

Advanced Analytics & Dashboard Design

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Exercise 6.1: Sourcing Open Data

Import Libraries

```
In [1]: import pandas as pd
import numpy as np
import numbers
import chart_studio
import plotly
from plotly.offline import init_notebook_mode, iplot
import chart_studio.plotly as py
import plotly.graph_objs as go
from plotly import tools
import folium
# from folium import plugins

init_notebook_mode(connected=True)
```

Load Data set and Set data frame

```
In [2]: gun_violence_df = pd.read_csv('gun-violence-data_01-2013_03-2018.csv')
```

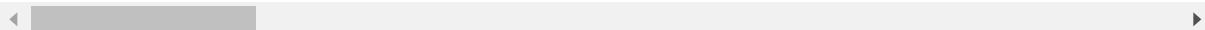
Glimpse of Data

```
In [3]: # head of data set  
gun_violence_df.head()
```

Out[3]:

	incident_id	date	state	city_or_county	address	n_killed	n_injured	
0	461105	2013-01-01	Pennsylvania	Mckeesport	1506 Versailles Avenue and Coursin Street	0	4	http://www.gunvic
1	460726	2013-01-01	California	Hawthorne	13500 block of Cerise Avenue	1	3	http://www.gunvio
2	478855	2013-01-01	Ohio	Lorain	1776 East 28th Street	1	3	http://www.gunvio
3	478925	2013-01-05	Colorado	Aurora	16000 block of East Ithaca Place	4	0	http://www.gunvio
4	478959	2013-01-07	North Carolina	Greensboro	307 Mourning Dove Terrace	2	2	http://www.gunvio

5 rows × 29 columns



```
In [4]: # Last values of the data set
gun_violence_df.tail()
```

Out[4]:

	incident_id	date	state	city_or_county	address	n_killed	n_injured	
239672	1083142	2018-03-31	Louisiana	Rayne	North Riceland Road and Highway 90	0	0	http://www.gu
239673	1083139	2018-03-31	Louisiana	Natchitoches	247 Keyser Ave	1	0	http://www.gu
239674	1083151	2018-03-31	Louisiana	Gretna	1300 block of Cook Street	0	1	http://www.gu
239675	1082514	2018-03-31	Texas	Houston	12630 Ashford Point Dr	1	0	http://www.gu
239676	1081940	2018-03-31	Maine	Norridgewock	434 Skowhegan Rd	2	0	http://www.gu

5 rows × 29 columns

Statistical Overview of the Data

```
In [5]: gun_violence_df.describe() ##describes only numeric data
```

Out[5]:

	incident_id	n_killed	n_injured	congressional_district	latitude	lc
count	2.396770e+05	239677.000000	239677.000000	227733.000000	231754.000000	231754
mean	5.593343e+05	0.252290	0.494007	8.001265	37.546598	-89.250000
std	2.931287e+05	0.521779	0.729952	8.480835	5.130763	14.500000
min	9.211400e+04	0.000000	0.000000	0.000000	19.111400	-171.000000
25%	3.085450e+05	0.000000	0.000000	2.000000	33.903400	-94.000000
50%	5.435870e+05	0.000000	0.000000	5.000000	38.570600	-86.000000
75%	8.172280e+05	0.000000	1.000000	10.000000	41.437375	-80.000000
max	1.083472e+06	50.000000	53.000000	53.000000	71.336800	97.000000

The information regarding the numerical columns of the statistics on gun violence is described in the table that was presented earlier. Because the information is only provided for the numeric columns, and there is no information provided about the data that is missing, we have developed a more in-depth tool that will describe the information for all of the attributes below.

Check for Missing Data


```

In [6]: # Function to describe more information for all the attributes
def brief(data):

    df = data.copy()

    print("This dataset has {} Rows {}".format(df.shape[0],df.shape[1]))
    print('\n')

    real_valued = {}
    symbolics = {}

    for i,col in enumerate(df.columns, 1):
        Missing = len(df[col]) - df[col].count()

        counter = 0
        for val in df[col].dropna():
            if isinstance(val, numbers.Number):
                counter += 1

        if counter != len(df[col].dropna()):
            arity = len(df[col].dropna().unique())
            symbolics[i] = [i, col, Missing, arity]
        else:
            Mean, Median, Sdev, Min, Max = df[col].mean(), df[col].median(), df[col].std(), df[col].min(), df[col].max()
            real_valued[i] = [i, col, Missing, Mean, Median, Sdev, Min, Max]

    #Create array containing list of real valued
    real_valued_array = [real_valued[keys] for keys in real_valued.keys()]
    real_valued_transformed = np.array(real_valued_array).T

    symbolic_array = [symbolics[keys] for keys in symbolics.keys()]
    symbolic_transformed = np.array(symbolic_array).T

    # return symbolic_transformed
    real_cols = ['Attribute_ID', 'Attribute_Name', 'Missing', 'Mean', 'Median']
    sym_cols = ['Attribute_ID', 'Attribute_Name', 'Missing', 'arity']

    index = range(1, len(real_valued.keys())+1)
    real_val_df = pd.DataFrame(data={unit[0]:unit[1] for unit in zip(real_cols, real_valued_transformed)})

    index_sym = range(1, len(symbolics.keys())+1)
    sym_val_df = pd.DataFrame(data={unit[0]:unit[1] for unit in zip(sym_cols, symbolic_transformed)})

    text = ("real valued attributes" + "\n" + "-----"
            + "\n" + str(real_val_df) + "\n" + "non-real valued attributes"
            + "\n" + "-----" + "\n" + str(sym_val_df))

    return text

```



```
In [7]: %time
print(brief(gun_violence_df))
```

CPU times: total: 0 ns

Wall time: 0 ns

This dataset has 239677 Rows 29 Attributes

real valued attributes

```
-----
Attribute_ID      Attribute_Name  Missing      Mean \
1                1      incident_id      0      559334.3464037017
2                6      n_killed      0      0.25228953967214207
3                7      n_injured      0      0.4940065171042695
4               10 incident_url_fields_missing      0      0.0
5               11 congressional_district      11944      8.001264638853394
6               15      latitude      7923      37.54659822311588
7               17      longitude      7923      -89.33834822915676
8               18      n_guns_involved      99451      1.3724416299402393
9               28      state_house_district      38772      55.44713172892661
10              29      state_senate_district      32335      20.477110281563792
```

```
Median      Sdev      Min      Max
1  543587.0  293128.684285221  92114  1083472
2      0.0    0.52177887298012      0      50
3      0.0    0.7299522740842754      0      53
4      0.0      0.0      False      False
5      5.0    8.480834796700318      0.0      53.0
6   38.5706   5.130763162136701   19.1114   71.3368
7  -86.2496   14.35954557699743  -171.429   97.4331
8      1.0    4.678202195031997      1.0    400.0
9     47.0    42.04811689079994      1.0    901.0
10    19.0    14.20455963079257      1.0     94.0
```

non-real valued attributes

```
-----
Attribute_ID      Attribute_Name  Missing  arity
1                2      date      0    1725
2                3      state      0     51
3                4      city_or_county      0   12898
4                5      address      16497  198037
5                8      incident_url      0   239677
6                9      source_url      468   213989
7               12      gun_stolen      99498    349
8               13      gun_type      99451   2502
9               14 incident_characteristics      326   18126
10              16      location_description      197588  27595
11              19      notes      81017  136652
12              20      participant_age      92298   18951
13              21      participant_age_group      42119    898
14              22      participant_gender      36362    873
15              23      participant_name      122253  113488
16              24      participant_relationship      223903    284
17              25      participant_status      27626   2150
18              26      participant_type      24863    259
19              27      sources      609   217280
```


Based on the analysis presented above, you can deduce that certain properties, such as `participant_name` and `participant_relationship`, are missing almost as many values as the total number of records contained in the dataset.

In [8]: `gun_violence_df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 239677 entries, 0 to 239676
Data columns (total 29 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   incident_id                          239677 non-null int64
1   date                                239677 non-null object
2   state                              239677 non-null object
3   city_or_county                     239677 non-null object
4   address                            223180 non-null object
5   n_killed                           239677 non-null int64
6   n_injured                           239677 non-null int64
7   incident_url                        239677 non-null object
8   source_url                          239209 non-null object
9   incident_url_fields_missing         239677 non-null bool
10  congressional_district               227733 non-null float64
11  gun_stolen                          140179 non-null object
12  gun_type                            140226 non-null object
13  incident_characteristics             239351 non-null object
14  latitude                            231754 non-null float64
15  location_description                 42089 non-null object
16  longitude                           231754 non-null float64
17  n_guns_involved                     140226 non-null float64
18  notes                               158660 non-null object
19  participant_age                      147379 non-null object
20  participant_age_group                197558 non-null object
21  participant_gender                   203315 non-null object
22  participant_name                     117424 non-null object
23  participant_relationship              15774 non-null object
24  participant_status                   212051 non-null object
25  participant_type                     214814 non-null object
26  sources                              239068 non-null object
27  state_house_district                 200905 non-null float64
28  state_senate_district                207342 non-null float64
dtypes: bool(1), float64(6), int64(3), object(19)
memory usage: 51.4+ MB
```

I contribute further to the analysis that was done earlier by providing additional information above. Given the facts shown above, it is very evident that some of the data will require some sort of cleaning.

Data Cleaning

```
In [9]: # added important missing data point found in the description on Kaggle
missing = ['sban_1', '2017-10-01', 'Nevada', 'Las Vegas', 'Mandalay Bay 3950 I
        '-115.171667', 47, 'Route 91 Harvest Festiva; concert, open fire f
gun_violence_df.loc[len(gun_violence_df)] = missing

print(gun_violence_df.shape)
drop_columns = gun_violence_df.columns[gun_violence_df.apply(lambda col: col.i
gun_violence_filtered = gun_violence_df.drop(drop_columns, axis=1)
print(gun_violence_filtered.shape)
print('Dropped Columns:', list(drop_columns))
```

(239678, 29)
(239678, 26)
Dropped Columns: ['location_description', 'participant_name', 'participant_re
lationship']

In []: