In [230]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

In [231]:
```
df = pd.read_csv('AviationData.csv',encoding='latin1',low_memory=False)
```

In [232]:
```
df.head()
```

Out[232]:

	Event.Id	Investigation.Type	Accident.Number	Event.Date	Location	Country
0	20001218X45444	Accident	SEA87LA080	1948-10-24	MOOSE CREEK, ID	United States
1	20001218X45447	Accident	LAX94LA336	1962-07-19	BRIDGEPORT, CA	United States
2	20061025X01555	Accident	NYC07LA005	1974-08-30	Saltville, VA	United States
3	20001218X45448	Accident	LAX96LA321	1977-06-19	EUREKA, CA	United States
4	20041105X01764	Accident	CHI79FA064	1979-08-02	Canton, OH	United States

5 rows × 31 columns

In [233]:
```
df.tail()
```

Out[233]:

	Event.Id	Investigation.Type	Accident.Number	Event.Date	Location	Country
88884	20221227106491	Accident	ERA23LA093	2022-12-26	Annapolis, MD	United States
88885	20221227106494	Accident	ERA23LA095	2022-12-26	Hampton, NH	United States
88886	20221227106497	Accident	WPR23LA075	2022-12-26	Payson, AZ	United States
88887	20221227106498	Accident	WPR23LA076	2022-12-26	Morgan, UT	United States
88888	20221230106513	Accident	ERA23LA097	2022-12-29	Athens, GA	United States

5 rows × 31 columns

In [234]:
```
df.shape
```

Out[234]: (88889, 31)

In [235]:
```
##Data information
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 88889 entries, 0 to 88888
Data columns (total 31 columns):
 #   Column                  Non-Null Count  Dtype  
---  ------                  --------------  -----  
 0   Event.Id                88889 non-null  object 
 1   Investigation.Type      88889 non-null  object 
 2   Accident.Number         88889 non-null  object 
 3   Event.Date              88889 non-null  object 
 4   Location                88837 non-null  object 
 5   Country                 88663 non-null  object 
 6   Latitude                34382 non-null  object 
 7   Longitude               34373 non-null  object 
 8   Airport.Code            50249 non-null  object 
 9   Airport.Name            52790 non-null  object 
 10  Injury.Severity         87889 non-null  object 
 11  Aircraft.damage         85695 non-null  object 
 12  Aircraft.Category       32287 non-null  object 
 13  Registration.Number     87572 non-null  object 
 14  Make                    88826 non-null  object 
 15  Model                   88797 non-null  object 
 16  Amateur.Built           88787 non-null  object 
 17  Number.of.Engines       82805 non-null  float64
 18  Engine.Type             81812 non-null  object 
 19  FAR.Description         32023 non-null  object 
 20  Schedule                12582 non-null  object 
 21  Purpose.of.flight       82697 non-null  object 
 22  Air.carrier             16648 non-null  object 
 23  Total.Fatal.Injuries    77488 non-null  float64
 24  Total.Serious.Injuries  76379 non-null  float64
 25  Total.Minor.Injuries    76956 non-null  float64
 26  Total.Uninjured         82977 non-null  float64
 27  Weather.Condition       84397 non-null  object 
 28  Broad.phase.of.flight   61724 non-null  object 
 29  Report.Status           82508 non-null  object 
 30  Publication.Date        75118 non-null  object 
dtypes: float64(5), object(26)
memory usage: 21.0+ MB
```

In [118]:
```
#DATA CLEANING
```

In [236]:
```
##Checking for the missing data
df.isnull().sum()
```

Out[236]:
```
Event.Id                    0
Investigation.Type          0
Accident.Number             0
Event.Date                  0
Location                   52
Country                   226
Latitude                54507
Longitude               54516
Airport.Code            38640
Airport.Name            36099
Injury.Severity          1000
Aircraft.damage          3194
Aircraft.Category       56602
Registration.Number      1317
Make                       63
Model                      92
Amateur.Built             102
Number.of.Engines        6084
Engine.Type              7077
FAR.Description         56866
Schedule                76307
Purpose.of.flight        6192
Air.carrier             72241
Total.Fatal.Injuries    11401
Total.Serious.Injuries  12510
Total.Minor.Injuries    11933
Total.Uninjured          5912
Weather.Condition        4492
Broad.phase.of.flight   27165
Report.Status            6381
Publication.Date        13771
dtype: int64
```

In [237]:
```
##Handling Missing Values
##Columns such as Injury.Severity, Aircraft.damage and Weather.condition a
##Drop columns that have excess missing values (>50%)
##Drop rows for the critical columns that have sparse missing data
df_cleaned = df.drop(columns=['Aircraft.Category', 'Schedule', 'Air.carrie
```

In [238]:
```
## Use mode for the categorical variables
df['Weather.Condition'].fillna(df['Weather.Condition'].mode()[0], inplace=
df['Aircraft.damage'].fillna('Unknown', inplace=True)
```

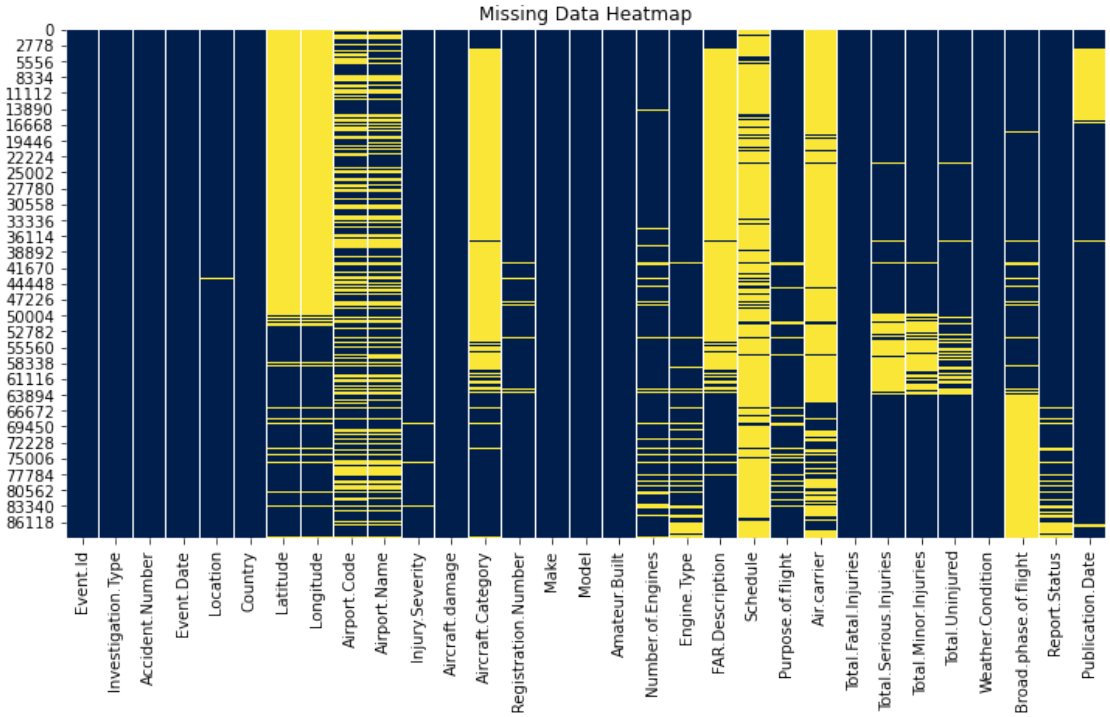
In [239]:
```
##Use Mean for the Numerical Variables
df['Total.Fatal.Injuries'].fillna(df['Total.Fatal.Injuries'].mean(), inpla
## Grouped Computation of the Numerical Variables
df['Total.Fatal.Injuries'] = df.groupby('Aircraft.damage')['Total.Fatal.Ir
    lambda x: x.fillna(x.mean()))
```

In [240]:
```
##Correlation Check
df.corr()
```

Out[240]:

	Number.of.Engines	Total.Fatal.Injuries	Total.Serious.Injuries	Total.Minc
Number.of.Engines	1.000000	0.091553	0.046157	
Total.Fatal.Injuries	0.091553	1.000000	0.135099	
Total.Serious.Injuries	0.046157	0.135099	1.000000	
Total.Minor.Injuries	0.098162	0.051716	0.326849	
Total.Uninjured	0.406058	-0.012921	0.052869	

In [242]:
```
##Missing Value Visualization
plt.figure(figsize=(12, 6))
sns.heatmap(df.isnull(), cbar=False, cmap="cividis")  # Alternative to vir
plt.title("Missing Data Heatmap")
plt.show()
```



In [243]:
```
#A heatmap to check for the colors
missing_data = df.isnull()
plt.figure(figsize=(12, 6))
plt.imshow(missing_data, aspect="auto", cmap="viridis", interpolation="nor
plt.colorbar(label="Missing Data")
plt.title("Missing Data Visualization")
plt.xlabel("Columns")
plt.ylabel("Rows")
plt.show()
```

In [248]:
```
##Check the state of the dataset after cleaning
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 88889 entries, 0 to 88888
Data columns (total 31 columns):
 #   Column                  Non-Null Count  Dtype  
---  ------                  --------------  -----  
 0   Event.Id                88889 non-null  object 
 1   Investigation.Type      88889 non-null  object 
 2   Accident.Number         88889 non-null  object 
 3   Event.Date              88889 non-null  object 
 4   Location                88837 non-null  object 
 5   Country                 88663 non-null  object 
 6   Latitude                34382 non-null  object 
 7   Longitude               34373 non-null  object 
 8   Airport.Code            50249 non-null  object 
 9   Airport.Name            52790 non-null  object 
 10  Injury.Severity         87889 non-null  object 
 11  Aircraft.damage         88889 non-null  object 
 12  Aircraft.Category       32287 non-null  object 
 13  Registration.Number     87572 non-null  object 
 14  Make                    88826 non-null  object 
 15  Model                   88797 non-null  object 
 16  Amateur.Built           88787 non-null  object 
 17  Number.of.Engines       82805 non-null  float64
 18  Engine.Type             81812 non-null  object 
 19  FAR.Description         32023 non-null  object 
 20  Schedule                12582 non-null  object 
 21  Purpose.of.flight       82697 non-null  object 
 22  Air.carrier             16648 non-null  object 
 23  Total.Fatal.Injuries    88889 non-null  float64
 24  Total.Serious.Injuries  76379 non-null  float64
 25  Total.Minor.Injuries    76956 non-null  float64
 26  Total.Uninjured         82977 non-null  float64
 27  Weather.Condition       88889 non-null  object 
 28  Broad.phase.of.flight   61724 non-null  object 
 29  Report.Status           82508 non-null  object 
 30  Publication.Date        75118 non-null  object 
dtypes: float64(5), object(26)
memory usage: 21.0+ MB
```

In [245]:
```
df.describe()
```

Out[245]:

	Number.of.Engines	Total.Fatal.Injuries	Total.Serious.Injuries	Total.Minor.Injuries	Tota
count	82805.000000	88889.000000	76379.000000	76956.000000	82
mean	1.146585	0.647855	0.279881	0.357061	
std	0.446510	5.122070	1.544084	2.235625	
min	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	0.000000	0.000000	0.000000	
50%	1.000000	0.000000	0.000000	0.000000	
75%	1.000000	0.647855	0.000000	0.000000	
max	8.000000	349.000000	161.000000	380.000000	

In [246]:
```
df.shape
```

Out[246]:
```
(88889, 31)
```

In [249]:
```
df.columns
```

Out[249]:
```
Index(['Event.Id', 'Investigation.Type', 'Accident.Number', 'Event.Dat
e',
       'Location', 'Country', 'Latitude', 'Longitude', 'Airport.Code',
       'Airport.Name', 'Injury.Severity', 'Aircraft.damage',
       'Aircraft.Category', 'Registration.Number', 'Make', 'Model',
       'Amateur.Built', 'Number.of.Engines', 'Engine.Type', 'FAR.Descrip
tion',
       'Schedule', 'Purpose.of.flight', 'Air.carrier', 'Total.Fatal.Inju
ries',
       'Total.Serious.Injuries', 'Total.Minor.Injuries', 'Total.Uninjure
d',
       'Weather.Condition', 'Broad.phase.of.flight', 'Report.Status',
       'Publication.Date'],
      dtype='object')
```

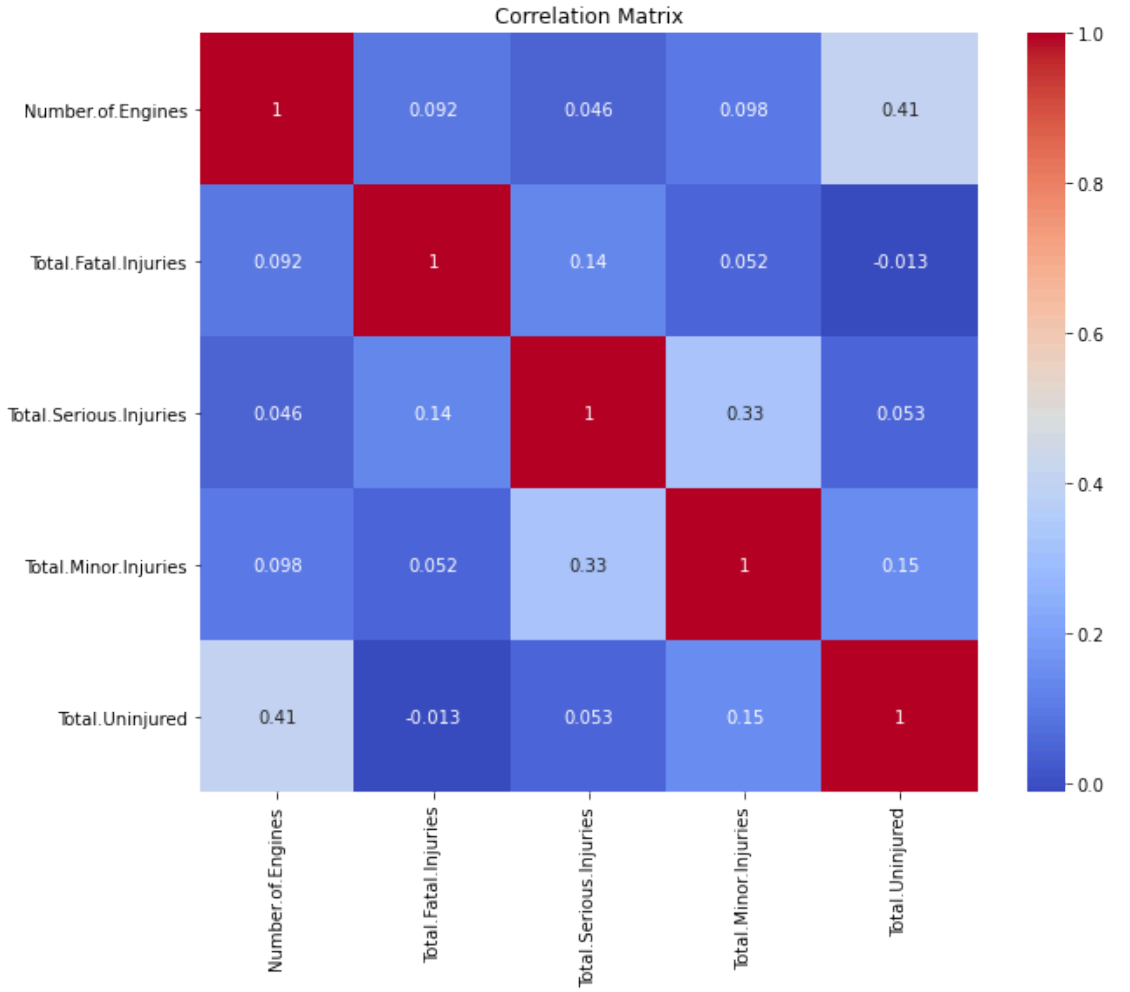
In [ ]:
```

```

In [250]:
```
#Explore the Data (EDA)
df.describe()
```

Out[250]:

	Number.of.Engines	Total.Fatal.Injuries	Total.Serious.Injuries	Total.Minor.Injuries	Tota
count	82805.000000	88889.000000	76379.000000	76956.000000	82
mean	1.146585	0.647855	0.279881	0.357061	
std	0.446510	5.122070	1.544084	2.235625	
min	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	0.000000	0.000000	0.000000	
50%	1.000000	0.000000	0.000000	0.000000	
75%	1.000000	0.647855	0.000000	0.000000	
max	8.000000	349.000000	161.000000	380.000000	

In [251]:
```
#Visualize Relationships using Correlation
plt.figure(figsize=(10, 8))
sns.heatmap(df.corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Matrix")
plt.show()
```



In [253]:
```
df.columns
```

Out[253]:
```
Index(['Event.Id', 'Investigation.Type', 'Accident.Number', 'Event.Dat
e',
       'Location', 'Country', 'Latitude', 'Longitude', 'Airport.Code',
       'Airport.Name', 'Injury.Severity', 'Aircraft.damage',
       'Aircraft.Category', 'Registration.Number', 'Make', 'Model',
       'Amateur.Built', 'Number.of.Engines', 'Engine.Type', 'FAR.Descrip
tion',
       'Schedule', 'Purpose.of.flight', 'Air.carrier', 'Total.Fatal.Inju
ries',
       'Total.Serious.Injuries', 'Total.Minor.Injuries', 'Total.Uninjure
d',
       'Weather.Condition', 'Broad.phase.of.flight', 'Report.Status',
       'Publication.Date'],
      dtype='object')
```

In [252]:
```
#Dropping columns that have more than 25% of the total data missing. This
columns_to_drop = ['Latitude', 'Longitude', 'Airport.Code', 'Airport.Name'
df_clean = df.drop(columns=columns_to_drop)
```

In [255]:
```
#Checking the columns after cleaning
df_clean.columns
```

Out[255]:
```
Index(['Event.Id', 'Investigation.Type', 'Event.Date', 'Location', 'Coun
try',
       'Injury.Severity', 'Aircraft.damage', 'Make', 'Model',
       'Number.of.Engines', 'Purpose.of.flight', 'Total.Fatal.Injuries',
       'Total.Serious.Injuries', 'Total.Minor.Injuries', 'Total.Uninjure
d',
       'Weather.Condition'],
      dtype='object')
```

In [256]:
```
#Check for duplicates
df.columns.duplicated()
```

Out[256]:
```
array([False, False, False, False, False, False, False, False, False,
       False, False, False, False, False, False, False, False, False,
       False, False, False, False, False, False, False, False, False,
       False, False, False, False])
```

In [257]:
```
#Further cleaning
columns_to_drop = ['Latitude', 'Longitude', 'Airport.Code', 'Airport.Name'
                   'FAR.Description', 'Schedule', 'Air.carrier', 'Broad.ph
                   'Report.Status', 'Amateur.Built', 'Accident.Number', 'R
                   'Publication.Date', 'Engine.Type']
df_clean = df.drop(columns=columns_to_drop)
print(df_clean.columns)
```

```
Index(['Event.Id', 'Investigation.Type', 'Event.Date', 'Location', 'Coun
try',
       'Injury.Severity', 'Aircraft.damage', 'Make', 'Model',
       'Number.of.Engines', 'Purpose.of.flight', 'Total.Fatal.Injuries',
       'Total.Serious.Injuries', 'Total.Minor.Injuries', 'Total.Uninjure
d',
       'Weather.Condition'],
      dtype='object')
```

In [258]:
```
df.isnull().sum()
```

Out[258]:
```
Event.Id                    0
Investigation.Type          0
Accident.Number             0
Event.Date                  0
Location                   52
Country                   226
Latitude                54507
Longitude               54516
Airport.Code            38640
Airport.Name            36099
Injury.Severity          1000
Aircraft.damage             0
Aircraft.Category       56602
Registration.Number      1317
Make                       63
Model                      92
Amateur.Built             102
Number.of.Engines        6084
Engine.Type              7077
FAR.Description         56866
Schedule                76307
Purpose.of.flight        6192
Air.carrier             72241
Total.Fatal.Injuries        0
Total.Serious.Injuries  12510
Total.Minor.Injuries    11933
Total.Uninjured          5912
Weather.Condition           0
Broad.phase.of.flight   27165
Report.Status            6381
Publication.Date        13771
dtype: int64
```

In [259]:
```
#Check the shape of the data
df_clean.shape
```

Out[259]:
```
(88889, 16)
```

In [260]:
```
df.head
```

Out[260]:

```
<bound method NDFrame.head of             Event.Id Investigation.Type Accident.Number  Event.Date  \
0      20001218X45444           Accident      SEA87LA080  1948-10-24
1      20001218X45447           Accident      LAX94LA336  1962-07-19
2      20061025X01555           Accident      NYC07LA005  1974-08-30
3      20001218X45448           Accident      LAX96LA321  1977-06-19
4      20041105X01764           Accident      CHI79FA064  1979-08-02
...               ...                ...             ...         ...
88884  20221227106491           Accident      ERA23LA093  2022-12-26
88885  20221227106494           Accident      ERA23LA095  2022-12-26
88886  20221227106497           Accident      WPR23LA075  2022-12-26
88887  20221227106498           Accident      WPR23LA076  2022-12-26
88888  20221230106513           Accident      ERA23LA097  2022-12-29

               Location        Country   Latitude   Longitude Airport.Code  \
0      MOOSE CREEK, ID  United States        NaN         NaN          NaN
1       BRIDGEPORT, CA  United States        NaN         NaN          NaN
2        Saltville, VA  United States  36.922223  -81.878056          NaN
3           EUREKA, CA  United States        NaN         NaN          NaN
4           Canton, OH  United States        NaN         NaN          NaN
...                ...            ...        ...         ...          ...
88884    Annapolis, MD  United States        NaN         NaN          NaN
88885      Hampton, NH  United States        NaN         NaN          NaN
88886       Payson, AZ  United States    341525N    1112021W           PA
88887       Morgan, UT  United States        NaN         NaN          NaN
88888       Athens, GA  United States        NaN         NaN          NaN

      Airport.Name  ... Purpose.of.flight        Air.carrier  \
0              NaN  ...          Personal                NaN
1              NaN  ...          Personal                NaN
2              NaN  ...          Personal                NaN
3              NaN  ...          Personal                NaN
4              NaN  ...          Personal                NaN
...            ...  ...               ...                ...
88884          NaN  ...          Personal                NaN
88885          NaN  ...               NaN                NaN
88886       PAYSON  ...          Personal                NaN
88887          NaN  ...          Personal  MC CESSNA 210N LLC
88888          NaN  ...          Personal                NaN

      Total.Fatal.Injuries Total.Serious.Injuries Total.Minor.Injuries
\
0                      2.0                    0.0                   0.0
1                      4.0                    0.0                   0.0
2                      3.0                    NaN                   NaN
```

```
3                    2.0                    0.0                    0.0
4                    1.0                    2.0                    NaN
...                  ...                    ...                    ...
88884                0.0                    1.0                    0.0
88885                0.0                    0.0                    0.0
88886                0.0                    0.0                    0.0
88887                0.0                    0.0                    0.0
88888                0.0                    1.0                    0.0

      Total.Uninjured Weather.Condition  Broad.phase.of.flight  \
0                 0.0               UNK                Cruise
1                 0.0               UNK              Unknown
2                 NaN               IMC                Cruise
3                 0.0               IMC                Cruise
4                 0.0               VMC              Approach
...               ...               ...                   ...
88884             0.0               VMC                   NaN
88885             0.0               VMC                   NaN
88886             1.0               VMC                   NaN
88887             0.0               VMC                   NaN
88888             1.0               VMC                   NaN

        Report.Status Publication.Date
0      Probable Cause              NaN
1      Probable Cause       19-09-1996
2      Probable Cause       26-02-2007
3      Probable Cause       12-09-2000
4      Probable Cause       16-04-1980
...               ...              ...
88884             NaN       29-12-2022
88885             NaN              NaN
88886             NaN       27-12-2022
88887             NaN              NaN
88888             NaN       30-12-2022

[88889 rows x 31 columns]>
```

In [265]:
```
print(df_clean.isnull().sum())
```

```
Event.Id                   0
Investigation.Type         0
Event.Date                 0
Location                  52
Country                  226
Injury.Severity         1000
Aircraft.damage            0
Make                      63
Model                     92
Number.of.Engines       6084
Purpose.of.flight       6192
Total.Fatal.Injuries       0
Total.Serious.Injuries  12510
Total.Minor.Injuries    11933
Total.Uninjured          5912
Weather.Condition           0
dtype: int64
```

In [266]:
```
df_clean.shape
```

Out[266]: (88889, 16)

In [267]:
```
#Replace Missing Values
categorical_columns = ['Location', 'Country', 'Aircraft.Category']
for col in categorical_columns:
    df[col].fillna(df[col].mode()[0], inplace=True)
```

In [276]:
```
numerical_columns = ['Latitude', 'Longitude', 'Total.Uninjured']
for col in numerical_columns:
    df[col] = pd.to_numeric(df[col], errors='coerce')
    if df[col].isnull().any():
        df[col].fillna(df[col].mean(), inplace=True)
```

In [270]:
```
columns_to_drop = ['FAR.Description', 'Schedule', 'Air.carrier']
df.drop(columns=columns_to_drop, inplace=True, errors='ignore')
```

In [271]:
```
columns_to_drop = ['Airport.Code', 'Airport.Name', 'Broad.phase.of.flight'
df.drop(columns=columns_to_drop, inplace=True, errors='ignore')
```

In [272]:
```
categorical_columns = ['Registration.Number', 'Make', 'Model', 'Amateur.Bu
for col in categorical_columns:
    if col in df.columns:
        df[col].fillna(df[col].mode()[0], inplace=True)
```

In [273]:
```
numerical_columns = ['Number.of.Engines', 'Engine.Type']
for col in numerical_columns:
    if col in df.columns:
        # Convert to numeric, coercing errors to NaN
        df[col] = pd.to_numeric(df[col], errors='coerce')
        # Fill missing values with the column mean
        df[col].fillna(df[col].mean(), inplace=True)
```

In [278]:
```
#Handling missing values based on column Type
df['Engine.Type'].fillna('Unknown', inplace=True)
df['Purpose.of.flight'].fillna(df['Purpose.of.flight'].mode()[0], inplace=
df['Report.Status'].fillna('Incomplete', inplace=True)
df['Injury.Severity'].fillna('Unknown', inplace=True)
df['Total.Serious.Injuries'].fillna(df['Total.Serious.Injuries'].mean(), i
df['Total.Minor.Injuries'].fillna(df['Total.Minor.Injuries'].mean(), inpla
```

In [290]:
```
df.isnull().sum()
```

Out[290]:
```
Event.Id                0
Investigation.Type      0
Accident.Number         0
Event.Date              0
Location                0
Country                 0
Latitude                0
Longitude               0
Injury.Severity         0
Aircraft.damage         0
Aircraft.Category       0
Registration.Number     0
Make                    0
Model                   0
Amateur.Built           0
Number.of.Engines       0
Engine.Type             0
Purpose.of.flight       0
Total.Fatal.Injuries    0
Total.Serious.Injuries  0
Total.Minor.Injuries    0
Total.Uninjured         0
Weather.Condition       0
Report.Status           0
dtype: int64
```

In [291]:
```
df.to_csv('cleaned_data.csv', index=False)
```

In [293]:
```
df.shape
```

Out[293]:
```
(88889, 24)
```

In [189]:
```
#Resolving Mixed Type Columns
print(df.dtypes)
```

```
Event.Id                  object
Investigation.Type        object
Accident.Number           object
Event.Date                object
Location                  object
Country                   object
Latitude                  object
Longitude                 object
Airport.Code              object
Airport.Name              object
Injury.Severity           object
Aircraft.damage           object
Aircraft.Category         object
Registration.Number       object
Make                      object
Model                     object
Amateur.Built             object
Number.of.Engines        float64
Engine.Type               object
FAR.Description           object
Schedule                  object
Purpose.of.flight         object
Air.carrier               object
Total.Fatal.Injuries     float64
Total.Serious.Injuries   float64
Total.Minor.Injuries     float64
Total.Uninjured          float64
Weather.Condition         object
Broad.phase.of.flight     object
Report.Status             object
Publication.Date          object
dtype: object
```

In [191]:
```
#Convert to a single type
df['Latitude'] = pd.to_numeric(df['Latitude'], errors='coerce')  # Convert
df['Longitude'] = pd.to_numeric(df['Longitude'], errors='coerce')
```

In [80]:
```
#The correlation between the five columns
df.corr()
```

Out[80]:

	Number.of.Engines	Total.Fatal.Injuries	Total.Serious.Injuries	Total.Minc
Number.of.Engines	1.000000	0.091553	0.046157	
Total.Fatal.Injuries	0.091553	1.000000	0.135099	
Total.Serious.Injuries	0.046157	0.135099	1.000000	
Total.Minor.Injuries	0.098162	0.051716	0.326849	
Total.Uninjured	0.406058	-0.012921	0.052869	
Event.Year	-0.018393	0.017132	0.033246	

In [284]:
```
#Summary and Exploration of Data
df.nunique()
```

Out[284]:
```
Event.Id                87951
Investigation.Type          2
Accident.Number         88863
Event.Date              14782
Location                27758
Country                   219
Latitude                 8879
Longitude                9272
Injury.Severity           110
Aircraft.damage             4
Aircraft.Category          15
Registration.Number     79105
Make                     8237
Model                   12318
Amateur.Built               2
Number.of.Engines           8
Engine.Type                 1
Purpose.of.flight          26
Total.Fatal.Injuries      126
Total.Serious.Injuries     51
Total.Minor.Injuries       58
Total.Uninjured           380
Weather.Condition           4
Report.Status           17076
dtype: int64
```

In [97]:
```
#Summary statistics for numerical columns
df.describe
```

Out[97]:

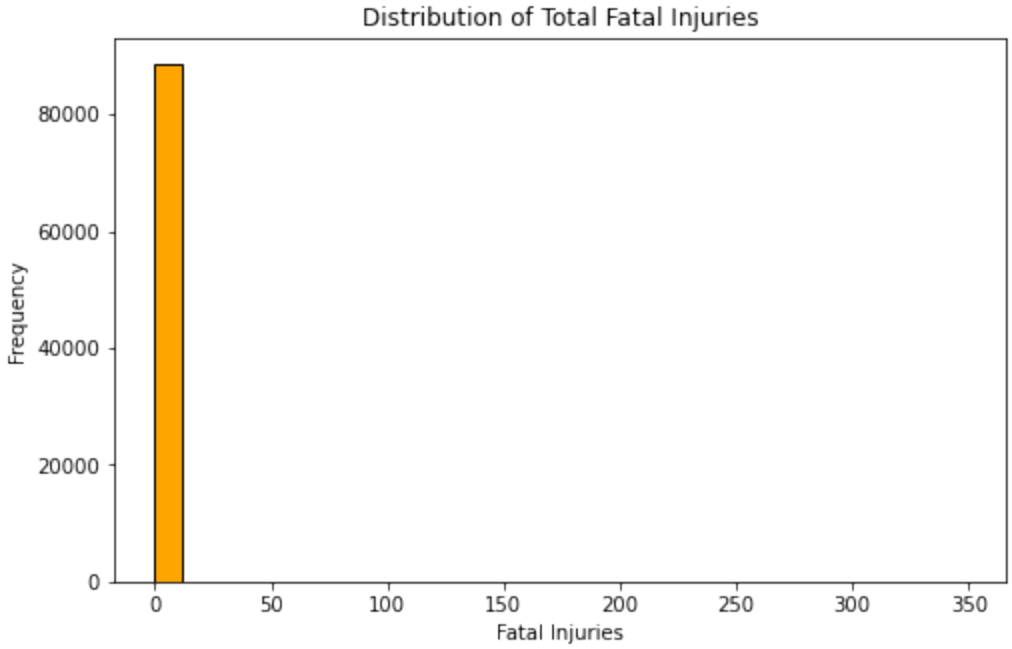
```
<bound method NDFrame.describe of             Event.Id Investigation.Type Accident.Number  Event.Date  \
0      20001218X45444           Accident      SEA87LA080  1948-10-24
1      20001218X45447           Accident      LAX94LA336  1962-07-19
2      20061025X01555           Accident      NYC07LA005  1974-08-30
3      20001218X45448           Accident      LAX96LA321  1977-06-19
4      20041105X01764           Accident      CHI79FA064  1979-08-02
...               ...                ...             ...         ...
88884  20221227106491           Accident      ERA23LA093  2022-12-26
88885  20221227106494           Accident      ERA23LA095  2022-12-26
88886  20221227106497           Accident      WPR23LA075  2022-12-26
88887  20221227106498           Accident      WPR23LA076  2022-12-26
88888  20221230106513           Accident      ERA23LA097  2022-12-29

               Location        Country   Latitude   Longitude Airport.Code  \
0      MOOSE CREEK, ID  United States        NaN         NaN          NaN
1       BRIDGEPORT, CA  United States        NaN         NaN          NaN
2        Saltville, VA  United States  36.922223  -81.878056          NaN
3           EUREKA, CA  United States        NaN         NaN          NaN
4           Canton, OH  United States        NaN         NaN          NaN
...                ...            ...        ...         ...          ...
88884    Annapolis, MD  United States        NaN         NaN          NaN
88885      Hampton, NH  United States        NaN         NaN          NaN
88886       Payson, AZ  United States    341525N    1112021W          PA
88887       Morgan, UT  United States        NaN         NaN          NaN
88888       Athens, GA  United States        NaN         NaN          NaN

      Airport.Name  ...         Air.carrier Total.Fatal.Injuries  \
0              NaN  ...                 NaN                 2.0
1              NaN  ...                 NaN                 4.0
2              NaN  ...                 NaN                 3.0
3              NaN  ...                 NaN                 2.0
4              NaN  ...                 NaN                 1.0
...            ...  ...                 ...                 ...
88884          NaN  ...                 NaN                 0.0
88885          NaN  ...                 NaN                 0.0
88886       PAYSON  ...                 NaN                 0.0
88887          NaN  ...  MC CESSNA 210N LLC                 0.0
88888          NaN  ...                 NaN                 0.0

      Total.Serious.Injuries Total.Minor.Injuries Total.Uninjured  \
0                        0.0                  0.0             0.0
1                        0.0                  0.0             0.0
2                        NaN                  NaN             NaN
3                        0.0                  0.0             0.0
```

```
4                      2.0                  NaN             0.0
...                    ...                  ...             ...
88884                  1.0                  0.0             0.0
88885                  0.0                  0.0             0.0
88886                  0.0                  0.0             1.0
88887                  0.0                  0.0             0.0
88888                  1.0                  0.0             1.0

      Weather.Condition Broad.phase.of.flight   Report.Status  \
0                   UNK                Cruise  Probable Cause
1                   UNK               Unknown  Probable Cause
2                   IMC                Cruise  Probable Cause
3                   IMC                Cruise  Probable Cause
4                   VMC              Approach  Probable Cause
...                 ...                   ...             ...
88884               VMC                   NaN             NaN
88885               VMC                   NaN             NaN
88886               VMC                   NaN             NaN
88887               VMC                   NaN             NaN
88888               VMC                   NaN             NaN

      Publication.Date Event.Year
0                  NaN       1948
1           19-09-1996       1962
2           26-02-2007       1974
3           12-09-2000       1977
4           16-04-1980       1979
...                ...        ...
88884       29-12-2022       2022
88885              NaN       2022
88886       27-12-2022       2022
88887              NaN       2022
88888       30-12-2022       2022

[88889 rows x 32 columns]>
```

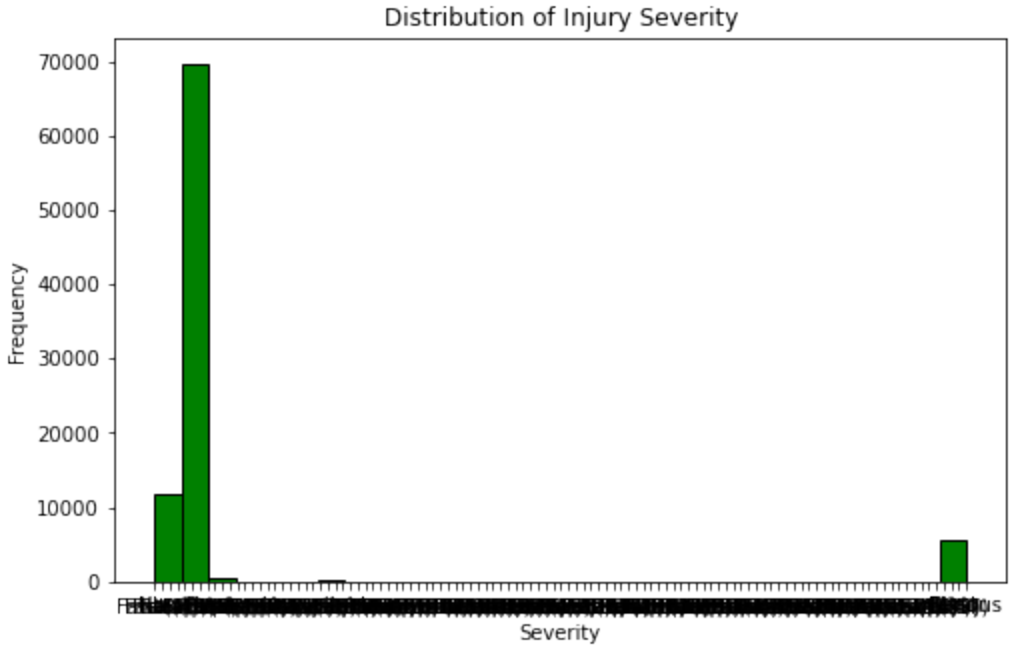
In [288]:
```
#Frequency distribution for Categorical variables
df['Weather.Condition'].value_counts()
df['Purpose.of.flight'].value_counts()
```

Out[288]:
```
Personal                    55640
Instructional               10601
Unknown                      6802
Aerial Application           4712
Business                     4018
Positioning                  1646
Other Work Use               1264
Ferry                         812
Aerial Observation            794
Public Aircraft               720
Executive/corporate           553
Flight Test                   405
Skydiving                     182
External Load                 123
Public Aircraft - Federal     105
Banner Tow                    101
Air Race show                  99
Public Aircraft - Local        74
Public Aircraft - State        64
Air Race/show                  59
Glider Tow                     53
Firefighting                   40
Air Drop                       11
ASHO                            6
PUBS                            4
PUBL                            1
Name: Purpose.of.flight, dtype: int64
```

In [58]:
```
#Histogram for Total Fatal Injuries
plt.figure(figsize=(8, 5))
df['Total.Fatal.Injuries'].hist(bins=30, color='orange', edgecolor='black'
plt.title("Distribution of Total Fatal Injuries")
plt.xlabel("Fatal Injuries")
plt.ylabel("Frequency")
plt.grid(False)
plt.show()
```

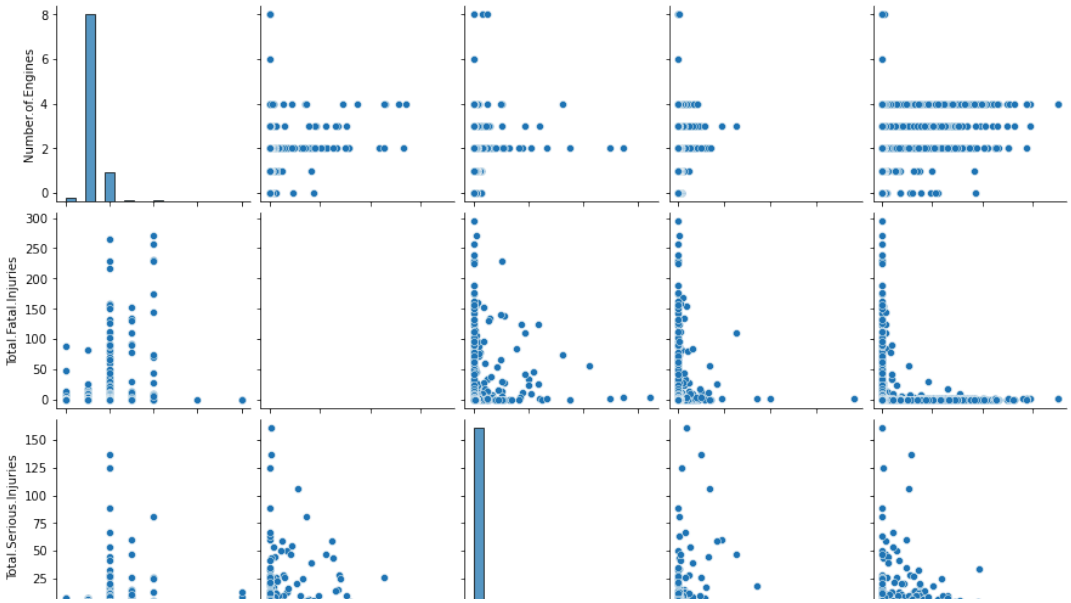


In [59]:
```
#Histogram for Injury Severity
plt.figure(figsize=(8, 5))
df['Injury.Severity'].hist(bins=30, color='green', edgecolor='black')
plt.title("Distribution of Injury Severity")
plt.xlabel("Severity")
plt.ylabel("Frequency")
plt.grid(False)
plt.show()
```

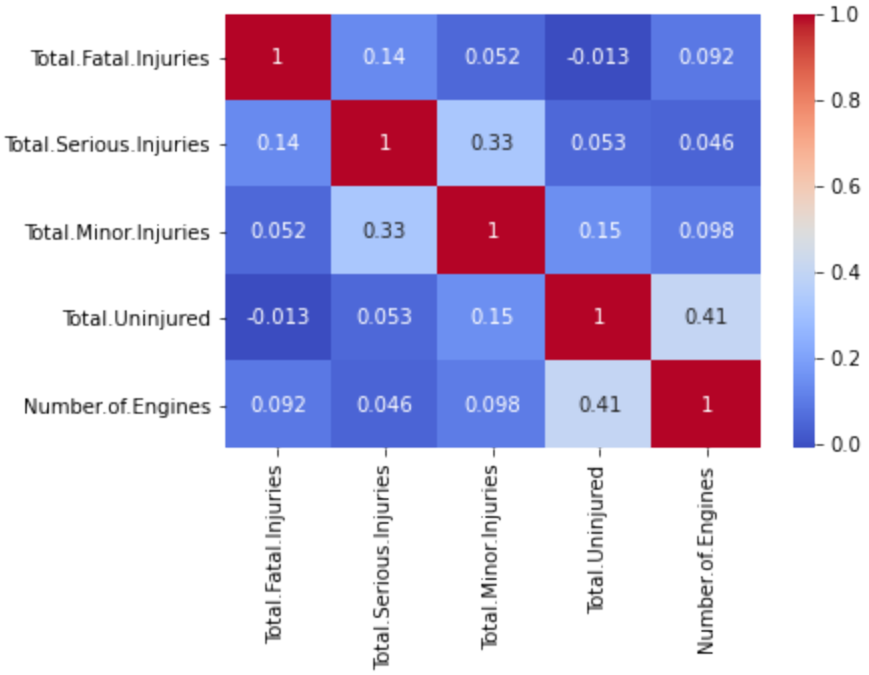


In [75]:
```
sns.pairplot(df)
```

Out[75]: <seaborn.axisgrid.PairGrid at 0x22b190f35b0>



In [101]:
```
#Relationships between injury counts and other numerical columns
correlation_matrix = df[['Total.Fatal.Injuries', 'Total.Serious.Injuries',
                         'Total.Minor.Injuries', 'Total.Uninjured',
                         'Number.of.Engines']].corr()
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
plt.show()
```



In [102]:
```
#Analysis of Injury Severity by Weather Conditions
df.groupby('Weather.Condition')[['Total.Fatal.Injuries', 'Total.Serious.Ir
```

Out[102]:

Weather.Condition	Total.Fatal.Injuries	Total.Serious.Injuries
IMC	2.020644	0.419777
UNK	2.886843	0.255405
Unk	1.244275	0.500000
VMC	0.522216	0.269129

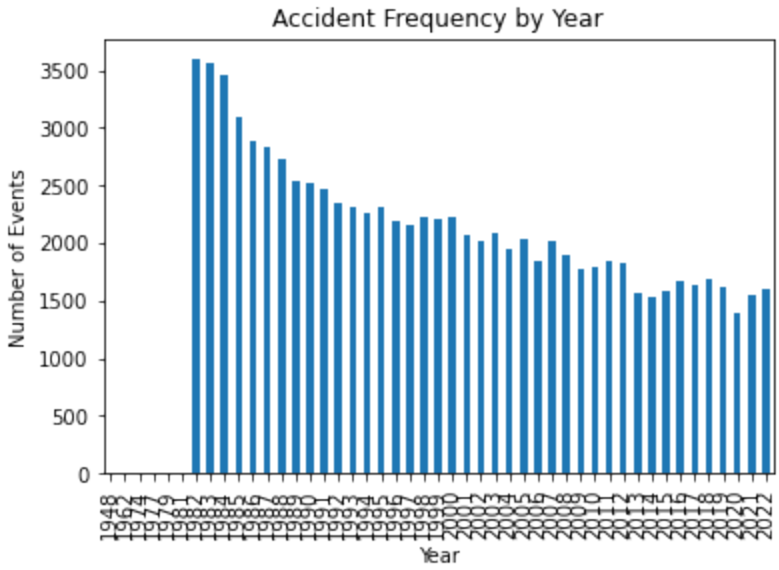
In [103]:
```
#Check if certain makes or models have higher fatality rates
df.groupby('Make')[['Total.Fatal.Injuries', 'Total.Serious.Injuries']].mea
```

Out[103]:

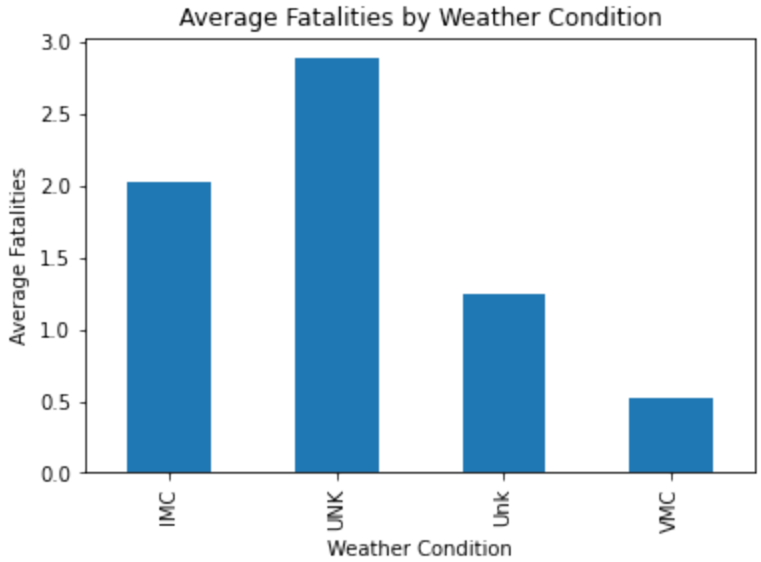
Make	Total.Fatal.Injuries	Total.Serious.Injuries
Tupolev	210.000000	NaN
TUPOLEV	44.500000	0.0
VIKING AIR LIMITED	23.000000	0.0
Aviocar CASA	18.000000	0.0
SUKHOI	14.333333	0.0
...	...	...
Honda Jet	0.000000	0.0
Honda Aircraft	0.000000	0.0
Honda	0.000000	0.0
Homer Davis	0.000000	0.0
unknown	0.000000	0.0

8237 rows × 2 columns

In [105]:
```
#Visualize Insights
#Histogram for Yearly Trends
df['Event.Year'].value_counts().sort_index().plot(kind='bar', title='Accid
plt.xlabel('Year')
plt.ylabel('Number of Events')
plt.show()
```



In [106]:
```
#Bar Chart for Weather Impact on Severity
df.groupby('Weather.Condition')['Total.Fatal.Injuries'].mean().plot(kind='
plt.xlabel('Weather Condition')
plt.ylabel('Average Fatalities')
plt.show()
```



In [107]:
```
#Scatter Plot for Engines vs. Fatalities
plt.scatter(df['Number.of.Engines'], df['Total.Fatal.Injuries'])
plt.title('Number of Engines vs. Total Fatal Injuries')
plt.xlabel('Number of Engines')
plt.ylabel('Total Fatal Injuries')
plt.show()
```