

| Action | Code | Explanation |
|---|---|-------------|
| Mount Drive | <pre>#mounting drive from google.colab import drive drive.mount('/content/drive')</pre> | |
| Import Panda | <pre>import pandas as pd</pre> | |
| Upload CSV file | <pre>df1=pd.read_csv("/content/drive/MyDrive/wk10_pandas_learner_vs1902/Resources/Flight Delays/Flights_Delay_Cause_2019-2020.csv") df2=pd.read_csv("/content/drive/MyDrive/wk10_pandas_learner_vs1902/Resources/Flight Delays/Flights_Delay_Cause_2020-2021.csv") df3=pd.read_csv("/content/drive/MyDrive/wk10_pandas_learner_vs1902/Resources/Flight Delays/Flights_Delay_Cause_2021-2022.csv")</pre> | |
| To see 1 st 5 & last 5 lines | <pre>df1.head() df1tail()</pre> | |
| To combine multiple CSV file in 1 dataset | <pre>#use function .concat() df= pd.concat([df1, df2, df3], ignore_index=True) #setting the parameterignore_index=True will reassign new indexes to the whole dataset, the intial indexes of each document will be ignored</pre> | |
| Explore wholedataset | <pre>df.head(10)</pre> | |
| Explore total rows & Columns | <pre>df.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 42461 entries, 0 to 42460 Data columns (total 16 columns):</pre> | |
| Attribution step | <pre>df.shape (42461, 16)</pre> | |

| | | |
|---------------------------------------|---|---|
| Datatype | <pre>df.dtypes year int64 month int64 carrier object carrier_name object airport object airport_name object arr_flights float64 arr_del15 float64 arr_cancelled float64 arr_diverted float64 arr_delay float64 carrier_delay float64 weather_delay float64 nas_delay float64 security_delay float64 delays float64 dtype: object</pre> | |
| Describe | <pre>df.describe()</pre> | |
| Convert data into a date datatype | <pre>.todatetime()</pre> | |
| Copy original dataset to avoid errors | <pre>df_copy = df.copy(deep=True)</pre> | This is our new copied database name from now on. |
| Add "Date" column | <p>the "month" and "year" from our dataset. We can concatenate our 2 columns month and year using the "+" sign:</p> <pre>df_copy['date'] = df_copy['month'] + df_copy['year'] df_copy.head()</pre> | Error spotted in date column |

| | <table><thead><tr><th></th><th>year</th><th>month</th><th>carrier</th><th>delays</th><th>date</th></tr></thead><tbody><tr><td>0</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2029</td></tr><tr><td>1</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2029</td></tr><tr><td>2</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2029</td></tr><tr><td>3</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2029</td></tr><tr><td>4</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2029</td></tr></tbody></table> | | year | month | carrier | delays | date | 0 | 2022 | 7 | 9E | NaN | 2029 | 1 | 2022 | 7 | 9E | NaN | 2029 | 2 | 2022 | 7 | 9E | NaN | 2029 | 3 | 2022 | 7 | 9E | NaN | 2029 | 4 | 2022 | 7 | 9E | NaN | 2029 | |
|--|--|--------------------------------------|---------|--------|---------|--------|------|---|------|---|----|-----|------|---|------|---|----|-----|------|---|------|---|----|-----|------|---|------|---|----|-----|------|---|------|---|----|-----|------|--|
| | year | month | carrier | delays | date | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 0 | 2022 | 7 | 9E | NaN | 2029 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 1 | 2022 | 7 | 9E | NaN | 2029 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 2 | 2022 | 7 | 9E | NaN | 2029 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 3 | 2022 | 7 | 9E | NaN | 2029 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 4 | 2022 | 7 | 9E | NaN | 2029 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| Change the month & year datatype to string | <pre>.astype()</pre> <p>#First, let's convert the month and year into a string datatype so we can concatenate them afterall: #We need to use the .astype() method to convert data in Pandas</p> <pre>df_copy=df_copy.astype({"month": str, "year":str, "date":str}) df_copy.info()</pre> <pre><class 'pandas.core.frame.DataFrame'> RangeIndex: 42461 entries, 0 to 42460 Data columns (total 17 columns): # Column Non-Null Count Dtype --- --- 0 year 42461 non-null object 1 month 42461 non-null object</pre> <p>OR</p> <pre>#alternative method to convert datatypes in Pandas: df_copy['month'].astype(str) df_copy['year'].astype(str) df_copy['date'].astype(str)</pre> | Datatype changed to object in pandas | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |

| | <pre>df_copy['date'] = df_copy['year']+"-"+df_copy['month'] df_copy.head()</pre> <div><table><thead><tr><th></th><th>year</th><th>month</th><th>carrier</th><th>delays</th><th>date</th></tr></thead><tbody><tr><td>0</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-7</td></tr><tr><td>1</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-7</td></tr><tr><td>2</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-7</td></tr><tr><td>3</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-7</td></tr><tr><td>4</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-7</td></tr></tbody></table></div> | | year | month | carrier | delays | date | 0 | 2022 | 7 | 9E | NaN | 2022-7 | 1 | 2022 | 7 | 9E | NaN | 2022-7 | 2 | 2022 | 7 | 9E | NaN | 2022-7 | 3 | 2022 | 7 | 9E | NaN | 2022-7 | 4 | 2022 | 7 | 9E | NaN | 2022-7 | |
|--|---|-------|---------|--------|------------|--------|------|---|------|---|----|-----|------------|---|------|---|----|-----|------------|---|------|---|----|-----|------------|---|------|---|----|-----|------------|---|------|---|----|-----|------------|--|
| | year | month | carrier | delays | date | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 0 | 2022 | 7 | 9E | NaN | 2022-7 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 1 | 2022 | 7 | 9E | NaN | 2022-7 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 2 | 2022 | 7 | 9E | NaN | 2022-7 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 3 | 2022 | 7 | 9E | NaN | 2022-7 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 4 | 2022 | 7 | 9E | NaN | 2022-7 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| Transform the date into datetime datatype: | <pre>df_copy['date'] = pd.to_datetime(df_copy['date']) df_copy.info()</pre> <div><div>df_copy.head()</div><div><table><thead><tr><th></th><th>year</th><th>month</th><th>carrier</th><th>delays</th><th>date</th></tr></thead><tbody><tr><td>0</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-07-01</td></tr><tr><td>1</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-07-01</td></tr><tr><td>2</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-07-01</td></tr><tr><td>3</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-07-01</td></tr><tr><td>4</td><td>2022</td><td>7</td><td>9E</td><td>NaN</td><td>2022-07-01</td></tr></tbody></table></div></div> | | year | month | carrier | delays | date | 0 | 2022 | 7 | 9E | NaN | 2022-07-01 | 1 | 2022 | 7 | 9E | NaN | 2022-07-01 | 2 | 2022 | 7 | 9E | NaN | 2022-07-01 | 3 | 2022 | 7 | 9E | NaN | 2022-07-01 | 4 | 2022 | 7 | 9E | NaN | 2022-07-01 | |
| | year | month | carrier | delays | date | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 0 | 2022 | 7 | 9E | NaN | 2022-07-01 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 1 | 2022 | 7 | 9E | NaN | 2022-07-01 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 2 | 2022 | 7 | 9E | NaN | 2022-07-01 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 3 | 2022 | 7 | 9E | NaN | 2022-07-01 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| 4 | 2022 | 7 | 9E | NaN | 2022-07-01 | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |
| Correct date format | <pre>df_copy['date'] = pd.to_datetime(df_copy['date']).dt.to_period('M')</pre> | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | | |

| | | |
|--|--|--|
| | <pre>16 date 42461 non-null period[M] dtypes: float64(10), object(6), period[M](1) memory usage: 5.5+ MB</pre> <p>The 'date' column was assigned the 'period' datatype which is another date format</p> | |
| Save added "Date" column in our database | <pre>df_copy.to_csv('flights_dataset.csv') and mount the drive from google.colab import drive drive.mount('/content/drive')</pre> | <p>To save in different location</p> <pre>from google.colab import drive; drive.mount('/content/drive').</pre> <p>Then, use a path within your Drive, like /content/drive/MyDrive/data/flights_dataset.csv.</p> |
| Clear unwanted columns | <pre>df_copy=df_copy.drop(['month', 'year'], axis=1) df_copy.head()</pre> | |
| 3. DATA CLEANING | | |
| Find missing values | <pre>isnull() df_copy.isnull().sum()</pre> | |

| | | |
|---------------------|--|---|
| | <p>Let's check how many missing values our dataset has using the isnull() function</p> <pre> [31] df_copy.isnull().sum() carrier 0 carrier_name 16 airport 0 airport_name 0 arr_flights 107 arr_del15 278 arr_cancelled 107 arr_diverted 107 arr_delay 107 carrier_delay 107 weather_delay 107 nas_delay 107 security_delay 107 delays 42461 date 0 dtype: int64 </pre> | |
| Drop column | <pre> df_copy= df_copy.dropna(subset=['carrier_name']) df_copy.isnull().sum() </pre> | |
| Get column names | <pre> Df_copy.columns Index(['carrier', 'carrier_name', 'airport', 'airport_name', 'arr_flights', 'arr_del15', 'arr_cancelled', 'arr_diverted', 'arr_delay', 'carrier_delay', 'weather_delay', 'nas_delay', 'security_delay', 'delays', 'date'], dtype='object') </pre> | |
| Built in box option | <pre> .box_plot df_copy.boxplot('weather_delay') #We create a function that will return a box plot: </pre> | <pre> # Identify and remove the row with the outlier </pre> |

| | | |
|---------------------------------|--|--|
| Name the chart X Axis | <pre>def box_plot(flights): return df_copy.boxplot(flights) #We call our function passing a data variable as argument: box_plot('nas_delay')</pre> | <pre>outlier_index= df_copy[df_copy['nas_delay']>175000].index df_cleaned= df_copy.drop(outlier_index) Int64Index([363], dtype='int64') #row number</pre> |
| | <pre>df_cleaned.boxplot('nas_delay')</pre> | |
| Python matplotlib visualisation | <pre>Subplots() import matplotlib.pyplot as plt figure, ax = plt.subplots(nrows=2, ncols=1) # subplot layer has 2 rows an 1 column which means 2 figures df_cleaned.hist('weather_delay',ax=ax[0], bins=25, color="purple") df_cleaned.boxplot("weather_delay",ax=ax[1], color="blue")</pre> |  |
| | <pre>def subplot_function(data1, data2): # I added one parameter per plot figure, ax = plt.subplots(nrows=2, ncols=1) # Get statistics min_val= df_cleaned[data1].min() max_val= df_cleaned[data1].max() mean_val= df_cleaned[data1].mean() med_val= df_cleaned[data1].median()</pre> | <pre>def subplot_function(data1, data2): # Define a function to create a multi-p lot figure """ Creates a figure with two subplots : a histogram for data1 and a boxplot for data2. Adds reference lines for min, mean</pre> |

```

# Add lines for the min, mean and median, and max
ax[0].axvline(x=min_val, color = 'gray',
linestyle='dashed', linewidth = 2)
ax[0].axvline(x=mean_val, color = 'cyan',
linestyle='dashed', linewidth = 2)
ax[0].axvline(x=med_val, color = 'red',
linestyle='dashed', linewidth = 2)
ax[0].axvline(x=max_val, color = 'blue',
linestyle='dashed', linewidth = 2)

#Change labels- 1st plot
ax[0].set_ylabel('Frequency')

df_cleaned.hist(data1,ax=ax[0], bins=10,
color="purple")

#Change labels- 2nd plot
ax[1].set_xlabel('')
ax[1].set_ylabel('Frequency')

df_cleaned.boxplot(data2,ax=ax[1], color="blue")

```

, median, and max values in the histogram.

Args:

data1 (str): Name of the column in df_cleaned to be used for the histogram.

data2 (str): Name of the column in df_cleaned to be used for the box plot.

"""

figure, ax = plt.subplots(nrows=2, ncols=1) # Create a figure with 2 rows and 1 column of subplots

Get statistics for data1 to add reference lines

```

min_val = df_cleaned[data1].min()
max_val = df_cleaned[data1].max()
mean_val = df_cleaned[data1].mean()

```

```

)
med_val = df_cleaned[data1].median()

```

Add vertical reference lines for min, mean, median, and max in the first subplot

```

ax[0].axvline(x=min_val, color='gray', linestyle='dashed', linewidth=2,
label='Min')

```

```

ax[0].axvline(x=mean_val, color='cyan', linestyle='dashed', linewidth=2,
label='Mean')

```

```

ax[0].axvline(x=med_val, color='red', linestyle='dashed', linewidth=2,
label='Median')

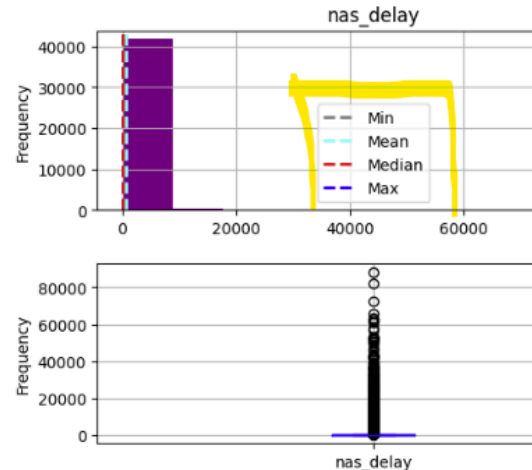
```

```

ax[0].axvline(x=max_val, color='blue', linestyle='dashed', linewidth=2,
label='Max')

```

ax[0].legend() # Add a legend to

| | | |
|---------------------|---|---|
| | | <p>explain the reference lines</p> <pre> # Create histogram for data1 in the first subplot ax[0].set_ylabel('Frequency') # Set y-axis label df_cleaned.hist(data1, ax=ax[0], bins=10, color="purple") # Create boxplot for data2 in the second subplot ax[1].set_xlabel('') # Remove x-axis label for cleaner presentation ax[1].set_ylabel('Frequency') # Set y-axis label df_cleaned.boxplot(data2, ax=ax[1], color="blue") </pre> |
| | subplot_function('nas_delay', "nas_delay") | |
| Add legend in chart | <pre> # Same plot with Labels def subplot_function(data1, data2): # Define a function to create a multi-plot figure """ Creates a figure with two subplots: a histogram for data1 and a boxplot for data2. Adds reference lines for min, mean, median, and max values in the histogram. Args: data1 (str): Name of the column in df_cleaned to be used for the histogram. data2 (str): Name of the column in df_cleaned to be used for the boxplot. """ </pre> | <p>✓ 1s</p> <p>subplot_function('nas_delay', "nas_delay")</p>  |

```

figure, ax = plt.subplots(nrows=2, ncols=1) #
Create a figure with 2 rows and 1 column of subplots

# Get statistics for data1 to add reference lines
min_val = df_cleaned[data1].min()
max_val = df_cleaned[data1].max()
mean_val = df_cleaned[data1].mean()
med_val = df_cleaned[data1].median()

# Add vertical reference lines for min, mean,
median, and max in the first subplot
ax[0].axvline(x=min_val, color='gray',
linestyle='dashed', linewidth=2, label='Min')
ax[0].axvline(x=mean_val, color='cyan',
linestyle='dashed', linewidth=2, label='Mean')
ax[0].axvline(x=med_val, color='red',
linestyle='dashed', linewidth=2, label='Median')
ax[0].axvline(x=max_val, color='blue',
linestyle='dashed', linewidth=2, label='Max')
ax[0].legend() # Add a legend to explain the
reference lines

# Create histogram for data1 in the first subplot
ax[0].set_ylabel('Frequency') # Set y-axis label
df_cleaned.hist(data1, ax=ax[0], bins=10,
color="purple")

# Create boxplot for data2 in the second subplot
ax[1].set_xlabel('') # Remove x-axis label for
cleaner presentation
ax[1].set_ylabel('Frequency') # Set y-axis label
df_cleaned.boxplot(data2, ax=ax[1], color="blue")

```

4 Data Analysis

Q1-Which Airlines have the most delayed flights?

We can group the airline data and calculate the average of delayed flights per airline

```
df_delays=df_cleaned.groupby(by="carrier_name", as_index=False)
#using the argument as_index=False in the groupby function will allow us to return a pandas dataframe,
otherwise a pandas series will be returned.
```

.size()

df_delays.size()



| | carrier_name | size |
|---|------------------------|------|
| 0 | Alaska Airlines Inc. | 1851 |
| 1 | Allegiant Air | 3158 |
| 2 | American Airlines Inc. | 2574 |
| 3 | Delta Air Lines Inc. | 3212 |
| 4 | Endeavor Air Inc. | 2720 |
| 5 | Envoy Air | 3655 |

we can see how many airlines has been grouped with the function .size():

Mean

df_delays=df_delays["arr_delay"].mean()

40.9383...

df_delays # here we get the top 5 of airlines with the most delays

plot the data using the Matplotlib and Seaborn visualisation libraries

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