US Crime Trends: Uncovering Patterns and Insights through Exploratory Data Analysis

Reading Data

In [2]:
```
import pandas as pd
```

In [3]:
```
df = pd.read_csv('Downloads/Crime_Data_from_2020_to_Present.csv')
```

In [4]:
```
df.head()
```

Out[4]:

	DR_NO	Date Rptd	DATE OCC	TIME OCC	AREA	AREA NAME	Rpt Dist No	Part 1-2	Crm Cd	Crm
0	190326475	03/01/2020 12:00:00 AM	03/01/2020 12:00:00 AM	2130	7	Wilshire	784	1	510	V
1	200106753	02/09/2020 12:00:00 AM	02/08/2020 12:00:00 AM	1800	1	Central	182	1	330	BU
2	200320258	11/11/2020 12:00:00 AM	11/04/2020 12:00:00 AM	1700	3	Southwest	356	1	480	BIKE -
3	200907217	05/10/2023 12:00:00 AM	03/10/2020 12:00:00 AM	2037	9	Van Nuys	964	1	343	SHOPI GRAN ($9
4	220614831	08/18/2022 12:00:00 AM	08/17/2020 12:00:00 AM	1200	6	Hollywood	666	2	354	T II

5 rows × 28 columns

In [38]:
```
df.columns
```

Out[38]:
```
Index(['DR_NO', 'Date Rptd', 'DATE OCC', 'TIME OCC', 'AREA', 'AREA NAME',
       'Rpt Dist No', 'Part 1-2', 'Crm Cd', 'Crm Cd Desc', 'Mocodes',
       'Vict Age', 'Vict Sex', 'Vict Descent', 'Premis Cd', 'Premis Desc',
       'Weapon Used Cd', 'Weapon Desc', 'Status', 'Status Desc', 'Crm Cd 1',
       'Crm Cd 2', 'Crm Cd 3', 'Crm Cd 4', 'LOCATION', 'Cross Street', 'LAT',
       'LON', 'Race', 'weekday', 'month', 'year'],
      dtype='object')
```

Description of all the columns:

- DR_NO: Report number for the crime incident.
- Date Rptd: Date the crime was reported.
- DATE OCC: Date the crime occurred.
- TIME OCC: Time the crime occurred.
- AREA: Area code where the crime occurred.
- AREA NAME: Name of the area where the crime occurred.
- Rpt Dist No: Reporting district number.
- Part 1-2: Indicates if the crime is Part 1 or Part 2 offense.
- Crm Cd: Crime code.
- Crm Cd Desc: Description of the crime.
- Mocodes: Modus Operandi (Method of operation) codes.
- Vict Age: Age of the victim.
- Vict Sex: Sex of the victim.
- Vict Descent: Descent of the victim.
- Premis Cd: Premises code where the crime occurred.
- Premis Desc: Description of the premises.
- Weapon Used Cd: Code indicating if a weapon was used.
- Weapon Desc: Description of the weapon used.
- Status: Status of the crime report.
- Status Desc: Description of the status.
- Crm Cd 1, Crm Cd 2, Crm Cd 3, Crm Cd 4: Additional crime codes.
- LOCATION: Location where the crime occurred.
- Cross Street: Cross street of the location.
- LAT: Latitude coordinate of the location.
- LON: Longitude cooes rdinate of the location.

Null Values

In [6]:
```
df.isnull().sum().sort_values(ascending=False)
```

Out[6]:
```
Crm Cd 4          883925
Crm Cd 3          881811
Crm Cd 2          819490
Cross Street      744208
Weapon Desc       577383
Weapon Used Cd    577383
Race              132154
Mocodes           123013
Vict Descent      117068
Vict Sex          117060
Premis Desc          540
Crm Cd 1              11
Premis Cd             10
Vict Age               0
Date Rptd              0
Crm Cd Desc            0
Crm Cd                 0
Part 1-2               0
Status                 0
Status Desc            0
Rpt Dist No            0
AREA NAME              0
AREA                   0
TIME OCC               0
LOCATION               0
DATE OCC               0
LAT                    0
LON                    0
DR_NO                  0
dtype: int64
```

Weekly Trends: Analysis of Crime Incidents on a Weekly Basis

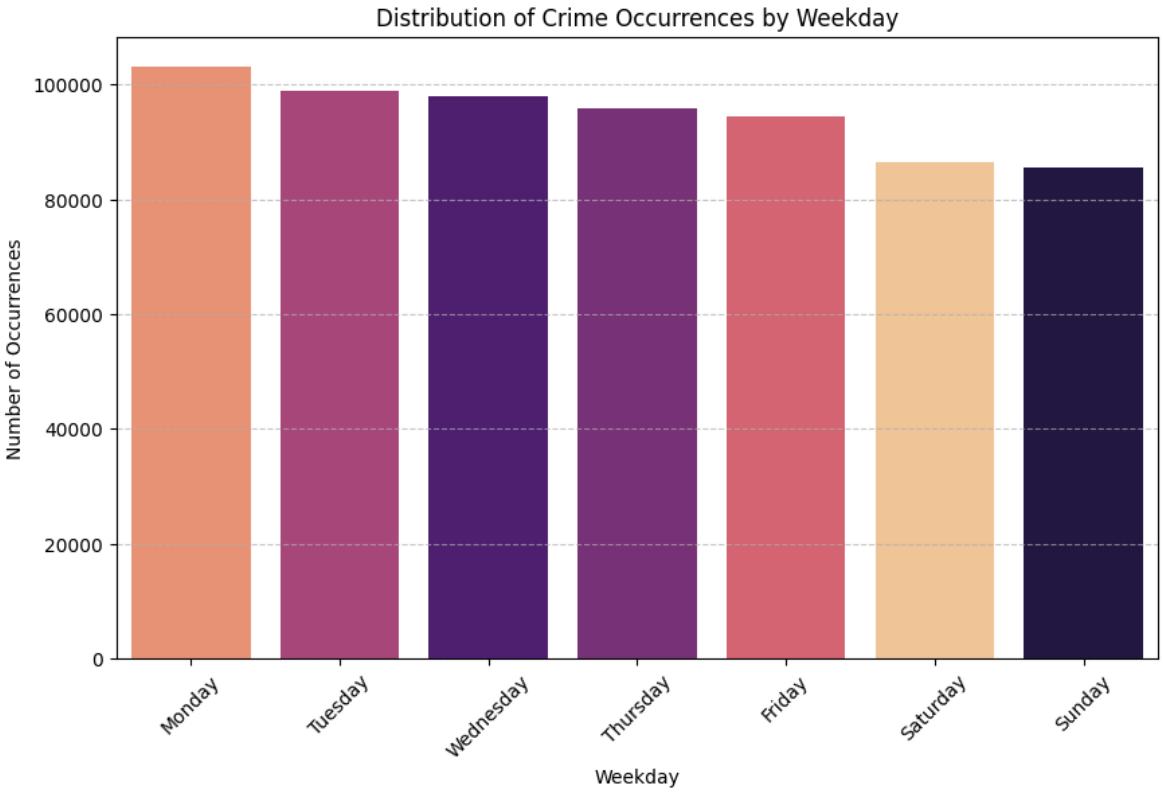
In [51]:
```
# Convert 'Date Rptd' to datetime format
df['Date Rptd'] = pd.to_datetime(df['Date Rptd'])

# Extract weekday from 'Date Rptd'
df['weekday'] = df['Date Rptd'].dt.weekday

# Map the numeric weekday to the corresponding weekday name
weekday_names = {0: 'Monday', 1: 'Tuesday', 2: 'Wednesday', 3: 'Thursday', 4: 'F
df['weekday'] = df['weekday'].map(weekday_names)
df_weekday = df['weekday'].value_counts().sort_values(ascending=False)
df_weekday
```

Out[51]:
```
weekday
Monday       138918
Tuesday      132630
Wednesday    131340
Thursday     128466
Friday       126346
Saturday     114031
Sunday       112173
Name: count, dtype: int64
```

🟡 Most crimes are reported on Mondays, with Sundays having the lowest reported crime rate.

In [167…

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

# Assuming df is your DataFrame containing the dataset

# Convert 'Date Rptd' to datetime format
df['Date Rptd'] = pd.to_datetime(df['Date Rptd'])

# Extract weekday from 'Date Rptd'
df['weekday'] = df['Date Rptd'].dt.strftime('%A')

# Plot the distribution of crime occurrences by weekday using Seaborn's countplo
plt.figure(figsize=(10, 6))
sns.countplot(data=df, x='weekday', hue='weekday', order=df['weekday'].value_cou
plt.title('Distribution of Crime Occurrences by Weekday')
plt.xlabel('Weekday')
plt.ylabel('Number of Occurrences')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```



Percentage wise distribution

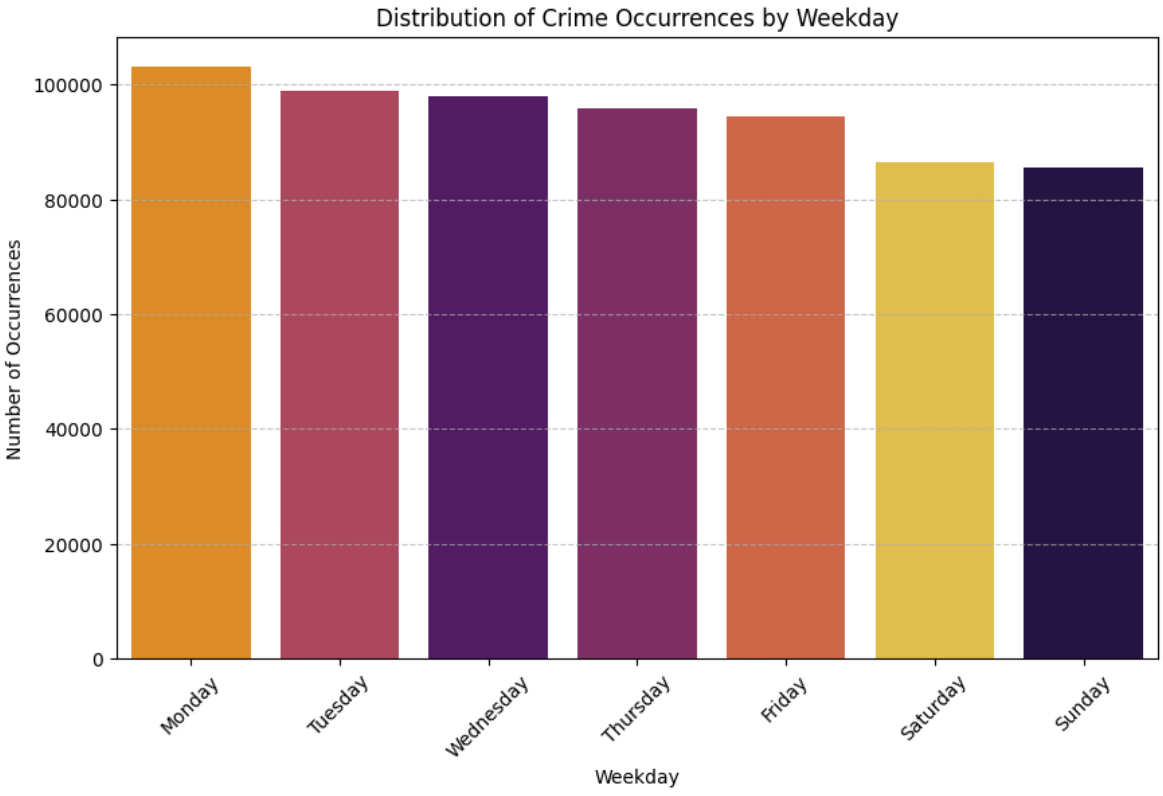
In [9]:

```
total_crimes = df_weekday.sum()
df_weekday_percentage = (df_weekday / total_crimes) * 100

print(df_weekday_percentage)
```

```
weekday
Monday       15.716181
Tuesday      15.005764
Wednesday    14.859608
Thursday     14.534150
Friday       14.294554
Saturday     12.899850
Sunday       12.689892
Name: count, dtype: float64
```

In [172…

```python
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

# Assuming df is your DataFrame containing the dataset

# Convert 'Date Rptd' to datetime format
df['Date Rptd'] = pd.to_datetime(df['Date Rptd'])

# Extract weekday from 'Date Rptd'
df['weekday'] = df['Date Rptd'].dt.strftime('%A')

# Plot the distribution of crime occurrences by weekday using Seaborn's countplo
plt.figure(figsize=(10, 6))
sns.countplot(data=df, x='weekday', hue='weekday', order=df['weekday'].value_cou
plt.title('Distribution of Crime Occurrences by Weekday')
plt.xlabel('Weekday')
plt.ylabel('Number of Occurrences')
plt.xticks(rotation=45)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```



Monthly Patterns: Examination of Crime Trends on a Monthly Scale

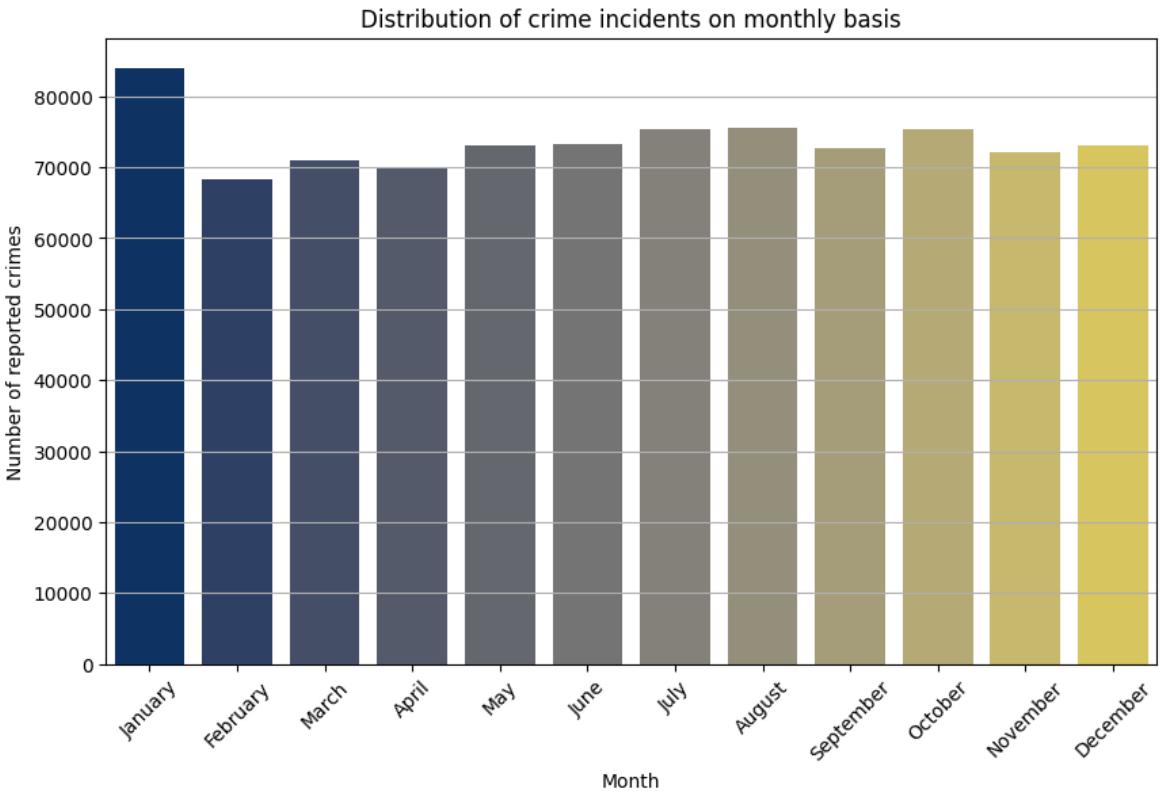
In [13]:
```
import calendar
df['month'] = df['Date Rptd'].dt.month
df_month = df['month'].value_counts().sort_values(ascending=False).reset_index()
df_month['month'] = df_month['month'].map(lambda x: calendar.month_name[x])
df_month
```

Out[13]:

	month	count
0	January	84053
11	February	68370
9	March	71015
10	April	69984
5	May	73122
4	June	73273
2	July	75446
1	August	75576
7	September	72634
3	October	75437
8	November	72048
6	December	73029

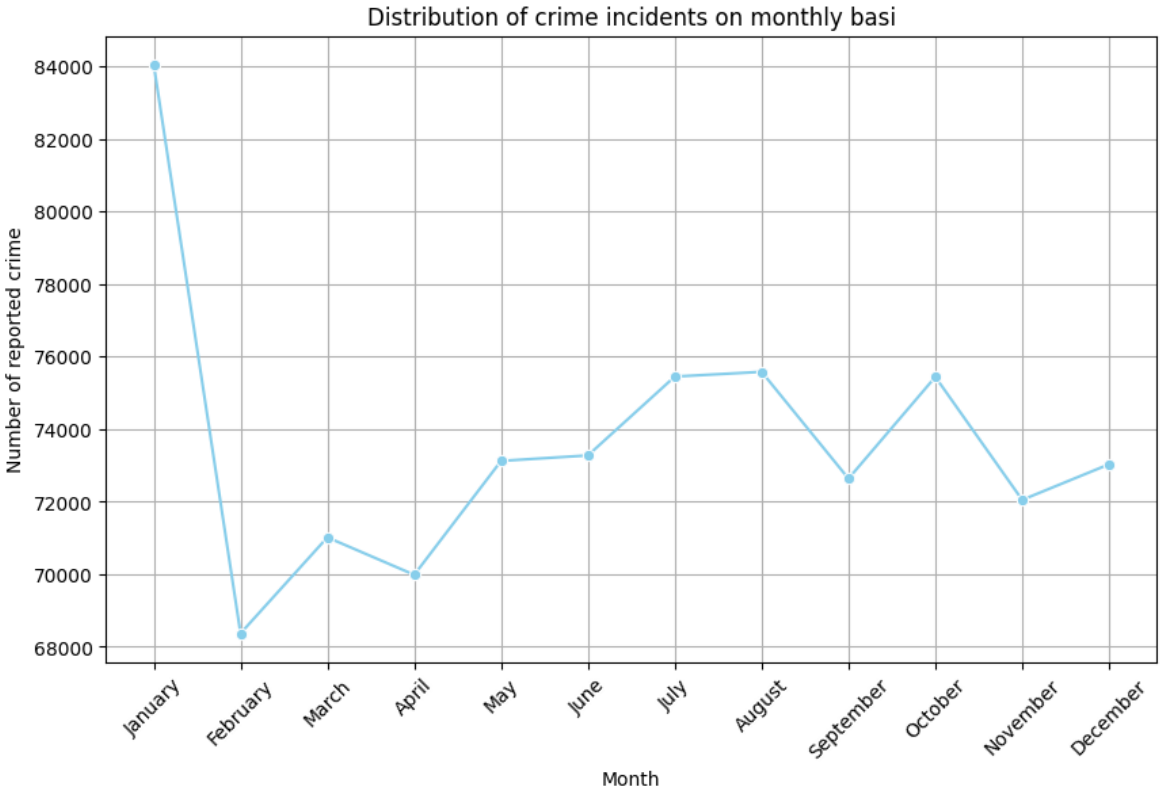
📊 January had the highest number of reported cases, whereas February had the lowest number.

In [116…
```
import seaborn as sns
import matplotlib.pyplot as plt

# Create a bar plot for the count of crimes per month
plt.figure(figsize=(10, 6))
sns.barplot(x='month', y='count', hue='month', data=df_month, palette='cividis',
plt.title('Distribution of crime incidents on monthly basis')
plt.xlabel('Month')
plt.ylabel('Number of reported crimes')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```



In [117…

```
import seaborn as sns
import matplotlib.pyplot as plt

# Create a line plot for the count of crimes per month
plt.figure(figsize=(10, 6))
sns.lineplot(x='month', y='count', data=df_month, marker='o', color='skyblue')
plt.title('Distribution of crime incidents on monthly basi')
plt.xlabel('Month')
plt.ylabel('Number of reported crime')
plt.xticks(rotation=45)
plt.grid(True)
plt.show()
```



Yearly Analysis: Trends and Patterns in Crime Over the Years

In [17]:
```
df['year'] = df['Date Rptd'].dt.year
df_year = df['year'].value_counts().reset_index().sort_values(by='year')
df_year
```

Out[17]:

	year	count
3	2020	192705
2	2021	208284
0	2022	235069
1	2023	234613
4	2024	13316

🟡 The year 2024 shows very less number of reported cases because this project is being done on Feb 2024 and we have only one month's data of 2024.

🟡 The year 2020 had the highest number of crimes reported.

In [120…
```
import seaborn as sns
import matplotlib.pyplot as plt

# Create a line plot for the count of crimes per year
plt.figure(figsize=(10, 6))
sns.lineplot(x='year', y='count', data=df_year, marker='o', color='red')
plt.title('Number of Crimes reported per Year')
plt.xlabel('Year')
```

```
plt.ylabel('Count of Crimes')
plt.xticks(rotation=45)
plt.grid(True)
plt.show()
```

Reporting vs. Occurrence: Comparison of Reported Date and Occurrence Date

In [121…
```
#Number of cases that were reported the same day they occurred
same_day = (df['Date Rptd'] == df['DATE OCC']).sum()
same_per = same_day/ len(df) *100
same_per
```

Out[121…
```
49.237247483889654
```

In [122…
```
#Number of cases that were not reported the same day they occurred
diff_day = (df['Date Rptd'] != df['DATE OCC']).sum()
diff_per = diff_day/ len(df)*100
diff_per
```

Out[122…
```
50.762752516110346
```

📊 Only 49.2% of the crimes were reported on the same day as the occurrence of the crime.

Area Distribution: Spatial Distribution of Crime Incidents

In [21]:
```
df['AREA NAME'].value_counts().sort_values( ascending=False)
```

Out[21]:
```
AREA NAME
Central        60123
77th Street    55450
Pacific        51520
Southwest      49598
Hollywood      46558
Southeast      44696
Olympic        44621
N Hollywood    44263
Newton         44214
Wilshire       42269
Rampart        41644
West LA        40495
Northeast      38019
Van Nuys       37351
West Valley    37157
Harbor         36422
Topanga        36110
Devonshire     35977
Mission        35202
Hollenbeck     32993
Foothill       29305
Name: count, dtype: int64
```

🔍 There were a lot of crimes in the Central area. I would like to investigate further the kind of crimes most common in that area.

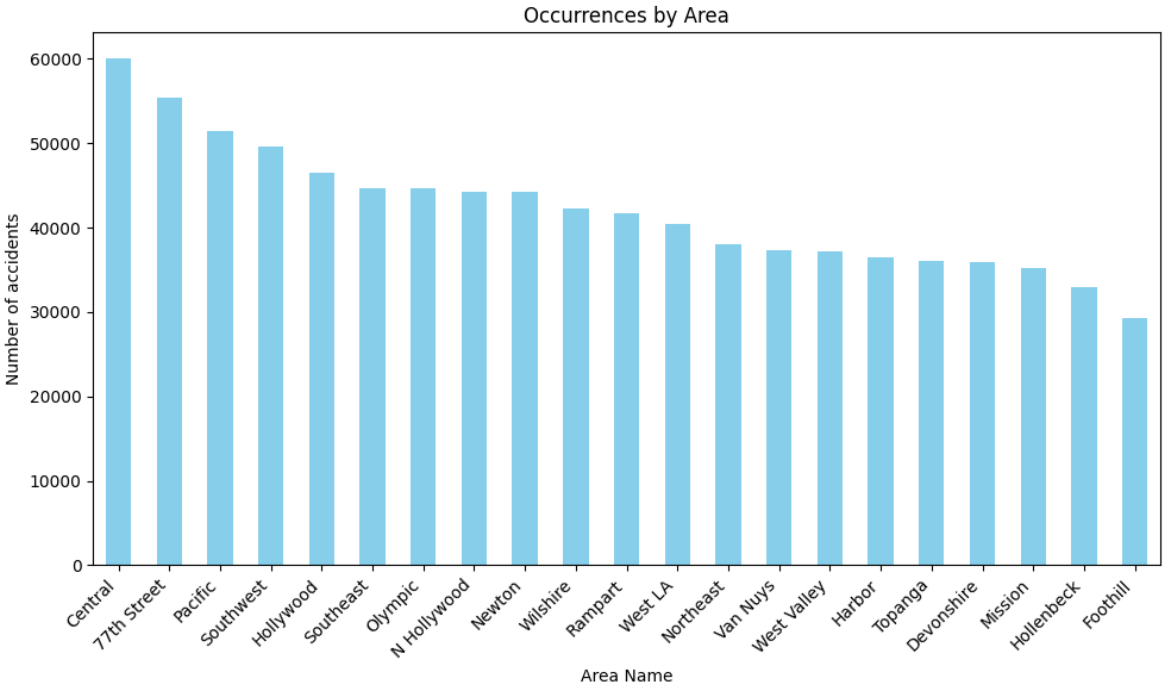
In [22]:
```
df['AREA NAME'].value_counts().sort_values( ascending=True)
```

Out[22]:
```
AREA NAME
Foothill       29305
Hollenbeck     32993
Mission        35202
Devonshire     35977
Topanga        36110
Harbor         36422
West Valley    37157
Van Nuys       37351
Northeast      38019
West LA        40495
Rampart        41644
Wilshire       42269
Newton         44214
N Hollywood    44263
Olympic        44621
Southeast      44696
Hollywood      46558
Southwest      49598
Pacific        51520
77th Street    55450
Central        60123
Name: count, dtype: int64
```

📉 Foothill area had the minimum number of crime rate.

