


```
[38]: #drop all damage values for the Make column
aircraft_df.Make = aircraft_df.Make.replace(['NONE', 'UNKNOWN', 'NaN'], 'NA')

[39]: #do the same thing for the Model column
aircraft_df.Model = aircraft_df.Model.replace(['NONE', 'UNKNOWN', 'NaN'], 'NA')

In [40]: #create a list of the top 10 aircraft makes and calculate their sum of appearances
top_10_makes_counts = aircraft_df.Make.value_counts().head(10).reset_index()
print(sum(count for value, count in top_10_makes_counts))

[[('CESSNA', 27149), ('PIPER', 14790), ('BEECH', 5372), ('BOEING', 2745), ('BELL', 272
2), ('MOONEY', 1334), ('ROBINSON', 1230), ('GRUMMAN', 1172), ('BELLANCA', 1045), ('HUG
BONINCA', 922)]

In [41]: #do the same thing for the Model column, including the corresponding Make value
model_counts = aircraft_df['Model'].value_counts().head(10).reset_index()
model_counts.columns = ['Model', 'Count']

result_df = pd.merge(model_counts, make_counts, on='Model')
top_10_model_counts_merged = result_df[['Make', 'Model', 'Count']].values.tolist()
print(top_10_model_counts_merged)

[('CESSNA', 152), ('PIPER', 14790), ('CESSNA', 172), ('BEECH', 5372), ('BOEING', 2745), ('BELL', 272
2), ('MOONEY', 1334), ('ROBINSON', 1230), ('GRUMMAN', 1172), ('BELLANCA', 1045), ('HUG
BONINCA', 922)]

In [42]: #check which columns still have empty values
columns_with_nan = aircraft_df.columns[aircraft_df.isna().any()]
columns_with_nan

Out[42]: Index(['Location', 'Country', 'Airport Code', 'Injury_Severity',
'Aircraft_Damage', 'Aircraft_Category', 'Registration_Number',
'Amateur_Built', 'Number_of_Engines', 'Engine_Type', 'FAR_Description',
'Schedule', 'Purpose_of_Flight', 'Total_Fatal_Injuries',
'Total_Serious_Injuries', 'Total_Minor_Injuries', 'Total_Uninjured',
'Weather_Condition', 'Report_Status', 'Publication_Date',
'dtype='object']

In [43]: #check Location value counts
aircraft_df.Location.value_counts(dropna=False)

Out[43]: Location
ANCHORAGE, AK    434
MIAMI, FL        200
ALBUQUERQUE, NM   196
HOUSTON, TX       193
CHICAGO, IL       186
...
ST. JOSEPH, LA    1
WINDSOR, ONTARIO, ON 1
ELK GARDEN, VA    1
PIURA, Peru      1
BRASOLETE, ...    1
Name: count, Length: 2735, dtype: int64

In [44]: #check Country value counts
aircraft_df.Country.value_counts(dropna=False)

Out[44]: Country
United States    82227
Brazil           371
Mexico           358
Canada           357
United Kingdom  244
...
Palau             1
Saint Vincent and the Grenadines 1
Malampa          1
Turks and Caicos Islands 1
Name: count, Length: 220, dtype: int64

In [45]: #check Airport Code value counts
aircraft_df.Airport_Code.value_counts(dropna=False)

Out[45]: Airport_Code
NaN      38708
WNSW     1487
PVT       485
APA       160
ORF       149
...
TM95      1
TOS       1
3M8       1
4C03      1
ERKH      1
Name: count, Length: 10372, dtype: int64

In [46]: #fill empty Airport Code entries with "NONE"
aircraft_df.Airport_Code = aircraft_df.Airport_Code.fillna("NONE")
aircraft_df.Airport_Code.value_counts(dropna=False)

Out[46]: Airport_Code
WNSW     1487
PVT       485
APA       160
ORF       149
...
TM95      1
TOS       1
3M8       1
4C03      1
ERKH      1
Name: count, Length: 10371, dtype: int64

In [47]: #check Injury Severity value counts
aircraft_df.Injury_Severity.value_counts(dropna=False)

Out[47]: Injury_Severity
Non-Fatal      67332
Fatal          5253
FATAL (217)    3708
Incident       2216
...
Fatal (80)      1
Fatal (217)     1
Fatal (169)     1
Fatal (88)      1
Fatal (189)     1
Name: count, Length: 110, dtype: int64

In [48]: #check Weather Condition value counts
aircraft_df.Weather_Condition.value_counts(dropna=False)

Out[48]: Weather_Condition
Non-Fatal      67332
Fatal          5253
FATAL (217)    3708
Incident       2216
...
Fatal (80)      1
Fatal (217)     1
Fatal (169)     1
Fatal (88)      1
Fatal (189)     1
Name: count, Length: 110, dtype: int64

In [49]: #capitalise Unk in Weather Condition column
aircraft_df.Location[aircraft_df.Weather_Condition == 'Unk', 'Weather_Condition'] = 'UNK'
#fill empty values with UNK
aircraft_df.Weather_Condition.fillna("UNK", inplace=True)
aircraft_df.Weather_Condition.value_counts(dropna=False)

Out[49]: Weather_Condition
VNC      17282
IMC       5973
UNK       5571
Name: count, dtype: int64

In [50]: #check the value counts of Amateur_Built
aircraft_df.Amateur_Built.value_counts(dropna=False)

Out[50]: Amateur_Built
No      80266
Yes     8460
NaN     100
Name: count, dtype: int64

In [51]: #check the value count of Number_of_Engines
aircraft_df.Number_of_Engines.value_counts(dropna=False)

Out[51]: Number_of_Engines
1.0    6569
2.0    11078
NaN     6035
10.0    1226
3.0     483
4.0     431
8.0      3
6.0      1
Name: count, dtype: int64

In [52]: #check the value count of Engine_Type
aircraft_df.Engine_Type.value_counts(dropna=False)

Out[52]: Engine_Type
Reciprocating    69519
NaN              7045
Turbo Shaft      3609
Turbo Prop       3391
Turbo Fan        2481
Unknown          2050
Turbo Jet        703
Geared Turbopan 12
Electric         10
LPK              2
WOW             1
Hybrid Rocket    1
UNK              1
Name: count, dtype: int64

In [53]: #replace UNK and NaN with Unknown in Engine_Type column
aircraft_df.Location[aircraft_df.Engine_Type == 'UNK', 'Engine_Type'] = 'Unknown'
aircraft_df.Engine_Type = aircraft_df.Engine_Type.fillna("Unknown")
aircraft_df.Engine_Type.value_counts(dropna=False)

Out[53]: Engine_Type
Reciprocating    69519
Unknown          9096
Turbo Shaft      3609
Turbo Prop       3391
Turbo Fan        2481
Unknown          2050
Turbo Jet        703
Geared Turbopan 12
Electric         10
LPK              2
WOW             1
Hybrid Rocket    1
UNK              1
Name: count, dtype: int64

In [54]: #rename Aircraft_Damage to Aircraft_Damage and check the value counts of Aircraft_Damage
aircraft_df = aircraft_df.rename(columns={'Aircraft_Damage': 'Aircraft_Damage'})
aircraft_df.Aircraft_Damage.value_counts(dropna=False)

Out[54]: Aircraft_Damage
Substantial    64124
Destroyed      18612
NaN            3176
Minor          2795
Unknown        119
Name: count, dtype: int64

In [55]: #fill empty values in Aircraft_Damage with Unknown
aircraft_df.Aircraft_Damage = aircraft_df.Aircraft_Damage.fillna("Unknown")
aircraft_df.Aircraft_Damage.value_counts(dropna=False)

Out[55]: Aircraft_Damage
Substantial    64124
Destroyed      18612
NaN            3176
Minor          2795
Unknown        119
Name: count, dtype: int64

In [56]: #check the value counts for Aircraft_Category
aircraft_df.Aircraft_Category.value_counts(dropna=False)

Out[56]: Aircraft_Category
NaN      56551
Airplane 27608
Helicopter 3437
Glider    508
Balloon   231
Gyrocraft 173
Weight-Shift 161
Powered Parachute 91
ULTRALIGHT 30
Unknown    18
WSFT       9
Powered-Lift 5
Blimp      4
UNK        2
Rocket     1
Ultralight 1
Name: count, dtype: int64

In [57]: #replace UNK and NaN with Unknown in Aircraft_Category
aircraft_df.Location[aircraft_df.Aircraft_Category == 'UNK', 'Aircraft_Category'] = 'Unknown'
aircraft_df.Aircraft_Category = aircraft_df.Aircraft_Category.fillna("Unknown")
aircraft_df.Aircraft_Category.value_counts(dropna=False)

Out[57]: Aircraft_Category
Unknown      56567
Airplane     27608
Helicopter    3437
Glider        508
Balloon       231
Gyrocraft     173
Weight-Shift  161
Powered Parachute 91
ULTRALIGHT    30
Unknown       18
WSFT           9
Powered-Lift  5
Blimp          4
Rocket         1
Ultralight     1
Name: count, dtype: int64

In [58]: #drop all rows except ones in which the Aircraft_Category is either Unknown, Airplane, Helicopter
aircraft_categories = ['Unknown', 'Airplane', 'Helicopter']
aircraft_df = aircraft_df[aircraft_df.Aircraft_Category.isin(aircraft_categories)].reset_index()
aircraft_df.Aircraft_Category.value_counts(dropna=False)

Out[58]: Aircraft_Category
Unknown      56567
Airplane     27608
Helicopter    3437
Name: count, dtype: int64

In [59]: #check value counts of Schedule
aircraft_df.Schedule.value_counts(dropna=False)

Out[59]: Schedule
NaN      75091
NSCH     4443
UNK      4099
SCHD     3979
Name: count, dtype: int64

In [60]: #fill NaN in Schedule column with UNK
aircraft_df.Schedule = aircraft_df.Schedule.fillna("UNK")
aircraft_df.Schedule.value_counts(dropna=False)

Out[60]: Schedule
UNK      79130
NSCH     4443
UNK      4099
SCHD     3979
Name: count, dtype: int64

In [61]: #check value counts of FAR_Description
aircraft_df.FAR_Description.value_counts(dropna=False)

Out[61]: FAR_Description
NaN      56848
091      17288
91: General Aviation 1573
NUSN     1010
137       989
135       746
121       674
Part 137: Agricultural 437
UNK       364
Part 135: Air Taxi & Commuter 298
PUBU      251
129       244
Part 121: Air Carrier 165
133       107
Part 129: Foreign 96
Non-U.S., Non-Commercial 91
Non-U.S., Commercial 91
Part 133: Rotorcraft Ext. Load 32
Unknown   18
Public Use 14
091K      14
ARMP       7
125        5
Part 125: 20+ Pax,6000+ lbs 5
107        4
Public Aircraft 2
Armed Forces 1
Part 91F: Special Flt Ops. 1
Part 91 Subpart K: Fractional 1
Name: count, dtype: int64

In [62]: #fill empty values in FAR_Description with Unknown
aircraft_df.FAR_Description = aircraft_df.FAR_Description.fillna("Unknown")
aircraft_df.FAR_Description.value_counts(dropna=False)

Out[62]: FAR_Description
Unknown      56870
091      17288
91: General Aviation 1573
NUSN     1010
137       989
135       746
121       674
Part 137: Agricultural 437
UNK       364
Part 135: Air Taxi & Commuter 298
PUBU      251
129       244
Part 121: Air Carrier 165
133       107
Part 129: Foreign 96
Non-U.S., Non-Commercial 91
Non-U.S., Commercial 91
Part 133: Rotorcraft Ext. Load 32
Public Use 14
091K      14
ARMP       7
125        5
Part 125: 20+ Pax,6000+ lbs 5
107        4
Public Aircraft 2
Armed Forces 1
Part 91F: Special Flt Ops. 1
Part 91 Subpart K: Fractional 1
Name: count, dtype: int64

In [63]: #check the value counts in Purpose_of_flight
aircraft_df.Purpose_of_flight.value_counts(dropna=False)

Out[63]: Purpose_of_flight
Personal      48555
Instructional 10473
Unknown      6777
Aerial Application 4712
Business      3986
Positioning    1644
Other Work Use 1207
Ferry          812
Aerial Observation 768
Public Aircraft 719
Executive/corporate 553
Flight Test    389
Skydiving      181
External Load  122
Public Aircraft - Federal 101
Banner Tow     105
Air Race/show  78
Public Aircraft - Local 64
Air Race/show  56
Glider Tow     40
Firefighting   42
Air Drop       11
ASRO           4
PUBS           1
PUBL          1
Name: count, dtype: int64

In [64]: #fill empty fields on Purpose_of_flight with Unknown
aircraft_df.Purpose_of_flight = aircraft_df.Purpose_of_flight.fillna("Unknown")
aircraft_df.Purpose_of_flight.value_counts(dropna=False)

Out[64]: Purpose_of_flight
Personal      48555
Unknown      12910
Instructional 10473
Aerial Application 4712
Business      3986
Positioning    1644
Other Work Use 1207
Ferry          812
Aerial Observation 768
Public Aircraft 719
Executive/corporate 553
Flight Test    389
Skydiving      181
External Load  122
Public Aircraft - Federal 101
Banner Tow     105
Air Race/show  78
Public Aircraft - Local 64
Air Race/show  56
Glider Tow     40
Firefighting   42
Air Drop       11
ASRO           4
PUBS           1
PUBL          1
Name: count, dtype: int64

In [65]: #make all entries containing Public Aircraft in Purpose_of_flight column one spelling
aircraft_df.Location[aircraft_df.Purpose_of_flight.str.contains('Public Aircraft', na=False)]
aircraft_df.Purpose_of_flight.value_counts(dropna=False)

Out[65]: Purpose_of_flight
Personal      48555
Unknown      12910
Instructional 10473
Aerial Application 4712
Business      3986
Positioning    1644
Other Work Use 1207
Ferry          812
Aerial Observation 768
Executive/corporate 553
Flight Test    389
Skydiving      181
External Load  122
Public Aircraft - Federal 101
Banner Tow     105
Air Race/show  78
Public Aircraft - Local 64
Air Race/show  56
Glider Tow     40
Firefighting   42
Air Drop       11
ASRO           4
PUBS           1
PUBL          1
Name: count, dtype: int64

In [66]: #drop all rows except ones in which the Purpose_of_flight is either Personal, Instru
Unknown, Aerial Application, Business, Positioning, Other Work Use, Ferry, Aerial Ob
purpose_categories = ['Personal', 'Unknown', 'Instructional', 'Aerial Application', '
aircraft_df = aircraft_df[aircraft_df.Purpose_of_flight.isin(purpose_categories)].reset_index()
aircraft_df.Purpose_of_flight.value_counts(dropna=False)

Out[66]: Purpose_of_flight
Personal      48555
Unknown      12910
Instructional 10473
Aerial Application 4712
Business      3986
Positioning    1644
Other Work Use 1207
Ferry          812
Aerial Observation 768
Executive/corporate 553
Flight Test    389
Skydiving      181
External Load  122
Public Aircraft - Federal 101
Banner Tow     105
Air Race/show  78
Public Aircraft - Local 64
Air Race/show  56
Glider Tow     40
Firefighting   42
Air Drop       11
ASRO           4
PUBS           1
PUBL          1
Name: count, dtype: int64

In [67]: #look at the most common Make values
aircraft_df.Make.value_counts()

Out[67]: Make
CESSNA      26978
PIPER       14790
BEECH       5351
BOEING      2735
BELL        2678
...
HAZELRIGE   1
MASKO       1
FRANK H MARCHETTI 1
SHANKLIN    1
ROYSE RALPH L 1
Name: count, Length: 7196, dtype: int64

In [68]: #calculate the fatalities for CESSNA accidents
cessna_fatal_injuries = aircraft_df.loc[aircraft_df.Make == 'CESSNA', 'Total_Fatal_In
cessna_fatal_injuries

Out[68]: 9518.0

In [69]: #check the sum of Total_Fatal_Injuries of the entire dataframe
aircraft_df.Total_Fatal_Injuries.sum()

Out[69]: 49418.0

In [70]: #make a series of the top 5 makes and their occurrences
top_5_makes_counts = aircraft_df.Make.value_counts().head(5)
top_5_makes_counts

Out[70]: Make
CESSNA      26978
PIPER       14790
BEECH       5351
BELL        2678
Name: count, dtype: int64

In [71]: #create a table of the top 5 for Makes, Accidents, and Total_Fatal_Injuries
merged_make_fatal_df = pd.DataFrame({'Make': top_5_makes_counts.index, 'Count': top_5
merged_make_fatal_df = merged_make_fatal_df.merge(make_fatalities, on='Make')
merged_make_fatal_df = merged_make_fatal_df.rename(columns={'Total_Fatal_Injuries_x':

Out[71]:
   Make      Accidents  Total_Fatal_Injuries
0  CESSNA      26978          9518.0
1  PIPER      14790          6673.0
2  BEECH      5351          3708.0
3  BOEING      2735          8738.0
4  BELL       2678          1301.0

In [72]: #create a series of the top 5 total fatal injuries by make
aircraft_df.Total_Fatal_Injuries = aircraft_df.groupby('Make')['Total_Fatal_Injuries'].sum().sort
total_fatal_injuries_df = pd.DataFrame({'Make': total_fatal_injuries.index, 'Total_Fa
total_fatal_injuries_df = total_fatal_injuries_df.head(5)
total_fatal_injuries_df

Out[72]:
   Make  Total_Fatal_Injuries
0  CESSNA          9518.0
1  BOEING          8738.0
2  PIPER           6673.0
3  BEECH           3708.0
4  AIRBUS          1325.0

In [73]: #plot a double bar chart to represent the top 5 makes to appear and their total fatal
injuries

#set bar width = 0.35
bar_width = 0.35

#set the index
index = np.arange(len(merged_make_fatal_df['Make']))

#plot the bar chart
plt.bar(index, merged_make_fatal_df['Accidents'], bar_width, label='Accidents')
plt.bar(index + bar_width, merged_make_fatal_df['Total_Fatal_Injuries'], bar_width, la
plt.xlabel('Make')
plt.ylabel('Count')
plt.title('Accidents and Total Fatal Injuries by Make')

#adjust the x ticks
plt.xticks(index + bar_width/2, merged_make_fatal_df['Make'])

#create a legend and show the graph
plt.legend()
plt.show()

Out[73]:
Figure: Accidents and Total Fatal Injuries by Make
A bar chart with two series: 'Accidents' (blue bars) and 'Total Fatal Injuries' (orange bars). The x-axis represents the 'Make' of the aircraft, and the y-axis represents the 'Count'. The data is as follows:


| Make   | Accidents | Total Fatal Injuries |
|--------|-----------|----------------------|
| CESSNA | 26978     | 9518.0               |
| PIPER  | 14790     | 6673.0               |
| BEECH  | 5351      | 3708.0               |
| BOEING | 2735      | 8738.0               |
| BELL   | 2678      | 1301.0               |



In [74]: # Create a figure with two subplots
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(10, 5))

# Plot the first bar graph for 'Accidents/Incidents'
ax1.bar(merged_make_fatal_df['Make'], merged_make_fatal_df['Accidents'])
ax1.set_xlabel('Make')
ax1.set_ylabel('Accidents and Incidents')
ax1.set_title('Top 5 Accidents and Incidents by Make')

# Plot the second bar graph for 'Total Fatal Injuries'
ax2.bar(merged_make_fatal_df['Make'], merged_make_fatal_df['Total_Fatal_Injuries'])
ax2.set_xlabel('Make')
ax2.set_ylabel('Total Fatal Injuries')
ax2.set_title('Top 5 Total Fatal Injuries by Make')

# Adjust the spacing between subplots
plt.tight_layout()

# Display the plots
plt.show()

Out[74]:
Figure: Top 5 Accidents and Incidents by Make, Top 5 Total Fatal Injuries by Make
Two side-by-side bar charts. The left chart shows 'Top 5 Accidents and Incidents by Make' with blue bars. The right chart shows 'Top 5 Total Fatal Injuries by Make' with orange bars. The data is as follows:


| Make   | Accidents | Total Fatal Injuries |
|--------|-----------|----------------------|
| CESSNA | 26978     | 9518.0               |
| BOEING | 2735      | 8738.0               |
| PIPER  | 14790     | 6673.0               |
| BEECH  | 5351      | 3708.0               |
| AIRBUS | 1325.0    | 1325.0               |



In [75]: #change the Event_Date column to datetimelike values
aircraft_df['Event_Date'] = pd.to_datetime(aircraft_df['Event_Date'])
aircraft_df = aircraft_df.groupby('Event_Date').dt.year
aircraft_df['Year']

Out[75]:
0      1948
1      1962
2      1974
3      1977
4      1979
...
86963    2022
86964    2022
86965    2022
86966    2022
86967    2022
Name: Year, Length: 86970, dtype: int32

In [76]: #group the data by Year, Weather_Condition, Aircraft_Damage, Total_Fatal_Injuries
injured_grouped_df = aircraft_df.groupby(['Year', 'Weather_Condition', 'Aircraft_Damage
injured_grouped_df = injured_grouped_df.reset_index(drop=True)

#convert the year to an integer type
injured_grouped_df.Year = injured_grouped_df.Year.astype(int)

#filter out entries prior to 1980
injured_grouped_df = injured_grouped_df[injured_grouped_df.Year >= 1980]

#create another column Total_Injuries that combines the three injuries columns
injured_grouped_df['Total_Injuries'] = injured_grouped_df[['Total_Fatal_Injuries', 'To
injured_grouped_df

Out[76]:
   Year Weather_Condition Aircraft_Damage  Total_Fatal_Injuries  Total_Serious_Injuries  Total_Minor
0    1981              IMC      Destroyed                4.0                0.0
1    1982              IMC      Destroyed               427.0                65.0
2    1982              IMC      Substantial                18.0                1.0
3    1982              IMC      Unknown                  2.0                0.0
...    ...              ...      ...                    ...                    ...
472   2022              UNK      Unknown                  53.0                39.0
473   2022              VMC      Destroyed               106.0                5.0
474   2022              VMC      Minor                   2.0                1.0
475   2022              VMC      Substantial              123.0               227.0
476   2022              VMC      Unknown                  9.0                9.0
477 rows x 8 columns

In [77]: #Filter the DataFrame to exclude Unknown from Aircraft_Damage
injured_grouped_filtered_df = injured_grouped_df[injured_grouped_df['Aircraft_Damage']
injured_grouped_filtered_df.Aircraft_Damage.value_counts(dropna=False)

Out[77]: Aircraft_Damage
Destroyed    124
Substantial  123
Minor       114
Name: count, dtype: int64

In [78]: #create a plot
plt.figure(figsize=(10, 6))

#plot a line graph of Total Uninjured People by Year in Accidents
sns.lineplot(data=injured_grouped_df, x='Year', y='Total Uninjured', ci=None)

#customize the labels and title
plt.xlabel('Year')
plt.ylabel('Number of Uninjured')
plt.title('Number of Uninjured People Involved in Aircraft Accidents by Year')

#customize the ticks and limit of the Y axis
plt.xticks(range(1980, 2023, 5))
plt.yticks(range(0, 2000, 200))
plt.ylim(0, 2000)

#show the graph
plt.show()

Out[78]:
Figure: Number of Uninjured People Involved in Aircraft Accidents by Year
A line graph showing the number of uninjured people involved in aircraft accidents by year from 1980 to 2020. The y-axis is labeled 'Number of Uninjured' and ranges from 0 to 1800. The x-axis is labeled 'Year' and ranges from 1980 to 2020. The line shows a fluctuating trend with a notable peak around 1985 and another around 2010.

In [79]: #create a plot
plt.figure(figsize=(10, 6))

#create a line graph for the three types of Injuries
sns.lineplot(data=injured_grouped_df, x='Year', y='Total_Serious_Injuries', label='Se
sns.lineplot(data=injured_grouped_df, x='Year', y='Total_Minor_Injuries', label='Mino
sns.lineplot(data=injured_grouped_df, x='Year', y='Total_Fatal_Injuries', label='Fata

#customize the labels and title
plt.xlabel('Year')
plt.ylabel('Number of People')
plt.title('Number of Injured People Involved in Aircraft Accidents by Year')

#customize the ticks and Y limit
plt.xticks(range(1980, 2023, 5))
plt.yticks(range(0, 500, 50))
plt.ylim(0, 500)

#show the legend and graph
plt.legend()
plt.show()

Out[79]:
Figure: Number of Injured People Involved in Aircraft Accidents by Year
A line graph showing the number of injured people involved in aircraft accidents by year from 1980 to 2020. The y-axis is labeled 'Number of People' and ranges from 0 to 450. The x-axis is labeled 'Year' and ranges from 1980 to 2020. The graph shows three lines: 'Serious Injuries' (blue), 'Fatal Injuries' (orange), and 'Minor Injuries' (green). The total number of injured people shows a fluctuating trend with a notable peak around 1985 and another around 2010.

In [80]: #create a plot for two separate graphs
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(18, 6))

#Create a pivot table to reshape the data
injured_pivot_df = injured_grouped_filtered_df.pivot_table(index='Year', columns='Airc
uninjured_pivot_df = injured_grouped_filtered_df.pivot_table(index='Year', columns='A

#Create a Total Injuries line plot using Seaborn
sns.lineplot(data=injured_pivot_df, ax=ax1)

#Set the labels and title
ax1.set_xlabel('Year')
ax1.set_ylabel('Number of Injuries')
ax1.set_title('Total Number of People Injured in Aircraft Accidents by Year')

#set the y limit
ax1.set_ylim(0, 3000)

#Create a total uninjured line plot using Seaborn
sns.lineplot(data=uninjured_pivot_df, ax=ax2)

#Set the labels and title
ax2.set_xlabel('Year')
ax2.set_ylabel('Number of Uninjured')
ax2.set_title('Total Number of Uninjured People Involved in Aircraft Accidents by Year')

#customize the y limit
ax2.set_ylim(0, 3000)

# Display the legend and show the plot
plt.legend(title='Aircraft Damage')
plt.show()

Out[80]:
Figure: Total Number of People Injured in Aircraft Accidents by Year, Total Number of Uninjured People Involved in Aircraft Accidents by Year
Two side-by-side line plots. The left plot shows 'Total Number of People Injured in Aircraft Accidents by Year' with three lines: 'Accident' (blue), 'Destroyed' (orange), and 'Substantial' (green). The right plot shows 'Total Number of Uninjured People Involved in Aircraft Accidents by Year' with three lines: 'Accident' (blue), 'Destroyed' (orange), and 'Substantial' (green). The data is as follows:


| Year | Accident | Destroyed | Substantial | Uninjured |
|------|----------|-----------|-------------|-----------|
| 1980 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 1985 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 1990 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 1995 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 2000 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 2005 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 2010 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 2015 | 26978    | 9518.0    | 1325.0      | 37821.0   |
| 2020 | 26978    | 9518.0    | 1325.0      | 37821.0   |



In [81]: #create a dataframe based on the Make and Aircraft_Damage columns
make_damage_grouped_df = aircraft_df.groupby(['Make', 'Aircraft_Damage']).size().sort
make_damage_grouped_df

# Filter the DataFrame to only show Destroyed from Aircraft_Damage
make_damage_destroyed_df = make_damage_grouped_df[make_damage_grouped_df['Aircraft_Da
make_damage_destroyed_df

#create a top 10
top10_make_damage_destroyed_df = make_damage_destroyed_df.head(10).reset_index(drop=True)
top10_make_damage_destroyed_df

#Filter the DataFrame to only show Minor from Aircraft_Damage
make_damage_minor_df = make_damage_grouped_df[make_damage_grouped_df['Aircraft_Damage
make_damage_minor_df

#create a top 10
top10_make_damage_minor_df = make_damage_minor_df.head(10).reset_index(drop=True)
print(top10_make_damage_minor_df)

Make Aircraft_Damage Count
0  CESSNA      Destroyed    5172
1  PIPER       Destroyed    3423
2  BEECH       Destroyed    1576
3  BELL        Destroyed    696
4  MOONEY      Destroyed    373
5  GRUMMAN     Destroyed    298
6  ROBINSON    Destroyed    282
7  BELLANCA    Destroyed    231
8  HUGBONINCA Destroyed    187
9  BOEING      Destroyed    166

Make Aircraft_Damage Count
0      BOEING      Minor    710
1      CESSNA      Minor    383
2      PIPER       Minor    204
3      BEECH      Minor    168
4  MCGRUNNELL DOUGLAS Minor    162
5      AIRBUS     Minor    59
6      DOUGLAS    Minor    54
7  AIRBUS INDOUS Minor    55
8      EMBRAER    Minor    49
9      BELL       Minor    47

In [82]: #create a plot for two separate graphs
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(25, 10))

#Create a Destroyed Make bar plot using Seaborn
sns.barplot(data=make_damage_destroyed_df, x='Make', y='Count', ax=ax1)

#set the labels and title
ax1.set_xlabel('Number of Destroyed Aircraft (Thousands)')
ax1.set_ylabel('Total Number of Destroyed Aircraft by Make')

#set the y limit
ax1.set_ylim(0, 3000)

#Create a total uninjured line plot using Seaborn
sns.lineplot(data=uninjured_pivot_df, ax=ax2)

#Set the labels and title
ax2.set_xlabel('Year')
ax2.set_ylabel('Number of Uninjured')
ax2.set_title('Total Number of Uninjured People Involved in Aircraft Accidents by Year')

#customize the y limit
ax2.set_ylim(0, 3000)

# Display the legend and show the plot
plt.legend(title='Aircraft Damage')
plt.show()

Out[82]:
Figure: Number of Destroyed Aircraft (Thousands) by Make, Total Number of Uninjured People Involved in Aircraft Accidents by Year
Two side-by-side plots. The left plot is a bar chart showing the 'Total Number of Destroyed Aircraft by Make' in thousands. The right plot is a line graph showing the 'Total Number of Uninjured People Involved in Aircraft Accidents by Year' from 1980 to 2020. The data is as follows:


| Make   | Count (Thousands) |
|--------|-------------------|
| CESSNA | 26.978            |
| PIPER  | 14.790            |
| BEECH  | 5.351             |
| BOEING | 2.735             |
| BELL   | 2.678             |


The right plot shows the 'Total Number of Uninjured People Involved in Aircraft Accidents by Year' from 1980 to 2020, with a fluctuating trend and a notable peak around 1985 and another around 2010.
```

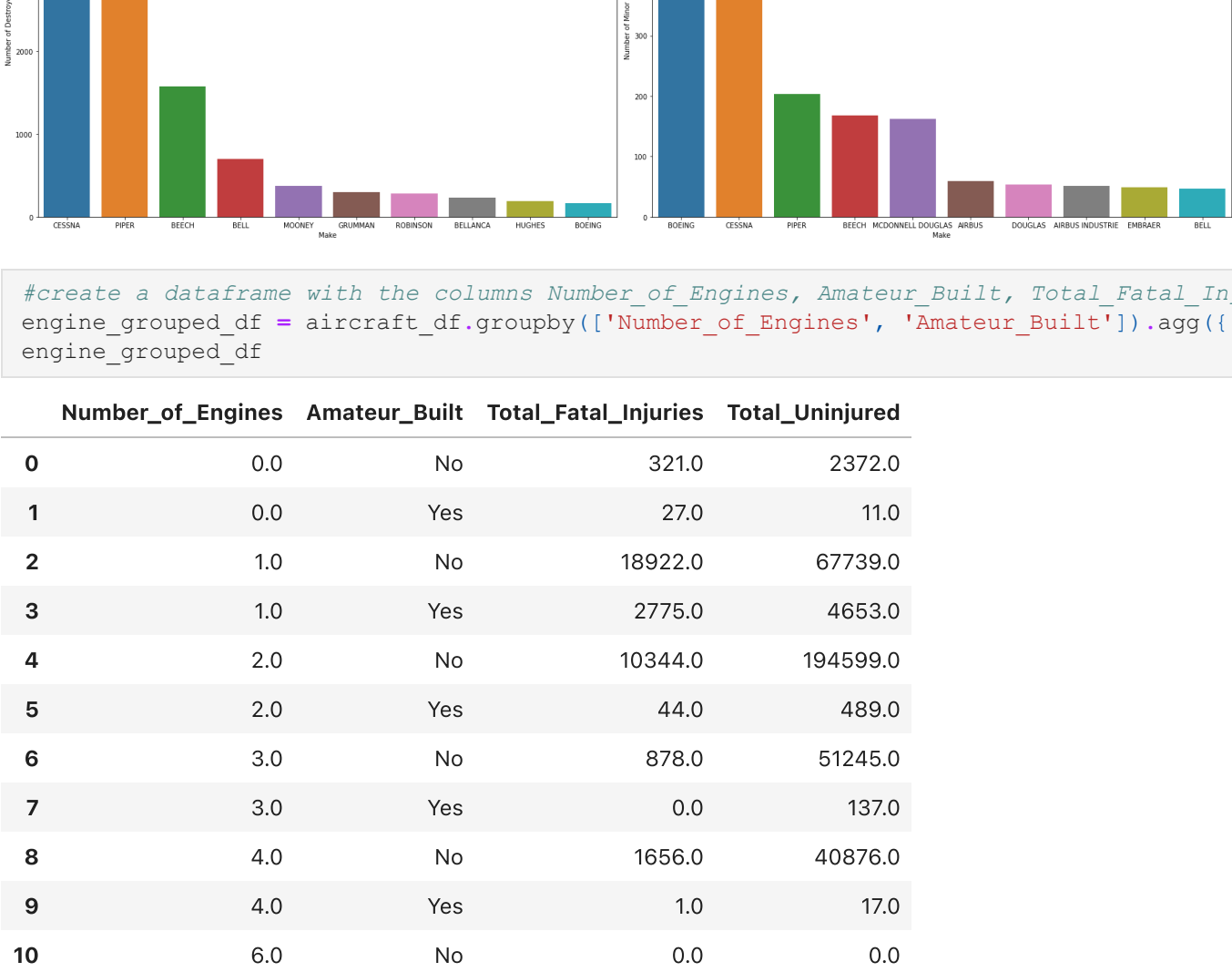


```
#create a Mini Make bar plot using Seaborn
sns.barplot(data=top10_make_damage_minor_df, x='Make', y='Count', ax=ax2)

#set the labels and title
ax2.set_xlabel('Make')
ax2.set_ylabel('Number of Minor Accidents (Hundreds)')
ax2.set_title('Top 10 Number of Minor Accidents by Make')

#adjust the spacing
plt.tight_layout()

#show the graph
plt.show()
```



```
In [83]: #create a dataframe with the columns Number_of_Engines, Amateur_Built, Total_Fatal_Injuries
engine_grouped_df = aircraft_df.groupby(['Number_of_Engines', 'Amateur_Built']).agg({'
```

```
Out[83]:
```

	Number_of_Engines	Amateur_Built	Total_Fatal_Injuries	Total_Uninjured
0	0.0	No	321.0	2372.0
1	0.0	Yes	27.0	11.0
2	1.0	No	18922.0	67739.0
3	1.0	Yes	2775.0	4653.0
4	2.0	No	10344.0	194599.0
5	2.0	Yes	44.0	489.0
6	3.0	No	878.0	51245.0
7	3.0	Yes	0.0	137.0
8	4.0	No	1656.0	40876.0
9	4.0	Yes	1.0	17.0
10	6.0	No	0.0	0.0
11	8.0	No	0.0	0.0

```
In [84]: #create a plot
plt.figure(figsize=(8, 6))

#create a scatter plot of fatal injuries by number of engines with matplotlib and color
plt.scatter(engine_grouped_df['Number_of_Engines'], engine_grouped_df['Total_Fatal_Injuries'])

#customize the labels and title
plt.xlabel('Number of Engines')
plt.ylabel('Number of Fatal Injuries')
plt.title('Proportion of Fatal Injuries by Number of Engines')

#customize the xlim and ylim
plt.xlim(-1.5, engine_grouped_df['Number_of_Engines'].max() + 1)
plt.ylim(-2000, 30000)

#customize the legend
legend_elements = [plt.Line2D([0], [0], marker='o', color='w', label='yes', markerfacecolor='r'),
plt.Line2D([0], [0], marker='o', color='b', label='no', markerfacecolor='b')]

#show the legend and plot
plt.legend(handles=legend_elements, title='Amateur Built')
plt.show()
```



```
In [85]: #create a modified dataset to use for Tableau
#aircraft_df.to_csv('modified_aircraft_data.csv', index=False)
```