

911 Calls Capstone Project

Data and Setup

Importing Required Packages

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

Reading in the csv file and creating Data Frame

```
from google.colab import drive
drive.mount('/content/drive')
df = pd.read_csv('./drive/MyDrive/Dataset/911.csv')
df.head()
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

	lat	lng	desc \
0	40.297876	-75.581294	REINDEER CT & DEAD END; NEW HANOVER; Station
...			
1	40.258061	-75.264680	BRIAR PATH & WHITEMARSH LN; HATFIELD TOWNSHIP...
2	40.121182	-75.351975	HAWS AVE; NORRISTOWN; 2015-12-10 @ 14:39:21-St...
3	40.116153	-75.343513	AIRY ST & SWEDE ST; NORRISTOWN; Station 308A;...
4	40.251492	-75.603350	CHERRYWOOD CT & DEAD END; LOWER POTTS GROVE; S...

	zip	title	timeStamp	twp \
0	19525.0	EMS: BACK PAINS/INJURY	2015-12-10 17:40:00	NEW HANOVER
1	19446.0	EMS: DIABETIC EMERGENCY	2015-12-10 17:40:00	HATFIELD TOWNSHIP
2	19401.0	Fire: GAS-ODOR/LEAK	2015-12-10 17:40:00	NORRISTOWN
3	19401.0	EMS: CARDIAC EMERGENCY	2015-12-10 17:40:01	NORRISTOWN
4	NaN	EMS: DIZZINESS	2015-12-10 17:40:01	LOWER POTTS GROVE

	addr	e
0	REINDEER CT & DEAD END	1
1	BRIAR PATH & WHITEMARSH LN	1
2	HAWS AVE	1
3	AIRY ST & SWEDE ST	1
4	CHERRYWOOD CT & DEAD END	1

Checking the info() of the df

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 99492 entries, 0 to 99491
Data columns (total 9 columns):
 #   Column      Non-Null Count  Dtype
---  -
 0   lat         99492 non-null  float64
 1   lng         99492 non-null  float64
 2   desc        99492 non-null  object
 3   zip         86637 non-null  float64
 4   title       99492 non-null  object
 5   timeStamp   99492 non-null  object
 6   twp         99449 non-null  object
 7   addr        98973 non-null  object
 8   e           99492 non-null  int64
dtypes: float64(3), int64(1), object(5)
memory usage: 6.8+ MB
```

Basic Questions

What are the top 5 zipcodes for 911 calls?

df['zip'].value_counts().head(5)

```
19401.0    6979
19464.0    6643
19403.0    4854
19446.0    4748
19406.0    3174
Name: zip, dtype: int64
```

What are the top 5 townships (twp) for 911 calls?

df['twp'].value_counts().head(5)

```
LOWER MERION    8443
ABINGTON        5977
NORRISTOWN      5890
UPPER MERION    5227
```

```
CHELtenham      4575
Name: twp, dtype: int64
```

How many unique title codes are there?

```
df['title'].unique
```

```
<bound method Series.unique of 0
1      EMS: DIABETIC EMERGENCY      EMS: BACK PAINS/INJURY
2      Fire: GAS-ODOR/LEAK
3      EMS: CARDIAC EMERGENCY
4      EMS: DIZZINESS
...
99487  Traffic: VEHICLE ACCIDENT -
99488  Traffic: VEHICLE ACCIDENT -
99489      EMS: FALL VICTIM
99490      EMS: NAUSEA/VOMITING
99491  Traffic: VEHICLE ACCIDENT -
Name: title, Length: 99492, dtype: object>
```

Creating new features

In the titles column there are "Reasons/Departments" specified before the title code. These are EMS, Fire, and Traffic. Using `.apply()` to create a new column called "Reason" that contains this string value.

```
def createReason(x):
    x = x.split(":")
    return x[0]
```

```
df['Reason'] = df['title'].apply(createReason)
df.head()
```

```
      lat      lng
desc \
0  40.297876 -75.581294  REINDEER CT & DEAD END;  NEW HANOVER; Station
...
1  40.258061 -75.264680  BRIAR PATH & WHITEMARSH LN;  HATFIELD
TOWNSHIP...
2  40.121182 -75.351975  HAWS AVE; NORRISTOWN; 2015-12-10 @ 14:39:21-
St...
3  40.116153 -75.343513  AIRY ST & SWEDE ST;  NORRISTOWN; Station
308A;...
4  40.251492 -75.603350  CHERRYWOOD CT & DEAD END;  LOWER POTTS GROVE;
S...
```

```
      zip      title      timeStamp
twp \
0  19525.0  EMS: BACK PAINS/INJURY  2015-12-10 17:40:00      NEW
HANOVER
```

1	19446.0	EMS: DIABETIC EMERGENCY	2015-12-10 17:40:00	HATFIELD TOWNSHIP
2	19401.0	Fire: GAS-ODOR/LEAK	2015-12-10 17:40:00	NORRISTOWN
3	19401.0	EMS: CARDIAC EMERGENCY	2015-12-10 17:40:01	NORRISTOWN
4	NaN	EMS: DIZZINESS	2015-12-10 17:40:01	LOWER POTTS GROVE

	addr	e	Reason
0	REINDEER CT & DEAD END	1	EMS
1	BRIAR PATH & WHITEMARSH LN	1	EMS
2	HAWS AVE	1	Fire
3	AIRY ST & SWEDE ST	1	EMS
4	CHERRYWOOD CT & DEAD END	1	EMS

What is the most common Reason for a 911 call based off of this new column?

```
df['Reason'].value_counts()
```

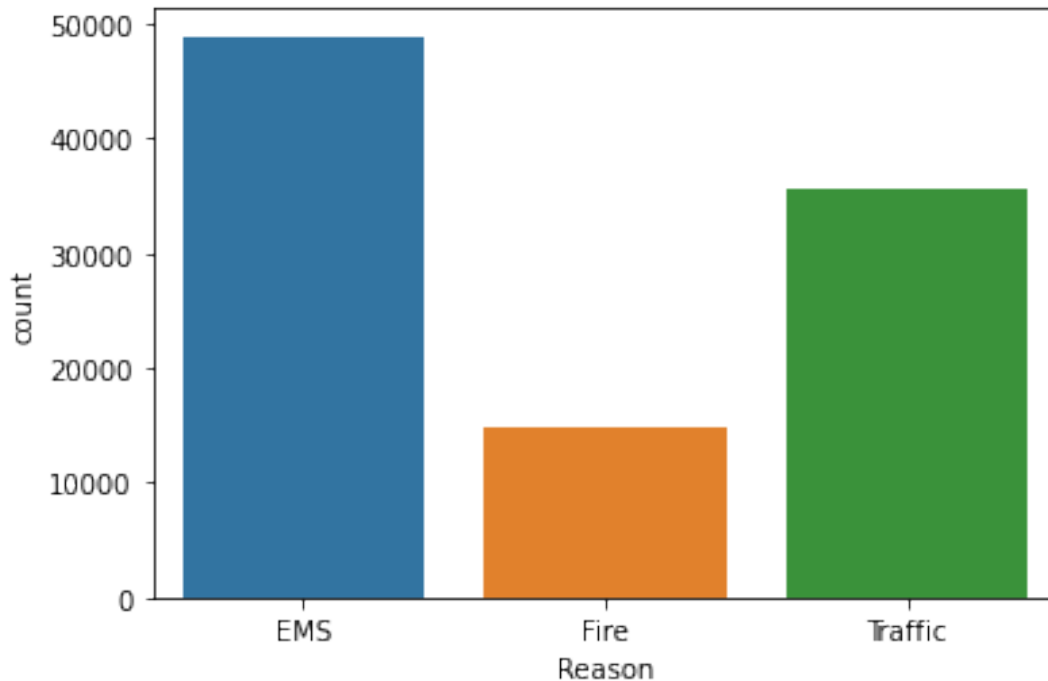
```
EMS      48877
Traffic  35695
Fire     14920
Name: Reason, dtype: int64
```

Creating a countplot of 911 calls by Reason.

```
import seaborn as sns
```

```
sns.countplot(data=df, x='Reason')
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fe9d1680ed0>
```



What is the data type of the objects in the timeStamp column?

```
df['timeStamp'].dtype
```

```
dtype('O')
```

Using pd.to_datetime to convert the column from strings to DateTime objects.

```
df['timeStamp'] = pd.to_datetime(df['timeStamp'])
df.head()
```

```

      lat      lng
desc \
0  40.297876 -75.581294  REINDEER CT & DEAD END;  NEW HANOVER; Station
...
1  40.258061 -75.264680  BRIAR PATH & WHITEMARSH LN;  HATFIELD
TOWNSHIP...
2  40.121182 -75.351975  HAWS AVE; NORRISTOWN; 2015-12-10 @ 14:39:21-
St...
3  40.116153 -75.343513  AIRY ST & SWEDE ST;  NORRISTOWN; Station
308A;...
4  40.251492 -75.603350  CHERRYWOOD CT & DEAD END;  LOWER POTTS GROVE;
S...
```

```

      zip      title      timeStamp
twp \
0  19525.0  EMS: BACK PAINS/INJURY 2015-12-10 17:40:00      NEW
HANOVER
```

```

1 19446.0 EMS: DIABETIC EMERGENCY 2015-12-10 17:40:00 HATFIELD
TOWNSHIP
2 19401.0 Fire: GAS-ODOR/LEAK 2015-12-10 17:40:00
NORRISTOWN
3 19401.0 EMS: CARDIAC EMERGENCY 2015-12-10 17:40:01
NORRISTOWN
4 NaN EMS: DIZZINESS 2015-12-10 17:40:01 LOWER
POTTSGROVE

```

```

          addr e Reason
0 REINDEER CT & DEAD END 1 EMS
1 BRIAR PATH & WHITEMARSH LN 1 EMS
2 HAWS AVE 1 Fire
3 AIRY ST & SWEDE ST 1 EMS
4 CHERRYWOOD CT & DEAD END 1 EMS

```

Now that the timestamp column is actually DateTime objects, using .apply() to create 3 new columns called Hour, Month, and Day of Week.

```

df['Hour'] = df['timeStamp'].apply(lambda x: x.hour)
df['Month'] = df['timeStamp'].apply(lambda x: x.month)
df['DayOfWeek'] = df['timeStamp'].apply(lambda x: x.dayofweek)

```

```
df.head()
```

```

          lat      lng
desc \
0 40.297876 -75.581294 REINDEER CT & DEAD END; NEW HANOVER; Station
...
1 40.258061 -75.264680 BRIAR PATH & WHITEMARSH LN; HATFIELD
TOWNSHIP...
2 40.121182 -75.351975 HAWS AVE; NORRISTOWN; 2015-12-10 @ 14:39:21-
St...
3 40.116153 -75.343513 AIRY ST & SWEDE ST; NORRISTOWN; Station
308A;...
4 40.251492 -75.603350 CHERRYWOOD CT & DEAD END; LOWER POTTSGROVE;
S...

```

```

          zip          title          timeStamp
twp \
0 19525.0 EMS: BACK PAINS/INJURY 2015-12-10 17:40:00 NEW
HANOVER
1 19446.0 EMS: DIABETIC EMERGENCY 2015-12-10 17:40:00 HATFIELD
TOWNSHIP
2 19401.0 Fire: GAS-ODOR/LEAK 2015-12-10 17:40:00
NORRISTOWN
3 19401.0 EMS: CARDIAC EMERGENCY 2015-12-10 17:40:01
NORRISTOWN
4 NaN EMS: DIZZINESS 2015-12-10 17:40:01 LOWER
POTTSGROVE

```

	addr	e	Reason	Hour	Month	DayOfWeek
0	REINDEER CT & DEAD END	1	EMS	17	12	3
1	BRIAR PATH & WHITEMARSH LN	1	EMS	17	12	3
2	HAWS AVE	1	Fire	17	12	3
3	AIRY ST & SWEDE ST	1	EMS	17	12	3
4	CHERRYWOOD CT & DEAD END	1	EMS	17	12	3

Using the .map() with dictionary to map the actual string names to the day of the week

```
dmap = {0: 'Mon', 1: 'Tue', 2: 'Wed', 3: 'Thu', 4: 'Fri', 5: 'Sat', 6: 'Sun'}
df['DayOfWeek'] = df['DayOfWeek'].map(dmap)
```

```
df.head()
```

	lat	lng	desc \
0	40.297876	-75.581294	REINDEER CT & DEAD END; NEW HANOVER; Station
...			
1	40.258061	-75.264680	BRIAR PATH & WHITEMARSH LN; HATFIELD TOWNSHIP...
2	40.121182	-75.351975	HAWS AVE; NORRISTOWN; 2015-12-10 @ 14:39:21-St...
3	40.116153	-75.343513	AIRY ST & SWEDE ST; NORRISTOWN; Station 308A;...
4	40.251492	-75.603350	CHERRYWOOD CT & DEAD END; LOWER POTTS GROVE; S...

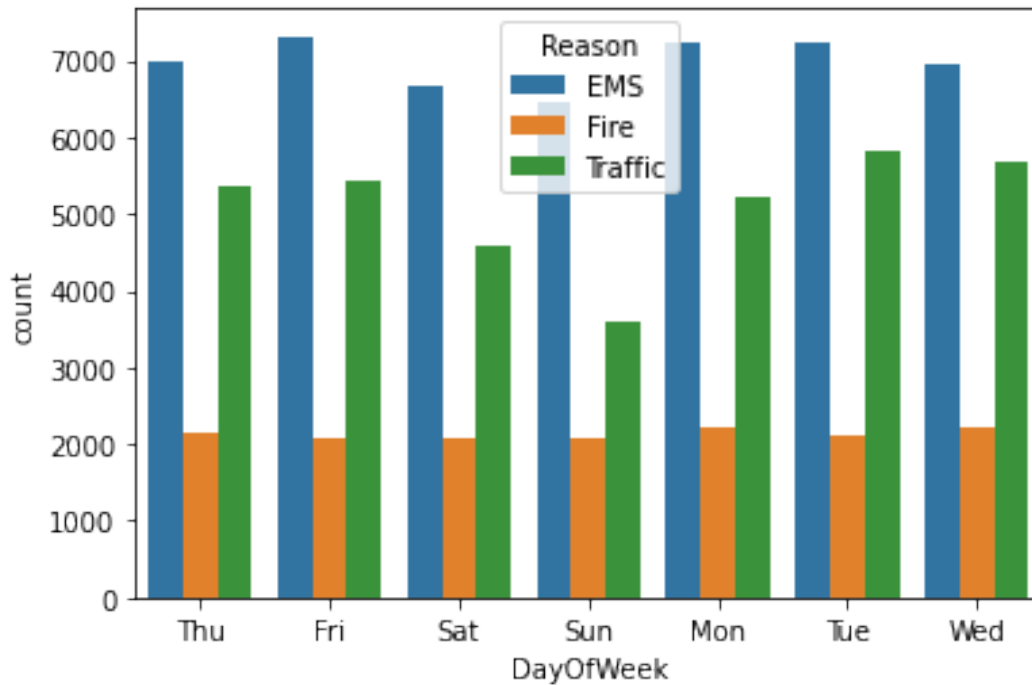
	zip	title	timeStamp	twp \
0	19525.0	EMS: BACK PAINS/INJURY	2015-12-10 17:40:00	NEW HANOVER
1	19446.0	EMS: DIABETIC EMERGENCY	2015-12-10 17:40:00	HATFIELD TOWNSHIP
2	19401.0	Fire: GAS-ODOR/LEAK	2015-12-10 17:40:00	NORRISTOWN
3	19401.0	EMS: CARDIAC EMERGENCY	2015-12-10 17:40:01	NORRISTOWN
4	NaN	EMS: DIZZINESS	2015-12-10 17:40:01	LOWER POTTS GROVE

	addr	e	Reason	Hour	Month	DayOfWeek
0	REINDEER CT & DEAD END	1	EMS	17	12	Thu
1	BRIAR PATH & WHITEMARSH LN	1	EMS	17	12	Thu
2	HAWS AVE	1	Fire	17	12	Thu
3	AIRY ST & SWEDE ST	1	EMS	17	12	Thu
4	CHERRYWOOD CT & DEAD END	1	EMS	17	12	Thu

Creating a countplot of the Day of Week column with the hue based off of the Reason column.

```
sns.countplot(data=df, x='DayOfWeek', hue='Reason')
```

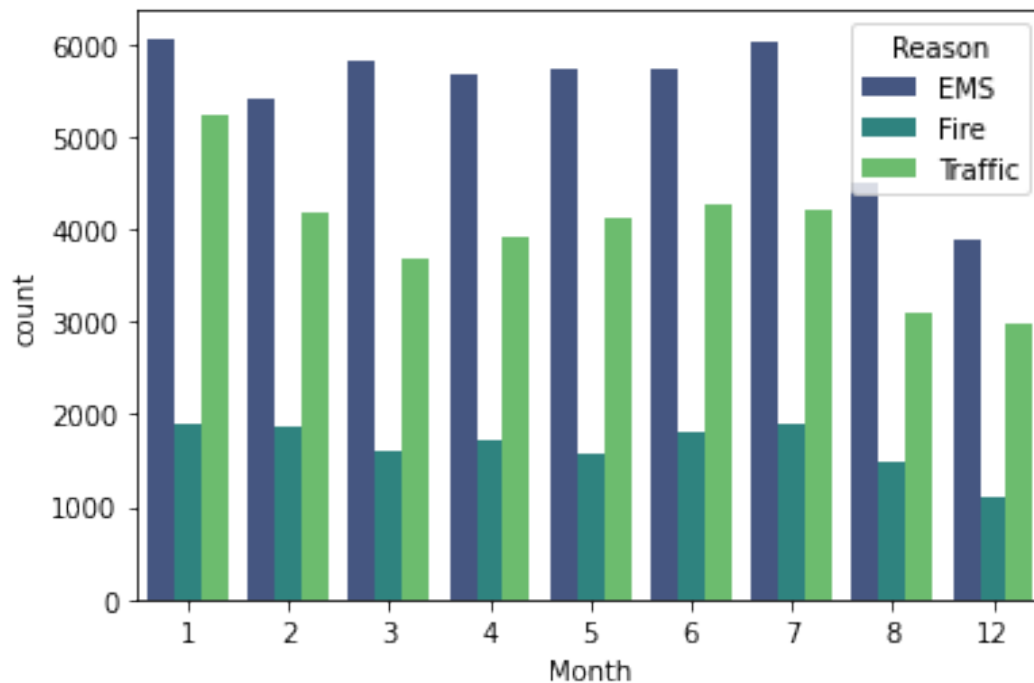
<matplotlib.axes._subplots.AxesSubplot at 0x7fe9d05ffb90>



Now doing the same for Month

```
sns.countplot(data=df, x='Month', hue='Reason', palette='viridis')
```

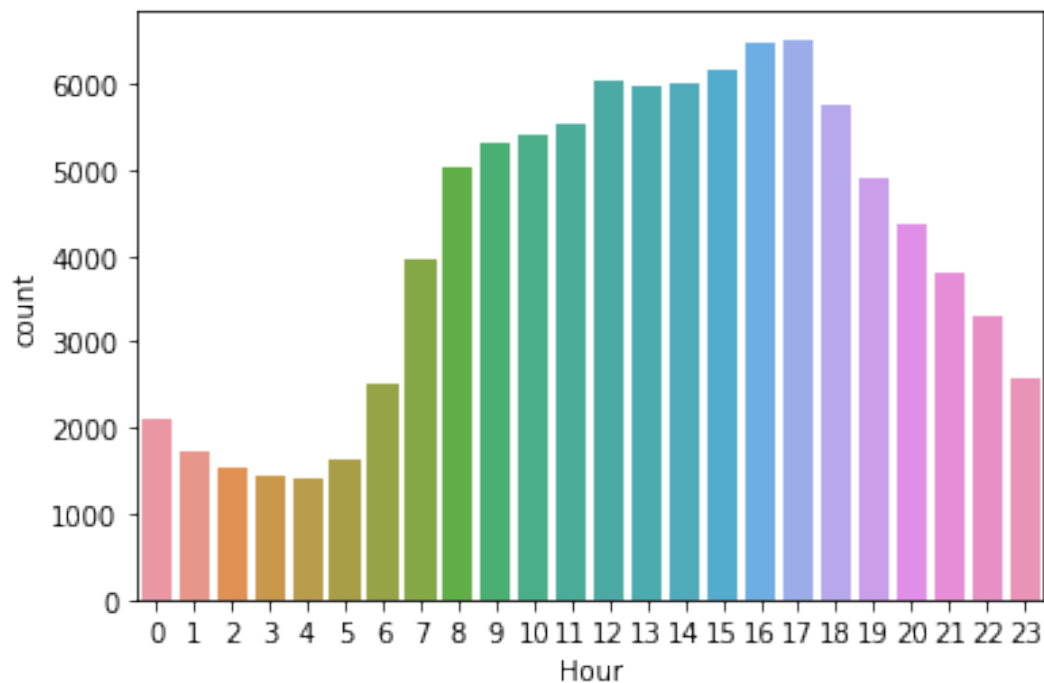
<matplotlib.axes._subplots.AxesSubplot at 0x7fe9d168e250>



Which Hour Has the most calls?

```
sns.countplot(data=df, x='Hour')
```

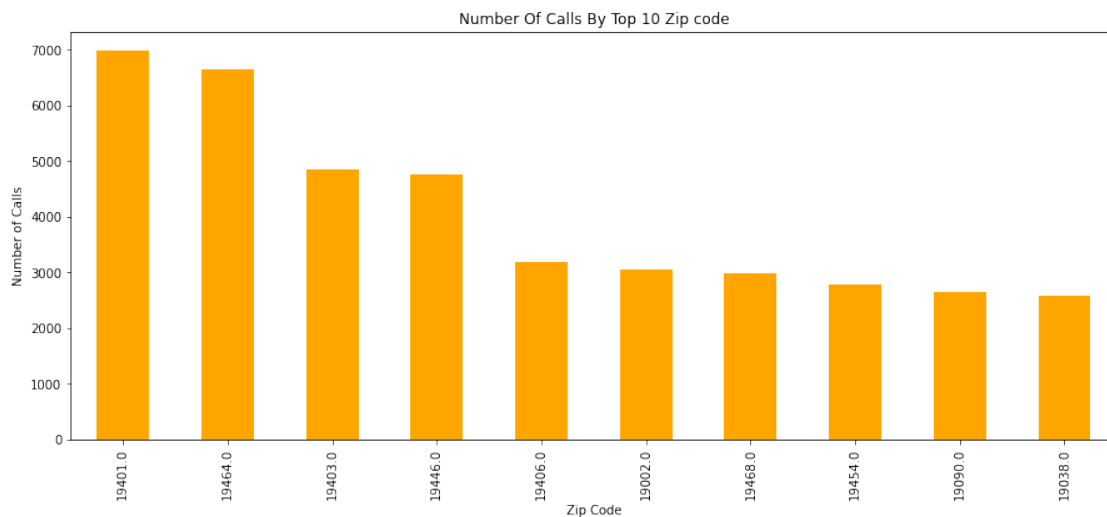
<matplotlib.axes._subplots.AxesSubplot at 0x7fe9d0405a90>



Which Zip code Receives More Calls?

```
df['zip'].value_counts().head(10).plot(kind='bar', figsize=(15,6),  
color='orange')  
plt.title("Number Of Calls By Top 10 Zip code")  
plt.xlabel('Zip Code')  
plt.ylabel('Number of Calls')
```

```
Text(0, 0.5, 'Number of Calls')
```



Observations

1. 19401 Zip code receives the most 911 calls.
2. 911 receives the most calls from LOWER MERION.
3. 911 receives most calls for EMS with 48877 calls in the dataset.
4. Wednesdays record most calls for EMS.
5. Tuesday record most calls for traffic.
6. Mondays record most calls for Fire.
7. January and July have received the highest and almost same number of calls for EMS.
8. Most calls for traffic were received in January.
9. Most calls for fire were received in July.
10. 911 receives most calls between 15, 16, and 17 hours of the day.