Top 20 areas with the highest crime rates:

In [23]:
```
import matplotlib.pyplot as plt

# Calculate counts of occurrences for each area and sort in descending order
area_counts = df['AREA NAME'].value_counts().sort_values(ascending=False)

# Create a bar plot
plt.figure(figsize=(10, 6))
area_counts.plot(kind='bar', color='skyblue')
plt.title('Occurrences by Area')
plt.xlabel('Area Name')
plt.ylabel('Number of accidents')
plt.xticks(rotation=45, ha='right')  # Rotate x-axis labels for better readabili
plt.tight_layout()  # Adjust layout to prevent clipping of labels
plt.show()
```



Location Analysis: Geographical Patterns of Crime Occurrence

In [24]:
```
df['LOCATION'].value_counts().sort_values( ascending=False)
```

Out[24]:
```
LOCATION
800 N  ALAMEDA                    ST      1667
6TH                          ST            1474
7TH                          ST            1403
10200    SANTA MONICA                BL    1400
7TH                                        1377
                                           ...
3500    STONEWOOD                   DR        1
10300    LAS LUNITAS                 AV        1
11600    VIKING                      AV        1
3700    STARGAZER                   AV        1
6400 N  FIGUEROAA                   ST        1
Name: count, Length: 64796, dtype: int64
```

🔍 The location "800 N Alameda St" has the highest count of incidents, with a total of 1667 reported cases.

Gender Disparities: Analysis of Crime Rates by Gender

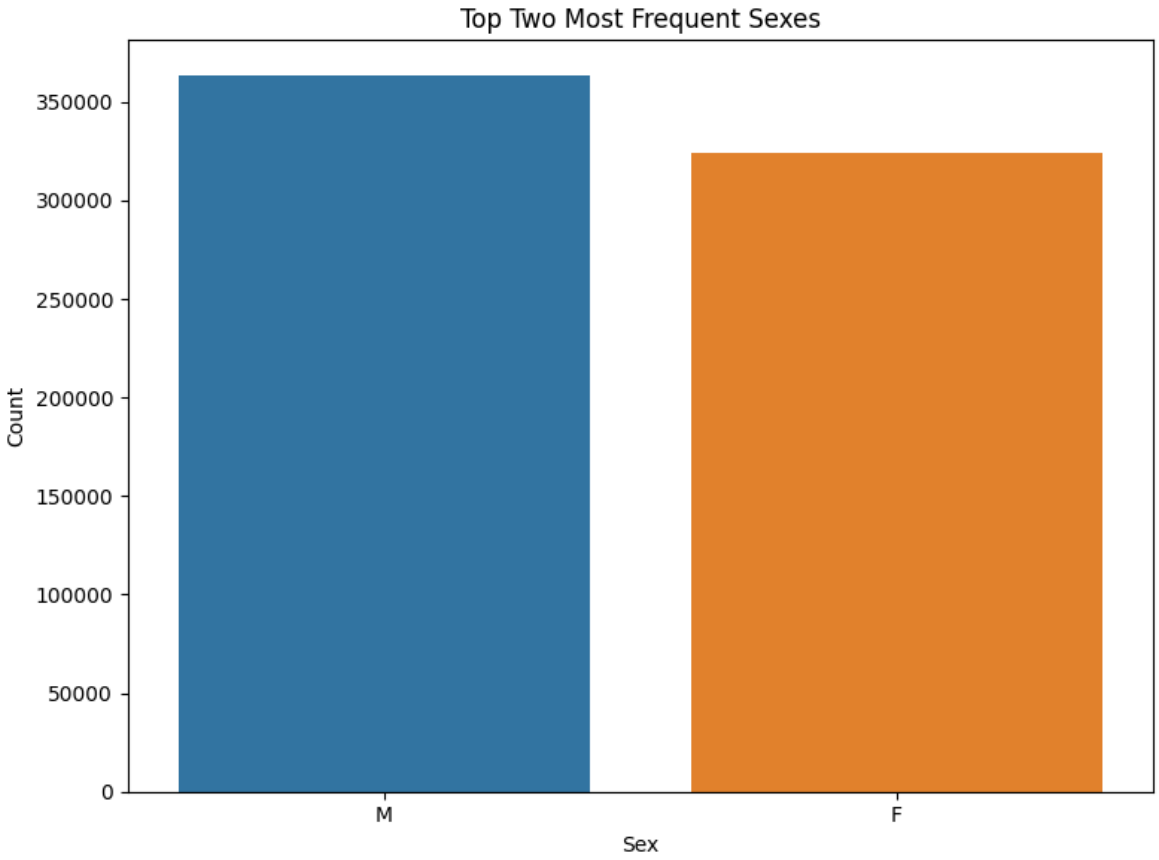
In [25]:
```
df['Vict Sex'].unique()
# I am not sure what X and H means here. So ignoring those two.
```

Out[25]:
```
array(['M', 'X', 'F', nan, 'H', '-'], dtype=object)
```

In [26]:
```
df['Vict Sex'].value_counts().sort_values( ascending=False)[:2]
```

Out[26]:
```
Vict Sex
M    363260
F    323891
Name: count, dtype: int64
```

In [125…
```
import seaborn as sns
import matplotlib.pyplot as plt

# Get the top two most frequent values and their counts
top_sex_counts = df['Vict Sex'].value_counts().sort_values(ascending=False)[:2]

# Create a bar plot using Seaborn
plt.figure(figsize=(8, 6))
sns.barplot(x=top_sex_counts.index, y=top_sex_counts.values, hue=top_sex_counts.
plt.title('Top Two Most Frequent Sexes')
plt.xlabel('Sex')
plt.ylabel('Count')
plt.xticks(rotation=0)  # Rotate x-axis labels if needed
plt.tight_layout()
plt.show()
```



📊 The number of male victimsweres higher.

In [28]:
```
df[df['Vict Sex'] == 'F']['Crm Cd Desc'].value_counts().sort_values(ascending=Fa
## This calculates the most common type of crime faced by women
```

Out[28]:
```
Crm Cd Desc
INTIMATE PARTNER - SIMPLE ASSAULT           33499
BATTERY - SIMPLE ASSAULT                    33043
THEFT OF IDENTITY                           31802
BURGLARY FROM VEHICLE                       22692
THEFT PLAIN - PETTY ($950 & UNDER)          19434
                                            ...
BRIBERY                                         2
FIREARMS RESTRAINING ORDER (FIREARMS RO)        1
LYNCHING - ATTEMPTED                            1
GRAND THEFT / AUTO REPAIR                       1
INCITING A RIOT                                 1
Name: count, Length: 132, dtype: int64
```

🔍 The prevalent types of crime experienced by women were: Assault by an intimate partner, General assault by anyone, and Identity theft.tity

In [29]:
```
df[df['Vict Sex'] == 'M']['Crm Cd Desc'].value_counts().sort_values(ascending=Fa
## This calculates the most common type of crime faced by men
```

Out[29]:
```
Crm Cd Desc
BATTERY - SIMPLE ASSAULT                                    36827
ASSAULT WITH DEADLY WEAPON, AGGRAVATED ASSAULT              36028
BURGLARY FROM VEHICLE                                       30559
BURGLARY                                                    26727
VANDALISM - FELONY ($400 & OVER, ALL CHURCH VANDALISMS)     24710
                                                            ...
DISHONEST EMPLOYEE ATTEMPTED THEFT                              2
PETTY THEFT - AUTO REPAIR                                       2
PURSE SNATCHING - ATTEMPT                                       1
PICKPOCKET, ATTEMPT                                             1
PIMPING                                                         1
Name: count, Length: 134, dtype: int64
```

🔍 The top crimes encountered by men were: Simple assault, Assault with a deadly weapon, and Burglary from a vehicle. It's unsurprising that men didn't face assault by their intimate partners.

Weapon Usage: Identification of Most Common Weapons in Male Assaults

In [30]:
```
#What kind of weapons were men assaulted with the most?

df[df['Vict Sex'] == 'M']['Weapon Desc'].value_counts().sort_values(ascending=Fa
```

Out[30]:
```
Weapon Desc
STRONG-ARM (HANDS, FIST, FEET OR BODILY FORCE)    65026
UNKNOWN WEAPON/OTHER WEAPON                       17026
HAND GUN                                          12771
VERBAL THREAT                                     10441
SEMI-AUTOMATIC PISTOL                              4646
                                                   ...
ANTIQUE FIREARM                                       5
MAC-11 SEMIAUTOMATIC ASSAULT WEAPON                   2
MAC-10 SEMIAUTOMATIC ASSAULT WEAPON                   2
M1-1 SEMIAUTOMATIC ASSAULT RIFLE                      1
M-14 SEMIAUTOMATIC ASSAULT RIFLE                      1
Name: count, Length: 78, dtype: int64
```

🔍 The prevalent weapon used for assaulting men was "strong arm".

Age Distribution: Understanding the Age Profile of Victims

In [129…
```
df['Vict Age'].value_counts().sort_values(ascending=False)
```

Out[129…
```
Vict Age
0      221363
30      20061
35      19708
31      19208
29      19164
        ...
95         93
96         91
98         68
97         65
120         1
Name: count, Length: 100, dtype: int64
```

In [142…
```
df['Vict Age'].unique()
```

Out[142…
```
array([ 47,  19,  28,  41,  25,  27,  24,  26,   8,   7,  13,  56,  22,
        23,  31,  30,  57,  10,  12,  46,  51,  37,  20,  29,  33,  34,
        15,  65,  59,   9,  35,  50,  64,  79,  40,  44,  32,  16,  21,
        42,   6,  72,  60,  36,  18,  17,   5,  39,  62,  58,  63,  14,
        71,  69,  49,  38,  61,  43,  45,  78,  67,  53,  54,  11,  48,
        73,  52,  68,  81,  80,   3,  55,  89,  70,  77,  83,  91,  82,
        75,  76,  74,   4,   2,  66,  84,  87,  86,  85,  90,  99,  96,
        88,  92,  94,  93,  95,  97,  98, 120], dtype=int64)
```

In [139…
```
# Drop rows where victim age is 0
df_age = df[df['Vict Age'] != 0]
df_age['Vict Age'].value_counts().sort_values(ascending=False)
```

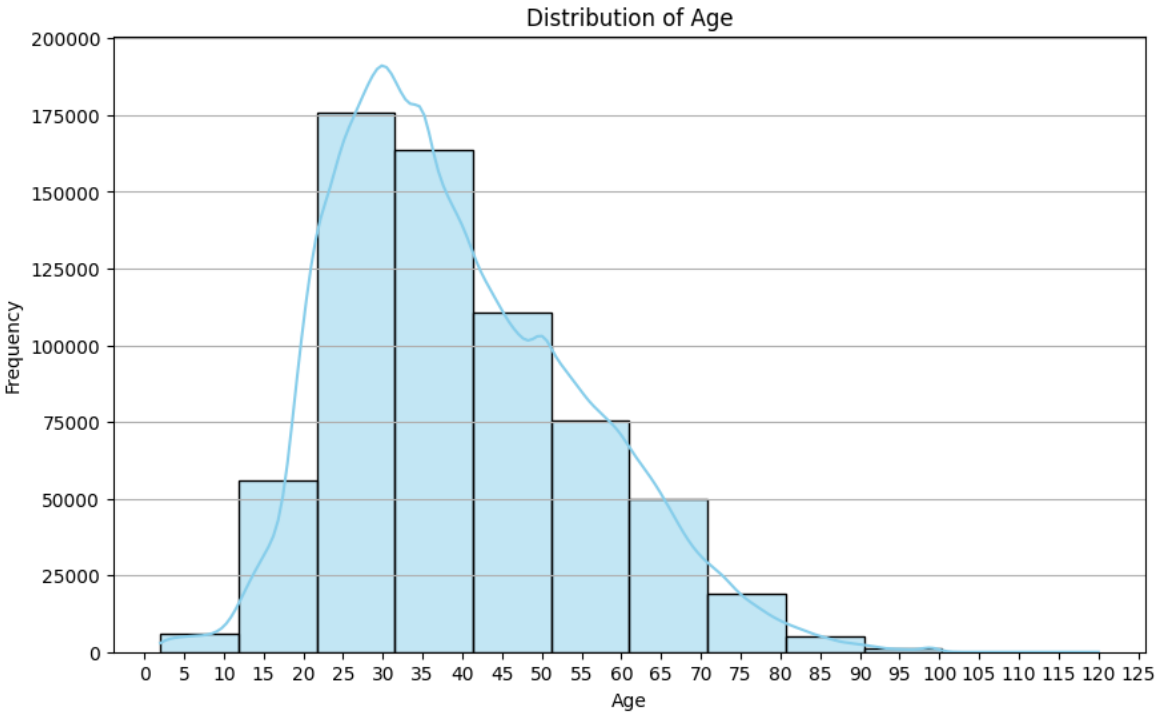
Out[139…

```
Vict Age
30     20061
35     19708
31     19208
29     19164
28     18870
       ...
95        93
96        91
98        68
97        65
120        1
Name: count, Length: 99, dtype: int64
```

📊 The maximum number of victims were aged 30.

In [145…

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(10, 6))
sns.histplot(data=df_age, x='Vict Age', bins=12, kde=True, color='skyblue')
plt.title('Distribution of Age')
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.xticks(range(0, 130, 5))  # Set x-axis ticks from 0 to 100 with intervals of
plt.grid(axis='y')
plt.show()
```



📊 The highest number of victims belonged to the age group 25-40 years.

📊 There were very few victims aged over 70 years.

In [147…

```
#What kind of crime was most common for victim of age group between 25-40 years?

df[(df['Vict Age'] >= 25) & (df['Vict Age'] <= 40)]['Crm Cd Desc'].value_counts(
```

Out[147…

```
Crm Cd Desc
BURGLARY FROM VEHICLE                                27812
THEFT OF IDENTITY                                    25472
BATTERY - SIMPLE ASSAULT                             25278
INTIMATE PARTNER - SIMPLE ASSAULT                    23832
ASSAULT WITH DEADLY WEAPON, AGGRAVATED ASSAULT       19997
                                                     ...
PICKPOCKET, ATTEMPT                                      1
INCEST (SEXUAL ACTS BETWEEN BLOOD RELATIVES)             1
FIREARMS RESTRAINING ORDER (FIREARMS RO)                 1
CHILD ABANDONMENT                                        1
GRAND THEFT / INSURANCE FRAUD                            1
Name: count, Length: 129, dtype: int64
```

🚗 Age group 25-40 reported a high number of BURGLARY FROM VEHICLE cases: 27812 incidents.

In [148…

```
#What kind of crime was committed against aged people?

df[(df['Vict Age'] > 70)]['Crm Cd Desc'].value_counts().sort_values(ascending=Fa
```

Out[148…

```
Crm Cd Desc
BURGLARY                                             3439
BATTERY - SIMPLE ASSAULT                             2703
THEFT OF IDENTITY                                    2653
THEFT PLAIN - PETTY ($950 & UNDER)                   1766
VANDALISM - FELONY ($400 & OVER, ALL CHURCH VANDALISMS)    1477
                                                     ...
DEFRAUDING INNKEEPER/THEFT OF SERVICES, OVER $950.01        1
RECKLESS DRIVING                                            1
CONSPIRACY                                                  1
WEAPONS POSSESSION/BOMBING                                  1
EMBEZZLEMENT, PETTY THEFT ($950 & UNDER)                    1
Name: count, Length: 98, dtype: int64
```

🔍 It seems that older individuals reported a significant number of burglary incidents.

Race-Based Crime Rates: Crime Rates Segmented by Victim Race

Adding a new column for victim's race

In [150…

```
# Mapping of abbreviations to full descriptions
descent_mapping = {
    'O': 'Other',
    'H': 'Hispanic',
    'B': 'Black or African American',
    'W': 'White',
    'A': 'Asian',
    'I': 'American Indian or Alaska Native',
    'P': 'Native Hawaiian or Other Pacific Islander',
    'X': 'Unknown',  # Placeholder for unknown or missing data
}

# Add a new column 'Vict_Descent_Desc' with full descriptions
df['Race'] = df['Vict Descent'].map(descent_mapping)
```

```
# Display the DataFrame to verify the new column has been added
print(df.Race)
```

```
1                              Other
2                            Unknown
3                              Other
4                           Hispanic
5                           Hispanic
                    ...
883982    Black or African American
883983                      Hispanic
883984                      Hispanic
883985                      Hispanic
883986                         White
Name: Race, Length: 662541, dtype: object
```

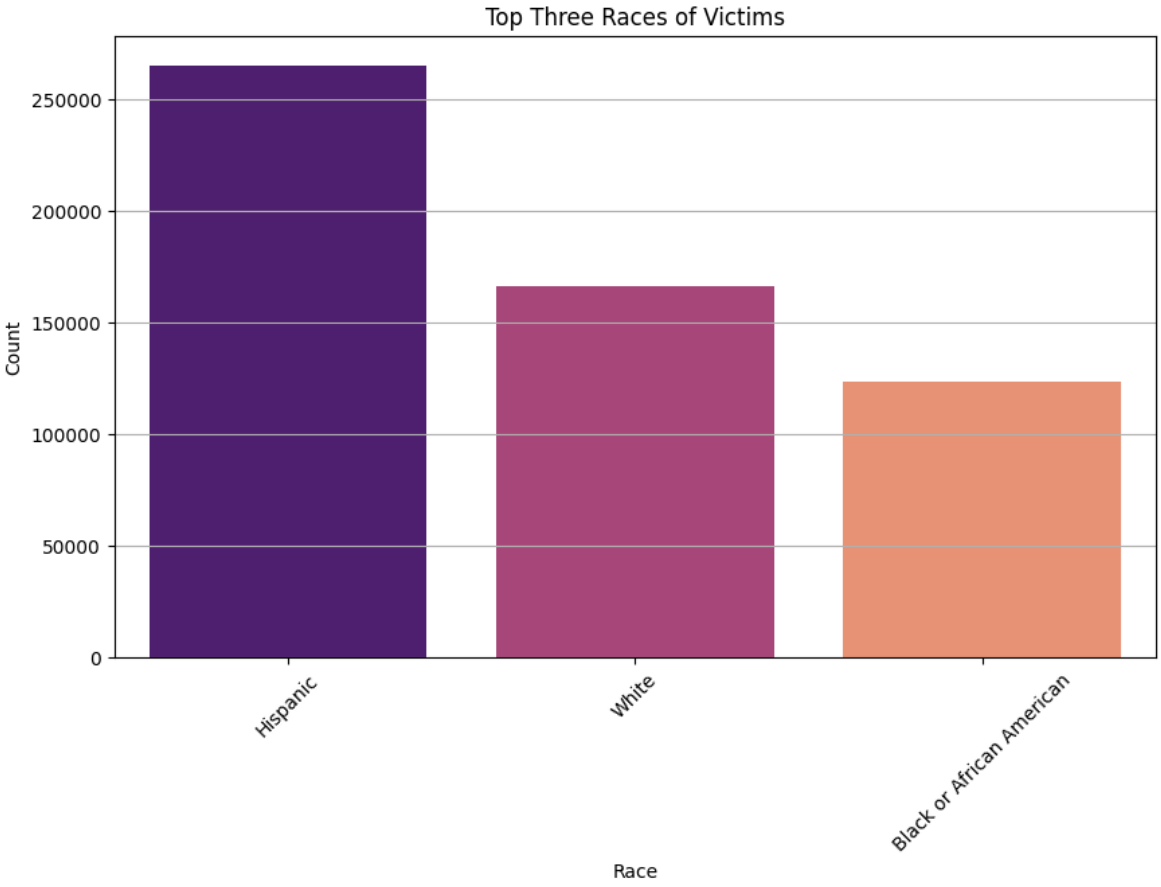
In [39]:
```
df['Race'].value_counts().sort_values(ascending=False)
```

Out[39]:
```
Race
Hispanic                                    269738
White                                       179118
Black or African American                   124819
Unknown                                      87717
Other                                        69904
Asian                                        19377
American Indian or Alaska Native               840
Native Hawaiian or Other Pacific Islander      237
Name: count, dtype: int64
```

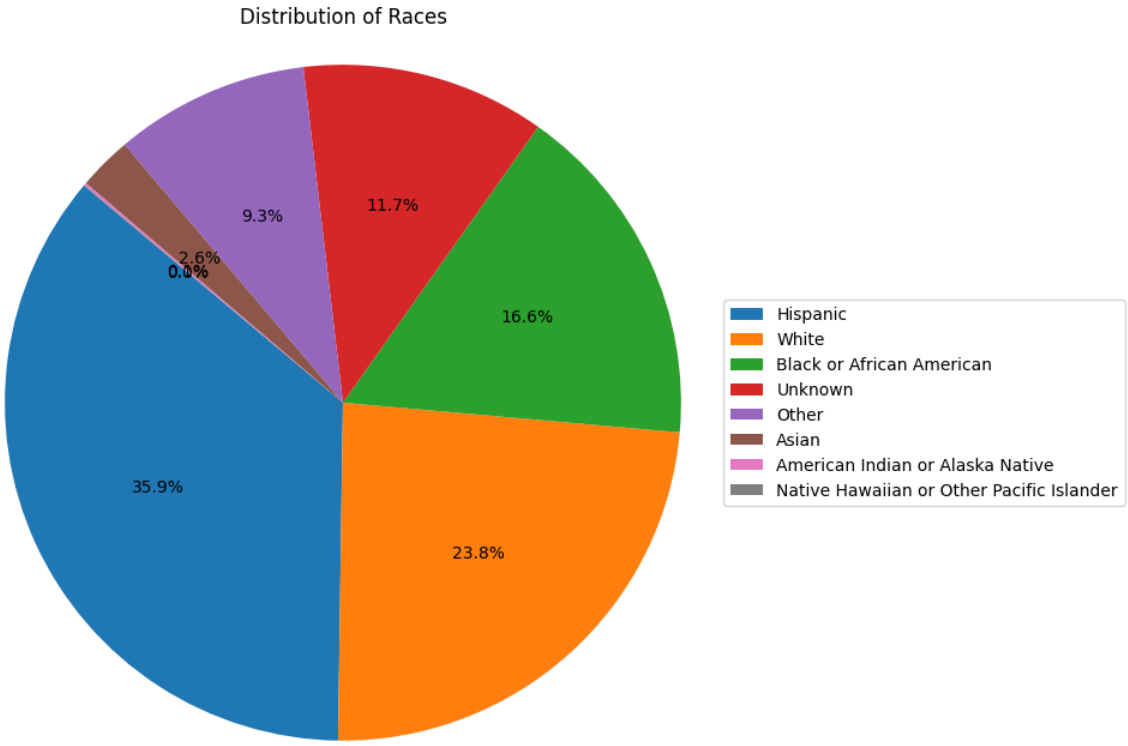
📊 Most victims belonged to the Hispanic race/ethnicity.

The top three races of the victims were:

In [155…
```
import seaborn as sns
import matplotlib.pyplot as plt

# Assuming df['Race'] contains race information
race_counts = df['Race'].value_counts().sort_values(ascending=False)[:3]

# Plot the bar plot using Seaborn
plt.figure(figsize=(10, 6))
sns.barplot(x=race_counts.index, y=race_counts.values, hue=race_counts.index, le
plt.title('Top Three Races of Victims')
plt.xlabel('Race')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```



In [41]:
```
import matplotlib.pyplot as plt

# Assuming df['Race'] contains race information
race_counts = df['Race'].value_counts().sort_values(ascending=False)

# Create a pie chart
plt.figure(figsize=(8, 8))
plt.pie(race_counts.values, labels=None, autopct='%1.1f%%', startangle=140)
plt.legend(race_counts.index, loc="center left", bbox_to_anchor=(1, 0, 0.5, 1))
plt.title('Distribution of Races')
plt.axis('equal')  # Equal aspect ratio ensures that pie is drawn as a circle.
plt.show()
```



Investigation Status: Overview of Crime Incidents by Investigation Status

In [42]:
```
df['Status Desc'].value_counts().sort_values(ascending=False)
```

Out[42]:
```
Status Desc
Invest Cont     707611
Adult Other      95153
Adult Arrest     76714
Juv Arrest        2869
Juv Other         1553
UNK                  4
Name: count, dtype: int64
```

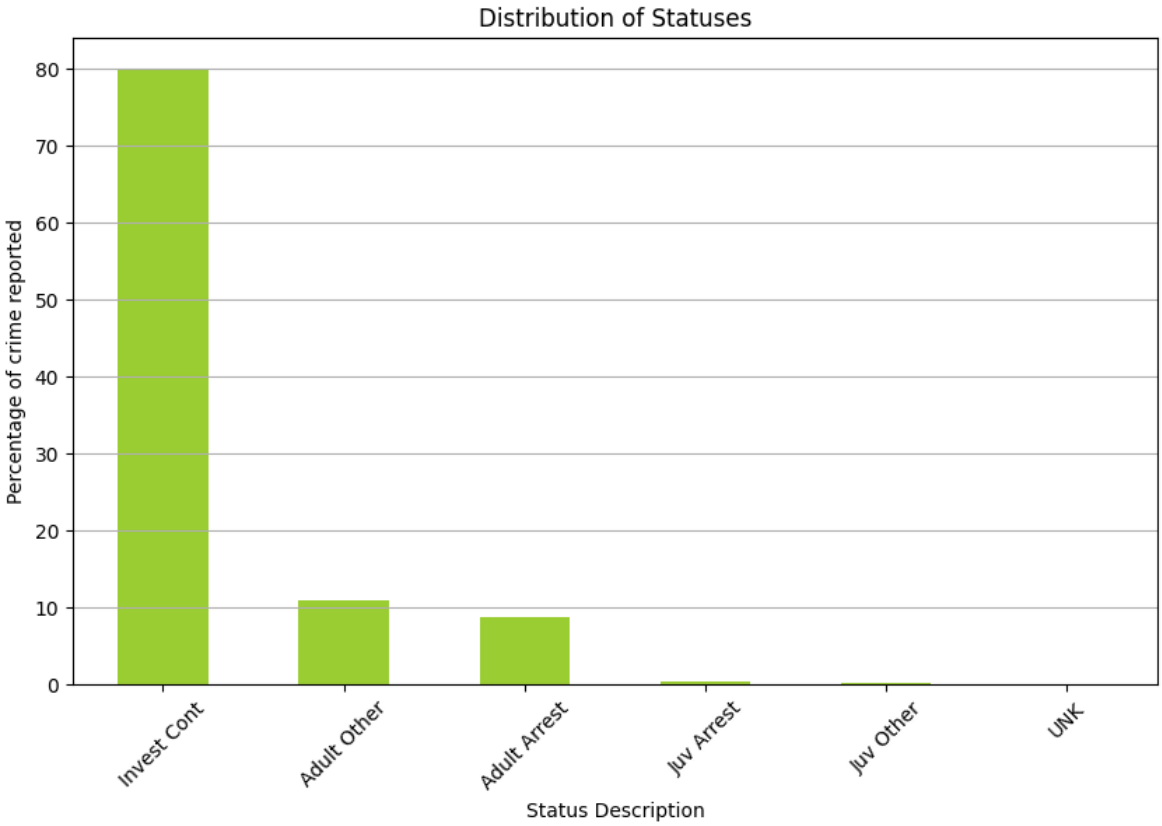
In [43]:
```
#Percentage of investigations that are still ongoing:

df['Status Desc'].value_counts().sort_values(ascending=False)/len(df)*100
```

Out[43]:
```
Status Desc
Invest Cont     80.055187
Adult Other     10.765083
Adult Arrest     8.678997
Juv Arrest       0.324583
Juv Other        0.175698
UNK              0.000453
Name: count, dtype: float64
```

🔍 Nearly 80% of reported crimes are still under investigation.

In [44]:
```
import matplotlib.pyplot as plt

status_counts = df['Status Desc'].value_counts().sort_values(ascending=False)/le


```

```
plt.figure(figsize=(10, 6))
status_counts.plot(kind='bar', color='yellowgreen')
plt.title('Distribution of Statuses')
plt.xlabel('Status Description')
plt.ylabel('Percentage of crime reported')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```



In [45]:
```
#In what kind  of crime Juv were arrested?

df[df['Status Desc'] == 'Juv Arrest']['Crm Cd Desc'].value_counts().sort_values(
```

Out[45]:
```
Crm Cd Desc
ROBBERY                                                572
ASSAULT WITH DEADLY WEAPON, AGGRAVATED ASSAULT         506
BATTERY - SIMPLE ASSAULT                               334
VEHICLE - STOLEN                                       149
VANDALISM - FELONY ($400 & OVER, ALL CHURCH VANDALISMS)  147
                                                       ...
STALKING                                                 1
BURGLARY FROM VEHICLE, ATTEMPTED                         1
VIOLATION OF TEMPORARY RESTRAINING ORDER                 1
BIKE - STOLEN                                            1
PANDERING                                                1
Name: count, Length: 80, dtype: int64
```

🔒 572 juveniles were arrested for robbery, indicating that robbery was the primary crime committed by juveniles.

In [110…
```
#In what kind  of crime Adults were arrested the most?

df[df['Status Desc'] == 'Adult Arrest']['Crm Cd Desc'].value_counts().sort_value
```

Out[110…

```
Crm Cd Desc
ASSAULT WITH DEADLY WEAPON, AGGRAVATED ASSAULT                13.963553
INTIMATE PARTNER - SIMPLE ASSAULT                             11.863545
VANDALISM - FELONY ($400 & OVER, ALL CHURCH VANDALISMS)        6.864458
ROBBERY                                                        6.784941
BURGLARY                                                       5.683448
                                                                ...
DRUGS, TO A MINOR                                              0.001304
BEASTIALITY, CRIME AGAINST NATURE SEXUAL ASSLT WITH ANIM       0.001304
DISHONEST EMPLOYEE ATTEMPTED THEFT                             0.001304
DISHONEST EMPLOYEE - PETTY THEFT                               0.001304
GRAND THEFT / AUTO REPAIR                                      0.001304
Name: count, Length: 121, dtype: float64
```

📉 13% of the arrested adults faced charges of 'Assault with deadly weapon'.

Analyzing the Types of Committed Crimes

In [108…

```
df['Crm Cd Desc'].value_counts().sort_values(ascending=False)
```

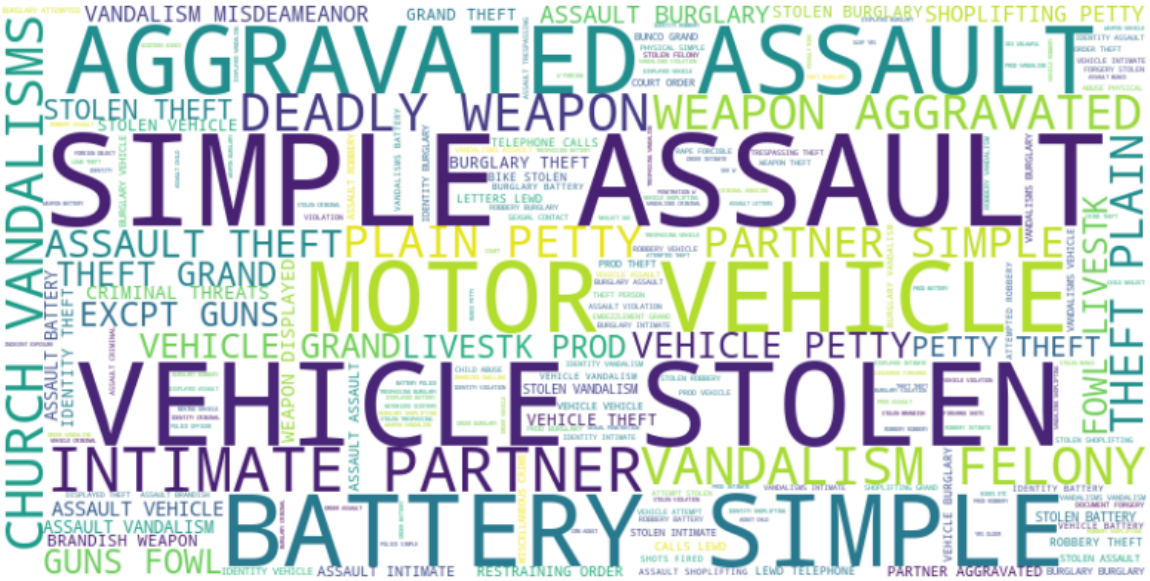
Out[108…

```
Crm Cd Desc
VEHICLE - STOLEN                                    94922
BATTERY - SIMPLE ASSAULT                            70311
THEFT OF IDENTITY                                   54926
BURGLARY FROM VEHICLE                               54536
BURGLARY                                            54070
                                                     ...
THEFT, COIN MACHINE - ATTEMPT                           5
FIREARMS EMERGENCY PROTECTIVE ORDER (FIREARMS EPO)      5
FIREARMS RESTRAINING ORDER (FIREARMS RO)                4
FAILURE TO DISPERSE                                     3
DISHONEST EMPLOYEE ATTEMPTED THEFT                      2
Name: count, Length: 138, dtype: int64
```

Visualization of Reported Crime Types

In [53]:

```
from wordcloud import WordCloud
crime_types = df['Crm Cd Desc']

# Join all crime types into a single string
all_crime_types = ' '.join(crime_types)

# Generate the word cloud
wordcloud = WordCloud(width=800, height=400, background_color='white').generate(

# Plot the word cloud
plt.figure(figsize=(10, 8))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis('off')
plt.show()
```



🔍 The most prevalent crime types identified from the word cloud are: 1. Simple assault 2. Vandalism 3. Vehicle theft.

- Most common type of crime in general : 1. Vehicle stealing 2. Simple Assualt 3. THEFT OF IDENTITY

Average daily crime rate

In [69]:
```
df['DATE OCC'].value_counts().mean()
```

Out[69]:
```
596.0242751180041
```

📅 On average, the daily count of crime incidents reaches around 59..

In [70]:
```
df['Date Rptd'].value_counts().mean()
```

Out[70]:
```
594.8209959623149
```

📅 Approximately 594 crime cases are reported on a daily basis.

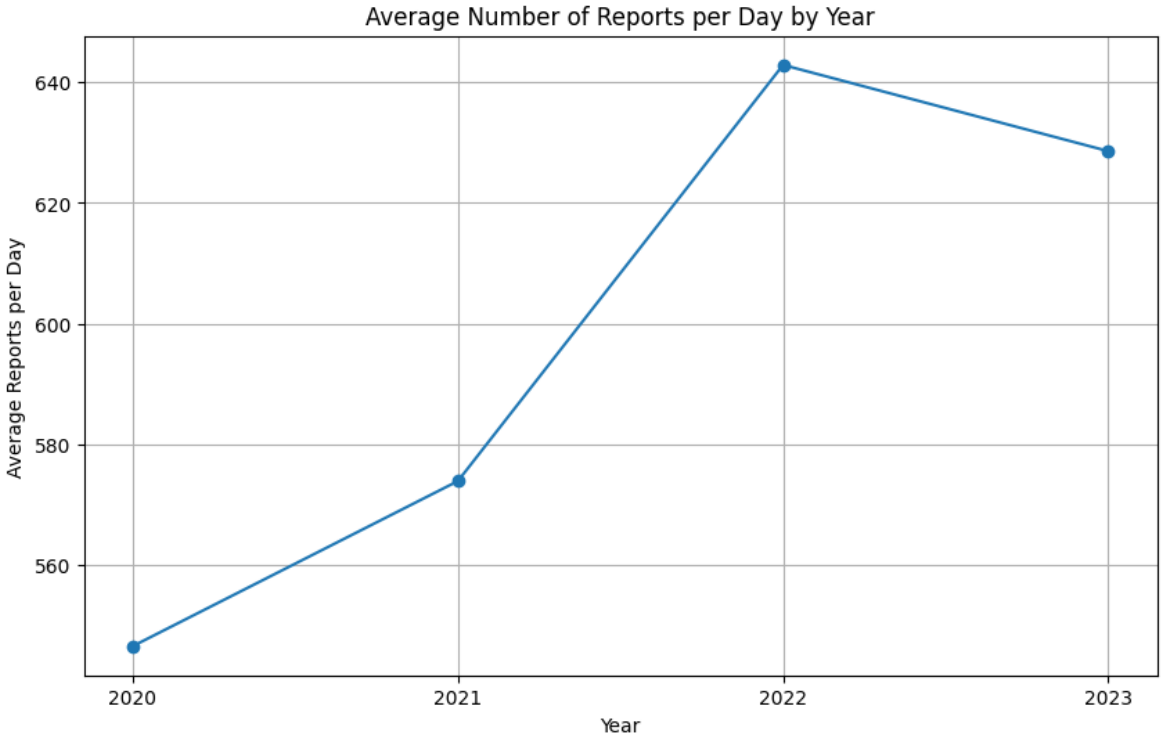
In [158…
```
(df['DATE OCC'].dt.year.value_counts()/365).sort_index()
#Avg daily crime incidents year-wise
```

Out[158…
```
DATE OCC
2020    414.383562
2021    435.706849
2022    489.449315
2023    456.147945
2024     19.493151
Name: count, dtype: float64
```

📅 The year 2022 recorded the highest average number of crime incidents per day.

In [90]:
```
import matplotlib.pyplot as plt

# Calculate the average number of reports per day for each year and sort by year
```

```
avg_reports_per_day = (df['DATE OCC'].dt.year.value_counts() / 365).sort_index()

# Plot the line graph
plt.figure(figsize=(10, 6))
plt.plot(avg_reports_per_day.index, avg_reports_per_day.values, marker='o', line
plt.title('Average Number of Reports per Day by Year')
plt.xlabel('Year')
plt.ylabel('Average Reports per Day')
plt.grid(True)
plt.xticks(avg_reports_per_day.index)  # Set x-axis ticks to the years
plt.show()
```



- The average number of cases has increased over the years.
- Dropped the year 2024 because very not all cases were present in the dataset.

Maximum crime reported and occurred on which day?

In [98]:
```
df['DATE OCC'].value_counts().sort_values(ascending=False)
```

Out[98]:
```
DATE OCC
2022-12-02    1132
2020-01-01    1118
2023-01-01    1097
2023-02-01    1076
2022-10-01    1070
              ...
2020-03-29     421
2020-03-22     417
2020-04-09     415
2024-01-21     366
2024-01-22     144
Name: count, Length: 1483, dtype: int64
```

- Maximum crime incidents occured on : 2022-12-02
- What happended that day?

In [102…
```
df[df['DATE OCC'] == '2022-12-02']['Crm Cd Desc'].value_counts()
```

Out[102…

Crm Cd Desc	
THEFT OF IDENTITY	471
VEHICLE - STOLEN	63
THEFT PLAIN - PETTY ($950 & UNDER)	59
THEFT FROM MOTOR VEHICLE - GRAND ($950.01 AND OVER)	57
BURGLARY	50
BURGLARY FROM VEHICLE	46
BATTERY - SIMPLE ASSAULT	45
VANDALISM - FELONY ($400 & OVER, ALL CHURCH VANDALISMS)	40
THEFT-GRAND ($950.01 & OVER)EXCPT,GUNS,FOWL,LIVESTK,PROD	34
THEFT FROM MOTOR VEHICLE - PETTY ($950 & UNDER)	31
ASSAULT WITH DEADLY WEAPON, AGGRAVATED ASSAULT	25
INTIMATE PARTNER - SIMPLE ASSAULT	23
CRIMINAL THREATS - NO WEAPON DISPLAYED	21
SHOPLIFTING - PETTY THEFT ($950 & UNDER)	18
ROBBERY	14
VANDALISM - MISDEAMEANOR ($399 OR UNDER)	12
BRANDISH WEAPON	11
VIOLATION OF RESTRAINING ORDER	10
BUNCO, GRAND THEFT	10
BIKE - STOLEN	9
LETTERS, LEWD - TELEPHONE CALLS, LEWD	7
PICKPOCKET	6
VEHICLE, STOLEN - OTHER (MOTORIZED SCOOTERS, BIKES, ETC)	6
TRESPASSING	5
VIOLATION OF COURT ORDER	5
RAPE, FORCIBLE	5
INTIMATE PARTNER - AGGRAVATED ASSAULT	4
BATTERY WITH SEXUAL CONTACT	4
CONTEMPT OF COURT	3
BURGLARY, ATTEMPTED	3
ATTEMPTED ROBBERY	2
BUNCO, PETTY THEFT	2
CHILD ANNOYING (17YRS & UNDER)	2
OTHER ASSAULT	2
OTHER MISCELLANEOUS CRIME	2
CHILD ABUSE (PHYSICAL) - SIMPLE ASSAULT	2
BATTERY POLICE (SIMPLE)	2
CHILD STEALING	1
KIDNAPPING - GRAND ATTEMPT	1
SEXUAL PENETRATION W/FOREIGN OBJECT	1
THEFT, PERSON	1
KIDNAPPING	1
SEX,UNLAWFUL(INC MUTUAL CONSENT, PENETRATION W/ FRGN OBJ	1
EMBEZZLEMENT, GRAND THEFT ($950.01 & OVER)	1
CONTRIBUTING	1
PIMPING	1
THEFT PLAIN - ATTEMPT	1
RAPE, ATTEMPTED	1
EXTORTION	1
DOCUMENT FORGERY / STOLEN FELONY	1
CRIMINAL HOMICIDE	1
CRM AGNST CHLD (13 OR UNDER) (14-15 & SUSP 10 YRS OLDER)	1
SHOTS FIRED AT MOVING VEHICLE, TRAIN OR AIRCRAFT	1
TELEPHONE PROPERTY - DAMAGE	1
SEX OFFENDER REGISTRANT OUT OF COMPLIANCE	1
THEFT FROM MOTOR VEHICLE - ATTEMPT	1
SHOPLIFTING-GRAND THEFT ($950.01 & OVER)	1
DISCHARGE FIREARMS/SHOTS FIRED	1

Name: count, dtype: int64

- Many people reported 'Theft of identity' on that day.