Police Dataset

Here,

The data from a Police Check Post is given.

The data is available as a CSV file. I am going to analyze this data set using the Pandas DataFrame.

In [2]:

import pandas as pd

In [3]:

data = pd.read_csv(r'C:\Users\harsh\Desktop\Projects\Python\Police Dataset Analysis\Police Data.csv')

In [4]:

data

Out[4]:

	stop_date	stop_time	country_name	driver_gender	driver_age_raw	driver_age	driver_race	vic			
0	1/2/2005	1:55	NaN	M	1985.0	20.0	White				
1	1/18/2005	8:15	NaN	M	1965.0	40.0	White				
2	1/23/2005	23:15	NaN	M	1972.0	33.0	White				
3	2/20/2005	17:15	NaN	M	1986.0	19.0	White	Call			
4	3/14/2005	10:00	NaN	F	1984.0	21.0	White				
					•••						
65530	12/6/2012	17:54	NaN	F	1987.0	25.0	White				
65531	12/6/2012	22:22	NaN	M	1954.0	58.0	White				
65532	12/6/2012	23:20	NaN	М	1985.0	27.0	Black	Equipmen			
65533	12/7/2012	0:23	NaN	NaN	NaN	NaN	NaN				
65534	12/7/2012	0:30	NaN	F	1985.0	27.0	White				
65535	65535 rows × 15 columns										

Q1) Remove the columns that only contain missing values.

↓

```
In [5]:
data.isnull().sum()
Out[5]:
stop date
                             0
stop_time
                        65535
country_name
driver_gender
                         4061
                         4054
driver_age_raw
driver_age
                         4307
driver_race
                         4060
violation_raw
                         4060
violation
                         4060
search_conducted
                             0
search_type
                        63056
                         4060
stop_outcome
is_arrested
                         4060
stop_duration
                         4060
                            0
drugs_related_stop
dtype: int64
In [6]:
data.drop(columns = 'country_name', inplace = True)
In [7]:
data.head(2)
Out[7]:
   stop_date stop_time driver_gender driver_age_raw driver_age driver_race violation_raw violation seai
    1/2/2005
                  1:55
                                            1985.0
                                                        20.0
                                                                  White
                                                                            Speeding
                                                                                     Speeding
    1/18/2005
                  8:15
                                            1965.0
                                                        40.0
                                                                  White
                                 Μ
                                                                            Speeding
                                                                                     Speeding
```

Q2) For speeding, were Men or Women stopped more often?

```
In [8]:

data[data.violation == 'Speeding'].driver_gender.value_counts()

Out[8]:

M    25517
F    11686
Name: driver_gender, dtype: int64
```

Q3) Does gender affect who gets searched during a stop?

```
In [9]:
```

```
data.groupby('driver_gender').search_conducted.sum()
```

Out[9]:

driver_gender
F 366
M 2113

Name: search_conducted, dtype: int64

Q4) What is the mean stop_duration?

In [10]:

```
data.head()
```

Out[10]:

r	driver_age_raw	driver_age	driver_race	violation_raw	violation	search_conducted	search_type	stop_outco
1	1985.0	20.0	White	Speeding	Speeding	False	NaN	Citat
1	1965.0	40.0	White	Speeding	Speeding	False	NaN	Citat
1	1972.0	33.0	White	Speeding	Speeding	False	NaN	Citat
1	1986.0	19.0	White	Call for Service	Other	False	NaN	Arrest Dri
=	1984.0	21.0	White	Speeding	Speeding	False	NaN	Citat
4	(•

In [11]:

data.stop_duration.value_counts()

Out[11]:

0-15 Min 47379 16-30 Min 11448 30+ Min 2647 2 1

Name: stop_duration, dtype: int64

In [12]:

```
data['stop_duration'] = data['stop_duration'].map( {'0-15 Min': 7.5,'16-30 Min': 23, '30+ Min': 45
```

```
In [13]:
```

data

Out[13]:

	stop_date	stop_time	driver_gender	driver_age_raw	driver_age	driver_race	violation_raw	Vic			
0	1/2/2005	1:55	М	1985.0	20.0	White	Speeding	Sp			
1	1/18/2005	8:15	M	1965.0	40.0	White	Speeding	Sp			
2	1/23/2005	23:15	M	1972.0	33.0	White	Speeding	Sp			
3	2/20/2005	17:15	M	1986.0	19.0	White	Call for Service				
4	3/14/2005	10:00	F	1984.0	21.0	White	Speeding	Sp			
•••											
65530	12/6/2012	17:54	F	1987.0	25.0	White	Speeding	Sp			
65531	12/6/2012	22:22	M	1954.0	58.0	White	Speeding	Sp			
65532	12/6/2012	23:20	М	1985.0	27.0	Black	Equipment/Inspection Violation	Equ			
65533	12/7/2012	0:23	NaN	NaN	NaN	NaN	NaN				
65534	12/7/2012	0:30	F	1985.0	27.0	White	Speeding	Sp			
65535 rows × 14 columns											
4								•			

In [14]:

data['stop_duration'].mean()

Out[14]:

12.001195627419722

Q5) Compare the age distributions for each violation.

In [15]:

data.groupby('violation').driver_age.describe()

Out[15]:

	count	mean	std	min	25%	50%	75%	max
violation								
Equipment	6507.0	31,682957	11,380671	16.0	23.0	28.0	39.0	81.0
Moving violation	11876.0	36.736443	13.258350	15.0	25.0	35.0	47.0	86.0
Other	3477.0	40.362381	12.754423	16.0	30.0	41.0	50.0	86.0
Registration/plates	2240.0	32.656696	11.150780	16.0	24.0	30.0	40.0	74.0
Seat belt	3.0	30.333333	10.214369	23.0	24.5	26.0	34.0	42.0
Speeding	37120.0	33.262581	12.615781	15.0	23.0	30.0	42.0	88.0