

**Analytics for Hospitals Health-Care Data**  
**TEAM ID: PNT2022TMID16326**

```
import os
for dirname, _, filenames in os.walk('input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
import seaborn as sns
from warnings import filterwarnings
filterwarnings('ignore')
pd.options.display.max_columns = None
pd.options.display.max_rows = None
spd.options.display.float_format = '{:.6f}'.format
from sklearn.model_selection import train_test_split
import statsmodels
import statsmodels.api as sm
from sklearn.preprocessing import StandardScaler
from sklearn import metrics
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report
from sklearn.metrics import cohen_kappa_score
from sklearn.metrics import confusion_matrix
from sklearn.metrics import roc_curve
from sklearn.metrics import accuracy_score
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn import tree
from sklearn.model_selection import GridSearchCV
from sklearn.ensemble import
AdaBoostClassifier, GradientBoostingClassifier
from catboost import CatBoostClassifier
from sklearn.feature_selection import RFE
plt.rcParams['figure.figsize'] = [15,8]
```

```
train = pd.read_csv("D:/HealthCare/train_data.csv")
test = pd.read_csv("D:/HealthCare/test_data.csv")
```

```
train.head()
```

	case_id	Hospital_code	Hospital_type_code	City_Code_Hospital	\
0	1	8	c	3	
1	2	2	c	5	
2	3	10	e	1	
3	4	26	b	2	
4	5	26	b	2	

	Hospital_region_code	Available Extra Rooms in Hospital	Department	\
0	Z	3		

radiotherapy		
1	Z	2
radiotherapy		
2	X	2
anesthesia		
3	Y	2
radiotherapy		
4	Y	2
radiotherapy		

Ward_Type	Ward_Facility_Code	Bed	Grade	patientid
City_Code_Patient \				
0	R	F	2.000000	31397
7.000000				
1	S	F	2.000000	31397
7.000000				
2	S	E	2.000000	31397
7.000000				
3	R	D	2.000000	31397
7.000000				
4	S	D	2.000000	31397
7.000000				

Type of Admission	Severity of Illness	Visitors with Patient	Age
\			
0	Emergency	Extreme	2 51-60
1	Trauma	Extreme	2 51-60
2	Trauma	Extreme	2 51-60
3	Trauma	Extreme	2 51-60
4	Trauma	Extreme	2 51-60

Admission_Deposit	Stay
0	4911.000000 0-10
1	5954.000000 41-50
2	4745.000000 31-40
3	7272.000000 41-50
4	5558.000000 41-50

```
print(train.shape)
print(test.shape)
```

```
(318438, 18)
(137057, 17)
```

```
train.dtypes
```

```

case_id                int64
Hospital_code          int64
Hospital_type_code     object
City_Code_Hospital     int64
Hospital_region_code   object
Available Extra Rooms in Hospital int64
Department            object
Ward_Type             object
Ward_Facility_Code     object
Bed Grade             float64
patientid             int64
City_Code_Patient     float64
Type of Admission      object
Severity of Illness    object
Visitors with Patient  int64
Age                   object
Admission_Deposit     float64
Stay                  object
dtype: object

```

```
train.nunique()
```

```

case_id                318438
Hospital_code          32
Hospital_type_code     7
City_Code_Hospital     11
Hospital_region_code   3
Available Extra Rooms in Hospital 18
Department            5
Ward_Type             6
Ward_Facility_Code     6
Bed Grade             4
patientid             92017
City_Code_Patient     37
Type of Admission      3
Severity of Illness    3
Visitors with Patient  28
Age                   10
Admission_Deposit     7300
Stay                  11
dtype: int64

```

```
train.duplicated().sum()
```

```
0
```

```

train['Hospital_code'] = train['Hospital_code'].astype(object)
train['City_Code_Hospital'] =
train['City_Code_Hospital'].astype(object)
train['Available Extra Rooms in Hospital'] = train['Available Extra
Rooms in Hospital'].astype(object)

```

```
train['Bed Grade'] = train['Bed Grade'].astype(object)
train['City_Code_Patient'] = train['City_Code_Patient'].astype(object)
```

```
train.dtypes
```

```
case_id                int64
Hospital_code          object
Hospital_type_code     object
City_Code_Hospital     object
Hospital_region_code   object
Available Extra Rooms in Hospital object
Department            object
Ward_Type             object
Ward_Facility_Code     object
Bed Grade             object
patientid             int64
City_Code_Patient     object
Type of Admission     object
Severity of Illness    object
Visitors with Patient  int64
Age                   object
Admission_Deposit      float64
Stay                  object
dtype: object
```

```
test['Hospital_code'] = test['Hospital_code'].astype(object)
test['City_Code_Hospital'] = test['City_Code_Hospital'].astype(object)
test['Available Extra Rooms in Hospital'] = test['Available Extra
Rooms in Hospital'].astype(object)
test['Bed Grade'] = test['Bed Grade'].astype(object)
test['City_Code_Patient'] = test['City_Code_Patient'].astype(object)
```

```
test.dtypes
```

```
case_id                int64
Hospital_code          object
Hospital_type_code     object
City_Code_Hospital     object
Hospital_region_code   object
Available Extra Rooms in Hospital object
Department            object
Ward_Type             object
Ward_Facility_Code     object
Bed Grade             object
patientid             int64
City_Code_Patient     object
Type of Admission     object
Severity of Illness    object
Visitors with Patient  int64
Age                   object
```

```

Admission_Deposit                                float64
dtype: object

train.drop(['case_id', 'patientid'], axis=1, inplace=True)

test.drop(['case_id', 'patientid'], axis=1, inplace=True)

train['Stay'] .replace ('More than 100 Days', '100+', inplace=True)

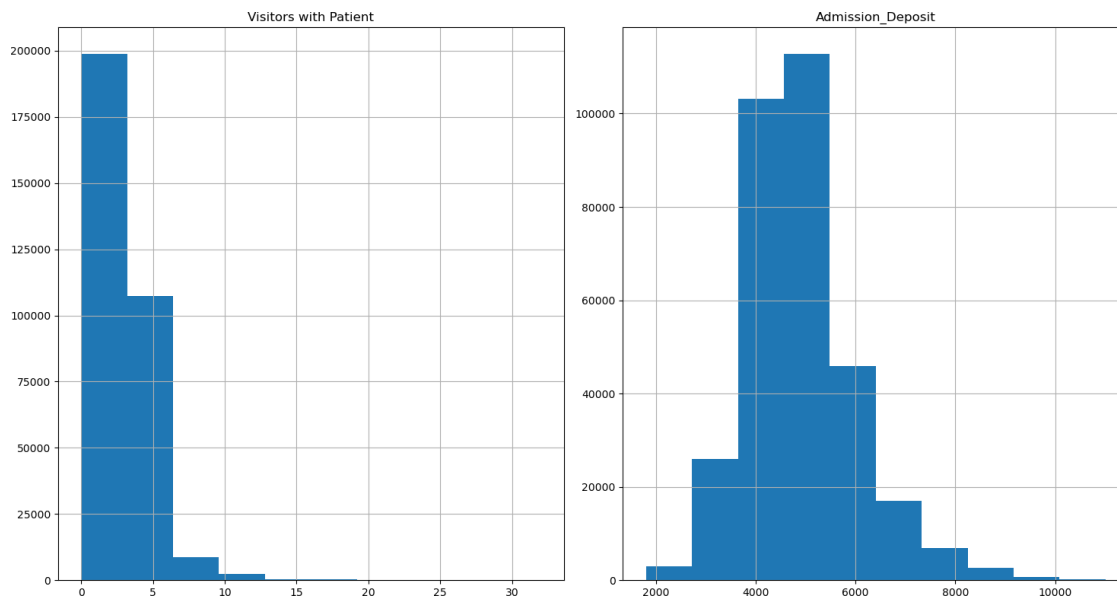
plt.figure(figsize=(15,8))

train.drop('Stay', axis = 1).hist()
plt.tight_layout()
plt.show()

print('Skewness:')
train.drop('Stay', axis = 1).skew()

```

<Figure size 1500x800 with 0 Axes>



Skewness:

```

Hospital_code                                -0.280783
City_Code_Hospital                          0.538809
Available Extra Rooms in Hospital           0.971930
Bed Grade                                   0.051754
City_Code_Patient                          1.581736
Visitors with Patient                      3.137125
Admission_Deposit                          0.931454
dtype: float64

```

```

train['Stay'] .replace('More than 100 Days', '>100', inplace=True)

```

```

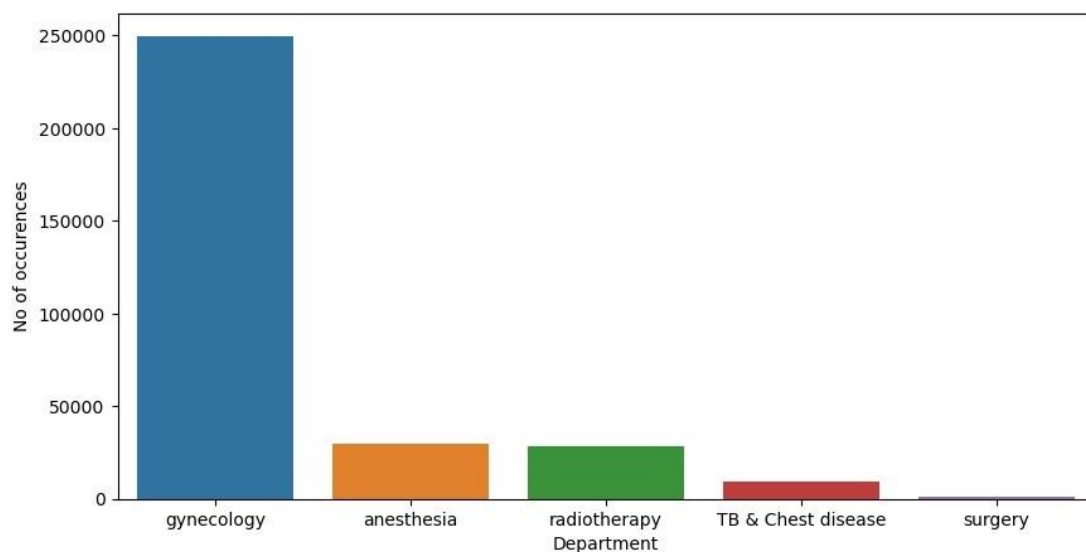
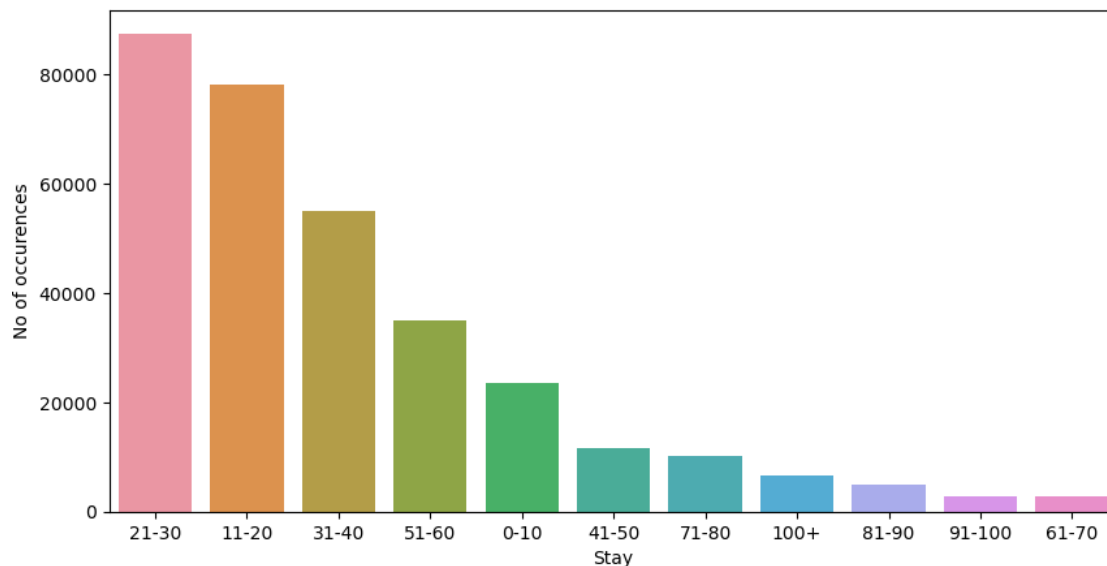
for i in [ 'Stay', 'Department', 'Available Extra Rooms in Hospital',
'Ward_Type' , 'Ward_Facility_Code', 'Age',
          'Type of Admission', 'Severity of Illness', 'Bed Grade',
'Hospital_region_code', 'Hospital_type_code' ,
          'City_Code_Hospital', 'Hospital_code', 'City_Code_Patient',
'Visitors with Patient']:
    count = train[i].value_counts()

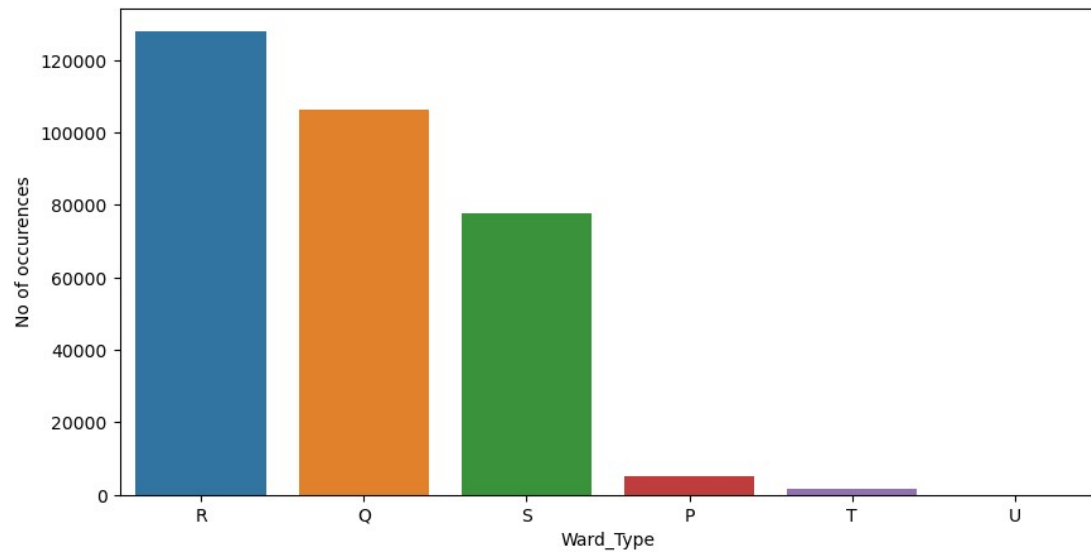
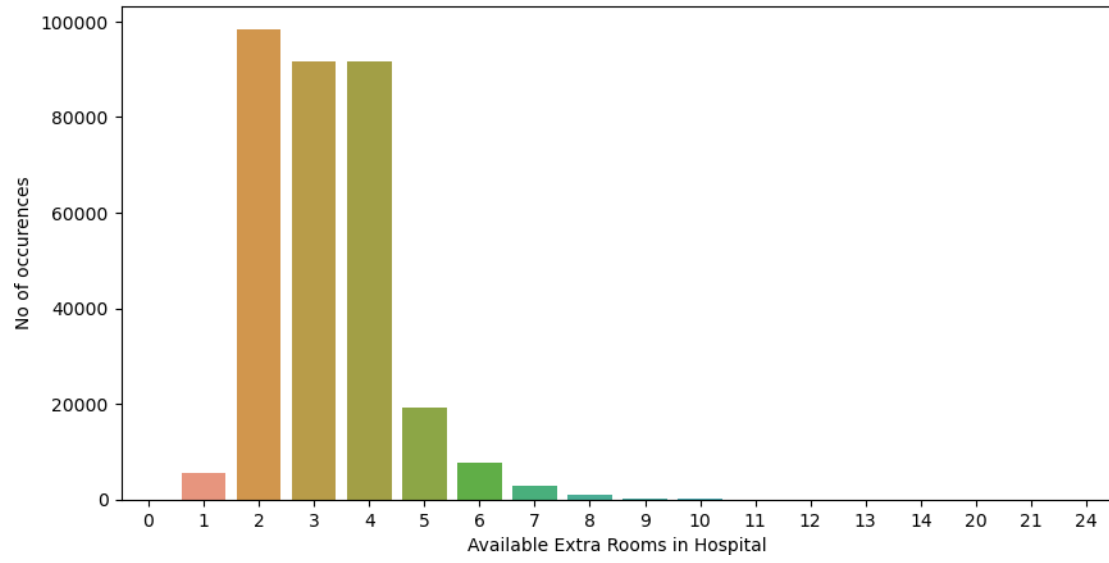
    plt.figure(figsize=(10,5))

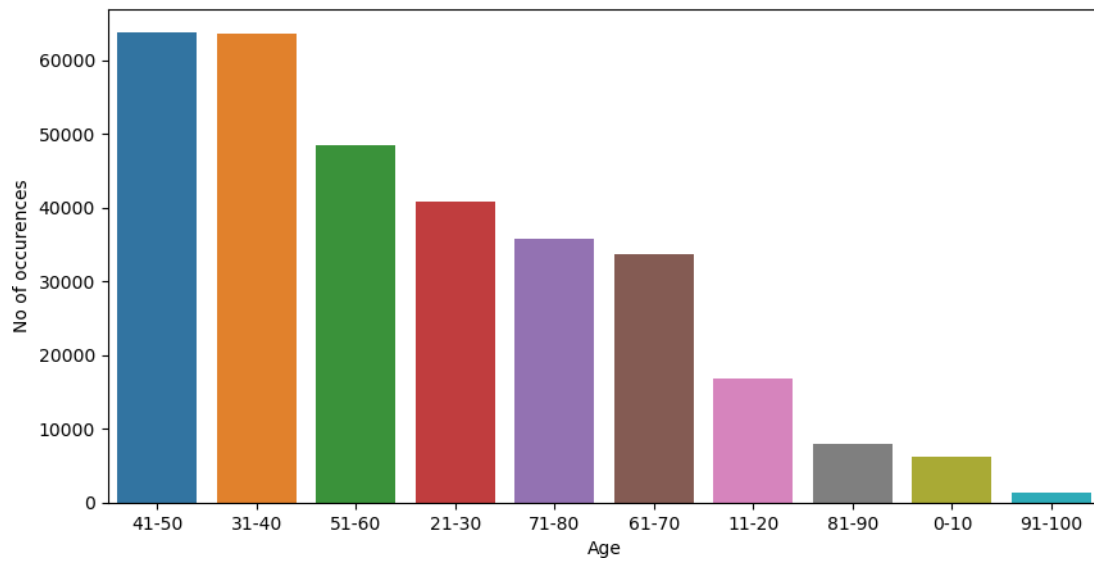
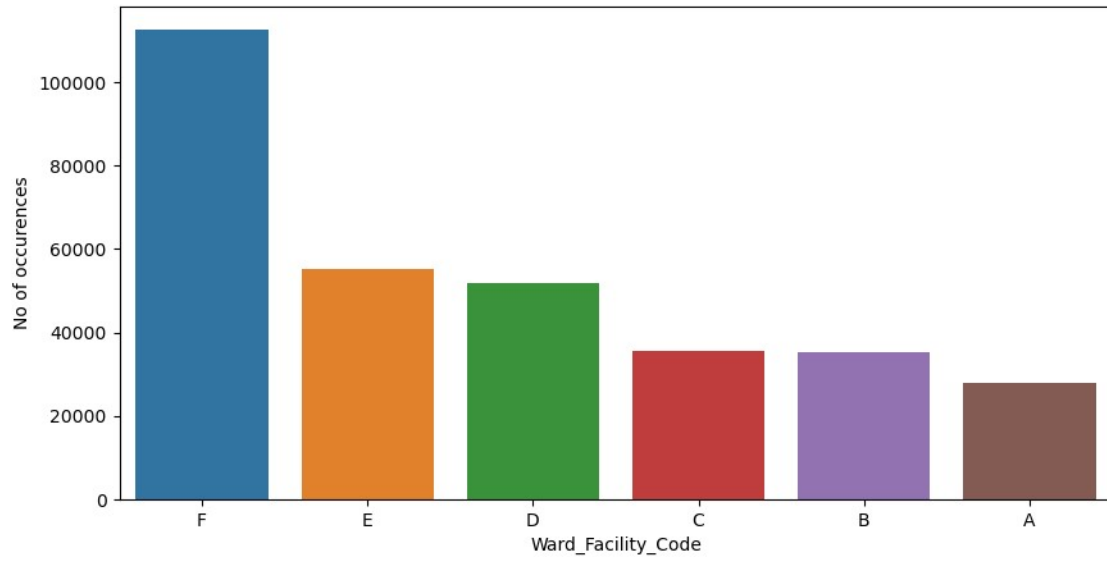
    sns.barplot(x=count.index.values,
y=count.values,data=train)

    plt.xlabel(i)
    plt.ylabel('No of occurences')

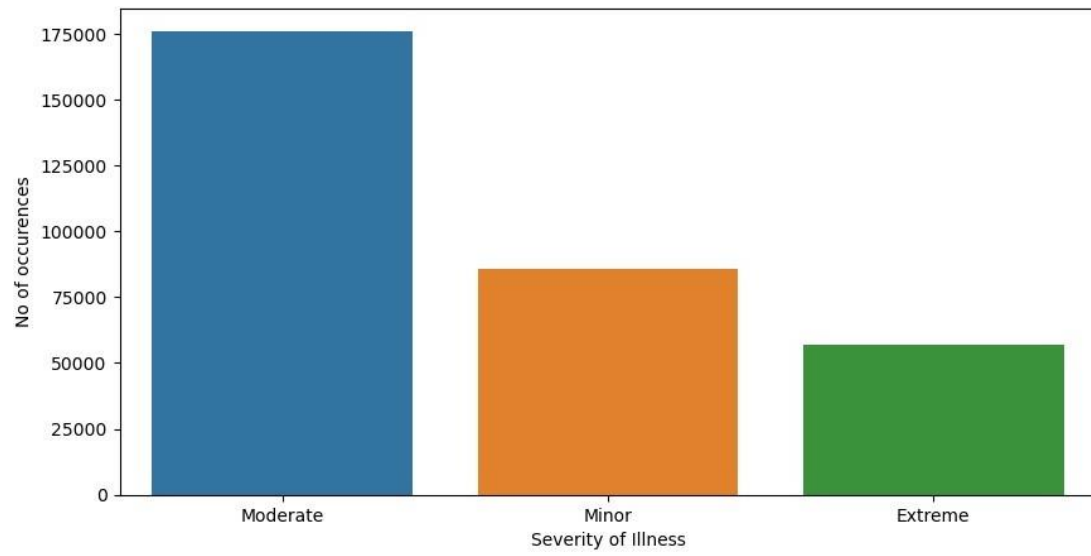
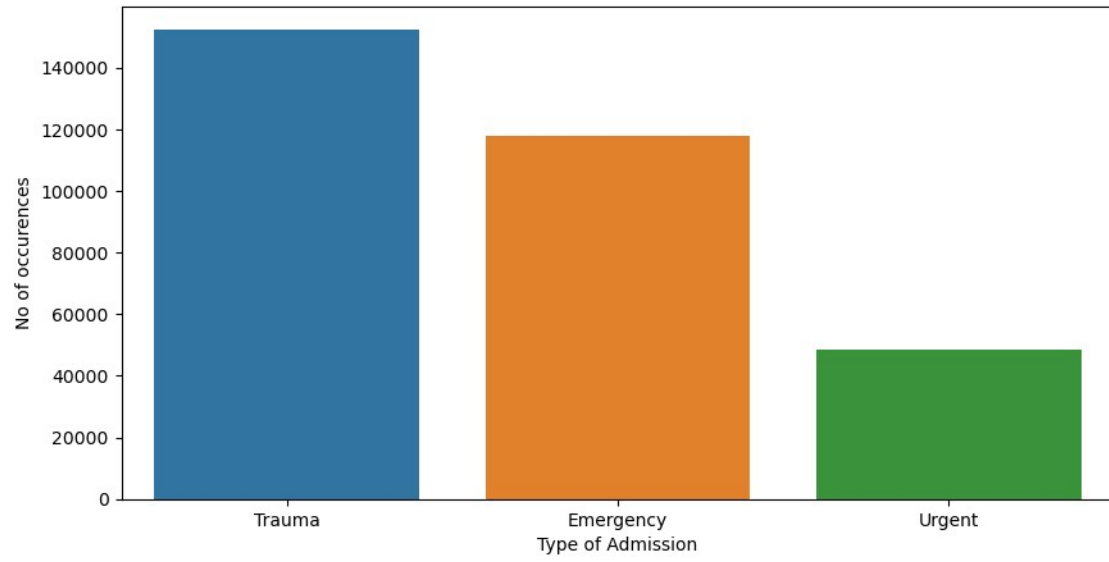
```

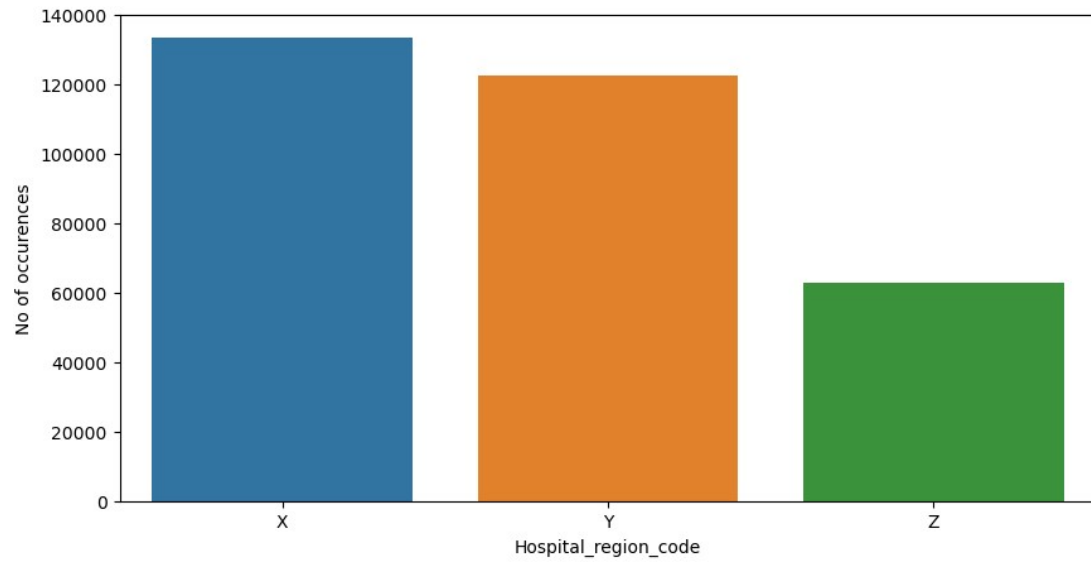
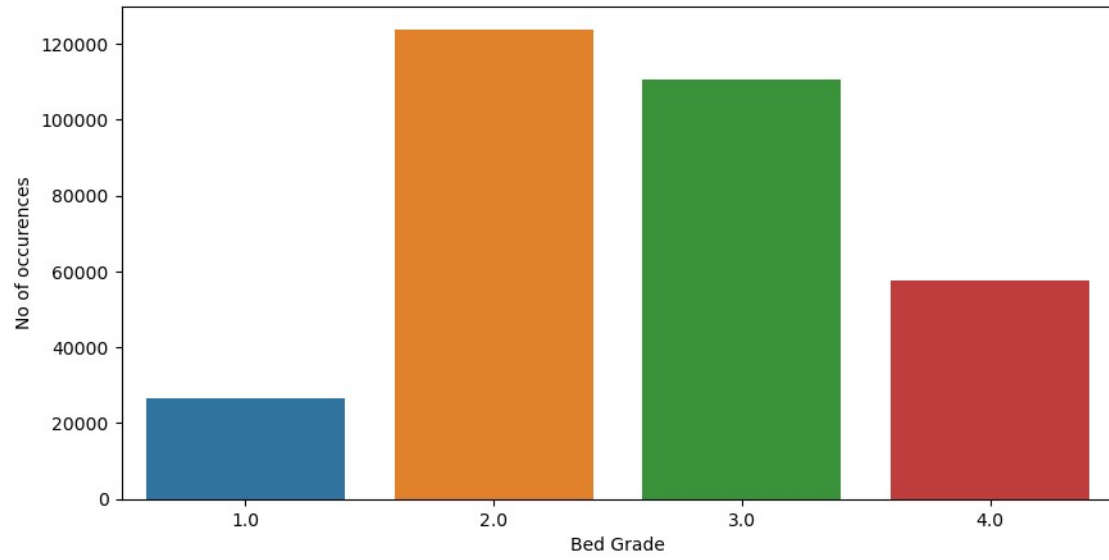


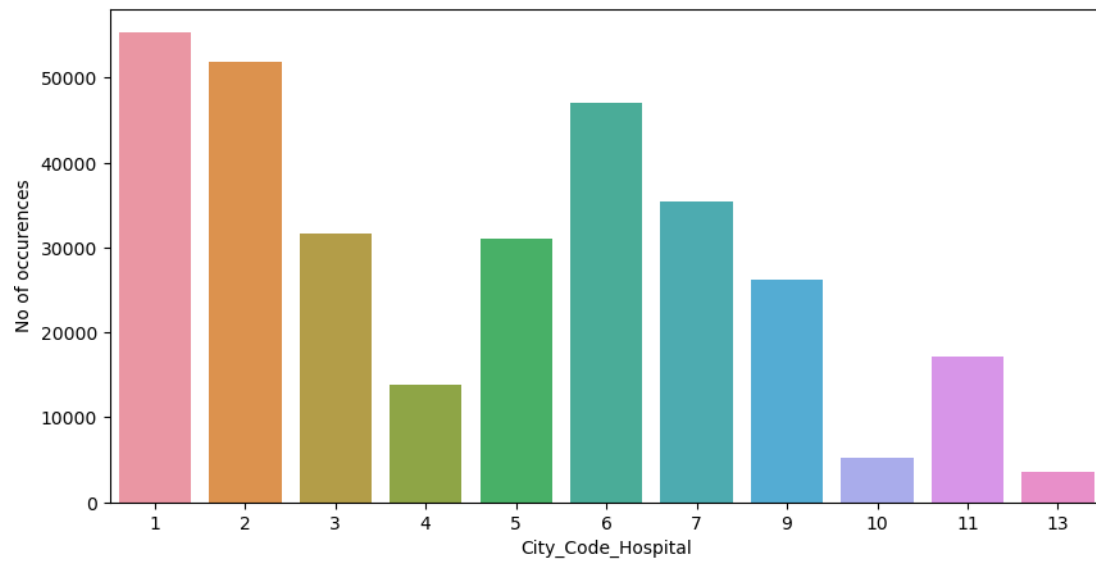
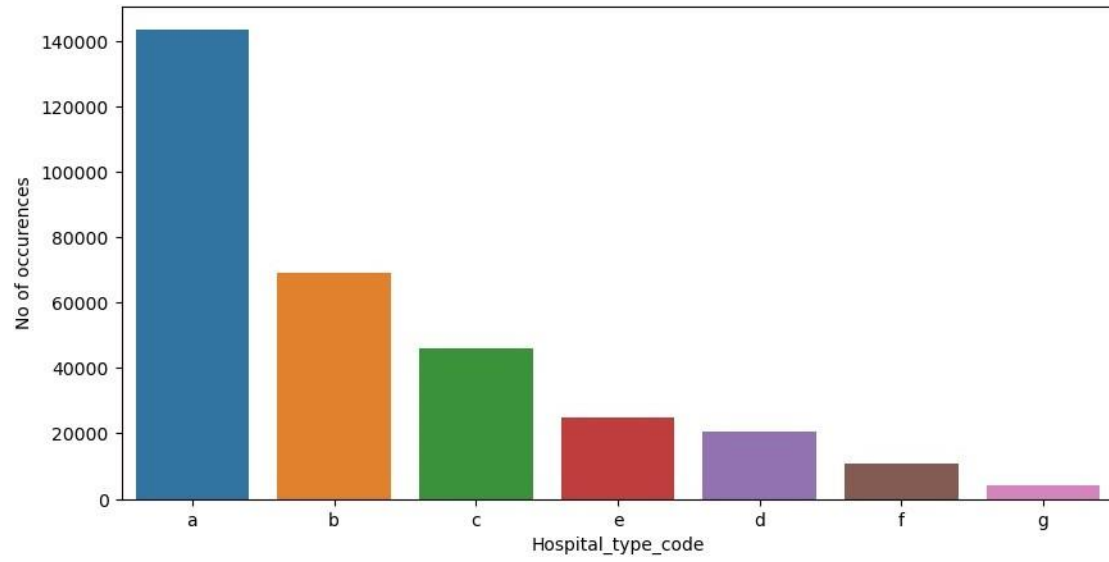


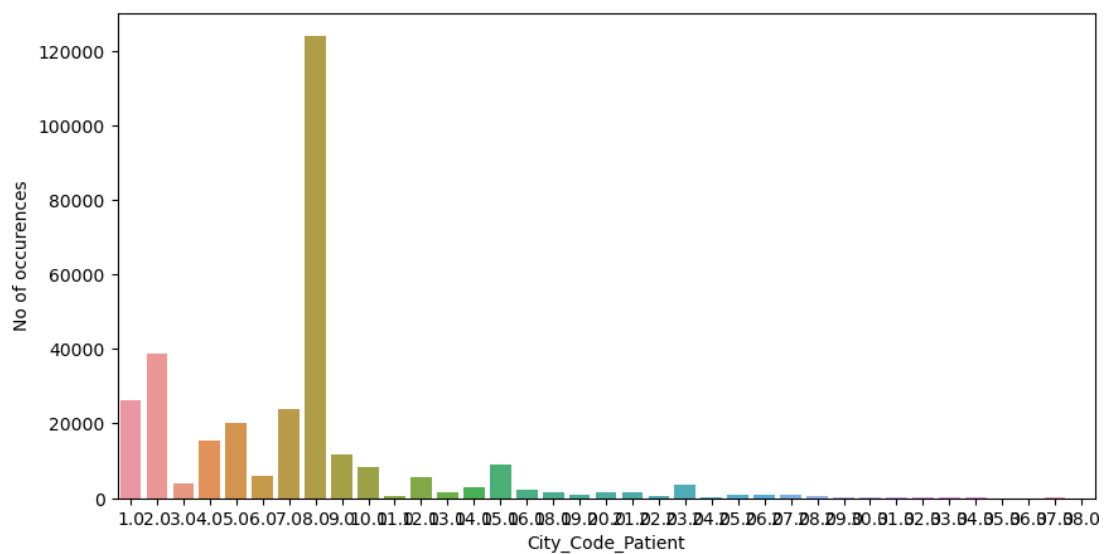
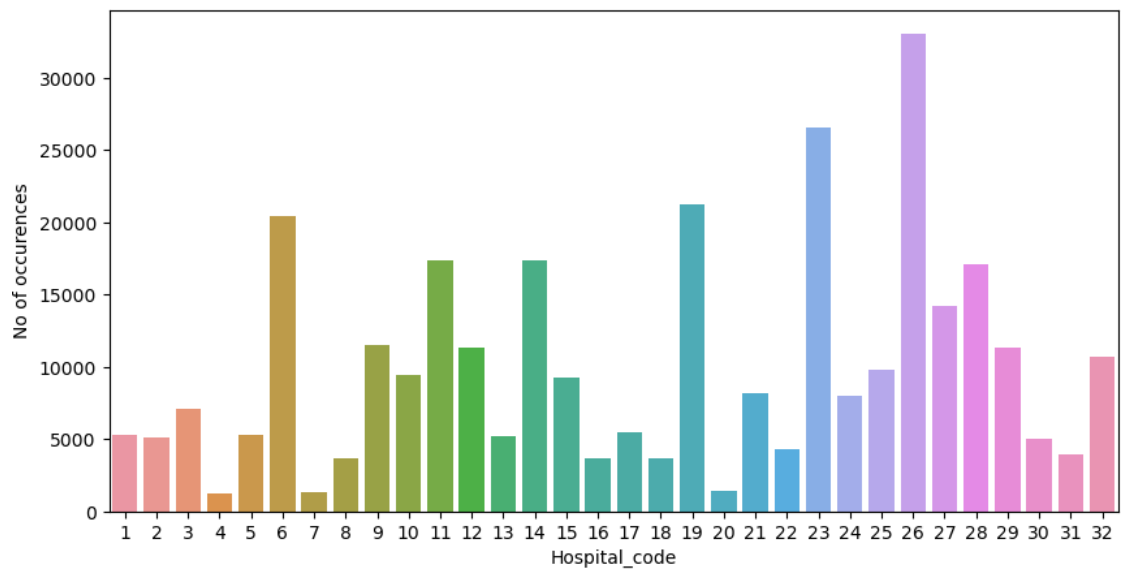


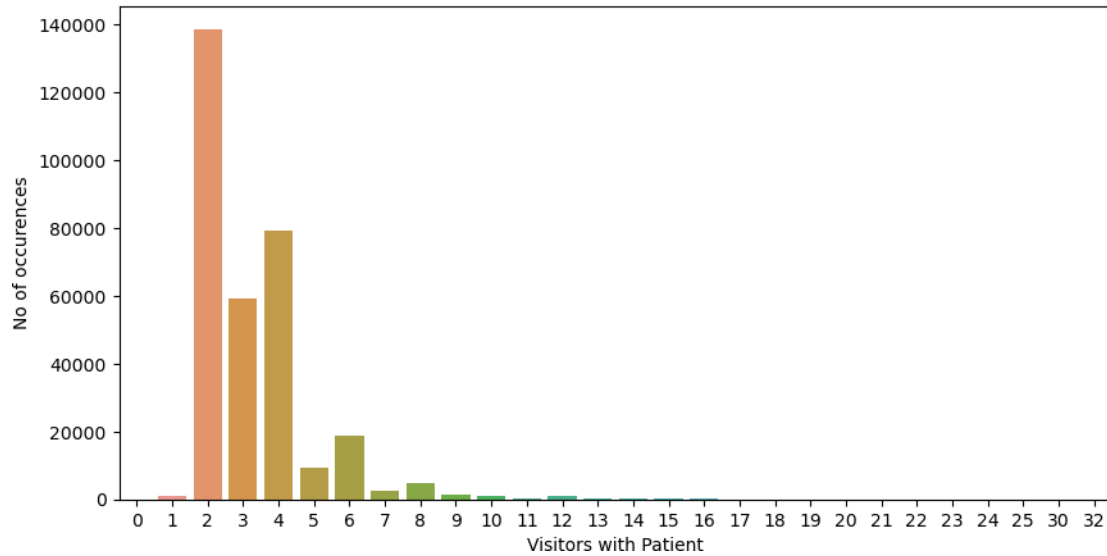








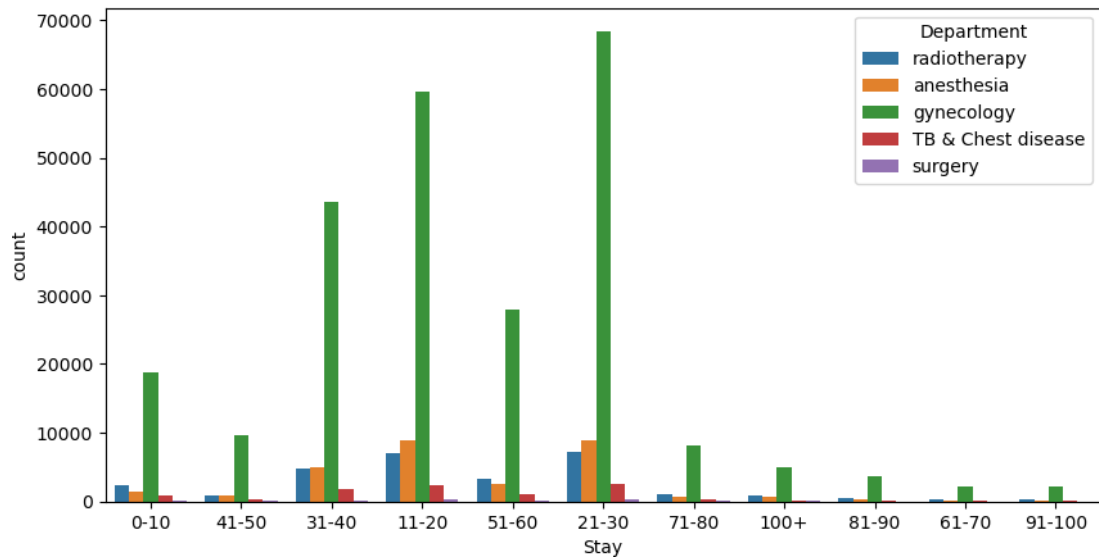


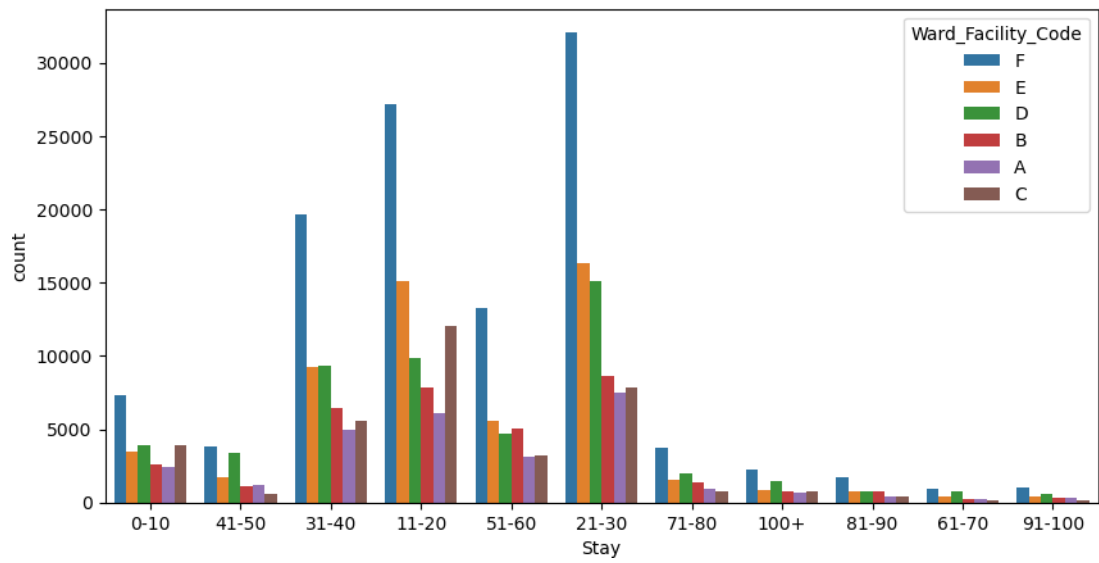
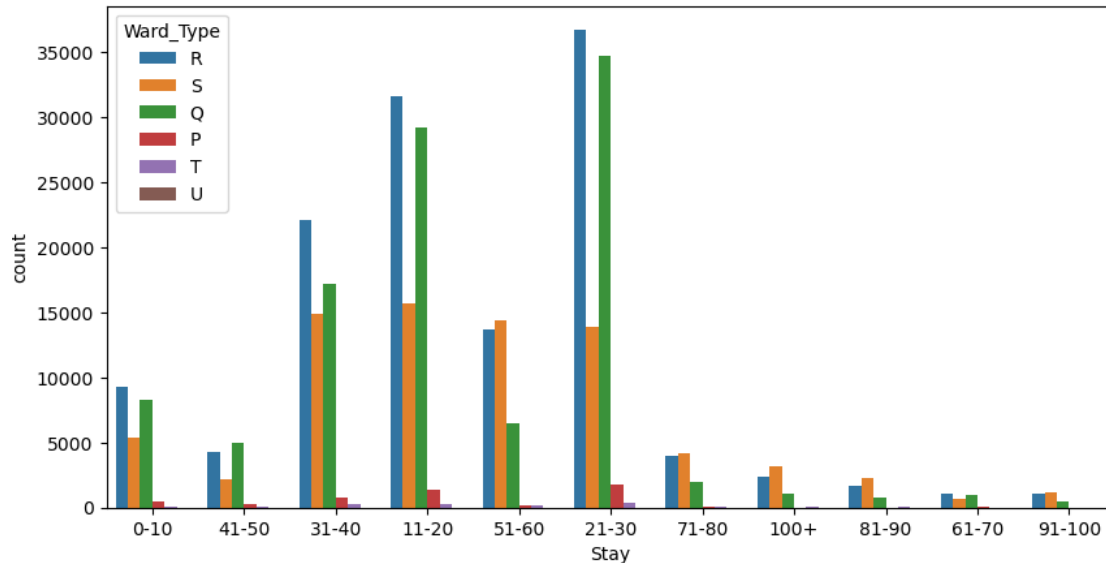


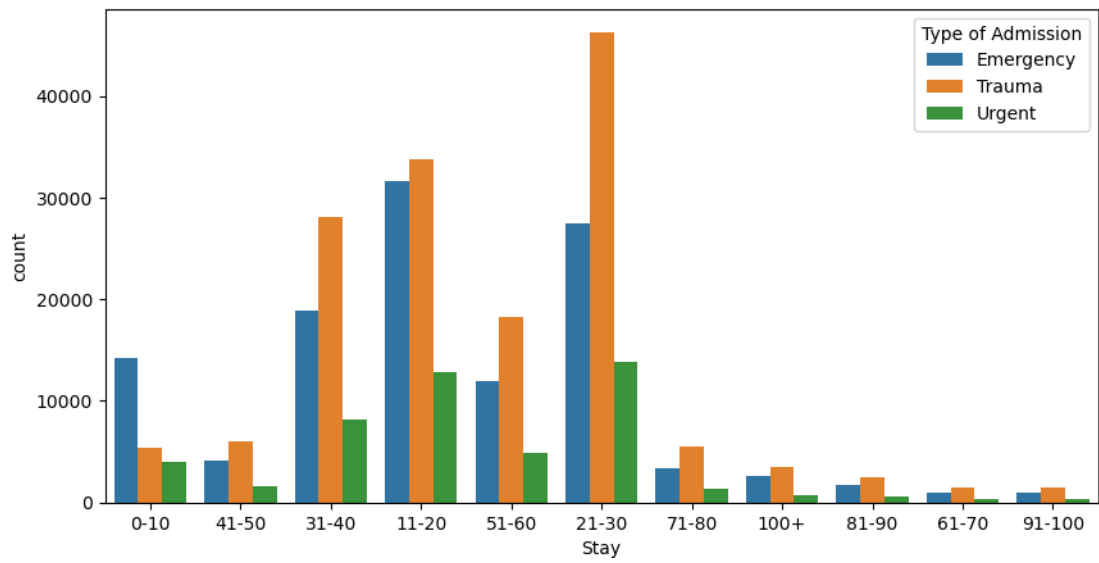
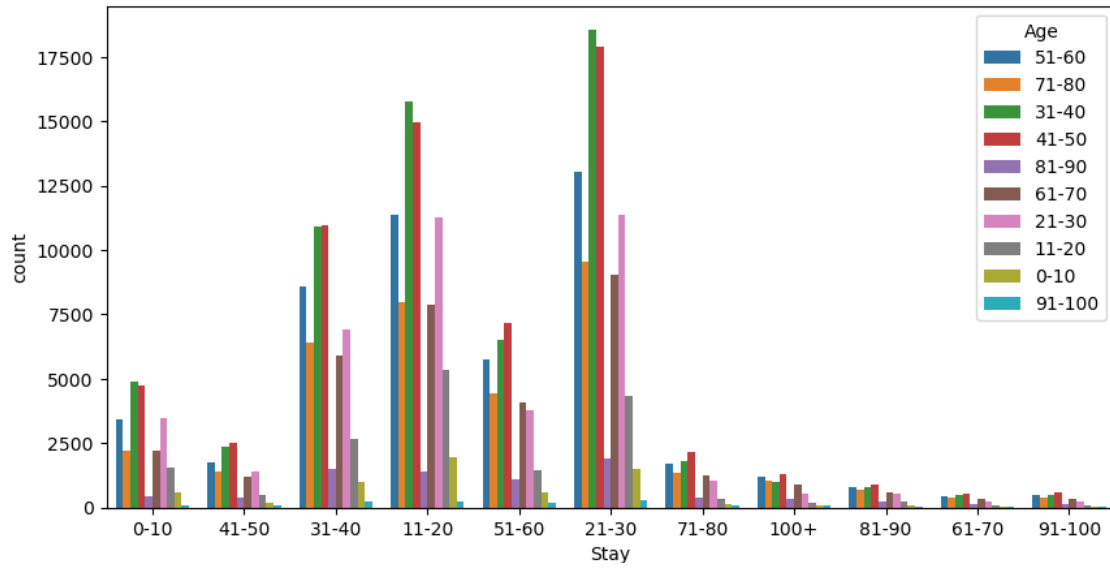
```
for i in [ 'Department', 'Ward_Type' , 'Ward_Facility_Code', 'Age',
'Type of Admission', 'Severity of Illness',
'Bed Grade', 'Hospital_region_code', 'Hospital_type_code' ]:
```

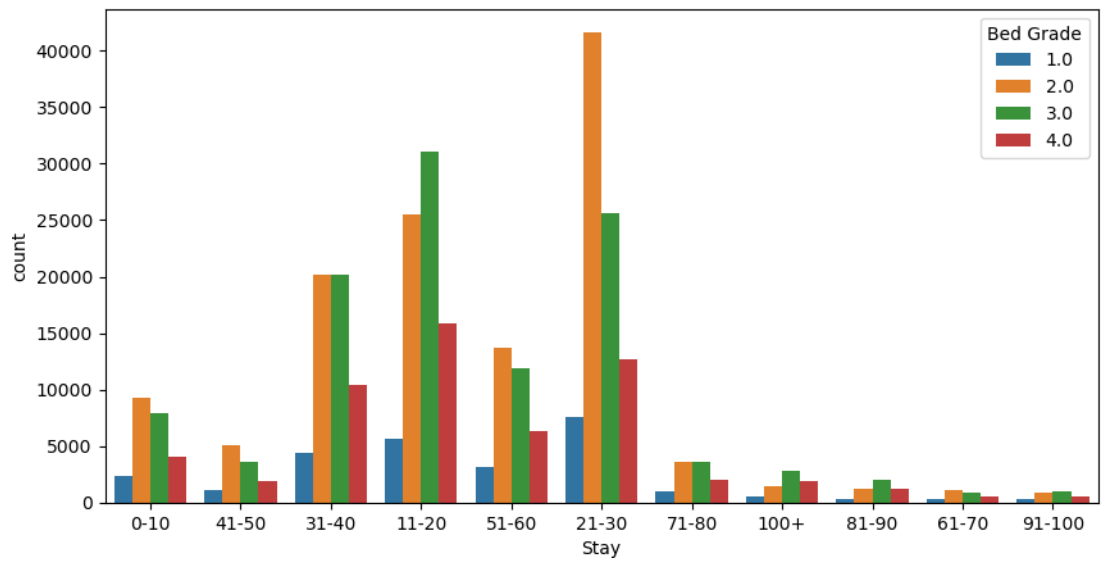
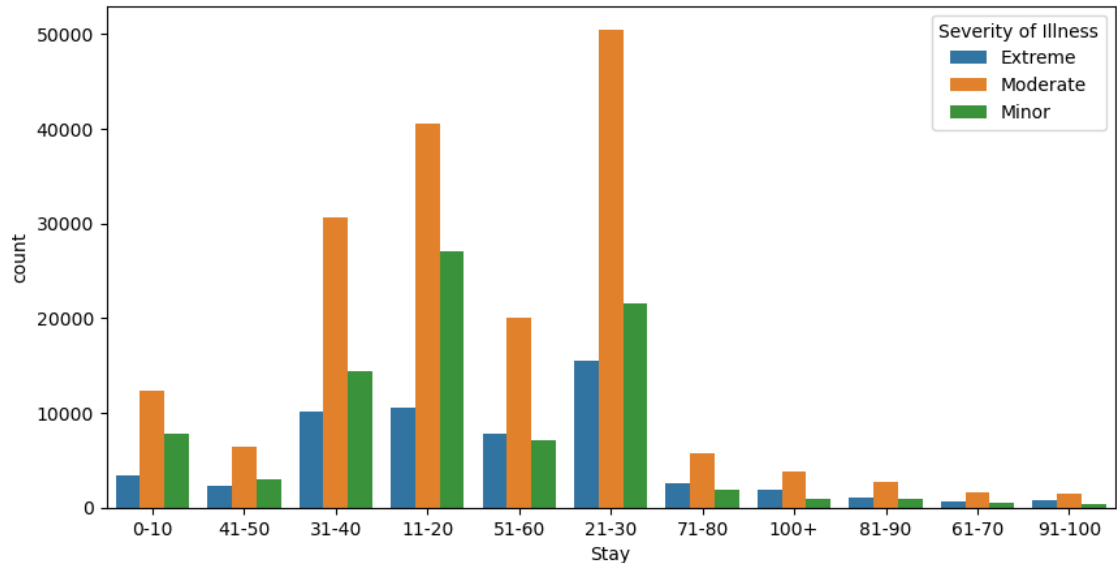
```
plt.figure(figsize=(10,5))
```

```
sns.countplot(x='Stay',hue=i,data=train)
```

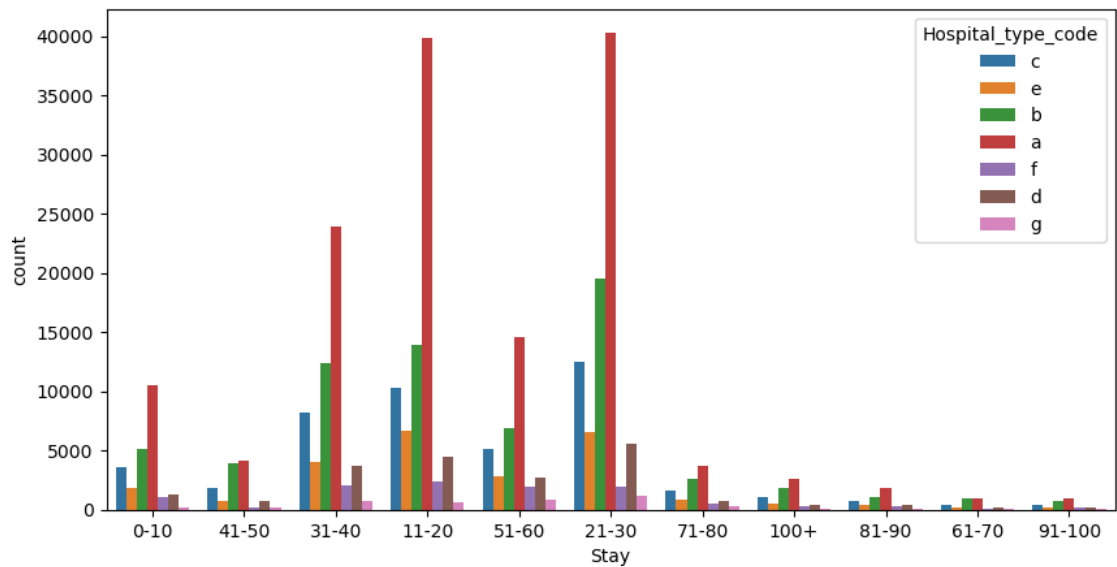
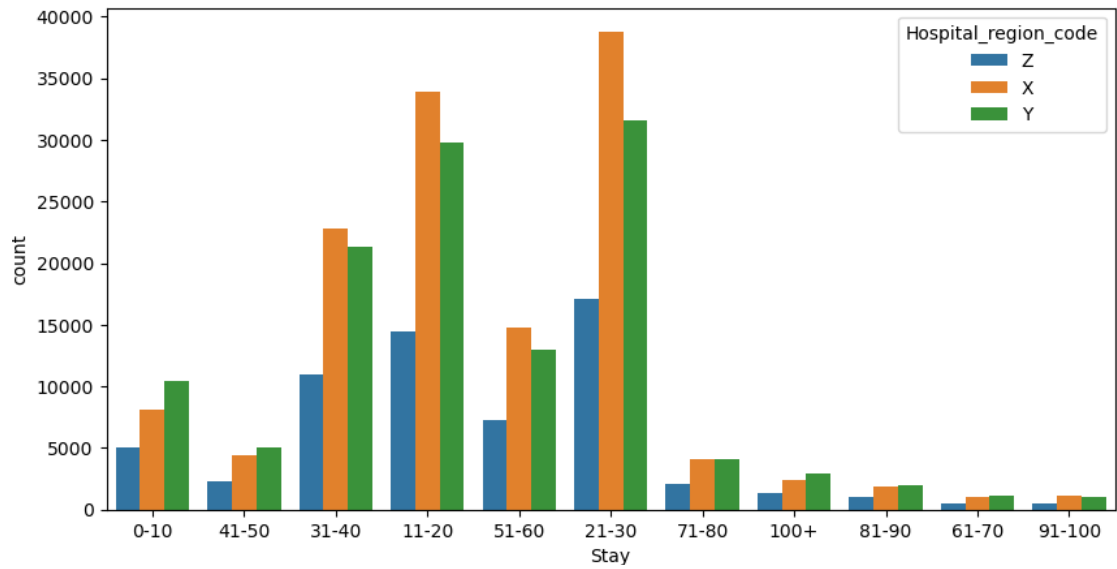






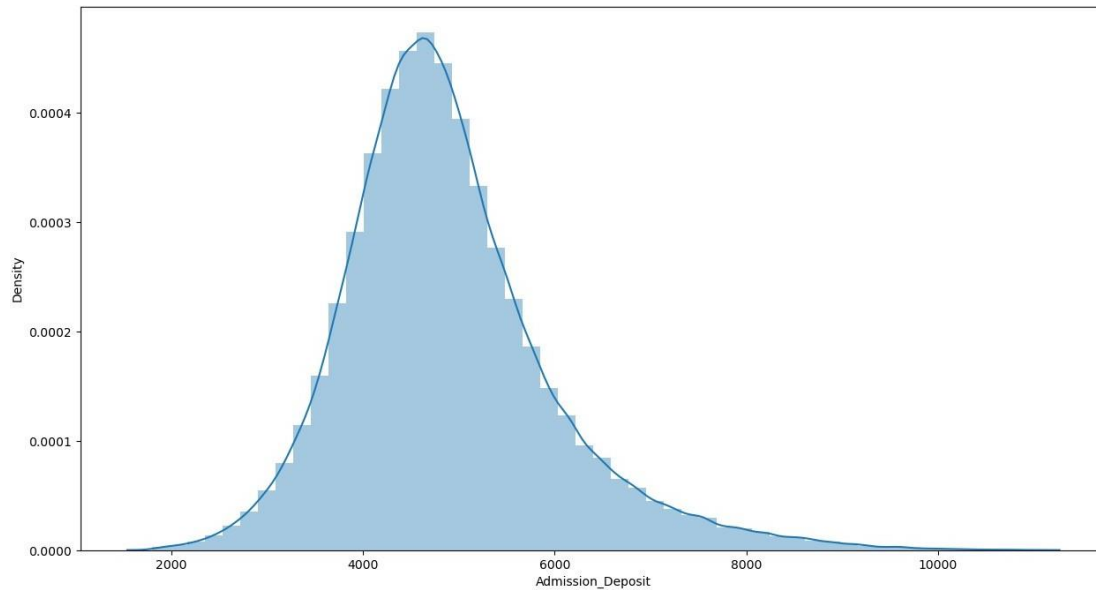






```
sns.distplot( train['Admission_Deposit'])
```

```
<AxesSubplot:xlabel='Admission_Deposit', ylabel='Density'>
```



```
Total = train.isnull().sum().sort_values(ascending=False)
```

```
Percent =
(train.isnull().sum()*100/train.isnull().count()).sort_values(ascending=False)
```

```
missing_data = pd.concat([Total, Percent], axis = 1, keys = ['Total',
'Percentage of Missing Values'])
missing_data
```

	Total	Percentage of Missing Values
City_Code_Patient	4532	1.423197
Bed Grade	113	0.035486
Hospital_code	0	0.000000
Hospital_type_code	0	0.000000
City_Code_Hospital	0	0.000000
Hospital_region_code	0	0.000000
Available Extra Rooms in Hospital	0	0.000000
Department	0	0.000000
Ward_Type	0	0.000000
Ward_Facility_Code	0	0.000000
Type of Admission	0	0.000000
Severity of Illness	0	0.000000
Visitors with Patient	0	0.000000
Age	0	0.000000
Admission_Deposit	0	0.000000
Stay	0	0.000000

```
Total = train.isnull().sum().sort_values(ascending=False)
```

```
Percent =
(train.isnull().sum()*100/train.isnull().count()).sort_values(ascending=False)
```

```
g=False)
```

```
missing_data = pd.concat([Total, Percent], axis = 1, keys = ['Total',  
'Percentage of Missing Values'])
```

```
missing_data
```

	Total	Percentage of Missing Values
City_Code_Patient	4532	1.423197
Bed Grade	113	0.035486
Hospital_code	0	0.000000
Hospital_type_code	0	0.000000
City_Code_Hospital	0	0.000000
Hospital_region_code	0	0.000000
Available Extra Rooms in Hospital	0	0.000000
Department	0	0.000000
Ward_Type	0	0.000000
Ward_Facility_Code	0	0.000000
Type of Admission	0	0.000000
Severity of Illness	0	0.000000
Visitors with Patient	0	0.000000
Age	0	0.000000
Admission_Deposit	0	0.000000
Stay	0	0.000000

```
Total = test.isnull().sum().sort_values(ascending=False)
```

```
Percent =
```

```
(test.isnull().sum()*100/test.isnull().count()).sort_values(ascending=  
False)
```

```
missing_data = pd.concat([Total, Percent], axis = 1, keys = ['Total',  
'Percentage of Missing Values'])
```

```
missing_data
```

	Total	Percentage of Missing Values
City_Code_Patient	2157	1.573798
Bed Grade	35	0.025537
Hospital_code	0	0.000000
Hospital_type_code	0	0.000000
City_Code_Hospital	0	0.000000
Hospital_region_code	0	0.000000
Available Extra Rooms in Hospital	0	0.000000
Department	0	0.000000
Ward_Type	0	0.000000
Ward_Facility_Code	0	0.000000
Type of Admission	0	0.000000
Severity of Illness	0	0.000000
Visitors with Patient	0	0.000000
Age	0	0.000000
Admission_Deposit	0	0.000000

```

train.dropna(subset = ['Bed Grade', 'City_Code_Patient'], inplace =
True)

test['Bed Grade'] = test['Bed Grade'].fillna(test['Bed Grade'].mode()
[0], inplace = True)
test['City_Code_Patient'] =
test['City_Code_Patient'].fillna(test['City_Code_Patient'].mode()[0],
inplace = True)

```

```

df_num_train = train.select_dtypes([np.number])
df_num_train.head()

```

```

Visitors with Patient  Admission_Deposit
0                      2          4911.000000
1                      2          5954.000000
2                      2          4745.000000
3                      2          7272.000000
4                      2          5558.000000

```

```

df_cat_train = train.select_dtypes([np.object])
df_cat_train.head()

```

```

Hospital_code Hospital_type_code City_Code_Hospital
Hospital_region_code \
0                8                c                3
Z
1                2                c                5
Z
2               10                e                1
X
3               26                b                2
Y
4               26                b                2
Y

```

```

Available Extra Rooms in Hospital  Department Ward_Type \
0                3  radiotherapy          R
1                2  radiotherapy          S
2                2    anesthesia          S
3                2  radiotherapy          R
4                2  radiotherapy          S

```

```

Ward_Facility_Code Bed Grade City_Code_Patient Type of Admission \
0                F  2.000000          7.000000      Emergency
1                F  2.000000          7.000000          Trauma
2                E  2.000000          7.000000          Trauma
3                D  2.000000          7.000000          Trauma
4                D  2.000000          7.000000          Trauma

```

```

Severity of Illness  Age  Stay
0          Extreme  51-60  0-10

```

1	Extreme	51-60	41-50
2	Extreme	51-60	31-40
3	Extreme	51-60	41-50
4	Extreme	51-60	41-50

```
df_num_test = test.select_dtypes([np.number])
df_num_test.head()
```

	Visitors with Patient	Admission_Deposit
0	2	3095.000000
1	4	4018.000000
2	3	4492.000000
3	3	4173.000000
4	4	4161.000000

```
df_cat_test = test.select_dtypes([np.object])
df_cat_test.head()
```

	Hospital_code	Hospital_type_code	City_Code_Hospital
	Hospital_region_code \		
0	21	c	3
Z			
1	29	a	4
X			
2	26	b	2
Y			
3	6	a	6
X			
4	28	b	11
X			

	Available Extra Rooms in Hospital	Department	Ward_Type
	Ward_Facility_Code \		
0	3	gynecology	S
A			
1	2	gynecology	S
F			
2	3	gynecology	Q
D			
3	3	gynecology	Q
F			
4	2	gynecology	R
F			

	Bed Grade	City_Code_Patient	Type of Admission	Severity of Illness
	Age			
0	None	None	Emergency	Moderate
71-80				
1	None	None	Trauma	Moderate
71-80				
2	None	None	Emergency	Moderate

71-80				
3	None	None	Trauma	Moderate
71-80				
4	None	None	Trauma	Moderate
71-80				

```
admission_encode = {'Trauma': 1, 'Urgent': 2, 'Emergency' : 3 }
severity_encode   = {'Minor': 1, 'Moderate': 2, 'Extreme': 3 }
```

```
df_cat_train['Type of Admission'] = df_cat_train['Type of
Admission'].map (admission_encode)
df_cat_train['Severity of Illness'] = df_cat_train['Severity of
Illness'].map (severity_encode)
```

```
df_cat_test['Type of Admission'] = df_cat_test['Type of
Admission'].map (admission_encode)
df_cat_test['Severity of Illness'] = df_cat_test['Severity of
Illness'].map (severity_encode)
```

```
df_cat_train['Stay']= df_cat_train['Stay'].replace({'0-10':1, '11-
20':2, '21-30':3, '31-40':4, '41-50':5, '51-60':6,'61-70':7,
                                                    '71-80':8,'81-90':9,'91-
100':10,'100+':11})
```

```
df_cat_train['Age']= df_cat_train['Age'].replace({'0-10':1, '11-20':2,
'21-30':3, '31-40':4, '41-50':5, '51-60':6,'61-70':7,
                                                    '71-80':8,'81-90':9,'91-
100':10})
```

```
df_cat_test['Age']= df_cat_test['Age'].replace({'0-10':1, '11-20':2,
'21-30':3, '31-40':4, '41-50':5, '51-60':6,'61-70':7,
                                                    '71-80':8,'81-90':9,'91-
100':10})
```

```
df_cat_train['Stay']=df_cat_train['Stay'].astype(int)
```

```
from sklearn.preprocessing import LabelEncoder
LE=LabelEncoder()
```

```
df_cat_train['Hospital_code']=LE.fit_transform(df_cat_train['Hospital_
code'])
df_cat_train['Hospital_type_code']=LE.fit_transform(df_cat_train['Hosp
ital_type_code'])
df_cat_train['City_Code_Hospital']=LE.fit_transform(df_cat_train['City
_Code_Hospital'])
df_cat_train['Hospital_region_code']=LE.fit_transform(df_cat_train['Ho
spital_region_code'])
df_cat_train['Department']=LE.fit_transform(df_cat_train['Department']
)
```

```

df_cat_train['Ward_Type']=LE.fit_transform(df_cat_train['Ward_Type'])
df_cat_train['Ward_Facility_Code']=LE.fit_transform(df_cat_train['Ward_Facility_Code'])
df_cat_train['City_Code_Patient']=LE.fit_transform(df_cat_train['City_Code_Patient'])
df_cat_train['Bed Grade']=LE.fit_transform(df_cat_train['Bed Grade'])

```

```
df_cat_train.head()
```

	Hospital_code	Hospital_type_code	City_Code_Hospital	\
0	7	2	2	
1	1	2	4	
2	9	4	0	
3	25	1	1	
4	25	1	1	

	Hospital_region_code	Available Extra Rooms in Hospital	Department
0	2	3	3
1	2	2	3
2	0	2	1
3	1	2	3
4	1	2	3

	Ward_Type	Ward_Facility_Code	Bed Grade	City_Code_Patient	\
0	2	5	1	6	
1	3	5	1	6	
2	3	4	1	6	
3	2	3	1	6	
4	3	3	1	6	

	Type of Admission	Severity of Illness	Age	Stay
0	3	3	6	1
1	1	3	6	5
2	1	3	6	4
3	1	3	6	5
4	1	3	6	5

```

from sklearn.preprocessing import LabelEncoder
LE=LabelEncoder()

```

```

df_cat_test['Hospital_code']=LE.fit_transform(df_cat_test['Hospital_code'])
df_cat_test['Hospital_type_code']=LE.fit_transform(df_cat_test['Hospital_type_code'])

```

```

df_cat_test['City_Code_Hospital']=LE.fit_transform(df_cat_test['City_C
ode_Hospital'])
df_cat_test['Hospital_region_code']=LE.fit_transform(df_cat_test['Hosp
ital_region_code'])
df_cat_test['Department']=LE.fit_transform(df_cat_test['Department'])
df_cat_test['Ward_Type']=LE.fit_transform(df_cat_test['Ward_Type'])
df_cat_test['Ward_Facility_Code']=LE.fit_transform(df_cat_test['Ward_F
acility_Code'])
df_cat_test['City_Code_Patient']=LE.fit_transform(df_cat_test['City_Co
de_Patient'])
df_cat_test['Bed Grade']=LE.fit_transform(df_cat_test['Bed Grade'])

```

```
df_cat_test.head()
```

	Hospital_code	Hospital_type_code	City_Code_Hospital	\
0	20	2	2	
1	28	0	3	
2	25	1	1	
3	5	0	5	
4	27	1	9	

	Hospital_region_code	Available Extra Rooms in Hospital	Department
0	2	3	2
1	0	2	2
2	1	3	2
3	0	3	2
4	0	2	2

	Ward_Type	Ward_Facility_Code	Bed Grade	City_Code_Patient	\
0	3	0	0	0	
1	3	5	0	0	
2	1	3	0	0	
3	1	5	0	0	
4	2	5	0	0	

	Type of Admission	Severity of Illness	Age
0	3	2	8
1	1	2	8
2	3	2	8
3	1	2	8
4	1	2	8

```

from sklearn.preprocessing import StandardScaler
sc = StandardScaler()

```



```

num_scaled = sc.fit_transform(df_num_train)

df_num_scaled = pd.DataFrame(num_scaled, columns =
df_num_train.columns)

num_scaled_test = sc.fit_transform(df_num_test)

df_num_scaled_test = pd.DataFrame(num_scaled_test, columns =
df_num_test.columns)

df_num_scaled.shape

(313793, 2)

df_cat_train = df_cat_train.reset_index(drop=True)
df_num_scaled = df_num_scaled.reset_index(drop=True)
df_cat_test = df_cat_test.reset_index(drop=True)
df_num_scaled_test = df_num_scaled_test.reset_index(drop=True)

df_cat_train.shape

(313793, 14)

df_full = pd.concat([df_num_scaled, df_cat_train],axis=1)
df_full_test = pd.concat([df_num_scaled_test, df_cat_test],axis=1)

df_full.shape

(313793, 16)

df_full.head()

```

	Visitors with Patient	Admission_Deposit	Hospital_code \
0	-0.727035	0.026796	7
1	-0.727035	0.986987	1
2	-0.727035	-0.126025	9
3	-0.727035	2.200344	25
4	-0.727035	0.622427	25

	Hospital_type_code	City_Code_Hospital	Hospital_region_code \
0	2	2	2
1	2	4	2
2	4	0	0
3	1	1	1
4	1	1	1

	Available Extra Rooms in Hospital	Department	Ward_Type \
0	3	3	2
1	2	3	3
2	2	1	3

3			2		3		2
4			2		3		3

	Ward_Facility_Code	Bed Grade	City_Code_Patient	Type of Admission
0	5	1		6
1	5	1		6
2	4	1		6
3	3	1		6
4	3	1		6

	Severity of Illness	Age	Stay
0	3	6	1
1	3	6	5
2	3	6	4
3	3	6	5
4	3	6	5

```
sns.heatmap(df_full.corr(), annot = True)
```

```
<AxesSubplot:>
```



```

X = df_full.drop('Stay',axis=1)
y = df_full['Stay']

X = sm.add_constant(X)

X_train, X_test, y_train, y_test = train_test_split(X, y, random_state
= 10, test_size = 0.3)

print('X_train', X_train.shape)
print('y_train', y_train.shape)

print('X_test', X_test.shape)
print('y_test', y_test.shape)

X_train (219655, 16)
y_train (219655,)
X_test (94138, 16)
y_test (94138,)

from sklearn.model_selection import KFold,cross_val_score
kfold=KFold(n_splits=10, shuffle=True, random_state=10)

LR = LogisticRegression()

LR.fit(X_train,y_train)

y_pred_LR=LR.predict(X_test)
accuracy_score(y_test,y_pred_LR)*100

37.94217000573626

print(classification_report(y_test,y_pred_LR))

```

	precision	recall	f1-score	support
1	0.48	0.02	0.03	6901
2	0.37	0.45	0.41	23205
3	0.40	0.64	0.49	25792
4	0.32	0.19	0.24	16289
5	0.00	0.00	0.00	3439
6	0.37	0.46	0.41	10470
7	0.00	0.00	0.00	822
8	0.00	0.00	0.00	3093
9	0.11	0.00	0.01	1412
10	0.00	0.00	0.00	782
11	0.49	0.32	0.39	1933
accuracy			0.38	94138
macro avg	0.23	0.19	0.18	94138
weighted avg	0.34	0.38	0.33	94138

```
decision_tree_classification = DecisionTreeClassifier(criterion =  
'entropy', random_state = 10)
```

```
decision_tree = decision_tree_classification.fit(X_train, y_train)
```

```
y_pred_DT=decision_tree.predict(X_test)
```

```
accuracy_score(y_test,y_pred_DT)*100
```

29.678769466102956

```
print(classification_report(y_test,y_pred_DT))
```

	precision	recall	f1-score	support
1	0.19	0.19	0.19	6901
2	0.34	0.34	0.34	23205
3	0.39	0.38	0.38	25792
4	0.25	0.25	0.25	16289
5	0.06	0.07	0.07	3439
6	0.31	0.30	0.31	10470
7	0.03	0.04	0.03	822
8	0.15	0.15	0.15	3093
9	0.22	0.23	0.22	1412
10	0.08	0.10	0.09	782
11	0.32	0.35	0.33	1933
accuracy			0.30	94138
macro avg	0.21	0.22	0.21	94138
weighted avg	0.30	0.30	0.30	94138

```
dt_tuned = DecisionTreeClassifier(criterion = 'gini', max_depth=11,  
random_state = 10)
```

```
decision_tree_tuned = dt_tuned.fit(X_train, y_train)
```

```
y_pred_DT_tuned = decision_tree_tuned.predict(X_test)
```

```
accuracy_score(y_test,y_pred_DT_tuned)*100
```

40.82198474579872

```
print(classification_report(y_test,y_pred_DT_tuned))
```

	precision	recall	f1-score	support
1	0.36	0.13	0.19	6901
2	0.41	0.46	0.44	23205
3	0.41	0.67	0.51	25792
4	0.39	0.23	0.29	16289
5	0.07	0.00	0.00	3439
6	0.41	0.45	0.43	10470
7	0.00	0.00	0.00	822

8	0.27	0.03	0.05	3093
9	0.35	0.21	0.27	1412
10	0.20	0.03	0.05	782
11	0.52	0.36	0.43	1933
accuracy			0.41	94138
macro avg	0.31	0.23	0.24	94138
weighted avg	0.38	0.41	0.37	94138

```
rf_classification = RandomForestClassifier(random_state = 10)
```

```
rf_model = rf_classification.fit(X_train, y_train)
```

```
y_pred_RF = rf_model.predict(X_test)
```

```
accuracy_score(y_test,y_pred_RF)*100
```

```
38.315026875438186
```

```
print(classification_report(y_test,y_pred_RF))
```

	precision	recall	f1-score	support
1	0.30	0.19	0.23	6901
2	0.39	0.44	0.42	23205
3	0.41	0.54	0.47	25792
4	0.33	0.27	0.29	16289
5	0.09	0.02	0.04	3439
6	0.40	0.44	0.42	10470
7	0.10	0.02	0.03	822
8	0.28	0.10	0.15	3093
9	0.37	0.22	0.28	1412
10	0.24	0.06	0.09	782
11	0.52	0.44	0.48	1933
accuracy			0.38	94138
macro avg	0.31	0.25	0.26	94138
weighted avg	0.36	0.38	0.37	94138

```
rf_classification_tuned = RandomForestClassifier(criterion = 'gini',
n_estimators = 47, random_state = 10)
```

```
rf_model_tuned = rf_classification_tuned.fit(X_train, y_train)
```

```
y_pred_RF_tuned = rf_model_tuned.predict(X_test)
```

```
accuracy_score(y_test,y_pred_RF_tuned)*100
```

```
37.94323227602031
```

```
print(classification_report(y_test,y_pred_RF_tuned))
```

	precision	recall	f1-score	support
1	0.29	0.19	0.23	6901
2	0.38	0.44	0.41	23205
3	0.41	0.53	0.46	25792
4	0.32	0.27	0.30	16289
5	0.09	0.02	0.04	3439
6	0.39	0.43	0.41	10470
7	0.10	0.02	0.03	822
8	0.28	0.11	0.15	3093
9	0.36	0.22	0.27	1412
10	0.22	0.05	0.08	782
11	0.52	0.43	0.47	1933
accuracy			0.38	94138
macro avg	0.31	0.25	0.26	94138
weighted avg	0.36	0.38	0.36	94138

```

dt_tuned = DecisionTreeClassifier(criterion = 'gini', max_depth=11,
random_state = 10)
ada_model_DT = AdaBoostClassifier(base_estimator=dt_tuned,
random_state = 10)

ada_model_DT.fit(X_train, y_train)

AdaBoostClassifier(base_estimator=DecisionTreeClassifier(max_depth=11,

random_state=10),

random_state=10)

y_pred_ada_model_DT = ada_model_DT.predict(X_test)
accuracy_score(y_test,y_pred_ada_model_DT)*100

29.91990482058255

print(classification_report(y_test,y_pred_ada_model_DT))

```

	precision	recall	f1-score	support
1	0.19	0.16	0.17	6901
2	0.33	0.34	0.33	23205
3	0.38	0.39	0.39	25792
4	0.22	0.27	0.24	16289
5	0.06	0.04	0.05	3439
6	0.31	0.36	0.33	10470
7	0.00	0.00	0.00	822
8	0.14	0.10	0.11	3093
9	0.22	0.10	0.13	1412
10	0.10	0.01	0.02	782
11	0.45	0.20	0.27	1933

accuracy			0.30	94138
macro avg	0.22	0.18	0.19	94138
weighted avg	0.29	0.30	0.29	94138

```
ada_model_DT_tuned = AdaBoostClassifier(base_estimator=dt_tuned,
n_estimators = 1, random_state = 10)
```

```
ada_model_DT_tuned.fit(X_train, y_train)
```

```
AdaBoostClassifier(base_estimator=DecisionTreeClassifier(max_depth=11,
random_state=10),
n_estimators=1, random_state=10)
```

```
y_pred_ada_model_DT_tuned = ada_model_DT.predict(X_test)
accuracy_score(y_test,y_pred_ada_model_DT_tuned)*100
```

29.91990482058255

```
print(classification_report(y_test,y_pred_ada_model_DT_tuned))
```

	precision	recall	f1-score	support
1	0.19	0.16	0.17	6901
2	0.33	0.34	0.33	23205
3	0.38	0.39	0.39	25792
4	0.22	0.27	0.24	16289
5	0.06	0.04	0.05	3439
6	0.31	0.36	0.33	10470
7	0.00	0.00	0.00	822
8	0.14	0.10	0.11	3093
9	0.22	0.10	0.13	1412
10	0.10	0.01	0.02	782
11	0.45	0.20	0.27	1933

accuracy			0.30	94138
macro avg	0.22	0.18	0.19	94138
weighted avg	0.29	0.30	0.29	94138

```
rf_classification_tuned = RandomForestClassifier(criterion = 'gini',
n_estimators = 47, random_state = 10)
```

```
ada_model_rf =
AdaBoostClassifier(base_estimator=rf_classification_tuned,
n_estimators=1, random_state = 10)
```

```
ada_model_rf.fit(X_train, y_train)
```

```
AdaBoostClassifier(base_estimator=RandomForestClassifier(n_estimators=
47,
```

```
random_state=10),
                    n_estimators=1, random_state=10)
```

```
y_pred_ada_model_RF = ada_model_rf.predict(X_test)
accuracy_score(y_test,y_pred_ada_model_RF)*100
```

```
38.081327412946955
```

```
print(classification_report(y_test,y_pred_ada_model_RF))
```

	precision	recall	f1-score	support
1	0.29	0.20	0.23	6901
2	0.39	0.44	0.41	23205
3	0.41	0.53	0.46	25792
4	0.32	0.27	0.30	16289
5	0.10	0.03	0.04	3439
6	0.40	0.43	0.41	10470
7	0.12	0.02	0.03	822
8	0.26	0.10	0.15	3093
9	0.39	0.24	0.29	1412
10	0.23	0.05	0.09	782
11	0.51	0.44	0.47	1933
accuracy			0.38	94138
macro avg	0.31	0.25	0.26	94138
weighted avg	0.36	0.38	0.36	94138

```
ada_model_rf_tuned =
AdaBoostClassifier(base_estimator=rf_classification_tuned,
n_estimators = 4, random_state = 10)
```

```
ada_model_rf_tuned.fit(X_train, y_train)
```

```
AdaBoostClassifier(base_estimator=RandomForestClassifier(n_estimators=
47,
```

```
random_state=10),
                    n_estimators=4, random_state=10)
```

```
from sklearn.ensemble import GradientBoostingClassifier
GB=GradientBoostingClassifier(random_state=10)
GB.fit(X_train, y_train)
```

```
GradientBoostingClassifier(random_state=10)
```

```
y_pred_GB = GB.predict(X_test)
accuracy_score(y_test,y_pred_GB)*100
```



41.5475153498056

```
print(classification_report(y_test,y_pred_GB))
```

	precision	recall	f1-score	support
1	0.41	0.12	0.19	6901
2	0.42	0.51	0.46	23205
3	0.42	0.66	0.52	25792
4	0.41	0.17	0.24	16289
5	0.14	0.00	0.00	3439
6	0.39	0.53	0.45	10470
7	0.00	0.00	0.00	822
8	0.30	0.01	0.02	3093
9	0.31	0.20	0.24	1412
10	0.17	0.01	0.01	782
11	0.52	0.40	0.45	1933
accuracy			0.42	94138
macro avg	0.32	0.24	0.23	94138
weighted avg	0.39	0.42	0.37	94138

```
GB_tuned=GradientBoostingClassifier(n_estimators=29, random_state=10)
GB_tuned.fit(X_train, y_train)
```

```
GradientBoostingClassifier(n_estimators=29, random_state=10)
```

```
from sklearn.naive_bayes import GaussianNB
```

```
NB = GaussianNB()
```

```
NB.fit(X_train,y_train)
```

```
GaussianNB()
```

```
y_pred_NB = NB.predict(X_test)
```

```
accuracy_score(y_test,y_pred_NB)*100
```

36.37850814761308

```
print(classification_report(y_test,y_pred_NB))
```

	precision	recall	f1-score	support
1	0.30	0.09	0.14	6901
2	0.36	0.41	0.39	23205
3	0.39	0.65	0.49	25792
4	0.32	0.15	0.21	16289
5	0.08	0.01	0.01	3439
6	0.33	0.38	0.36	10470
7	0.04	0.00	0.00	822
8	0.10	0.01	0.02	3093
9	0.12	0.02	0.04	1412

10	0.50	0.00	0.00	782
11	0.42	0.37	0.39	1933
accuracy			0.36	94138
macro avg	0.27	0.19	0.18	94138
weighted avg	0.33	0.36	0.32	94138

```
from sklearn.neighbors import KNeighborsClassifier
```

```
knn=KNeighborsClassifier(n_neighbors=565,weights='distance')
knn.fit(X_train,y_train)
```

```
KNeighborsClassifier(n_neighbors=565, weights='distance')
```

```
y_pred_KNN = NB.predict(X_test)
accuracy_score(y_test,y_pred_KNN)*100
```

```
36.37850814761308
```

```
print(classification_report(y_test,y_pred_NB))
```

	precision	recall	f1-score	support
1	0.30	0.09	0.14	6901
2	0.36	0.41	0.39	23205
3	0.39	0.65	0.49	25792
4	0.32	0.15	0.21	16289
5	0.08	0.01	0.01	3439
6	0.33	0.38	0.36	10470
7	0.04	0.00	0.00	822
8	0.10	0.01	0.02	3093
9	0.12	0.02	0.04	1412
10	0.50	0.00	0.00	782
11	0.42	0.37	0.39	1933
accuracy			0.36	94138
macro avg	0.27	0.19	0.18	94138
weighted avg	0.33	0.36	0.32	94138

```
from catboost import CatBoostClassifier
```

```
cb =
CatBoostClassifier(random_state=10,use_best_model=True,iterations=1000
)
cb.fit(X_train,y_train,use_best_model=True,verbose=100,eval_set=(X_test,y_test))
```

```
Learning rate set to 0.120271
```

```
0: learn: 2.1972797 test: 2.1978440 best: 2.1978440 (0) total:
```

```

532ms remaining: 8m 50s
100: learn: 1.5115382 test: 1.5242638 best: 1.5242638 (100) total:
37.5s remaining: 5m 34s
200: learn: 1.4818453 test: 1.5070444 best: 1.5070444 (200) total: 1m
17s remaining: 5m 9s
300: learn: 1.4638516 test: 1.5016071 best: 1.5016071 (300) total: 1m
57s remaining: 4m 32s
400: learn: 1.4484799 test: 1.4983697 best: 1.4983697 (400) total: 2m
35s remaining: 3m 52s
500: learn: 1.4345747 test: 1.4970449 best: 1.4970449 (500) total: 3m
14s remaining: 3m 13s
600: learn: 1.4237650 test: 1.4965192 best: 1.4964737 (579) total: 3m
56s remaining: 2m 37s
700: learn: 1.4123374 test: 1.4963260 best: 1.4961744 (652) total: 4m
34s remaining: 1m 57s
800: learn: 1.4018164 test: 1.4964416 best: 1.4961744 (652) total: 5m
13s remaining: 1m 17s
900: learn: 1.3915056 test: 1.4967396 best: 1.4961744 (652) total: 5m
52s remaining: 38.8s
999: learn: 1.3815565 test: 1.4971529 best: 1.4961744 (652) total: 6m
37s remaining: 0us

```

```

bestTest = 1.496174357
bestIteration = 652

```

Shrink model to first 653 iterations.

```
<catboost.core.CatBoostClassifier at 0x1f1be732bb0>
```

```

cb_pred = cb.predict(X_test)
accuracy_score(y_test,cb_pred)*100

```

```
42.54180033567741
```

```
print(classification_report(y_test,cb_pred))
```

	precision	recall	f1-score	support
1	0.41	0.16	0.23	6901
2	0.43	0.51	0.47	23205
3	0.43	0.66	0.52	25792
4	0.41	0.24	0.30	16289
5	0.24	0.00	0.01	3439
6	0.41	0.48	0.44	10470
7	0.12	0.00	0.00	822
8	0.42	0.03	0.05	3093
9	0.36	0.21	0.27	1412
10	0.29	0.01	0.02	782
11	0.52	0.43	0.47	1933
accuracy			0.43	94138

macro avg	0.37	0.25	0.25	94138
weighted avg	0.41	0.43	0.39	94138

```
cb_pred_train= cb.predict(X_train)
accuracy_score(y_train,cb_pred_train)*100
```

```
45.57692745441715
```

```
print(classification_report(y_train,cb_pred_train))
```

	precision	recall	f1-score	support
1	0.46	0.19	0.27	16349
2	0.45	0.54	0.49	53890
3	0.45	0.68	0.54	60524
4	0.47	0.26	0.34	38023
5	0.63	0.01	0.02	8102
6	0.45	0.53	0.48	23993
7	0.79	0.01	0.02	1876
8	0.71	0.06	0.10	7003
9	0.55	0.31	0.40	3349
10	0.84	0.07	0.12	1931
11	0.62	0.51	0.56	4615
accuracy			0.46	219655
macro avg	0.58	0.29	0.30	219655
weighted avg	0.48	0.46	0.42	219655

```
ls = df_full_test.columns.tolist()
```

```
in_data = df_full_test[ls]
```

```
out_data = cb.predict (in_data)
```

```
test = pd.read_csv("D:/HealthCare/test_data.csv")
submit = pd.DataFrame()
```

```
submit ['case_id'] = test['case_id']
submit ['Stay'] = out_data
```

```
stay_decode = { 1 : '0-10', 2 : '11-20', 3 : '21-30', 4 :
'31-40', 5 : '41-50', 6 : '51-60', 7 : '61-70', 8 : '71-80', 9 : '81-
90',
10 : '91-100', 11 : 'More than 100 Days' }
```

```
submit ['Stay'] = submit ['Stay'].map(stay_decode)
```

```
submit.head(15)
```

	case_id	Stay
0	318439	0-10
1	318440	51-60
2	318441	21-30
3	318442	21-30
4	318443	51-60
5	318444	21-30
6	318445	21-30
7	318446	21-30
8	318447	21-30
9	318448	21-30
10	318449	21-30
11	318450	51-60
12	318451	21-30
13	318452	21-30
14	318453	31-40

```
count = submit ['Stay'].value_counts()
count
```

21-30	85076
51-60	23685
11-20	12285
31-40	6693
0-10	5242
More than 100 Days	2407
81-90	1329
71-80	174
41-50	127
91-100	32
61-70	7

Name: Stay, dtype: int64

```
sns.barplot(x=count.index.values, y=count.values,data=submit)
```

<AxesSubplot:>

