			
Multi variate Analysis			

Outliers and Anomalies	-								
Data Preproc	es	sing Code	e Scr	een	shots				
-						discharge_disposition_id	admission_source_id	time_in_hospital	num_lab_procedu
	0	Caucasian	Female	[0- 30)	Others	Others	Referral	-1.632460	-0.108
	1	Caucasian	Female	[0- 30)	Emergency	Discharged to home	Emergency Room	-0.271144	0.808
Loading Data	2	AfricanAmerican	Female	[0- 30)	Emergency	Discharged to home	Emergency Room	-0.825587	-1.629
	3	Caucasian	Male	[30- 40)	Emergency	Discharged to home	Emergency Room	-0.825587	0.044
	4	Caucasian	Male	[40- 50)	Emergency	Discharged to home	Emergency Room	-1.632460	0.400
Handling Missing Data	<pre>data['num_medications'].fillna(data['num_medications'].mean(),inplace=True) data['number_outpatient'].fillna(data['number_outpatient'].mean(),inplace=True) data['number_emergency'].fillna(data['number_emergency'].mean(),inplace=True) data['number_diagnoses'].fillna(data['number_diagnoses'].mean(),inplace=True) data['age_derived'].fillna(data['age_derived'].mean(),inplace=True) data['count_Steady'].fillna(data['count_Steady'].mean(),inplace=True) data['count_Down'].fillna(data['count_Down'].mean(),inplace=True) data['count_Up'].fillna(data['count_Up'].mean(),inplace=True) data['number_inpatient'].fillna(data['number_inpatient'].mean(),inplace=True)</pre>								

	: le=LabelEncoder()
	data['gender']=le.fit_transform(data['gender'])
Data Transformatio	data['age']=le.fit_transform(data['age'])
n	data['admission_type_id']=le.fit_transform(data['admission_type_id'])
	data['discharge_disposition_id']=le.fit_transform(data['discharge_disposition_id'])
	data['admission_source_id']=le.fit_transform(data['admission_source_id'])
	data['diag_1']=le.fit_transform(data['diag_1'])
	data['diag_2']=le.fit_transform(data['diag_2'])
	data['diag_3']=le.fit_transform(data['diag_3'])
	data['metformin']=le.fit_transform(data['metformin'])
	data['repaglinide']=le.fit_transform(data['repaglinide'])
	data['glipizide']=le.fit_transform(data['glipizide'])
	data['insulin']=le.fit_transform(data['insulin'])
	data['change']=le.fit_transform(data['change'])
	data['diabetesMed']=le.fit_transform(data['diabetesMed'])
	data['readmitted']=le.fit_transform(data['readmitted'])
Feature Engineering	Attached the codes in final submission