

Mile stone

Team Name: 29DataSets

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Repository: <https://github.com/Bonampak1/29DataSets>
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Trends of crime data over the years across Utah cities

Background and Motivation (Revised after feedback)

The idea for this project came after some thought about using a combination of the datasets with an Opioid dataset to evaluate if there were any specific trends involved with Opioid use and crime. We would use the location and years and see if there was an overlap. The revised project idea is to see if there are general trends of crime data within specific cities and specific years and crime per day and time and what type of crime per day and build visualizations on that data. We were both intersted in this topic because we were both from Utah but did not know much about the crime in UT.

New Hypothesis: Does a specific day and time of day impact what time of crime is committed? Does the location also impact this result?

Data

We will be using data from <https://opendata.utah.gov/> (<https://opendata.utah.gov/>), because there were multiple dataset CSVs we have to use, they were all downloaded and uploaded into our repository from above. The data is in the form of 29 CSV files.

Data Processing

We do anticipate some data clean up with these files. For example, not all the files have the same columns, in the case of 'Cache_County_Sheriff_Police_Crime_Data' and 'Brigham_City_Police_Crime_Data', the 'Cache_County' CSV file contains an extra column 'country'. On that note, some of the CSV files are per county and others are per city. If we plan to do trends across different counties, we would have to merge the data accordingly on city per county. There is also a 'state' column which we likely not need because the datasets are all for the state of Utah. There are also some 'NAs' on Zip Codes that will have to be accounted for and we will have to align the years across all 29 datasets so that we have coverage of data across those years.

Common columns:

- Incident_id (except SLC)
- Case_number(except SLC)
- incident_datetime (except SLC)
- incident_type_primary (except SLC)
- incident_description (except SLC)
- clearance_type (except SLC)
- address_1 (except SLC)
- city (except SLC)
- state (except SLC)
- Zip (except SLC & Syracuse)
- Latitude (except SLC)
- Longitude (except SLC)
- created_at (except SLC)
- updated_at (except SLC)
- location (except SLC)
- hour_of_day (except SLC)
- day_of_week (except SLC)
- parent_incident_type (except SLC)

Columns to get rid of:

- Clearance Type (did not know what this was and is NA in 3-4 datasets)
- State (it is Utah for all the data)
- Zip code (since we have latitude and longitude and because some files do not have this column)

DATA MERGING

We will need to merge all the county data into one dataframe to use for analytics and visualizations

```
In [46]: # import libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import altair as alt
```

```
In [47]: # 1.Read in data
st_g1 = pd.read_csv('./1.Saint_George_Police_Data_20231018.csv', dtype={
st_g1.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 99999 entries, 0 to 99998
Data columns (total 21 columns):
#   Column                                     Non-Null
Count Dtype
---  ---
-----
0    incident_id                             99999 no
n-null int64
1    case_number                             99999 no
n-null object
2    incident_datetime                       99999 no
n-null object
3    incident_type_primary                   99999 no
n-null object
4    incident_description                    99975 no
n-null object
5    clearance_type                          0 non-nu
ll      float64
6    address_1                             99978 no
n-null object
7    address_2                              0 non-nu
ll      float64
8    city                                   99999 no
n-null object
9    state                                  99999 no
n-null object
10   zip                                    25454 no
n-null object
11   country                               0 non-nu
ll      float64
12   latitude                             99998 no
n-null float64
13   longitude                             99998 no
n-null float64
14   created_at                            99994 no
n-null object
15   updated_at                            99999 no
n-null object
16   location                              99984 no
n-null object
17   hour_of_day                           99999 no
n-null int64
18   day_of_week                           99999 no
n-null object
19   parent_incident_type                   99999 no
n-null object
20   St George Police Department Districts Shapes - qdt2-uyjz 79156 no
n-null float64
dtypes: float64(6), int64(2), object(13)
memory usage: 16.0+ MB
```

```
In [48]: # Read in data
st_g2 = pd.read_csv('./2.Saint_George_Police_Data_20231018.csv')
st_g2.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 90243 entries, 0 to 90242
Data columns (total 21 columns):
 #   Column                                Non-Null
Count Dtype
---  ---
0    incident_id                        90243 no
n-null int64
1    case_number                        90243 no
n-null object
2    incident_datetime                 90243 no
n-null object
3    incident_type_primary             90243 no
n-null object
4    incident_description               90144 no
n-null object
5    clearance_type                    0 non-nu
ll    float64
6    address_1                         90228 no
n-null object
7    address_2                         0 non-nu
ll    float64
8    city                              90243 no
n-null object
9    state                             90243 no
n-null object
10   zip                               32649 no
n-null object
11   country                           0 non-nu
ll    float64
12   latitude                          90243 no
n-null float64
13   longitude                         90243 no
n-null float64
14   created_at                        90243 no
n-null object
15   updated_at                       90243 no
n-null object
16   location                          90240 no
n-null object
17   hour_of_day                       90243 no
n-null int64
18   day_of_week                       90243 no
n-null object
19   parent_incident_type              90243 no
n-null object
20   St George Police Department Districts Shapes - qdt2-uyjz 59944 no
n-null float64
dtypes: float64(6), int64(2), object(13)
memory usage: 14.5+ MB
```



```
In [51]: # 2.Read in data
beaver = pd.read_csv('./Beaver_County_Police_Crime_Data_20231011.csv')
beaver.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2676 entries, 0 to 2675
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed Column                        2676 non-null   int64
1   address_1                            2653 non-null   object
2   case_number                          2676 non-null   int64
3   city                                 2676 non-null   object
4   clearance_type                       0 non-null      float64
5   country                             1 non-null      object
6   created_at                           2676 non-null   object
7   day_of_week                          2676 non-null   object
8   hour_of_day                          2676 non-null   int64
9   incident_datetime                   2676 non-null   object
10  incident_description                  2676 non-null   object
11  incident_id                          2676 non-null   int64
12  incident_type_primary                 2676 non-null   object
13  latitude                             2676 non-null   float64
14  location                             2676 non-null   object
15  longitude                            2676 non-null   float64
16  parent_incident_type                 2676 non-null   object
17  state                                2676 non-null   object
18  updated_at                           2676 non-null   object
19  zip                                  2514 non-null   float64
dtypes: float64(4), int64(4), object(12)
memory usage: 418.2+ KB
```

```
In [52]: # Re-arrange columns
beaver = beaver[['incident_id', 'case_number', 'incident_datetime', 'parent_incident_type', 'incident_type_primary', 'incident_type_secondary', 'incident_description', 'address_1', 'city', 'latitude', 'longitude', 'location', 'hour_of_day', 'day_of_week', 'created_at', 'updated_at']]
beaver.head()
```

Out[52]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary	incident_description	address_1	city	latitude	longitude	location	hour_of_day	day_of_week	created_at	updated_at
0	45510703	36871	8/31/2010 0:00	Theft		0	A									
1	834901401	55101	2/10/2018 13:22	Traffic	PI Accident											
2	834901390	55112	2/12/2018 14:08	Traffic	Traffic Control											
3	834901372	55130	2/15/2018 10:54	Traffic	Livestock Probl		L									
4	834901370	55132	2/15/2018 23:54	Community Policing	Suspicious											

```
In [53]: # Look at the unique values of "clearance_type" to decide if we want the
# unique_values_clearance_type = pd.unique(beaver['clearance_type'])
# unique_values_clearance_type
```

```
In [54]: # Look at the unique values of "parent_incident_type" to decide if we want
unique_values_parent_incident_type = pd.unique(beaver['parent_incident_type'])
unique_values_parent_incident_type
```

```
Out[54]: array(['Theft', 'Traffic', 'Community Policing', 'Other',
'Weapons Offense', 'Other Sexual Offense', 'Disorder', 'Drugs',
'Alarm', 'Property Crime', 'Fire', 'Emergency', 'Family Offense',
'Liquor', 'Breaking & Entering', 'Missing Person', 'Assault',
'Theft of Vehicle', 'Robbery', 'Vehicle Recovery'], dtype=object)
```


In [55]: `# 3.Read in data`
`Brigham = pd.read_csv('./Brigham_City_Police_Crime_Data_20231018.csv')`
`Brigham.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 19 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             15000 non-null  int64
1   address_1                             15000 non-null  object
2   case_number                           15000 non-null  object
3   city                                  14999 non-null  object
4   clearance_type                         0 non-null      float64
5   created_at                            15000 non-null  object
6   day_of_week                           15000 non-null  object
7   hour_of_day                           15000 non-null  int64
8   incident_datetime                     15000 non-null  object
9   incident_description                   15000 non-null  object
10  incident_id                            15000 non-null  int64
11  incident_type_primary                  15000 non-null  object
12  latitude                               15000 non-null  float64
13  location                               15000 non-null  object
14  longitude                              15000 non-null  float64
15  parent_incident_type                  15000 non-null  object
16  state                                 15000 non-null  object
17  updated_at                            15000 non-null  object
18  zip                                    14957 non-null  float64
dtypes: float64(4), int64(3), object(12)
memory usage: 2.2+ MB
```

In [56]: `# 3.Re-arrange columns`
`Brigham=Brigham[['incident_id', 'case_number', 'incident_datetime', 'parent_incident_type', 'incident_type_primary', 'incident_description', 'address_1', 'city', 'latitude', 'longitude', 'hour_of_day', 'day_of_week', 'created_at', 'updated_at', 'zip']]`
`Brigham.head()`

Out[56]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_description
0	913039167	19-B02493	04/05/2019 11:41:06 AM	Weapons Offense	Weapon Offense	Weapon Offense
1	832915287	17-B09872	12/22/2017 08:33:10 AM	Other	Suspicious	Suspicious
2	832915358	17-B09944	12/22/2017 11:00:00 AM	Traffic	PD Accident	PD Accident
3	832915293	17-B09878	12/22/2017 12:16:37 PM	Other	Citizen Assist	Citizen Assist
4	832915296	17-B09881	12/22/2017 02:53:57 PM	Other	911 Unknown	911 Unknown

```
In [57]: # 4.Read in data
cache = pd.read_csv('./Cache_County_Sheriff_Police_Crime_Data_20231018.csv')
cache.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 20 columns):

#	Column	Non-Null Count	Dtype
0	column 1	15000 non-null	int64
1	address_1	15000 non-null	object
2	case_number	15000 non-null	object
3	city	15000 non-null	object
4	clearance_type	0 non-null	float64
5	country	0 non-null	float64
6	created_at	15000 non-null	object
7	day_of_week	15000 non-null	object
8	hour_of_day	15000 non-null	int64
9	incident_datetime	15000 non-null	object
10	incident_description	15000 non-null	object
11	incident_id	15000 non-null	int64
12	incident_type_primary	15000 non-null	object
13	latitude	15000 non-null	float64
14	location	15000 non-null	object
15	longitude	15000 non-null	float64
16	parent_incident_type	15000 non-null	object
17	state	15000 non-null	object
18	updated_at	15000 non-null	object
19	zip	0 non-null	float64

dtypes: float64(5), int64(3), object(12)
memory usage: 2.3+ MB

In [58]:

```
# 4.Re-arrange columns
cache = cache[['incident_id', 'case_number', 'incident_datetime','parent',
               'incident_description','address_1', 'city', 'latitude', 'location', 'hour_of_day', 'day_of_week', 'created_at', '
cache.head()
```

Out [58]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incider
0	758696378	16-C3608	05/04/2016 06:31:41 PM	Drugs	[CCSO] C/S DRUGS	
1	756956187	16-C3027	04/16/2016 09:27:09 PM	Community Policing	[CCSO] SUSP INCIDENT	
2	757683093	16-C3198	04/22/2016 05:31:43 AM	Alarm	[CCSO] ALARM,INTRU	
3	757683094	16-C3199	04/22/2016 07:38:07 AM	Community Policing	[CCSO] ANIMAL PROBLEM	A
4	750003534	16-C1221	02/15/2016 01:03:09 PM	Community Policing	[CCSO] ANIMAL PROBLEM	A

```
In [59]: # 5.Read in data
ephraim = pd.read_csv('./Ephraim_City_Police_Crime_Data_20231018.csv')
ephraim.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4061 entries, 0 to 4060
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed Column                        4061 non-null   int64
1   address_1                            4035 non-null   object
2   case_number                          4061 non-null   object
3   city                                 4061 non-null   object
4   clearance_type                       0 non-null      float64
5   country                             0 non-null      float64
6   created_at                           4061 non-null   object
7   day_of_week                          4061 non-null   object
8   hour_of_day                          4061 non-null   int64
9   incident_datetime                   4061 non-null   object
10  incident_description                 4061 non-null   object
11  incident_id                         4061 non-null   int64
12  incident_type_primary               4061 non-null   object
13  latitude                            4035 non-null   float64
14  location                            4035 non-null   object
15  longitude                           4035 non-null   float64
16  parent_incident_type                4061 non-null   object
17  state                              4061 non-null   object
18  updated_at                          4061 non-null   object
19  zip                                 3987 non-null   float64
dtypes: float64(5), int64(3), object(12)
memory usage: 634.7+ KB
```

Out[60]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	758597485	15110222	11/06/2015 12:20:00 PM	Property Crime	TRESPASSING	TRIFLING
1	758597159	16010069	01/03/2016 05:56:00 PM	Community Policing	SUSPICIOUS ACTIVITY	PROPERTY CRIME
2	758597083	16010500	01/15/2016 03:23:00 AM	Community Policing	SUSPICIOUS ACTIVITY	ARSON
3	758597081	16010515	01/15/2016 09:43:00 AM	Other	MISCELLANEOUS	MISCELLANEOUS
4	490848716	14070315	07/10/2014 06:54:00 PM	Community Policing	SUSPICIOUS ACTIVITY	ACQUAINTANCE

```
In [61]: # 6.Read in data
iron = pd.read_csv('./Iron_County_Sheriffs_Office_Crime_Police_Data_2023')
iron.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8513 entries, 0 to 8512
Data columns (total 20 columns):

#	Column	Non-Null Count	Dtype
0	Unnamed Column	8513 non-null	int64
1	address_1	8504 non-null	object
2	case_number	8513 non-null	object
3	city	8495 non-null	object
4	clearance_type	0 non-null	float64
5	country	0 non-null	float64
6	created_at	8513 non-null	object
7	day_of_week	8513 non-null	object
8	hour_of_day	8513 non-null	int64
9	incident_datetime	8513 non-null	object
10	incident_description	8512 non-null	object
11	incident_id	8513 non-null	int64
12	incident_type_primary	8513 non-null	object
13	latitude	8509 non-null	float64
14	location	8509 non-null	object
15	longitude	8509 non-null	float64
16	parent_incident_type	8513 non-null	object
17	state	8487 non-null	object
18	updated_at	8513 non-null	object
19	zip	1794 non-null	float64

dtypes: float64(5), int64(3), object(12)
memory usage: 1.3+ MB

```
In [62]: # 6.Re-arrange columns
iron = iron[['incident_id', 'case_number', 'incident_datetime','parent_in',
            'incident_description','address_1', 'city', 'latitude', 'lon',
            'location', 'hour_of_day', 'day_of_week', 'created_at', 'id'],
            columns=['incident_id', 'case_number', 'incident_datetime', 'parent_incident_type', 'incident_type_primary', 'incident_type_secondary']]
iron.head()
```

Out[62]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	409580467	13-03269	12/09/2013 08:00:50 PM	Drugs	Drugs	
1	515955081	14-02014	08/05/2014 10:02:01 AM	Alarm	Alarm	
2	548497904	14-02349	09/08/2014 01:43:34 PM	Drugs	Drugs	
3	571312495	14-02613	10/03/2014 08:16:05 PM	Theft	Theft	
4	575721633	14-02657	10/08/2014 05:48:31 PM	Traffic	DUI	

```
In [63]: # 7.Read in data
juab = pd.read_csv('./Juab_County_Sheriff_Police_Crime_Data_20231018.csv')
juab.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1640 entries, 0 to 1639
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  ---                                -
0   Unnamed Column                        1640 non-null   int64
1   address_1                            1640 non-null   object
2   case_number                          1640 non-null   object
3   city                                 1640 non-null   object
4   clearance_type                       0 non-null      float64
5   country                             0 non-null      float64
6   created_at                           1640 non-null   object
7   day_of_week                          1640 non-null   object
8   hour_of_day                          1640 non-null   int64
9   incident_datetime                   1640 non-null   object
10  incident_description                  1640 non-null   object
11  incident_id                          1640 non-null   int64
12  incident_type_primary                 1640 non-null   object
13  latitude                             1640 non-null   float64
14  location                             1640 non-null   object
15  longitude                            1640 non-null   float64
16  parent_incident_type                 1640 non-null   object
17  state                                1640 non-null   object
18  updated_at                           1640 non-null   object
19  zip                                  0 non-null      float64
dtypes: float64(5), int64(3), object(12)
memory usage: 256.4+ KB
```


In [64]:

```
# 7.Re-arrange columns
juab = juab[['incident_id', 'case_number', 'incident_datetime','parent_inci
            'incident_description','address_1', 'city', 'latitude', 'lon
            'location', 'hour_of_day', 'day_of_week', 'created_at', 'i
juab.head()
```

Out [64]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_description
0	751460206	16JC0307	03/01/2016 11:03:02 AM	Community Policing	ANIMAL PROBLEM	Description
1	842131619	18JC0358	03/27/2018 09:03:14 AM	Other	NON-CRIMINAL CIVIL COMPLAINT	Description
2	751760013	16JC0309	03/02/2016 08:03:13 AM	Community Policing	ANIMAL PROBLEM	Description
3	752212713	16JC0314	03/03/2016 10:03:46 AM	Community Policing	CITIZEN ASSIST	Description
4	752212707	16JC0321	03/04/2016 05:03:57 PM	Community Policing	CITIZEN ASSIST	Description

```
In [65]: # 8.Read in data
kaysville = pd.read_csv('./Kaysville_City_Police_Crime_Data_20231018.csv')
kaysville.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 19 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             15000 non-null  int64
1   address_1                            15000 non-null  object
2   case_number                          15000 non-null  object
3   city                                  15000 non-null  object
4   clearance_type                       0 non-null      float64
5   created_at                           15000 non-null  object
6   day_of_week                          15000 non-null  object
7   hour_of_day                          15000 non-null  int64
8   incident_datetime                   15000 non-null  object
9   incident_description                 15000 non-null  object
10  incident_id                          15000 non-null  int64
11  incident_type_primary                15000 non-null  object
12  latitude                             15000 non-null  float64
13  location                             15000 non-null  object
14  longitude                            15000 non-null  float64
15  parent_incident_type                 15000 non-null  object
16  state                                15000 non-null  object
17  updated_at                           15000 non-null  object
18  zip                                  13271 non-null  float64
dtypes: float64(4), int64(3), object(12)
memory usage: 2.2+ MB
```

```
In [66]: # 8.Re-arrange columns
kaysville = kaysville[['incident_id', 'case_number', 'incident_datetime',
                        'incident_description', 'address_1', 'city', 'latitude', 'longitude',
                        'location', 'hour_of_day', 'day_of_week', 'created_at', 'parent_incident_type']]
kaysville.head()
```

Out[66]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_description
0	833928479	C1858059	02/11/2018 07:23:38 AM	Vehicle Stop	CAD: Traffic Stop	Vehicle Stop
1	833928478	C1858068	02/11/2018 08:57:37 AM	Vehicle Stop	CAD: Traffic Stop	Vehicle Stop
2	833928477	C1858071	02/11/2018 09:06:37 AM	Vehicle Stop	CAD: Traffic Stop	Vehicle Stop
3	833928476	C1858073	02/11/2018 09:15:27 AM	Vehicle Stop	CAD: Traffic Stop	Vehicle Stop
4	833928475	C1858076	02/11/2018 09:21:26 AM	Vehicle Stop	CAD: Traffic Stop	Vehicle Stop

```
In [67]: # 9.Read in data
park_city = pd.read_csv('./Park_City_Police_Crime_Data_20231018.csv')
park_city.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed Column                        15000 non-null  int64
1   address_1                             14995 non-null  object
2   case_number                           15000 non-null  object
3   city                                  14969 non-null  object
4   clearance_type                         0 non-null      float64
5   country                               0 non-null      float64
6   created_at                            15000 non-null  object
7   day_of_week                           15000 non-null  object
8   hour_of_day                           15000 non-null  int64
9   incident_datetime                     15000 non-null  object
10  incident_description                   14996 non-null  object
11  incident_id                           15000 non-null  int64
12  incident_type_primary                 15000 non-null  object
13  latitude                              15000 non-null  float64
14  location                              15000 non-null  object
15  longitude                             15000 non-null  float64
16  parent_incident_type                  15000 non-null  object
17  state                                15000 non-null  object
18  updated_at                            15000 non-null  object
19  zip                                   393 non-null    float64
dtypes: float64(5), int64(3), object(12)
memory usage: 2.3+ MB
```

```
In [68]: # 9.Re-arrange columns
park_city = park_city[['incident_id', 'case_number', 'incident_datetime',
                        'incident_description', 'address_1', 'city', 'latitude', 'longitude',
                        'location', 'hour_of_day', 'day_of_week', 'created_at', 'parent_incident_type', 'incident_type_primary', 'incident_type_secondary']]
park_city.head()
```

Out[68]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	828265191	17-20940	12/04/2017 01:12:09 PM	Theft	FRAUD	
1	95483627	A10-24288	10/26/2010 09:10:01 AM	Property Crime	CRIM MISCHIEF	
2	734697515	15-19052	10/21/2015 12:10:55 AM	Disorder	SUSPICIOUS	
3	760351917	16-09075	05/03/2016 12:05:00 AM	Property Crime	CRIM MISCHIEF	
4	760351928	16-09086	05/12/2016 10:05:41 AM	Disorder	HARASSMENT	

```
In [69]: # 10.Read in data
perry = pd.read_csv('./Perry_City_Police_Crime_Data_20231018.csv')
perry.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 679 entries, 0 to 678
Data columns (total 19 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             679 non-null    int64
1   address_1                            679 non-null    object
2   case_number                          679 non-null    object
3   city                                 679 non-null    object
4   clearance_type                       0 non-null      float64
5   created_at                           679 non-null    object
6   day_of_week                          679 non-null    object
7   hour_of_day                          679 non-null    int64
8   incident_datetime                   679 non-null    object
9   incident_description                 679 non-null    object
10  incident_id                          679 non-null    int64
11  incident_type_primary               679 non-null    object
12  latitude                            679 non-null    float64
13  location                            679 non-null    object
14  longitude                           679 non-null    float64
15  parent_incident_type                679 non-null    object
16  state                              679 non-null    object
17  updated_at                          679 non-null    object
18  zip                                 661 non-null    float64
dtypes: float64(4), int64(3), object(12)
memory usage: 100.9+ KB
```

```
In [70]: # 10.Re-arrange columns
perry = perry[['incident_id', 'case_number', 'incident_datetime', 'parent',
               'incident_description', 'address_1', 'city', 'latitude', 'location', 'hour_of_day', 'day_of_week', 'created_at', '
perry.head()
```

Out[70]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incider
0	742623389	15-P01549	12/30/2015 12:48:02 PM	Property Crime	Vandalism	
1	742754753	16-P00005	01/01/2016 05:10:47 PM	Traffic	Hit & Run PD	
2	742754754	16-P00006	01/02/2016 04:44:18 AM	Property Crime	Property Damage	Prc
3	742923954	16-P00013	01/04/2016 04:13:13 AM	Property Crime	Property Damage	Prc
4	746843931	16-P00074	01/19/2016 09:11:04 AM	Theft	fraud	

```
In [71]: # 11.Read in data: Pleasant_View_Police_Crime_Data_20231018
pleasant_view = pd.read_csv('./Pleasant_View_Police_Crime_Data_20231018.csv')
pleasant_view.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 19 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             15000 non-null  int64
1   address_1                             15000 non-null  object
2   case_number                           15000 non-null  object
3   city                                  14999 non-null  object
4   clearance_type                         0 non-null      float64
5   created_at                             15000 non-null  object
6   day_of_week                           15000 non-null  object
7   hour_of_day                           15000 non-null  int64
8   incident_datetime                     15000 non-null  object
9   incident_description                   15000 non-null  object
10  incident_id                           15000 non-null  int64
11  incident_type_primary                  15000 non-null  object
12  latitude                               15000 non-null  float64
13  location                               15000 non-null  object
14  longitude                              15000 non-null  float64
15  parent_incident_type                   15000 non-null  object
16  state                                  15000 non-null  object
17  updated_at                             15000 non-null  object
18  zip                                    14957 non-null  float64
dtypes: float64(4), int64(3), object(12)
memory usage: 2.2+ MB
```



```
In [72]: # 11.Re-arrange columns
pleasant_view = pleasant_view[['incident_id', 'case_number', 'incident_datetime',
                                'parent_incident_type', 'incident_type_primary', 'incident_type_secondary',
                                'incident_description', 'address_1', 'city', 'latitude', 'longitude',
                                'location', 'hour_of_day', 'day_of_week', 'created_at', 'updated_at']]
pleasant_view.head()
```

Out[72]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	913039167	19-B02493	04/05/2019 11:41:06 AM	Weapons Offense	Weapon Offense	Weapon Offense
1	832915287	17-B09872	12/22/2017 08:33:10 AM	Other	Suspicious	
2	832915358	17-B09944	12/22/2017 11:00:00 AM	Traffic	PD Accident	
3	832915293	17-B09878	12/22/2017 12:16:37 PM	Other	Citizen Assist	
4	832915296	17-B09881	12/22/2017 02:53:57 PM	Other	911 Unknown	

```
In [73]: # 12.Read in data: Price_Police_Crime_Data_20231018
price = pd.read_csv('./Price_Police_Crime_Data_20231018.csv')
price.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8618 entries, 0 to 8617
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed Column                        8618 non-null   int64
1   address_1                            8617 non-null   object
2   case_number                          8618 non-null   int64
3   city                                 8618 non-null   object
4   clearance_type                       0 non-null      float64
5   country                             0 non-null      float64
6   created_at                           8582 non-null   object
7   day_of_week                          8618 non-null   object
8   hour_of_day                          8618 non-null   int64
9   incident_datetime                   8618 non-null   object
10  incident_description                 8618 non-null   object
11  incident_id                         8618 non-null   int64
12  incident_type_primary               8618 non-null   object
13  latitude                            8617 non-null   float64
14  location                            8617 non-null   object
15  longitude                           8617 non-null   float64
16  parent_incident_type                8618 non-null   object
17  state                               8618 non-null   object
18  updated_at                          8618 non-null   object
19  zip                                  8432 non-null   object
dtypes: float64(4), int64(4), object(12)
memory usage: 1.3+ MB
```

In [74]:

```
# 12.Re-arrange columns
price = price[['incident_id', 'case_number', 'incident_datetime', 'parent_incident_type', 'incident_type_primary', 'incident_type_secondary', 'incident_description', 'address_1', 'city', 'latitude', 'longitude', 'location', 'hour_of_day', 'day_of_week', 'created_at', 'updated_at']]
price.head()
```

Out [74]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	100181896	20120235	02/22/2012 06:00:00 PM	Disorder	[PCPD] Public Intoxication	Public Intoxication
1	100329494	20120233	02/22/2012 02:00:00 PM	Traffic	"[PCPD] Traffic Accident, Vehicle Damage"	ACCIDENT, Vehicle Damage
2	100329495	20120237	02/23/2012 12:45:00 PM	Theft	[PCPD] Theft Other	THEFT
3	100329496	20120240	02/23/2012 04:08:00 PM	Drugs	[PCPD] Other Controlled Substances	SUBSTANCE CONTROL
4	100329497	20120239	02/23/2012 04:08:00 PM	Traffic	"[PCPD] Traffic Accident, Vehicle Damage"	ACCIDENT, Vehicle Damage

```
In [75]: # 13.Read in data: Roy_City_Police_Crime_Data_20231018
roy = pd.read_csv('./Roy_City_Police_Crime_Data_20231018.csv')
roy.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             15000 non-null  int64
1   address_1                             14983 non-null  object
2   case_number                           15000 non-null  object
3   city                                  15000 non-null  object
4   clearance_type                         0 non-null      float64
5   country                               0 non-null      float64
6   created_at                            15000 non-null  object
7   day_of_week                           15000 non-null  object
8   hour_of_day                           15000 non-null  int64
9   incident_datetime                     15000 non-null  object
10  incident_description                   15000 non-null  object
11  incident_id                           15000 non-null  int64
12  incident_type_primary                  15000 non-null  object
13  latitude                               14983 non-null  float64
14  location                               14983 non-null  object
15  longitude                              14983 non-null  float64
16  parent_incident_type                   15000 non-null  object
17  state                                 15000 non-null  object
18  updated_at                             15000 non-null  object
19  zip                                    14826 non-null  float64
dtypes: float64(5), int64(3), object(12)
memory usage: 2.3+ MB
```

```
In [76]: # 13.Re-arrange columns
roy = roy[['incident_id', 'case_number', 'incident_datetime', 'parent_incident_type', 'incident_type_primary', 'incident_type_secondary', 'incident_description', 'address_1', 'city', 'latitude', 'longitude', 'location', 'hour_of_day', 'day_of_week', 'created_at', 'updated_at']]
roy.head()
```

Out[76]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	751766860	16-5647	03/04/2016 11:13:53 AM	Other	Alarm	ALARM
1	751029830	16-5048	02/27/2016 01:54:08 PM	Traffic	Traffic	TRAFFIC
2	751029833	16-5053	02/27/2016 02:16:06 PM	Other	Other	KEEPER
3	751029835	16-5061	02/27/2016 04:44:29 PM	Traffic	Traffic	TRAFFIC
4	751029838	16-5066	02/27/2016 05:31:02 PM	Traffic	Traffic	TRAFFIC

```
In [77]: # 14.Read in data: Salt Lake County Crime Data 2013_20231018
slc = pd.read_csv('./Salt_Lake_County_Crime_Data_2013_20231018.csv')
slc.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14 entries, 0 to 13
Data columns (total 12 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Agency                                14 non-null    object
1   Population                            14 non-null    int64
2   Homicide                             13 non-null    float64
3   Rape                                 13 non-null    float64
4   Robbery                              13 non-null    float64
5   Aggravated Assault                   13 non-null    float64
6   Burglary                             13 non-null    float64
7   Larceny                              13 non-null    float64
8   Motor Vehicle Theft                  13 non-null    float64
9   Arson                                13 non-null    float64
10  Total Crime Index                     13 non-null    float64
11  Crime Rate per 1,000                  11 non-null    float64
dtypes: float64(10), int64(1), object(1)
memory usage: 1.4+ KB
```

```
In [78]: # 15.Read in data: Smithfield_Police_Crime_Data_20231018
smithfield = pd.read_csv('./Smithfield_Police_Crime_Data_20231018.csv')
smithfield.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6211 entries, 0 to 6210
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             6211 non-null   int64
1   address_1                            6211 non-null   object
2   case_number                          6211 non-null   int64
3   city                                 6211 non-null   object
4   clearance_type                       0 non-null      float64
5   country                             0 non-null      float64
6   created_at                           6211 non-null   object
7   day_of_week                          6211 non-null   object
8   hour_of_day                          6211 non-null   int64
9   incident_datetime                   6211 non-null   object
10  incident_description                 6211 non-null   object
11  incident_id                         6211 non-null   int64
12  incident_type_primary               6211 non-null   object
13  latitude                            6211 non-null   float64
14  location                            6211 non-null   object
15  longitude                           6211 non-null   float64
16  parent_incident_type                6211 non-null   object
17  state                               6211 non-null   object
18  updated_at                          6211 non-null   object
19  zip                                 6080 non-null   float64
dtypes: float64(5), int64(4), object(11)
memory usage: 970.6+ KB
```

In [79]:

```
# 15.Re-arrange columns
smithfield = smithfield[['incident_id', 'case_number', 'incident_datetime',
                          'incident_description', 'address_1', 'city', 'latitude', 'longitude',
                          'location', 'hour_of_day', 'day_of_week', 'created_at', 'parent_incident_type']]
smithfield.head()
```

Out[79]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	290521772	132895	07/16/2013 09:47:00 AM	Theft	-FRAUD	SCAM
1	717774978	151248	05/21/2015 03:17:00 PM	Alarm	-ALARM	ALARM
2	715047696	151062	05/05/2015 10:06:00 AM	Family Offense	-DOMESTIC/FAMILY INCIDENT	DOMESTIC VIOLENCE
3	715139075	151069	05/05/2015 02:52:00 PM	Traffic	-TRAFFIC CRASH	TRAFIC CRASH
4	715139076	151071	05/05/2015 08:45:00 PM	Drugs	-CONTROLLED SUBSTANCE	

```
In [80]: # 16.Read in data: South_Ogden_Police_Crime_Data_20231018
s_ogden = pd.read_csv('./South_Ogden_Police_Crime_Data_20231018.csv')
s_ogden.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             15000 non-null  int64
1   address_1                             14987 non-null  object
2   case_number                           15000 non-null  object
3   city                                  15000 non-null  object
4   clearance_type                         0 non-null      float64
5   country                               0 non-null      float64
6   created_at                            15000 non-null  object
7   day_of_week                           15000 non-null  object
8   hour_of_day                           15000 non-null  int64
9   incident_datetime                     15000 non-null  object
10  incident_description                   15000 non-null  object
11  incident_id                           15000 non-null  int64
12  incident_type_primary                 15000 non-null  object
13  latitude                              15000 non-null  float64
14  location                              14987 non-null  object
15  longitude                             15000 non-null  float64
16  parent_incident_type                  15000 non-null  object
17  state                                 15000 non-null  object
18  updated_at                            15000 non-null  object
19  zip                                    4032 non-null   float64
dtypes: float64(5), int64(3), object(12)
memory usage: 2.3+ MB
```



```
In [81]: # 16.Re-arrange columns
s_ogden = s_ogden[['incident_id', 'case_number', 'incident_datetime', 'pa
                'incident_description', 'address_1', 'city', 'latitude', 'l
                'location', 'hour_of_day', 'day_of_week', 'created_at', 'i
s_ogden.head()
```

Out[81]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incider
0	748390201	16-1345	01/30/2016 09:11:18 PM	Traffic	Traffic	TRAF
1	640185082	14-17947	12/19/2014 12:13:20 AM	Other	Alarm	ALAI
2	650733862	15-16	01/01/2015 02:10:20 PM	Other	Other	KEE
3	660620611	15-456	01/12/2015 01:00:30 PM	Theft	Theft	
4	667748129	15-712	01/19/2015 07:46:02 AM	Other	Alarm	ALAI

```
In [82]: # 17.Read in data: Sunset_Police_Crime_Data_20231018
sunset = pd.read_csv('./Sunset_Police_Crime_Data_20231018.csv')
sunset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Unnamed Column                        15000 non-null  int64
1   address_1                            14982 non-null  object
2   case_number                          15000 non-null  int64
3   city                                 14992 non-null  object
4   clearance_type                       0 non-null      float64
5   country                             0 non-null      float64
6   created_at                           15000 non-null  object
7   day_of_week                          15000 non-null  object
8   hour_of_day                          15000 non-null  int64
9   incident_datetime                    15000 non-null  object
10  incident_description                  15000 non-null  object
11  incident_id                          15000 non-null  int64
12  incident_type_primary                 15000 non-null  object
13  latitude                             14983 non-null  float64
14  location                             14982 non-null  object
15  longitude                             14983 non-null  float64
16  parent_incident_type                  15000 non-null  object
17  state                                15000 non-null  object
18  updated_at                           15000 non-null  object
19  zip                                   14331 non-null  object
dtypes: float64(4), int64(4), object(12)
memory usage: 2.3+ MB
```

```
In [83]: # 17.Re-arrange columns
sunset = sunset[['incident_id', 'case_number', 'incident_datetime', 'parent_incident_type', 'incident_type_primary', 'incident_type_secondary', 'incident_description', 'address_1', 'city', 'latitude', 'longitude', 'location', 'hour_of_day', 'day_of_week', 'created_at', 'updated_at']]
sunset.head()
```

Out[83]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	745069660	1602372	01/15/2016 07:32:55 AM	Community Policing	[CFS] SCHOOL ZONE ENFORCEMENT	SCHOOL ZONE ENFORCEMENT
1	746646114	1603443	01/22/2016 05:58:00 AM	Other	[CFS] 1050 PD	
2	746646113	1603448	01/22/2016 07:40:10 AM	Community Policing	[CFS] CITIZEN REQUESTING INFORMATION	
3	788938238	1702064	01/12/2017 03:09:02 AM	Other	[CFS] 1050 PD	
4	788938237	1702067	01/12/2017 03:45:25 AM	Other	[CFS] 1075	

```
In [84]: # 18.Read in data: Syracuse_Police_Crime_Data_20231018
syracuse = pd.read_csv('./Syracuse_Police_Crime_Data_20231018.csv')
syracuse.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15000 entries, 0 to 14999
Data columns (total 18 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             15000 non-null  int64
1   address_1                             15000 non-null  object
2   case_number                           15000 non-null  object
3   city                                  15000 non-null  object
4   clearance_type                         0 non-null      float64
5   created_at                             15000 non-null  object
6   day_of_week                           15000 non-null  object
7   hour_of_day                           15000 non-null  int64
8   incident_datetime                     15000 non-null  object
9   incident_description                   15000 non-null  object
10  incident_id                           15000 non-null  int64
11  incident_type_primary                  15000 non-null  object
12  latitude                              15000 non-null  float64
13  location                              15000 non-null  object
14  longitude                              15000 non-null  float64
15  parent_incident_type                  15000 non-null  object
16  state                                 15000 non-null  object
17  updated_at                            15000 non-null  object
dtypes: float64(3), int64(3), object(12)
memory usage: 2.1+ MB
```



```
In [86]: # 19.Read in data: Utah_County_Sheriff_Crime_Incident_Data_20231018
utah = pd.read_csv('./Utah_County_Sheriff_Crime_Incident_Data_20231018.csv')
utah.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 72400 entries, 0 to 72399
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   incident_id                          72400 non-null  int64
1   case_number                          72400 non-null  object
2   incident_datetime                    72400 non-null  object
3   incident_type_primary                72400 non-null  object
4   incident_description                 72400 non-null  object
5   clearance_type                       0 non-null      float64
6   address_1                           72381 non-null  object
7   address_2                           0 non-null      float64
8   city                                72399 non-null  object
9   state                                72400 non-null  object
10  zip                                  3266 non-null   float64
11  country                             22156 non-null  object
12  latitude                             72400 non-null  float64
13  longitude                             72400 non-null  float64
14  created_at                           72400 non-null  object
15  updated_at                           72400 non-null  object
16  location                             72398 non-null  object
17  hour_of_day                           72400 non-null  int64
18  day_of_week                           72400 non-null  object
19  parent_incident_type                 72400 non-null  object
dtypes: float64(5), int64(2), object(13)
memory usage: 11.0+ MB
```

```
/var/folders/q4/rr159kcn4vb9yhnfzgzd85xh0000gn/T/ipykernel_3964/3430271
747.py:2: DtypeWarning: Columns (11) have mixed types. Specify dtype op
tion on import or set low_memory=False.
utah = pd.read_csv('./Utah_County_Sheriff_Crime_Incident_Data_2023101
8.csv')
```

```
In [87]: # 19.Re-arrange columns
        utah = utah[['incident_id', 'case_number', 'incident_datetime', 'parent_inci
                  'incident_description', 'address_1', 'city', 'latitude', 'lon
                  'location', 'hour_of_day', 'day_of_week', 'created_at', 'l
        utah.head()
```

Out [87]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident_type_secondary
0	4155486	08UC11086	10/05/2008 12:00:00 AM	Other	Weapons Offense	
1	781823043	16UC11433	11/19/2016 12:11:51 AM	Liquor	ALCOHOL OFFENSE	Description
2	781161909	16UC08720	09/04/2016 12:09:09 PM	Community Policing	ANIMAL PROBLEM	Description
3	752214707	16UC02132	03/06/2016 06:03:29 PM	Alarm	ALARM	Description
4	771139313	16UC08685	09/03/2016 01:09:26 AM	Traffic	DRIVING UNDER INFLUENCE	Description

```
In [88]: # 20.Read in data: Woods_Cross_Police_Crime_Data_20231018
woods_cross = pd.read_csv('./Woods_Cross_Police_Crime_Data_20231018.csv')
woods_cross.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 49624 entries, 0 to 49623
Data columns (total 20 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   column 1                             49624 non-null  int64
1   address_1                            49598 non-null  object
2   case_number                          49624 non-null  int64
3   city                                 49600 non-null  object
4   clearance_type                       0 non-null      float64
5   country                             0 non-null      float64
6   created_at                          49624 non-null  object
7   day_of_week                         49624 non-null  object
8   hour_of_day                         49624 non-null  int64
9   incident_datetime                   49624 non-null  object
10  incident_description                 49624 non-null  object
11  incident_id                         49624 non-null  int64
12  incident_type_primary               49624 non-null  object
13  latitude                           49602 non-null  float64
14  location                           49587 non-null  object
15  longitude                           49602 non-null  float64
16  parent_incident_type                49624 non-null  object
17  state                              49624 non-null  object
18  updated_at                         49624 non-null  object
19  zip                                 45007 non-null  object
dtypes: float64(4), int64(4), object(12)
memory usage: 7.6+ MB
```

In [89]:

```
# 20.Re-arrange columns
woods_cross = woods_cross[['incident_id', 'case_number', 'incident_datet',
                            'incident_description', 'address_1', 'city', 'latitude', 'l',
                            'location', 'hour_of_day', 'day_of_week', 'created_at', 'i',
woods_cross.head()
```

Out [89]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incider
0	745069660	1602372	01/15/2016 07:32:55 AM	Community Policing	[CFS] SCHOOL ZONE ENFORCEMENT	S EM
1	746646114	1603443	01/22/2016 05:58:00 AM	Other	[CFS] 1050 PD	
2	746646113	1603448	01/22/2016 07:40:10 AM	Community Policing	[CFS] CITIZEN REQUESTING INFORMATION	I
3	788938238	1702064	01/12/2017 03:09:02 AM	Other	[CFS] 1050 PD	
4	788938237	1702067	01/12/2017 03:45:25 AM	Other	[CFS] 1075	


```
In [90]: # Concatenate all the dataframes vertically
df = pd.concat([st_george, beaver, Brigham, cache, ephraim, iron, juab, l
               perry, pleasant_view, price, roy, smithfield, s_ogden, su
               utah, woods_cross], ignore_index=True)
df
```

Out [90]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	i
0	691160812	15P003395	2/11/2015 16:02	Theft	FRAUD	
1	712757140	15P008661	4/13/2015 13:04	Drugs	DRUGS	
2	715044076	15P010661	5/5/2015 22:05	Drugs	DRUGS	
3	825529671	17P027337	11/5/2017 5:11	Traffic	DUI	
4	736930480	15P027109	11/12/2015 23:11	Traffic	DUI	
...
479659	808705608	1729112	06/09/2017 08:46:12 AM	Other	[CFS] 1090 COM	
479660	808705606	1729120	06/09/2017 09:26:42 AM	Community Policing	[CFS] OUTSIDE ASSIST	
479661	808705604	1729143	06/09/2017 11:40:39 AM	Community Policing	[CFS] VACATION WATCH	
479662	808705600	1729197	06/09/2017 04:05:21 PM	Other	[CFS] C/S VIOLATIONS	
479663	808705594	1729279	06/10/2017 01:57:04 AM	Disorder	[CFS] NOISE DISTURBANCE	

479664 rows × 15 columns

FEATURE SELECTION

We will need to extract the most relevant features to get a more accurate dataset for our visualization. Also our primary focus will be on the most common types of crimes so we will determine the top 5-10 crimes and place the rest in an 'Misc' category

```
In [91]: # # Look at the unique values of "parent_incident_type" to decide if we want to use it
unique_values = pd.unique(df['parent_incident_type'])
unique_values
```

```
Out[91]: array(['Theft', 'Drugs', 'Traffic', 'Disorder', 'Pedestrian Stop',
               'Assault', 'Missing Person', 'Theft from Vehicle',
               'Property Crime', 'Liquor', 'Family Offense', 'Community Policing',
               'Vehicle Stop', 'Breaking & Entering', 'Other',
               'Other Sexual Offense', 'Weapons Offense', 'Vehicle Recovery',
               'Kidnapping', 'Assault with Deadly Weapon', 'Alarm', 'Fire',
               'Emergency', 'Theft of Vehicle', 'Robbery', 'Death', 'Arson',
               'Sexual Assault', 'Proactive Policing', 'Quality of Life',
               'Homicide', 'Sexual Offense', 'Property Crime Residential'],
              dtype=object)
```

```
In [92]: # Do we want to use the 'parent_incident_type' column
value_counts = df['parent_incident_type'].value_counts()
```

```
In [93]: import matplotlib.pyplot as plt

# Assuming 'crime_type_column' is the name of the column in your DataFrame
crime_type_counts = df['parent_incident_type'].value_counts()

# Extract unique values and their counts
unique_values = crime_type_counts.index
value_counts = crime_type_counts.values

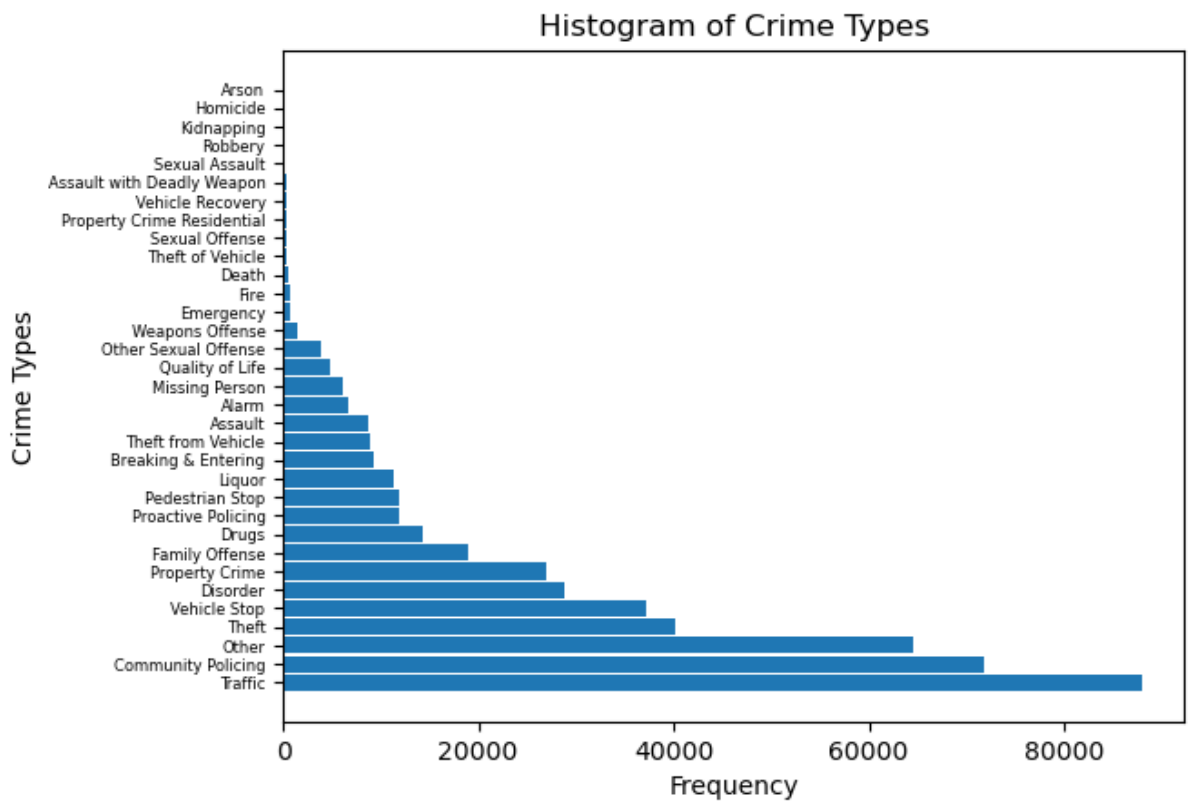
# Set the positions and labels for y-ticks
y_tick_positions = range(len(unique_values))
y_ticks = unique_values

# Create the horizontal bar chart
plt.barh(y_tick_positions, value_counts, height=0.9, align='center')

# Set labels and title
plt.xlabel('Frequency')
plt.ylabel('Crime Types')
plt.title('Histogram of Crime Types')

# Set the y-ticks with custom positions, labels, and font size
plt.yticks(y_tick_positions, y_ticks, fontsize=6) # Adjust the font size

# Show the plot
plt.show()
```



```
In [94]: # List of categories to keep
categories_to_keep = ['Traffic', 'Community Policing', 'Other', 'Theft',

# Function to categorize as 'Misc' if not in the list
def categorize_as_other(category):
    return category if category in categories_to_keep else 'Misc'

# Apply the function to the 'Category' column
df['Crime_Category'] = df['parent_incident_type'].apply(categorize_as_otl

# Display the updated DataFrame
df.head()
```

Out[94]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incider
0	691160812	15P003395	2/11/2015 16:02	Theft	FRAUD	
1	712757140	15P008661	4/13/2015 13:04	Drugs	DRUGS	
2	715044076	15P010661	5/5/2015 22:05	Drugs	DRUGS	
3	825529671	17P027337	11/5/2017 5:11	Traffic	DUI	
4	736930480	15P027109	11/12/2015 23:11	Traffic	DUI	

OUTLIERS

We will also need to determine any outliers in the data and determine if we should remove them or replace them with Median values.

```
In [95]: from scipy import stats

# Assuming 'numeric_column' is the name of the column with numerical data
z_scores = stats.zscore(df['hour_of_day'])
threshold = 3 # You can adjust this threshold as needed

# Find indices of potential outliers
outlier_indices = np.where(np.abs(z_scores) > threshold)

# List the actual data points that are potential outliers
outliers = df.iloc[outlier_indices]
outliers
```

```
Out[95]:
```

incident_id	case_number	incident_datetime	parent_incident_type	incident_type_primary	incident

```
In [96]: # Detect missing values
df.isna().sum()
# Drop the 88 rows with missing longitude and latitude
# Drop location column
# Drop all N/A
```

```
Out[96]: incident_id      0
case_number      0
incident_datetime  0
parent_incident_type  0
incident_type_primary  0
incident_description  128
address_1      193
city           84
latitude       88
longitude      88
location      136
hour_of_day     0
day_of_week     0
created_at      41
updated_at      0
Crime_Category  0
dtype: int64
```

DATA CLEANUP

Some of the cities within the files were not cities from UTAH, so they needed to be cleanup/removed

```
In [97]: # Drop the 'location' and 'incident_type_primary' columns
df = df.drop(['location', 'incident_type_primary'], axis=1)
```

In [98]:

Drop rows with missing values
df = df.dropna()
df

Out [98]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
0	691160812	15P003395	2/11/2015 16:02	Theft	FRAUD	
1	712757140	15P008661	4/13/2015 13:04	Drugs	DRUGS	1
2	715044076	15P010661	5/5/2015 22:05	Drugs	DRUGS	I-1
3	825529671	17P027337	11/5/2017 5:11	Traffic	DUI	RI
4	736930480	15P027109	11/12/2015 23:11	Traffic	DUI	
...	
479659	808705608	1729112	06/09/2017 08:46:12 AM	Other	1090 COM	
479660	808705606	1729120	06/09/2017 09:26:42 AM	Community Policing	OUTSIDE ASSIST	60
479661	808705604	1729143	06/09/2017 11:40:39 AM	Community Policing	VACATION WATCH	70
479662	808705600	1729197	06/09/2017 04:05:21 PM	Other	C/S VIOLATIONS	90
479663	808705594	1729279	06/10/2017 01:57:04 AM	Disorder	NOISE DISTURBANCE	

479221 rows × 14 columns

```
In [99]: # Check to see if rows with missing values were deleted
df.isna().sum()
```

```
Out[99]: incident_id          0
case_number          0
incident_datetime    0
parent_incident_type 0
incident_description  0
address_1            0
city                0
latitude            0
longitude           0
hour_of_day          0
day_of_week          0
created_at          0
updated_at          0
Crime_Category       0
dtype: int64
```

```
In [100]: # Convert the 'city' column to uppercase
df.loc[:, 'city'] = df['city'].str.upper()
```

```
In [101]: # Prints unique values for cities  
cities = df['city'].unique()  
cities
```



```

Out[101]: array(['ST GEORGE', 'WASHINGTON', 'HURRICANE', 'LAVERKIN', 'SANTA CLAR
A',
                'LEEDS', 'WASH CO OTHER', 'IVINS', 'WASHCO STGEORGE', 'COVINGTO
N',
                'APPLE VALLEY', 'WASHCO WASHINGT', 'NEW HARMONY',
                'WINCHESTER HILL', 'PINTURA', 'MOHAVE COUNTY', 'ZION NAT PARK',
                'DIAMOND VALLEY', 'SPRINGDALE', 'BROWSE', 'GUNLOCK', 'VIRGIN',
                'TOQUERVILLE', 'HILDALE', 'CENTRAL', 'CLARK COUNTY', 'PINTO',
                'DAMMERON VALLEY', 'LINCOLN COUNTY', 'PINE VALLEY', 'ROCKVILLE',
                'BROOKSIDE', 'ENTERPRISE', 'VEYO', 'DIXIE COLLEGE', 'IRON COUNT
Y',
                'ENTERPRISE WCSO', 'WASHCO TOQUERVI', 'WASHCO HURRICAN',
                'WASHCO LEEDS', 'WASHCO NEWHARMO', 'WASHCO PINE VLY',
                'WASHCO VEYO', 'WASHCO IVINS', 'SHIVWITS RESERV',
                'WASHCO ENTERPRI', 'MILFORD', 'BEAVER', 'BRIGHAM CITY', 'MANTU
A',
                'GARLAND', 'PERRY', 'BOX ELDER CO', 'WILLARD', 'RIVERDALE',
                'WELLSVILLE', 'FAR WEST', 'TREMONTON', 'HONEYVILLE', 'CORINNE',
                'BRIGHAM', 'THATCHER', 'LOGAN', 'MILLVILLE', 'CACHE COUNTY',
                'HYRUM', 'NORTH LOGAN', 'NIBLEY', 'RICHMOND', 'PROVIDENCE',
                'AMALGA', 'SMITHFIELD', 'PARADISE', 'MENDON', 'LEWISTON',
                'TRENTON', 'NEWTON', 'RIVER HEIGHTS', 'CORNISH', 'HYDE PARK',
                'CLARKSTON', 'PRESTON', 'BENSON', 'LOGAN CANYON', 'EPHRAIM',
                'MANTI', 'MOUNT PLEASANT', 'SANPETE COUNTY', 'FAIRVIEW', 'CHESTE
R',
                'SPRING CITY', 'MORONI', 'CENTERFIELD', 'WALES', 'STERLING',
                'FOUNTAIN GREEN', 'GUNNISON', 'PAROWAN', 'IRON CO CC GRID',
                'CEDAR CITY', 'IRON CO BERYLG', 'ENOCH', 'IRON CO NC G',
                'KANARRAVILLE', 'PARAGONAH', 'IRON CO PAR GRI', 'IRON CO PARGG',
                'IRON CO KANG', 'IRON CO SUMMIT', 'WASH CO NEW HAR', 'BRIAN HEA
D',
                'IRON CO BHG', 'IRON CO MODG', 'DELTA', 'BERYL', 'SUMMIT',
                'SAINT GEORGE', 'NEWCASTLE', 'IRON CO ENTERG', 'MILLARD CO DEL
T',
                'NEW CASTLE', 'BRIANHEAD', 'HAMILTON FORT', 'MODENA', 'LUND',
                'BEAVER CO BG', 'HAMBLIN VALLEY', 'WASH CO ENTERPR', 'KANARAVILL
E',
                'MILLARD CO FILL', 'KANE COUNTY DUC', 'MONA', 'JUAB NEPHI',
                'EUREKA', 'LEVAN', 'NEPHI', 'JUAB CO EUR GR', 'JUAB CO NEP GR',
                'LITTLE SAHARA', 'JUAB CO LEV GR', 'JUAB CO MON GR', 'JUAB MON
A',
                'JUAB COUNTY', 'JUAB CO WES GR', 'JUAB WEST DESER', 'JUAB LEVA
N',
                'GOSHEN', 'UTAH COUNTY', 'ROCKY RIDGE', 'LEHI', 'JUAB EUREKA',
                'PROVO', 'KAYSVILLE', 'PARK CITY', 'COALVILLE', 'HOYTSTVILLE',
                'ROCKPORT', 'FRANCIS', 'SUMMIT COUNTY', 'HEBER CITY', 'KAMAS',
                'SAMAK', 'WEBER CANYON', 'MARION', 'OAKLEY', 'MURRAY',
                'SALT LAKE CITY', 'WANSHIP', 'MCKINNON', 'MIDWAY', 'PEOA',
                'HENEFER', 'PRICE', 'HELPER', 'PRICE, UTAH (CARBON)', 'WELLINGTO
N',
                'CARBONVILLE', 'EMERY COUNTY', 'OTHER', 'EAST CARBON',
                'SPRING GLEN', 'TEMP', 'KENILWORTH', 'ROY', 'SOUTH OGDEN',
                'WOODS CROSS', 'BOUNTIFUL', 'NORTH SALT LAKE', 'WEST BOUNTIFUL',
                'DAVIS COUNTY', 'LAYTON', 'FARMINGTON', 'CENTERVILLE', 'SYRACUS
E',
                'CLEARFIELD', 'SOUTH WEBER', 'COUNTY NW', 'GENOLA', 'VINEYARD',
                'WOODLAND HILLS', 'EAGLE MOUNTAIN', 'ELK RIDGE', 'CEDAR FORT',

```

```

'SPANISH FORK', 'AMERICAN FORK', 'OREM', 'FAIRFIELD', 'LINDON',
'SANTAQUIN', 'SPRINGVILLE', 'PAYSON', 'MAPLETON', 'CEDAR HILLS',
'HIGHLAND', 'PLEASANT GROVE', 'SALEM', 'ALPINE', 'SARATOGA SPRIN
G',
'DRAPER', 'TOOELE COUNTY', 'SPRING LAKE', 'BLUFFDALE',
'THISTLE/BIRDSEY', 'UTAH VALLEY U', 'PALMYRA', 'CARBON COUNTY',
'SUNDANCE', 'ELBERTA', 'SALT LAKE CNTY', 'LAKE SHORE',
'COVERED BRIDGE', 'LELAND', 'BENJAMIN', 'WASATCH COUNTY',
'BYU CAMPUS', 'UINTAH', 'DUCHESNE', 'MOAB', 'SALT LAKE', 'EMER
Y',
'KANAB', 'NEVADA', 'SALT LAKE COUNT', 'SANTA ANA', 'HEBER',
'FILLMORE', 'OGDEN', 'IDAHO FALLS', 'RICHFIELD',
'SARATOGA SPRINGS', 'WEBER COUNTY', 'TOOELE', 'SAN FRANCISCO',
'POCATELLO', 'PALM SPRINGS', 'FLORENCE', 'LOS ANGELES',
'CEDAR VALLEY', 'PROVOST', 'YINEYARD', 'CLINTON', 'FRUIT HEIGHT
S'],
dtype=object)

```

```

In [102]: unique_city_count = df['city'].nunique()
print("Number of unique cities:", unique_city_count)

```

Number of unique cities: 262

```

In [103]: cities_to_exclude = ['SANTA ANA', 'NEVADA', 'IDAHO FALLS', 'SAN FRANCISCO',
                                'PALM SPRINGS', 'FLORENCE', 'POCATELLO', 'LOS ANGELES']

df = df[~df['city'].isin(cities_to_exclude)]

```

```
In [104]: # Prints unique values for cities  
cities = df['city'].unique()  
cities
```

```

Out[104]: array(['ST GEORGE', 'WASHINGTON', 'HURRICANE', 'LAVERKIN', 'SANTA CLAR
A',
                'LEEDS', 'WASH CO OTHER', 'IVINS', 'WASHCO STGEORGE', 'COVINGTO
N',
                'APPLE VALLEY', 'WASHCO WASHINGT', 'NEW HARMONY',
                'WINCHESTER HILL', 'PINTURA', 'MOHAVE COUNTY', 'ZION NAT PARK',
                'DIAMOND VALLEY', 'SPRINGDALE', 'BROWSE', 'GUNLOCK', 'VIRGIN',
                'TOQUERVILLE', 'HILDALE', 'CENTRAL', 'CLARK COUNTY', 'PINTO',
                'DAMMERON VALLEY', 'LINCOLN COUNTY', 'PINE VALLEY', 'ROCKVILLE',
                'BROOKSIDE', 'ENTERPRISE', 'VEYO', 'DIXIE COLLEGE', 'IRON COUNT
Y',
                'ENTERPRISE WCSO', 'WASHCO TOQUERVI', 'WASHCO HURRICAN',
                'WASHCO LEEDS', 'WASHCO NEWHARMO', 'WASHCO PINE VLY',
                'WASHCO VEYO', 'WASHCO IVINS', 'SHIVWITS RESERV',
                'WASHCO ENTERPRI', 'MILFORD', 'BEAVER', 'BRIGHAM CITY', 'MANTU
A',
                'GARLAND', 'PERRY', 'BOX ELDER CO', 'WILLARD', 'RIVERDALE',
                'WELLSVILLE', 'FAR WEST', 'TREMONTON', 'HONEYVILLE', 'CORINNE',
                'BRIGHAM', 'THATCHER', 'LOGAN', 'MILLVILLE', 'CACHE COUNTY',
                'HYRUM', 'NORTH LOGAN', 'NIBLEY', 'RICHMOND', 'PROVIDENCE',
                'AMALGA', 'SMITHFIELD', 'PARADISE', 'MENDON', 'LEWISTON',
                'TRENTON', 'NEWTON', 'RIVER HEIGHTS', 'CORNISH', 'HYDE PARK',
                'CLARKSTON', 'PRESTON', 'BENSON', 'LOGAN CANYON', 'EPHRAIM',
                'MANTI', 'MOUNT PLEASANT', 'SANPETE COUNTY', 'FAIRVIEW', 'CHESTE
R',
                'SPRING CITY', 'MORONI', 'CENTERFIELD', 'WALES', 'STERLING',
                'FOUNTAIN GREEN', 'GUNNISON', 'PAROWAN', 'IRON CO CC GRID',
                'CEDAR CITY', 'IRON CO BERYLG', 'ENOCH', 'IRON CO NC G',
                'KANARRAVILLE', 'PARAGONAH', 'IRON CO PAR GRI', 'IRON CO PARGG',
                'IRON CO KANG', 'IRON CO SUMMIT', 'WASH CO NEW HAR', 'BRIAN HEA
D',
                'IRON CO BHG', 'IRON CO MODG', 'DELTA', 'BERYL', 'SUMMIT',
                'SAINT GEORGE', 'NEWCASTLE', 'IRON CO ENTERG', 'MILLARD CO DEL
T',
                'NEW CASTLE', 'BRIANHEAD', 'HAMILTON FORT', 'MODENA', 'LUND',
                'BEAVER CO BG', 'HAMBLIN VALLEY', 'WASH CO ENTERPR', 'KANARAVILL
E',
                'MILLARD CO FILL', 'KANE COUNTY DUC', 'MONA', 'JUAB NEPHI',
                'EUREKA', 'LEVAN', 'NEPHI', 'JUAB CO EUR GR', 'JUAB CO NEP GR',
                'LITTLE SAHARA', 'JUAB CO LEV GR', 'JUAB CO MON GR', 'JUAB MON
A',
                'JUAB COUNTY', 'JUAB CO WES GR', 'JUAB WEST DESER', 'JUAB LEVA
N',
                'GOSHEN', 'UTAH COUNTY', 'ROCKY RIDGE', 'LEHI', 'JUAB EUREKA',
                'PROVO', 'KAYSVILLE', 'PARK CITY', 'COALVILLE', 'HOYTSTVILLE',
                'ROCKPORT', 'FRANCIS', 'SUMMIT COUNTY', 'HEBER CITY', 'KAMAS',
                'SAMAK', 'WEBER CANYON', 'MARION', 'OAKLEY', 'MURRAY',
                'SALT LAKE CITY', 'WANSHIP', 'MCKINNON', 'MIDWAY', 'PEOA',
                'HENEFER', 'PRICE', 'HELPER', 'PRICE, UTAH (CARBON)', 'WELLINGTO
N',
                'CARBONVILLE', 'EMERY COUNTY', 'OTHER', 'EAST CARBON',
                'SPRING GLEN', 'TEMP', 'KENILWORTH', 'ROY', 'SOUTH OGDEN',
                'WOODS CROSS', 'BOUNTIFUL', 'NORTH SALT LAKE', 'WEST BOUNTIFUL',
                'DAVIS COUNTY', 'LAYTON', 'FARMINGTON', 'CENTERVILLE', 'SYRACUS
E',
                'CLEARFIELD', 'SOUTH WEBER', 'COUNTY NW', 'GENOLA', 'VINEYARD',
                'WOODLAND HILLS', 'EAGLE MOUNTAIN', 'ELK RIDGE', 'CEDAR FORT',

```

```

'SPANISH FORK', 'AMERICAN FORK', 'OREM', 'FAIRFIELD', 'LINDON',
'SANTAQUIN', 'SPRINGVILLE', 'PAYSON', 'MAPLETON', 'CEDAR HILLS',
'HIGHLAND', 'PLEASANT GROVE', 'SALEM', 'ALPINE', 'SARATOGA SPRIN
G',
'DRAPER', 'TOOELE COUNTY', 'SPRING LAKE', 'BLUFFDALE',
'THISTLE/BIRDSEY', 'UTAH VALLEY U', 'PALMYRA', 'CARBON COUNTY',
'SUNDANCE', 'ELBERTA', 'SALT LAKE CNTY', 'LAKE SHORE',
'COVERED BRIDGE', 'LELAND', 'BENJAMIN', 'WASATCH COUNTY',
'BYU CAMPUS', 'UINTAH', 'DUCHESNE', 'MOAB', 'SALT LAKE', 'EMER
Y',
'KANAB', 'SALT LAKE COUNT', 'HEBER', 'FILLMORE', 'OGDEN',
'RICHFIELD', 'SARATOGA SPRINGS', 'WEBER COUNTY', 'TOOELE',
'CEDAR VALLEY', 'PROVOST', 'YINEYARD', 'CLINTON', 'FRUIT HEIGHT
S'],
dtype=object)

```

```

In [105]: cities_to_exclude2 = ['WASH CO OTHER', 'WASHCO STGEORGE', 'COVINGTON', 'V
'BROWSE', 'CLARK COUNTY', 'LINCOLN COUNTY', 'ENTERPRISE',
'WASHCO LEEDS', 'WASHCO NEWHARMO', 'WASHCO PINE VLY',
'SHIVWITS RESERV', 'WASHCO ENTERPRI', 'BOX ELDER CO',
'IRON CO NC G', 'IRON CO PAR GRI', 'IRON CO PARGG', 'IRON CO BHG', 'IRON CO MODG', 'IRON CO ENTERG', 'M
df = df[~df['city'].isin(cities_to_exclude2)]

```

```

In [106]: # Find rows where 'city' column is equal to "New Castle"
new_castle_rows = df[df['city'] == 'NEW CASTLE']
new_castle_rows

```

Out[106]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
232315	761991986	16-01490	06/11/2016 12:00:00 AM	Other	Domestic Prob	20
234120	820251554	17-02744	09/25/2017 12:00:00 AM	Emergency	Medical	

```

In [107]: # Replace 'NEW CASTLE' with 'NEWCASTLE' in the 'city' column
df['city'] = df['city'].str.replace('NEW CASTLE', 'NEWCASTLE')

```

```

In [108]: # Replace 'NEW CASTLE' with 'NEWCASTLE' in the 'city' column
df['city'] = df['city'].str.replace('BRIANHEAD', 'BRIAN HEAD')

```

```
In [109]: # Prints unique values for cities  
cities = df['city'].unique()  
cities
```

```

Out[109]: array(['ST GEORGE', 'WASHINGTON', 'HURRICANE', 'LAVERKIN', 'SANTA CLAR
A',
                'LEEDS', 'IVINS', 'APPLE VALLEY', 'NEW HARMONY', 'WINCHESTER HIL
L',
                'PINTURA', 'ZION NAT PARK', 'DIAMOND VALLEY', 'SPRINGDALE',
                'GUNLOCK', 'VIRGIN', 'TOQUERVILLE', 'HILDALE', 'CENTRAL', 'PINT
O',
                'DAMMERON VALLEY', 'PINE VALLEY', 'ROCKVILLE', 'BROOKSIDE',
                'ENTERPRISE', 'VEYO', 'DIXIE COLLEGE', 'IRON COUNTY', 'MILFORD',
                'BEAVER', 'BRIGHAM CITY', 'MANTUA', 'GARLAND', 'PERRY', 'WILLAR
D',
                'RIVERDALE', 'WELLSVILLE', 'FAR WEST', 'TREMONTON', 'HONEYVILL
E',
                'CORINNE', 'BRIGHAM', 'THATCHER', 'LOGAN', 'MILLVILLE', 'HYRUM',
                'NORTH LOGAN', 'NIBLEY', 'RICHMOND', 'PROVIDENCE', 'AMALGA',
                'SMITHFIELD', 'PARADISE', 'MENDON', 'LEWISTON', 'TRENTON',
                'NEWTON', 'RIVER HEIGHTS', 'CORNISH', 'HYDE PARK', 'CLARKSTON',
                'PRESTON', 'BENSON', 'LOGAN CANYON', 'EPHRAIM', 'MANTI',
                'MOUNT PLEASANT', 'SANPETE COUNTY', 'FAIRVIEW', 'CHESTER',
                'SPRING CITY', 'MORONI', 'CENTERFIELD', 'WALES', 'STERLING',
                'FOUNTAIN GREEN', 'GUNNISON', 'PAROWAN', 'CEDAR CITY', 'ENOCH',
                'KANARRAVILLE', 'PARAGONAH', 'BRIAN HEAD', 'DELTA', 'BERYL',
                'SUMMIT', 'SAINT GEORGE', 'NEWCASTLE', 'HAMILTON FORT', 'MODEN
A',
                'LUND', 'BEAVER CO BG', 'HAMBLIN VALLEY', 'WASH CO ENTERPR',
                'KANARAVILLE', 'MILLARD CO FILL', 'KANE COUNTY DUC', 'MONA',
                'JUAB NEPHI', 'EUREKA', 'LEVAN', 'NEPHI', 'JUAB CO EUR GR',
                'JUAB CO NEP GR', 'LITTLE SAHARA', 'JUAB CO LEV GR',
                'JUAB CO MON GR', 'JUAB MONA', 'JUAB COUNTY', 'JUAB CO WES GR',
                'JUAB WEST DESER', 'JUAB LEVAN', 'GOSHEN', 'UTAH COUNTY',
                'ROCKY RIDGE', 'LEHI', 'JUAB EUREKA', 'PROVO', 'KAYSVILLE',
                'PARK CITY', 'COALVILLE', 'HOYTSTVILLE', 'ROCKPORT', 'FRANCIS',
                'SUMMIT COUNTY', 'HEBER CITY', 'KAMAS', 'SAMAK', 'WEBER CANYON',
                'MARION', 'OAKLEY', 'MURRAY', 'SALT LAKE CITY', 'WANSHIP',
                'MCKINNON', 'MIDWAY', 'PEOA', 'HENEFER', 'PRICE', 'HELPER',
                'PRICE, UTAH (CARBON)', 'WELLINGTON', 'CARBONVILLE',
                'EMERY COUNTY', 'OTHER', 'EAST CARBON', 'SPRING GLEN', 'TEMP',
                'KENILWORTH', 'ROY', 'SOUTH OGDEN', 'WOODS CROSS', 'BOUNTIFUL',
                'NORTH SALT LAKE', 'WEST BOUNTIFUL', 'DAVIS COUNTY', 'LAYTON',
                'FARMINGTON', 'CENTERVILLE', 'SYRACUSE', 'CLEARFIELD',
                'SOUTH WEBER', 'COUNTY NW', 'GENOLA', 'VINEYARD', 'WOODLAND HILL
S',
                'EAGLE MOUNTAIN', 'ELK RIDGE', 'CEDAR FORT', 'SPANISH FORK',
                'AMERICAN FORK', 'OREM', 'FAIRFIELD', 'LINDON', 'SANTAQUIN',
                'SPRINGVILLE', 'PAYSON', 'MAPLETON', 'CEDAR HILLS', 'HIGHLAND',
                'PLEASANT GROVE', 'SALEM', 'ALPINE', 'SARATOGA SPRING', 'DRAPE
R',
                'TOOELE COUNTY', 'SPRING LAKE', 'BLUFFDALE', 'THISTLE/BIRDSEY',
                'UTAH VALLEY U', 'PALMYRA', 'CARBON COUNTY', 'SUNDANCE', 'ELBERT
A',
                'SALT LAKE CNTY', 'LAKE SHORE', 'COVERED BRIDGE', 'LELAND',
                'BENJAMIN', 'WASATCH COUNTY', 'BYU CAMPUS', 'UINTAH', 'DUCHESN
E',
                'MOAB', 'SALT LAKE', 'EMERY', 'KANAB', 'SALT LAKE COUNT', 'HEBE
R',
                'FILLMORE', 'OGDEN', 'RICHFIELD', 'SARATOGA SPRINGS',

```

```
In [110]: cities_to_exclude3 = ['BEAVER CO BG', 'WASH CO ENTERPR', 'WASH CO ENTERPR',
                                'JUAB NEPHI', 'JUAB CO EUR GR', 'JUAB CO NEP GR', 'JUAB CO NEP GR',
                                'JUAB COUNTY', 'JUAB CO WES GR', 'JUAB WEST DESER',
                                'JUAB WEST DESER']

df = df[~df['city'].isin(cities_to_exclude3)]
```



```
In [111]: # Prints unique values for cities  
cities = df['city'].unique()  
cities
```

```

Out[111]: array(['ST GEORGE', 'WASHINGTON', 'HURRICANE', 'LAVERKIN', 'SANTA CLAR
A',
                'LEEDS', 'IVINS', 'APPLE VALLEY', 'NEW HARMONY', 'WINCHESTER HIL
L',
                'PINTURA', 'ZION NAT PARK', 'DIAMOND VALLEY', 'SPRINGDALE',
                'GUNLOCK', 'VIRGIN', 'TOQUERVILLE', 'HILDALE', 'CENTRAL', 'PINT
O',
                'DAMMERON VALLEY', 'PINE VALLEY', 'ROCKVILLE', 'BROOKSIDE',
                'ENTERPRISE', 'VEYO', 'DIXIE COLLEGE', 'IRON COUNTY', 'MILFORD',
                'BEAVER', 'BRIGHAM CITY', 'MANTUA', 'GARLAND', 'PERRY', 'WILLAR
D',
                'RIVERDALE', 'WELLSVILLE', 'FAR WEST', 'TREMONTON', 'HONEYVILL
E',
                'CORINNE', 'BRIGHAM', 'THATCHER', 'LOGAN', 'MILLVILLE', 'HYRUM',
                'NORTH LOGAN', 'NIBLEY', 'RICHMOND', 'PROVIDENCE', 'AMALGA',
                'SMITHFIELD', 'PARADISE', 'MENDON', 'LEWISTON', 'TRENTON',
                'NEWTON', 'RIVER HEIGHTS', 'CORNISH', 'HYDE PARK', 'CLARKSTON',
                'PRESTON', 'BENSON', 'LOGAN CANYON', 'EPHRAIM', 'MANTI',
                'MOUNT PLEASANT', 'SANPETE COUNTY', 'FAIRVIEW', 'CHESTER',
                'SPRING CITY', 'MORONI', 'CENTERFIELD', 'WALES', 'STERLING',
                'FOUNTAIN GREEN', 'GUNNISON', 'PAROWAN', 'CEDAR CITY', 'ENOCH',
                'KANARRAVILLE', 'PARAGONAH', 'BRIAN HEAD', 'DELTA', 'BERYL',
                'SUMMIT', 'SAINT GEORGE', 'NEWCASTLE', 'HAMILTON FORT', 'MODEN
A',
                'LUND', 'HAMBLIN VALLEY', 'KANARAVILLE', 'MONA', 'EUREKA', 'LEVA
N',
                'NEPHI', 'LITTLE SAHARA', 'GOSHEN', 'UTAH COUNTY', 'ROCKY RIDG
E',
                'LEHI', 'JUAB EUREKA', 'PROVO', 'KAYSVILLE', 'PARK CITY',
                'COALVILLE', 'HOYTSTVILLE', 'ROCKPORT', 'FRANCIS', 'SUMMIT COUNT
Y',
                'HEBER CITY', 'KAMAS', 'SAMAK', 'WEBER CANYON', 'MARION', 'OAKLE
Y',
                'MURRAY', 'SALT LAKE CITY', 'WANSHIP', 'MCKINNON', 'MIDWAY',
                'PEOA', 'HENEFER', 'PRICE', 'HELPER', 'PRICE, UTAH (CARBON)',
                'WELLINGTON', 'CARBONVILLE', 'EMERY COUNTY', 'OTHER',
                'EAST CARBON', 'SPRING GLEN', 'TEMP', 'KENILWORTH', 'ROY',
                'SOUTH OGDEN', 'WOODS CROSS', 'BOUNTIFUL', 'NORTH SALT LAKE',
                'WEST BOUNTIFUL', 'DAVIS COUNTY', 'LAYTON', 'FARMINGTON',
                'CENTERVILLE', 'SYRACUSE', 'CLEARFIELD', 'SOUTH WEBER',
                'COUNTY NW', 'GENOLA', 'VINEYARD', 'WOODLAND HILLS',
                'EAGLE MOUNTAIN', 'ELK RIDGE', 'CEDAR FORT', 'SPANISH FORK',
                'AMERICAN FORK', 'OREM', 'FAIRFIELD', 'LINDON', 'SANTAQUIN',
                'SPRINGVILLE', 'PAYSON', 'MAPLETON', 'CEDAR HILLS', 'HIGHLAND',
                'PLEASANT GROVE', 'SALEM', 'ALPINE', 'SARATOGA SPRING', 'DRAPE
R',
                'TOOELE COUNTY', 'SPRING LAKE', 'BLUFFDALE', 'THISTLE/BIRDSEY',
                'UTAH VALLEY U', 'PALMYRA', 'CARBON COUNTY', 'SUNDANCE', 'ELBERT
A',
                'SALT LAKE CNTY', 'LAKE SHORE', 'COVERED BRIDGE', 'LELAND',
                'BENJAMIN', 'WASATCH COUNTY', 'BYU CAMPUS', 'UINTAH', 'DUCESN
E',
                'MOAB', 'SALT LAKE', 'EMERY', 'KANAB', 'SALT LAKE COUNT', 'HEBE
R',
                'FILLMORE', 'OGDEN', 'RICHFIELD', 'SARATOGA SPRINGS',

```

```
'WEBER COUNTY', 'TOOELE', 'CEDAR VALLEY', 'PROVOST', 'YINEYARD',  
'CLINTON', 'FRUIT HEIGHTS'], dtype=object)
```

```
In [112]: # Replace 'PRICE, UTAH (CARBON)' with 'PRICE' in the 'city' column  
df['city'] = df['city'].str.replace('PRICE, UTAH (CARBON)', 'PRICE')
```

```
In [113]: cities_to_exclude4 = ['JUAB EUREKA', 'OTHER', 'DAVIS COUNTY', 'SUMMIT COUNTY',  
                                'EMERY COUNTY', 'COUNTY NW', 'CARBON COUNTY', 'TOOELE',  
                                'UTAH COUNTY', 'WASATCH COUNTY'] # List of cities to  
  
df = df[~df['city'].isin(cities_to_exclude4)]
```

```
In [114]: # Prints unique values for cities  
cities = df['city'].unique()  
cities
```

```

Out[114]: array(['ST GEORGE', 'WASHINGTON', 'HURRICANE', 'LAVERKIN', 'SANTA CLAR
A',
                'LEEDS', 'IVINS', 'APPLE VALLEY', 'NEW HARMONY', 'WINCHESTER HIL
L',
                'PINTURA', 'ZION NAT PARK', 'DIAMOND VALLEY', 'SPRINGDALE',
                'GUNLOCK', 'VIRGIN', 'TOQUERVILLE', 'HILDALE', 'CENTRAL', 'PINT
O',
                'DAMMERON VALLEY', 'PINE VALLEY', 'ROCKVILLE', 'BROOKSIDE',
                'ENTERPRISE', 'VEYO', 'DIXIE COLLEGE', 'IRON COUNTY', 'MILFORD',
                'BEAVER', 'BRIGHAM CITY', 'MANTUA', 'GARLAND', 'PERRY', 'WILLAR
D',
                'RIVERDALE', 'WELLSVILLE', 'FAR WEST', 'TREMONTON', 'HONEYVILL
E',
                'CORINNE', 'BRIGHAM', 'THATCHER', 'LOGAN', 'MILLVILLE', 'HYRUM',
                'NORTH LOGAN', 'NIBLEY', 'RICHMOND', 'PROVIDENCE', 'AMALGA',
                'SMITHFIELD', 'PARADISE', 'MENDON', 'LEWISTON', 'TRENTON',
                'NEWTON', 'RIVER HEIGHTS', 'CORNISH', 'HYDE PARK', 'CLARKSTON',
                'PRESTON', 'BENSON', 'LOGAN CANYON', 'EPHRAIM', 'MANTI',
                'MOUNT PLEASANT', 'FAIRVIEW', 'CHESTER', 'SPRING CITY', 'MORON
I',
                'CENTERFIELD', 'WALES', 'STERLING', 'FOUNTAIN GREEN', 'GUNNISO
N',
                'PAROWAN', 'CEDAR CITY', 'ENOCH', 'KANARRAVILLE', 'PARAGONAH',
                'BRIAN HEAD', 'DELTA', 'BERYL', 'SUMMIT', 'SAINT GEORGE',
                'NEWCASTLE', 'HAMILTON FORT', 'MODENA', 'LUND', 'HAMBLIN VALLE
Y',
                'KANARAVILLE', 'MONA', 'EUREKA', 'LEVAN', 'NEPHI', 'LITTLE SAHAR
A',
                'GOSHEN', 'ROCKY RIDGE', 'LEHI', 'PROVO', 'KAYSVILLE', 'PARK CIT
Y',
                'COALVILLE', 'HOYTSVILLE', 'ROCKPORT', 'FRANCIS', 'HEBER CITY',
                'KAMAS', 'SAMAK', 'WEBER CANYON', 'MARION', 'OAKLEY', 'MURRAY',
                'SALT LAKE CITY', 'WANSHIP', 'MCKINNON', 'MIDWAY', 'PEOA',
                'HENEFER', 'PRICE', 'HELPER', 'WELLINGTON', 'CARBONVILLE',
                'EAST CARBON', 'SPRING GLEN', 'KENILWORTH', 'ROY', 'SOUTH OGDE
N',
                'WOODS CROSS', 'BOUNTIFUL', 'NORTH SALT LAKE', 'WEST BOUNTIFUL',
                'LAYTON', 'FARMINGTON', 'CENTERVILLE', 'SYRACUSE', 'CLEARFIELD',
                'SOUTH WEBER', 'GENOLA', 'VINEYARD', 'WOODLAND HILLS',
                'EAGLE MOUNTAIN', 'ELK RIDGE', 'CEDAR FORT', 'SPANISH FORK',
                'AMERICAN FORK', 'OREM', 'FAIRFIELD', 'LINDON', 'SANTAQUIN',
                'SPRINGVILLE', 'PAYSON', 'MAPLETON', 'CEDAR HILLS', 'HIGHLAND',
                'PLEASANT GROVE', 'SALEM', 'ALPINE', 'SARATOGA SPRING', 'DRAPE
R',
                'SPRING LAKE', 'BLUFFDALE', 'THISTLE/BIRDSEY', 'UTAH VALLEY U',
                'PALMYRA', 'SUNDANCE', 'ELBERTA', 'LAKE SHORE', 'COVERED BRIDG
E',
                'LELAND', 'BENJAMIN', 'BYU CAMPUS', 'UINTAH', 'DUCHESNE', 'MOA
B',
                'SALT LAKE', 'EMERY', 'KANAB', 'HEBER', 'FILLMORE', 'OGDEN',
                'RICHFIELD', 'SARATOGA SPRINGS', 'TOOELE', 'CEDAR VALLEY',
                'PROVOST', 'YINEYARD', 'CLINTON', 'FRUIT HEIGHTS'], dtype=objec
t)

```

```
In [115]: # Find rows where 'city' column is equal to "SALT LAKE"
SALT_LAKE_rows = df[df['city'] == 'SALT LAKE']
SALT_LAKE_rows
```

Out[115]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
382709	2850990	08UC07934	07/21/2008 12:00:00 AM	Proactive Policing	Description: WARRANT	
382929	2973079	08UC08343	07/30/2008 12:00:00 AM	Proactive Policing	Description: WARRANT	
383539	3110001	08UC08828	08/11/2008 12:00:00 AM	Proactive Policing	Description: WARRANT	BI
384479	3235791	08UC09102	08/18/2008 12:00:00 AM	Proactive Policing	Description: WARRANT	BI

```
In [116]: # Drop rows where 'city' column is equal to "SALT LAKE" from the original df
df = df.drop(SALT_LAKE_rows.index)
```

```
In [117]: # Prints unique values for cities
cities = df['city'].unique()
cities
```

```
Out[117]: array(['ST GEORGE', 'WASHINGTON', 'HURRICANE', 'LAVERKIN', 'SANTA CLARA',
        'LEEDS', 'IVINS', 'APPLE VALLEY', 'NEW HARMONY', 'WINCHESTER HILL',
        'PINTURA', 'ZION NAT PARK', 'DIAMOND VALLEY', 'SPRINGDALE',
        'GUNLOCK', 'VIRGIN', 'TOQUERVILLE', 'HILDALE', 'CENTRAL', 'PINTO',
        'DAMMERON VALLEY', 'PINE VALLEY', 'ROCKVILLE', 'BROOKSIDE',
        'ENTERPRISE', 'VEYO', 'DIXIE COLLEGE', 'IRON COUNTY', 'MILFORD',
        'BEAVER', 'BRIGHAM CITY', 'MANTUA', 'GARLAND', 'PERRY', 'WILLARD',
        'RIVERDALE', 'WELLSVILLE', 'FAR WEST', 'TREMONTON', 'HONEYVILLE',
        'CORINNE', 'BRIGHAM', 'THATCHER', 'LOGAN', 'MILLVILLE', 'HYRUM',
        'NORTH LOGAN', 'NIBLEY', 'RICHMOND', 'PROVIDENCE', 'AMALGA',
        'SMITHFIELD', 'PARADISE', 'MENDON', 'LEWISTON', 'TRENTON',
        'NEWTON', 'RIVER HEIGHTS', 'CORNISH', 'HYDE PARK', 'CLARKSTON',
        ...])
```

```
In [118]: # Find duplicate rows
duplicate_rows = df[df.duplicated()]
# Print the resulting DataFrame containing duplicate rows
print("Duplicate Rows:")
duplicate_rows
```

Duplicate Rows:

Out[118]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
	99999	544003266	22978	12/1/1997 20:12	Pedestrian Stop	SUSP OTHER
267811	913039167	19-B02493	04/05/2019 11:41:06 AM	Weapons Offense	Weapon Offense	10 N
267812	832915287	17-B09872	12/22/2017 08:33:10 AM	Other	Suspicious	90 ;
267813	832915358	17-B09944	12/22/2017 11:00:00 AM	Traffic	PD Accident	80 W
267814	832915293	17-B09878	12/22/2017 12:16:37 PM	Other	Citizen Assist	1 N
...
445050	378861433	1403980	01/26/2014 05:38:58 PM	Other	HARASSMENT	40
445051	378861434	1403979	01/26/2014 05:33:38 PM	Community Policing	WARRANT SERVICE	H
445052	378861435	1403976	01/26/2014 05:06:35 PM	Community Policing	WARRANT SERVICE	10 S
445053	378861438	1403958	01/26/2014 12:28:34 PM	Community Policing	CIVIL STANDBY	92
445054	378861439	1403954	01/26/2014 11:26:17 AM	Other	1050 PD	50 S O

29905 rows × 14 columns

```
In [119]: # Identify rows that appear more than once
duplicated_rows = df[df.duplicated(keep=False)]

# Sort the resulting DataFrame
duplicated_rows = duplicated_rows.sort_values(by=list(df.columns))

# Print the resulting sorted DataFrame containing rows that appear more
print("Rows that Appear More than Once (Sorted):")
duplicated_rows.head(20)
```

Rows that Appear More than Once (Sorted):

Out [119]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
327731	83579655	1005671	02/15/2010 02:33:45 AM	Other	1047 PER	40
430131	83579655	1005671	02/15/2010 02:33:45 AM	Other	1047 PER	40
327948	83581148	1013984	04/20/2010 04:14:27 PM	Other	1090 RES	40
430349	83581148	1013984	04/20/2010 04:14:27 PM	Other	1090 RES	40
328143	83582141	1020691	06/11/2010 12:45:18 AM	Other	1075	40
430544	83582141	1020691	06/11/2010 12:45:18 AM	Other	1075	40
328545	83583929	1034040	09/17/2010 12:34:26 PM	Other	1075	40
430945	83583929	1034040	09/17/2010 12:34:26 PM	Other	1075	40
328690	83584765	1039807	11/01/2010 09:23:20 PM	Theft	THEFT	50
431090	83584765	1039807	11/01/2010 09:23:20 PM	Theft	THEFT	50
328135	83587220	1110337	03/19/2011 07:23:44 PM	Community Policing	OUTSIDE ASSIST	50
430536	83587220	1110337	03/19/2011 07:23:44 PM	Community Policing	OUTSIDE ASSIST	50
328136	83587221	1110444	03/20/2011 03:48:26 PM	Community Policing	SUSPICIOUS CIRCUMSTANCE	50
430537	83587221	1110444	03/20/2011 03:48:26 PM	Community Policing	SUSPICIOUS CIRCUMSTANCE	50
328137	83587222	1109999	03/17/2011 11:28:57 AM	Other	911 TRACE	50

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
	430538	83587222	1109999	03/17/2011 11:28:57 AM	Other	911 TRACE
	328138	83587223	1110821	03/23/2011 01:27:29 PM	Other	VIN INSPECTION
	430539	83587223	1110821	03/23/2011 01:27:29 PM	Other	VIN INSPECTION
	328139	83587224	1111401	03/28/2011 05:58:33 AM	Other	1090 COM 90
	430540	83587224	1111401	03/28/2011 05:58:33 AM	Other	1090 COM 90

```
In [120]: # Convert 'case_number' column to a consistent type (e.g., string) and to
df['case_number'] = df['case_number'].astype(str)
duplicates = df[df.duplicated(subset='case_number', keep=False)].sort_val

# Print or further process the sorted duplicates
duplicates
```

Out[120]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
	430131	83579655	1005671	02/15/2010 02:33:45 AM	Other	1047 PER
	327731	83579655	1005671	02/15/2010 02:33:45 AM	Other	1047 PER
	327948	83581148	1013984	04/20/2010 04:14:27 PM	Other	1090 RES
	430349	83581148	1013984	04/20/2010 04:14:27 PM	Other	1090 RES
	328143	83582141	1020691	06/11/2010 12:45:18 AM	Other	1075

	192907	913011626	57840	4/5/2019 14:57	Traffic	Traffic Hazard
	130591	544164324	57844	4/8/1999 14:04	Theft	THEFT-RETAIL
	192910	913011622	57844	4/5/2019 20:19	Other	Lockout
	192914	913082732	57849	4/6/2019 18:56	Traffic	Livestock Probl
	130306	544164219	57849	4/8/1999 15:04	Traffic	ABANDONED VEHIC

77935 rows × 14 columns

```
In [121]: # Drop duplicates and keep only one instance
df = df.drop_duplicates(keep='first')

# Print the resulting DataFrame without duplicates
print("DataFrame without Duplicates:")
df
```

DataFrame without Duplicates:

Out[121]:

	incident_id	case_number	incident_datetime	parent_incident_type	incident_description	ad
0	691160812	15P003395	2/11/2015 16:02	Theft	FRAUD	
1	712757140	15P008661	4/13/2015 13:04	Drugs	DRUGS	1
2	715044076	15P010661	5/5/2015 22:05	Drugs	DRUGS	I-1
3	825529671	17P027337	11/5/2017 5:11	Traffic	DUI	RI
4	736930480	15P027109	11/12/2015 23:11	Traffic	DUI	
...	
479659	808705608	1729112	06/09/2017 08:46:12 AM	Other	1090 COM	
479660	808705606	1729120	06/09/2017 09:26:42 AM	Community Policing	OUTSIDE ASSIST	60
479661	808705604	1729143	06/09/2017 11:40:39 AM	Community Policing	VACATION WATCH	70
479662	808705600	1729197	06/09/2017 04:05:21 PM	Other	C/S VIOLATIONS	90
479663	808705594	1729279	06/10/2017 01:57:04 AM	Disorder	NOISE DISTURBANCE	

418575 rows × 14 columns

```
In [122]: # Prints unique values for case_number column
unique_case_numbers = df['case_number'].nunique()
print("Number of unique case_numbers:", unique_case_numbers)
```

Number of unique case_numbers: 409264

```
In [123]: # Prints unique values for incident_id column
unique_incident_ids = df['incident_id'].nunique()
print("Number of unique incident_ids:", unique_incident_ids)
```

Number of unique incident_ids: 418575

```
In [124]: # Drop 'case_number' column since it contains duplicate values using "in
# Drop the 'case_number' column
df = df.drop('case_number', axis=1)
df
```

Out[124]:

	incident_id	incident_datetime	parent_incident_type	incident_description	address_1	
0	691160812	2/11/2015 16:02	Theft	FRAUD	BY 21	GE
1	712757140	4/13/2015 13:04	Drugs	DRUGS	1 Block N 200 E	GE
2	715044076	5/5/2015 22:05	Drugs	DRUGS	I-15 SB X8 ONR	GE
3	825529671	11/5/2017 5:11	Traffic	DUI	4800 Block S RIVER RD	GE
4	736930480	11/12/2015 23:11	Traffic	DUI	I-15 NB MM 8	GE
...
479659	808705608	06/09/2017 08:46:12 AM	Other	1090 COM	1400 Block S 1800 WEST	W C
479660	808705606	06/09/2017 09:26:42 AM	Community Policing	OUTSIDE ASSIST	600 Block S 700 WEST	W C
479661	808705604	06/09/2017 11:40:39 AM	Community Policing	VACATION WATCH	2000 Block S 700 WEST	W C
479662	808705600	06/09/2017 04:05:21 PM	Other	C/S VIOLATIONS	900 Block W 500 SOUTH	BOUN
479663	808705594	06/10/2017 01:57:04 AM	Disorder	NOISE DISTURBANCE	1800 Block S 1200 WEST	W C

418575 rows × 13 columns

VALIDITY OF DATA VALUES

We will also need to validate the locations of the files, on initial analysis, it was noticed the the latitude and longitudes did not all reside in Utah. As shown in the image/code below. We will likely have to filter latitude and longitude ranges by: 35 thru 42, and -114 thru -108

```
In [125]: #first we need to select our latitude and longitude range
#selecting our latitude range
loc_df_1 = df[(df["latitude"] >= 35) & (df["latitude"] <= 42)]

#selecting our longitude range
loc_df_2 = loc_df_1[(loc_df_1["longitude"] >= -114) & (loc_df_1["longitude"] <= -110)]

loc_df_2.head()
```

Out[125]:

	incident_id	incident_datetime	parent_incident_type	incident_description	address_1	city	lat
0	691160812	2/11/2015 16:02	Theft	FRAUD	BY 21	ST GEORGE	35.8811
6	112983061	4/2/2012 17:04	Disorder	CIVIL	BY 21	ST GEORGE	35.8811
7	112983974	4/10/2012 13:04	Theft	THEFT-MISDEMEAN	BY 21	ST GEORGE	35.8811
8	112984779	4/15/2012 0:04	Traffic	HIT AND RUN	BY 21	ST GEORGE	35.8811
9	112989665	5/21/2012 10:05	Traffic	PARKING PROBLEM	BY 21	ST GEORGE	35.8811

DATA SIZE

After merging all the datasets together, we ended up with about 4millions rows of data. This was a lot of data to process, so we needed to train/test split this data to get a smaller amount without losing the variety of the dataset.

```
In [145]: from sklearn.model_selection import train_test_split
# Assuming 'target_column' is your target variable
X = loc_df_2.drop('incident_id', axis=1) # Features
y = loc_df_2['Crime_Category'] # Target variable

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1,

X_test['Total'] = X_test.groupby('Crime_Category')['Crime_Category'].tra

print(len(X_test))
print(X_test.head())
```

37918

	incident_datetime	parent_incident_type	\
323604	06/07/2013 04:57:57 PM	Breaking & Entering	
475748	01/02/2012 08:48:33 PM	Vehicle Stop	
27969	10/27/2011 0:10	Theft	
104652	9/4/1998 17:09	Property Crime	
284811	03/26/2013 10:45:00 AM	Drugs	

	incident_description	\
323604	BURGLARY (10-91)	
475748	T / TRAFFIC STOP	
27969	FRAUD	
104652	CRIM MISCHIEF	
284811	"CONTROLLED SUBSTANCE / Amphetamine, Sell"	

	address_1	city	latitude	longitude
323604	800 Block 5450 S	SOUTH OGDEN	41.164021	-111.960445
475748	1700 Block W 500 S	WOODS CROSS	40.884085	-111.928060
27969	700 Block S INDIAN HILLS DR	ST GEORGE	37.096300	-113.611000
104652	1400 Block N DIXIE DOWNS	ST GEORGE	37.132946	-113.622726
284811	1 Block W MAIN STREET	PRICE	39.599536	-110.811287

	hour_of_day	day_of_week	created_at	\
323604	16	Friday	06/08/2013 09:23:37 AM	
475748	20	Monday	01/03/2012 09:21:11 AM	
27969	0	Thursday	5/23/2012 23:28	
104652	17	Friday	9/4/2014 20:44	
284811	10	Tuesday	04/16/2013 07:08:49 AM	

	updated_at	Crime_Category	Total
323604	06/21/2013 09:23:33 AM	Breaking & Entering	727
475748	01/09/2012 09:21:12 AM	Vehicle Stop	3381
27969	9/5/2014 3:22	Theft	3330
104652	9/5/2014 1:00	Property Crime	2205
284811	09/21/2013 07:08:01 AM	Drugs	1168

Analysis Questions

Primary questions: Our primary focus will be on the most common types of crimes, that we discovered in our initial data analysis

1) Are there crime trends in different cities? 2) Are there crime trends with different years? 3) What different types of crime occurs, and which one is most prevalent per city? 4) Does the day in the week effect the amount of crime and what type of crime occurs?

We would like to learn how to display and visualize this data in an unbiased and straightforward fashion. In the long-run, if a visualization worked it could be used to determine which areas in Utah to live. However, this could arise another issue with how a visualization can be harmful than helpful.

DESIGN IDEAS

MAP of Crime Data

We started with a map of the crime data, which is helpful in it's own way, but we will need to think about how to make this filterable and easier to read. Right now we can kind of see a trend that crime is occurring along the freeway, but you really have to zoom in to see what types of crimes are occurring and where.

```
In [182]: from IPython.display import HTML

HTML("""
    <video alt="test" controls style="width: 600px; height: 400px;">
        <source src="map.mp4" type="video/mp4">
    </video>
""")
```

Out[182]:

0:00 / 0:53

Time Series and Location

The plan for this visualization is to be able to see at where the crime occurred at a specific year and at the specific location. The goal is to be able to add a corresponding bar chart to be able to break down the type of crime that also occurred at that location.

Just get the Years of our Dates so we can do a Time Series visualization

```
In [147]: # Convert the datetime column to datetime format
X_test['datetime_column'] = pd.to_datetime(X_test['created_at'], format=

# Format the datetime column as desired
X_test['formatted_column'] = X_test['datetime_column'].dt.strftime('%m/%

X_test['final_date_column'] = pd.to_datetime(X_test['formatted_column'])

# Extract the year and create a new column
X_test['year_column'] = X_test['final_date_column'].dt.year

print(X_test['year_column'])
```

```
323604    2013
475748    2012
27969     2012
104652    2014
284811    2013
...
298521    2012
445849    2014
320939    2013
138519    2014
106708    2014
Name: year_column, Length: 37918, dtype: int32
```

```
In [148]: # Group by year and count rows
X_test['Total_Year'] = X_test['year_column'].groupby(X_test['year_column'])
```

```
In [149]: skip = 50

line = alt.Chart(X_test.iloc[::skip, :]).mark_line(tooltip=True).encode(
    x=alt.X('year_column:0'),
    y=alt.Y('Total_Year:Q', scale=alt.Scale(zero=False))
)

# Create a selection that chooses the nearest point & selects based on x
nearest = alt.selection(type='single', nearest=True, on='mouseover',
                        fields=['year_column'])

# Transparent selectors across the chart. This is what tells us
# the x-value of the cursor
selectors = alt.Chart(X_test.iloc[::skip, :]).mark_point().encode(
    x='year_column:0',
    opacity=alt.value(0),
).add_params(
    nearest
)

# Draw points on the line, and highlight based on selection
points = line.mark_point(color='red').encode(
    opacity=alt.condition(nearest, alt.value(1), alt.value(0))
)

# Draw a rule at the location of the selection
rules = alt.Chart(X_test.iloc[::skip, :]).mark_rule(color='gray').encode(
    x='year_column:0',
).transform_filter(
    nearest
)

#line.mark_line() + selectors + points + rules
```

/Users/roannarague/opt/anaconda3/lib/python3.8/site-packages/altair/utis/deprecation.py:65: AltairDeprecationWarning: 'selection' is deprecated.

Use 'selection_point()' or 'selection_interval()' instead; these functions also include more helpful docstrings.

warnings.warn(message, AltairDeprecationWarning, stacklevel=1)
 /Users/roannarague/opt/anaconda3/lib/python3.8/site-packages/altair/vegalite/v5/api.py:450: AltairDeprecationWarning: The types 'single' and 'multi' are now

combined and should be specified using "selection_point()".
 warnings.warn(

```

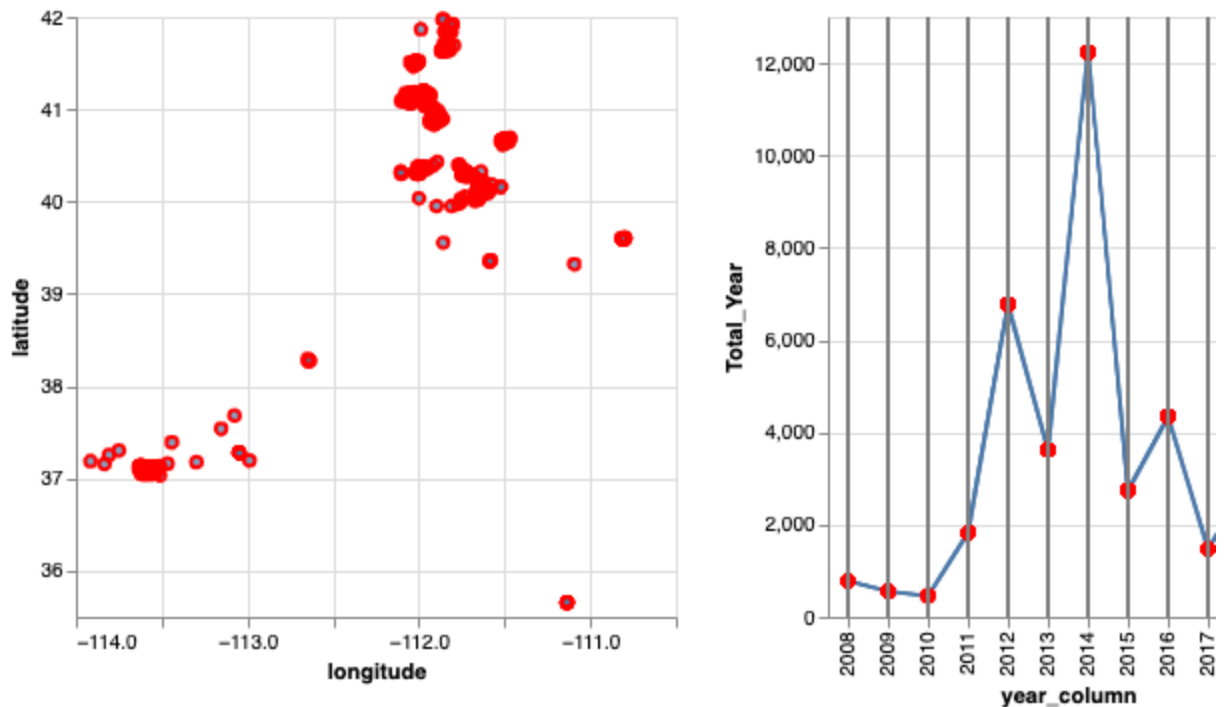
In [150]: map = alt.Chart(X_test.iloc[:, :]).mark_circle().encode(
            x=alt.X('longitude:Q', scale=alt.Scale(zero=False)),
            y=alt.Y('latitude:Q', scale=alt.Scale(zero=False)),
            order='year_column'
        ).project('albersUsa')

map_point = map.mark_point(color='red').encode(
    opacity=alt.condition(nearest, alt.value(1), alt.value(0))
)

map + map_point | line + line.mark_line() + selectors + points + rules

```

Out[150]:



The second graph combines the Type of Crime as a table and then highlights where this crime occurred on the map when you click on it. The idea is to somehow include this as a bar chart instead.

```
In [179]: crime_cat = alt.selection(type='single', nearest=True, on='mouseover',
                                   fields=['Crime_Category'])

splits = alt.Chart(X_test.iloc[:, :]).mark_text(fontWeight='lighter',
y=alt.Y('Crime_Category:0', axis=None),
text='Crime_Category:N',
opacity=alt.value(1)
).properties(
    height=200
).add_params(
    crime_cat
)

map = alt.Chart(X_test.iloc[:, :]).mark_circle().encode(
    x=alt.X('longitude:Q', scale=alt.Scale(zero=False)),
    y=alt.Y('latitude:Q', scale=alt.Scale(zero=False)),
    order='created_at'
).project('albersUsa')

map_point = map.mark_circle(color='red', size=20).encode(
    opacity=alt.condition(crime_cat, alt.value(1), alt.value(0))
)

(splits.properties(title="Crime Category") | (map+map_point)).configure_
strokeWidth=0,
)
```

/Users/roannarague/opt/anaconda3/lib/python3.8/site-packages/altair/utis/deprecation.py:65: AltairDeprecationWarning:

'selection' is deprecated.

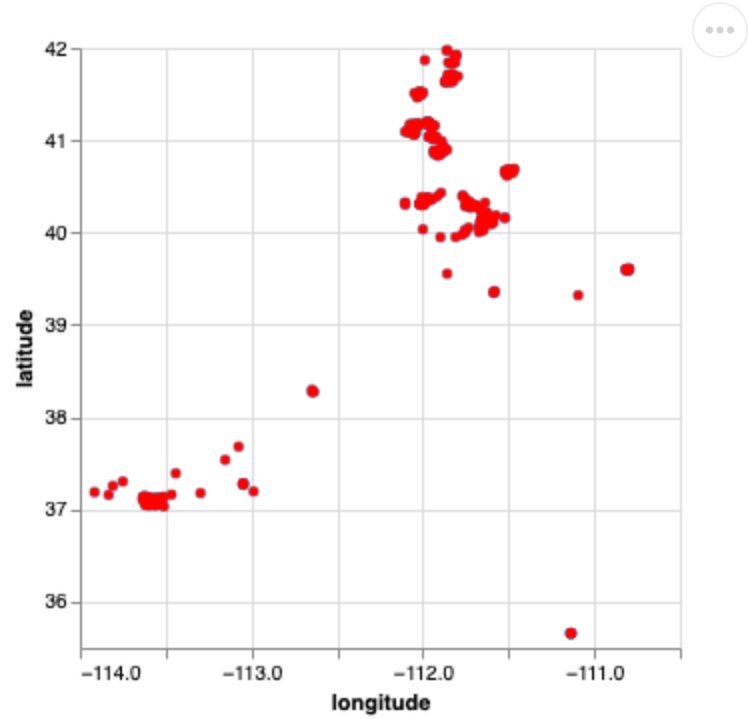
Use 'selection_point()' or 'selection_interval()' instead; these functions also include more helpful docstrings.

/Users/roannarague/opt/anaconda3/lib/python3.8/site-packages/altair/vegalite/v5/api.py:450: AltairDeprecationWarning:

The types 'single' and 'multi' are now combined and should be specified using "selection_point()".

Out[179]: **Crime Category**

- Assault
- Breaking & Entering
- Community Policing
- Disorder
- Drugs
- Family Offense
- Liquor
- Misc
- Other
- Pedestrian Stop
- Proactive Policing
- Property Crime
- Theft
- Theft from Vehicle
- Traffic
- Vehicle Stop



```
In [178]: crime_cat = alt.selection(type='single', nearest=True, on='mouseover',
                                   fields=['day_of_week'])

splits = alt.Chart(X_test.iloc[:, :]).mark_text(fontWeight='lighter',
y=alt.Y('day_of_week:0', axis=None),
text='day_of_week:N',
opacity=alt.value(1)
).properties(
    height=200
).add_params(
    crime_cat
)

map = alt.Chart(X_test.iloc[:, :]).mark_circle().encode(
    x=alt.X('longitude:Q', scale=alt.Scale(zero=False)),
    y=alt.Y('latitude:Q', scale=alt.Scale(zero=False)),
    order='created_at'
).project('albersUsa')

map_point = map.mark_circle(color='red', size=40).encode(
    opacity=alt.condition(crime_cat, alt.value(1), alt.value(0))
)

(splits.properties(title="Day Of Week") | (map+map_point)).configure_view(
    strokeWidth=0,
)
```

/Users/roannarague/opt/anaconda3/lib/python3.8/site-packages/altair/utils/deprecation.py:65: AltairDeprecationWarning:

'selection' is deprecated.

Use 'selection_point()' or 'selection_interval()' instead; these functions also include more helpful docstrings.

/Users/roannarague/opt/anaconda3/lib/python3.8/site-packages/altair/vegalite/v5/api.py:450: AltairDeprecationWarning:

The types 'single' and 'multi' are now combined and should be specified using "selection_point()".

Out[178]: **Day Of Week**

Friday

Monday

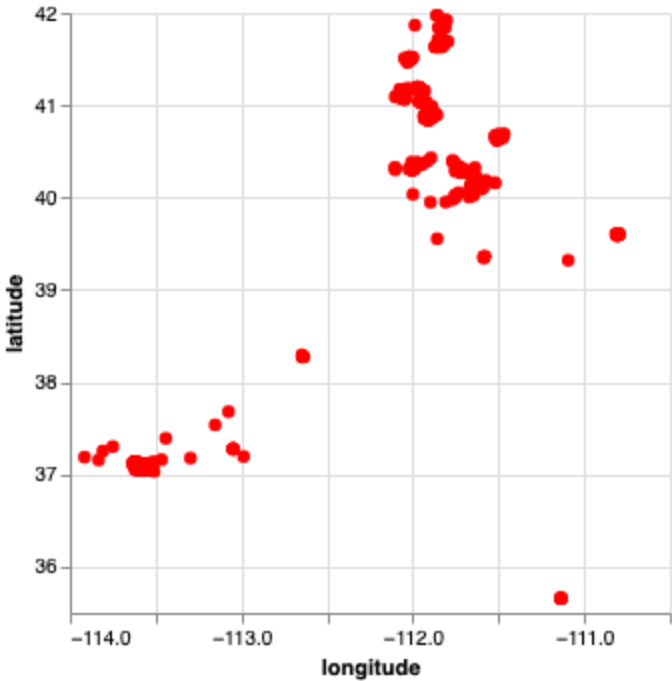
Saturday

Sunday

Thursday

Tuesday

Wednesday



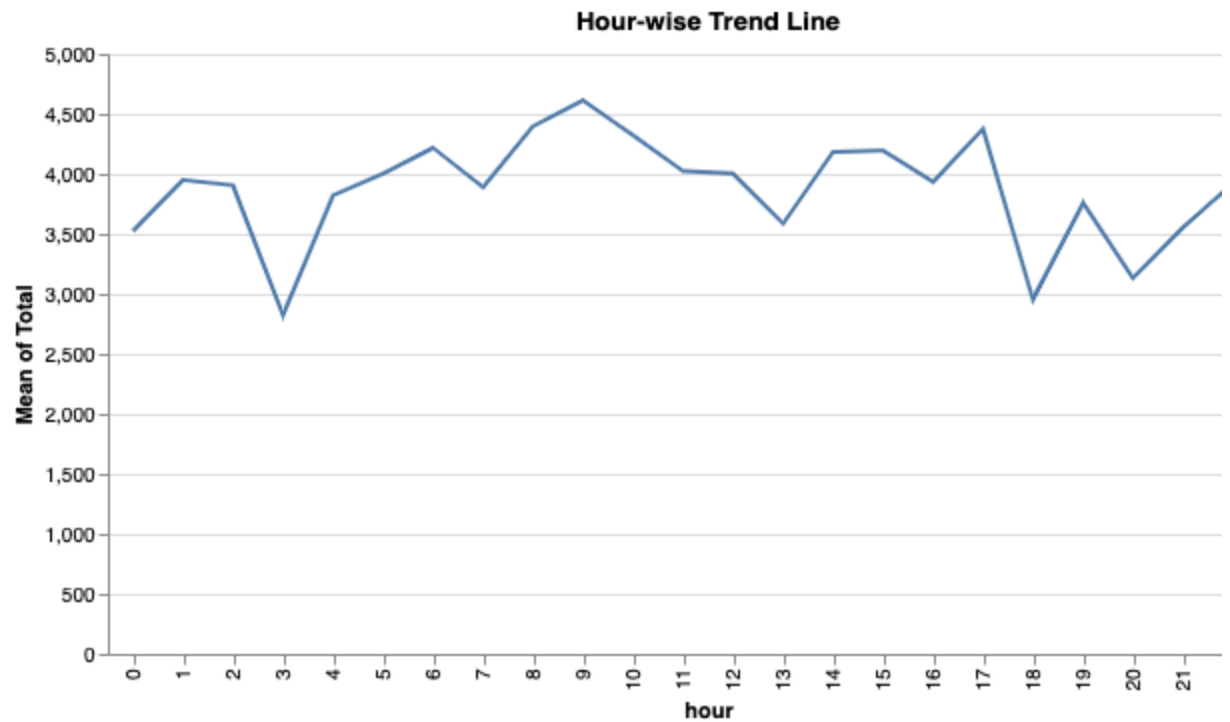
```
In [152]: alt.data_transformers.disable_max_rows()

# Extract hour from timestamp
X_test['hour'] = X_test['final_date_column'].dt.hour

# Create an Altair chart
chart = alt.Chart(X_test).mark_line().encode(
    x='hour:Q',
    y='mean(Total):Q',
    tooltip=['hour:Q', 'mean(Total):Q']
).properties(
    title='Hour-wise Trend Line',
    width=600
)

# Show the chart
chart
```

Out[152]:



Bubble Chart plotted on the Map of where the Crime Occured

```
In [175]: # Create an Altair bubble chart
chart = alt.Chart(X_test).mark_circle().encode(
    longitude='longitude:Q',
    latitude='latitude:Q',
    size='Total:Q',
    color='day_of_week:N',
    tooltip=['Crime_Category:N', 'Total:Q']
).properties(
    title='Bubble Chart of Crime Types',
    width=600
)

# Show the chart
chart
```

Out[175]:

Bubble Chart of Crime Types



```
In [174]: import pandas as pd
import plotly.express as px

# Calculate the total count of each crime category
crime_category_counts = X_test['Crime_Category'].value_counts().reset_index
crime_category_counts.columns = ['Crime_Category', 'Total']

# Create a packed bubble chart
fig = px.scatter(crime_category_counts, x='Total', y='Total', size='Total',
                 hover_name='Crime_Category', title='Packed Bubble Chart',
                 labels={'Count': 'Total Count'})

# Show the chart
fig.show()
```

NETWORK GRAPH

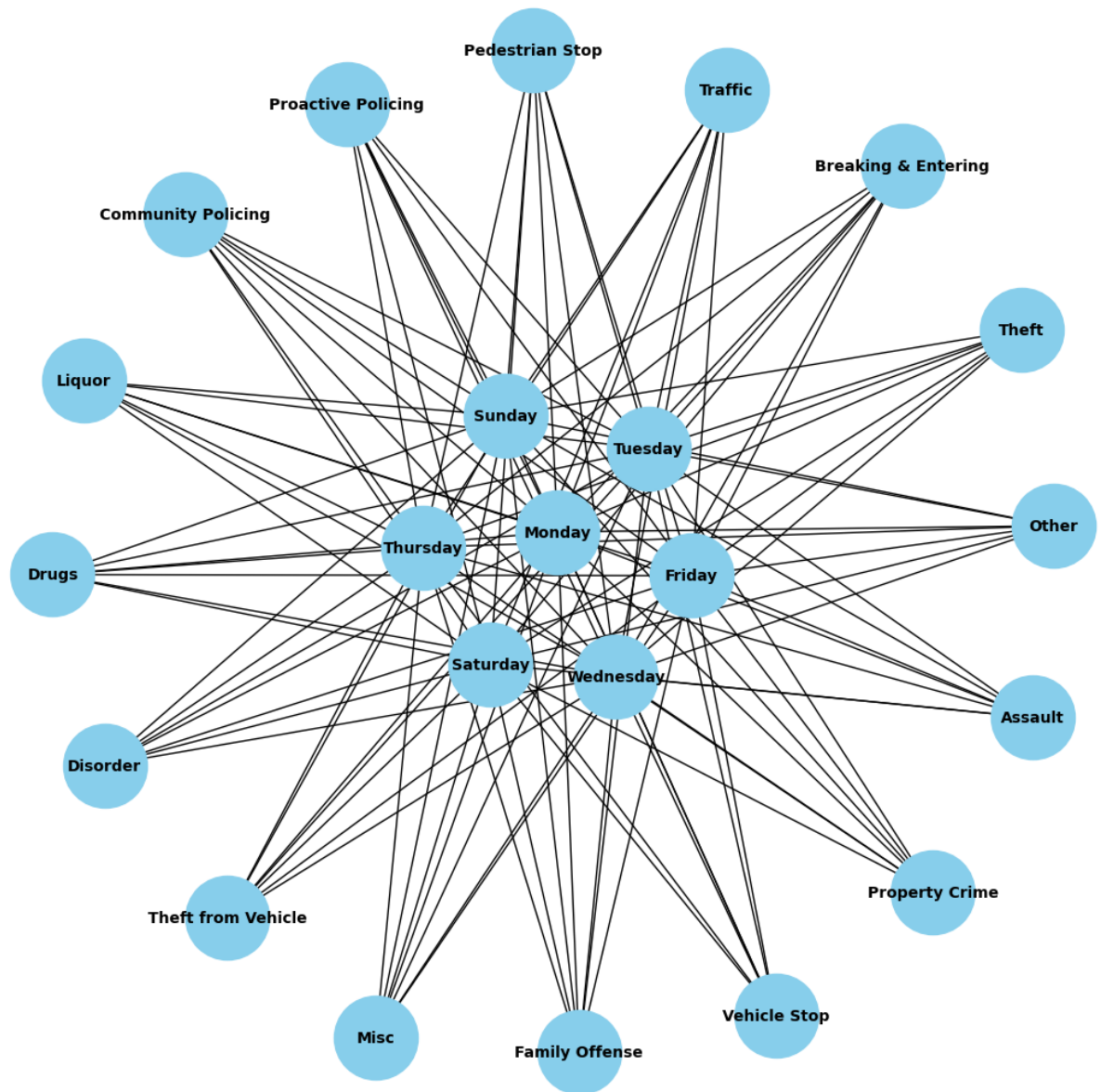
This graph is to visualize any relation between the type of crime and the day of the week.

```
In [180]: import networkx as nx
```

```
In [181]: # Build your graph
G = nx.from_pandas_edgelist(X_test, 'day_of_week', 'Crime_Category')

# Set the figure size
plt.figure(figsize=(10, 10))

# Plot it
nx.draw(G, with_labels=True, font_size=10, node_size=3000, node_color='slateblue')
plt.show()
```



Final Result Plan

The goal is to combine the above graphs to create a filterable dashboard to be able to filter and check to see if there are trends with the data that answer the following questions:

1) Are there crime trends in different cities? 2) Are there crime trends with different years? 3) What different types of crime occurs, and which one is most prevalent per city? 4) Does the day in the week effect the amount of crime and what type of crime occurs? 5) Does the time of day impact where a crime occurs and what type?

Type *Markdown* and LaTeX: α^2

In []:

