

Final Project

May 29, 2024

```
[1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
from sklearn.preprocessing import LabelEncoder
from scipy.stats import skew, norm
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split, KFold, cross_val_score
from sklearn.metrics import mean_absolute_error, r2_score
from sklearn.pipeline import make_pipeline
from sklearn.linear_model import LinearRegression, Ridge, Lasso, RidgeCV
from sklearn.preprocessing import OneHotEncoder, RobustScaler
import warnings
from sklearn.preprocessing import StandardScaler
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.metrics import mean_squared_error
from sklearn.svm import SVR

warnings.filterwarnings("ignore")
```

```
[2]: flights = pd.read_csv('Flight_Fare.csv')
flights
```

```
[2]:
```

	Unnamed: 0	airline	flight	source_city	departure_time	stops	\
0	0	SpiceJet	SG-8709	Delhi	Evening	zero	
1	1	SpiceJet	SG-8157	Delhi	Early_Morning	zero	
2	2	AirAsia	I5-764	Delhi	Early_Morning	zero	
3	3	Vistara	UK-995	Delhi	Morning	zero	
4	4	Vistara	UK-963	Delhi	Morning	zero	
...	
300148	300148	Vistara	UK-822	Chennai	Morning	one	
300149	300149	Vistara	UK-826	Chennai	Afternoon	one	
300150	300150	Vistara	UK-832	Chennai	Early_Morning	one	
300151	300151	Vistara	UK-828	Chennai	Early_Morning	one	
300152	300152	Vistara	UK-822	Chennai	Morning	one	

	arrival_time	destination_city	class	duration	days_left	price
0	Night	Mumbai	Economy	2.17	1	5953
1	Morning	Mumbai	Economy	2.33	1	5953
2	Early_Morning	Mumbai	Economy	2.17	1	5956
3	Afternoon	Mumbai	Economy	2.25	1	5955
4	Morning	Mumbai	Economy	2.33	1	5955
...
300148	Evening	Hyderabad	Business	10.08	49	69265
300149	Night	Hyderabad	Business	10.42	49	77105
300150	Night	Hyderabad	Business	13.83	49	79099
300151	Evening	Hyderabad	Business	10.00	49	81585
300152	Evening	Hyderabad	Business	10.08	49	81585

[300153 rows x 12 columns]

```
[3]: flights.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300153 entries, 0 to 300152
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Unnamed: 0            300153 non-null  int64
1   airline               300153 non-null  object
2   flight               300153 non-null  object
3   source_city          300153 non-null  object
4   departure_time       300153 non-null  object
5   stops               300153 non-null  object
6   arrival_time         300153 non-null  object
7   destination_city     300153 non-null  object
8   class               300153 non-null  object
9   duration             300153 non-null  float64
10  days_left            300153 non-null  int64
11  price               300153 non-null  int64
dtypes: float64(1), int64(3), object(8)
memory usage: 27.5+ MB
```

```
[4]: flights.drop(columns = ['Unnamed: 0'], inplace = True)
```

```
[5]: flights
```

```
[5]:
```

	airline	flight	source_city	departure_time	stops	arrival_time \
0	SpiceJet	SG-8709	Delhi	Evening	zero	Night
1	SpiceJet	SG-8157	Delhi	Early_Morning	zero	Morning
2	AirAsia	I5-764	Delhi	Early_Morning	zero	Early_Morning
3	Vistara	UK-995	Delhi	Morning	zero	Afternoon
4	Vistara	UK-963	Delhi	Morning	zero	Morning
...

300148	Vistara	UK-822	Chennai	Morning	one	Evening
300149	Vistara	UK-826	Chennai	Afternoon	one	Night
300150	Vistara	UK-832	Chennai	Early_Morning	one	Night
300151	Vistara	UK-828	Chennai	Early_Morning	one	Evening
300152	Vistara	UK-822	Chennai	Morning	one	Evening

	destination_city	class	duration	days_left	price
0	Mumbai	Economy	2.17	1	5953
1	Mumbai	Economy	2.33	1	5953
2	Mumbai	Economy	2.17	1	5956
3	Mumbai	Economy	2.25	1	5955
4	Mumbai	Economy	2.33	1	5955
...
300148	Hyderabad	Business	10.08	49	69265
300149	Hyderabad	Business	10.42	49	77105
300150	Hyderabad	Business	13.83	49	79099
300151	Hyderabad	Business	10.00	49	81585
300152	Hyderabad	Business	10.08	49	81585

[300153 rows x 11 columns]

```
[6]: new_columns = []
for c in flights.columns:
    new_columns.append(c.replace(' ', '_'))

flights.columns = new_columns

flights.head()
```

```
[6]: airline flight source_city departure_time stops arrival_time \
0 SpiceJet SG-8709 Delhi Evening zero Night
1 SpiceJet SG-8157 Delhi Early_Morning zero Morning
2 AirAsia I5-764 Delhi Early_Morning zero Early_Morning
3 Vistara UK-995 Delhi Morning zero Afternoon
4 Vistara UK-963 Delhi Morning zero Morning

destination_city class duration days_left price
0 Mumbai Economy 2.17 1 5953
1 Mumbai Economy 2.33 1 5953
2 Mumbai Economy 2.17 1 5956
3 Mumbai Economy 2.25 1 5955
4 Mumbai Economy 2.33 1 5955
```

```
[7]: flights.drop(columns = ['flight'], inplace = True)
```

```
[8]: column_list = flights.columns.tolist()
```

```
print(column_list)
```

```
['airline', 'source_city', 'departure_time', 'stops', 'arrival_time',  
'destination_city', 'class', 'duration', 'days_left', 'price']
```

```
[9]: flights
```

```
[9]:      airline source_city departure_time stops  arrival_time \  
0      SpiceJet      Delhi      Evening  zero      Night  
1      SpiceJet      Delhi  Early_Morning  zero      Morning  
2      AirAsia      Delhi  Early_Morning  zero  Early_Morning  
3      Vistara      Delhi      Morning  zero      Afternoon  
4      Vistara      Delhi      Morning  zero      Morning  
...      ...      ...      ...      ...  
300148  Vistara      Chennai      Morning  one      Evening  
300149  Vistara      Chennai      Afternoon  one      Night  
300150  Vistara      Chennai  Early_Morning  one      Night  
300151  Vistara      Chennai  Early_Morning  one      Evening  
300152  Vistara      Chennai      Morning  one      Evening  
  
      destination_city      class  duration  days_left  price  
0              Mumbai  Economy      2.17      1  5953  
1              Mumbai  Economy      2.33      1  5953  
2              Mumbai  Economy      2.17      1  5956  
3              Mumbai  Economy      2.25      1  5955  
4              Mumbai  Economy      2.33      1  5955  
...      ...      ...      ...      ...  
300148      Hyderabad  Business     10.08     49  69265  
300149      Hyderabad  Business     10.42     49  77105  
300150      Hyderabad  Business     13.83     49  79099  
300151      Hyderabad  Business     10.00     49  81585  
300152      Hyderabad  Business     10.08     49  81585
```

```
[300153 rows x 10 columns]
```

```
[10]: exchange_rate = 0.012  
  
flights['dollars'] = (flights['price'] * exchange_rate)
```

```
[11]: flights
```

```
[11]:      airline source_city departure_time stops  arrival_time \  
0      SpiceJet      Delhi      Evening  zero      Night  
1      SpiceJet      Delhi  Early_Morning  zero      Morning  
2      AirAsia      Delhi  Early_Morning  zero  Early_Morning  
3      Vistara      Delhi      Morning  zero      Afternoon  
4      Vistara      Delhi      Morning  zero      Morning  
...      ...      ...      ...      ...
```

300148	Vistara	Chennai	Morning	one	Evening
300149	Vistara	Chennai	Afternoon	one	Night
300150	Vistara	Chennai	Early_Morning	one	Night
300151	Vistara	Chennai	Early_Morning	one	Evening
300152	Vistara	Chennai	Morning	one	Evening

	destination_city	class	duration	days_left	price	dollars
0	Mumbai	Economy	2.17	1	5953	71.436
1	Mumbai	Economy	2.33	1	5953	71.436
2	Mumbai	Economy	2.17	1	5956	71.472
3	Mumbai	Economy	2.25	1	5955	71.460
4	Mumbai	Economy	2.33	1	5955	71.460
...
300148	Hyderabad	Business	10.08	49	69265	831.180
300149	Hyderabad	Business	10.42	49	77105	925.260
300150	Hyderabad	Business	13.83	49	79099	949.188
300151	Hyderabad	Business	10.00	49	81585	979.020
300152	Hyderabad	Business	10.08	49	81585	979.020

[300153 rows x 11 columns]

```
[12]: flights.drop(columns = ['price'], inplace = True)
```

```
[13]: flights
```

```
[13]:
```

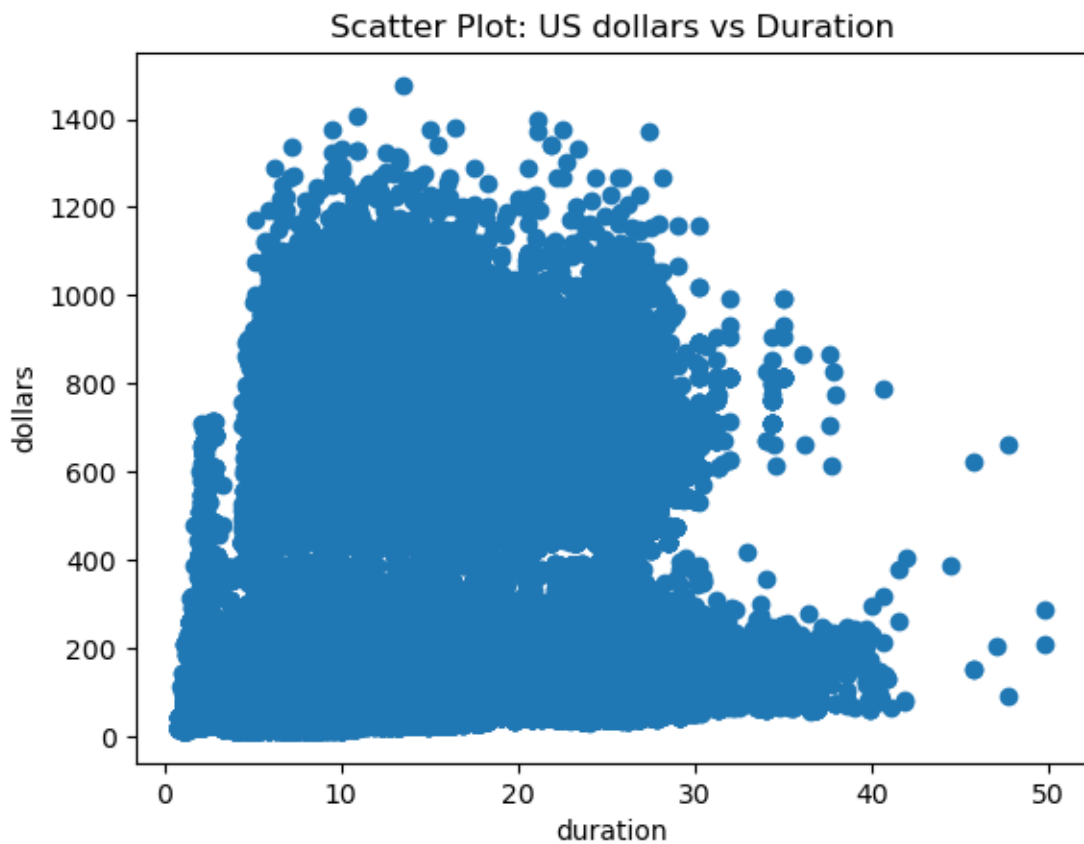
	airline	source_city	departure_time	stops	arrival_time	\
0	SpiceJet	Delhi	Evening	zero	Night	
1	SpiceJet	Delhi	Early_Morning	zero	Morning	
2	AirAsia	Delhi	Early_Morning	zero	Early_Morning	
3	Vistara	Delhi	Morning	zero	Afternoon	
4	Vistara	Delhi	Morning	zero	Morning	
...	
300148	Vistara	Chennai	Morning	one	Evening	
300149	Vistara	Chennai	Afternoon	one	Night	
300150	Vistara	Chennai	Early_Morning	one	Night	
300151	Vistara	Chennai	Early_Morning	one	Evening	
300152	Vistara	Chennai	Morning	one	Evening	

	destination_city	class	duration	days_left	dollars
0	Mumbai	Economy	2.17	1	71.436
1	Mumbai	Economy	2.33	1	71.436
2	Mumbai	Economy	2.17	1	71.472
3	Mumbai	Economy	2.25	1	71.460
4	Mumbai	Economy	2.33	1	71.460
...
300148	Hyderabad	Business	10.08	49	831.180
300149	Hyderabad	Business	10.42	49	925.260

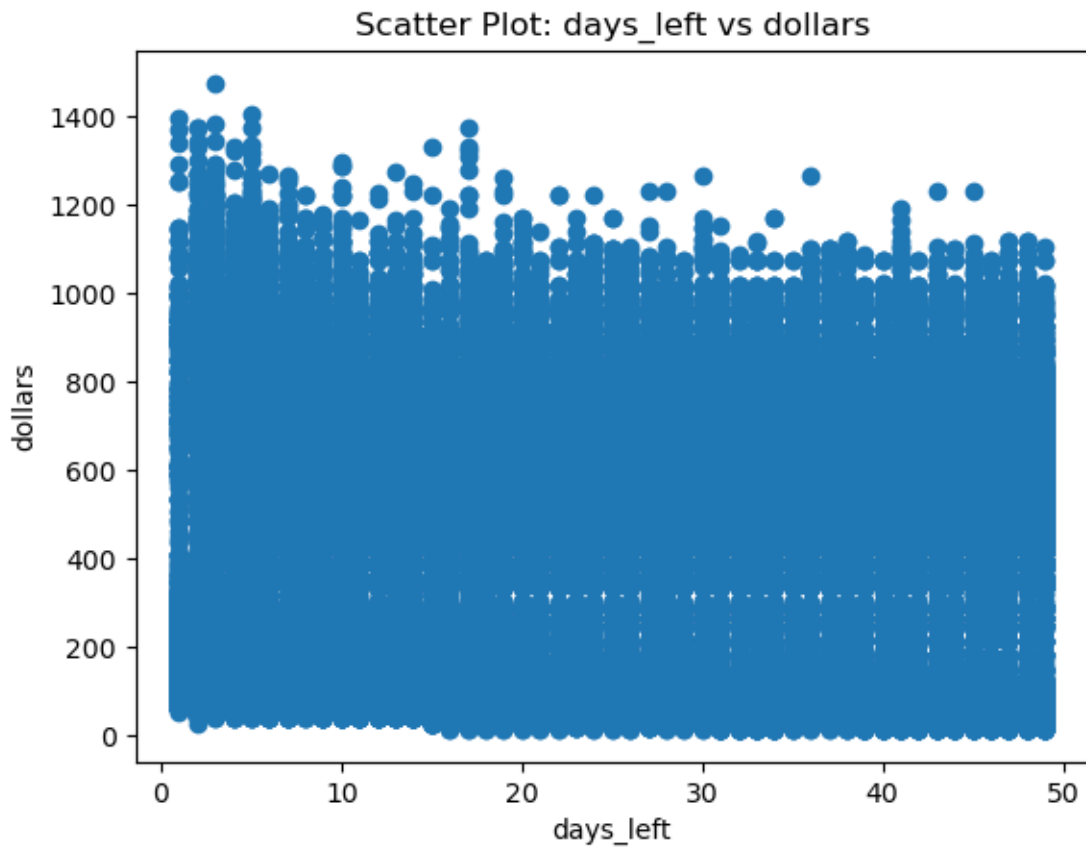
300150	Hyderabad	Business	13.83	49	949.188
300151	Hyderabad	Business	10.00	49	979.020
300152	Hyderabad	Business	10.08	49	979.020

[300153 rows x 10 columns]

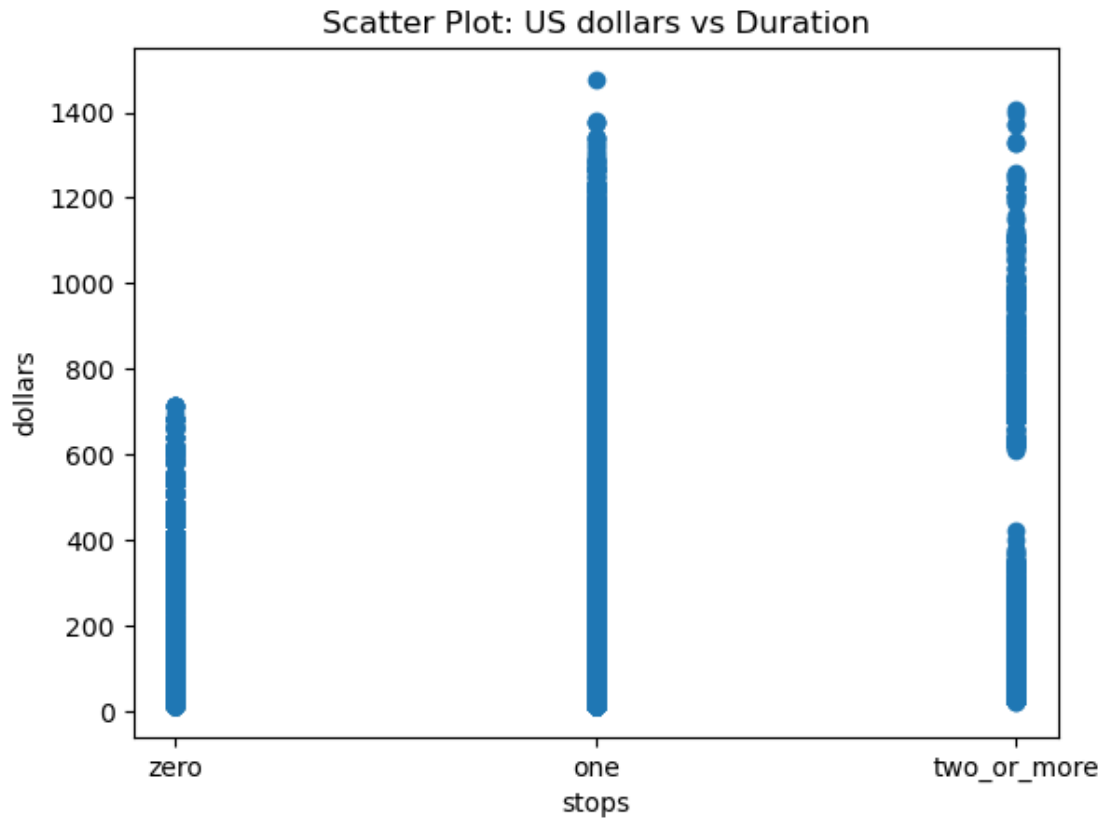
```
[14]: # Scatterplot
plt.scatter(flights['duration'], flights['dollars'])
plt.xlabel('duration')
plt.ylabel('dollars')
plt.title('Scatter Plot: US dollars vs Duration')
plt.show()
```



```
[15]: plt.scatter(flights['days_left'], flights['dollars'])
plt.xlabel('days_left')
plt.ylabel('dollars')
plt.title('Scatter Plot: days_left vs dollars')
plt.show()
```



```
[16]: plt.scatter(flights['stops'], flights['dollars'])  
plt.xlabel('stops')  
plt.ylabel('dollars')  
plt.title('Scatter Plot: US dollars vs Duration')  
plt.show()
```

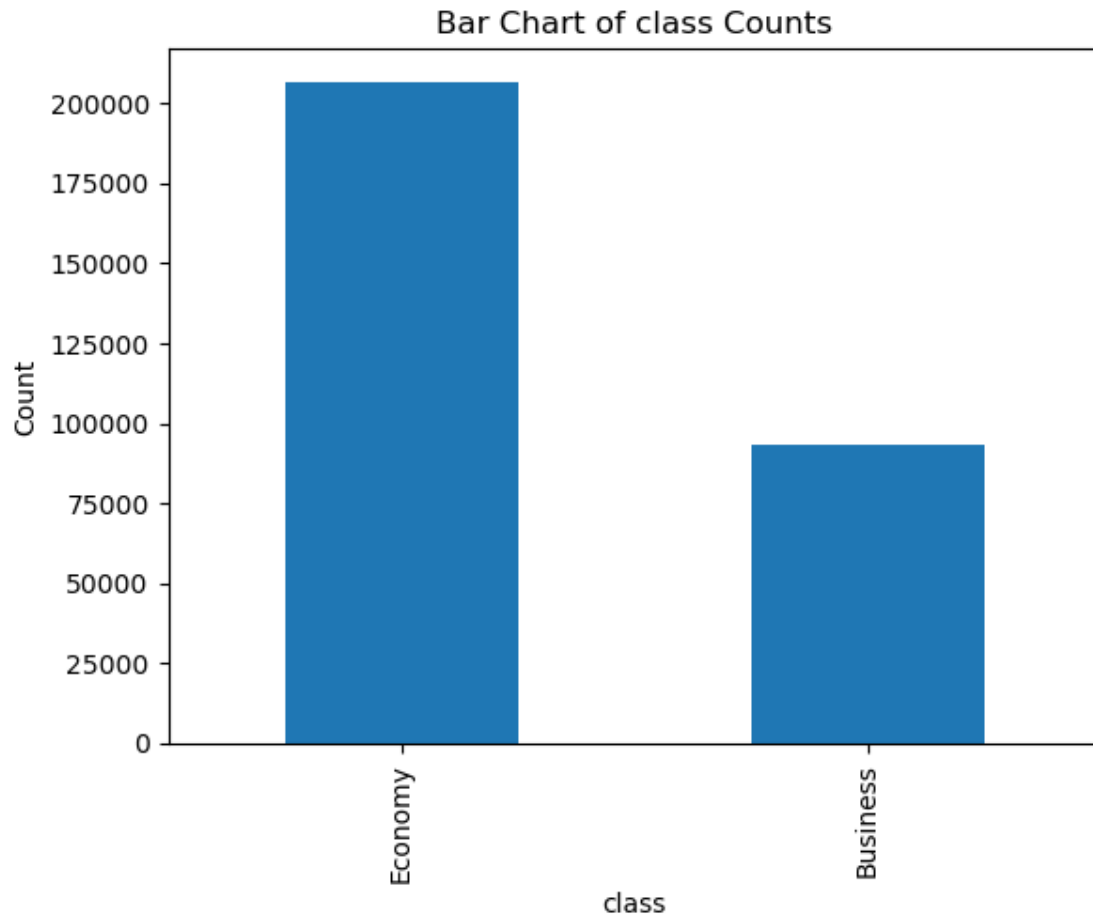


```
[17]: counts = flights['class'].value_counts()

# Create bar chart
counts.plot(kind='bar')

# Add labels and title
plt.xlabel('class')
plt.ylabel('Count')
plt.title('Bar Chart of class Counts')

# Show the plot
plt.show()
```

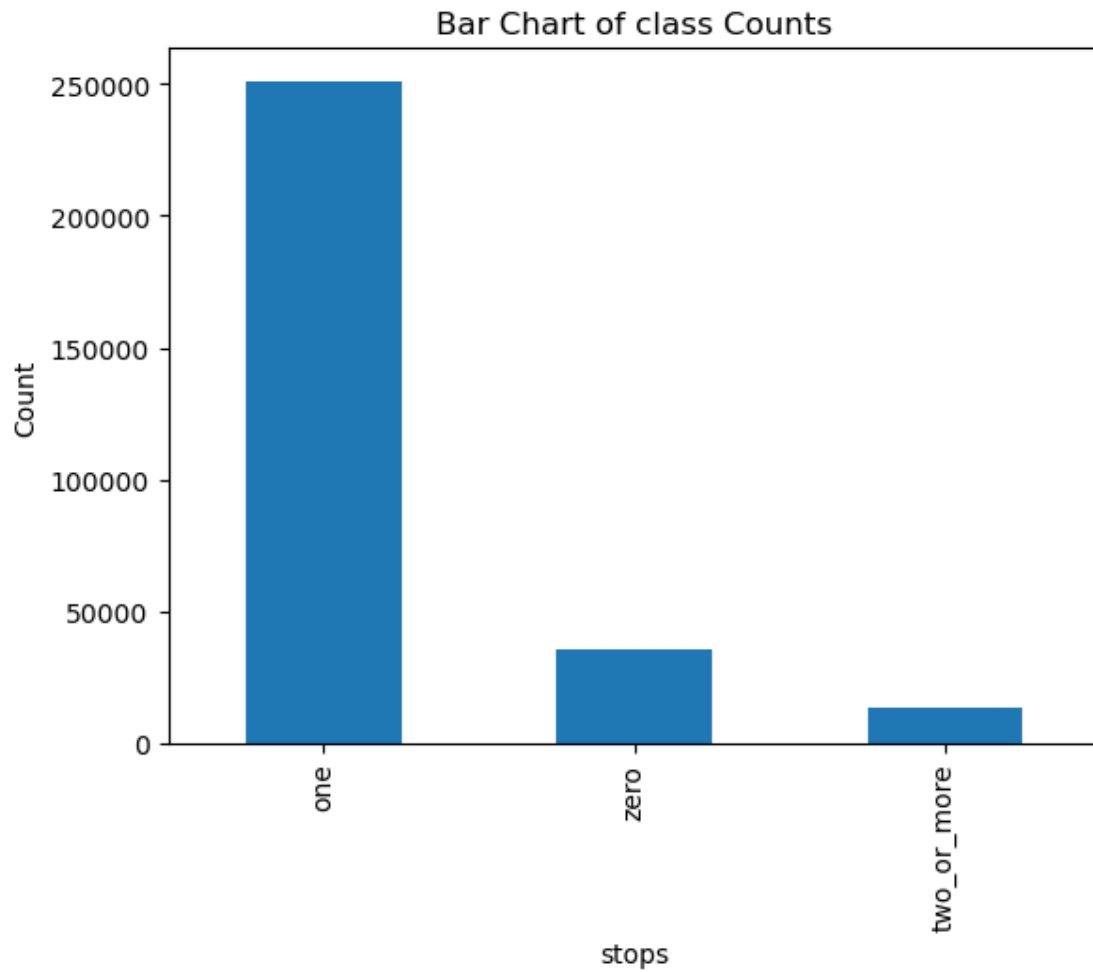



```
[18]: counts = flights['stops'].value_counts()

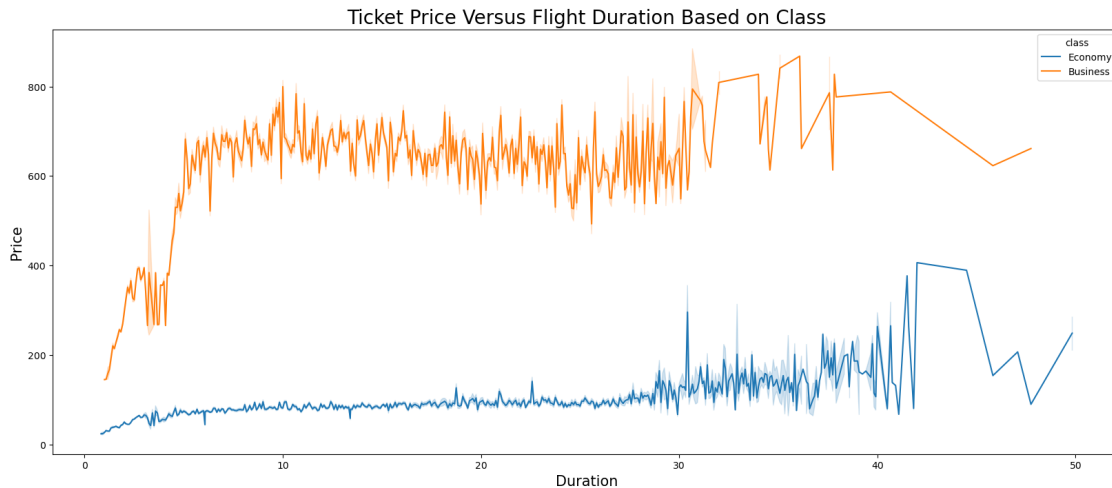
# Create bar chart
counts.plot(kind='bar')

# Add labels and title
plt.xlabel('stops')
plt.ylabel('Count')
plt.title('Bar Chart of class Counts')

# Show the plot
plt.show()
```



```
[19]: plt.figure(figsize=(20,8))
sns.lineplot(data=flights,x='duration',y='dollars',hue = 'class')
plt.title('Ticket Price Versus Flight Duration Based on Class',fontsize=20)
plt.xlabel('Duration',fontsize=15)
plt.ylabel('Price',fontsize=15)
plt.show()
```



```
[20]: grouped_data = flights.groupby(['destination_city', 'class']).sum().unstack()
grouped_data
```

```
[20]:
```

		airline \
class		Business
destination_city		
Bangalore	Air_IndiaAir_IndiaAir_IndiaAir_IndiaAir_IndiaV...	
Chennai	Air_IndiaVistaraAir_IndiaAir_IndiaAir_IndiaAir...	
Delhi	Air_IndiaAir_IndiaAir_IndiaAir_IndiaVistaraVis...	
Hyderabad	Air_IndiaAir_IndiaVistaraVistaraAir_IndiaAir_I...	
Kolkata	Air_IndiaAir_IndiaVistaraAir_IndiaAir_IndiaAir...	
Mumbai	Air_IndiaAir_IndiaAir_IndiaAir_IndiaAir_IndiaV...	

		Economy
class		
destination_city		
Bangalore	SpiceJetAirAsiaVistaraVistaraVistaraVistaraVis...	
Chennai	SpiceJetVistaraVistaraIndigoIndigoAir_IndiaAir...	
Delhi	VistaraVistaraGO_FIRSTIndigoIndigoAir_IndiaAir...	
Hyderabad	VistaraIndigoIndigoIndigoAir_IndiaAir_IndiaAir...	
Kolkata	GO_FIRSTGO_FIRSTIndigoIndigoGO_FIRSTGO_FIRSTIn...	
Mumbai	SpiceJetSpiceJetAirAsiaVistaraVistaraVistaraVi...	

		source_city \
class		Business
destination_city		
Bangalore	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...	
Chennai	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...	
Delhi	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...	
Hyderabad	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...	

Kolkata	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Mumbai	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...

class	Economy
destination_city	
Bangalore	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Chennai	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Delhi	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...
Hyderabad	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Kolkata	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Mumbai	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...

	departure_time
class	Business
destination_city	
Bangalore	EveningNightNightEveningEveningEveningEveningE...
Chennai	NightEveningEveningNightEveningEveningEveningE...
Delhi	EveningNightEveningEveningEveningEveningNightN...
Hyderabad	EveningNightNightEveningEveningNightEveningNig...
Kolkata	NightEveningEveningNightEveningNightEveningNig...
Mumbai	EveningEveningEveningNightEveningEveningNightE...

class	Economy
destination_city	
Bangalore	Early_MorningMorningMorningMorningAfternoonEve...
Chennai	Early_MorningEarly_MorningAfternoonMorningAfte...
Delhi	MorningEveningEveningEveningAfternoonEveningNi...
Hyderabad	MorningMorningMorningAfternoonMorningMorningEv...
Kolkata	Early_MorningEveningEarly_MorningEarly_Morning...
Mumbai	EveningEarly_MorningEarly_MorningMorningMornin...

	stops
class	Business
destination_city	
Bangalore	zerozerooneoneoneonezerooneonezerooneoneoneone...
Chennai	zerozerooneoneoneoneoneoneoneoneoneoneoneonetw...
Delhi	zerozerooneonezerozerooneoneoneoneoneoneoneone...
Hyderabad	zerozerozerozerooneoneoneoneoneoneoneoneoneone...
Kolkata	zerozerozerooneoneoneoneoneoneoneoneoneoneoneo...
Mumbai	zerozerooneoneoneonezerooneoneonezerooneoneone...

class	Economy
destination_city	
Bangalore	zerozerozerozerozerozerozerozerozerozerozeroze...

Chennai	zerozerozerozerozerozerozerozerooneoneoneoneze...
Delhi	zerozerozerozerozerozerozerozerooneonezerozero...
Hyderabad	zerozerozerozerozerozerozerooneonezerozerozero...
Kolkata	zerozerozerozerooneoneoneonezerozerozerozeroze...
Mumbai	zerozerozerozerozerozerozerozerozerozerozeroze...

	arrival_time \
class	Business
destination_city	
Bangalore	NightLate_NightAfternoonAfternoonMorningEvenin...
Chennai	NightNightEarly_MorningNightEarly_MorningMorni...
Delhi	NightNightMorningNightNightNightAfternoonEveni...
Hyderabad	EveningNightNightEveningAfternoonEarly_Morning...
Kolkata	NightEveningEveningMorningMorningAfternoonNigh...
Mumbai	EveningNightNightNightNightNightAfternoonEveni...

class	Economy \
destination_city	
Bangalore	MorningAfternoonMorningMorningEveningEveningNi...
Chennai	MorningMorningEveningMorningAfternoonMorningAf...
Delhi	AfternoonEveningNightNightAfternoonNightNightN...
Hyderabad	AfternoonAfternoonMorningAfternoonMorningMorni...
Kolkata	Early_MorningNightMorningMorningEarly_MorningE...
Mumbai	NightMorningEarly_MorningAfternoonMorningAfter...

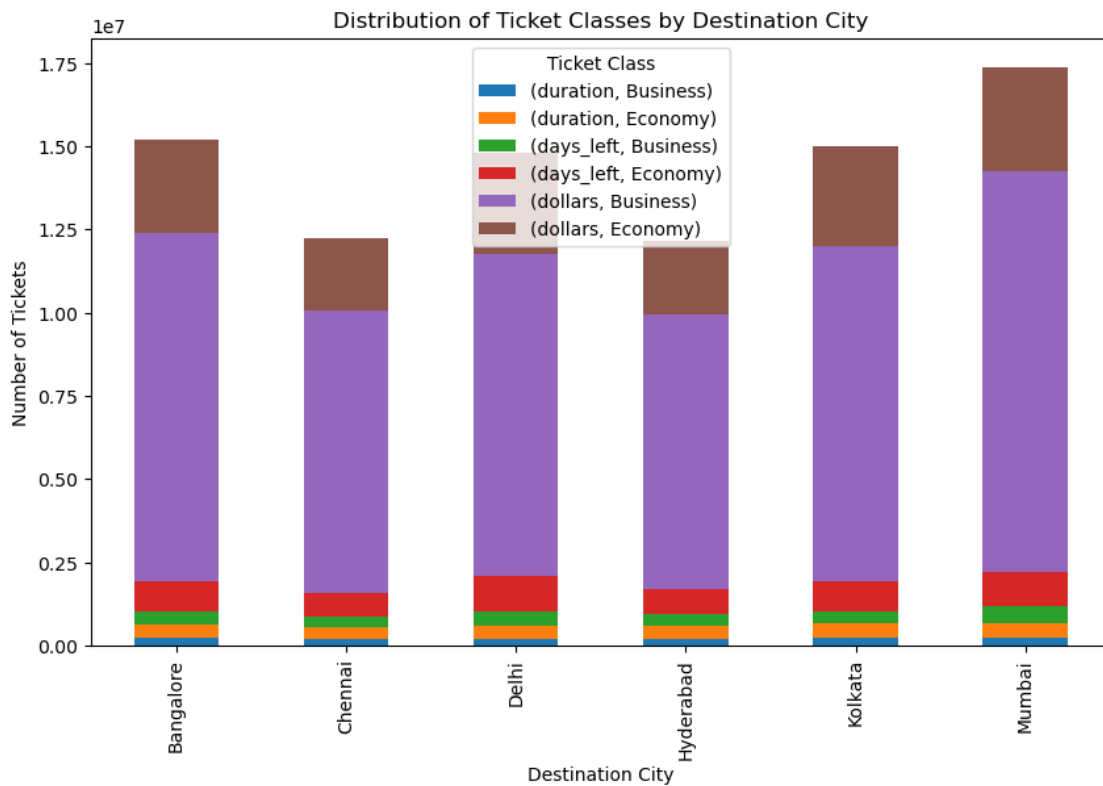
	duration		days_left		dollars \
class	Business	Economy	Business	Economy	Business
destination_city					
Bangalore	221787.05	393992.87	415745	914504	1.046375e+07
Chennai	197240.94	341223.79	341263	709130	8.465924e+06
Delhi	193921.23	409122.21	428920	1072687	9.640890e+06
Hyderabad	204660.31	367096.68	355120	755934	8.275319e+06
Kolkata	228157.72	426431.74	380447	901871	1.004807e+07
Mumbai	235404.20	449137.32	485034	1044749	1.204781e+07

class	Economy
destination_city	
Bangalore	2769368.280
Chennai	2168616.888
Delhi	3049506.048
Hyderabad	2198188.560
Kolkata	3004862.592
Mumbai	3108814.164

```
[21]: grouped_data.plot(kind='bar', stacked=True, figsize=(10, 6))

# Add labels and title
plt.xlabel('Destination City')
plt.ylabel('Number of Tickets')
plt.title('Distribution of Ticket Classes by Destination City')

# Show plot
plt.legend(title='Ticket Class')
plt.show()
```



```
[22]: df = flights.groupby(['airline', 'class']).sum().unstack()
df
```

```
[22]:
```

class	source_city \
airline	Business
AirAsia	NaN
Air_India	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
GO_FIRST	NaN
Indigo	NaN
SpiceJet	NaN

Vistara	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
---------	---

class	Economy
airline	
AirAsia	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Air_India	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
GO_FIRST	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Indigo	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
SpiceJet	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...
Vistara	DelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiDelhiD...

	departure_time
class	Business
airline	
AirAsia	NaN
Air_India	EveningEveningEveningNightEveningNightEveningN...
GO_FIRST	NaN
Indigo	NaN
SpiceJet	NaN
Vistara	EveningNightEveningEveningEveningEveningEvenin...

class	Economy
airline	
AirAsia	Early_MorningEveningEveningMorningEveningEarly...
Air_India	Early_MorningEarly_MorningEveningEarly_Morning...
GO_FIRST	Early_MorningAfternoonAfternoonMorningEarly_Mo...
Indigo	Early_MorningMorningAfternoonMorningMorningEve...
SpiceJet	EveningEarly_MorningEveningEveningEveningEveni...
Vistara	MorningMorningMorningMorningAfternoonAfternoon...

	stops
class	Business
airline	
AirAsia	NaN
Air_India	zerozerooneoneoneoneoneoneoneoneoneoneoneoneon...
GO_FIRST	NaN
Indigo	NaN
SpiceJet	NaN
Vistara	zeroonezerooneoneoneoneoneoneoneoneoneoneoneon...

class	Economy
airline	
AirAsia	zerooneoneoneoneonezerooneoneonezerooneoneonezero...
Air_India	zerozerooneoneoneoneoneoneoneoneoneoneoneoneon...

GO_FIRST	zerozerozerozerooneoneoneoneoneoneoneoneonezeroze...
Indigo	zerozerozerozerozerozerozerozerooneoneoneoneoneone...
SpiceJet	zerozerozerooneoneoneoneonezerozerozerozerooneoneoneo...
Vistara	zerozerozerozerozerozerozerozerooneoneoneoneoneonezeroone...

	arrival_time \
class	Business
airline	
AirAsia	NaN
Air_India	EveningNightNightNightNightAfternoonEveningMor...
GO_FIRST	NaN
Indigo	NaN
SpiceJet	NaN
Vistara	NightAfternoonNightNightMorningNightMorningNig...

	Economy
class	
airline	
AirAsia	Early_MorningEarly_MorningMorningAfternoonEarl...
Air_India	MorningMorningNightAfternoonMorningMorningMorn...
GO_FIRST	MorningEveningEveningAfternoonEveningNightMorn...
Indigo	MorningAfternoonEveningMorningMorningEveningNi...
SpiceJet	NightMorningNightNightMorningAfternoonMorningM...
Vistara	AfternoonMorningAfternoonMorningEveningEvening...

	destination_city \
class	Business
airline	
AirAsia	NaN
Air_India	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...
GO_FIRST	NaN
Indigo	NaN
SpiceJet	NaN
Vistara	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...

	Economy	duration \	Business
class			
airline			
AirAsia	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...		NaN
Air_India	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...	481459.50	
GO_FIRST	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...		NaN
Indigo	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...		NaN
SpiceJet	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...		NaN
Vistara	MumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumbaiMumb...	799711.95	

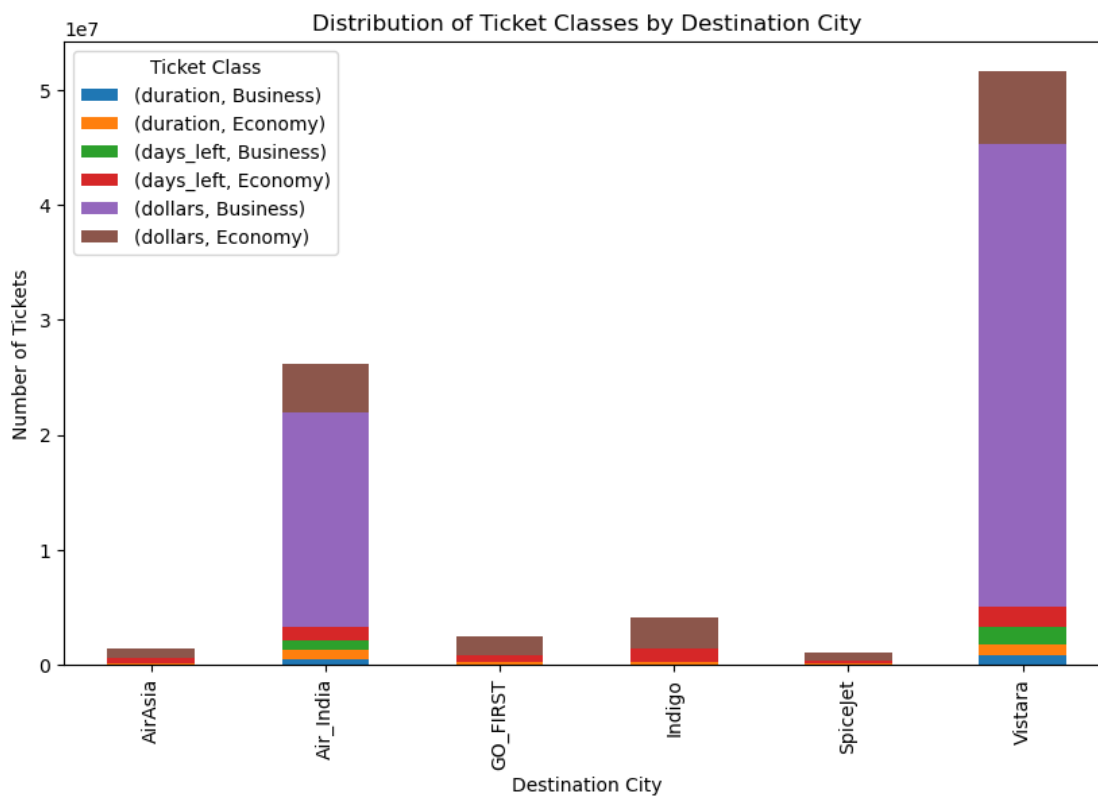
	days_left	dollars
class	Economy Business Economy	Business Economy

airline					
AirAsia	143943.72	NaN	446481.0	NaN	790297.068
Air_India	772709.10	830615.0	1231926.0	1.860620e+07	4212154.344
GO_FIRST	202888.42	NaN	635645.0	NaN	1571687.664
Indigo	249888.88	NaN	1132517.0	NaN	2754962.484
SpiceJet	113356.28	NaN	217371.0	NaN	668177.784
Vistara	904218.21	1575914.0	1734935.0	4.033557e+07	6302077.188

```
[23]: df.plot(kind='bar', stacked=True, figsize=(10, 6))

# Add labels and title
plt.xlabel('Destination City')
plt.ylabel('Number of Tickets')
plt.title('Distribution of Ticket Classes by Destination City')

# Show plot
plt.legend(title='Ticket Class')
plt.show()
```



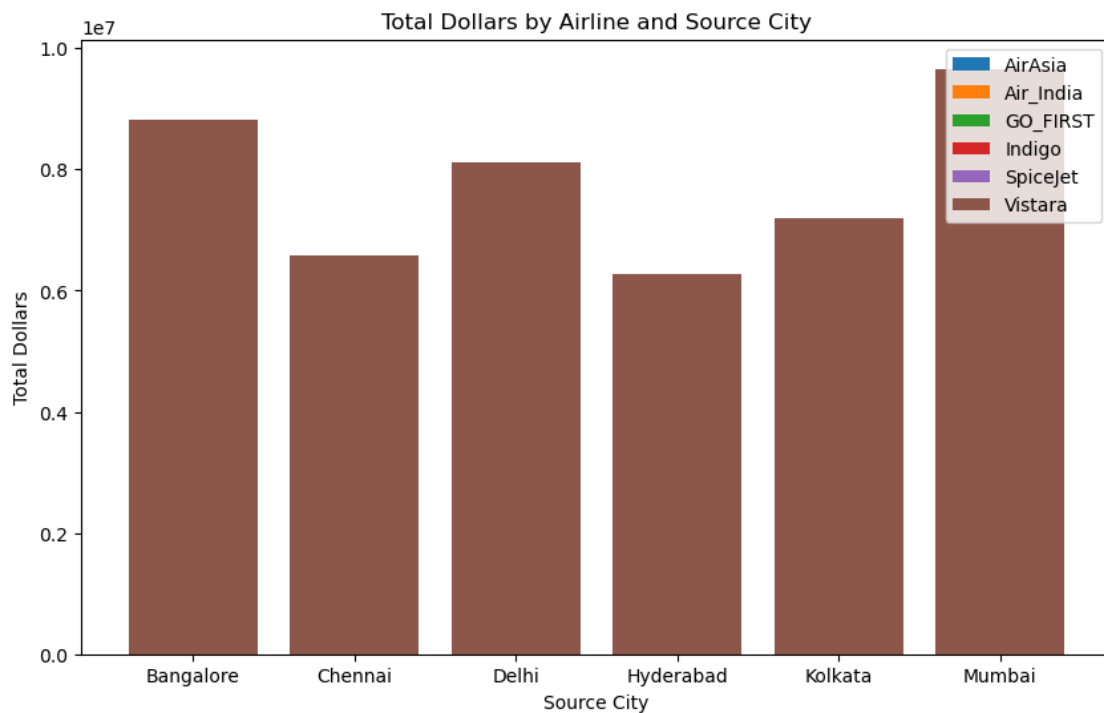
```
[24]: airlines = flights.groupby(['airline', 'source_city'])['dollars'].sum().
      ↪reset_index()
```

```

# Create a bar plot
plt.figure(figsize=(10, 6))
for airline in airlines['airline'].unique():
    subset = airlines[airlines['airline'] == airline]
    plt.bar(subset['source_city'], subset['dollars'], label=airline)

plt.xlabel('Source City')
plt.ylabel('Total Dollars')
plt.title('Total Dollars by Airline and Source City')
plt.legend()
plt.show()

```



```

[25]: from scipy.stats import zscore

# Calculate Z-scores
flights['Z_Score'] = zscore(flights['duration'])

# Filter outliers
df_filtered = flights[flights['Z_Score'].abs() <= 3]
df_filtered

```

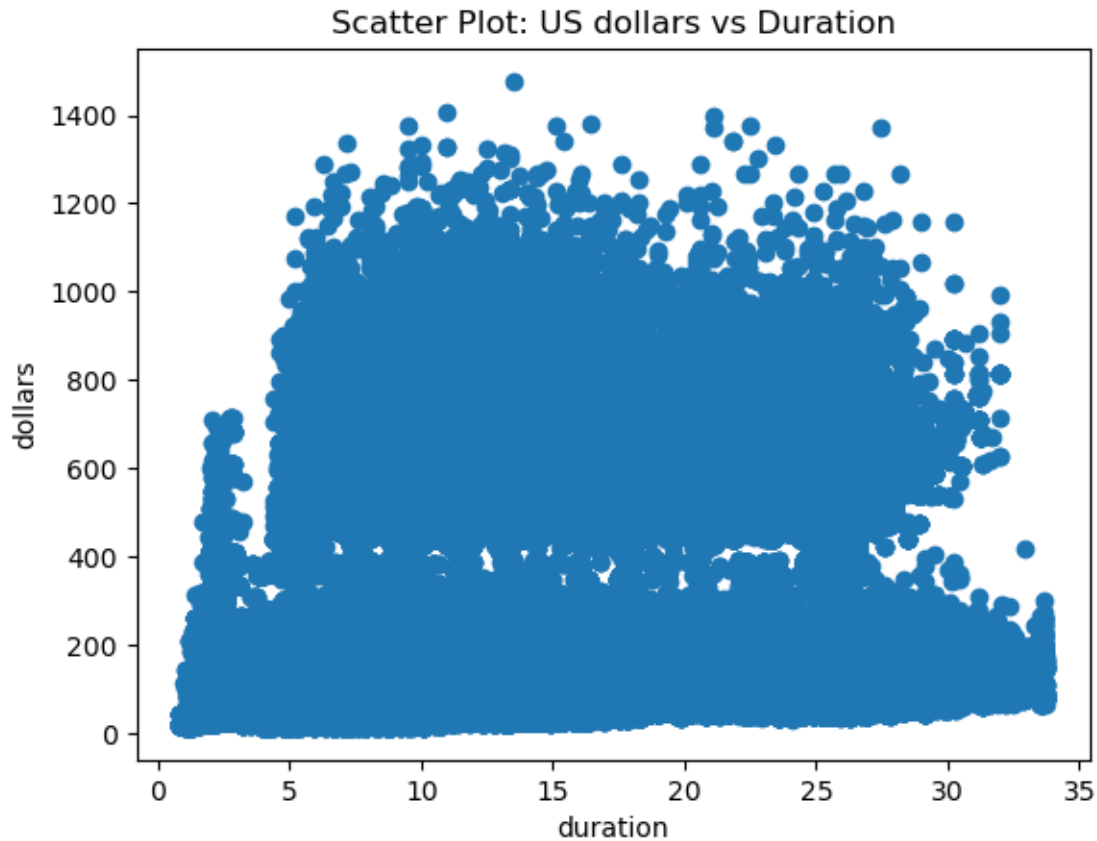
```
[25]:
```

	airline	source_city	departure_time	stops	arrival_time	\
0	SpiceJet	Delhi	Evening	zero	Night	
1	SpiceJet	Delhi	Early_Morning	zero	Morning	
2	AirAsia	Delhi	Early_Morning	zero	Early_Morning	
3	Vistara	Delhi	Morning	zero	Afternoon	
4	Vistara	Delhi	Morning	zero	Morning	
...	
300148	Vistara	Chennai	Morning	one	Evening	
300149	Vistara	Chennai	Afternoon	one	Night	
300150	Vistara	Chennai	Early_Morning	one	Night	
300151	Vistara	Chennai	Early_Morning	one	Evening	
300152	Vistara	Chennai	Morning	one	Evening	

	destination_city	class	duration	days_left	dollars	Z_Score
0	Mumbai	Economy	2.17	1	71.436	-1.397531
1	Mumbai	Economy	2.33	1	71.436	-1.375284
2	Mumbai	Economy	2.17	1	71.472	-1.397531
3	Mumbai	Economy	2.25	1	71.460	-1.386407
4	Mumbai	Economy	2.33	1	71.460	-1.375284
...
300148	Hyderabad	Business	10.08	49	831.180	-0.297695
300149	Hyderabad	Business	10.42	49	925.260	-0.250421
300150	Hyderabad	Business	13.83	49	949.188	0.223718
300151	Hyderabad	Business	10.00	49	979.020	-0.308819
300152	Hyderabad	Business	10.08	49	979.020	-0.297695

[299431 rows x 11 columns]

```
[26]: plt.scatter(df_filtered['duration'],df_filtered['dollars'])
plt.xlabel('duration')
plt.ylabel('dollars')
plt.title('Scatter Plot: US dollars vs Duration')
plt.show()
```



```
[27]: df_filtered.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 299431 entries, 0 to 300152
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   airline                299431 non-null object
1   source_city            299431 non-null object
2   departure_time         299431 non-null object
3   stops                 299431 non-null object
4   arrival_time           299431 non-null object
5   destination_city       299431 non-null object
6   class                 299431 non-null object
7   duration               299431 non-null float64
8   days_left              299431 non-null int64
9   dollars                299431 non-null float64
10  Z_Score                299431 non-null float64
dtypes: float64(3), int64(1), object(7)
memory usage: 27.4+ MB
```

```
[28]: flights.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300153 entries, 0 to 300152
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   airline                300153 non-null object
1   source_city            300153 non-null object
2   departure_time         300153 non-null object
3   stops                 300153 non-null object
4   arrival_time           300153 non-null object
5   destination_city       300153 non-null object
6   class                 300153 non-null object
7   duration               300153 non-null float64
8   days_left             300153 non-null int64
9   dollars               300153 non-null float64
10  Z_Score               300153 non-null float64
dtypes: float64(3), int64(1), object(7)
memory usage: 25.2+ MB
```

```
[29]: flight_status = df_filtered.drop(['Z_Score'], axis = True)
```

```
[30]: flight_status
```

```
[30]:
```

	airline	source_city	departure_time	stops	arrival_time	\
0	SpiceJet	Delhi	Evening	zero	Night	
1	SpiceJet	Delhi	Early_Morning	zero	Morning	
2	AirAsia	Delhi	Early_Morning	zero	Early_Morning	
3	Vistara	Delhi	Morning	zero	Afternoon	
4	Vistara	Delhi	Morning	zero	Morning	
...	
300148	Vistara	Chennai	Morning	one	Evening	
300149	Vistara	Chennai	Afternoon	one	Night	
300150	Vistara	Chennai	Early_Morning	one	Night	
300151	Vistara	Chennai	Early_Morning	one	Evening	
300152	Vistara	Chennai	Morning	one	Evening	

	destination_city	class	duration	days_left	dollars
0	Mumbai	Economy	2.17	1	71.436
1	Mumbai	Economy	2.33	1	71.436
2	Mumbai	Economy	2.17	1	71.472
3	Mumbai	Economy	2.25	1	71.460
4	Mumbai	Economy	2.33	1	71.460
...
300148	Hyderabad	Business	10.08	49	831.180
300149	Hyderabad	Business	10.42	49	925.260
300150	Hyderabad	Business	13.83	49	949.188

300151	Hyderabad	Business	10.00	49	979.020
300152	Hyderabad	Business	10.08	49	979.020

[299431 rows x 10 columns]

```
[31]: num_nans = flight_status.isna().sum()
      num_nans
```

```
[31]: airline          0
      source_city      0
      departure_time   0
      stops            0
      arrival_time     0
      destination_city 0
      class            0
      duration         0
      days_left        0
      dollars          0
      dtype: int64
```

```
[32]: flight_pred = pd.DataFrame(flight_status)

# Separate features (X) and target variable (y)
X = flight_status[['airline', 'source_city', 'departure_time', 'stops',
    ↪ 'arrival_time', 'destination_city', 'class', 'duration', 'days_left']]
y = flight_status['dollars']

# Split the dataset into training and testing sets (80% train, 20% test)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
    ↪ random_state=42)

# Display the shapes of the training and testing sets
print("Shapes of training and testing sets:")
print("X_train:", X_train.shape)
print("X_test:", X_test.shape)
print("y_train:", y_train.shape)
print("y_test:", y_test.shape)
```

Shapes of training and testing sets:

```
X_train: (239544, 9)
X_test: (59887, 9)
y_train: (239544,)
y_test: (59887,)
```

```
[33]: # Define the preprocessing steps for numerical and categorical data
      numeric_features = ['duration', 'days_left']
      numeric_transformer = StandardScaler()
```

```

categorical_features = ['airline', 'source_city', 'departure_time', 'stops', '
    ↪arrival_time', 'destination_city', 'class']
categorical_transformer = OneHotEncoder(drop='first')

# Combine the preprocessing steps
preprocessor = ColumnTransformer(
    transformers=[
        ('num', numeric_transformer, numeric_features),
        ('cat', categorical_transformer, categorical_features)
    ])

# Create a pipeline that first transforms the data, then fits the model
model_pipeline = Pipeline(steps=[
    ('preprocessor', preprocessor),
    ('regressor', LinearRegression())
])

# Train the model
model_pipeline.fit(X_train, y_train)

# Make predictions on the test set
y_pred = model_pipeline.predict(X_test)

# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)

print(f"Mean Squared Error: {mse}")
print(f"R^2 Score: {r2}")

```

Mean Squared Error: 6608.2335637394

R^2 Score: 0.9105275221250047

```

[34]: df2 = flights
column=['airline','source_city','departure_time','stops','arrival_time','destination_city','cl

df2[column] = df2[column].apply(LabelEncoder().fit_transform)
x=df2.drop(['dollars'],axis=1)
y=df2['dollars']

x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.
    ↪2,random_state=42)
print('x_train size',x_train.shape)
print('y_train size',y_train.shape)
print('x_test size',x_test.shape)
print('y_test size',y_test.shape)

```

x_train size (240122, 10)

y_train size (240122,)

```
x_test size (60031, 10)
y_test size (60031,)
```

```
[35]: from sklearn.ensemble import RandomForestRegressor
model = RandomForestRegressor(random_state=42)
model.fit(x_train, y_train)

y_pred2 = model.predict(x_test)

mse = mean_squared_error(y_test, y_pred2)
r2 = r2_score(y_test, y_pred2)

print("Random Forest Regression Results:")
print(f"Mean Squared Error: {mse}")
print(f"R^2 Score: {r2}")
```

```
Random Forest Regression Results:
Mean Squared Error: 1118.4422787940296
R^2 Score: 0.9849326336371773
```

```
[36]: from sklearn.svm import SVR
from sklearn.metrics import mean_squared_error
svm_model = SVR(kernel='linear')
svm_model.fit(x_train, y_train)
y_pred3 = svm_model.predict(x_test)
mse = mean_squared_error(y_test, y_pred3)
r3 = r2_score(y_test, y_pred3)
print("Mean Squared Error:", mse)
print(f"R^2 Score: {r3}")
```

```
Mean Squared Error: 7557.682114415242
R^2 Score: 0.8981848527807508
```

```
[37]: import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import Lasso
from sklearn.metrics import mean_squared_error, r2_score

lasso_model = Lasso(alpha=0.1, random_state=42)

lasso_model.fit(x_train, y_train)
y_pred4 = lasso_model.predict(x_test)
mse = mean_squared_error(y_test, y_pred4)
r4 = r2_score(y_test, y_pred4)
print("Mean Squared Error:", mse)
print(f"R^2 Score: {r4}")
```

```
Mean Squared Error: 7084.74906922515
```


R² Score: 0.9045560849775951

[]: