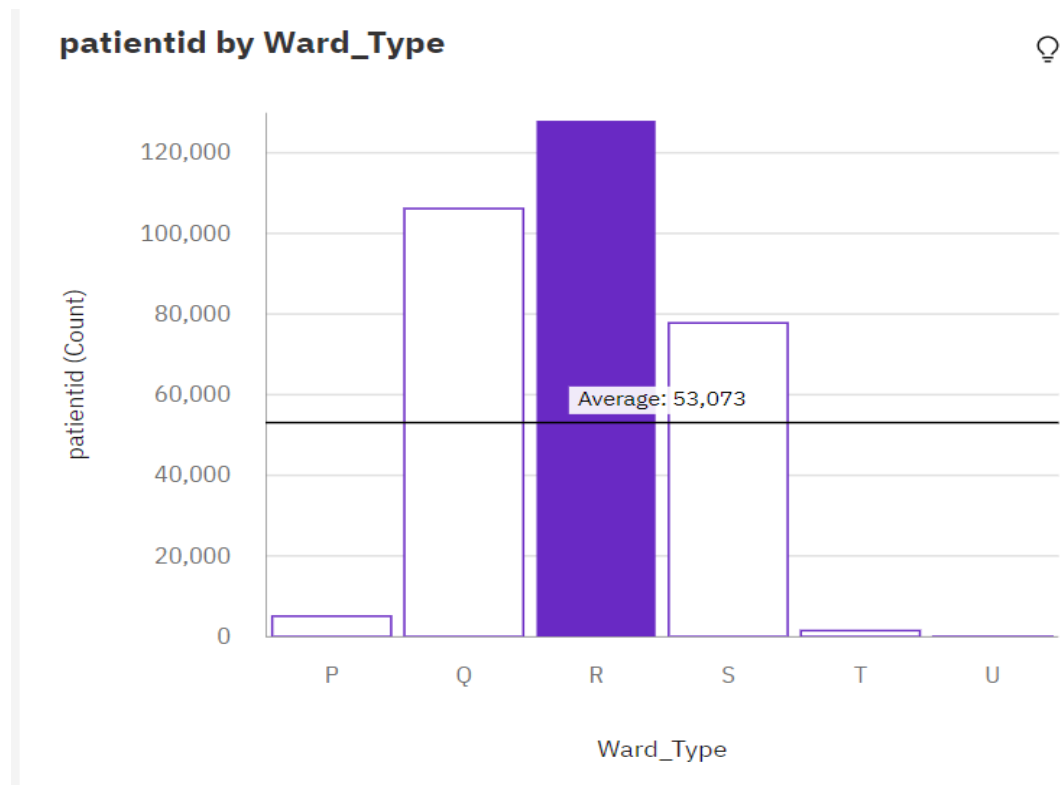


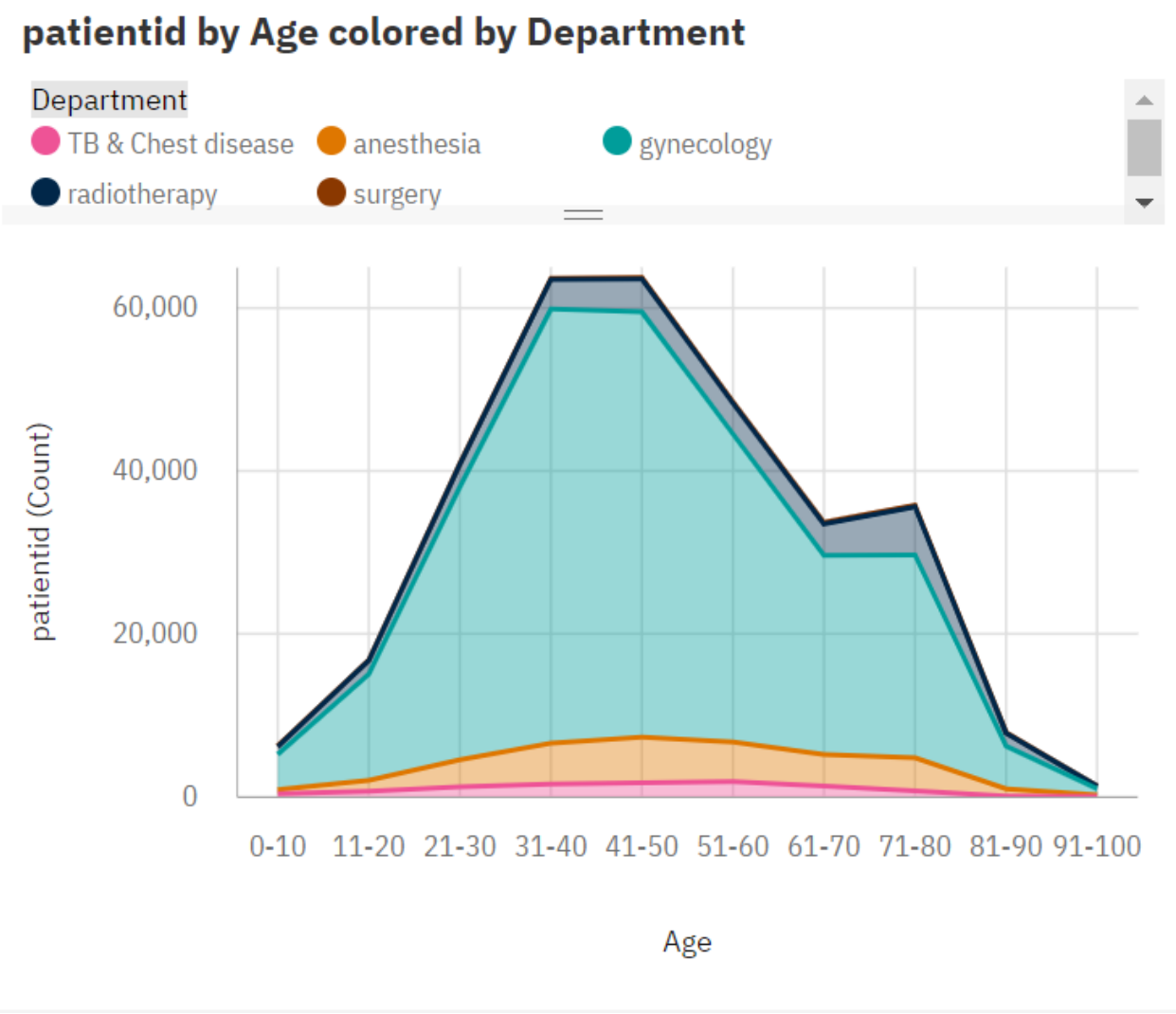
## Sprint -1

### Data Visualization

#### No of Patients in each ward



No of patients by age colored by department



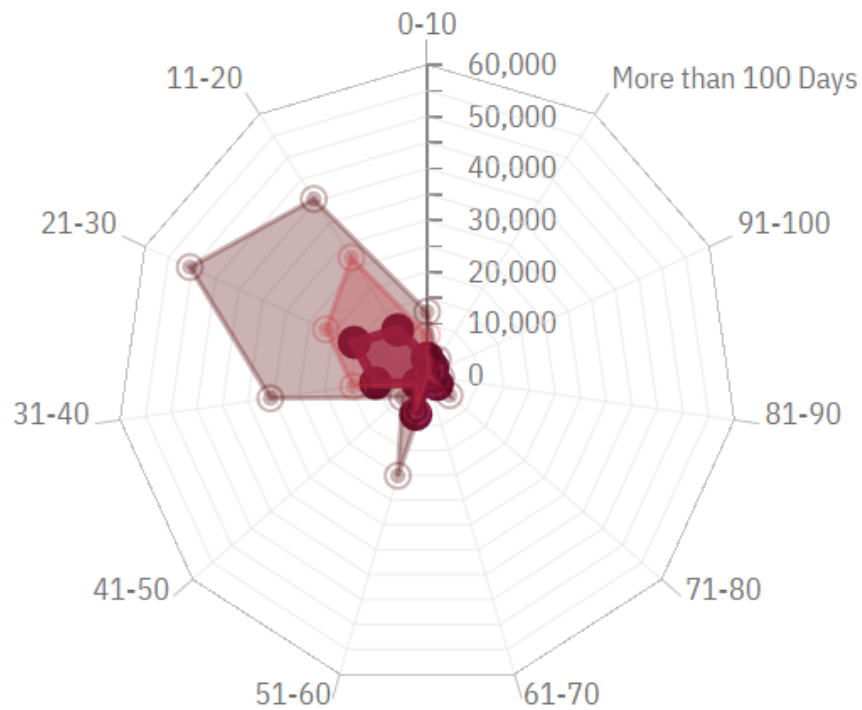
## No of patients by no of days in stay colored by severity of Illness

patientid by Stay colored by Severity of Illness

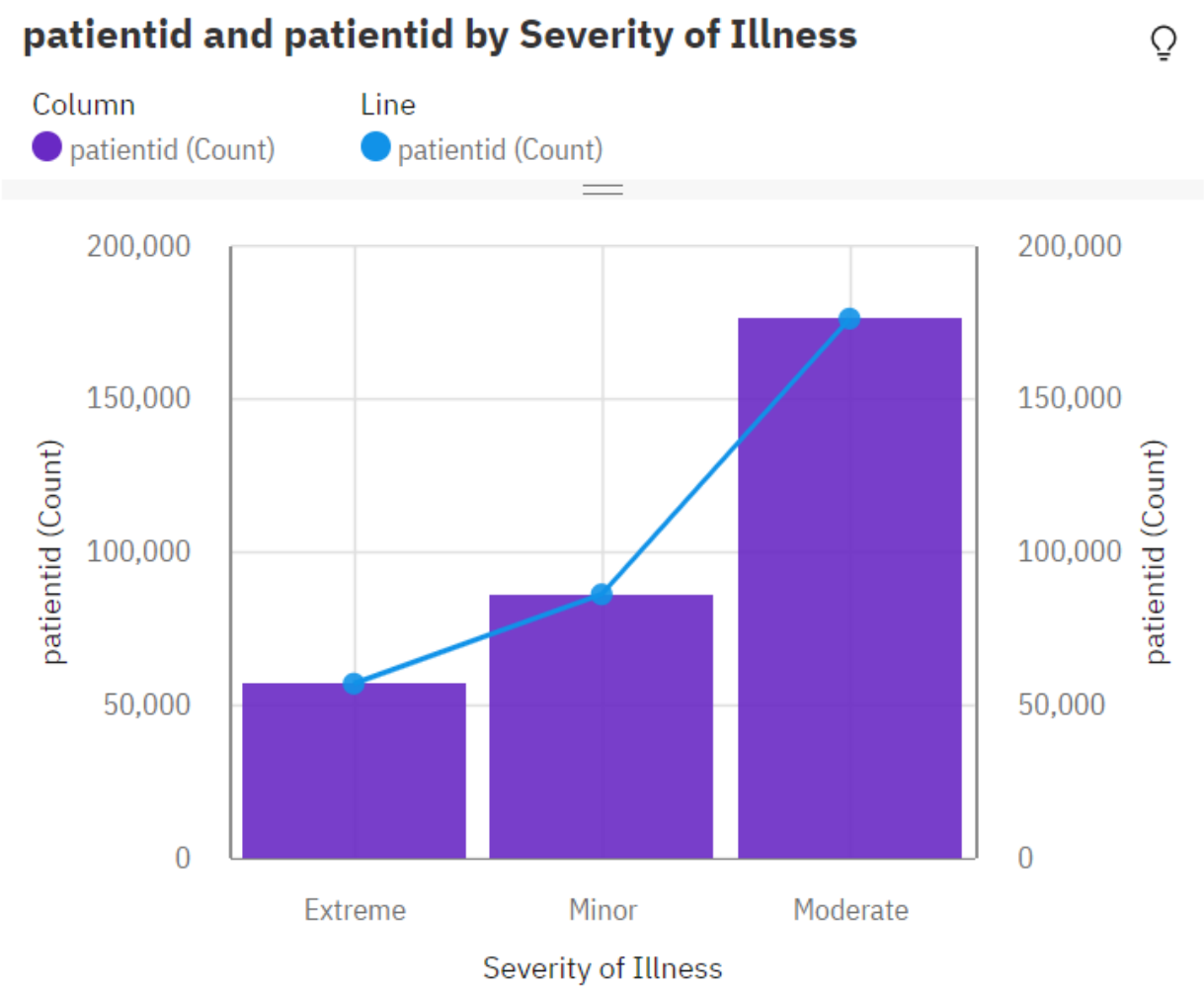


Severity of Illness

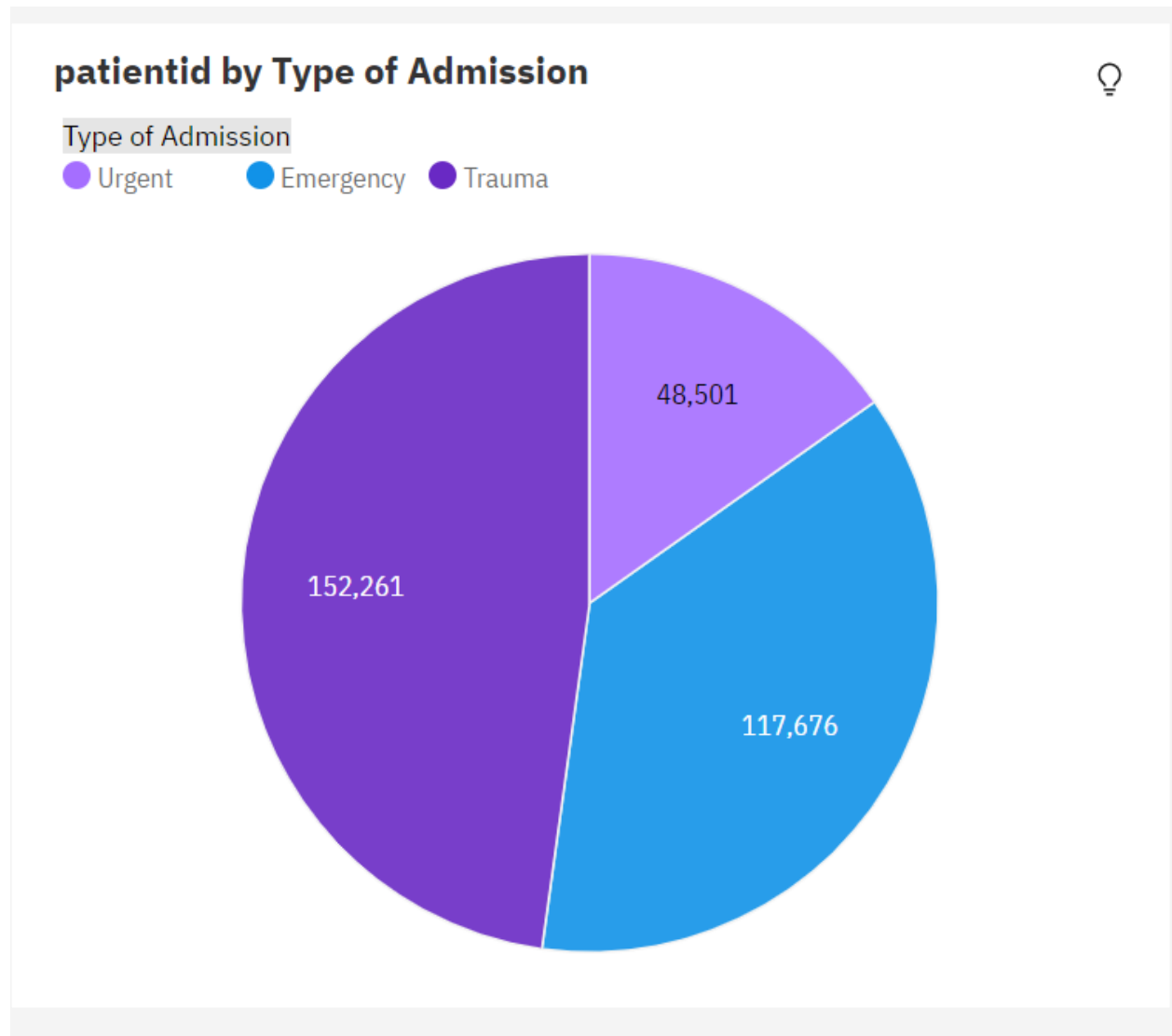
● Extreme ● Minor ● Moderate



# No of patients by severity of disease



## No of patients and type of admission:

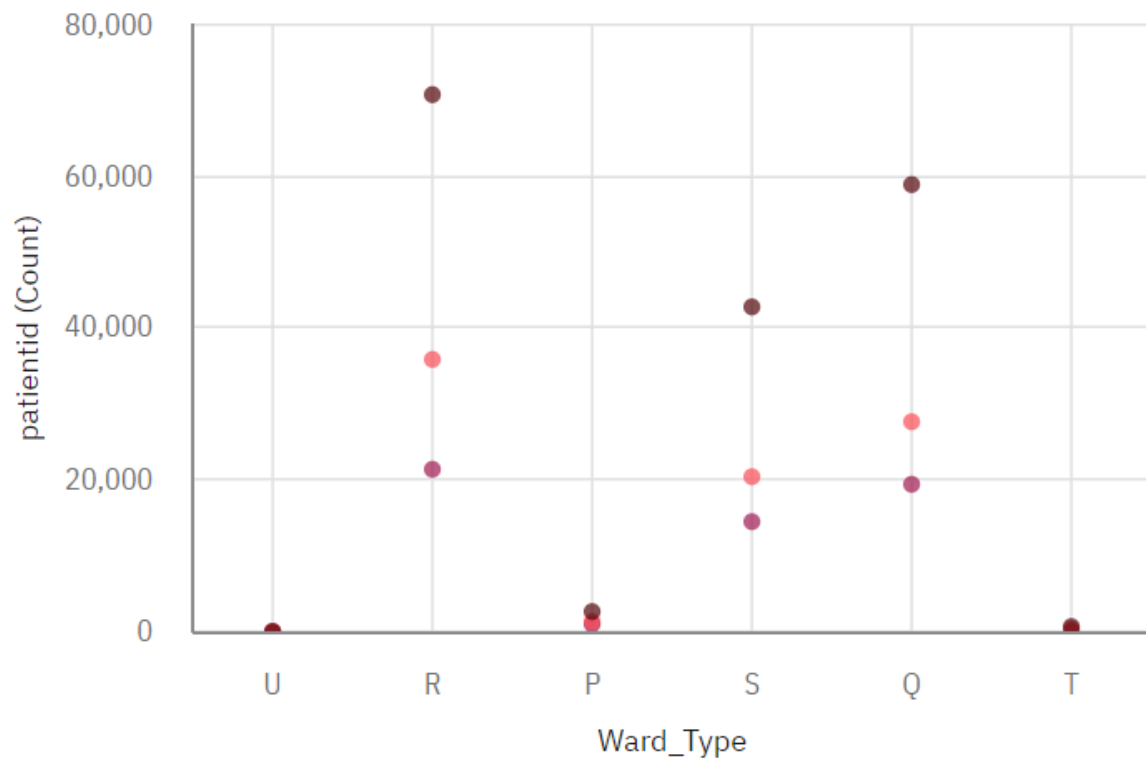


## No of patients by ward type colored by severity of illness

### Ward\_Type by patientid colored by Severity of Illness

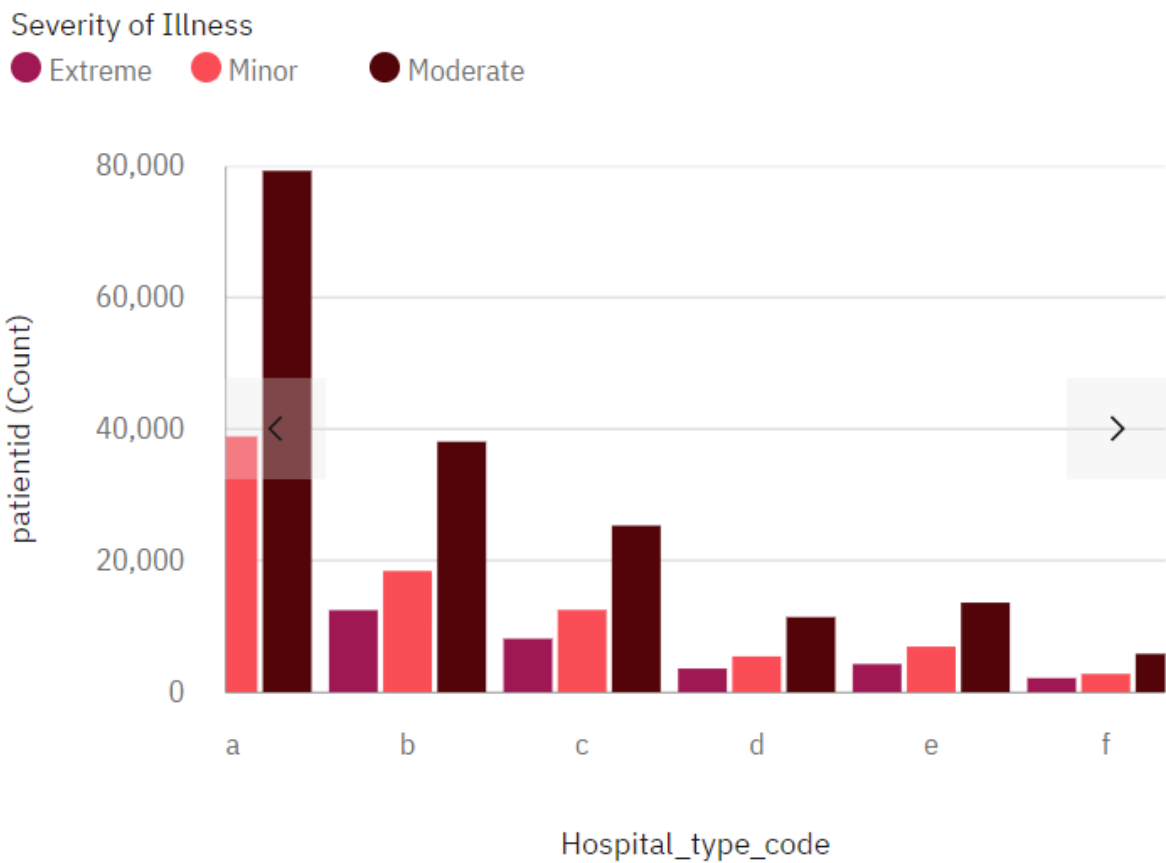
Severity of Illness

Extreme Minor Moderate

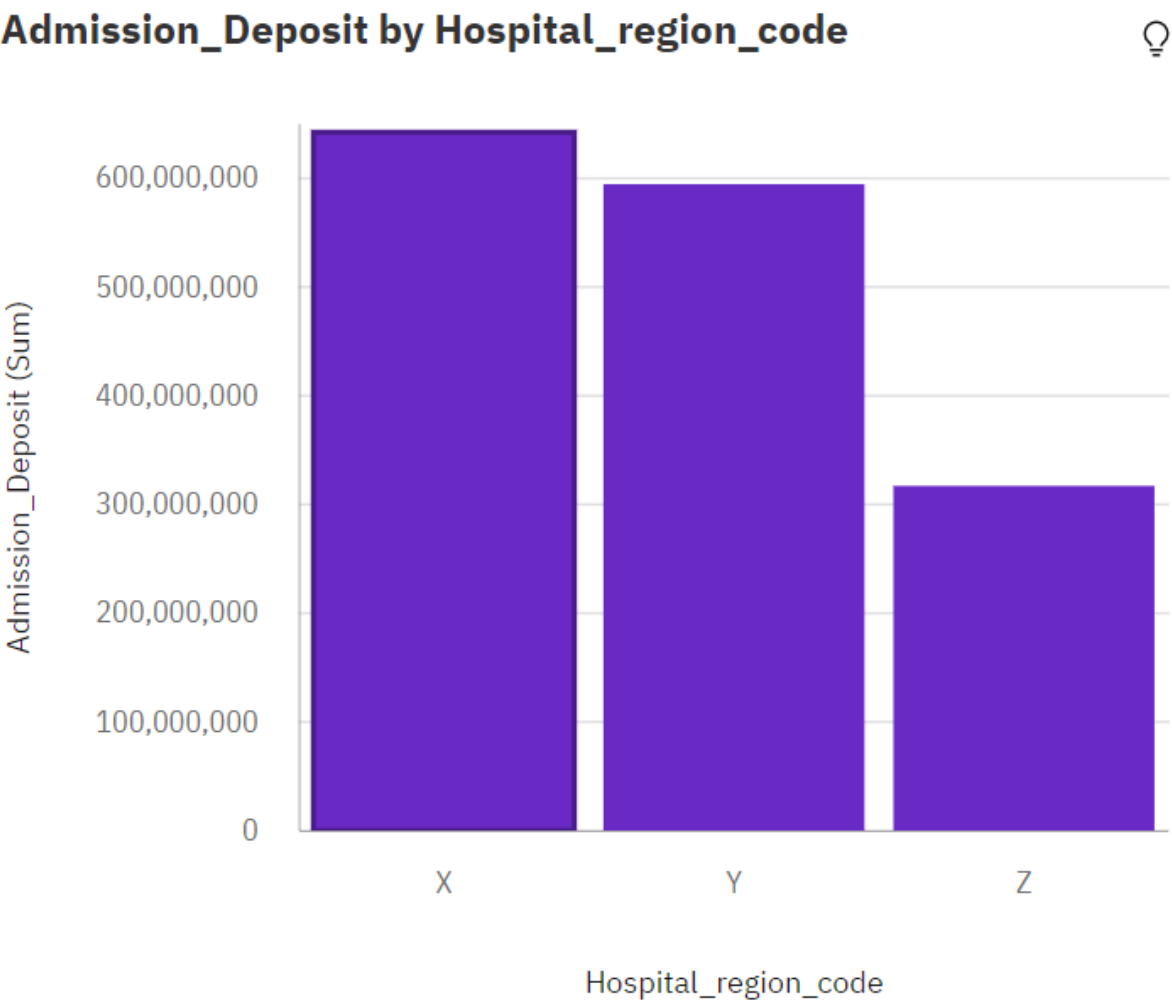


No of patients in a hospital type colored by Severity of Illness

patientid by Hospital\_type\_code colored by Severity of Illness



**Admission Deposit by Hospital region code:**





### No of patients by city code hospital colored by age

patientid by City\_Code\_Hospital colored by Age

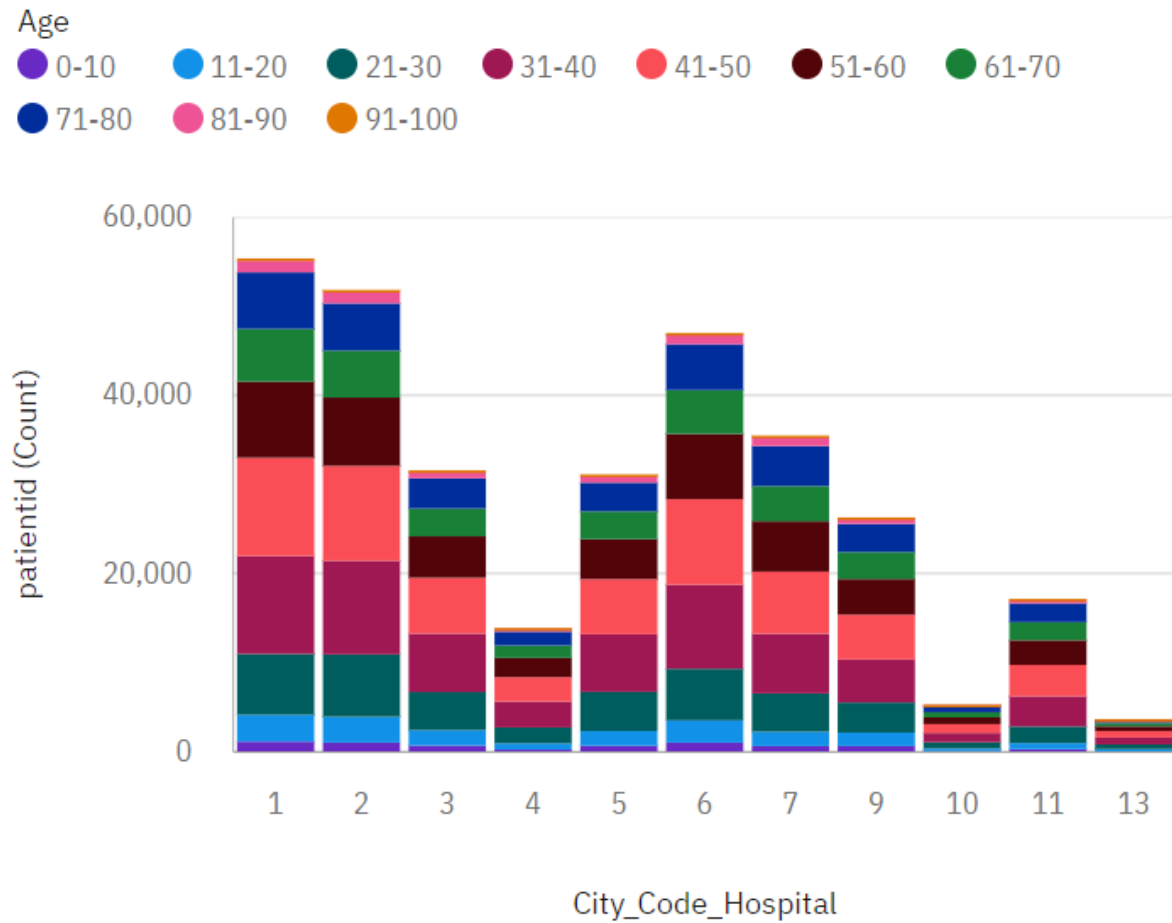
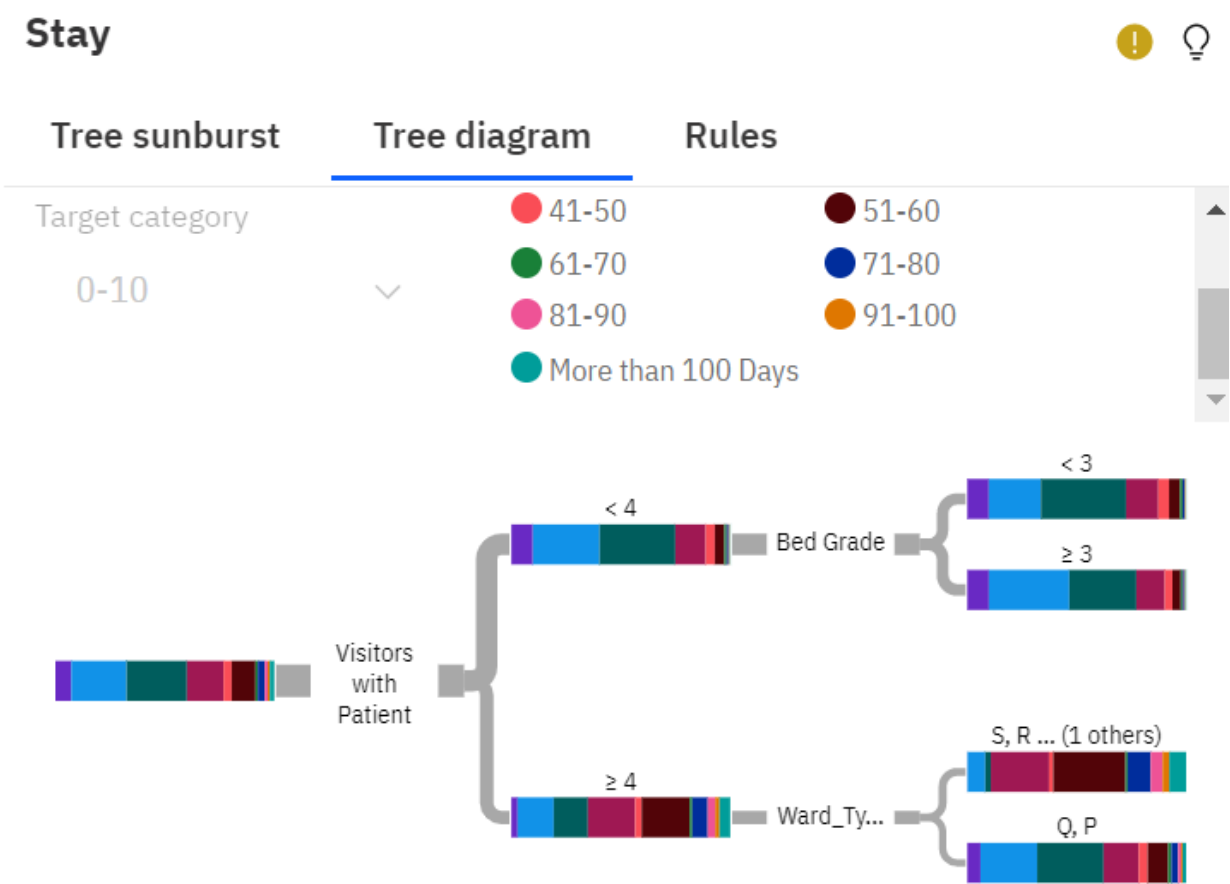


Table for age and no of patients in each age category:

Age and Hospital\_region\_code

Hospital_region_code	Age	Age
X	0-10	2,571
	11-20	6,900
	21-30	16,298
	31-40	26,716
	41-50	26,897
	51-60	20,786
	61-70	14,246
	71-80	15,167
	81-90	3,236
	91-100	519
Summary		133,336

Decision Tree:



Data Cleaning

# DATA CLEANING - HANDLING NULL VALUES

```
df["Bed Grade"].value_counts()
```

[9]



```
2.0    123671
3.0    110583
4.0     57566
1.0     26505
Name: Bed Grade, dtype: int64
```

```
df["Bed Grade"].fillna(2.0, inplace=True)
```

[10]



```
df["Bed Grade"].isna().sum()
```

[11]



```
0
```

[12]



```
df["City_Code_Hospital"].value_counts()
```

[12]



```
1    55351
2    51809
6    46991
7    35463
3    31569
5    31105
9    26277
11   17137
4    13857
10    5249
13    3630
Name: City_Code_Hospital, dtype: int64
```

.

```
df["City_Code_Patient"].fillna(1, inplace=True)
```

[13]



[14]

```
df.isna().sum()
```

Hospital_code	0
Hospital_type_code	0
City_Code_Hospital	0
Hospital_region_code	0
Available Extra Rooms in Hospital	0
Department	0
Ward_Type	0
Ward_Facility_Code	0
Bed Grade	0
City_Code_Patient	0
Type of Admission	0
Severity of Illness	0
Visitors with Patient	0
Age	0
Admission_Deposit	0
Stay	0
dtype: int64	

## Dropping unnecessary attributes






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

[8]

## Correlations in data

df.corr()

[7]

	<div>case_id float64</div> <div>-0.045972258824...</div> 	<div>Hospital_code fl...</div> <div>-0.059638371212...</div> 	<div>City_Code_Hosp...</div> <div>-0.049309083896...</div> 	<div>Available Extra ...</div> <div>-0.143739099539...</div> 	<div>Bed Grade TIO2</div> <div>-0.115867687</div> 
case_id	1.0	-0.043022506716 140094	-0.011351672166 076155	0.042580492972 68307	0.01370191 81
Hospital_code	-0.043022506716 140094	1.0	0.128293624838 98515	-0.059638371212 683096	-0.01373895 93
City_Code_Hosp...	-0.011351672166 076155	0.128293624838 98515	1.0	-0.045770970394 17465	-0.04930908 30
Available Extr...	0.042580492972 68307	-0.059638371212 683096	-0.045770970394 17465	1.0	-0.11586768 2
Bed Grade	0.013701912168 819724	-0.013738959637 931984	-0.049309083896 303674	-0.115867687620 21678	
patientid	-0.004149891023 962355	0.002290615096 8089894	0.000750373459 9816436	0.000920916992 943411	0.00164487 639
City_Code_Pa...	0.065196066629 81017	-0.015529844452 98396	-0.023988370312 9584	-0.009680986398 408104	-0.00810544 75
Visitors with...	0.001308943118 4283259	-0.028500291008 472723	0.018184441177 590032	0.096714353688 21796	0.08894536 4
Admissi	-0.045972258824	0.045445524001	-0.034455292791	-0.143739099539	0.07383255

Ready

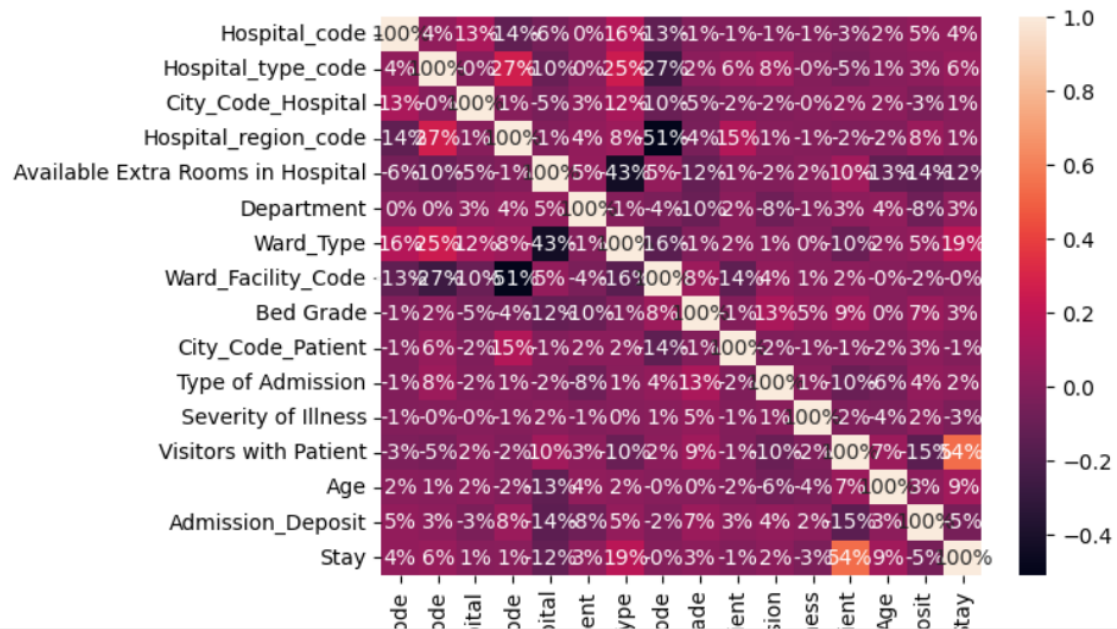
Run notebook

[42]

```
import seaborn
```

```
seaborn.heatmap(df.corr(), annot=True, fmt='.0%')
```

<AxesSubplot: >



# DATA TRANSFORMATION



```
from sklearn import preprocessing
```

[16]



```
le = preprocessing.LabelEncoder()
```

[17]



```
le.fit(df["Stay"])
```

[18]



▼ LabelEncoder  
LabelEncoder()

```
transformed = le.transform(df["Stay"])
```

[19]



```
df["Stay"] = transformed
```

[20]





```
le.fit(df["Age"])
```

[21]



▼ LabelEncoder

LabelEncoder()

```
age_transformed = le.transform(df["Age"])
```

[22]



```
df["Age"] = age_transformed
```

[23]



```
le.fit(df["Hospital_region_code"])  
hrc_transformed = le.transform(df["Hospital_region_code"])  
df["Hospital_region_code"] = hrc_transformed
```

[24]



```
le.fit(df["Department"])  
dept_transformed = le.transform(df["Department"])  
df["Department"] = dept_transformed
```

[25]

```
le.fit(df["Hospital_region_code"])
hrc_transformed = le.transform(df["Hospital_region_code"])
df["Hospital_region_code"] = hrc_transformed
```

[24]

```
le.fit(df["Department"])
dept_transformed = le.transform(df["Department"])
df["Department"] = dept_transformed
```

[25]

```
le.fit(df["Ward_Type"])
wt_transformed = le.transform(df["Ward_Type"])
df["Ward_Type"] = wt_transformed
```

[26]

```
le.fit(df["Ward_Facility_Code"])
wfc_transformed = le.transform(df["Ward_Facility_Code"])
df["Ward_Facility_Code"] = wfc_transformed
```

[27]

## Transformed data

df.head()

[31]

	Hospital_code in...	Hospital_type_c...	City_Code_Hosp...	Hospital_region...	Visualize Available Extra
0	8	2	3	2	
1	2	2	5	2	
2	10	4	1	0	
3	26	1	2	1	
4	26	1	2	1	

5 rows, showing 10 per page << < Page 1 of 1 > >> 