

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import folium
from folium.plugins import HeatMap
```

```
In [2]: data = pd.read_csv(r"C:\Users\Praveen T\Downloads\US_Accidents_March23.csv\US_Acciden
data
```

Out[2]:

	ID	Source	Severity	Start_Time	End_Time	Start_Lat	Start_Lng	End_Lat	End_Lng
0	A-1	Source2	3	2016-02-08 05:46:00	2016-02-08 11:00:00	39.865147	-84.058723	NaN	NaN
1	A-2	Source2	2	2016-02-08 06:07:59	2016-02-08 06:37:59	39.928059	-82.831184	NaN	NaN
2	A-3	Source2	2	2016-02-08 06:49:27	2016-02-08 07:19:27	39.063148	-84.032608	NaN	NaN
3	A-4	Source2	3	2016-02-08 07:23:34	2016-02-08 07:53:34	39.747753	-84.205582	NaN	NaN
4	A-5	Source2	2	2016-02-08 07:39:07	2016-02-08 08:09:07	39.627781	-84.188354	NaN	NaN
...
7728389	A-7777757	Source1	2	2019-08-23 18:03:25	2019-08-23 18:32:01	34.002480	-117.379360	33.99888	-117.379360
7728390	A-7777758	Source1	2	2019-08-23 19:11:30	2019-08-23 19:38:23	32.766960	-117.148060	32.76555	-117.148060
7728391	A-7777759	Source1	2	2019-08-23 19:00:21	2019-08-23 19:28:49	33.775450	-117.847790	33.77740	-117.847790
7728392	A-7777760	Source1	2	2019-08-23 19:00:21	2019-08-23 19:29:42	33.992460	-118.403020	33.98311	-118.399000
7728393	A-7777761	Source1	2	2019-08-23 18:52:06	2019-08-23 19:21:31	34.133930	-117.230920	34.13736	-117.230920

7728394 rows × 46 columns

```
In [3]: data.head()
```

Out[3]:

	ID	Source	Severity	Start_Time	End_Time	Start_Lat	Start_Lng	End_Lat	End_Lng	Distance(m)
0	A-1	Source2	3	2016-02-08 05:46:00	2016-02-08 11:00:00	39.865147	-84.058723	NaN	NaN	0.0
1	A-2	Source2	2	2016-02-08 06:07:59	2016-02-08 06:37:59	39.928059	-82.831184	NaN	NaN	0.0
2	A-3	Source2	2	2016-02-08 06:49:27	2016-02-08 07:19:27	39.063148	-84.032608	NaN	NaN	0.0
3	A-4	Source2	3	2016-02-08 07:23:34	2016-02-08 07:53:34	39.747753	-84.205582	NaN	NaN	0.0
4	A-5	Source2	2	2016-02-08 07:39:07	2016-02-08 08:09:07	39.627781	-84.188354	NaN	NaN	0.0

5 rows × 46 columns

In [4]: `data.tail()`

Out[4]:

	ID	Source	Severity	Start_Time	End_Time	Start_Lat	Start_Lng	End_Lat	End_Lng	Distance(m)
7728389	A-7777757	Source1	2	2019-08-23 18:03:25	2019-08-23 18:32:01	34.00248	-117.37936	33.99888	-117.370	
7728390	A-7777758	Source1	2	2019-08-23 19:11:30	2019-08-23 19:38:23	32.76696	-117.14806	32.76555	-117.153	
7728391	A-7777759	Source1	2	2019-08-23 19:00:21	2019-08-23 19:28:49	33.77545	-117.84779	33.77740	-117.857	
7728392	A-7777760	Source1	2	2019-08-23 19:00:21	2019-08-23 19:29:42	33.99246	-118.40302	33.98311	-118.395	
7728393	A-7777761	Source1	2	2019-08-23 18:52:06	2019-08-23 19:21:31	34.13393	-117.23092	34.13736	-117.239	

5 rows × 46 columns

In [5]: `data.info()`

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7728394 entries, 0 to 7728393
Data columns (total 46 columns):
#   Column                                Dtype
---  -
0   ID                                    object
1   Source                               object
2   Severity                             int64
3   Start_Time                           object
4   End_Time                             object
5   Start_Lat                           float64
6   Start_Lng                           float64
7   End_Lat                             float64
8   End_Lng                             float64
9   Distance(mi)                         float64
10  Description                           object
11  Street                               object
12  City                                 object
13  County                              object
14  State                               object
15  Zipcode                             object
16  Country                             object
17  Timezone                            object
18  Airport_Code                         object
19  Weather_Timestamp                    object
20  Temperature(F)                       float64
21  Wind_Chill(F)                        float64
22  Humidity(%)                          float64
23  Pressure(in)                         float64
24  Visibility(mi)                       float64
25  Wind_Direction                       object
26  Wind_Speed(mph)                      float64
27  Precipitation(in)                    float64
28  Weather_Condition                     object
29  Amenity                              bool
30  Bump                                 bool
31  Crossing                             bool
32  Give_Way                             bool
33  Junction                             bool
34  No_Exit                             bool
35  Railway                             bool
36  Roundabout                          bool
37  Station                             bool
38  Stop                                 bool
39  Traffic_Calming                      bool
40  Traffic_Signal                       bool
41  Turning_Loop                         bool
42  Sunrise_Sunset                       object
43  Civil_Twilight                       object
44  Nautical_Twilight                    object
45  Astronomical_Twilight                 object
dtypes: bool(13), float64(12), int64(1), object(20)
memory usage: 2.0+ GB

```

```
In [6]: data.describe()
```

Out[6]:		Severity	Start_Lat	Start_Lng	End_Lat	End_Lng	Distance(mi)	Temp
	count	7.728394e+06	7.728394e+06	7.728394e+06	4.325632e+06	4.325632e+06	7.728394e+06	7.50
	mean	2.212384e+00	3.620119e+01	-9.470255e+01	3.626183e+01	-9.572557e+01	5.618423e-01	6.10
	std	4.875313e-01	5.076079e+00	1.739176e+01	5.272905e+00	1.810793e+01	1.776811e+00	1.90
	min	1.000000e+00	2.455480e+01	-1.246238e+02	2.456601e+01	-1.245457e+02	0.000000e+00	-8.90
	25%	2.000000e+00	3.339963e+01	-1.172194e+02	3.346207e+01	-1.177543e+02	0.000000e+00	4.90
	50%	2.000000e+00	3.582397e+01	-8.776662e+01	3.618349e+01	-8.802789e+01	3.000000e-02	6.40
	75%	2.000000e+00	4.008496e+01	-8.035368e+01	4.017892e+01	-8.024709e+01	4.640000e-01	7.60
	max	4.000000e+00	4.900220e+01	-6.711317e+01	4.907500e+01	-6.710924e+01	4.417500e+02	2.00

```
In [7]: # Convert Start_Time and End_Time to datetime
data['Start_Time'] = pd.to_datetime(data['Start_Time'])
data['End_Time'] = pd.to_datetime(data['End_Time'])

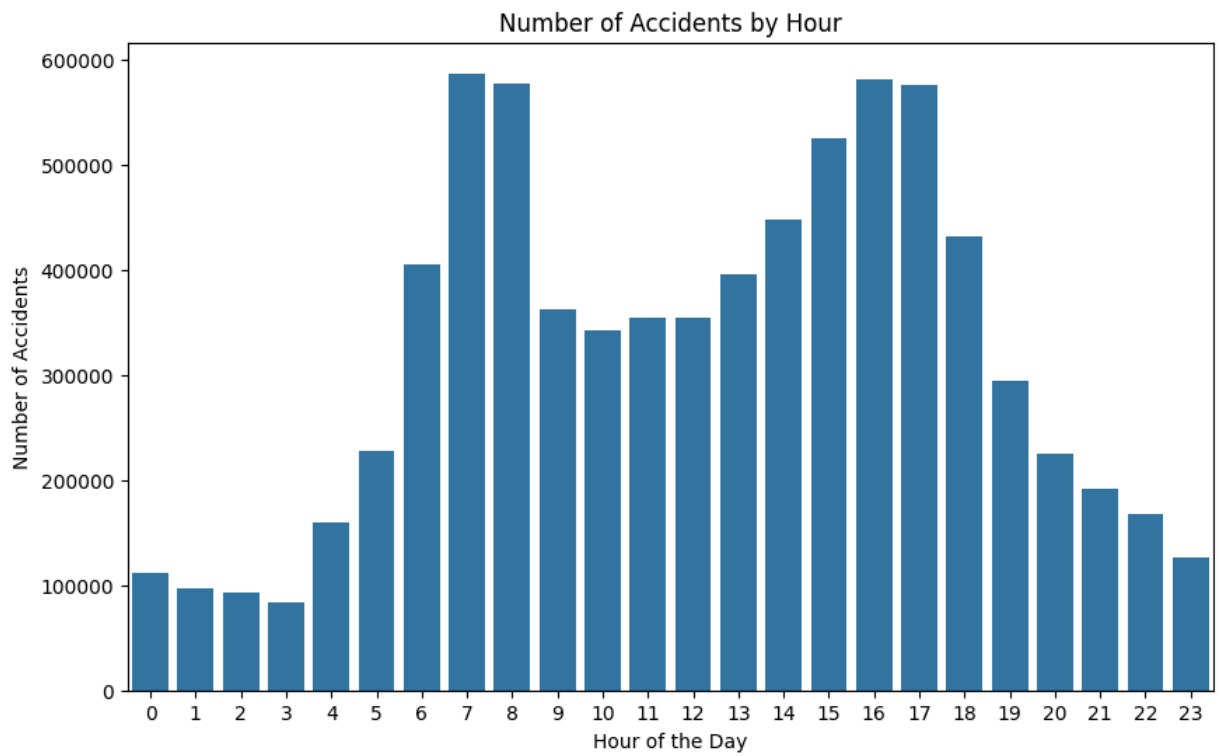
# Extract hour, day of week, and month
data['hour'] = data['Start_Time'].dt.hour
data['day_of_week'] = data['Start_Time'].dt.day_name()
data['month'] = data['Start_Time'].dt.month

data.head()
```

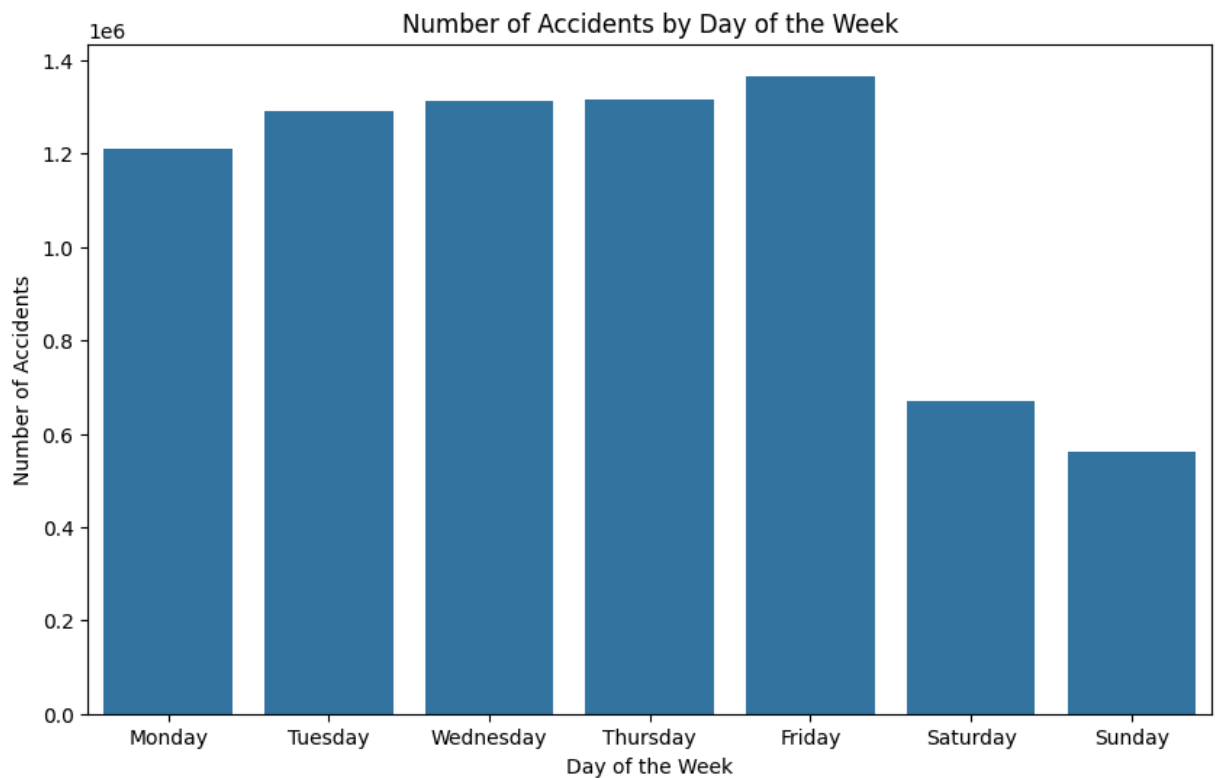
Out[7]:	ID	Source	Severity	Start_Time	End_Time	Start_Lat	Start_Lng	End_Lat	End_Lng	Distance(m)
0	A-1	Source2	3	2016-02-08 05:46:00	2016-02-08 11:00:00	39.865147	-84.058723	NaN	NaN	0.0
1	A-2	Source2	2	2016-02-08 06:07:59	2016-02-08 06:37:59	39.928059	-82.831184	NaN	NaN	0.0
2	A-3	Source2	2	2016-02-08 06:49:27	2016-02-08 07:19:27	39.063148	-84.032608	NaN	NaN	0.0
3	A-4	Source2	3	2016-02-08 07:23:34	2016-02-08 07:53:34	39.747753	-84.205582	NaN	NaN	0.0
4	A-5	Source2	2	2016-02-08 07:39:07	2016-02-08 08:09:07	39.627781	-84.188354	NaN	NaN	0.0

5 rows × 49 columns

```
In [8]: # Accidents by hour
plt.figure(figsize=(10, 6))
sns.countplot(data=data, x='hour')
plt.title('Number of Accidents by Hour')
plt.xlabel('Hour of the Day')
plt.ylabel('Number of Accidents')
plt.show()
```

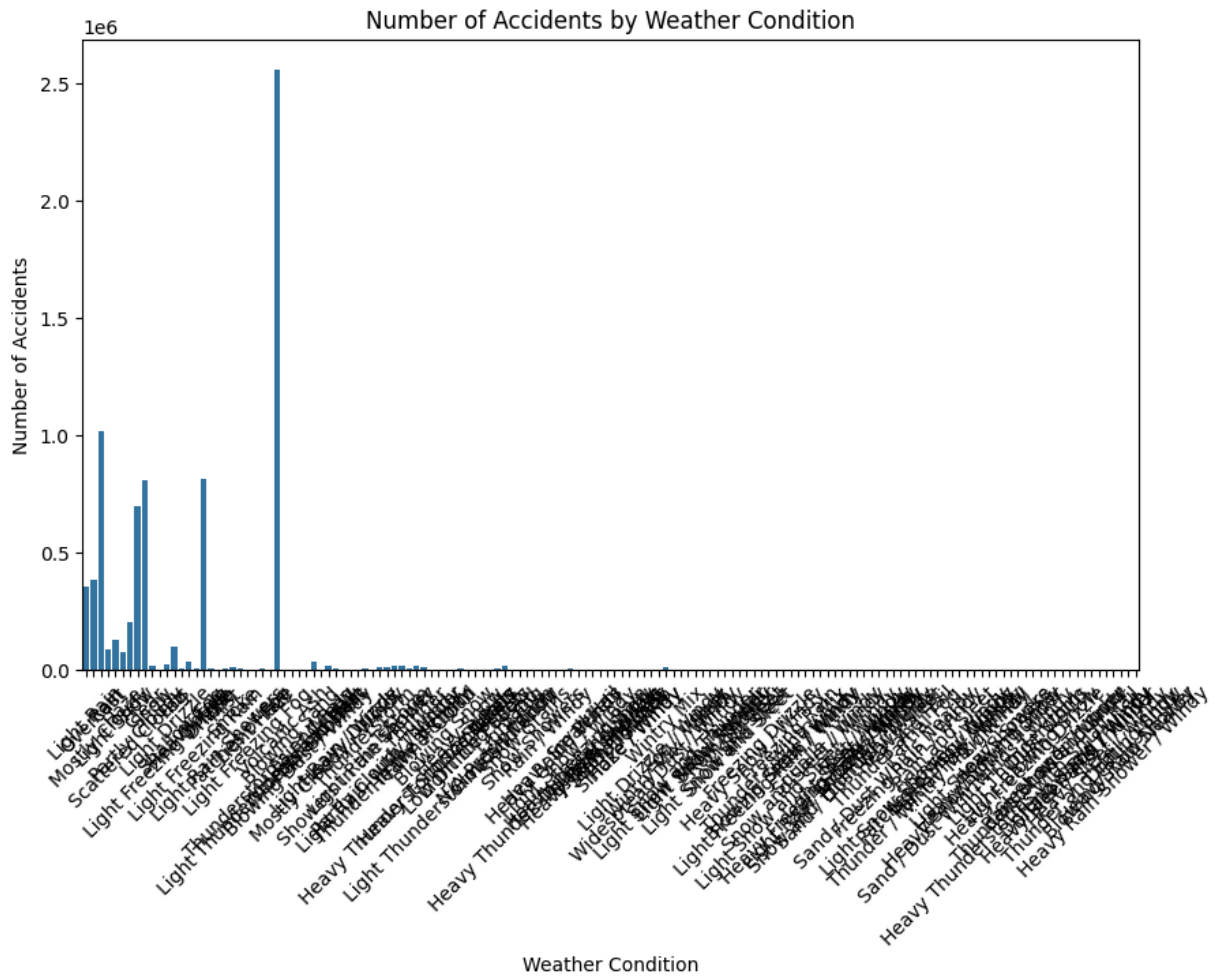


```
In [9]: # Accidents by day of the week
plt.figure(figsize=(10, 6))
sns.countplot(data=data, x='day_of_week', order=['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Sunday'])
plt.title('Number of Accidents by Day of the Week')
plt.xlabel('Day of the Week')
plt.ylabel('Number of Accidents')
plt.show()
```

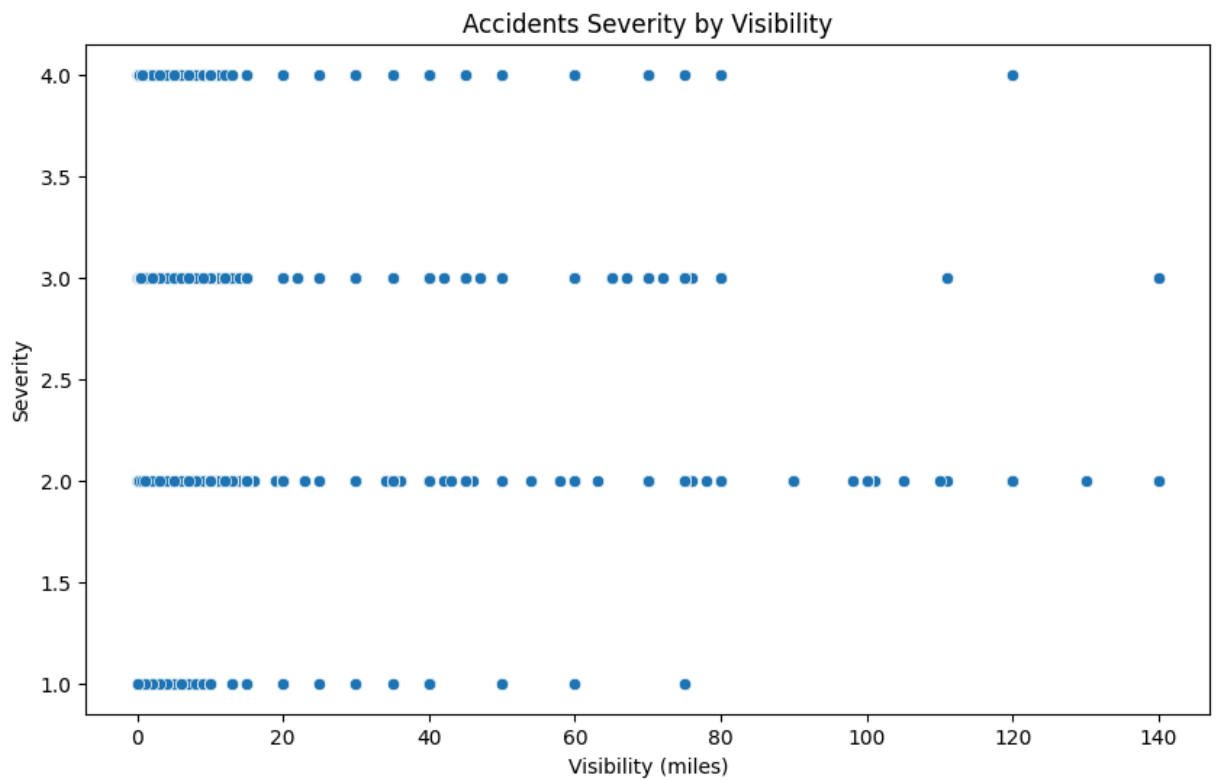


```
In [10]: # Accidents by weather condition
plt.figure(figsize=(10, 6))
sns.countplot(data=data, x='Weather_Condition')
plt.title('Number of Accidents by Weather Condition')
plt.xlabel('Weather Condition')
plt.ylabel('Number of Accidents')
```

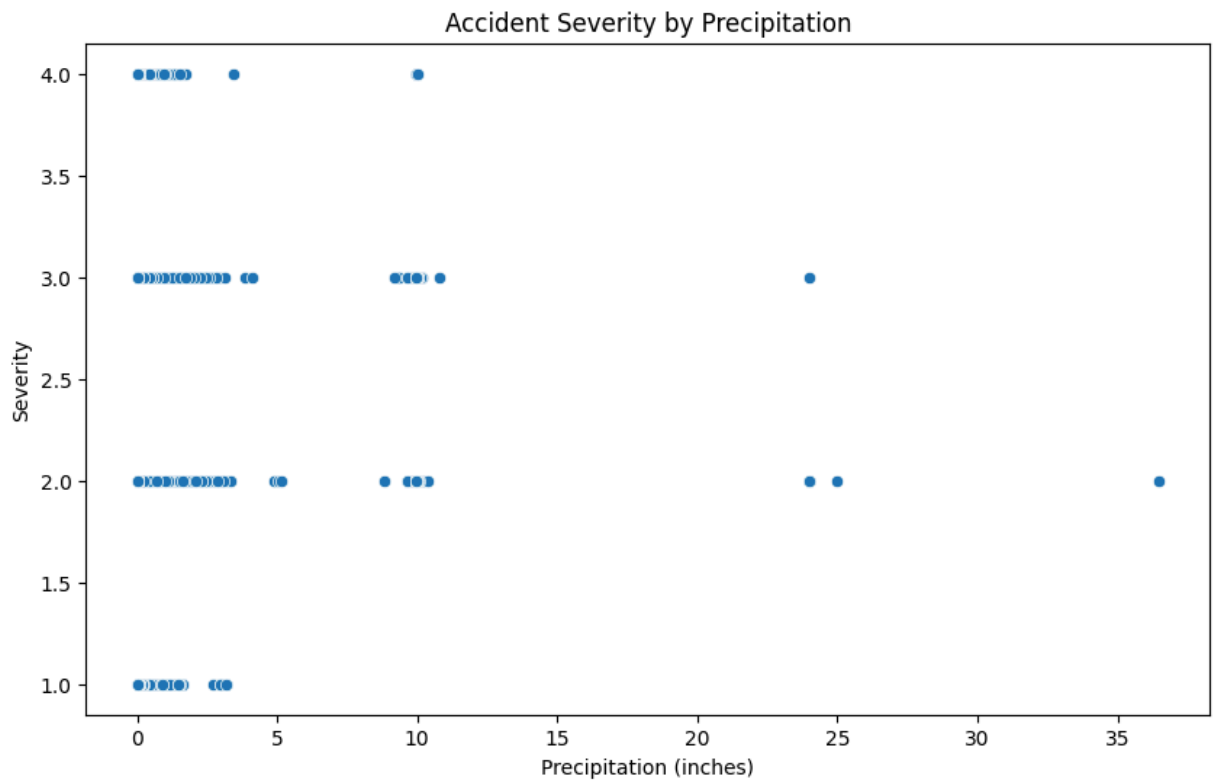
```
plt.xticks(rotation=45)
plt.show()
```



```
In [11]: # Accidents by visibility
plt.figure(figsize=(10, 6))
sns.scatterplot(data=data, x='Visibility(mi)', y='Severity')
plt.title('Accidents Severity by Visibility')
plt.xlabel('Visibility (miles)')
plt.ylabel('Severity')
plt.show()
```

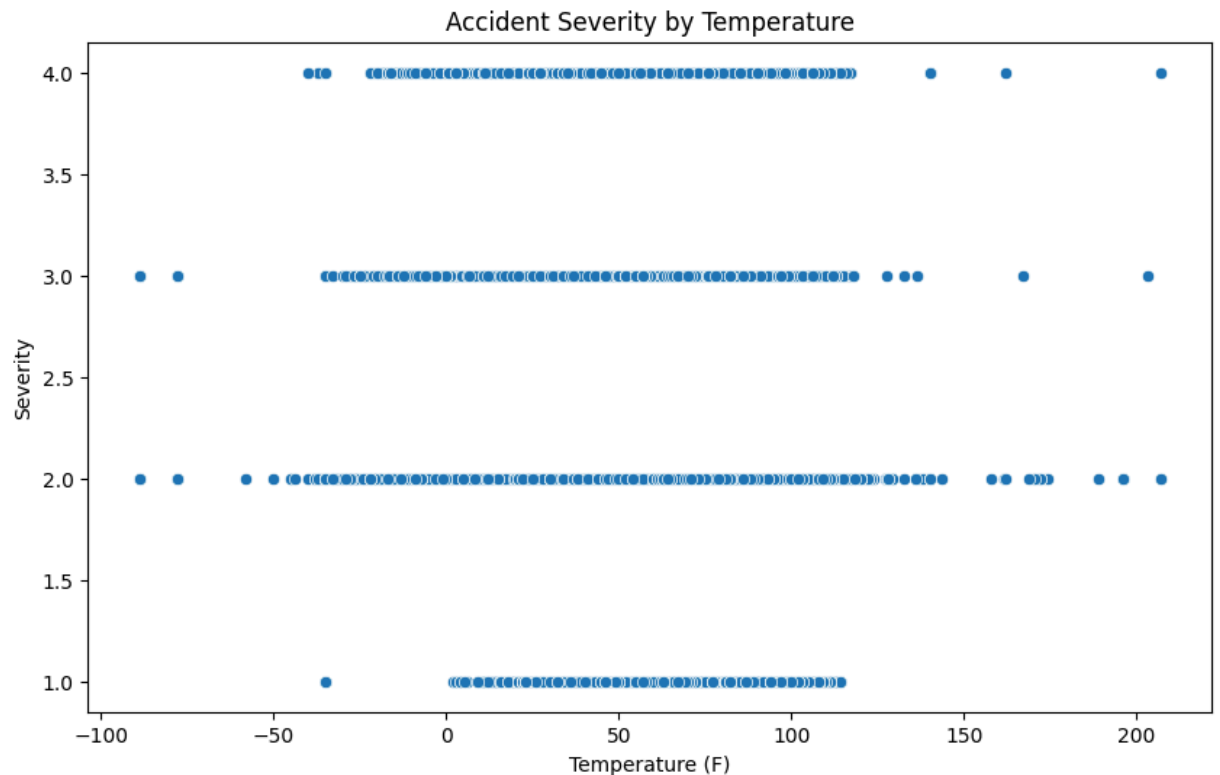


```
In [12]: # Accidents by precipitation
plt.figure(figsize=(10, 6))
sns.scatterplot(data=data, x='Precipitation(in)', y='Severity')
plt.title('Accident Severity by Precipitation')
plt.xlabel('Precipitation (inches)')
plt.ylabel('Severity')
plt.show()
```



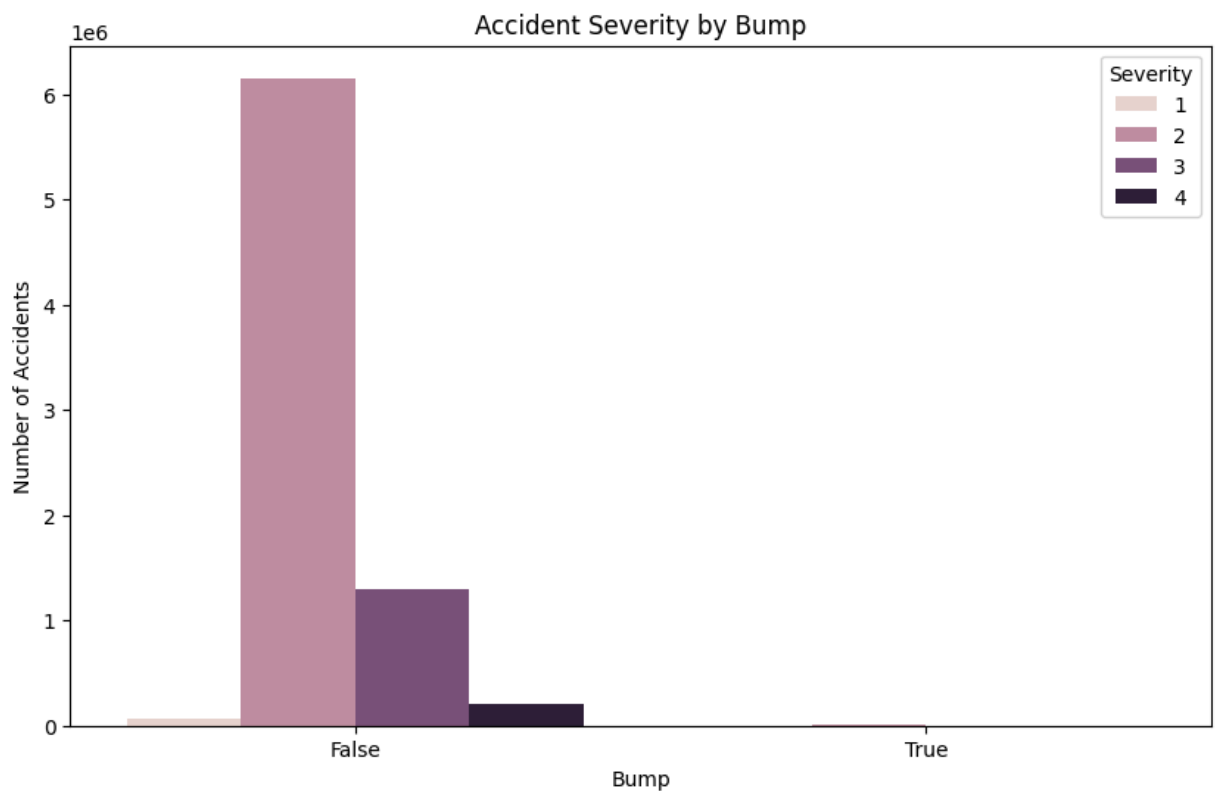
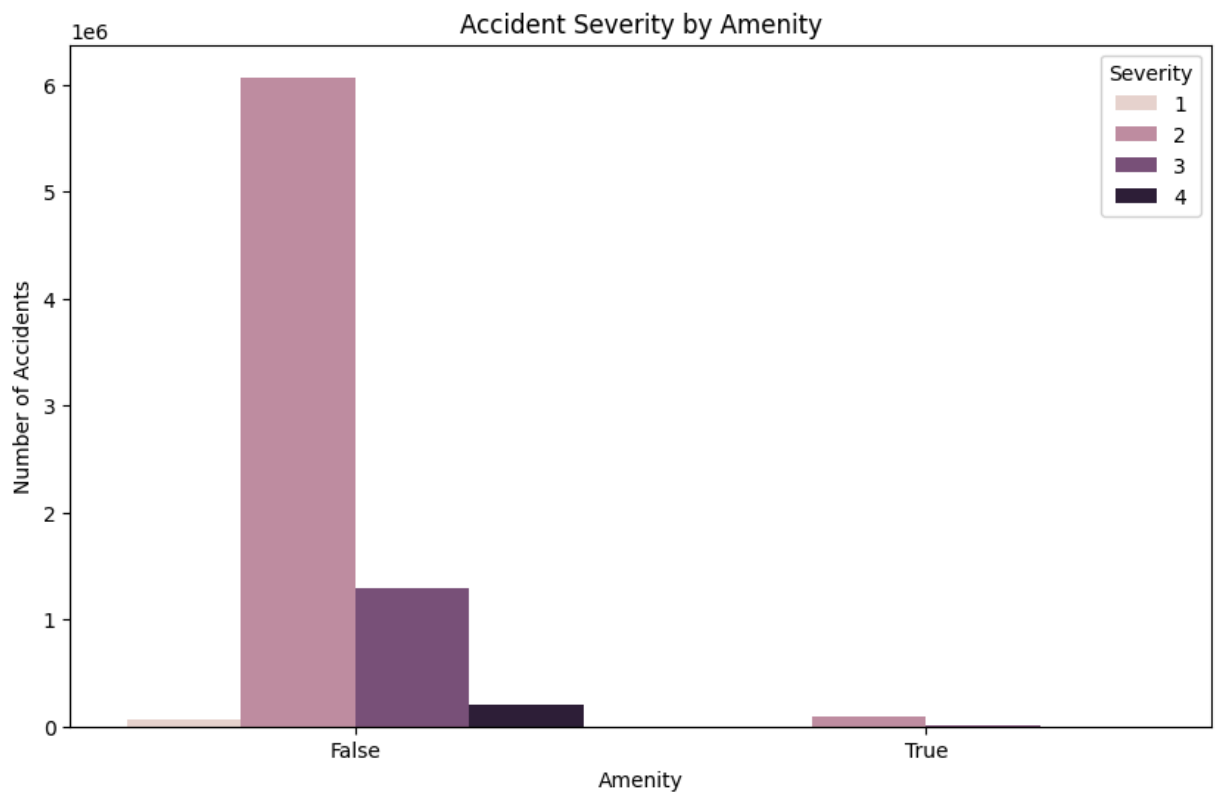
```
In [14]: # Accidents by temperature
plt.figure(figsize=(10, 6))
sns.scatterplot(data=data, x='Temperature(F)', y='Severity')
plt.title('Accident Severity by Temperature')
plt.xlabel('Temperature (F)')
```

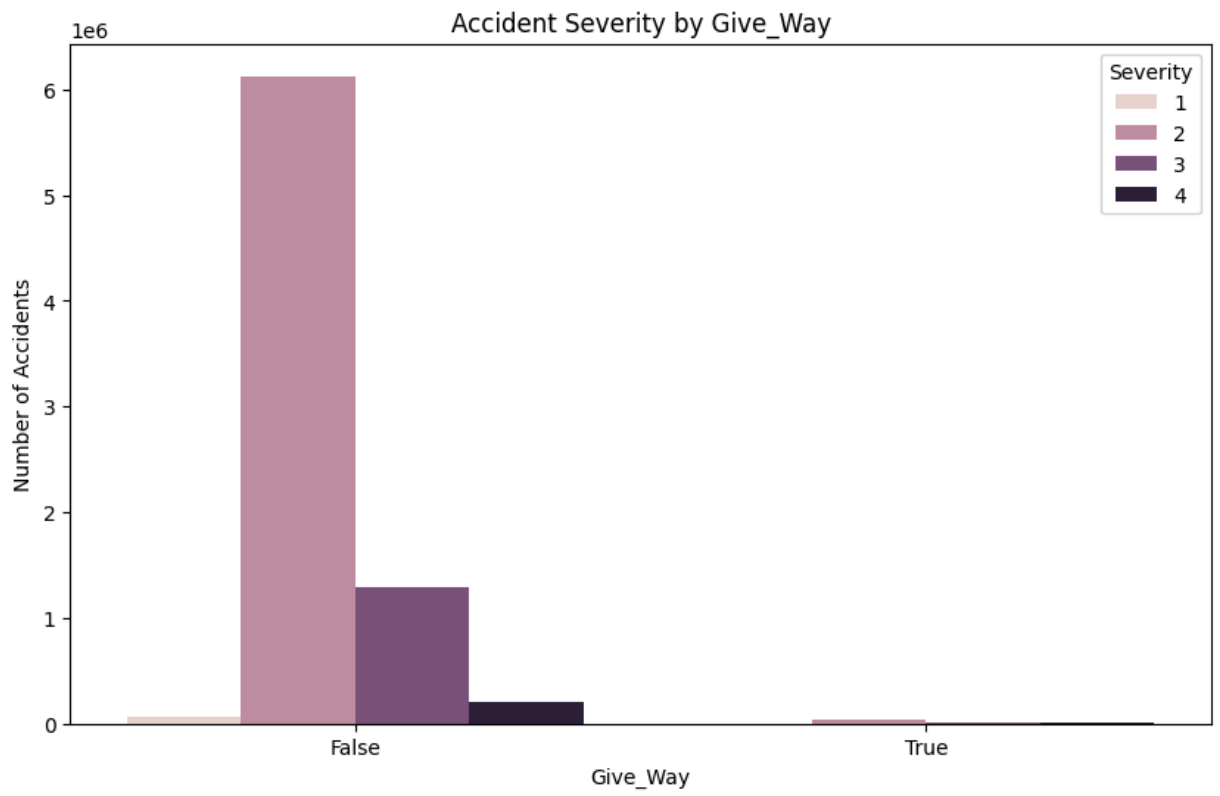
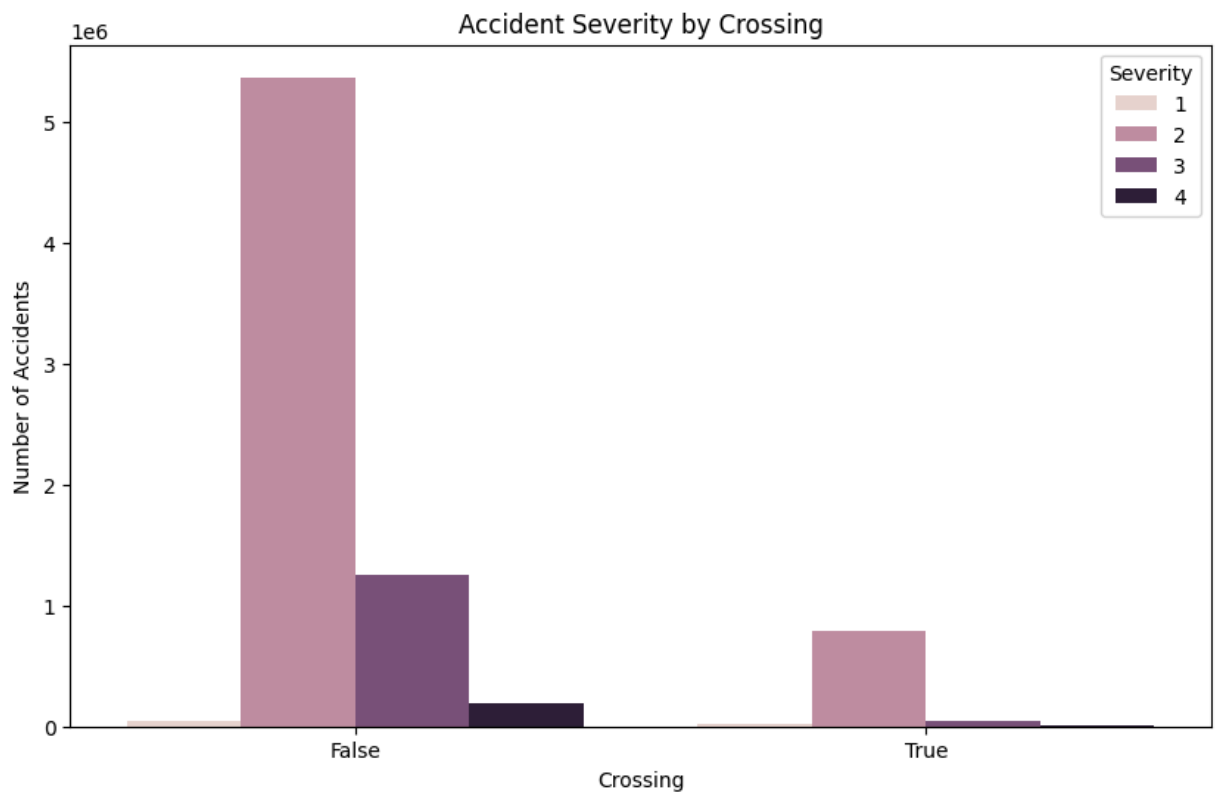
```
plt.ylabel('Severity')
plt.show()
```

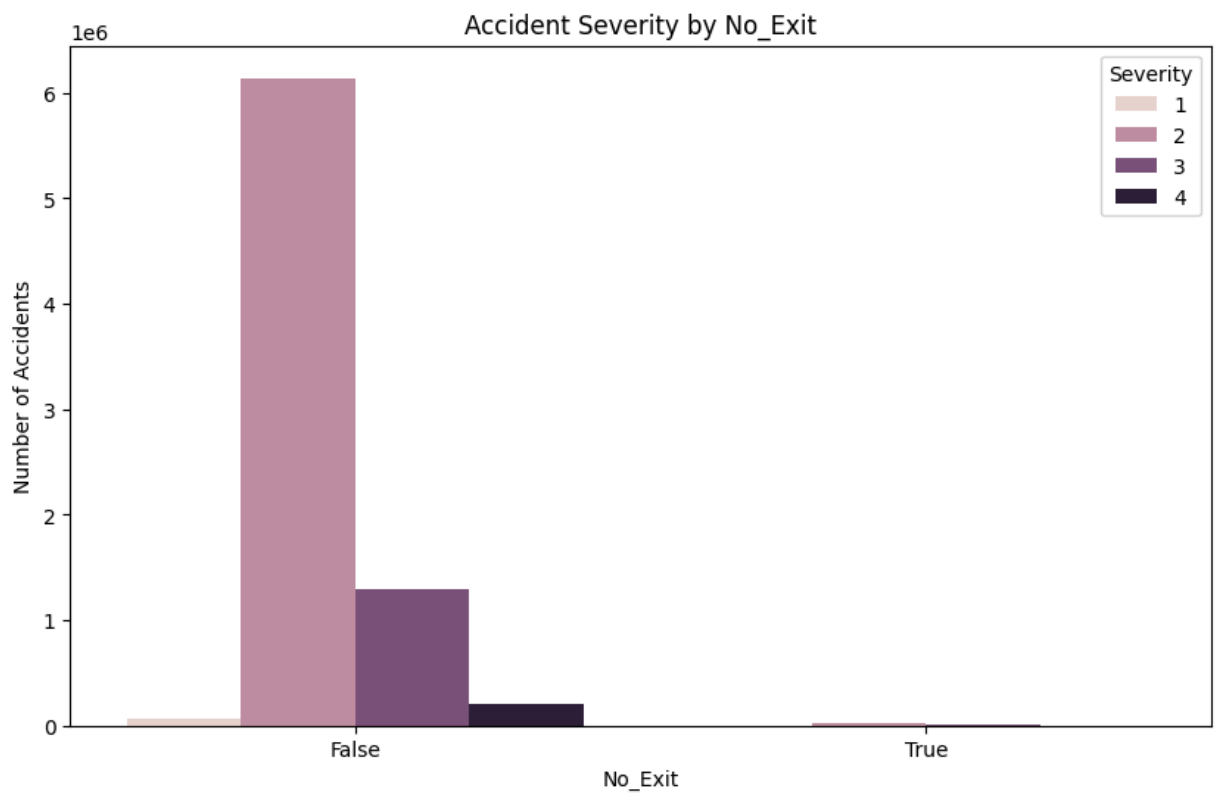
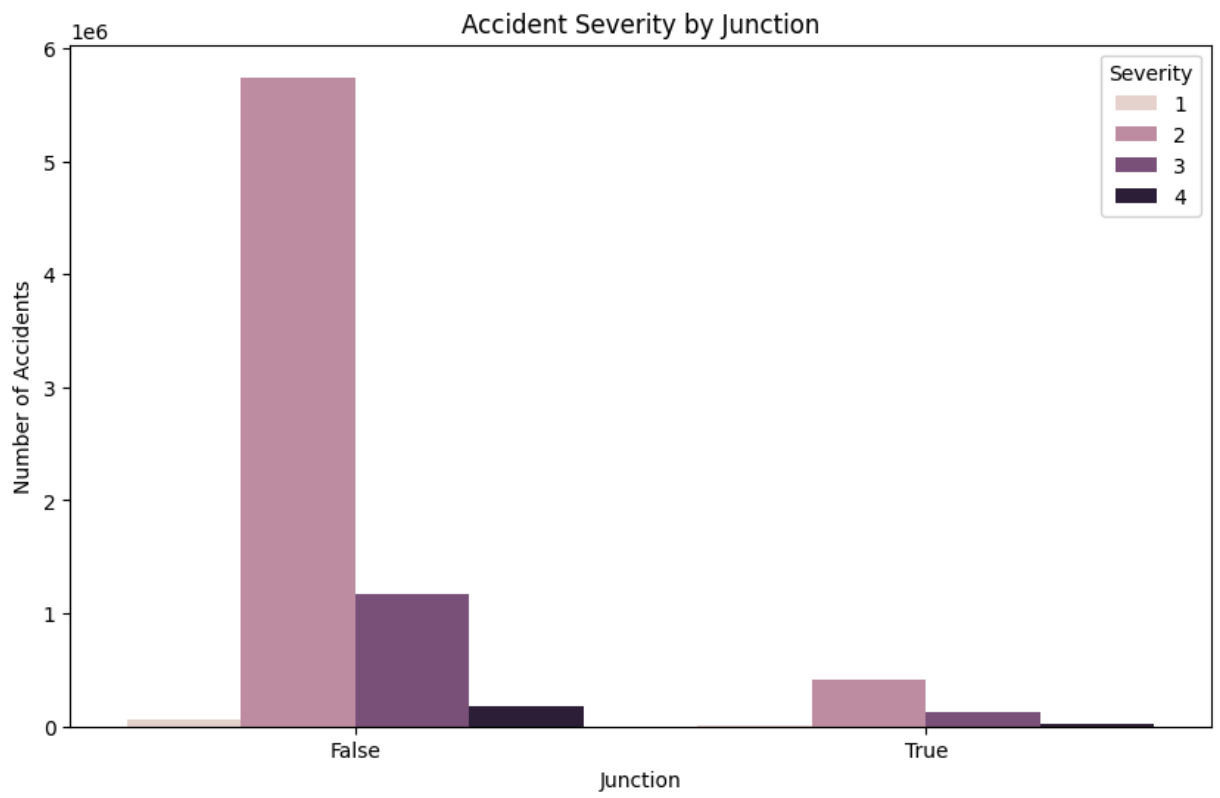


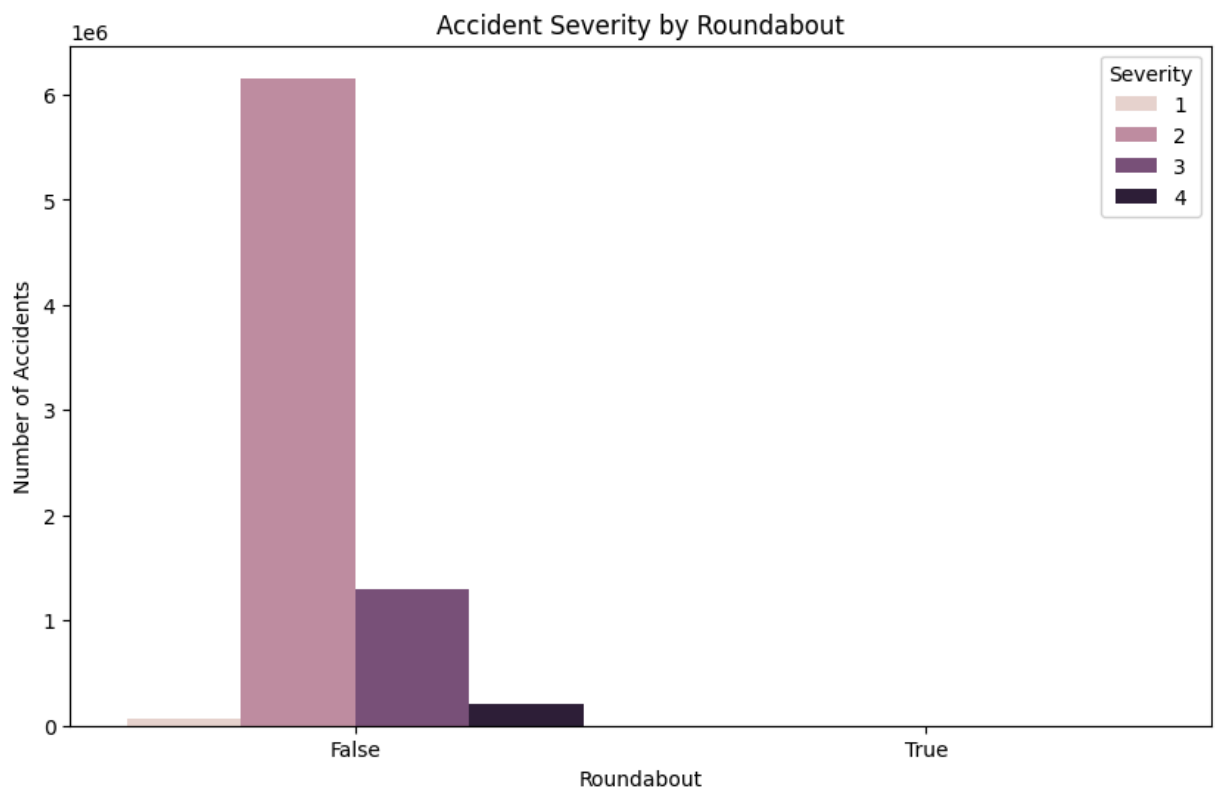
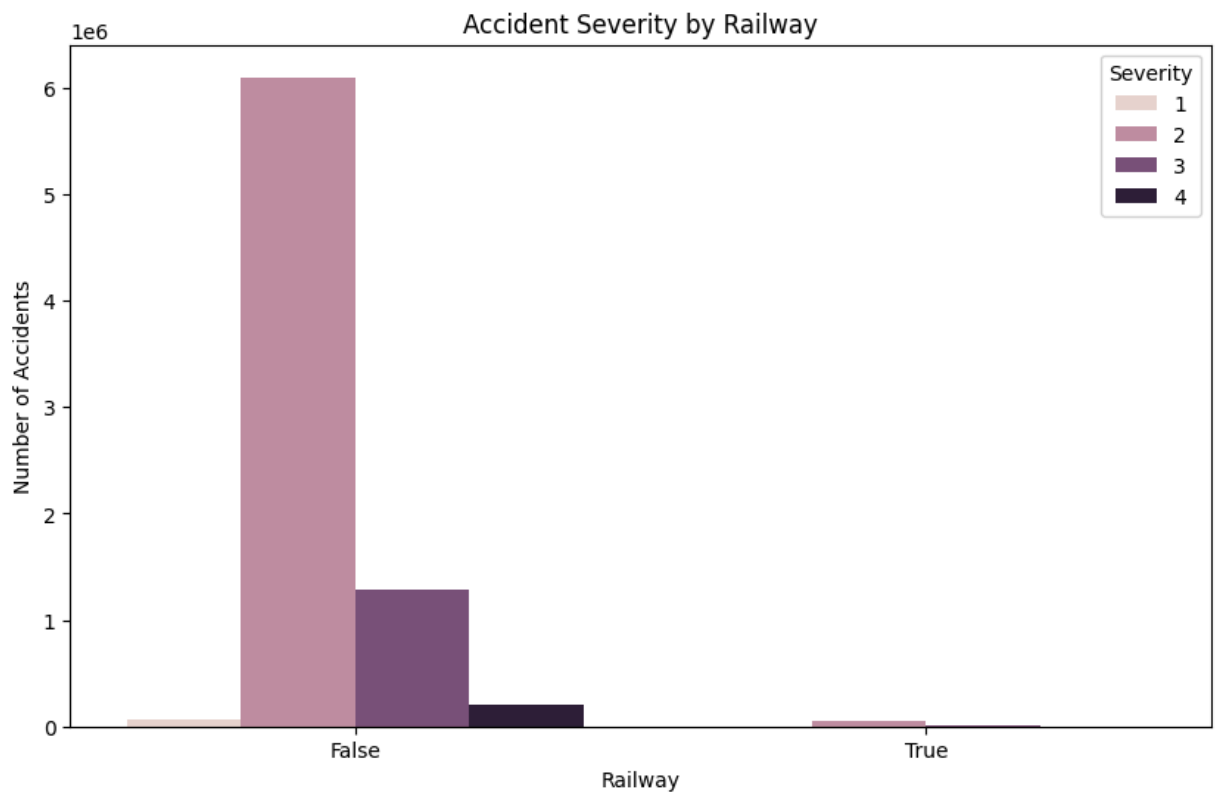
```
In [13]: # List of road condition features
road_conditions = ['Amenity', 'Bump', 'Crossing', 'Give_Way', 'Junction', 'No_Exit',
                  'Railway', 'Roundabout', 'Station', 'Stop', 'Traffic_Calming',
                  'Traffic_Signal', 'Turning_Loop']

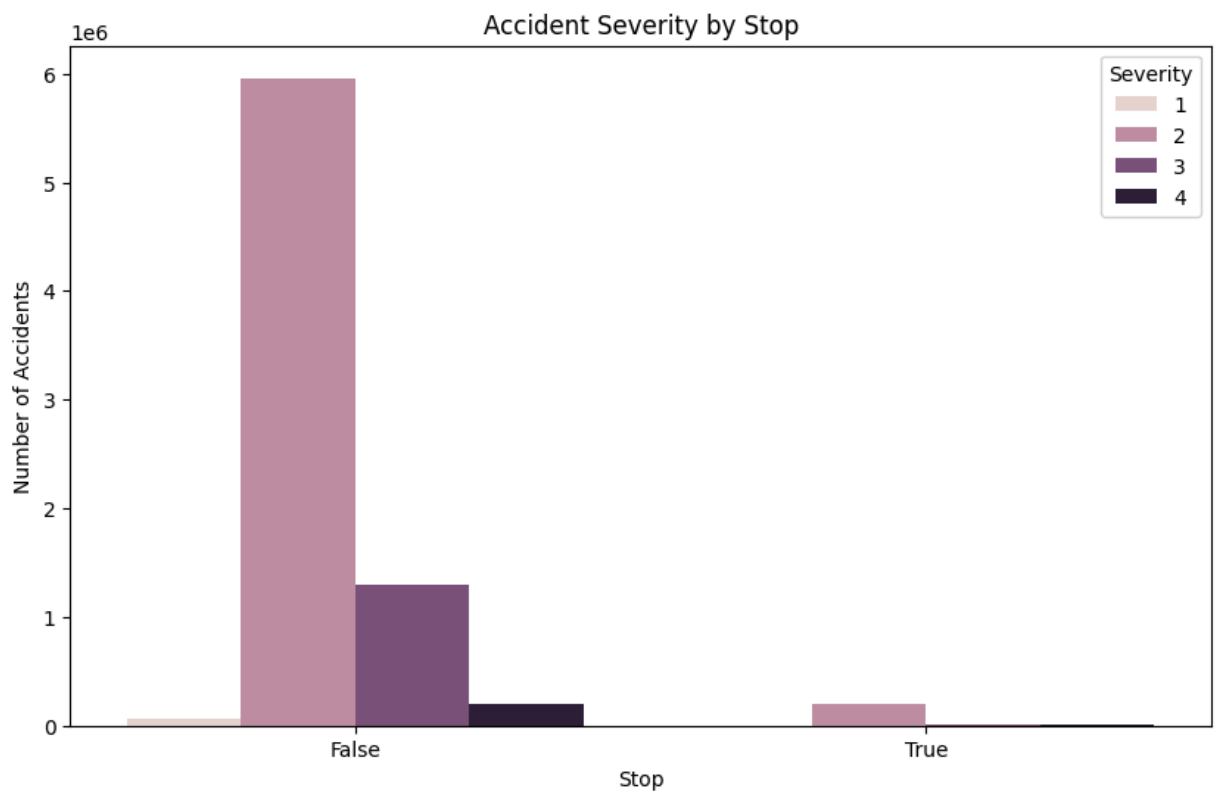
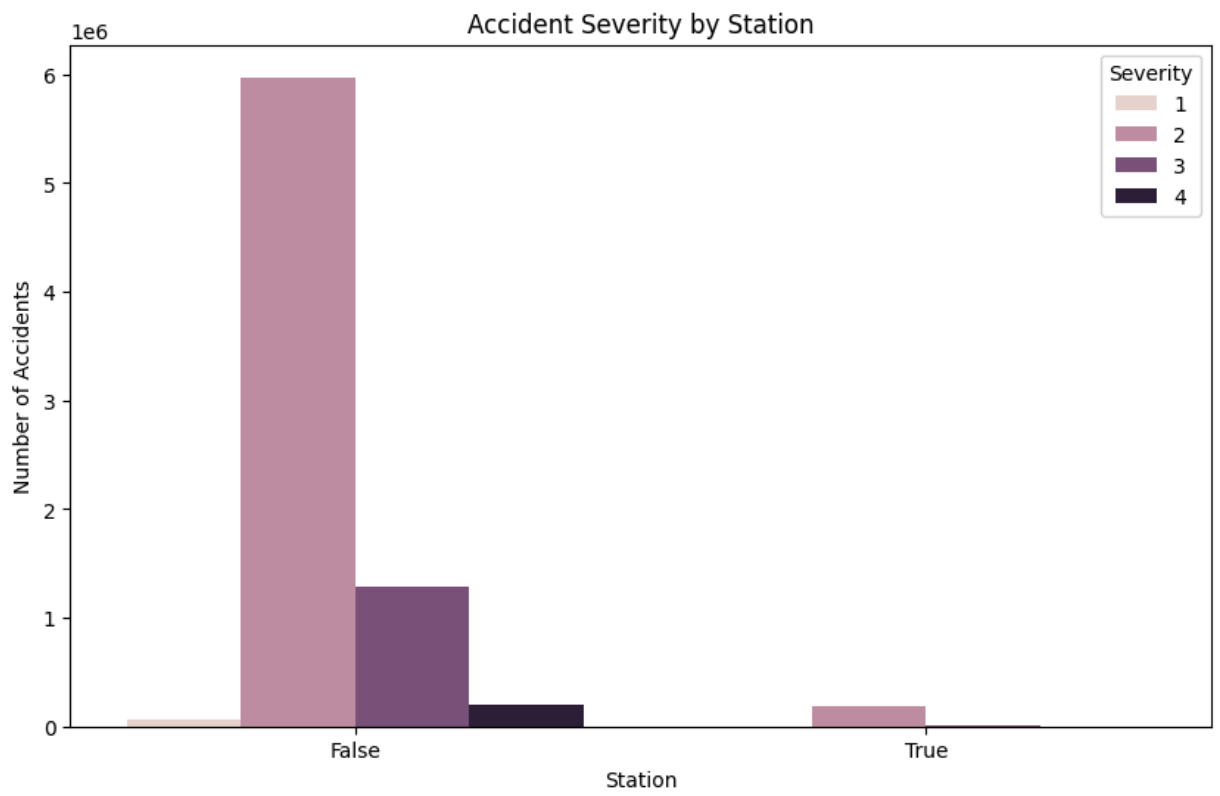
# Plotting accident severity by road conditions
for condition in road_conditions:
    plt.figure(figsize=(10, 6))
    sns.countplot(data=data, x=condition, hue='Severity')
    plt.title(f'Accident Severity by {condition}')
    plt.xlabel(condition)
    plt.ylabel('Number of Accidents')
    plt.legend(title='Severity')
    plt.show()
```

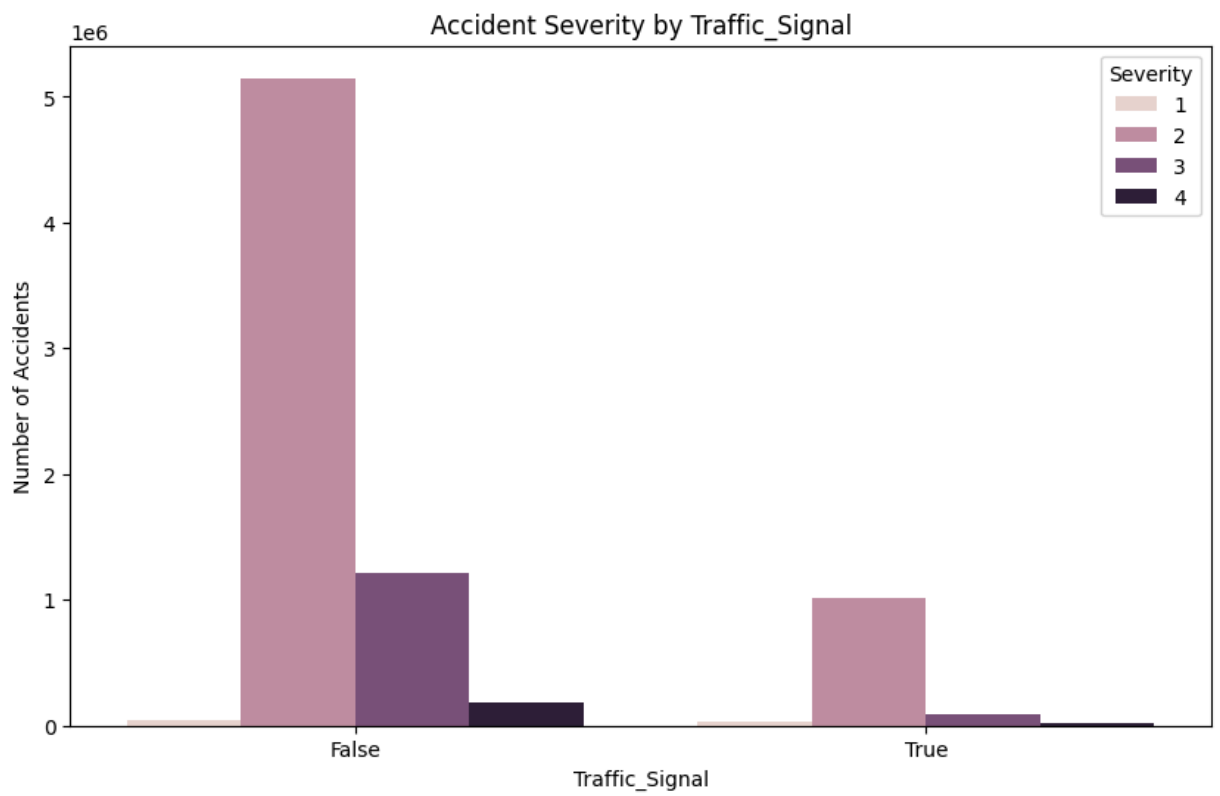
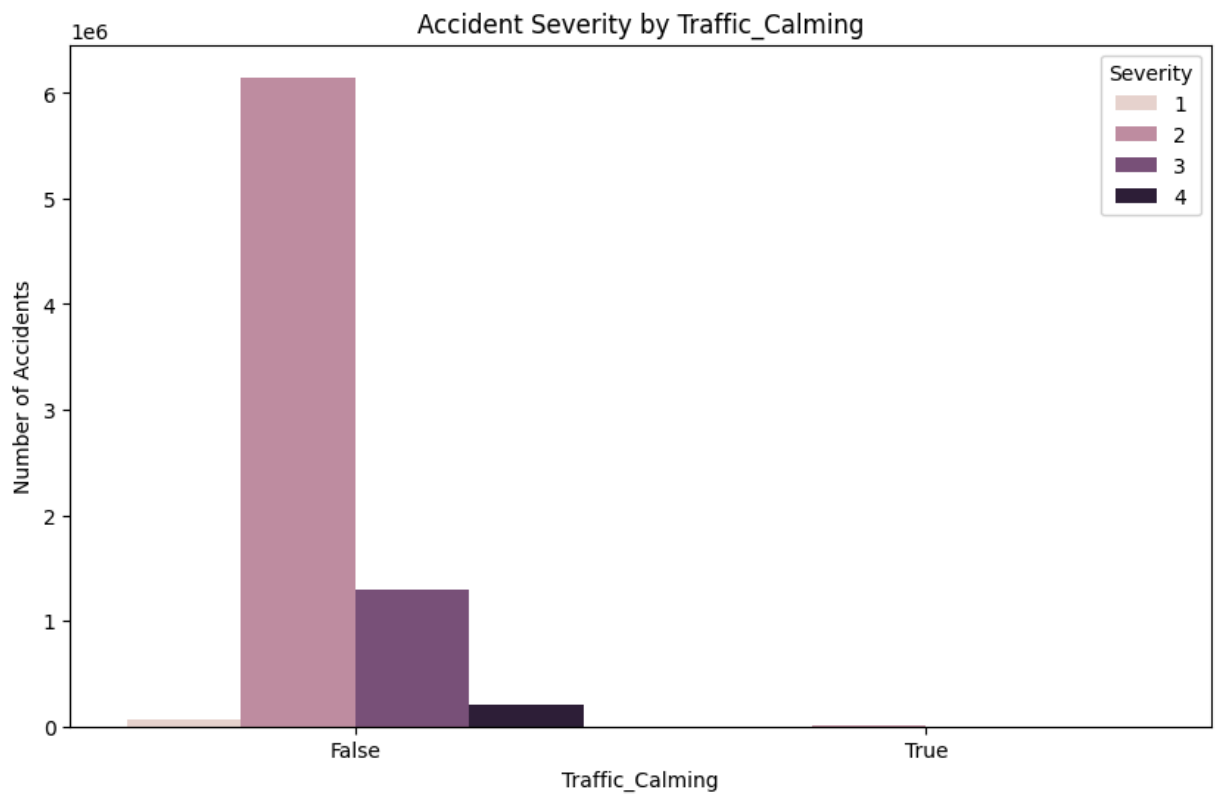



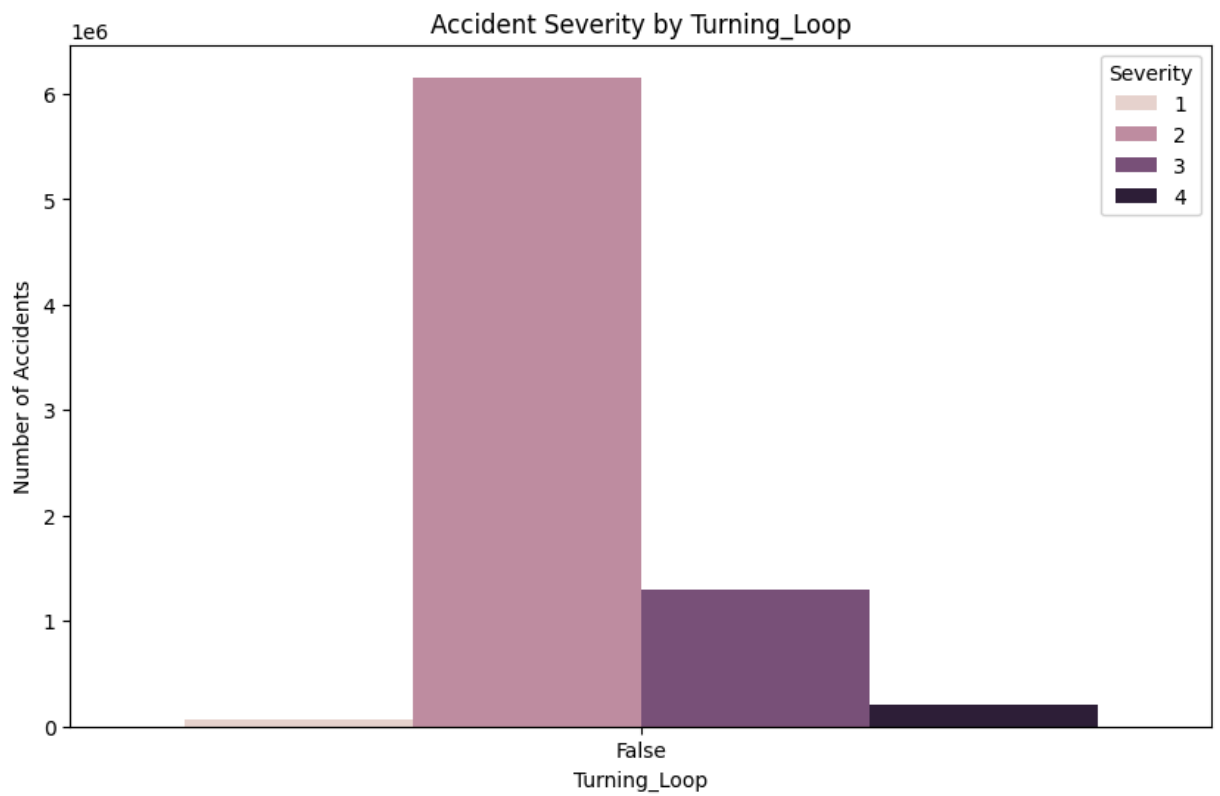












```
In [15]: # Create a base map
m = folium.Map(location=[data['Start_Lat'].mean(), data['Start_Lng'].mean()], zoom_start=13)

# Add accident data to the map
heat_data = [[row['Start_Lat'], row['Start_Lng']] for index, row in data.iterrows()]
HeatMap(heat_data).add_to(m)

# Save the map as an HTML file
m.save('accident_hotspots.html')
```

```
In [ ]:
```