Project - Dallas Shootings

Goal of Project

- A newspaper want a visualization of the shootings in Dallas with focus on subjects
- We will read data from from database and join into broader datasets
- We will explore ideas to visualize it and create a map with the shootings

Step 1: Acquire

- Explore problem
- Identify data
- Import data

Step 1.a: Import libraries

• Execute the cell below (SHIFT + ENTER)

```
import sqlite3
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
```

Step 1.b: Connect to SQLite database

 Use sqlite3.connect(<filename>) to connect to the database files/dallasois.sqlite

```
In [17]: conn = sqlite3.connect('files/dallas-ois.sqlite')
```

Step 1. c: Read data into DataFrames

- The database consists of 3 tables: incident, officers, subjects
- Read the data from each table into a DataFrame.
- HINT: Use pandas read_sql(sql_stmt, conn), which takes the SQL statement
 (sql_stmt) and the data base connection conn from previous step.
- HINT: The structure of the SQL statement is as follows.

```
SELECT * FROM table
```

```
incidents = pd.read_sql('SELECT * FROM incidents', conn)
officers = pd.read_sql('SELECT * FROM officers', conn)
subjects = pd.read_sql('SELECT * FROM subjects', conn)
```

Step 1.d: Explore the length of the DataFrames

- What is the length of the DataFrames
 - HINT: Apply len(...) on the DataFrames
- We want to explore data based on officers and data based on subjects, both with incident data.
- Notice: It is difficult to create one dataset for both problems
- Explore data further to understand why

In [19]:	len(ir	ncidents)	, len(officers),	len(subjects)				
Out[19]:	(219,	370, 223)						
In [20]:	incide	ents.head	l()						
Out[20]:	case	e_number	date	location	subject_statuses	subject_weapon	subjects	subject_count	0
	0	44523A	2013- 02-23	3000 Chihuahua Street	Injured	Handgun	Curry, James L/M	1	N Fil Briaı
	1	121982X	2010- 05-03	1300 N. Munger Boulevard	Injured	Handgun	Chavez, Gabriel L/M	1	I
	2	605484T	2007- 08-12	200 S. Stemmons Freeway	Other	Shotgun	Salinas, Nick L/M	1	F Jerr _!
	3	384832T	2007- 05-26	7900 S. Loop 12	Shoot and Miss	Unarmed	Smith, James B/M; Dews, Antonio B/M; Spearman,	3	N N
	4	244659R	2006- 04-03	6512 South Loop 12	Injured	Hands	Watkins, Caleb B/M	1	Arms N
4									•

Step 1.e: Read data into dataset

- Create first dataset subject_incidents as officers joined with incidents.
 - What does this dataset give us?
 - HINT:

SELECT * FROM table_1 JOIN table_2 ON
table_1.column_name_1=table_2.column_name_2

- HINT: You can join on columns case_number
- Is all data represented?

In [21]:	<pre>subjects_incidents = pd.read_sql('SELECT * FROM subjects s JOIN incidents i ON s.case</pre>										
In [22]:	len(s	ubjects_i	ncide	ents)							
Out[22]:	223										
In [23]:	subje	cts_incid	lents.	head()							
Out[23]:	cas	se_number	race	gender	last_name	first_name	full_name	case_number	date	location	sul
	0	44523A	L	М	Curry	James	Curry, James	44523A	2013- 02-23	3000 Chihuahua Street	
	1	121982X	L	М	Chavez	Gabriel	Chavez, Gabriel	121982X	2010- 05-03	1300 N. Munger Boulevard	
	2	605484T	L	М	Salinas	Nick	Salinas, Nick	605484T	2007- 08-12	200 S. Stemmons Freeway	
	3	384832T	В	М	Smith	James	Smith, James	384832T	2007- 05-26	7900 S. Loop 12	S
	4	384832T	В	М	Dews	Antonio	Dews, Antonio	384832T	2007- 05-26	7900 S. Loop 12	S
	5 rows	× 21 coluı	mns								
4											•

Step 2: Prepare

- Explore data
- Visualize ideas
- Cleaning data

Step 2.a: Check the data types

- A step to get to understand the data better is to explore the data types
- Get the data types by .dtypes

subjects_incidents.dtypes		
4]: case_number	object	
race	object	
gender	object	
last_name	object	
first_name	object	
full_name	object	
case_number	object	
date	object	
location	object	
subject_statuses	object	
subject_weapon	object	
subjects	object	
subject_count	int64	
officers	object	
officer_count	int64	
<pre>grand_jury_disposition</pre>	object	
attorney_general_forms_url	object	
summary_url	object	
summary_text	object	
latitude	float64	
longitude	float64	
dtype: object		

Step 2.b: Check for null (missing) values

- Data often is missing entries there can be many reasons for this
- We need to deal with that (will do later in course)
- Use .isna().sum()

```
In [25]: subjects_incidents.isnull().sum()
```

```
0
          case_number
Out[25]:
                                            0
          race
          gender
                                            0
          last_name
          first name
                                           18
          full_name
                                           0
          case number
          date
          location
          subject statuses
          subject_weapon
                                            0
          subjects
          subject_count
                                            0
          officers
          officer count
                                           0
          grand_jury_disposition
                                           88
          attorney_general_forms_url
                                          221
          summary_url
                                           3
          summary_text
                                            3
                                            9
          latitude
                                            9
          longitude
          dtype: int64
```

Step 2.c: Explore subject_statuses column

- As part of finding useful features let's explore subject_statuses
- We know that subject_statuses is categorical.
- Therefore we can use groupby(...) and count()

In [26]:	subjects_incid	dents.groupby	('sub	oject_st	atuses').	count()			
Out[26]:		case_number	race	gender	last_name	first_name	full_name	case_number	date l
	subject_statuses								
	1 Deceased 1 Injured	2	2	2	2	2	2	2	2
	2 Injured	1	1	1	1	1	1	1	1
	Deceased	69	69	69	69	67	69	69	69
	Deceased Injured	2	2	2	2	2	2	2	2
	Injured	60	60	60	60	60	60	60	60
	Other	2	2	2	2	2	2	2	2
	Shoot and Miss	87	87	87	87	71	87	87	87
4									

Step 2.d: Explore race column

• Repeat of previous step on column race

```
In [27]: subjects_incidents.groupby('race').count()
```

Out[27]:		case_number	gender	last_name	first_name	full_name	case_number	date	location	subject_s
	race									
	Α	2	2	2	2	2	2	2	2	
	В	111	111	111	104	111	111	111	111	
	L	72	72	72	61	72	72	72	72	
	W	38	38	38	38	38	38	38	38	
4										•

Step 2.e: Explore more columns

• Feel free to explore more columns (also called features)

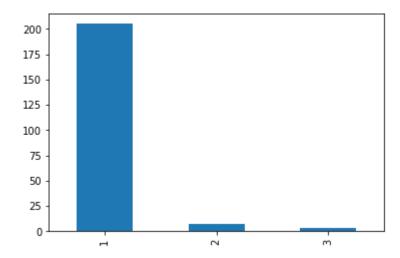
Step 2.f: Visualize ideas

- We want to make a visual plot of the shooting incident
- Let's explore if we can make a plot based on longitude and latitude.
 - HINT: use plot.scatter(x='longitude', y='latitude')
- Then the analysis phase will focus on how to make useful insights with the feature selections.

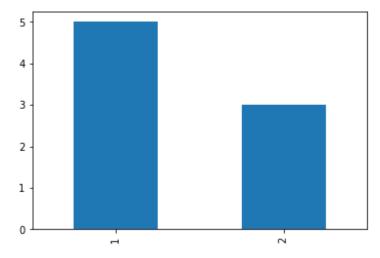
```
200 -
150 -
100 -
50 -
2
```

In [53]: subjects_incidents['subject_count'][subjects_incidents['gender']=='M'].value_counts().

Out[53]: <AxesSubplot:>

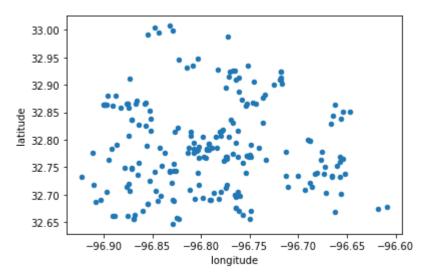


In [54]: subjects_incidents['subject_count'][subjects_incidents['gender']=='F'].value_counts().
Out[54]:



In [28]: subjects_incidents.plot.scatter(x='longitude', y='latitude')

Out[28]: <AxesSubplot:xlabel='longitude', ylabel='latitude'>



Step 3: Analyze

- Feature selection
- Model selection
- Analyze data

Step 3.a: Feature selection

- Here we will continue with the features selected feel free to explore other features.
- Create a dataset with the features: race, subject_statuses, latitude,
 longitude
 - HINT: Select the features of the DataFrame subject_incidents by filtering with a list of the columns.
- To make further processing easier apply dropna() to remove missing data.

```
In [29]: dataset = subjects_incidents[['race', 'subject_statuses', 'latitude', 'longitude']].dr
```

```
In [30]: len(dataset)
Out[30]: 214
```

Step 3.b: How to visualize features

- We want to visualize the two features: race and subject statuses
- A way to visualize data is by color and size.
- Idea:
 - Map the race features to colors
 - Map the subject_statuses to a size

Step 3.b.1: Convert column

- The race column has the following categories: B, W, A, L
- We can map that to color values (docs)
- A simple way to map columns is by using apply on a lambda-function.
 - Create a dict with the mapping:

```
mapping = {'B': 'blue', 'W': 'yellow', 'A': 'red', 'L': 'cyan'}
```

Do the mapping with apply and lambda as follows

```
dataset['race'] = dataset['race'].apply(lambda x: mapping[x])
```

```
In [31]: mapping = {'B': 'blue', 'W': 'yellow', 'A': 'red', 'L': 'cyan'}
    dataset['race'] = dataset['race'].apply(lambda x: mapping[x])
In [32]: dataset.head()
```

```
2... [52]
```

Out[32]:

		race	subject_statuses	latitude	longitude
	9	cyan	Deceased	32.68642	-96.908674
	10	blue	Deceased	32.86400	-96.898998
	11	yellow	Shoot and Miss	32.81482	-96.826787
	12	blue	Injured	32.77540	-96.767489
	13	blue	Shoot and Miss	32.74417	-96.828470

Step 3.b.2: Convert column

- The subject_statuses has the following categories: '1 Deceased 1 Injured', '2 Injured', 'Deceased', 'Deceased Injured', 'Injured', 'Other', 'Shoot and Miss'
- The main categories are: 'Deceased', 'Injured', 'Shoot and Miss'
- A simple way is quite similar to last step.
 - Create a mapping of the main categories:

```
mapping = {'Deceased': 1000, 'Injured': 500, 'Shoot and Miss': 250}
```

Do the mapping with apply and lambda as follows

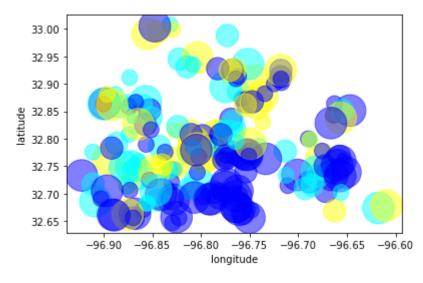
```
dataset['subject_statuses'] =
dataset['subject_statuses'].apply(lambda x: mapping.get(x, 100))
```

```
In [33]: mapping = {'Deceased': 1000, 'Injured': 500, 'Shoot and Miss': 250}
dataset['subject_statuses'] = dataset['subject_statuses'].apply(lambda x: mapping.get()
```

Step 3.c: Visualize the data

- Now we can visualize the data
- This can be done with a scatter plot dataset.plot.scatter(x='longitude', y='latitude', s='subject_statuses', c='race', alpha=.5)
- Where s= is the size feature and c= is the color feature, figsize= sets the size of the figure, alpha= sets the transparency of the dots.

```
In [34]: dataset.plot.scatter(x='longitude', y='latitude', s='subject_statuses', c='race', alph
Out[34]: <AxesSubplot:xlabel='longitude', ylabel='latitude'>
```



Step 4.a: Present a chart

- The goal here is to present your message
- Visualize one chart
- Add a headline (title) to give the audience a message

```
In [9]: pip install folium
```

Requirement already satisfied: folium in c:\users\kingofkings\anaconda3\lib\site-pack ages (0.12.1.post1)

Requirement already satisfied: numpy in c:\users\kingofkings\anaconda3\lib\site-packa ges (from folium) (1.21.5)

Requirement already satisfied: requests in c:\users\kingofkings\anaconda3\lib\site-pa ckages (from folium) (2.27.1)

Requirement already satisfied: jinja2>=2.9 in c:\users\kingofkings\anaconda3\lib\site -packages (from folium) (2.11.3)

Requirement already satisfied: branca>=0.3.0 in c:\users\kingofkings\anaconda3\lib\si te-packages (from folium) (0.5.0)

Requirement already satisfied: MarkupSafe>=0.23 in c:\users\kingofkings\anaconda3\lib \site-packages (from jinja2>=2.9->folium) (2.0.1)

Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\kingofkings\anaconda 3\lib\site-packages (from requests->folium) (1.26.9)

Requirement already satisfied: certifi>=2017.4.17 in c:\users\kingofkings\anaconda3\l ib\site-packages (from requests->folium) (2021.10.8)

Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\kingofkings\anac onda3\lib\site-packages (from requests->folium) (2.0.4)

Requirement already satisfied: idna<4,>=2.5 in c:\users\kingofkings\anaconda3\lib\sit e-packages (from requests->folium) (3.3)

Note: you may need to restart the kernel to use updated packages.

```
In [40]: import folium
```

```
In [41]: m = folium.Map(location=[32.8, -96.8])
```

In [42]:

Out[42]:



fill_color=row['race']
).add_to(m)

In [44]:

Out[44]:

