911 Calls project

October 24, 2016

1 911 Calls Project

For this capstone project we will be analyzing some 911 call data from Kaggle. The data contains the following fields:

- lat : String variable, Latitude
- $\bullet\,$ l
ng: String variable, Longitude
- desc: String variable, Description of the Emergency Call
- zip: String variable, Zipcode
- title: String variable, Title
- timeStamp: String variable, YYYY-MM-DD HH:MM:SS
- twp: String variable, Township
- addr: String variable, Address
- e: String variable, Dummy variable (always 1)

Just go along with this notebook and try to complete the instructions or answer the questions in bold using your Python and Data Science skills!

1.1 Data and Setup

```
** Import numpy and pandas **
In [138]: import numpy as np
    import pandas as pd

** Import visualization libraries and set %matplotlib inline. **
In [139]: import matplotlib.pyplot as plt
    import seaborn as sns
    sns.set_style('whitegrid')
    %matplotlib inline

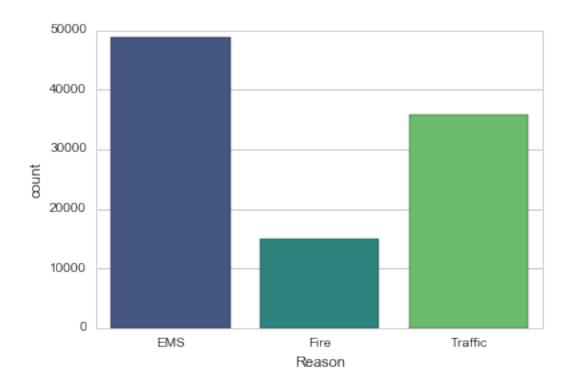
** Read in the csv file as a dataframe called df **
In [140]: df = pd.read_csv('911.csv')

** Check the info() of the df **
In [141]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 99492 entries, 0 to 99491
Data columns (total 9 columns):
            99492 non-null float64
lng
             99492 non-null float64
            99492 non-null object
desc
             86637 non-null float64
zip
title
             99492 non-null object
             99492 non-null object
timeStamp
twp
             99449 non-null object
addr
             98973 non-null object
             99492 non-null int64
dtypes: float64(3), int64(1), object(5)
memory usage: 6.8+ MB
  ** Check the head of df **
In [142]: df.head(5)
Out[142]:
                                                                                 desc
                              lng
                                   REINDEER CT & DEAD END; NEW HANOVER; Station ...
          0 40.297876 -75.581294
          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
            19525.0
                       EMS: BACK PAINS/INJURY
                                               2015-12-10 17:40:00
                                                                           NEW HANOVER
            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
            19401.0
                       EMS: CARDIAC EMERGENCY 2015-12-10 17:40:01
                                                                           NORRISTOWN
                               EMS: DIZZINESS 2015-12-10 17:40:01
                 NaN
                                                                     LOWER POTTSGROVE
                                   addr
                 REINDEER CT & DEAD END
          0
            BRIAR PATH & WHITEMARSH LN
          2
                               HAWS AVE 1
          3
                     AIRY ST & SWEDE ST
          4
               CHERRYWOOD CT & DEAD END 1
     Basic Questions
1.2
** The top 5 zipcodes for 911 calls? **
In [143]: df['zip'].value_counts().head(10)
Out[143]: 19401.0
                     6979
          19464.0
                     6643
          19403.0
                     4854
          19446.0
                     4748
          19406.0
                     3174
          19002.0
                     3050
          19468.0
                     2990
          19454.0
                     2781
          19090.0
                     2635
          19038.0
                     2578
          Name: zip, dtype: int64
```

```
** The top 5 townships (twp) for 911 calls? **
In [144]: df['twp'].value_counts().head(8)
Out[144]: LOWER MERION
                              8443
          ABINGTON
                              5977
          NORRISTOWN
                              5890
          UPPER MERION
                              5227
          CHELTENHAM
                              4575
          POTTSTOWN
                              4146
          UPPER MORELAND
                              3434
          LOWER PROVIDENCE
                              3225
          Name: twp, dtype: int64
  Unique title codes
In [145]: df['title'].nunique()
Out[145]: 110
     Creating new features
In [146]: df['Reason1'] = df['title'].apply(lambda title: title.split(':')[1])
In [11]: df['Reason1'].value_counts()
Out[11]: VEHICLE ACCIDENT -
                                           23066
          DISABLED VEHICLE -
                                            7702
          VEHICLE ACCIDENT
                                            5573
                                            5510
          FIRE ALARM
          RESPIRATORY EMERGENCY
                                            5112
          CARDIAC EMERGENCY
                                            5012
          FALL VICTIM
                                            4863
          ROAD OBSTRUCTION -
                                            3144
          SUBJECT IN PAIN
                                            2687
          HEAD INJURY
                                            2631
          UNKNOWN MEDICAL EMERGENCY
                                            1874
          SYNCOPAL EPISODE
                                            1718
          SEIZURES
                                            1630
          FIRE INVESTIGATION
                                            1478
          GENERAL WEAKNESS
                                            1464
          ABDOMINAL PAINS
                                            1436
          ALTERED MENTAL STATUS
                                            1386
          MEDICAL ALERT ALARM
                                            1344
          HEMORRHAGING
                                            1259
          CVA/STROKE
                                            1202
          OVERDOSE
                                            1191
          NAUSEA/VOMITING
                                            1131
          UNCONSCIOUS SUBJECT
                                            1116
          HAZARDOUS ROAD CONDITIONS -
                                            1086
          GAS-ODOR/LEAK
                                            1069
          BUILDING FIRE
                                            1008
          DIABETIC EMERGENCY
                                             837
          DIZZINESS
                                             799
          BACK PAINS/INJURY
                                             739
          FIRE SPECIAL SERVICE
                                             721
```

```
S/B AT HELICOPTER LANDING
                                             116
          RESCUE - GENERAL
                                              87
          BURN VICTIM
                                              86
          ANIMAL BITE
                                              83
          HEAT EXHAUSTION
                                              78
          VEHICLE LEAKING FUEL -
                                              77
          DEBRIS/FLUIDS ON HIGHWAY
                                              74
          VEHICLE LEAKING FUEL
                                              73
          DEBRIS/FLUIDS ON HIGHWAY -
                                              67
          SHOOTING
                                              39
          POISONING
                                              36
          RESCUE - WATER
                                              36
          EYE INJURY
                                              36
          STABBING
                                              32
          TRANSFERRED CALL
                                              30
          PUMP DETAIL
                                              25
          RESCUE - TECHNICAL
                                              19
          AMPUTATION
                                              14
          HAZARDOUS MATERIALS INCIDENT
                                              11
          TRAIN CRASH
                                               8
          DROWNING
                                               4
          INDUSTRIAL ACCIDENT
                                               3
          ELECTROCUTION
                                               2
          PLANE CRASH
                                               2
          POLICE INFORMATION
                                               2
          WARRANT SERVICE
                                               2
          SUSPICIOUS
                                               2
          ACTIVE SHOOTER
                                               2
          BOMB DEVICE FOUND
                                               1
          DISABLED VEHICLE
         Name: Reason1, dtype: int64
In [147]: df['Reason'] = df['title'].apply(lambda title: title.split(':')[0])
  ** The most common Reason for a 911 call based off of this new column? **
In [148]: df['Reason'].value_counts()
Out[148]: EMS
                     48877
          Traffic
                     35695
          Fire
                     14920
          Name: Reason, dtype: int64
  Graph Results
In [149]: sns.countplot(x='Reason',data=df,palette='viridis')
Out[149]: <matplotlib.axes._subplots.AxesSubplot at 0xe82a06cb70>
```



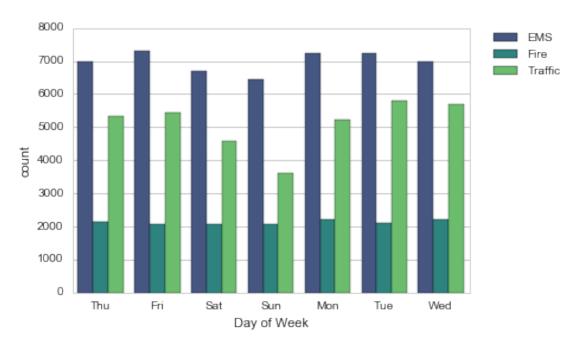
```
In [150]: type(df['timeStamp'].iloc[0])
Out[150]: str
  ** convert the column from strings to DateTime objects. **
In [151]: df['timeStamp'] = pd.to_datetime(df['timeStamp'])
In [152]: df['Hour'] = df['timeStamp'].apply(lambda time: time.hour)
          df['Month'] = df['timeStamp'].apply(lambda time: time.month)
          df['Day of Week'] = df['timeStamp'].apply(lambda time: time.dayofweek)
In [153]: dmap = {0:'Mon',1:'Tue',2:'Wed',3:'Thu',4:'Fri',5:'Sat',6:'Sun'}
          #dmap1 = {1:'Jan', 2:'Feb', 3:'Mar', 4:'Apr', 5:'May', 6:'Jun', 7:'Jul', 8:'Aug', 9:'Sep', 10
          #dmap2 = {17: 'Late', 16: 'Late', 15: 'Middle'}
In [154]: df['Day of Week'] = df['Day of Week'].map(dmap)
In [155]: df['Month'] = df['Month'].map(dmap1)
In [156]: df['Hour'] = df['Hour']
In [157]: del df['lat']
In [158]: del df['lng']
In [159]: del df['desc']
In [160]: del df['e']
In [161]: df.head(5)
```

```
Out[161]:
                 zip
                                         title
                                                          timeStamp
                                                                                    twp
                       EMS: BACK PAINS/INJURY 2015-12-10 17:40:00
             19525.0
                                                                            NEW HANOVER
                      EMS: DIABETIC EMERGENCY 2015-12-10 17:40:00
             19446.0
                                                                     HATFIELD TOWNSHIP
                          Fire: GAS-ODOR/LEAK 2015-12-10 17:40:00
                                                                             NORRISTOWN
             19401.0
          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
                                                       Reason1 Reason Hour Month
          0
                 REINDEER CT & DEAD END
                                            BACK PAINS/INJURY
                                                                  EMS
                                                                          17
                                                                               Dec
             BRIAR PATH & WHITEMARSH LN
                                           DIABETIC EMERGENCY
                                                                  EMS
                                                                          17
                                                                               Dec
          1
          2
                                HAWS AVE
                                                 GAS-ODOR/LEAK
                                                                 Fire
                                                                          17
                                                                               Dec
                     AIRY ST & SWEDE ST
                                            CARDIAC EMERGENCY
                                                                  EMS
          3
                                                                          17
                                                                               Dec
          4
               CHERRYWOOD CT & DEAD END
                                                     DIZZINESS
                                                                  EMS
                                                                          17
                                                                               Dec
            Day of Week
          0
                    Thu
          1
                    Thu
          2
                    Thu
          3
                    Thu
                    Thu
```

In [162]: sns.countplot(x='Day of Week',data=df,hue='Reason',palette='viridis')

plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)

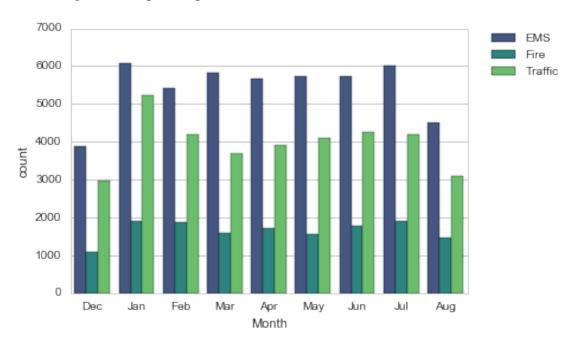
Out[162]: <matplotlib.legend.Legend at 0xe82a0790b8>



^{**} Seaborn to create a count plot of the Day of Week column with the hue based off of the Reason column.

```
In [163]: sns.countplot(x='Month',data=df,hue='Reason',palette='viridis')
    plt.legend(bbox_to_anchor=(1.05, 1), loc=2, borderaxespad=0.)
```

Out[163]: <matplotlib.legend.Legend at 0xe82a37d400>



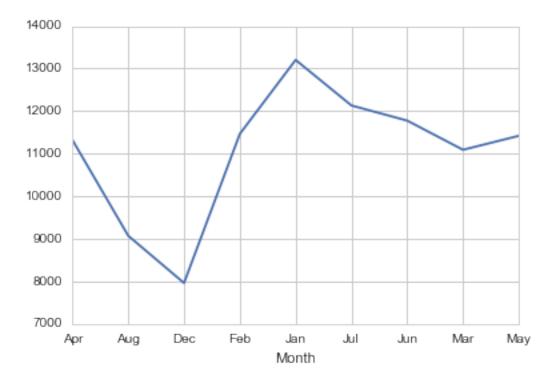
** Create a gropuby object called by Month, where you group the DataFrame by the month column and use the count () method for aggregation. Use the head () method on this returned DataFrame. **

Out[164]:		zip	title	timeStamp	twp	addr	Reason1	Reason	Hour	\
	Month									
	Apr	9895	11326	11326	11323	11283	11326	11326	11326	
	Aug	7832	9078	9078	9073	9025	9078	9078	9078	
	Dec	6907	7969	7969	7963	7916	7969	7969	7969	
	Feb	9930	11467	11467	11465	11396	11467	11467	11467	
	Jan	11527	13205	13205	13203	13096	13205	13205	13205	
		Day of	Week							
	Month									
	Apr		11326							
	Aug		9078							
	Dec		7969							
	Feb		11467							
	Jan		13205							

^{**} Now create a simple plot off of the dataframe indicating the count of calls per month. **

In [165]: byMonth['twp'].plot()

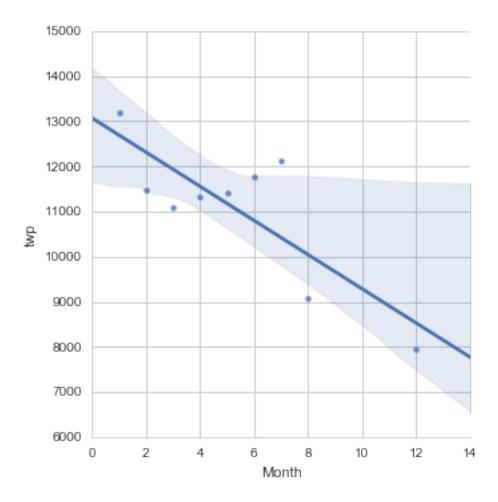
Out[165]: <matplotlib.axes._subplots.AxesSubplot at 0xe82a2c5320>



^{**} Now see if you can use seaborn's lmplot() to create a linear fit on the number of calls per month. Keep in mind you may need to reset the index to a column. In the Group by object Month is set as the index, the reset option makes month a column in the object **

In [123]: sns.lmplot(x='Month',y='twp',data=byMonth.reset_index())

Out[123]: <seaborn.axisgrid.FacetGrid at 0xe825eb5320>



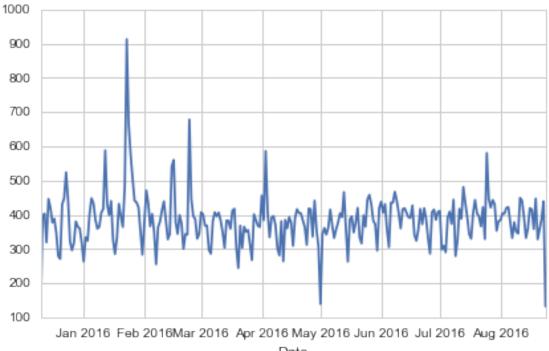
Create a new column called 'Date' that contains the date from the timeStamp column. You'll need to use apply along with the .date() method. There is a difference between timeStamp and date objects

```
In [166]: t = df['timeStamp'].iloc[0]
In [167]: t
Out[167]: Timestamp('2015-12-10 17:40:00')
In [168]: t.date()
Out[168]: datetime.date(2015, 12, 10)
In [169]: df['Date']=df['timeStamp'].apply(lambda t: t.date())
In [170]: df.head()
Out[170]:
                                        title
                                                        timeStamp
                zip
                     EMS: BACK PAINS/INJURY 2015-12-10 17:40:00
          0 19525.0
                                                                         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
                      EMS: CARDIAC EMERGENCY 2015-12-10 17:40:01
          3 19401.0
                                                                          NORRISTOWN
```

4	NaN	EMS: D	IZZINESS	2015-	12-10	17:4	0:01	LOWER	POTTSG	ROVE
		ad	dr		Reas	son1	Reason	Hour	Month	\
0	REINDEE	R CT & DEAD E	IND BAC	CK PAI	NS/INJ	JURY	EMS	17	Dec	
1	BRIAR PATH 8	k WHITEMARSH	LN DIAE	BETIC	EMERGE	ENCY	EMS	17	Dec	
2		HAWS A	VE	GAS-	ODOR/I	LEAK	Fire	17	Dec	
3	AIR	Y ST & SWEDE	ST CAF	RDIAC	EMERGE	ENCY	EMS	17	Dec	
4	CHERRYWOOI	O CT & DEAD E	ND		DIZZIN	IESS	EMS	17	Dec	
	Day of Week	Date								
0	Thu	2015-12-10								
1	Thu	2015-12-10								
2	Thu	2015-12-10								
3	Thu	2015-12-10								
4	Thu	2015-12-10								

^{**} Now groupby this Date column with the count() aggregate and create a plot of counts of 911 calls.**

In [172]: df.groupby('Date').count()['twp'].plot() plt.tight_layout()

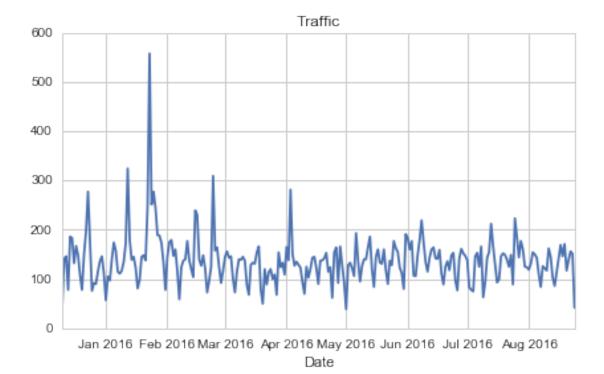


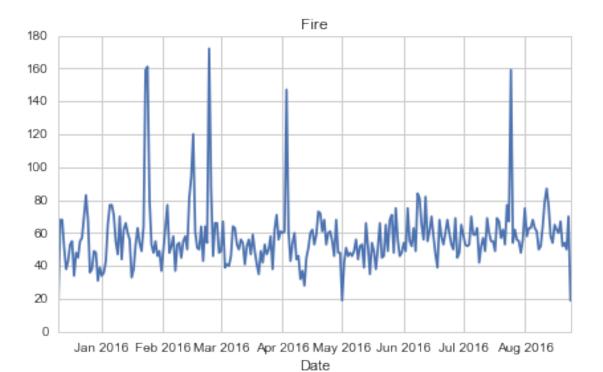
Date

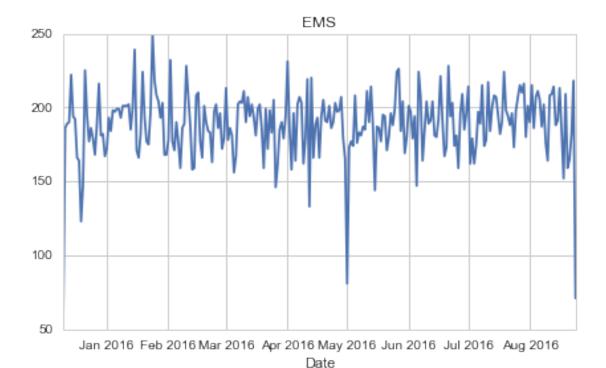
In [173]:	In [173]: df.groupby('Date').count().head()											
Out[173]:		zip	title	timeStamp	twp	addr	Reason1	Reason	Hour	Month	\	
	Date											
	2015-12-10	100	115	115	115	113	115	115	115	115		
	2015-12-11	333	396	396	395	391	396	396	396	396		
	2015-12-12	333	403	403	403	401	403	403	403	403		

2015-12-13	280	319	319	319	317	319	319	319	319
2015-12-14	387	447	447	446	445	447	447	447	447
	Day of	Week							
Date									
2015-12-10		115							
2015-12-11		396							
2015-12-12		403							
2015-12-13		319							
2015-12-14		447							

** Now recreate this plot but create 3 separate plots with each plot representing a Reason for the 911 call**







Creating heatmaps with seaborn. First group by a list of two variables and count option. Second, use the unstack option to create a crosstab Select on variable for counting

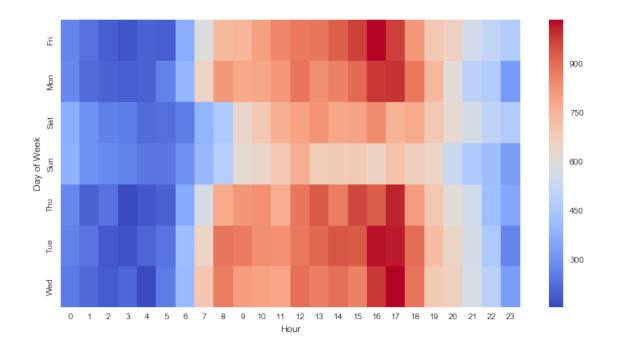
In [177]: df.groupby(['Day of Week', 'Hour']).count()

Out[177]:	D (zip	title	timeStamp	twp	addr	Reason1	Reason	Month	\
	•	Week Hour	040	075	075	075	075	075	075	075	
	Fri	0	248	275	275	275	275	275	275	275	
		1	200	235	235	235	232	235	235	235	
		2	165	191	191	191	191	191	191	191	
		3	164	175	175	175	175	175	175	175	
		4	184	201	201	201	201	201	201	201	
		5	166	194	194	194	194	194	194	194	
		6	319	372	372	372	369	372	372	372	
		7	526	598	598	598	593	598	598 740	598	
		8	637	742	742	742	737	742	742	742	
		9	663	752	752	752	748	752	752	752	
		10	722	803	803	803	800	803	803	803	
		11	756	859	859	859	858	859	859	859	
		12	764	885	885	885	877	885	885	885	
		13	767	890	890	890	885	890	890	890	
		14	808	932	932	931	926	932	932	932	
		15	840	980	980	980	976	980	980	980	
		16	897	1039	1039	1039	1038	1039	1039	1039	
		17	826	980	980	980	971	980	980	980	
		18	714	820	820	819	818	820	820	820	
		19	616	696	696	696	693	696	696	696	
		20	569	667	667	667	666	667	667	667	
		21 22	491	559 514	559 E14	558 514	553 513	559 514	559	559	
		23	445	514	514	514	513	514	514	514	
	Mon	0	400	474	474	474 282	469	474	474	474	
	Mon	1	243 198	282 221	282 221	202	282 221	282 221	282 221	282 221	
		2	183	201	201	201	201	201	201	201	
		3	173	194	194	194	194	194	194	194	
		4	185			204					
		5	222	204 267	204 267	267	203 267	204 267	204 267	204 267	
		3									
	Tue	18	761	905	905	905	900	905	905	905	
	Tue	19	666	731	731	731	722	731	731	731	
		20	579	647	647	647	646	647	647	647	
		21	494	571	571	570	565	571	571	571	
		22	411	462	462	462	460	462	462	462	
		23	245	274	274	274	273	274	274	274	
	Wed	0	226	250	250	249	249	250	250	250	
	#0a	1	193	216	216	216	216	216	216	216	
		2	169	189	189	189	188	189	189	189	
		3	195	209	209	207	209	209	209	209	
		4	138	156	156	156	155	156	156	156	
		5	224	255	255	254	254	255	255	255	
		J	227	200	200	204	204	200	200	200	

6	334	410	410	409	409	410	410	410
7	591	701	701	701	692	701	701	701
8	770	875	875	875	872	875	875	875
9	714	808	808	808	805	808	808	808
10	713	800	800	800	799	800	800	800
11	698	789	789	789	784	789	789	789
12	812	903	903	902	901	903	903	903
13	773	872	872	872	870	872	872	872
14	787	904	904	903	895	904	904	904
15	760	867	867	867	862	867	867	867
16	860	990	990	989	983	990	990	990
17	881	1037	1037	1036	1024	1037	1037	1037
18	771	894	894	894	891	894	894	894
19	590	686	686	686	682	686	686	686
20	597	668	668	668	662	668	668	668
21	508	575	575	574	572	575	575	575
22	432	490	490	490	485	490	490	490
23	294	335	335	334	335	335	335	335

				Date
Day	of	Week	Hour	
Fri			0	275
			1	235
			2	191
			3	175
			4	201
			5	194
			6	372
			7	598
			8	742
			9	752
			10	803
			11	859
			12	885
			13	890
			14	932
			15	980
			16	1039
			17	980
			18	820
			19	696
			20	667
			21	559
			22	514
			23	474
Mon			0	282
			1	221
			2	201
			3	194
			4	204
			5	267
 Tue			18	 905
rue			19	731
			13	131

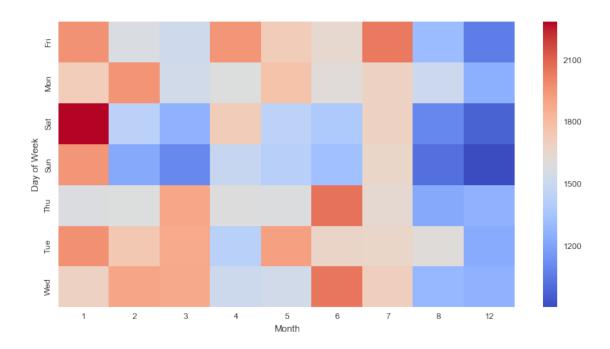
```
21
                                571
                        22
                                462
                        23
                                274
           Wed
                        0
                                250
                        1
                                216
                        2
                                189
                        3
                                209
                        4
                                156
                        5
                                255
                        6
                                410
                        7
                                701
                        8
                                875
                        9
                                808
                        10
                                800
                        11
                                789
                        12
                                903
                        13
                                872
                        14
                                904
                        15
                                867
                        16
                                990
                        17
                               1037
                        18
                                894
                        19
                                686
                        20
                                668
                        21
                                575
                        22
                                490
                        23
                                335
           [168 rows x 9 columns]
In [178]: dayHour = df.groupby(by=['Day of Week', 'Hour']).count()['Reason'].unstack()
           dayHour.head()
Out[178]: Hour
                           0
                                1
                                      2
                                           3
                                                 4
                                                       5
                                                             6
                                                                  7
                                                                        8
                                                                             9
                                                                                         14
                                                                                              15
                                                                                 . . .
           Day of Week
                                                                                 . . .
                                                      194
           Fri
                          275
                               235
                                     191
                                          175
                                                201
                                                           372
                                                                 598
                                                                       742
                                                                            752 ...
                                                                                       932
                                                                                             980
                          282
           Mon
                               221
                                     201
                                           194
                                                204
                                                      267
                                                           397
                                                                 653
                                                                       819
                                                                            786
                                                                                       869
                                                                                             913
                                                      231
                          375
                               301
                                                224
                                                           257
           Sat
                                     263
                                          260
                                                                 391
                                                                       459
                                                                            640
                                                                                        789
                                                                                             796
                                                                                . . .
           Sun
                          383
                               306
                                     286
                                          268
                                                242
                                                      240
                                                           300
                                                                 402
                                                                       483
                                                                            620
                                                                                        684
                                                                                             691
           Thu
                          278
                               202
                                     233
                                          159
                                                182
                                                      203
                                                           362
                                                                 570
                                                                       777
                                                                            828 ...
                                                                                        876
                                                                                             969
                                                                    23
           Hour
                            16
                                  17
                                        18
                                              19
                                                   20
                                                         21
                                                               22
           Day of Week
           Fri
                          1039
                                 980
                                       820
                                             696
                                                  667
                                                        559
                                                              514
                                                                   474
           Mon
                           989
                                 997
                                       885
                                             746
                                                  613
                                                        497
                                                              472
                                                                   325
                                             696
           Sat
                           848
                                 757
                                       778
                                                  628
                                                        572
                                                              506
                                                                   467
           Sun
                           663
                                 714
                                       670
                                             655
                                                  537
                                                        461
                                                              415
                                                                   330
           Thu
                           935
                                1013
                                            698
                                                        553
                                       810
                                                  617
                                                              424
                                                                   354
           [5 rows x 24 columns]
In [179]: plt.figure(figsize=(12,6))
           sns.heatmap(dayHour,cmap='coolwarm')
Out[179]: <matplotlib.axes._subplots.AxesSubplot at 0xe82a5aefd0>
```



** Table for Months. **

Out[105]: Month	1	2	3	4	5	6	7	8	12
Day of Week									
Fri	1970	1581	1525	1958	1730	1649	2045	1310	1065
Mon	1727	1964	1535	1598	1779	1617	1692	1511	1257
Sat	2291	1441	1266	1734	1444	1388	1695	1099	978
Sun	1960	1229	1102	1488	1424	1333	1672	1021	907
Thu	1584	1596	1900	1601	1590	2065	1646	1230	1266

Out[180]: <matplotlib.axes._subplots.AxesSubplot at 0xe82a695eb8>



In []: