

# Classification: Persistent vs Non-Persistent

```
In [1]: ##Import Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report

import warnings
warnings.filterwarnings("ignore")
```

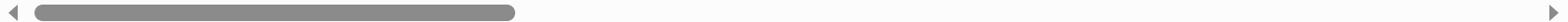
```
In [2]: #Load data
data=pd.read_csv(r"C:\Users\DD\Desktop\Persistent_vs_NonPersistent\Persistent_vs_NonPersistent.csv")
```

In [3]: data

Out[3]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ni
0	P1	Persistent	Male	Caucasian	Not Hispanic	West	>75	GENERAL PRACTITIONER	Others	OB/GYN
1	P2	Non-Persistent	Male	Asian	Not Hispanic	West	55-65	GENERAL PRACTITIONER	Others	OB/GYN
2	P3	Non-Persistent	Female	Other/Unknown	Hispanic	Midwest	65-75	GENERAL PRACTITIONER	Others	OB/GYN
3	P4	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
4	P5	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
...	...	...	...	...	...	...	...	...	...	...
3419	P3420	Persistent	Female	Caucasian	Not Hispanic	South	>75	GENERAL PRACTITIONER	Others	OB/GYN
3420	P3421	Persistent	Female	Caucasian	Not Hispanic	South	>75	Unknown	Others	OB/GYN
3421	P3422	Persistent	Female	Caucasian	Not Hispanic	South	>75	ENDOCRINOLOGY	Specialist	
3422	P3423	Non-Persistent	Female	Caucasian	Not Hispanic	South	55-65	Unknown	Others	OB/GYN
3423	P3424	Non-Persistent	Female	Caucasian	Not Hispanic	South	65-75	Unknown	Others	OB/GYN

3424 rows × 69 columns



In [4]: data.head()

Out[4]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ntm_Spec
0	P1	Persistent	Male	Caucasian	Not Hispanic	West	>75	GENERAL PRACTITIONER	Others	OB/GYN/Others/F
1	P2	Non-Persistent	Male	Asian	Not Hispanic	West	55-65	GENERAL PRACTITIONER	Others	OB/GYN/Others/F
2	P3	Non-Persistent	Female	Other/Unknown	Hispanic	Midwest	65-75	GENERAL PRACTITIONER	Others	OB/GYN/Others/F
3	P4	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN/Others/F
4	P5	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN/Others/F

5 rows × 69 columns

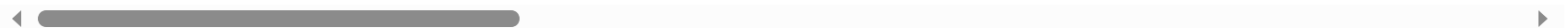


In [5]: data.tail(3)

Out[5]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ntm_Sp
3421	P3422	Persistent	Female	Caucasian	Not Hispanic	South	>75	ENDOCRINOLOGY	Specialist	
3422	P3423	Non-Persistent	Female	Caucasian	Not Hispanic	South	55-65	Unknown	Others	OB/GYN/Othe
3423	P3424	Non-Persistent	Female	Caucasian	Not Hispanic	South	65-75	Unknown	Others	OB/GYN/Othe

3 rows × 69 columns



In [6]: `data.info()`

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 3424 entries, 0 to 3423
```

```
Data columns (total 69 columns):
```

#	Column	Non-Null Count	Dtype
0	Ptid	3424 non-null	object
1	Persistence_Flag	3424 non-null	object
2	Gender	3424 non-null	object
3	Race	3424 non-null	object
4	Ethnicity	3424 non-null	object
5	Region	3424 non-null	object
6	Age_Bucket	3424 non-null	object
7	Ntm_Speciality	3424 non-null	object
8	Ntm_Specialist_Flag	3424 non-null	object
9	Ntm_Speciality_Bucket	3424 non-null	object
10	Gluko_Record_Prior_Ntm	3424 non-null	object
11	Gluko_Record_During_Rx	3424 non-null	object
12	Dexa_Freq_During_Rx	3309 non-null	float64
13	Dexa_During_Rx	3424 non-null	object
14	Frag_Frac_Prior_Ntm	3424 non-null	object
15	Frag_Frac_During_Rx	3424 non-null	object
16	Risk_Segment_Prior_Ntm	3424 non-null	object
17	Tscore_Bucket_Prior_Ntm	3424 non-null	object
18	Risk_Segment_During_Rx	3424 non-null	object
19	Tscore_Bucket_During_Rx	3424 non-null	object
20	Change_T_Score	3424 non-null	object
21	Change_Risk_Segment	3424 non-null	object
22	Adherent_Flag	3424 non-null	object
23	Idn_Indicator	3424 non-null	object
24	Injectable_Experience_During_Rx	3424 non-null	object
25	Comorb_Encounter_For_Screening_For_Malignant_Neoplasms	3424 non-null	object
26	Comorb_Encounter_For_Immunization	3424 non-null	object
27	Comorb_Encntr_For_General_Exam_W_O_Complaint,_Susp_Or_Reprtd_Dx	3424 non-null	object
28	Comorb_Vitamin_D_Deficiency	3424 non-null	object
29	Comorb_Other_Joint_Disorder_Not_Elsewhere_Classified	3424 non-null	object
30	Comorb_Encntr_For_Oth_Sp_Exam_W_O_Complaint_Suspected_Or_Reprtd_Dx	3424 non-null	object
31	Comorb_Long_Term_Current_Drug_Therapy	3424 non-null	object
32	Comorb_Dorsalgia	3424 non-null	object
33	Comorb_Personal_History_Of_Other_Diseases_And_Conditions	3424 non-null	object
34	Comorb_Other_Disorders_Of_Bone_Density_And_Structure	3424 non-null	object
35	Comorb_Disorders_of_lipoprotein_metabolism_and_other_lipidemias	3424 non-null	object
36	Comorb_Osteoporosis_without_current_pathological_fracture	3424 non-null	object
37	Comorb_Personal_history_of_malignant_neoplasm	3424 non-null	object

38	Comorb_Gastro_esophageal_reflux_disease	3424	non-null	object
39	Concom_Cholesterol_And_Triglyceride_Regulating_Preparations	3424	non-null	object
40	Concom_Narcotics	3424	non-null	object
41	Concom_Systemic_Corticosteroids_Plain	3424	non-null	object
42	Concom_Anti_Depressants_And_Mood_Stabilisers	3424	non-null	object
43	Concom_Fluoroquinolones	3424	non-null	object
44	Concom_Cephalosporins	3424	non-null	object
45	Concom_Macrolides_And_Similar_Types	3424	non-null	object
46	Concom_Broad_Spectrum_Penicillins	3424	non-null	object
47	Concom_Anaesthetics_General	3424	non-null	object
48	Concom_Viral_Vaccines	3424	non-null	object
49	Risk_Type_1_Insulin_Dependent_Diabetes	3424	non-null	object
50	Risk_Osteogenesis_Imperfecta	3424	non-null	object
51	Risk_Rheumatoid_Arthritis	3276	non-null	object
52	Risk_Untreated_Chronic_Hyperthyroidism	3424	non-null	object
53	Risk_Untreated_Chronic_Hypogonadism	3424	non-null	object
54	Risk_Untreated_Early_Menopause	3424	non-null	object
55	Risk_Patient_Parent_Fractured_Their_Hip	3406	non-null	object
56	Risk_Smoking_Tobacco	3424	non-null	object
57	Risk_Chronic_Malnutrition_Or_Malabsorption	3424	non-null	object
58	Risk_Chronic_Liver_Disease	3424	non-null	object
59	Risk_Family_History_Of_Osteoporosis	3424	non-null	object
60	Risk_Low_Calcium_Intake	3424	non-null	object
61	Risk_Vitamin_D_Insufficiency	3424	non-null	object
62	Risk_Poor_Health_Frailty	3424	non-null	object
63	Risk_Excessive_Thinness	3424	non-null	object
64	Risk_Hysterectomy_Oophorectomy	3424	non-null	object
65	Risk_Estrogen_Deficiency	3424	non-null	object
66	Risk_Immobilization	3424	non-null	object
67	Risk_Recurring_Falls	3424	non-null	object
68	Count_Of_Risks	3424	non-null	int64

dtypes: float64(1), int64(1), object(67)

memory usage: 1.8+ MB

In [7]: `data.info`

```

Out[7]: <bound method DataFrame.info of
\
0      P1      Persistent      Male      Caucasian      Not Hispanic      West
1      P2      Non-Persistent      Male      Asian      Not Hispanic      West
2      P3      Non-Persistent      Female      Other/Unknown      Hispanic      Midwest
3      P4      Non-Persistent      Female      Caucasian      Not Hispanic      Midwest
4      P5      Non-Persistent      Female      Caucasian      Not Hispanic      Midwest
...      ...      ...      ...      ...      ...      ...
3419  P3420      Persistent      Female      Caucasian      Not Hispanic      South
3420  P3421      Persistent      Female      Caucasian      Not Hispanic      South
3421  P3422      Persistent      Female      Caucasian      Not Hispanic      South
3422  P3423      Non-Persistent      Female      Caucasian      Not Hispanic      South
3423  P3424      Non-Persistent      Female      Caucasian      Not Hispanic      South

      Age_Bucket      Ntm_Speciality      Ntm_Specialist_Flag      \
0      >75      GENERAL PRACTITIONER      Others
1      55-65      GENERAL PRACTITIONER      Others
2      65-75      GENERAL PRACTITIONER      Others
3      >75      GENERAL PRACTITIONER      Others
4      >75      GENERAL PRACTITIONER      Others
...      ...      ...      ...
3419      >75      GENERAL PRACTITIONER      Others
3420      >75      Unknown      Others
3421      >75      ENDOCRINOLOGY      Specialist
3422      55-65      Unknown      Others
3423      65-75      Unknown      Others

      Ntm_Speciality_Bucket      ...      Risk_Family_History_Of_Osteoporosis      \
0      OB/GYN/Others/PCP/Unknown      ...      N
1      OB/GYN/Others/PCP/Unknown      ...      N
2      OB/GYN/Others/PCP/Unknown      ...      N
3      OB/GYN/Others/PCP/Unknown      ...      N
4      OB/GYN/Others/PCP/Unknown      ...      N
...      ...      ...      ...
3419      OB/GYN/Others/PCP/Unknown      ...      N
3420      OB/GYN/Others/PCP/Unknown      ...      N
3421      Endo/Onc/Uro      ...      N
3422      OB/GYN/Others/PCP/Unknown      ...      N
3423      OB/GYN/Others/PCP/Unknown      ...      N

      Risk_Low_Calcium_Intake      Risk_Vitamin_D_Insufficiency      \
0      N      N
1      N      N

```



2	Y	N
3	N	N
4	N	N
...	...	...
3419	N	Y
3420	N	N
3421	N	Y
3422	N	N
3423	N	Y

	Risk_Poor_Health_Frailty	Risk_Excessive_Thinness	\
0	N	N	
1	N	N	
2	N	N	
3	N	N	
4	N	N	
...	...	...	
3419	N	N	
3420	N	N	
3421	N	N	
3422	N	N	
3423	N	N	

	Risk_Hysterectomy_Oophorectomy	Risk_Estrogen_Deficiency	\
0	N	N	
1	N	N	
2	N	N	
3	N	N	
4	N	N	
...	...	...	
3419	N	N	
3420	N	N	
3421	N	N	
3422	N	N	
3423	N	N	

	Risk_Immobilization	Risk_Recurring_Falls	Count_Of_Risks
0	N	N	0
1	N	N	0
2	N	N	2
3	N	N	1
4	N	N	1
...	...	...	...

3419	N	N	1
3420	N	N	0
3421	N	N	1
3422	N	N	0
3423	N	N	1

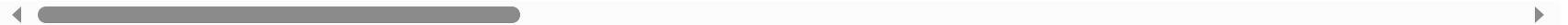
[3424 rows x 69 columns]&gt;

In [8]: data.describe(include="all")

Out[8]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ntm_Spe
<b>count</b>	3424	3424	3424	3424	3424	3424	3424	3424	3424	
<b>unique</b>	3424	2	2	4	3	5	4	36	2	
<b>top</b>	P1	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN/Others
<b>freq</b>	1	2135	3230	3148	3235	1383	1439	1535	2013	
<b>mean</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
<b>std</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
<b>min</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
<b>25%</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
<b>50%</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
<b>75%</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
<b>max</b>	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	

11 rows × 69 columns



```
In [9]: data.isnull().sum()
```

```
Out[9]: Ptid          0
        Persistency_Flag  0
        Gender          0
        Race            0
        Ethnicity       0
        ..
        Risk_Hysterectomy_Oophorectomy  0
        Risk_Estrogen_Deficiency        0
        Risk_Immobilization             0
        Risk_Recurring_Falls            0
        Count_Of_Risks                  0
        Length: 69, dtype: int64
```

```
In [10]: #any null value present
        data.isnull().values.any()
```

```
Out[10]: True
```

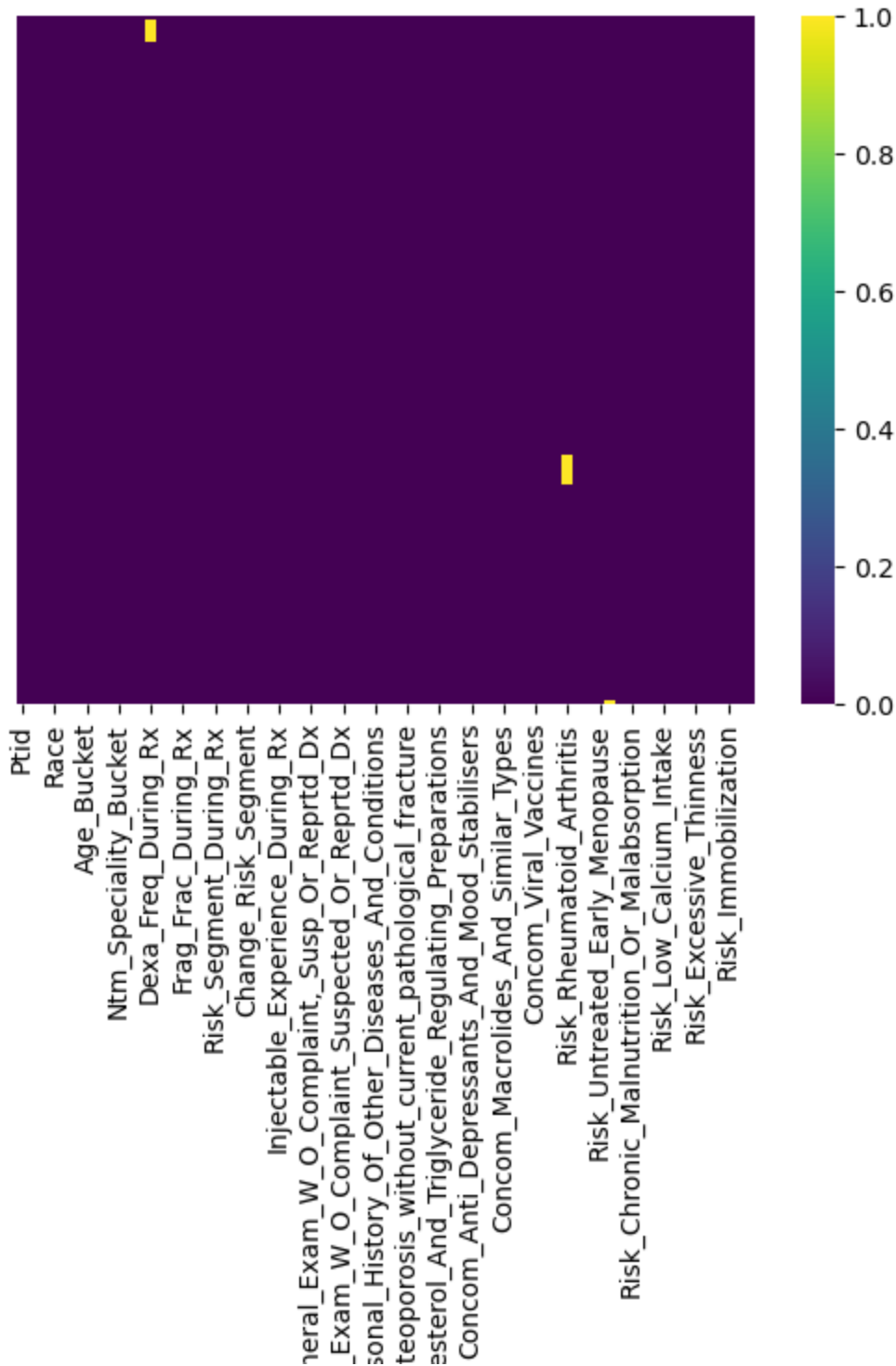
```
In [11]: data.isnull().sum()
```

```
Out[11]: Ptid          0
        Persistency_Flag  0
        Gender          0
        Race            0
        Ethnicity       0
        ..
        Risk_Hysterectomy_Oophorectomy  0
        Risk_Estrogen_Deficiency        0
        Risk_Immobilization             0
        Risk_Recurring_Falls            0
        Count_Of_Risks                  0
        Length: 69, dtype: int64
```

```
In [12]: sns.heatmap(data.isnull(),yticklabels=False,cmap="viridis")
```

```
Out[12]: <Axes: >
```





Comorb\_Encntr\_For\_Gei  
Comorb\_Encntr\_For\_Oth\_Sp  
Comorb\_Per  
Comorb\_Os  
Concom\_Choli

## Filling missing values

```
In [13]: data["Dexa_Freq_During_Rx"].fillna(data["Dexa_Freq_During_Rx"].median(),inplace=True)

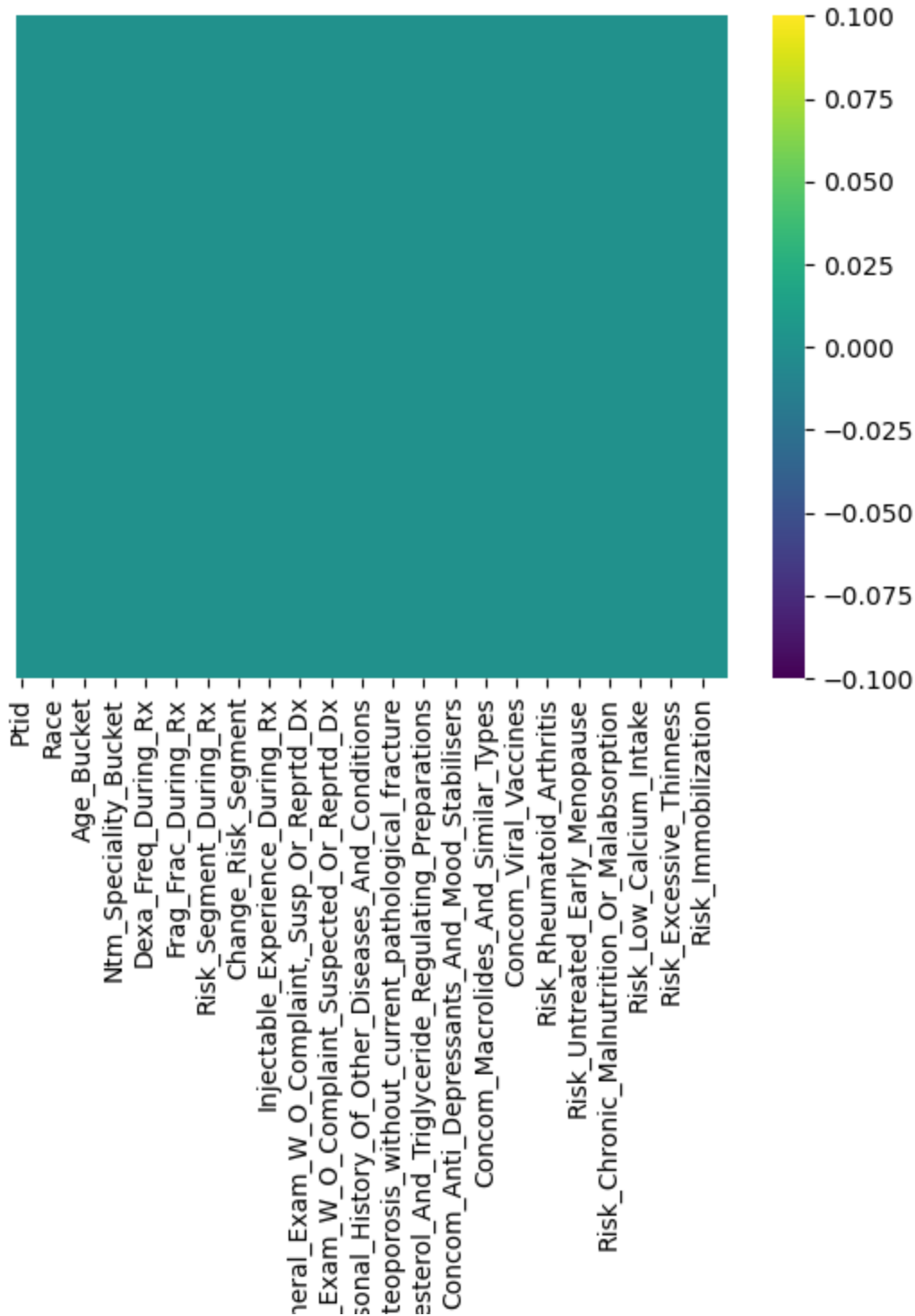
data["Risk_Rheumatoid_Arthritis"].fillna(data["Risk_Rheumatoid_Arthritis"].mode()[0],inplace=True)
data["Risk_Patient_Parent_Fractured_Their_Hip"].fillna(data["Risk_Patient_Parent_Fractured_Their_Hip"].mode()[
```

```
In [14]: sns.heatmap(data.isnull(),yticklabels=False,cmap="viridis")
```

```
Out[14]: <Axes: >
```







Comorb\_Encntr\_For\_Gel  
Comorb\_Encntr\_For\_Oth\_Sp  
Comorb\_Per  
Comorb\_Os  
Concom\_Choli

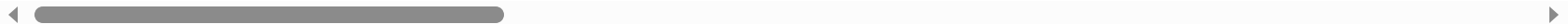
```
In [15]: cat_data=data.select_dtypes(include="object")  
num_data=data.select_dtypes(exclude="object")
```

In [16]: cat\_data

Out[16]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ni
0	P1	Persistent	Male	Caucasian	Not Hispanic	West	>75	GENERAL PRACTITIONER	Others	OB/GYN
1	P2	Non-Persistent	Male	Asian	Not Hispanic	West	55-65	GENERAL PRACTITIONER	Others	OB/GYN
2	P3	Non-Persistent	Female	Other/Unknown	Hispanic	Midwest	65-75	GENERAL PRACTITIONER	Others	OB/GYN
3	P4	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
4	P5	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
...	...	...	...	...	...	...	...	...	...	...
3419	P3420	Persistent	Female	Caucasian	Not Hispanic	South	>75	GENERAL PRACTITIONER	Others	OB/GYN
3420	P3421	Persistent	Female	Caucasian	Not Hispanic	South	>75	Unknown	Others	OB/GYN
3421	P3422	Persistent	Female	Caucasian	Not Hispanic	South	>75	ENDOCRINOLOGY	Specialist	
3422	P3423	Non-Persistent	Female	Caucasian	Not Hispanic	South	55-65	Unknown	Others	OB/GYN
3423	P3424	Non-Persistent	Female	Caucasian	Not Hispanic	South	65-75	Unknown	Others	OB/GYN

3424 rows × 67 columns



```
In [17]: num_data
```

Out[17]:

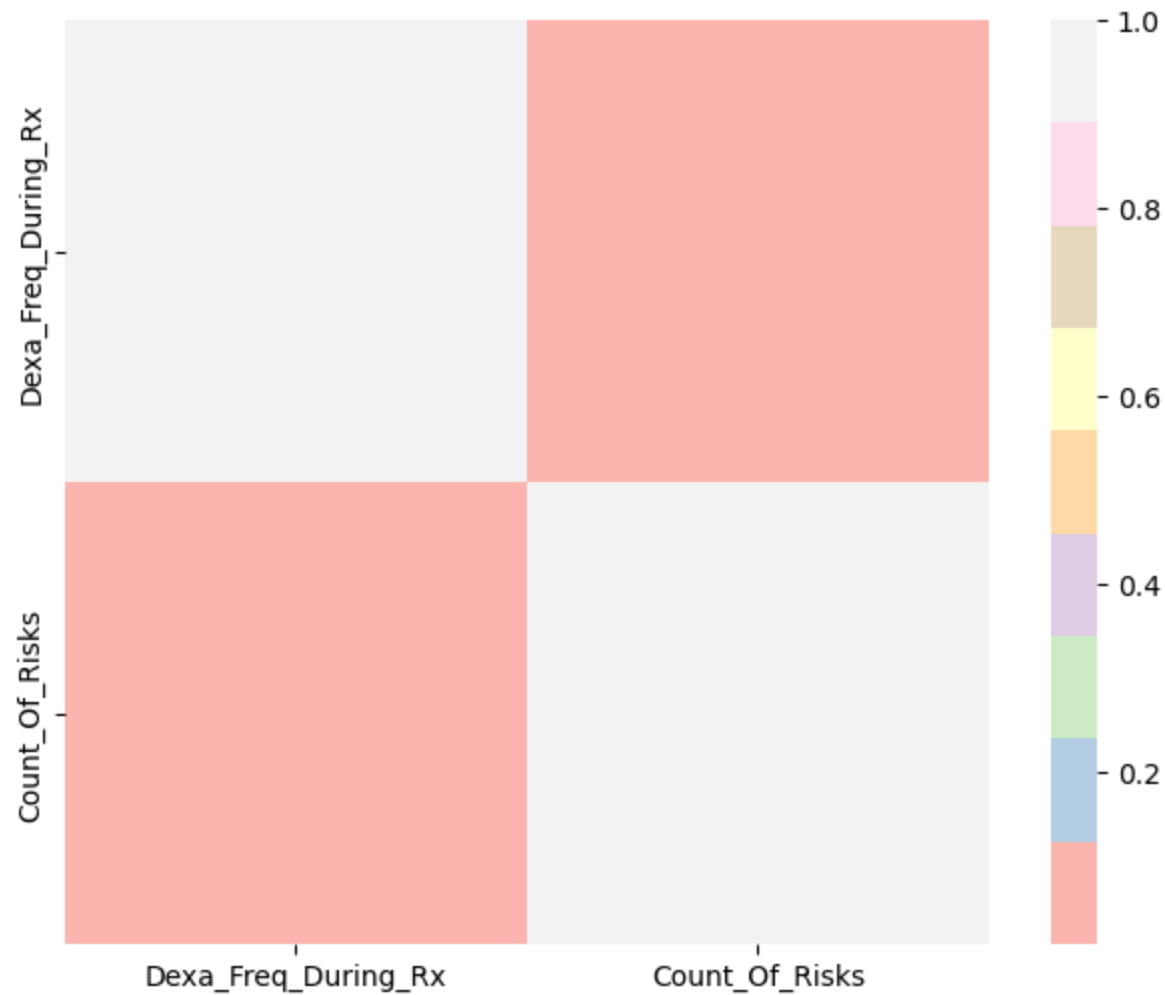
	Dexa_Freq_During_Rx	Count_Of_Risks
0	0.0	0
1	0.0	0
2	0.0	2
3	0.0	1
4	0.0	1
...	...	...
3419	0.0	1
3420	0.0	0
3421	7.0	1
3422	0.0	0
3423	0.0	1

3424 rows × 2 columns

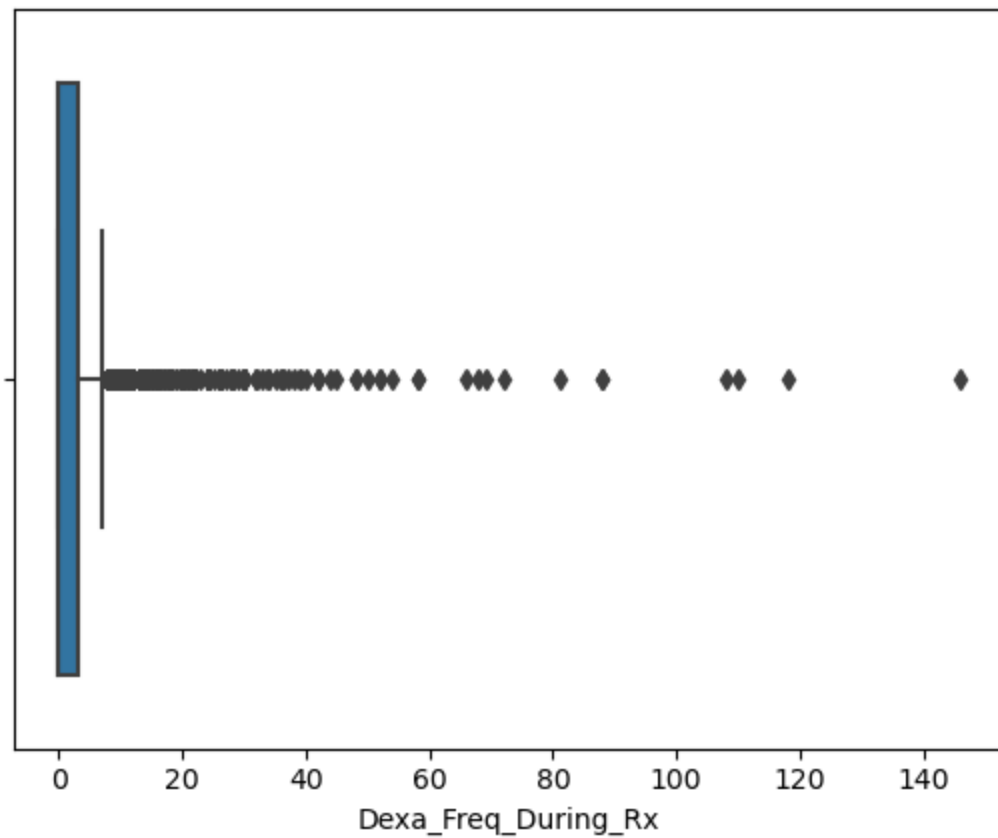
# Data Visualization

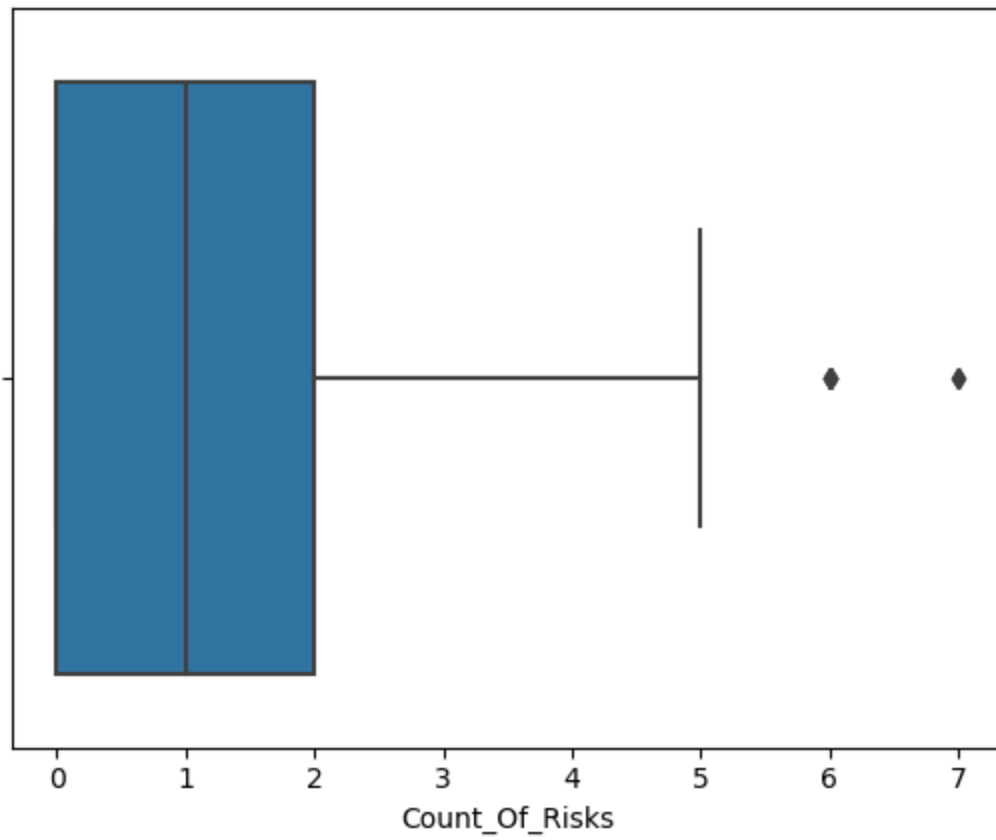
In [18]: *#Examining a correlation matrix of all the features*

```
corrmat=num_data.corr()  
plt.figure(figsize=(8,6))  
sns.heatmap(corrmat,cmap="Pastel1",square=True)  
plt.show()
```



```
In [19]: for i in num_data.columns:  
          sns.boxplot(x=data[i]) # Use data instead of num_data here  
          plt.show()
```

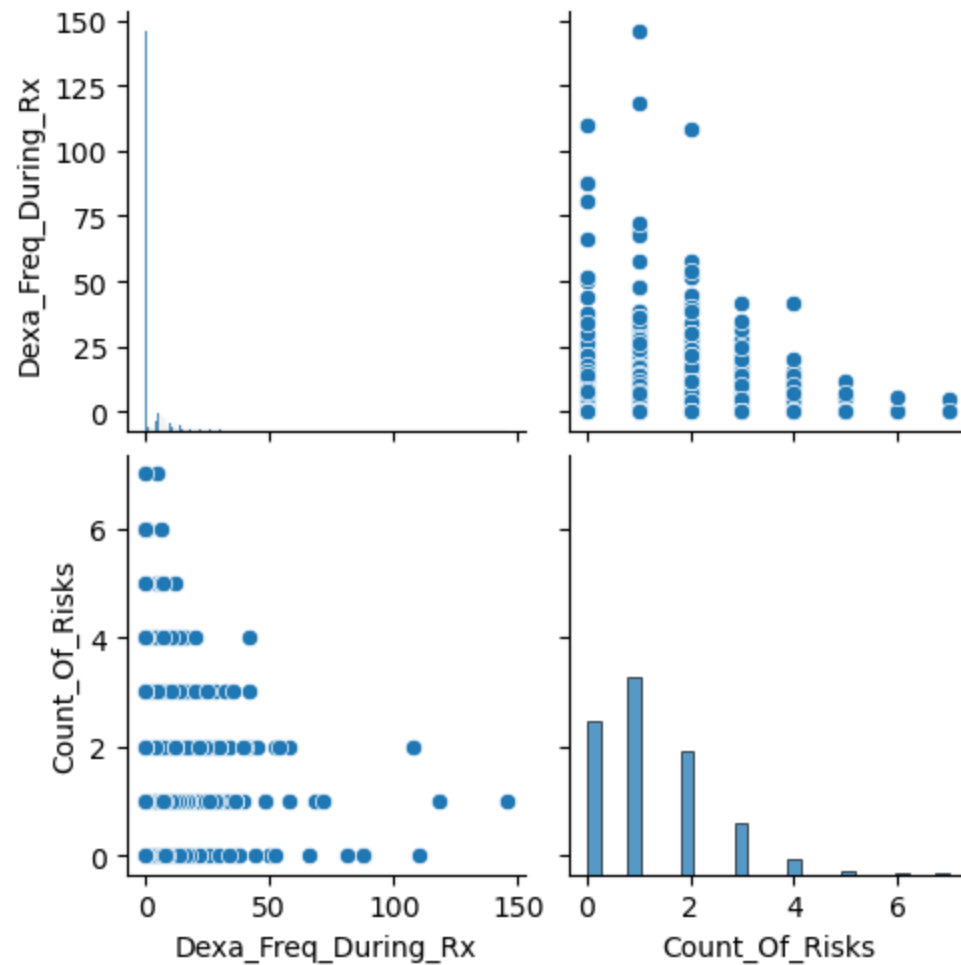






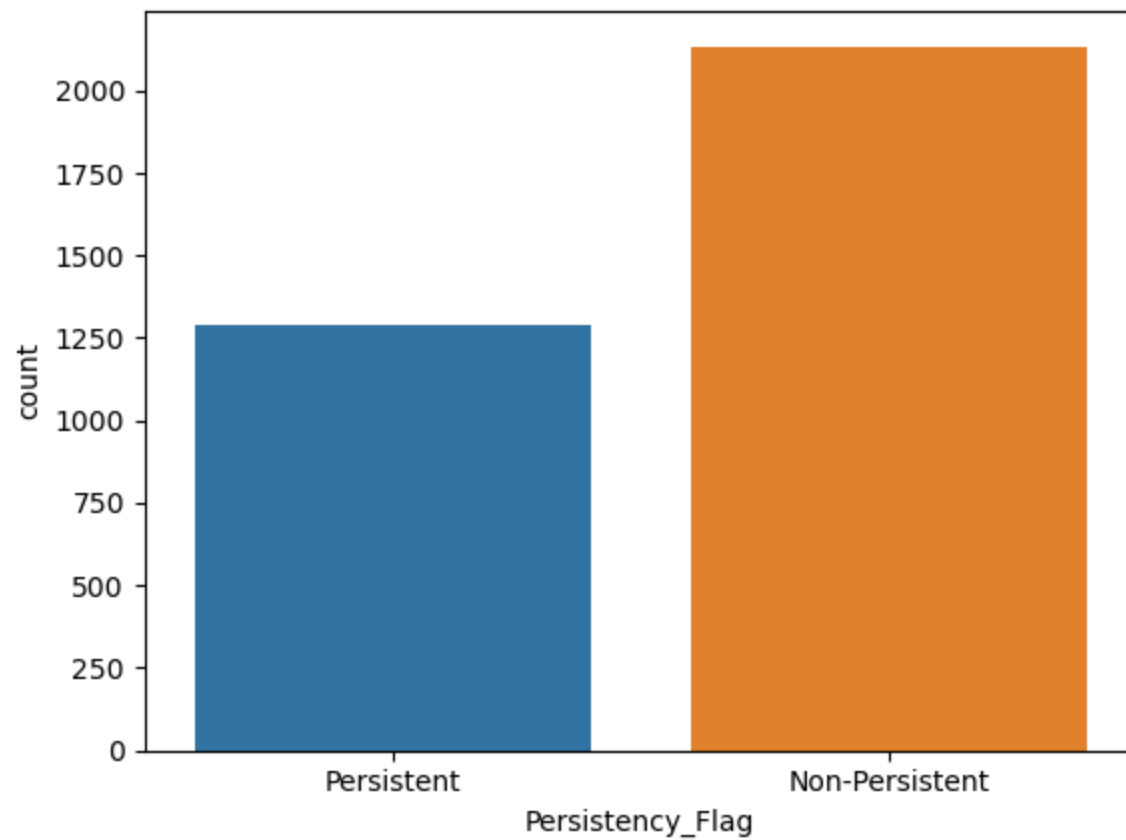
```
In [20]: sns.pairplot(data)
```

```
Out[20]: <seaborn.axisgrid.PairGrid at 0x23b2d065d50>
```

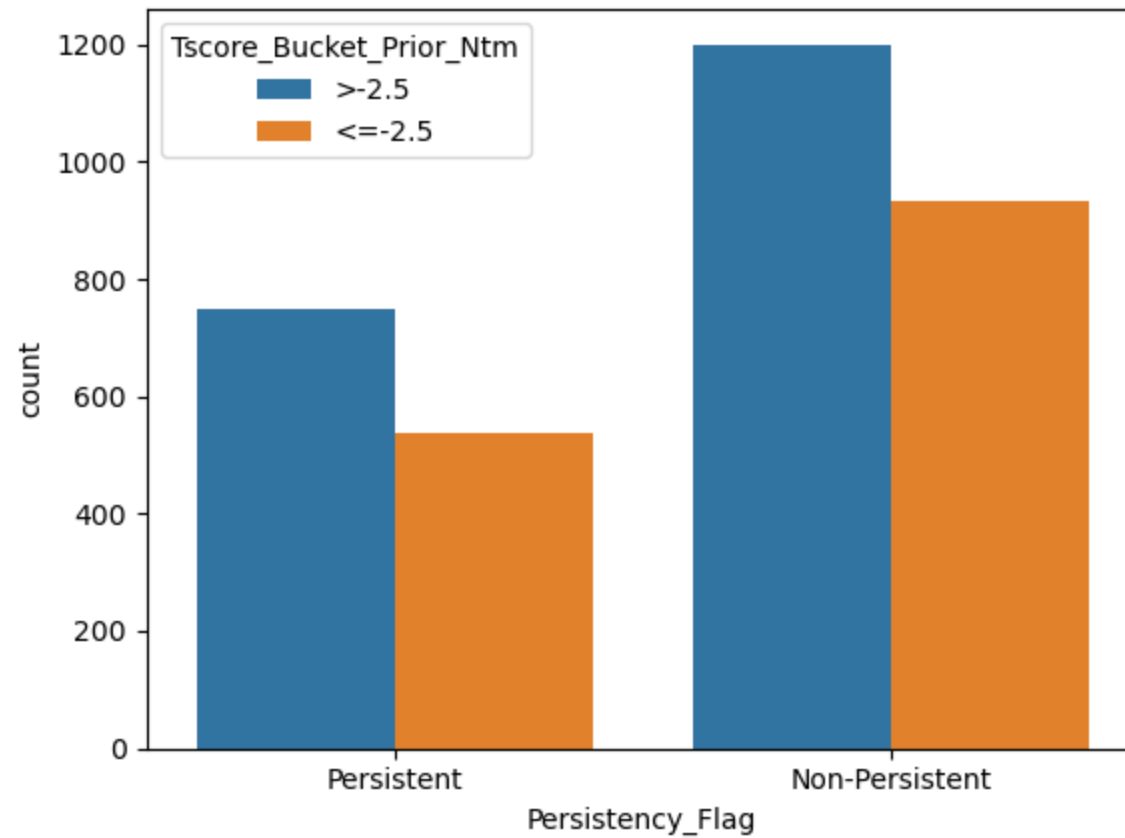


```
In [21]: sns.countplot(x="Persistency_Flag",data=data)
```

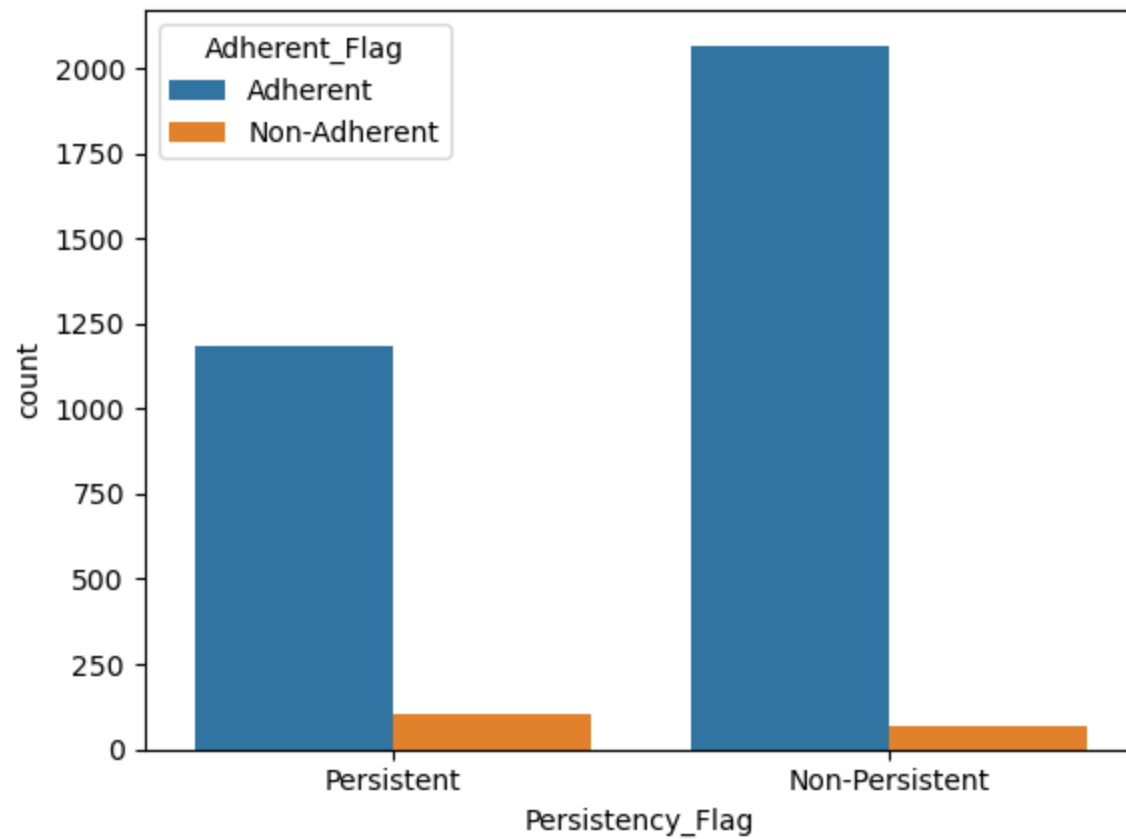
```
Out[21]: <Axes: xlabel='Persistency_Flag', ylabel='count'>
```



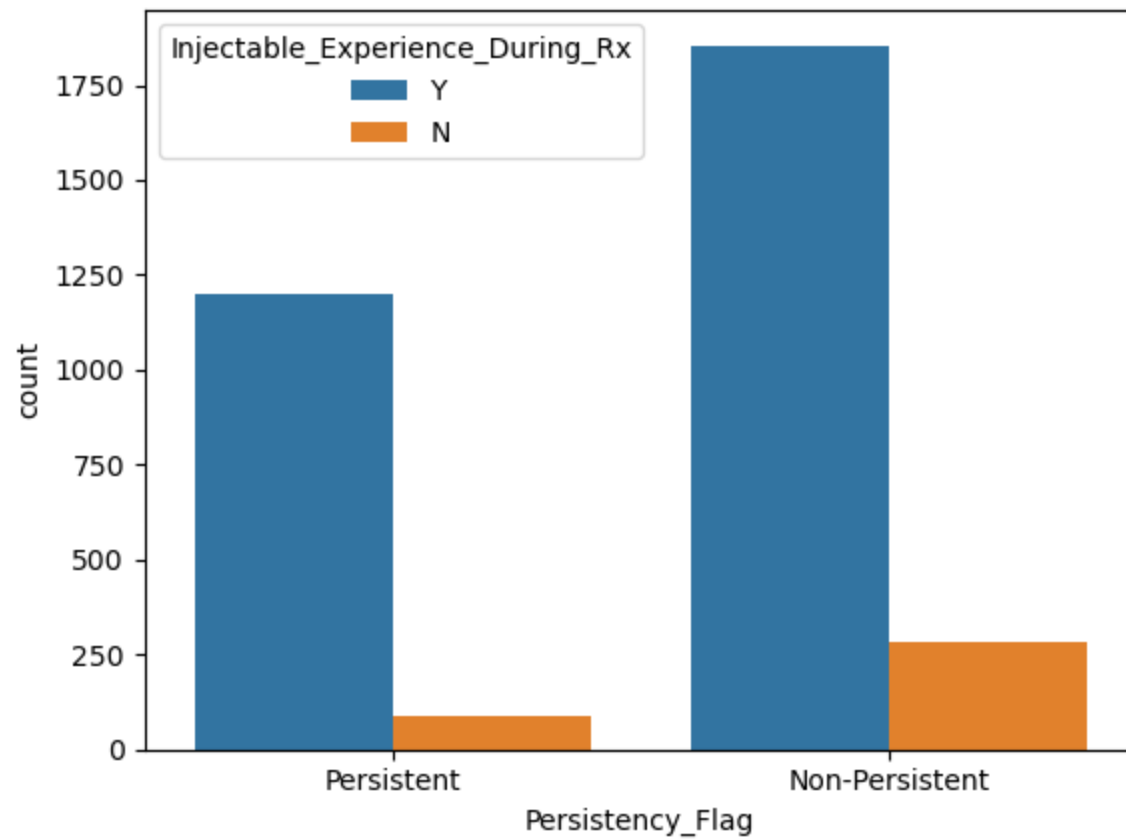
```
In [22]: sns.countplot(x="Persistency_Flag",hue="Tscore_Bucket_Prior_Ntm",data=data)  
plt.show()
```



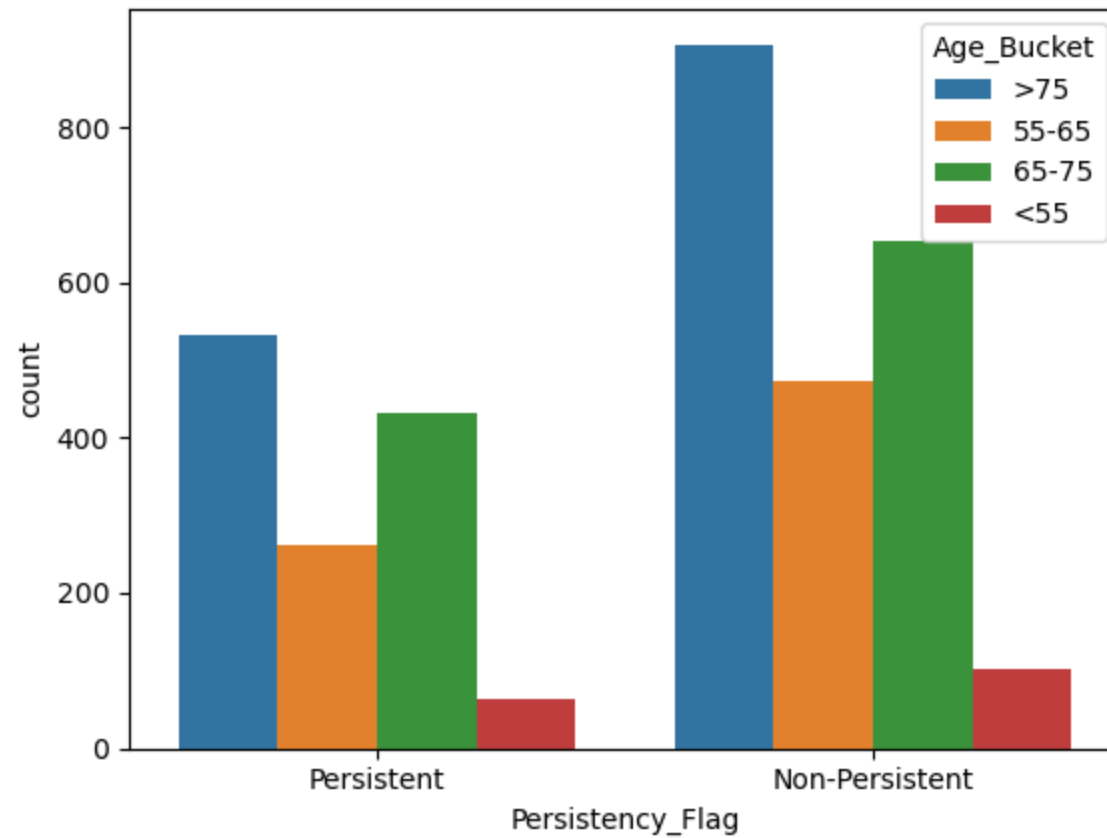
```
In [23]: sns.countplot(x="Persistency_Flag", hue='Adherent_Flag', data=data)  
plt.show()
```



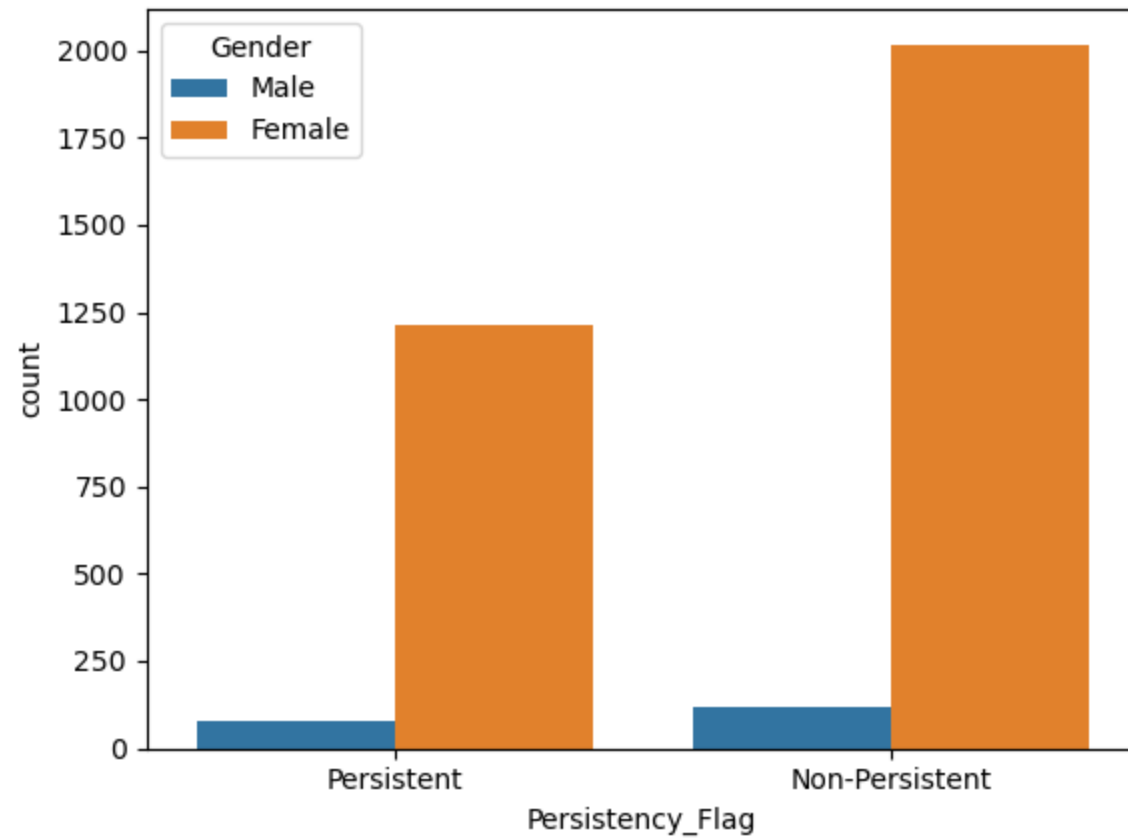
```
In [24]: sns.countplot(x="Persistency_Flag", hue='Injectable_Experience_During_Rx', data=data)  
plt.show()
```



```
In [25]: sns.countplot(x="Persistency_Flag", hue='Age_Bucket', data=data)  
plt.show()
```

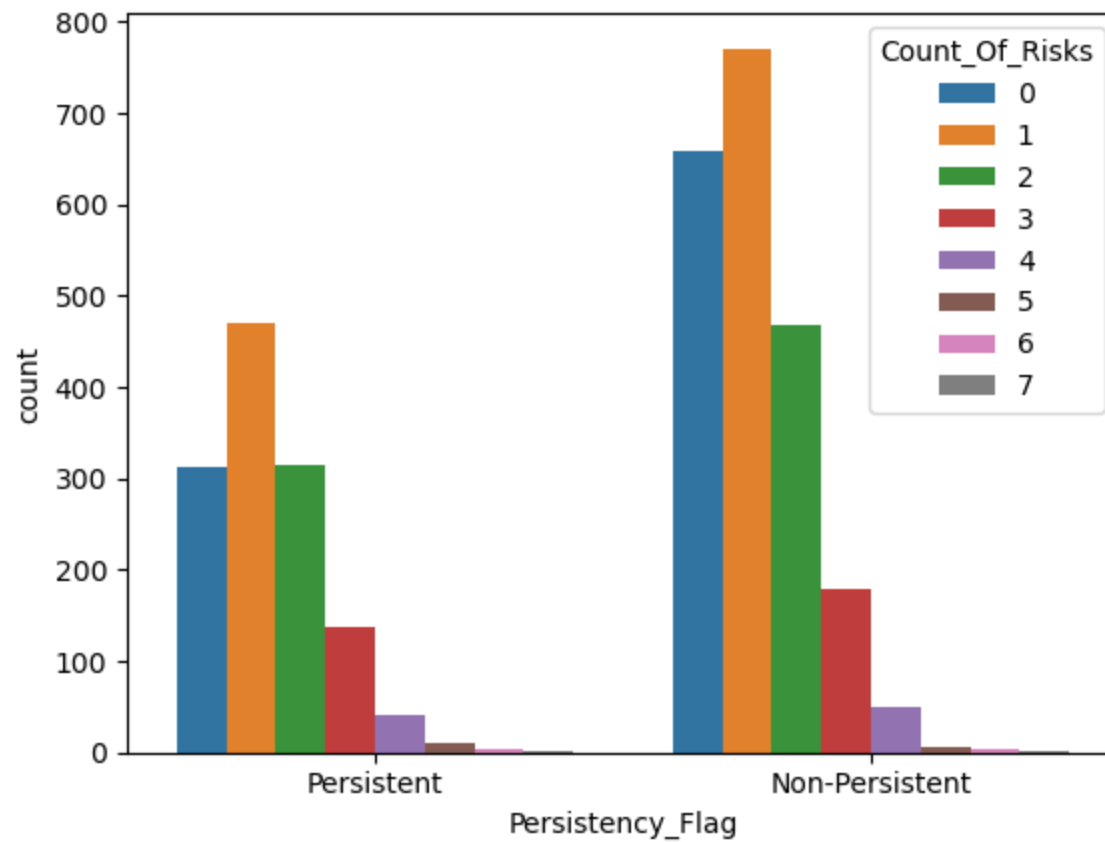


```
In [26]: sns.countplot(x="Persistency_Flag", hue='Gender', data=data)  
plt.show()
```



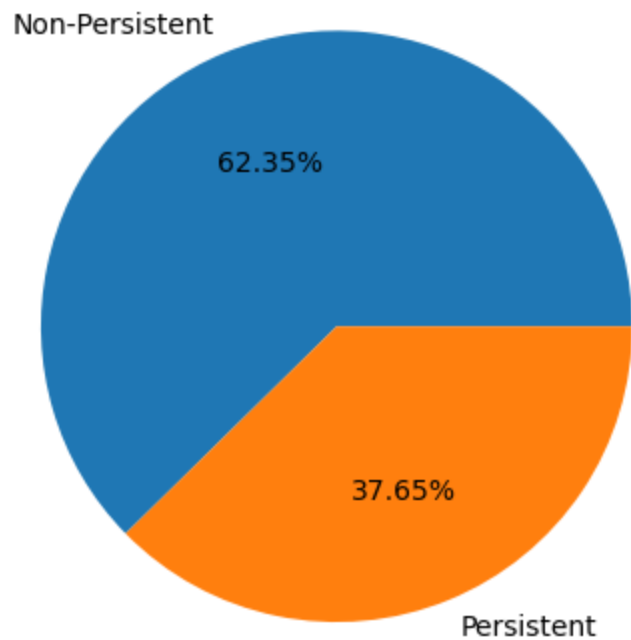
```
In [27]: sns.countplot(x="Persistency_Flag", hue='Count_Of_Risks', data=data)
```

```
Out[27]: <Axes: xlabel='Persistency_Flag', ylabel='count'>
```

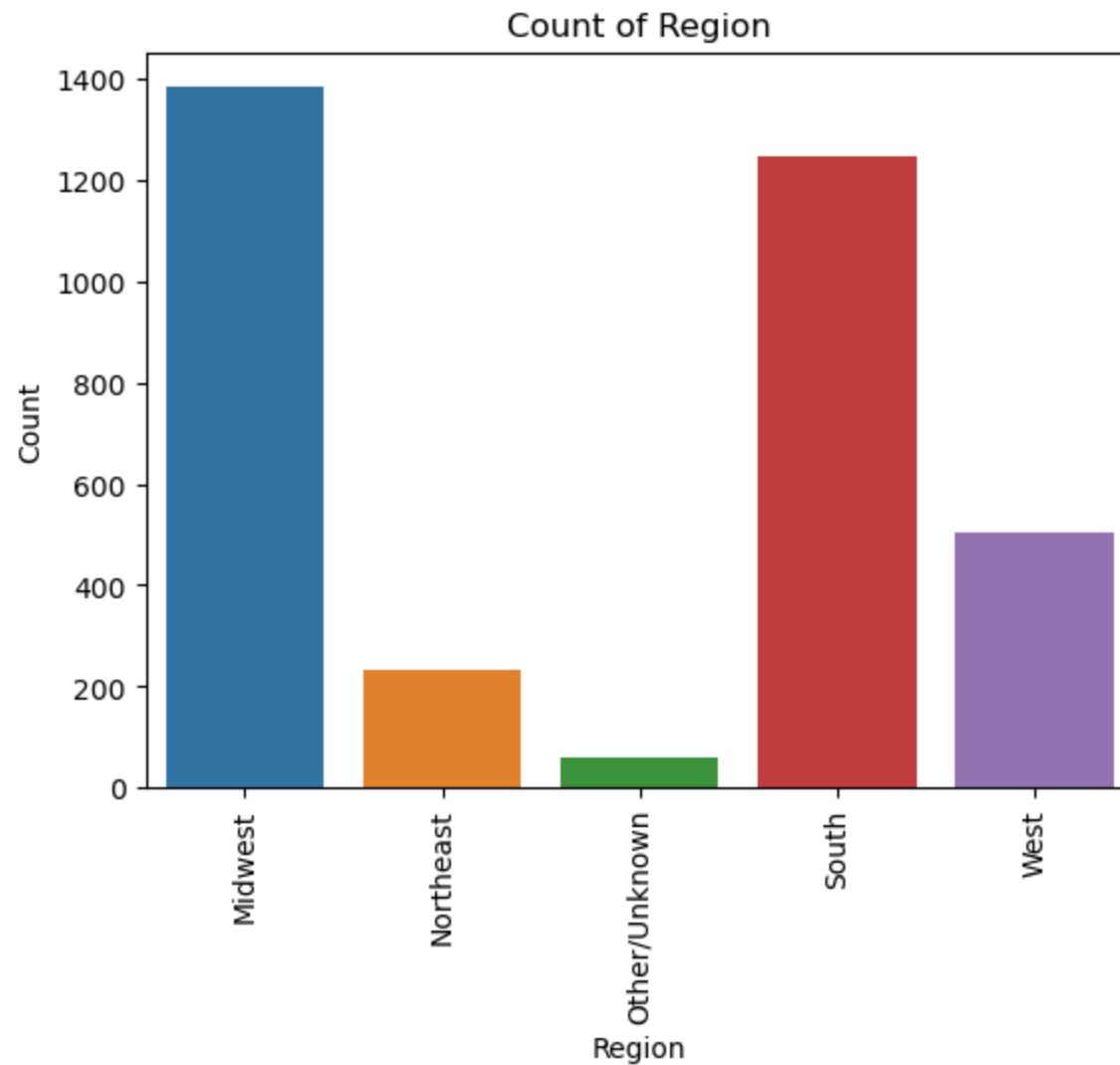




```
In [28]: b=data.groupby("Persistency_Flag")["Persistency_Flag"].count()  
plt.pie(b,labels=b.index,autopct="%.2f%%")  
plt.show()
```

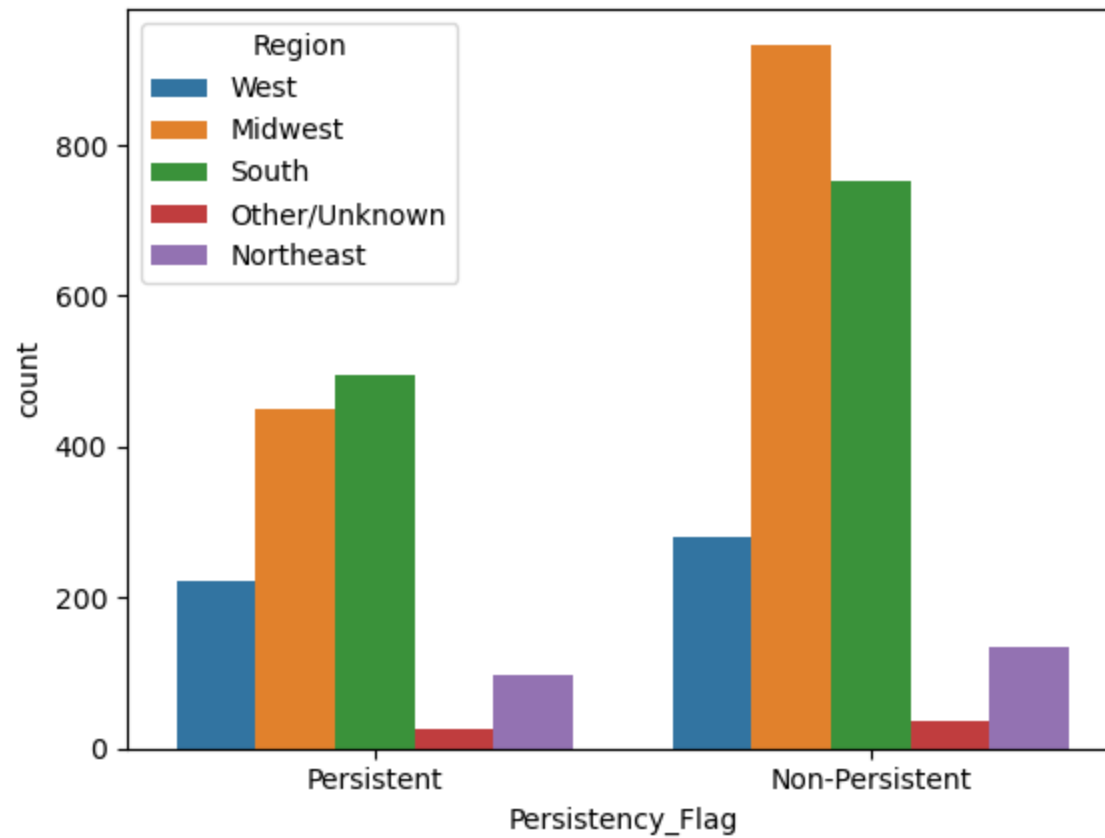


```
In [29]: a=data.groupby("Region")["Region"].count()  
sns.barplot(x=a.index,y=a.values)  
plt.xticks(rotation=90)  
plt.title("Count of Region")  
plt.xlabel("Region")  
plt.ylabel("Count")  
plt.show()
```



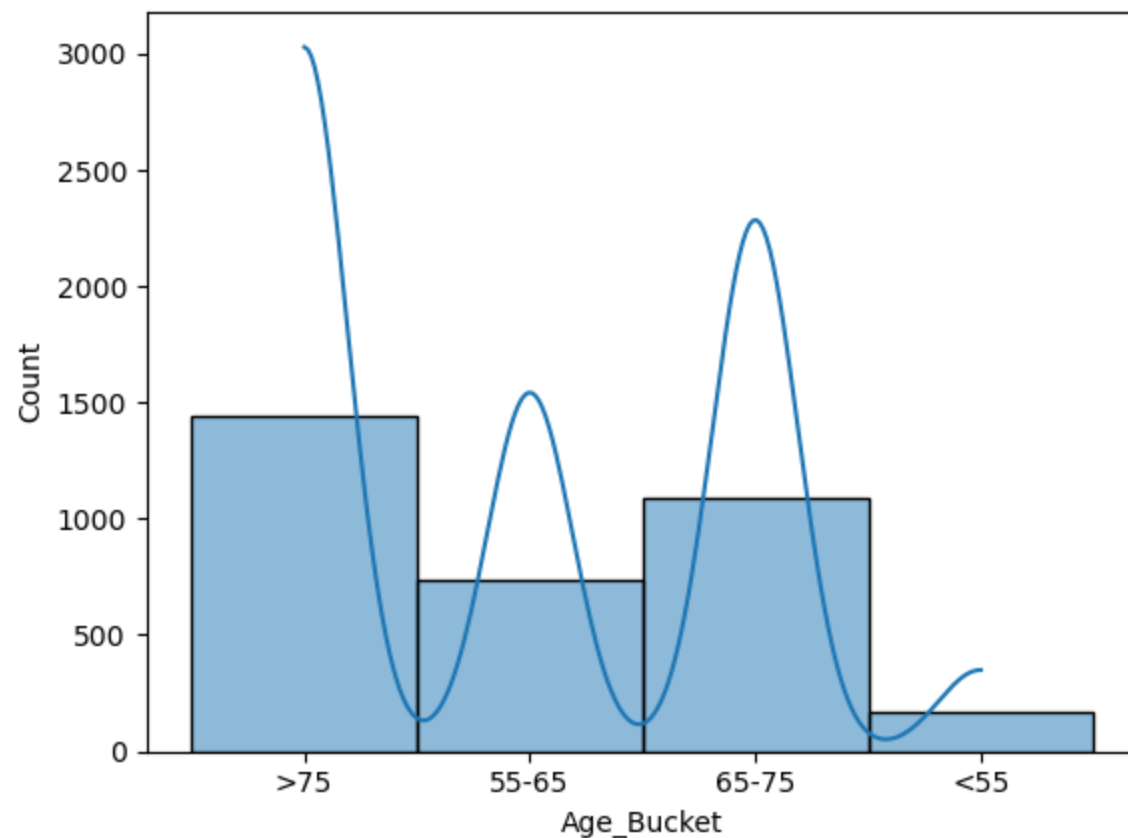
```
In [30]: sns.countplot(x="Persistency_Flag", hue='Region', data=data)
```

```
Out[30]: <Axes: xlabel='Persistency_Flag', ylabel='count'>
```



```
In [31]: sns.histplot(data["Age_Bucket"],bins=10,kde=True)
```

```
Out[31]: <Axes: xlabel='Age_Bucket', ylabel='Count'>
```



## Outlier detection and removal

```
In [32]: from scipy import stats  
z_scores=stats.zscore(data["Dexa_Freq_During_Rx"])  
z_score_outliers=(z_scores<-3)|(z_scores>3)
```

```
In [33]: z_score_outlier_rows=data[z_score_outliers]
print("outliers detected by Z-score:",z_score_outlier_rows)
```

outliers detected by Z-score:			Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region
\								
198	P199	Persistent	Female	Caucasian	Not Hispanic	South		
241	P242	Persistent	Female	Caucasian	Not Hispanic	Midwest		
541	P542	Persistent	Female	Caucasian	Not Hispanic	Midwest		
651	P652	Persistent	Female	Caucasian	Not Hispanic	Midwest		
1265	P1266	Persistent	Female	Caucasian	Not Hispanic	West		
1360	P1361	Persistent	Female	Caucasian	Not Hispanic	South		
1370	P1371	Non-Persistent	Female	Caucasian	Not Hispanic	South		
1398	P1399	Persistent	Female	Asian	Not Hispanic	South		
1734	P1735	Persistent	Male	Caucasian	Not Hispanic	Midwest		
1838	P1839	Persistent	Female	Caucasian	Not Hispanic	Northeast		
1854	P1855	Persistent	Male	Caucasian	Not Hispanic	Northeast		
1901	P1902	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest		
1909	P1910	Persistent	Female	Caucasian	Not Hispanic	Midwest		
1920	P1921	Persistent	Female	Other/Unknown	Unknown	Midwest		
1949	P1950	Persistent	Female	Caucasian	Not Hispanic	Midwest		
1993	P1994	Persistent	Female	Caucasian	Not Hispanic	South		
2006	P2007	Non-Persistent	Female	Caucasian	Not Hispanic	South		

```
In [34]: data.shape
```

```
Out[34]: (3424, 69)
```

```
In [35]: x=(z_scores>-3)&(z_scores<3)
```

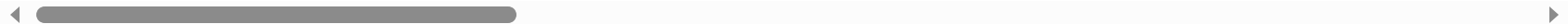
```
In [36]: new_data=data[x] # create a new data frame
```

In [37]: new\_data

Out[37]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ni
0	P1	Persistent	Male	Caucasian	Not Hispanic	West	>75	GENERAL PRACTITIONER	Others	OB/GYN
1	P2	Non-Persistent	Male	Asian	Not Hispanic	West	55-65	GENERAL PRACTITIONER	Others	OB/GYN
2	P3	Non-Persistent	Female	Other/Unknown	Hispanic	Midwest	65-75	GENERAL PRACTITIONER	Others	OB/GYN
3	P4	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
4	P5	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
...	...	...	...	...	...	...	...	...	...	...
3419	P3420	Persistent	Female	Caucasian	Not Hispanic	South	>75	GENERAL PRACTITIONER	Others	OB/GYN
3420	P3421	Persistent	Female	Caucasian	Not Hispanic	South	>75	Unknown	Others	OB/GYN
3421	P3422	Persistent	Female	Caucasian	Not Hispanic	South	>75	ENDOCRINOLOGY	Specialist	
3422	P3423	Non-Persistent	Female	Caucasian	Not Hispanic	South	55-65	Unknown	Others	OB/GYN
3423	P3424	Non-Persistent	Female	Caucasian	Not Hispanic	South	65-75	Unknown	Others	OB/GYN

3367 rows × 69 columns



In [38]: z\_scores=stats.zscore(new\_data["Count\_Of\_Risks"])  
z\_score\_outlier=(z\_scores<-3)|(z\_scores>3)

```
In [39]: z_score_outlier_row=new_data[z_score_outlier]
print("outliers detected by Z-score:",z_score_outlier_row)
```

outliers detected by Z-score:				Ptid	Persistency_Flag	Gender	Race	Ethnicity \
302	P303	Persistent	Female			Caucasian	Not Hispanic	
342	P343	Persistent	Female			Caucasian	Not Hispanic	
352	P353	Persistent	Female			Caucasian	Not Hispanic	
495	P496	Persistent	Female			Caucasian	Not Hispanic	
557	P558	Persistent	Female			Caucasian	Not Hispanic	
731	P732	Persistent	Female			Caucasian	Not Hispanic	
741	P742	Persistent	Female			Caucasian	Not Hispanic	
754	P755	Persistent	Female			Caucasian	Not Hispanic	
787	P788	Non-Persistent	Female			Caucasian	Not Hispanic	
817	P818	Persistent	Female	African	American	Not Hispanic		
1059	P1060	Non-Persistent	Female			Caucasian	Not Hispanic	
1112	P1113	Persistent	Female			Caucasian	Not Hispanic	
1247	P1248	Non-Persistent	Male			Caucasian	Not Hispanic	
1759	P1760	Persistent	Female			Caucasian	Not Hispanic	
1798	P1799	Non-Persistent	Female			Caucasian	Hispanic	
2091	P2092	Non-Persistent	Female			Caucasian	Not Hispanic	
2592	P2593	Non-Persistent	Female			Caucasian	Not Hispanic	
2601	P2602	Non-Persistent	Female			Caucasian	Not Hispanic	

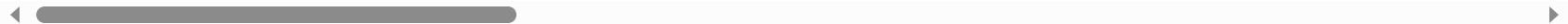
```
In [40]: p=(z_scores>-3)&(z_scores<3)
data_new=new_data[p]
```

In [41]: data\_new

Out[41]:

	Ptid	Persistency_Flag	Gender	Race	Ethnicity	Region	Age_Bucket	Ntm_Speciality	Ntm_Specialist_Flag	Ni
0	P1	Persistent	Male	Caucasian	Not Hispanic	West	>75	GENERAL PRACTITIONER	Others	OB/GYN
1	P2	Non-Persistent	Male	Asian	Not Hispanic	West	55-65	GENERAL PRACTITIONER	Others	OB/GYN
2	P3	Non-Persistent	Female	Other/Unknown	Hispanic	Midwest	65-75	GENERAL PRACTITIONER	Others	OB/GYN
3	P4	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
4	P5	Non-Persistent	Female	Caucasian	Not Hispanic	Midwest	>75	GENERAL PRACTITIONER	Others	OB/GYN
...	...	...	...	...	...	...	...	...	...	...
3419	P3420	Persistent	Female	Caucasian	Not Hispanic	South	>75	GENERAL PRACTITIONER	Others	OB/GYN
3420	P3421	Persistent	Female	Caucasian	Not Hispanic	South	>75	Unknown	Others	OB/GYN
3421	P3422	Persistent	Female	Caucasian	Not Hispanic	South	>75	ENDOCRINOLOGY	Specialist	
3422	P3423	Non-Persistent	Female	Caucasian	Not Hispanic	South	55-65	Unknown	Others	OB/GYN
3423	P3424	Non-Persistent	Female	Caucasian	Not Hispanic	South	65-75	Unknown	Others	OB/GYN

3344 rows × 69 columns



In [ ]:



```
In [42]: from sklearn.preprocessing import OneHotEncoder,StandardScaler
categorical_cols=['PtId', 'Gender', 'Race', 'Ethnicity', 'Region', 'Age_Bucket', 'Ntm_Speciality',
'Ntm_Specialist_Flag', 'Ntm_Speciality_Bucket', 'Gluko_Record_Prior_Ntm',
'Gluko_Record_During_Rx', 'Dexa_During_Rx', 'Frag_Frac_Prior_Ntm',
'Frag_Frac_During_Rx', 'Risk_Segment_Prior_Ntm', 'Tscore_Bucket_Prior_Ntm',
'Risk_Segment_During_Rx', 'Tscore_Bucket_During_Rx', 'Change_T_Score',
'Change_Risk_Segment', 'Adherent_Flag', 'Idn_Indicator',
'Injectable_Experience_During_Rx', 'Comorb_Encounter_For_Screening_For_Malignant_Neoplasms',
'Comorb_Encounter_For_Immunization', 'Comorb_Encntr_For_General_Exam_W_O_Complaint,_Susp_Or_Reprtd_Dx', 'Comor
'Comorb_Other_Joint_Disorder_Not_Elsewhere_Classified',
'Comorb_Encntr_For_Oth_Sp_Exam_W_O_Complaint_Suspected_Or_Reprtd_Dx',
'Comorb_Long_Term_Current_Drug_Therapy', 'Comorb_Dorsalgia',
'Comorb_Personal_History_Of_Other_Diseases_And_Conditions',
'Comorb_Other_Disorders_Of_Bone_Density_And_Structure',
'Comorb_Disorders_of_lipoprotein_metabolism_and_other_lipidemias',
'Comorb_Osteoporosis_without_current_pathological_fracture', 'Comorb_Personal_history_of_malignant_neoplasm',
'Comorb_Gastro_esophageal_reflux_disease', 'Concom_Cholesterol_And_Triglyceride_Regulating_Preparations',
'Concom_Narcotics', 'Concom_Systemic_Corticosteroids_Plain', 'Concom_Anti_Depressants_And_Mood_Stabilisers',
'Concom_Fluoroquinolones', 'Concom_Cephalosporins', 'Concom_Macrolides_And_Similar_Types',
'Concom_Broad_Spectrum_Penicillins', 'Concom_Anaesthetics_General', 'Concom_Viral_Vaccines',
'Risk_Type_1_Insulin_Dependent_Diabetes', 'Risk_Osteogenesis_Imperfecta', 'Risk_Rheumatoid_Arthritis',
'Risk_Untreated_Chronic_Hyperthyroidism', 'Risk_Untreated_Chronic_Hypogonadism',
'Risk_Untreated_Early_Menopause', 'Risk_Patient_Parent_Fractured_Their_Hip',
'Risk_Smoking_Tobacco', 'Risk_Chronic_Malnutrition_Or_Malabsorption', 'Risk_Chronic_Liver_Disease',
'Risk_Family_History_Of_Osteoporosis', 'Risk_Low_Calcium_Intake', 'Risk_Vitamin_D_Insufficiency',
'Risk_Poor_Health_Frailty', 'Risk_Excessive_Thinness', 'Risk_Hysterectomy_Oophorectomy',
'Risk_Estrogen_Deficiency', 'Risk_Immobilization', 'Risk_Recurring_Falls','Count_Of_Risks']
encoder=OneHotEncoder(drop='first',sparse=False)
```


```
In [43]: encoder=OneHotEncoder(drop='first',sparse=False)
encoder_cols=pd.DataFrame(encoder.fit_transform(data[categorical_cols]),columns=encoder.get_feature_names_out(
```

```
In [44]: encoder_cols
```

Out[44]:

	Ptid_P10	Ptid_P100	Ptid_P1000	Ptid_P1001	Ptid_P1002	Ptid_P1003	Ptid_P1004	Ptid_P1005	Ptid_P1006	Ptid_P1007	...	Risk_E
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
...	...	...	...	...	...	...	...	...	...	...	...	
3419	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3420	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3421	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3422	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3423	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	

3424 rows × 3544 columns



```
In [45]: x = encoder_cols
y = data['Persistence_Flag']
```

In [46]:

x

Out[46]:

	Ptid_P10	Ptid_P100	Ptid_P1000	Ptid_P1001	Ptid_P1002	Ptid_P1003	Ptid_P1004	Ptid_P1005	Ptid_P1006	Ptid_P1007	...	Risk_E
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
...	...	...	...	...	...	...	...	...	...	...	...	
3419	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3420	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3421	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3422	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	
3423	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	

3424 rows × 3544 columns



In [47]:

y

Out[47]:

```

0      Persistent
1      Non-Persistent
2      Non-Persistent
3      Non-Persistent
4      Non-Persistent
...
3419    Persistent
3420    Persistent
3421    Persistent
3422    Non-Persistent
3423    Non-Persistent
Name: Persistency_Flag, Length: 3424, dtype: object

```

```
In [48]: from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report
```

```
In [49]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42)
```

```
In [50]: scaler = StandardScaler()
x_train = scaler.fit_transform(x_train) # train ko sahi karna he bas test ko nahi 5.1ase1 tar 0.51eraise karto
x_test = scaler.fit_transform(x_test) # trans.. data tranform kaarto = fittrans..fit karta transform karto
```

```
In [51]: svc = SVC(kernel='linear')
```

```
In [52]: svc.fit(x_train, y_train)
```

```
Out[52]: SVC
SVC(kernel='linear')
```

```
In [53]: y_pred = svc.predict(x_test)
```

```
In [54]: acc = accuracy_score(y_test, y_pred)
acc
```

```
Out[54]: 0.8072992700729927
```

```
In [55]: print("Accuracy: {:.2f}%".format(acc*100))
Accuracy: 80.73%
```

```
In [56]: print(classification_report(y_test,y_pred))#report =classification Learn karyaLa help karto
```

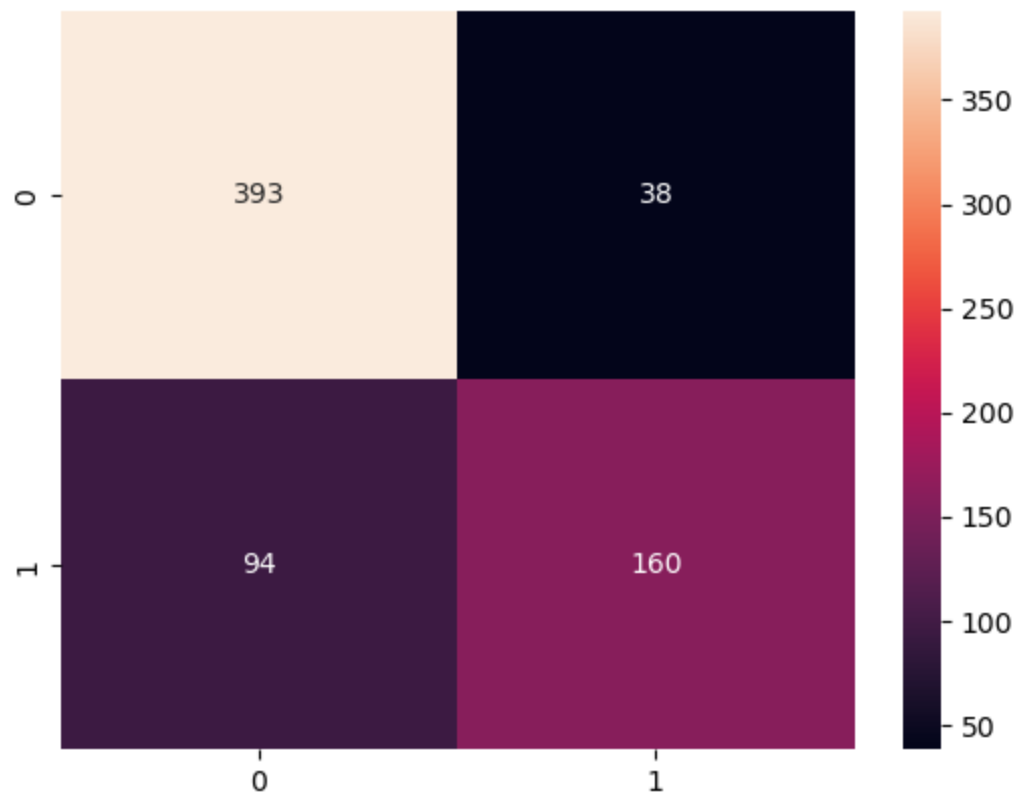
	precision	recall	f1-score	support
Non-Persistent	0.81	0.91	0.86	431
Persistent	0.81	0.63	0.71	254
accuracy			0.81	685
macro avg	0.81	0.77	0.78	685
weighted avg	0.81	0.81	0.80	685

```
In [57]: from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
print("Confusion Matrix")
print(cm)
```

```
Confusion Matrix
[[393  38]
 [ 94 160]]
```

```
In [111]: sns.heatmap(cm, annot=True,fmt='.3g')
```

```
Out[111]: <Axes: >
```



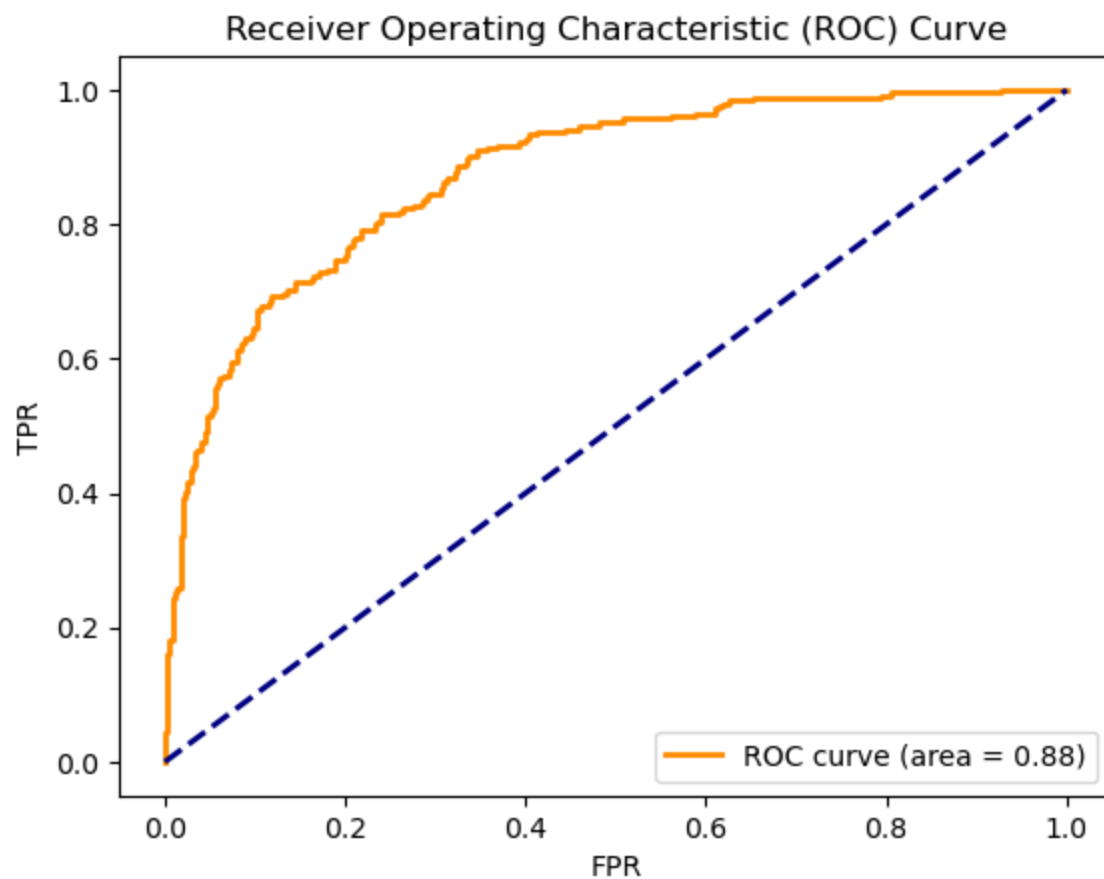
```
In [99]: from sklearn.svm import SVC
from sklearn.metrics import roc_curve, auc
from sklearn.preprocessing import label_binarize
from sklearn.multiclass import OneVsRestClassifier

yb = label_binarize(y, classes=[0,1])
nc=yb.shape[1]
classifier = OneVsRestClassifier(SVC(kernel="linear", probability=True, random_state=42,decision_function_shape="raw"))
y_score=classifier.fit(x_train,y_train).decision_function(x_test)
```

```
In [100]: fpr = dict()
          tpr = dict()
          roc_auc = dict()

          for i in range(nc):
              fpr[i], tpr[i], _ = roc_curve(y_test, y_score, pos_label='Persistent')
              roc_auc[i] = auc(fpr[i], tpr[i])
```

```
In [101]: plt.figure()
plt.plot(fpr[0], tpr[0], color='darkorange', lw=2, label='ROC curve (area = {:.2f})'.format(roc_auc[0]))
plt.plot([0, 1], [0, 1], 'k--', color='navy', lw=2)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('Receiver Operating Characteristic (ROC) Curve')
plt.legend(loc='lower right')
plt.show()
```





```
In [59]: from sklearn.metrics import roc_curve, auc
from sklearn.preprocessing import label_binarize
from sklearn.multiclass import OneVsRestClassifier
```

## GridSearchCV

```
In [60]: from sklearn.model_selection import GridSearchCV
```

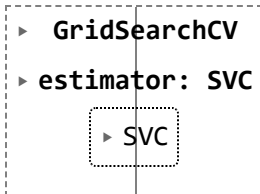
```
In [61]: param_grid = {
    'C' : [0.1, 1, 10, 100],
    'kernel' : ['linear', 'rbf', 'poly', 'sigmoid']
}
```

```
In [62]: svc = SVC()
```

```
In [63]: grid_search = GridSearchCV(svc, param_grid, cv=5)
```

```
In [64]: grid_search.fit(x_train, y_train)
```

```
Out[64]:
```



```
  ▸ GridSearchCV
    ▸ estimator: SVC
      ▸ SVC
```

```
In [65]: best_param = grid_search.best_params_
print("Best hyperparameter : ", best_param)
```

```
Best hyperparameter : {'C': 10, 'kernel': 'sigmoid'}
```

```
In [66]: best_svm = SVC(C=best_param['C'], kernel=best_param['kernel'])
```

```
In [67]: best_svm.fit(x_train, y_train)
```

```
Out[67]: SVC
SVC(C=10, kernel='sigmoid')
```

```
In [68]: y_pred = best_svm.predict(x_test)
acc = accuracy_score(y_test, y_pred)
print("Accuracy : {:.2f}%".format(acc * 100))
```

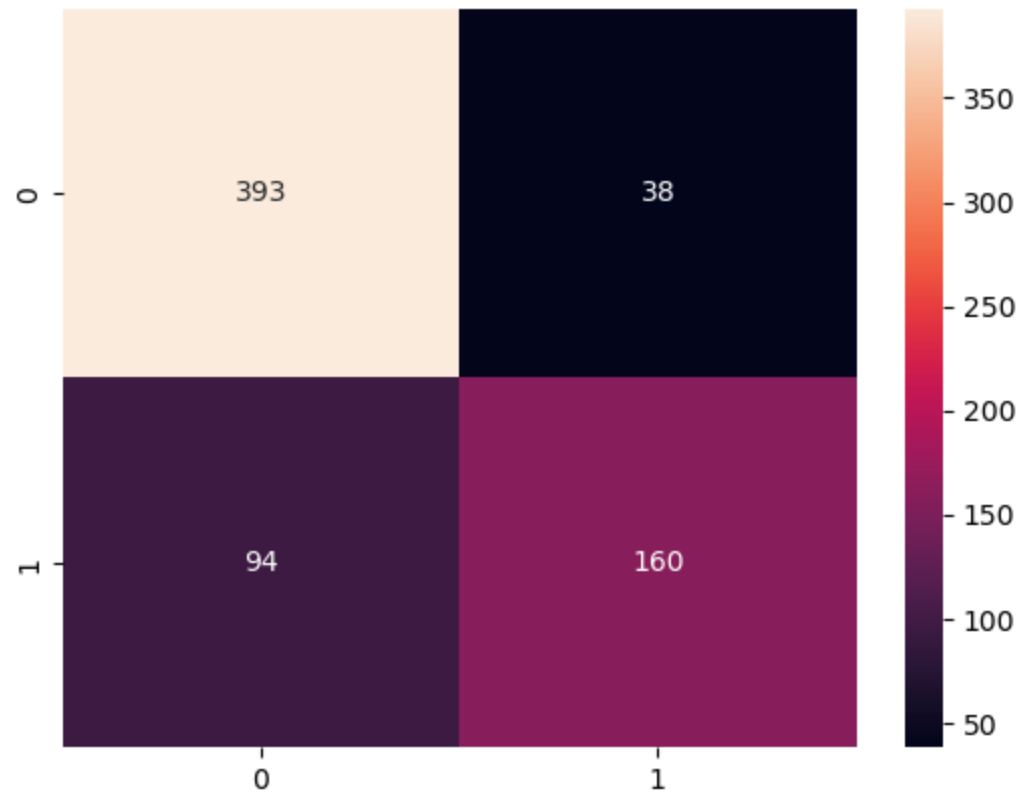
Accuracy : 81.17%

```
In [69]: cm=confusion_matrix(y_test,y_pred)
print("Confusion Matrix : ")
print(cm)
```

Confusion Matrix :

[[392	39]
[ 90	164]]

```
In [110]: sns.heatmap(cm, annot=True,fmt='.3g')  
plt.show()
```



## Random Search

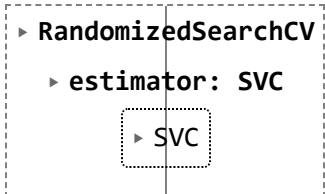
```
In [94]: from sklearn.model_selection import RandomizedSearchCV
```

```
In [95]: param_grid = {  
    'C' : [0.1, 1, 10, 100],  
    'kernel' : ['linear', 'rbf', 'poly', 'sigmoid']  
}
```

```
In [96]: svc = SVC()
```

```
In [97]: random_search = RandomizedSearchCV(svc, param_grid, cv=5)
```

```
In [98]: random_search.fit(x_train, y_train)
```

```
Out[98]:   
  ▸ RandomizedSearchCV  
    ▸ estimator: SVC  
      ▸ SVC
```

```
In [103]: best_parameters = random_search.best_params_  
best_model = random_search.best_estimator_  
print('Hyperparameters:', best_parameters)  
  
Hyperparameters: {'kernel': 'linear', 'C': 0.1}
```

```
In [104]: y_pred = best_model.predict(x_test)
```

```
In [106]: acc=accuracy_score(y_test,y_pred)  
print("Accuracy:", acc)
```

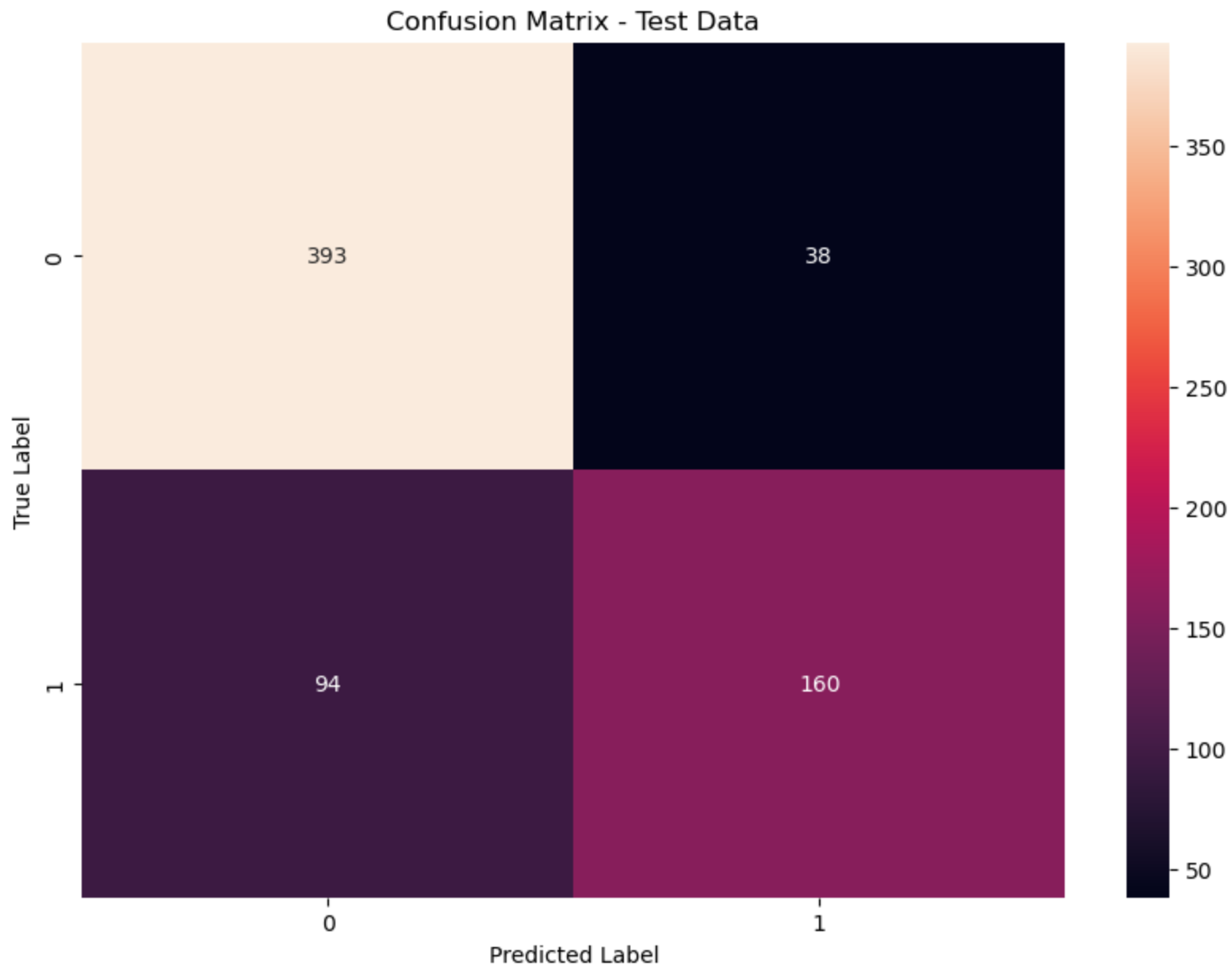
Accuracy: 0.8072992700729927

```
In [108]: print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
Non-Persistent	0.81	0.91	0.86	431
Persistent	0.81	0.63	0.71	254
accuracy			0.81	685
macro avg	0.81	0.77	0.78	685
weighted avg	0.81	0.81	0.80	685

```
In [109]: cm = confusion_matrix(y_test,y_pred)
print('Confusion Matrix: ',cm)
plt.figure(figsize = (10,7))
sns.heatmap(cm, annot=True, fmt='.3g')
plt.title('Confusion Matrix - Test Data')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
```

```
Confusion Matrix: [[393  38]
 [ 94 160]]
```



## Naive Bayes

```
In [112]: from sklearn import model_selection, naive_bayes, metrics, feature_extraction
```

```
In [114]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.25, random_state=42)
```

```
In [115]: from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
x_train = scaler.fit_transform(x_train)
x_test = scaler.transform(x_test)
```

```
In [116]: bayes = naive_bayes.MultinomialNB()
```

```
In [117]: bayes.fit(x_train, y_train)
```

```
Out[117]: ▾ MultinomialNB
MultinomialNB()
```

```
In [118]: y_pred_nb = bayes.predict(x_test)
```

```
In [119]: accuracy = metrics.accuracy_score(y_test, y_pred_nb)
accuracy
```

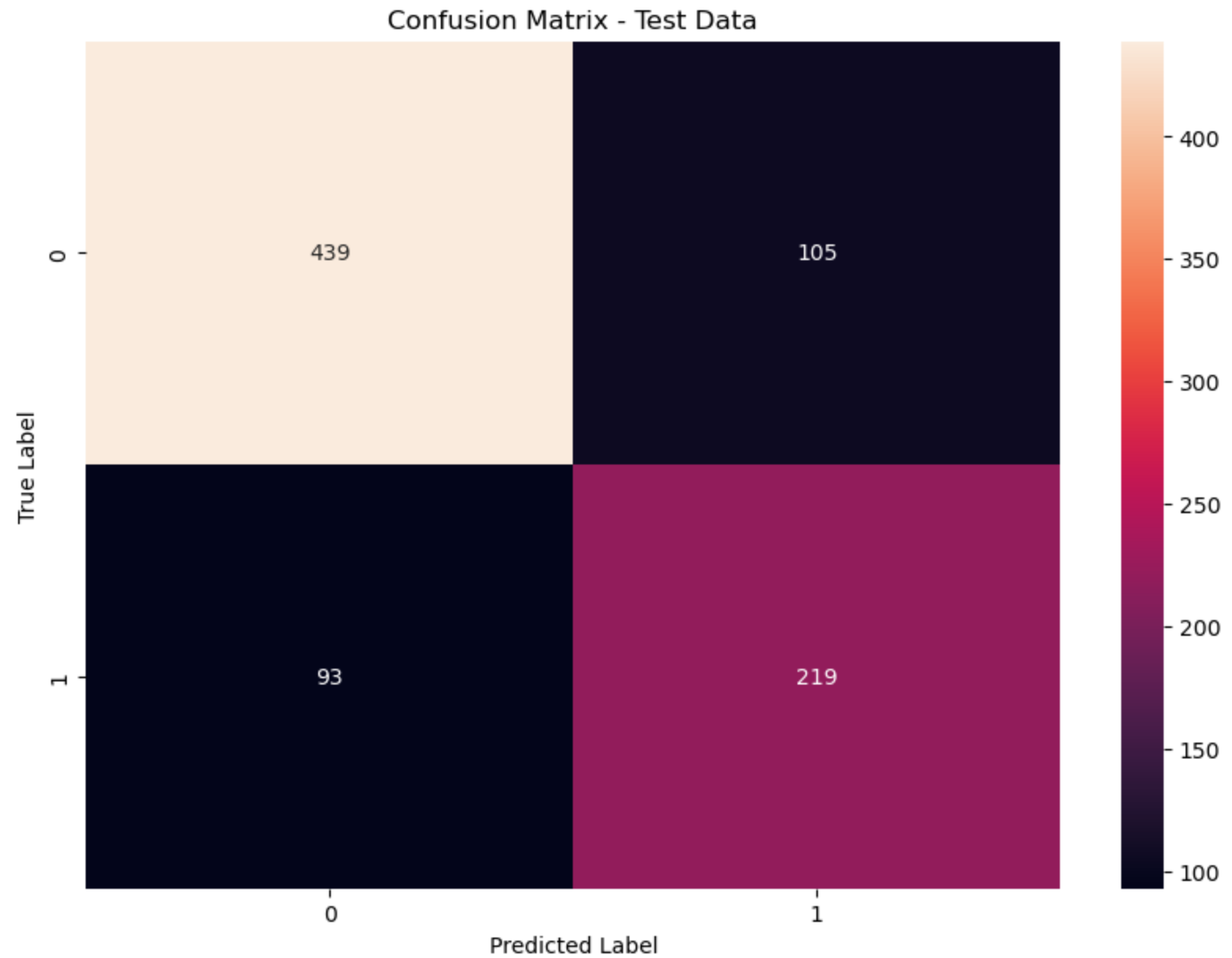
```
Out[119]: 0.7686915887850467
```

```
In [120]: print(metrics.classification_report(y_test, y_pred_nb))
```

	precision	recall	f1-score	support
Non-Persistent	0.83	0.81	0.82	544
Persistent	0.68	0.70	0.69	312
accuracy			0.77	856
macro avg	0.75	0.75	0.75	856
weighted avg	0.77	0.77	0.77	856



```
In [121]: cm=confusion_matrix(y_test,y_pred_nb)
cm
plt.figure(figsize = (10,7))
sns.heatmap(cm, annot=True, fmt='.3g')
plt.title('Confusion Matrix - Test Data')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
```

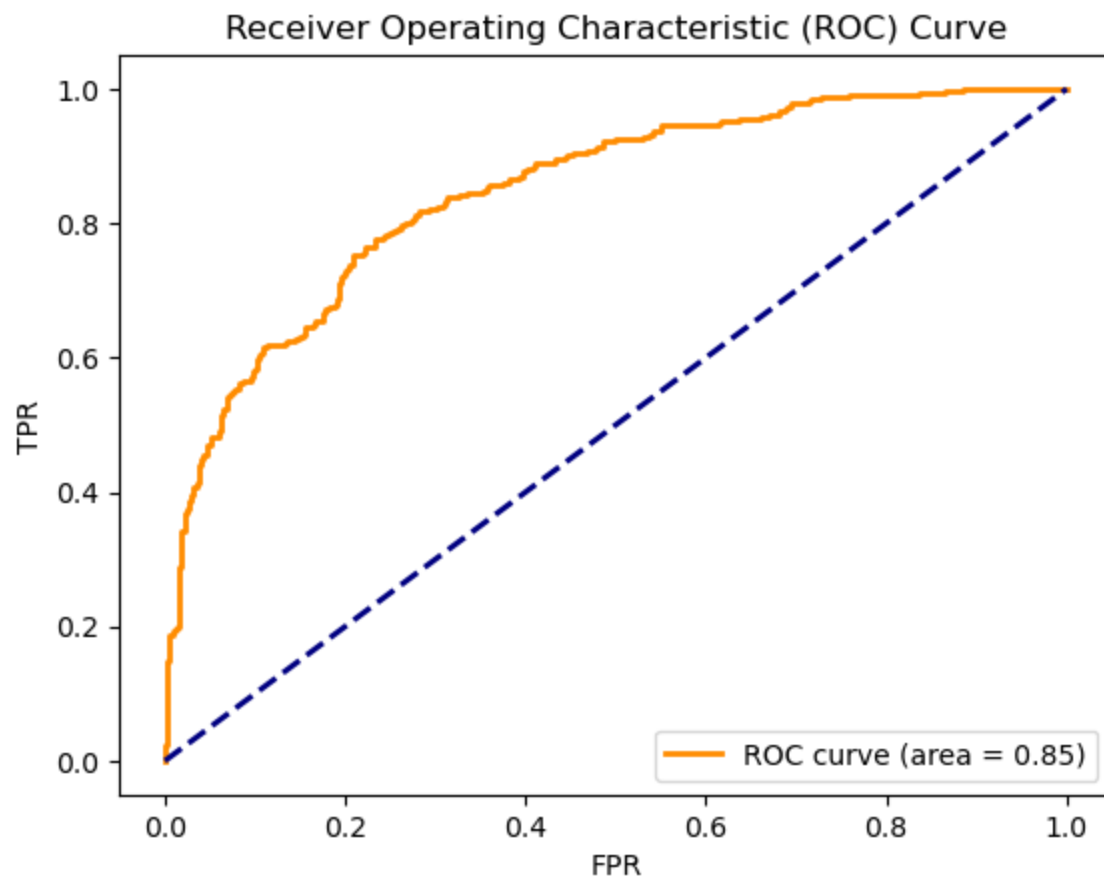


```
In [125]: from sklearn.metrics import roc_curve, auc

fpr = dict()
tpr = dict()
roc_auc = dict()

for i in range(nc):
    y_score = bayes.predict_proba(x_test)[: , 1] # Assuming 'Persistent' is the positive class
    fpr[i], tpr[i], _ = roc_curve(y_test, y_score, pos_label='Persistent')
    roc_auc[i] = auc(fpr[i], tpr[i])
```

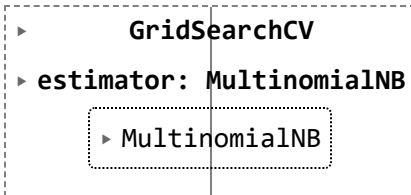
```
In [126]: plt.figure()
plt.plot(fpr[0], tpr[0], color='darkorange', lw=2, label='ROC curve (area = {:.2f})'.format(roc_auc[0]))
plt.plot([0, 1], [0, 1], 'k--', color='navy', lw=2)
plt.xlabel('FPR')
plt.ylabel('TPR')
plt.title('Receiver Operating Characteristic (ROC) Curve')
plt.legend(loc='lower right')
plt.show()
```



## Tuning for Naive Bayes Model

```
In [127]: param_grid = {  
    'alpha': [0.1, 1, 10, 100],  
    'fit_prior': [True, False]  
}
```

```
In [128]: bayes = naive_bayes.MultinomialNB()  
grid_search = GridSearchCV(bayes, param_grid, cv=5)  
grid_search.fit(x_train, y_train)
```

```
Out[128]: 
```

```
In [129]: best_param = grid_search.best_params_  
best_nb = naive_bayes.MultinomialNB(alpha = best_param['alpha'], fit_prior = best_param['fit_prior'])  
best_nb.fit(x_train, y_train)  
y_pred = best_nb.predict(x_test)
```

```
In [130]: print("Best Hyperparameter : ", best_param)  
  
Best Hyperparameter :  {'alpha': 1, 'fit_prior': True}
```

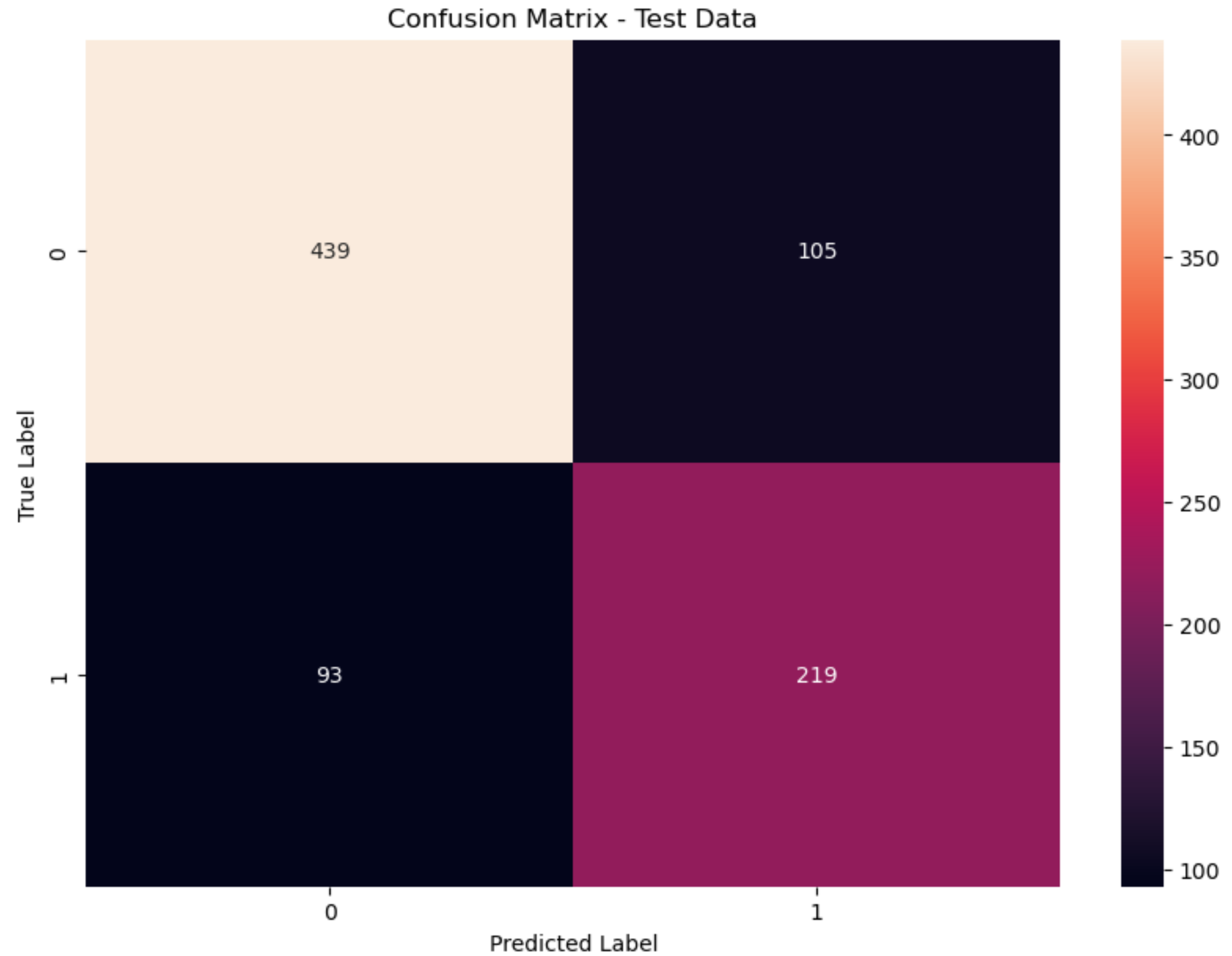
```
In [131]: acc = accuracy_score(y_test, y_pred)  
print('Accuracy', acc)
```

Accuracy 0.7686915887850467

```
In [132]: print (classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
Non-Persistent	0.83	0.81	0.82	544
Persistent	0.68	0.70	0.69	312
accuracy			0.77	856
macro avg	0.75	0.75	0.75	856
weighted avg	0.77	0.77	0.77	856

```
In [133]: cm=confusion_matrix(y_test,y_pred)
cm
plt.figure(figsize = (10,7))
sns.heatmap(cm, annot=True, fmt='.3g')
plt.title('Confusion Matrix - Test Data')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
```



## Randomized Search



In [134]:

```
from scipy.stats import uniform
param_dist = {
    'alpha': uniform(0.1, 2.0), # Example: Uniform distribution for alpha
    'fit_prior': [True, False]
}
```

In [135]:

```
bayes = naive_bayes.MultinomialNB()
```

In [136]:

```
from sklearn.utils.validation import check_non_negative
check_non_negative(x, "MultinomialNB (input x)")
```

In [138]:

```
randomized_search = RandomizedSearchCV(bayes, param_distributions=param_dist, n_iter=10, scoring='accuracy', cv=5)
randomized_search.fit(x, y) # X is your input data, y is your target labels
```

Out[138]:

```
RandomizedSearchCV
  estimator: MultinomialNB
    MultinomialNB
```

In [139]:

```
best_param = randomized_search.best_params_
print("Best Hyperparameter : ", best_param)
```

```
Best Hyperparameter : {'alpha': 1.4195573464854765, 'fit_prior': True}
```

In [140]:

```
best_nb = naive_bayes.MultinomialNB(alpha = best_param['alpha'], fit_prior = best_param['fit_prior'])
best_nb.fit(x_train, y_train)
y_pred = best_nb.predict(x_test)
```

In [141]:

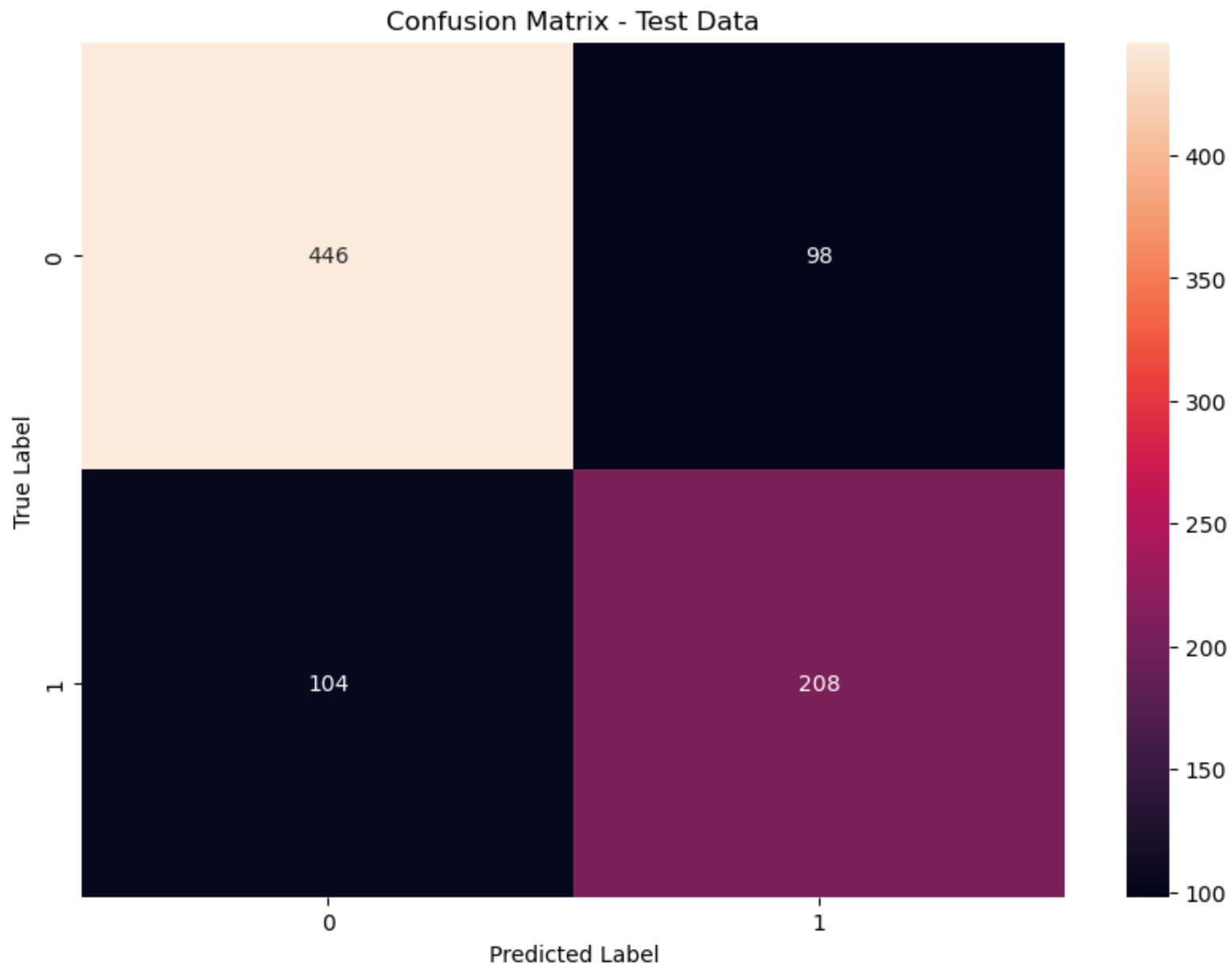
```
acc = accuracy_score(y_test, y_pred)
print('Accuracy', acc)
```

```
Accuracy 0.764018691588785
```

```
In [142]: print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
Non-Persistent	0.81	0.82	0.82	544
Persistent	0.68	0.67	0.67	312
accuracy			0.76	856
macro avg	0.75	0.74	0.74	856
weighted avg	0.76	0.76	0.76	856

```
In [143]: cm=confusion_matrix(y_test,y_pred)
cm
plt.figure(figsize = (10,7))
sns.heatmap(cm, annot=True, fmt='.3g')
plt.title('Confusion Matrix - Test Data')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
```



In [ ]:

