# DATA VISUALIZATION ON BIRD-STRIKE

## INTRODUCTION

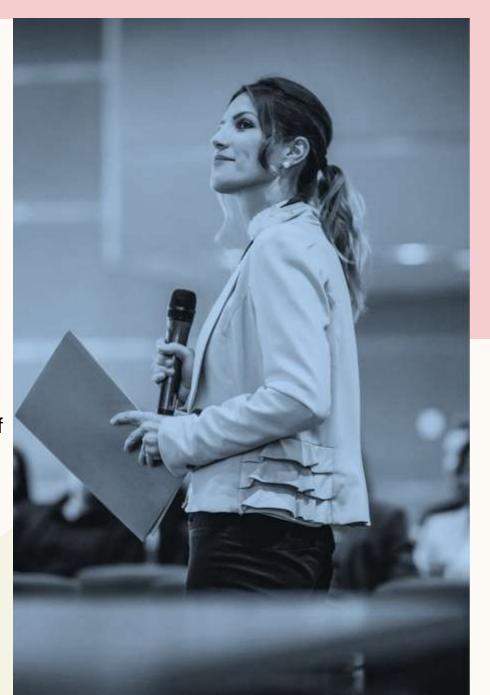
In the realm of transport and communication, analytics play a pivotal role in addressing pressing environmental and safety challenges.

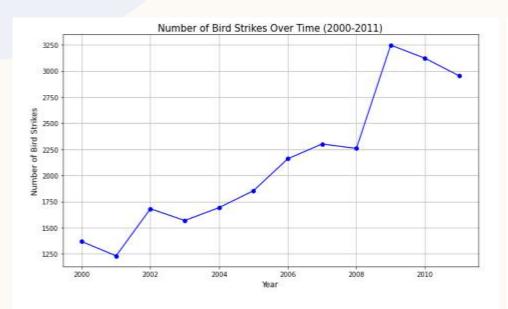
Among these, **bird strikes**—collisions between birds and aircraft during various flight phases—pose a significant concern. These incidents can lead to considerable damage to aircraft structures and, especially for jet engines, may result in dangerous loss of thrust if birds are ingested. This risk is heightened during take-off, initial climb, approach, and landing, when aircraft are at lower altitudes with increased bird activity.

The Federal Aviation Administration (FAA) has collected extensive data on bird strikes from 2000 to 2011. By analyzing this data, we can gain insights into the frequency and impact of these collisions. Our goal is to understand these trends better and explore innovative solutions, including technological advancements and artificial intelligence, to

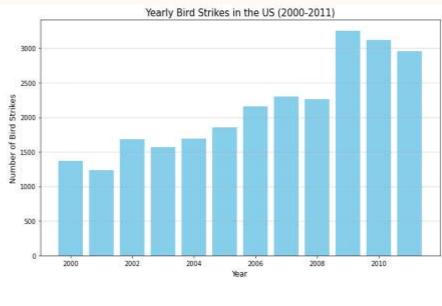
#### **CASE STUDIES**

- Visuals Depicting the Number of Bird Strikes
- Yearly Analysis & Bird Strikes in the US
- Top 10 US Airlines in terms of having encountered bird strikes
- Airports with most incidents of bird strikes Top 50
- Yearly Cost Incurred due to Bird Strikes:
- When do most bird strikes occur?
- Altitude of Aeroplanes at the time of strike
- Phase of flight at the time of the strike.
- Average Altitude of the Aeroplanes in different phases at the time of strike
- Effect of Bird Strikes & Impact on Flight
- Effect of Strike at Different Altitude
- Were Pilots Informed? & Prior Warning and Effect of Strike Relation

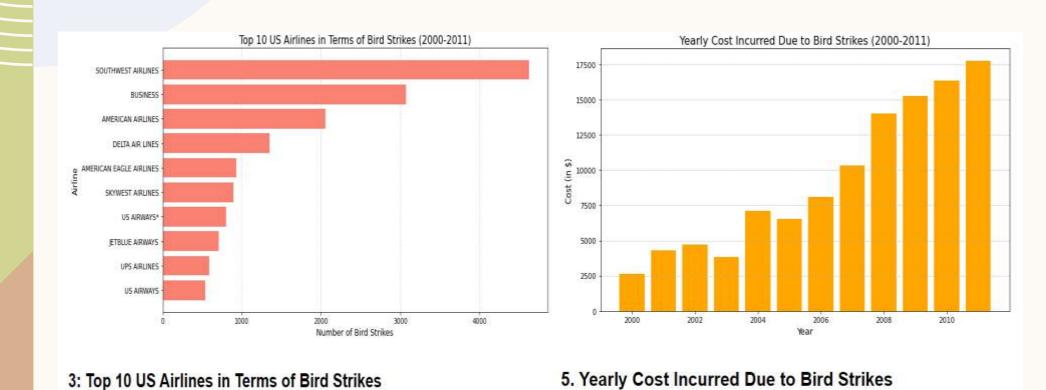


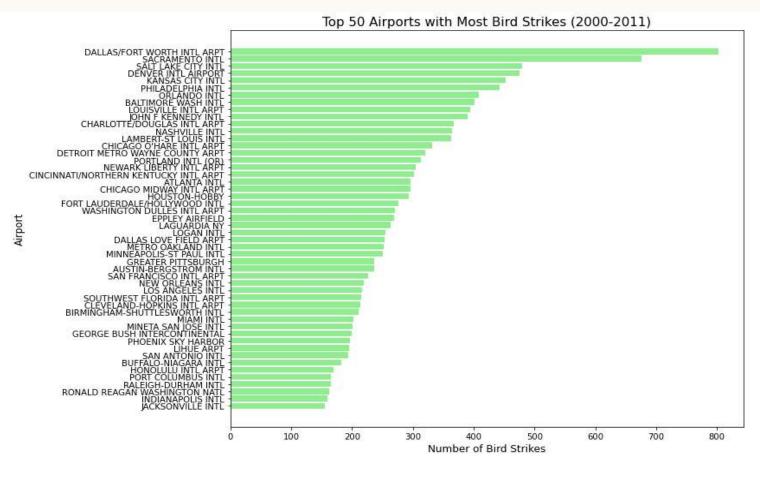


1. Visuals Depicting the Number of Bird Strikes

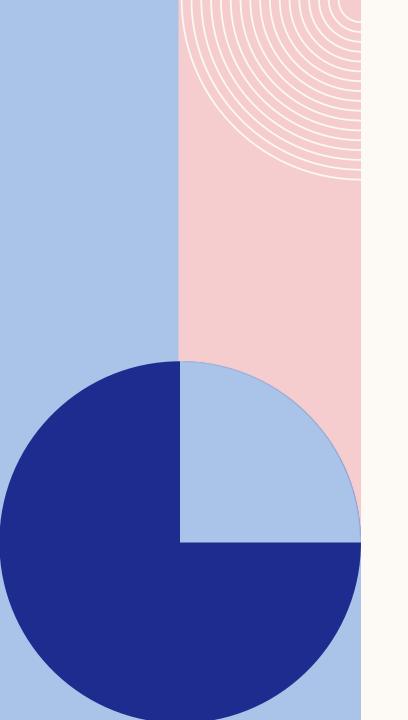


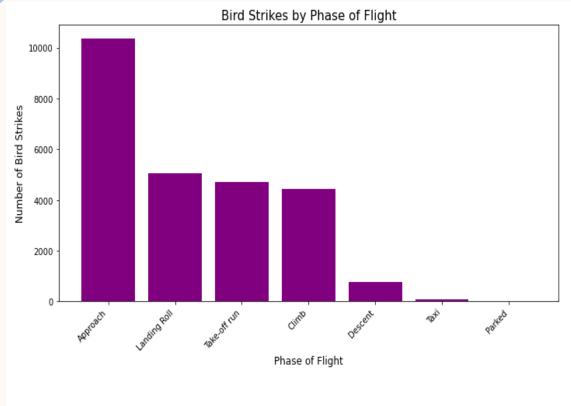
2. Yearly Analysis & Bird Strikes in the US



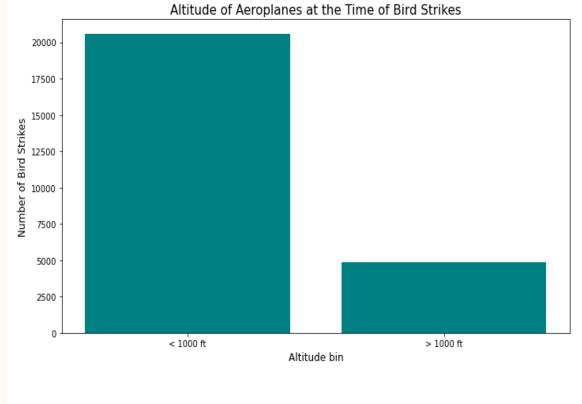




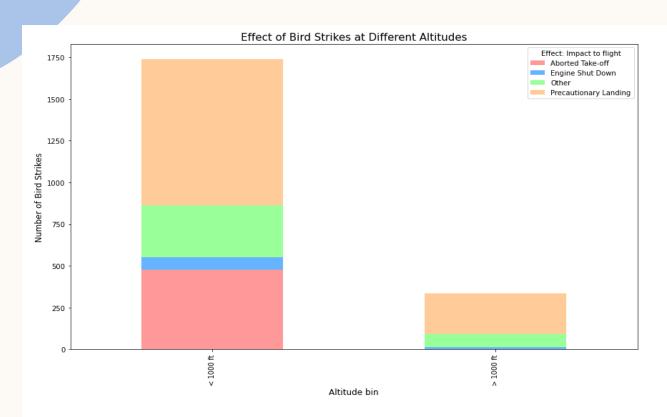


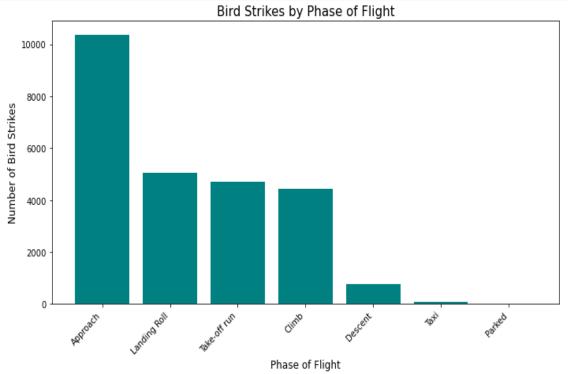


6. When Do Most Bird Strikes Occur?



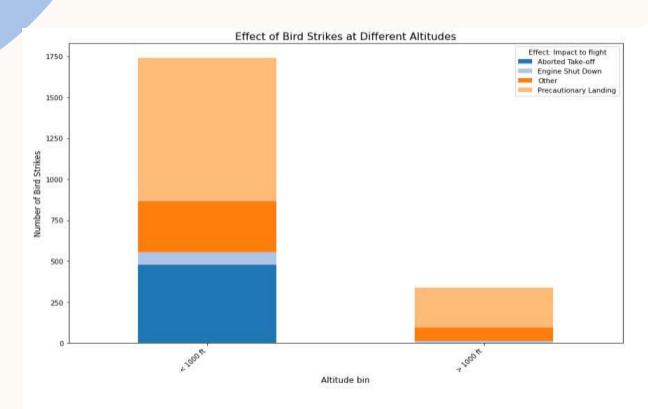
7 Altitude of aeroplanes at the time of strike

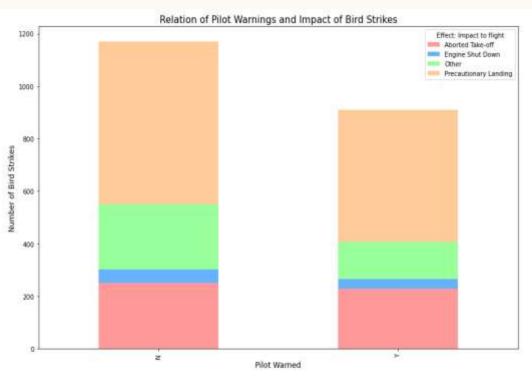




9 Average Altitude of the aeroplanes in different phases at the time of strike  $\P$ 

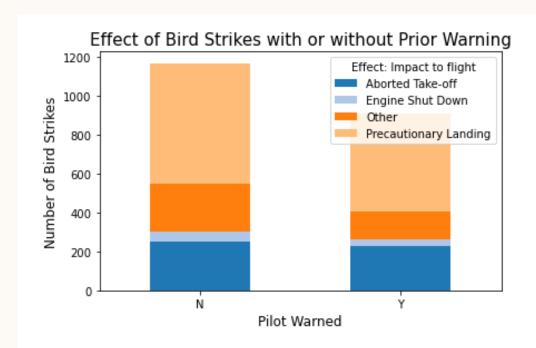
8 Phase of flight at the time of the strike.





11 Effect of Strike at Different Altitude

10 Effect of Bird Strikes & Impact on Flight



12 Were Pilots Informed? & Prior Warning and Effect of Strike Relation

## THANK YOU

# Import necessary libraries import pandas as pd # Load the dataset file\_path = 'Bird Strikes.csv' bird\_strikes\_df = pd.read\_csv(file\_path) # Display the first few rows of the dataset to inspect its structure bird\_strikes\_df.head(), bird\_strikes\_df.columns C:\Users\vivek\anaconda3\lib\site-packages\pandas\core\computation\expressions.py:21: UserWarning: Pandas requires version '2.8.4' or newer of 'numexpr' (version '2.7.3' currently installed). from pandas.core.computation.check import NUMEXPR\_INSTALLED C:\Users\vivek\anaconda3\lib\site-packages\pandas\core\arrays\masked.py:60: UserWarning: Pandas requires version '1.3.6' or newer of 'bottleneck' (version '1.3.2' currently install from pandas.core import ( Record ID Aircraft: Type Airport: Name Altitude bin \ Out[1]: Airplane 202152 LAGUARDIA NY > 1000 ft Airplane DALLAS/FORT WORTH INTL ARPT < 1000 ft 208159 2 207601 Airplane LAKEFRONT AIRPORT < 1000 ft < 1000 ft 3 215953 Airplane SEATTLE-TACOMA INTL 219878 Airplane NORFOLK INTL < 1000 ft Aircraft: Make/Model Wildlife: Number struck \ 0 B-737-400 Over 100 MD-80 Over 100 1 2 C-500 Over 100 B-737-400 Over 100 3 Over 100 4 CL-RJ100/200 Wildlife: Number Struck Actual Effect: Impact to flight FlightDate \ Engine Shut Down 11/23/00 0:00 0 859 424 7/25/01 0:00 2 261 9/14/01 0:00 3 806 Precautionary Landing 9/5/02 0:00 4 942 6/23/03 0:00 NaN Effect: Indicated Damage  $\dots$  Remains of wildlife sent to Smithsonian  $\setminus$ 0 Caused damage ... Caused damage ... False 1 2 False No damage ... False 3 No damage ... False No damage ... Remarks Wildlife: Size \ 0 FLT 753. PILOT REPTD A HUNDRED BIRDS ON UNKN T... 102 CARCASSES FOUND. 1 LDG LIGHT ON NOSE GEAR ... Small FLEW UNDER A VERY LARGE FLOCK OF BIRDS OVER AP... Small 3 NOTAM WARNING. 26 BIRDS HIT THE A/C, FORCING A... Small NO DMG REPTD. Small Wildlife: Species Pilot warned of birds or wildlife? \ Conditions: Sky No Cloud Unknown bird - medium 0 N Rock pigeon Υ 1 Some Cloud European starling Ν 2 No Cloud 3 European starling Υ Some Cloud European starling 4 No Cloud Cost: Total \$ Feet above ground Number of people injured Is Aircraft Large? 30,736 1,500 0 0 Yes 0 1 0 0 No 2 0 50 0 No 0 50 0 3 Yes No [5 rows x 26 columns], Index(['Record ID', 'Aircraft: Type', 'Airport: Name', 'Altitude bin', 'Aircraft: Make/Model', 'Wildlife: Number struck', 'Wildlife: Number Struck Actual', 'Effect: Impact to flight', 'FlightDate', 'Effect: Indicated Damage', 'Aircraft: Number of engines?', 'Aircraft: Airline/Operator', 'Origin State', 'When: Phase of flight', 'Conditions: Precipitation', 'Remains of wildlife collected?', 'Remains of wildlife sent to Smithsonian', 'Remarks', 'Wildlife: Size', 'Conditions: Sky', 'Wildlife: Species', 'Pilot warned of birds or wildlife?', 'Cost: Total \$', 'Feet above ground', 'Number of people injured', 'Is Aircraft Large?'], dtype='object')) Case studies import matplotlib.pyplot as plt import numpy as np # Check for any NaN or invalid values in 'Year' and 'Number of Bird Strikes' bird\_strikes\_per\_year.dropna(subset=['Year', 'Number of Bird Strikes'], inplace=True) # Convert to integer and numeric types, if necessary bird\_strikes\_per\_year['Year'] = pd.to\_numeric(bird\_strikes\_per\_year['Year'], errors='coerce').astype(int) bird\_strikes\_per\_year['Number of Bird Strikes'] = pd.to\_numeric(bird\_strikes\_per\_year['Number of Bird Strikes'], errors='coerce') # Ensure the columns are numpy arrays to avoid multi-dimensional indexing issues years = bird\_strikes\_per\_year['Year'].to\_numpy() strikes = bird\_strikes\_per\_year['Number of Bird Strikes'].to\_numpy() # Plot the data plt.figure(figsize=(10, 6)) plt.plot(years, strikes, marker='o', color='b') plt.title('Number of Bird Strikes Over Time (2000-2011)', fontsize=15) plt.xlabel('Year', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.grid(True) plt.tight\_layout() plt.show() Number of Bird Strikes Over Time (2000-2011) 3250 3000 2750 of Bird Strikes 2000 1750 1500 1250 2000 2002 2004 2006 2008 2010 Year 1. Visuals Depicting the Number of Bird Strikes In [5]: plt.figure(figsize=(10, 6)) plt.bar(bird\_strikes\_per\_year['Year'], bird\_strikes\_per\_year['Number of Bird Strikes'], color='skyblue') plt.title('Yearly Bird Strikes in the US (2000-2011)', fontsize=15) plt.xlabel('Year', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.grid(axis='y', linestyle='--', alpha=0.7) plt.tight\_layout() plt.show() Yearly Bird Strikes in the US (2000-2011) 3000 2500 Number of Bird Strikes 2000 500 2004 2006 2008 Year 2. Yearly Analysis & Bird Strikes in the US In [6]: # Group by the 'Aircraft: Airline/Operator' column to find the top 10 airlines with the most bird strikes top\_airlines = bird\_strikes\_df.groupby('Aircraft: Airline/Operator').size().reset\_index(name='Number of Bird Strikes') # Sort the airlines by the number of bird strikes in descending order and take the top 10 top\_10\_airlines = top\_airlines.sort\_values(by='Number of Bird Strikes', ascending=False).head(10) # Plot the top 10 airlines plt.figure(figsize=(12, 6)) plt.barh(top\_10\_airlines['Aircraft: Airline/Operator'], top\_10\_airlines['Number of Bird Strikes'], color='salmon') plt.title('Top 10 US Airlines in Terms of Bird Strikes (2000-2011)', fontsize=15) plt.xlabel('Number of Bird Strikes', fontsize=12) plt.ylabel('Airline', fontsize=12) plt.gca().invert\_yaxis() # Invert y-axis to have the top airline at the top plt.grid(axis='x', linestyle='--', alpha=0.7) plt.tight\_layout() plt.show() Top 10 US Airlines in Terms of Bird Strikes (2000-2011) SOUTHWEST AIRLINES BUSINESS AMERICAN AIRLINES DELTA AIR LINES a AMERICAN EAGLE AIRLINES US AIRWAYS\* JETBLUE AIRWAYS UPS AIRLINES 3000 4000 2000 Number of Bird Strikes 3: Top 10 US Airlines in Terms of Bird Strikes In [7]: top\_airports = bird\_strikes\_df.groupby('Airport: Name').size().reset\_index(name='Number of Bird Strikes') top\_50\_airports = top\_airports.sort\_values(by='Number of Bird Strikes', ascending=False).head(50) plt.figure(figsize=(12, 8)) plt.barh(top\_50\_airports['Airport: Name'], top\_50\_airports['Number of Bird Strikes'], color='lightgreen') plt.title('Top 50 Airports with Most Bird Strikes (2000-2011)', fontsize=15) plt.xlabel('Number of Bird Strikes', fontsize=12) plt.ylabel('Airport', fontsize=12) plt.gca().invert\_yaxis() plt.tight\_layout() plt.show() Top 50 Airports with Most Bird Strikes (2000-2011) DALLAS/FORT\_WOR DETROIT METRO WA POF NEWARK L CINCINNATI/NORTHERN KEN Airport RALEIGH-DURHAM RONALD REAGAN WASHINGTON 100 200 400 600 Number of Bird Strikes 4. Airports with Most Incidents of Bird Strikes – Top 50 In [8]: bird\_strikes\_df['Cost: Total \$'] = pd.to\_numeric(bird\_strikes\_df['Cost: Total \$'], errors='coerce') yearly\_cost = bird\_strikes\_df.groupby('Year')['Cost: Total \$'].sum().reset\_index() plt.figure(figsize=(10, 6)) plt.bar(yearly\_cost['Year'], yearly\_cost['Cost: Total \$'], color='orange') plt.title('Yearly Cost Incurred Due to Bird Strikes (2000-2011)', fontsize=15) plt.xlabel('Year', fontsize=12) plt.ylabel('Cost (in \$)', fontsize=12) plt.grid(axis='y', linestyle='--', alpha=0.7) plt.tight\_layout() plt.show() Yearly Cost Incurred Due to Bird Strikes (2000-2011) 17500 15000 12500 Cost (in \$) 10000 7500 5000 2500 2004 2006 2010 2008 Year 5. Yearly Cost Incurred Due to Bird Strikes In [30]: import pandas as pd import matplotlib.pyplot as plt # Load the dataset bird\_strikes\_data = pd.read\_csv('Bird Strikes.csv') # Perform the analysis with the correct column name phase\_of\_flight\_counts = bird\_strikes\_data['When: Phase of flight'].value\_counts().reset\_index() # Renaming the columns for clarity phase\_of\_flight\_counts.columns = ['Flight Phase', 'Number of Bird Strikes'] # Plotting the bar chart plt.figure(figsize=(10, 6)) plt.bar(phase\_of\_flight\_counts['Flight Phase'], phase\_of\_flight\_counts['Number of Bird Strikes'], color='purple') plt.title('Bird Strikes by Phase of Flight', fontsize=15) plt.xlabel('Phase of Flight', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.xticks(rotation=45, ha='right') plt.tight\_layout() plt.show() Bird Strikes by Phase of Flight 10000 8000 Number of Bird Strikes 6000 4000 2000 Phase of Flight 6. When Do Most Bird Strikes Occur? altitude\_counts = bird\_strikes\_data['Altitude bin'].value\_counts().reset\_index(name='Number of Bird Strikes') altitude\_counts.columns = ['Altitude bin', 'Number of Bird Strikes'] plt.figure(figsize=(10, 6)) plt.bar(altitude\_counts['Altitude bin'], altitude\_counts['Number of Bird Strikes'], color='teal') plt.title('Altitude of Aeroplanes at the Time of Bird Strikes', fontsize=15) plt.xlabel('Altitude bin', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.tight\_layout() plt.show() Altitude of Aeroplanes at the Time of Bird Strikes 20000 17500 Number of Bird Strikes 12500 10000 7500 5000 2500 < 1000 ft > 1000 ft Altitude bin 7 Altitude of aeroplanes at the time of strike In [25]: import pandas as pd import matplotlib.pyplot as plt # Load the dataset bird\_strikes\_data = pd.read\_csv('Bird Strikes.csv') # Count the occurrences of each phase of flight phase\_of\_flight\_counts = bird\_strikes\_data['When: Phase of flight'].value\_counts().reset\_index(name='Number of Bird Strikes') phase\_of\_flight\_counts.columns = ['Phase of Flight', 'Number of Bird Strikes'] # Plotting the bar chart plt.figure(figsize=(10, 6)) plt.bar(phase\_of\_flight\_counts['Phase of Flight'], phase\_of\_flight\_counts['Number of Bird Strikes'], color='teal') plt.title('Bird Strikes by Phase of Flight', fontsize=15) plt.xlabel('Phase of Flight', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.xticks(rotation=45, ha='right') # Rotate labels for better readability plt.tight\_layout() plt.show() Bird Strikes by Phase of Flight 10000 8000 Number of Bird Strikes 6000 4000 2000 Phase of Flight 8 Phase of flight at the time of the strike. In [17]: altitude\_effect = bird\_strikes\_df.groupby(['Altitude bin', 'Effect: Impact to flight']).size().unstack(fill\_value=0) altitude\_effect.plot(kind='bar', stacked=True, figsize=(12, 8), color=['#ff9999','#66b3ff','#99ff99','#ffcc99']) plt.title('Effect of Bird Strikes at Different Altitudes', fontsize=15) plt.xlabel('Altitude bin', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.tight\_layout() plt.show() Effect of Bird Strikes at Different Altitudes Effect: Impact to flight 1750 Aborted Take-off Engine Shut Down Other Precautionary Landing 1500 1250 Number of Bird Strikes 1000 750 500 250 Altitude bin 9 Average Altitude of the aeroplanes in different phases at the time of strike In [18]: warning\_effect\_relation = bird\_strikes\_df.groupby(['Pilot warned of birds or wildlife?', 'Effect: Impact to flight']).size().unstack(fill\_value=0) warning\_effect\_relation.plot(kind='bar', stacked=True, figsize=(12, 8), color=['#ff9999','#66b3ff','#99ff99','#ffcc99']) plt.title('Relation of Pilot Warnings and Impact of Bird Strikes', fontsize=15) plt.xlabel('Pilot Warned', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.tight\_layout() plt.show() Relation of Pilot Warnings and Impact of Bird Strikes 1200 Effect: Impact to flight Aborted Take-off Engine Shut Down Precautionary Landing 1000 800 Number of Bird Strikes 400 200 Pilot Warned 10 Effect of Bird Strikes & Impact on Flight In [27]: import pandas as pd import matplotlib.pyplot as plt # Load the dataset bird\_strikes\_data = pd.read\_csv('Bird Strikes.csv') # Group by 'Altitude bin' and 'Effect: Impact to flight' and count occurrences effect\_by\_altitude = bird\_strikes\_data.groupby(['Altitude bin', 'Effect: Impact to flight']).size().unstack(fill\_value=0) # Plotting the stacked bar chart without using colormap plt.figure(figsize=(12, 8)) effect\_by\_altitude.plot(kind='bar', stacked=True, ax=plt.gca(), color=plt.get\_cmap('tab20').colors) plt.title('Effect of Bird Strikes at Different Altitudes', fontsize=15) plt.xlabel('Altitude bin', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.xticks(rotation=45, ha='right') # Rotate labels for better readability plt.tight\_layout() plt.show() Effect of Bird Strikes at Different Altitudes Effect: Impact to flight 1750 Aborted Take-off Engine Shut Down Other Precautionary Landing 1500 1250 Number of Bird Strikes 500 250 Altitude bin 11 Effect of Strike at Different Altitude In [28]: import pandas as pd import matplotlib.pyplot as plt # Load the dataset bird\_strikes\_data = pd.read\_csv('Bird Strikes.csv') # Count occurrences of effects by whether pilots were warned or not warning\_effect\_counts = bird\_strikes\_data.groupby(['Pilot warned of birds or wildlife?', 'Effect: Impact to flight']).size().unstack(fill\_value=0) # Plotting the stacked bar chart plt.figure(figsize=(12, 8)) warning\_effect\_counts.plot(kind='bar', stacked=True, color=plt.get\_cmap('tab20').colors) plt.title('Effect of Bird Strikes with or without Prior Warning', fontsize=15) plt.xlabel('Pilot Warned', fontsize=12) plt.ylabel('Number of Bird Strikes', fontsize=12) plt.xticks(rotation=0) # Rotate labels for better readability plt.tight\_layout() plt.show() <Figure size 864x576 with 0 Axes> Effect of Bird Strikes with or without Prior Warning 1200 Effect: Impact to flight f Bird Strikes Aborted Take-off Engine Shut Down Other Precautionary Landing 600 оę Number 400 200 Pilot Warned 12 Were Pilots Informed? & Prior Warning and Effect of Strike Relation

In [ ]: