



Data Collection and Preprocessing Phase

Date	25th June 2025
Team ID	LTVIP2025TMID60530
Project Title	Revolutionizing Liver Care: Predicting Liver Cirrhosis Using Advanced Machine Learning Techniques.

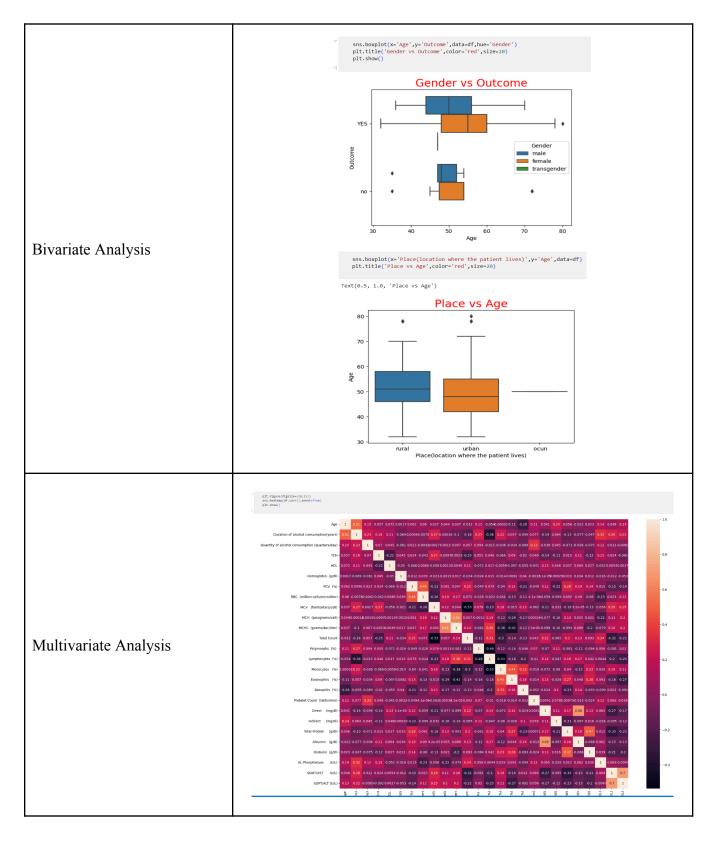
Data Exploration and Preprocessing Template

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
	Dimension: 949 rows × 39 columns Descriptive statistics:
Data Overview	Quantity of S.NO Age Consumption of alcohol alcohol TCH HDL Hemoglobin PCV (%) Selfs/microliter (femtoliters/cell) (%) Gount (mg/dl)
	count 950,000000 950,000000 950,000000 950,000000 591,000000 592,000000 950,0000000 950,0000000 950,000000 950,000000 950,000000 950,000000 950,000000 950,0000000 950,000000 95
	mean 475.500000 50.632632 20.606316 5.159947 197.544839 35.486254 10.263979 33.810000 3.390704 87.651435 0.498557 475.130042 4.040737 2.457542
	std 274.385677 8.808272 7.980664 22.908785 26.694968 7.982057 1.942300 5.751592 0.937089 13.844181 0.712546 6515.406159 2.757443 1.093691
	min 1,000000 32,000000 4,000000 1,000000 10,000000 25,000000 4,000000 1,000000 10,000000 60,000000 0,000000 0,520000 0,800000 0,200000
	25% 238.25000 44.00000 15.00000 2.00000 180.00000 9.00000 2.00000 78.00000 0.00000 1.20000 2.70000 2.00000
	50% 475,50000 50,000000 20,000000 2,000000 194,000000 35,000000 10,000000 35,000000 35,00000 87,000000 0,000000 1,420000 3,700000 2,300000
	75% 712.75000 57.00000 26.00000 3.00000 210.00000 38.00000 11.50000 38.00000 4.00000 94.00000 1,00000 1.70000 42.0000 3.00000 1,00000 1.70000
	max 950,000000 80,000000 45,000000 180,000000 296,000000 81,000000 15,900000 48,000000 5,700000 126,000000 4,00000 9000,000000 25,000000 66,00000
	sns.countplot(data-df,x='Place(location where the patient lives)') plt.title("location",color='y',size=20,loc='left') Location 600
	500 -
Univariate Analysis	400 -
Ullivariate Alialysis	u 30 -
	# 300 -
	20 -
	200 -
	10 -
	100-
	nural urban ocun
	Place(location where the patient lives)

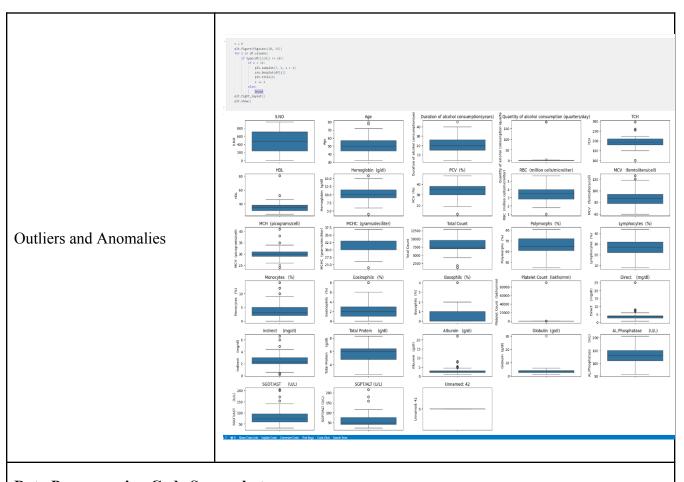












Data Preprocessing Code Screenshots

Loading Data

	# Loading the Dataset of = pd.read_excel('C:\SI project\Codes\Data\HealthCareData.xlsx') of.head()																		
	S.NO	Age	Gender	Place(location where the patient lives)	Duration of alcohol consumption(years)	Quantity of alcohol consumption (quarters/day)	Type of alcohol consumed	Hepatitis B infection	· c	Diabetes Result	Blood pressure (mmhg)	Obesity	Family history of cirrhosis/ hereditary	тсн	TG	LDL	HDL	Hemoglobin (g/dl)	PCV (%)
0	1	55	male	rural	12	2	branded liquor	negative	negative	YES	138/90	yes	no	205.0	115	120	35.0	12.0	40.0
1	2	55	male	rural	12	2	branded liquor	negative	negative	YES	138/90	yes	no	205.0	115	120	35.0	9.2	40.0
2	3	55	male	rural	12	2	branded liquor	negative	negative	YES	138/90	no	no	205.0	115	120	35.0	10.2	40.0
3	4	55	male	rural	12	2	branded liquor	negative	negative	NO	138/90	no	no	NaN	NaN	NaN	NaN	7.2	40.0
4	5	55	female	rural	12	2	branded liquor	negative	negative	YES	138/90	no	no	205.0	115	120	35.0	10.2	40.0





```
df('Tct')-df('Tct').fillna(df('Tct').mean())
df('Yct)-df('Not').fillna(df('Yct)'.mean())
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Handling Missing Data
                                                                                                                                                                                          df['A/G Ratio']=df['A/G Ratio'].fillna(df['A/G Ratio'].mode()[0])
                                                                                                                                                                                                                                                         from sklearn.preprocessing import StandardScaler
                                                                                                                                                                                                                                                         sc = StandardScaler()
                                                                                                                                                                                                                                                       x_train = sc.fit_transform(x_train)
#x_test = sc.transform(x_test)
                                                                                                                                                                                                                                                         x_train
                                                                                                                                                                                                                                          array([[ 2.44060333, -1.84159498, 1.29329571, ..., 1.08599342,
                                                                                                                                                                                                                                                                     4.92950302, 6.81450659],

[ 0.15458485, 0.50365769, 1.29329571, ..., -0.83331467,

-0.20286021, -0.14674577],
                                                                                                                                                                                                                                                                     [-1.44562809, 0.50365769, 1.29329571, ..., 0.49543709, -0.20286021, -0.14674577],
                                                                                                                                                                                                                                                                     ...,
[ 0.72608947, 0.50365769, -0.76458992, ..., 0.27397846,
    -0.20286021, -0.14674577],
[ 0.49748762, -1.84159498, -0.76458992, ..., 2.61774893,
    -0.20286021, -0.14674577],
[ 0.15458485, 0.50365769, -0.76458992, ..., 0.20015892,
    -0.20286021, -0.14674577]])
Data Transformation
                                                                                                                                                                                                                                                                            from sklearn.preprocessing import LabelEncoder
                                                                                                                                                                                                                                                                           le = LabelEncoder()
                                                                                                                                                                                                                                                                            for column in df.columns:
                                                                                                                                                                                                                                                                                            # Check if the column has categorical data
                                                                                                                                                                                                                                                                                            if df[column].dtype == 'object':
    # Perform label encoding
                                                                                                                                                                                                                                                                                                           df[column] = le.fit_transform(df[column])
```





```
categorical features = df.select dtypes(include=[np.object])
                                           categorical features.columns
                                        Index(['Gender', 'Place(location where the patient lives)',
                                                'Type of alcohol consumed', 'Hepatitis B infection',
                                               'Hepatitis C infection', 'Diabetes Result', 'Blood pressure (mmhg)',
                                               'Obesity', 'Family history of cirrhosis/ hereditary', 'TG', 'LDL',
                                               'Total Bilirubin
                                                                  (mg/dl)', 'A/G Ratio',
                                               'USG Abdomen (diffuse liver or not)', 'Outcome'],
                                              dtype='object')
Feature Engineering
                                           numeric_features = df.select_dtypes(include=[np.number])
                                           numeric_features.columns
                                        Index(['S.NO', 'Age', 'Duration of alcohol consumption(years)',
                                                'Quantity of alcohol consumption (quarters/day)', 'TCH', 'HDL',
                                               'Hemoglobin (g/dl)', 'PCV (%)', 'RBC (million cells/microliter)',
                                               'MCV (femtoliters/cell)', 'MCH (picograms/cell)',
                                               'MCHC (grams/deciliter)', 'Total Count', 'Polymorphs (%)',
                                               'Lymphocytes (%)', 'Monocytes (%)', 'Eosinophils (%)',
                                               'Basophils (%)', 'Platelet Count (lakhs/mm)', 'Direct (mg/dl)',
                                                              (mg/dl)', 'Total Protein (g/dl)', 'Albumin (g/dl)',
                                               'Globulin (g/dl)', 'AL.Phosphatase
                                                                                          (U/L)', 'SGOT/AST
                                                                                                                   (U/L)',
                                               'SGPT/ALT (U/L)'],
                                              dtype='object')
                                           # Save the cleaned and processed DataFrame to a CSV file
                                           df.to_csv('cleaned_data.csv', index=False)
                                           df.head()
                                       ✓ 0.0s
                                                                                    Quantity of
                                                       Place(location
                                                                                               Type of
                                                                                                               Blood
                                                                  Duration of alcohol
                                                                                       alcohol
                                                                                                      Diabetes
                                                                                                              pressure Obesity
                                            Age Gender
                                                          where the
                                                                                               alcohol
                                                                  consumption(years)
                                                                                   consumption
                                                                                                        Result
                                                                                                              (mmhg)
                                                        patient lives)
                                                                                             consumed
Save Processed Data
                                                                                  (quarters/day)
                                         0 55.0
                                                                             12.0
                                                                                                                  32
                                                                                          2.0
                                                                             12.0
                                                                                          2.0
                                                                                                   2
                                                                                                                  32
                                         1 55.0
                                                                                                                  32
                                                                                                                          0
                                            55.0
                                                                             12.0
                                                                                          2.0
                                                                                          2.0
                                                                                                   2
                                                                                                                  32
                                                                                                                         0
                                         3 55.0
                                                                             12.0
                                                                                                           0
                                          4 55.0
                                                                                                                          0
```