Data Collection and Preprocessing Phase

Date	4 June 2024
Team ID	SWTID1720076203
Project Title	Anemia Sense: Leveraging Machine Learning for Precise Anemia Recognitions
Maximum Marks	6 Marks

Data Exploration and Preprocessing

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

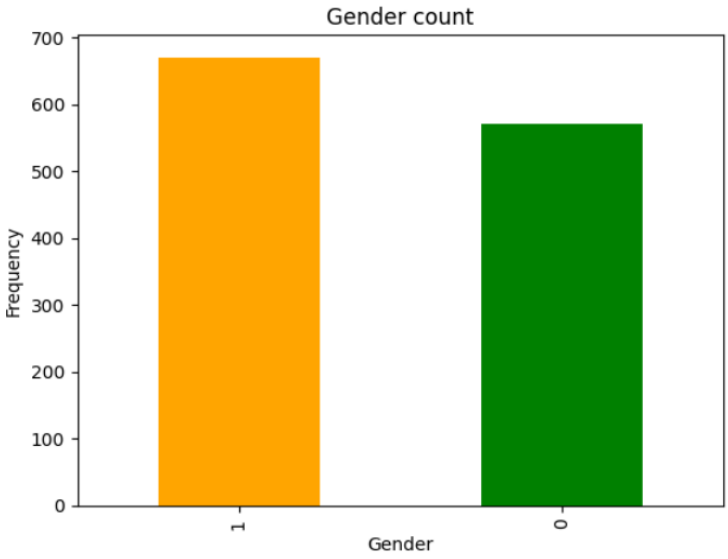
Section	Description
Data Overview	Dimension: 1421 rows x 6 columns

```
#Descriptive statistical

df.describe()
```

	Gender	Hemoglobin	MCH	MCHC	MCV	Result
count	1240.000000	1240.000000	1240.000000	1240.000000	1240.000000	1240.000000
mean	0.540323	13.218145	22.903952	30.277984	85.620968	0.500000
std	0.498573	1.976190	3.993624	1.394515	9.673794	0.500202
min	0.000000	6.600000	16.000000	27.800000	69.400000	0.000000
25%	0.000000	11.500000	19.400000	29.100000	77.300000	0.000000
50%	1.000000	13.000000	22.700000	30.400000	85.300000	0.500000
75%	1.000000	14.900000	26.200000	31.500000	94.225000	1.000000
max	1.000000	16.900000	30.000000	32.500000	101.600000	1.000000

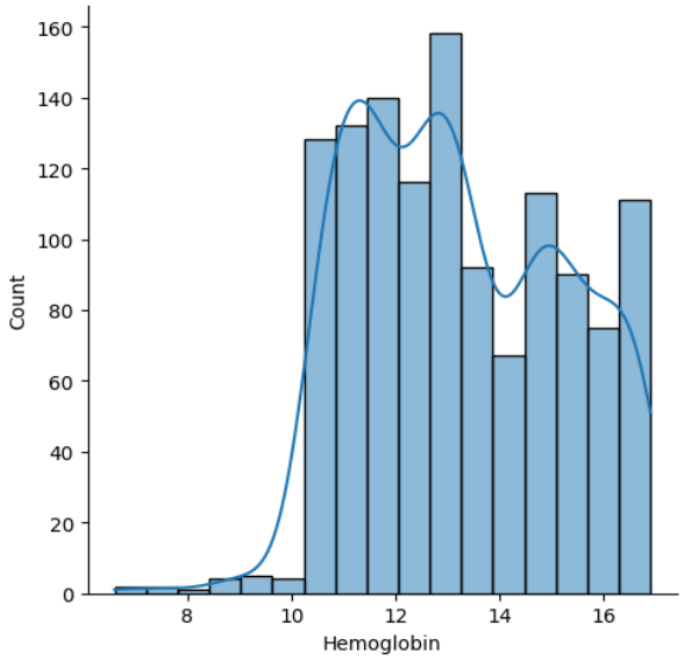
Univariate Analysis

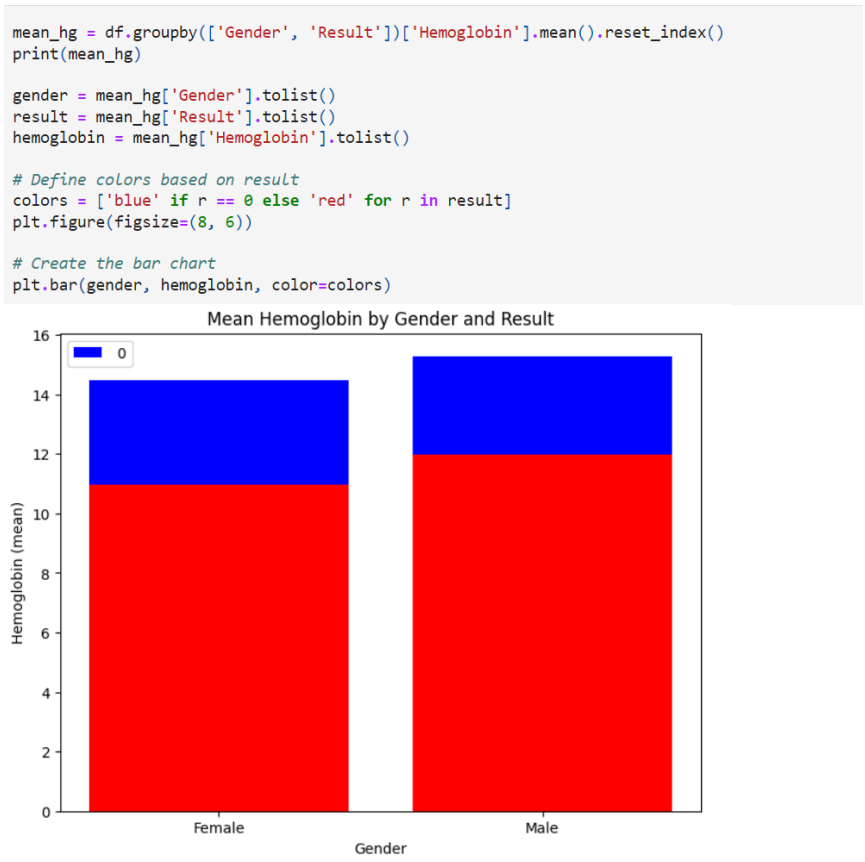
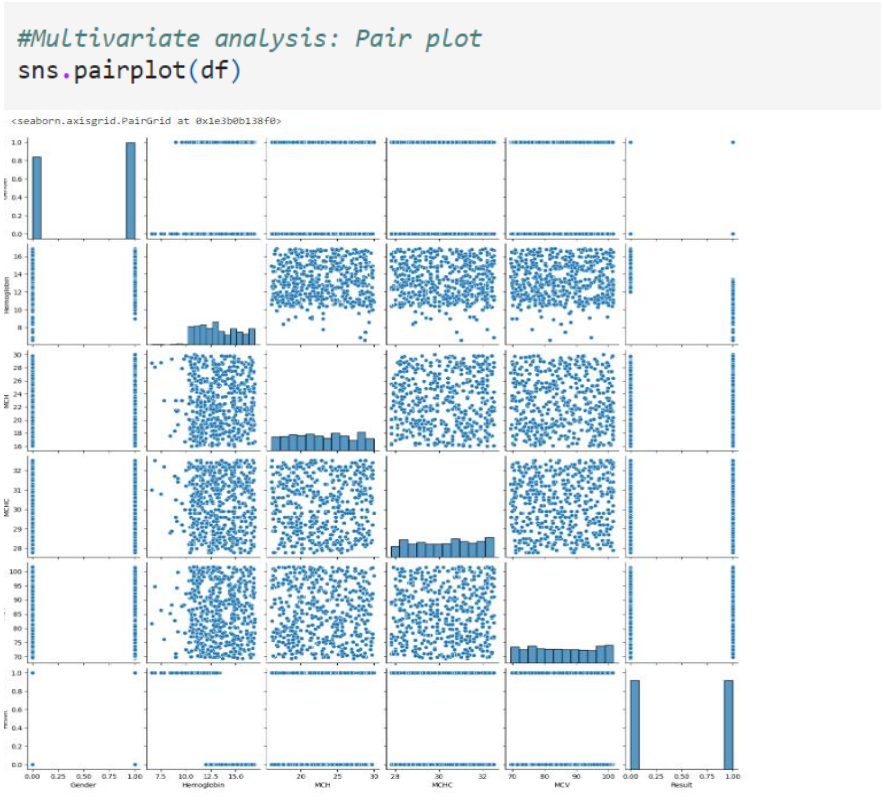
```
#Univariate Analysis: Bar graph

output=df['Gender'].value_counts()
output.plot(kind='bar',color=['orange','green'])
plt.xlabel('Gender')
plt.ylabel('Frequency')
plt.title('Gender count')
plt.show()
```




```
#Univariate analysis: displot

sns.displot(df['Hemoglobin'],kde=True)
```

```
<seaborn.axisgrid.FacetGrid at 0x1e3b0ce2b10>
```



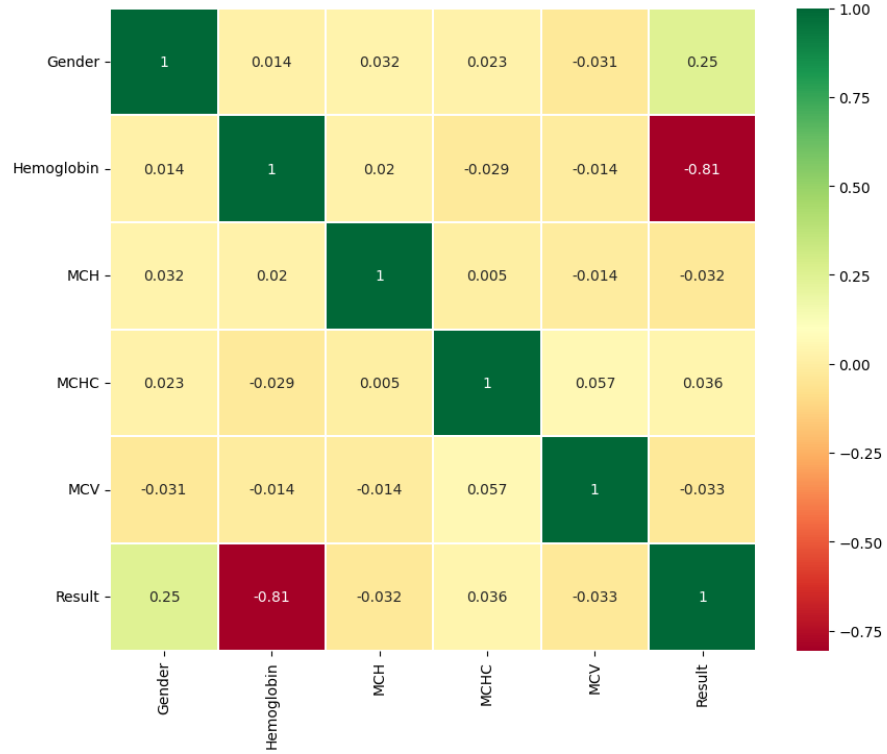
Bivariate Analysis	
Multivariate Analysis	

```
mean_hg = df.groupby(['Gender', 'Result'])['Hemoglobin'].mean().reset_index()
print(mean_hg)

gender = mean_hg['Gender'].tolist()
result = mean_hg['Result'].tolist()
hemoglobin = mean_hg['Hemoglobin'].tolist()

# Define colors based on result
colors = ['blue' if r == 0 else 'red' for r in result]
plt.figure(figsize=(8, 6))

# Create the bar chart
plt.bar(gender, hemoglobin, color=colors)
```



```
#Multivariate analysis: Pair plot
sns.pairplot(df)
```

```
#Multivariate analysis: Heatmap

sns.heatmap(df.corr(),annot=True,cmap="RdYlGn",linewidth=0.2)
fig=plt.gcf()
fig.set_size_inches(10,8)
plt.show()
```



Outliers and Anomalies	-

Data Preprocessing Code Screenshots

Loading Data

```
df = pd.read_csv("data/anemia.csv")
df.head()
```

	Gender	Hemoglobin	MCH	MCHC	MCV	Result
0	1	14.9	22.7	29.1	83.7	0
1	0	15.9	25.4	28.3	72.0	0
2	0	9.0	21.5	29.6	71.2	1
3	0	14.9	16.0	31.4	87.5	0
4	1	14.7	22.0	28.2	99.5	0

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<td>Handling Missing Data</td>
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```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1421 entries, 0 to 1420
Data columns (total 6 columns):
 #   Column      Non-Null Count  Dtype  
---  ------      --------------  -----  
 0   Gender      1421 non-null   int64  
 1   Hemoglobin  1421 non-null   float64
 2   MCH         1421 non-null   float64
 3   MCHC        1421 non-null   float64
 4   MCV         1421 non-null   float64
 5   Result      1421 non-null   int64  
dtypes: float64(4), int64(2)
memory usage: 66.7 KB
```

```
df.isnull().sum()
```

```
Gender        0
Hemoglobin    0
MCH           0
MCHC          0
MCV           0
Result        0
dtype: int64
```

There are no missing values in the dataset

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<td>Data Transformation</td>
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```
# female count is observed to be more than male so we balance it using undesampling
from sklearn.utils import resample
majorclass = df[df['Result'] == 0]
minorclass = df[df['Result'] == 1]
major_downsample = resample(majorclass, replace=False, n_samples=len(minorclass),random_state=42)
df = pd.concat([major_downsample,minorclass])
df['Result'].value_counts()
```

```
Result
0    620
1    620
Name: count, dtype: int64
```

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<td>Feature Engineering</td>
<td>Attached the codes in final submission.</td>
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<td>Save Processed Data</td>
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