

```
** Title:- Airlines Flight data analysis
```

```
#First we need to import the necessary libraries
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# import the CSV Dataset
```

```
data = pd.read_csv("C:\\Users\\david\\Downloads\\
airlines_flights_data.csv")
data
```

	index	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]
```

```
pip install mysql-connector-python
```

```
Requirement already satisfied: mysql-connector-python in c:\users\
david\anaconda3\lib\site-packages (9.4.0)
```

```
Note: you may need to restart the kernel to use updated packages.
```

```
import mysql.connector
```

```
try:
```

```
    connection = mysql.connector.connect(
        host='localhost',      # Your MySQL host, e.g., localhost
        user='root',           # Your MySQL username
        password='Jyothi@1110', # Your MySQL password
                                # Your database name)
    )
```

```
    if connection.is_connected():
```

```
        print("Successfully connected to MySQL database")
```

```
except mysql.connector.Error as err:
```

```
    print(f"Error: {err}")
```

```
Successfully connected to MySQL database
```

```
data.head()
```

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

```
data.tail()
```

	index	airline	flight	source_city	departure_time	stops
arrival_time \						

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

	destination_city	class	duration	days_left	price
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

data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 300153 entries, 0 to 300152
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   index                 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
```

data.describe()

	index	duration	days_left	price
count	300153.000000	300153.000000	300153.000000	300153.000000
mean	150076.000000	12.221021	26.004751	20889.660523
std	86646.852011	7.191997	13.561004	22697.767366
min	0.000000	0.830000	1.000000	1105.000000
25%	75038.000000	6.830000	15.000000	4783.000000
50%	150076.000000	11.250000	26.000000	7425.000000

75%	225114.000000	16.170000	38.000000	42521.000000
max	300152.000000	49.830000	49.000000	123071.000000

```
data.isnull().sum()
```

```
index          0
airline        0
flight         0
source_city    0
departure_time 0
stops         0
arrival_time   0
destination_city 0
class         0
duration       0
days_left     0
price         0
dtype: int64
```

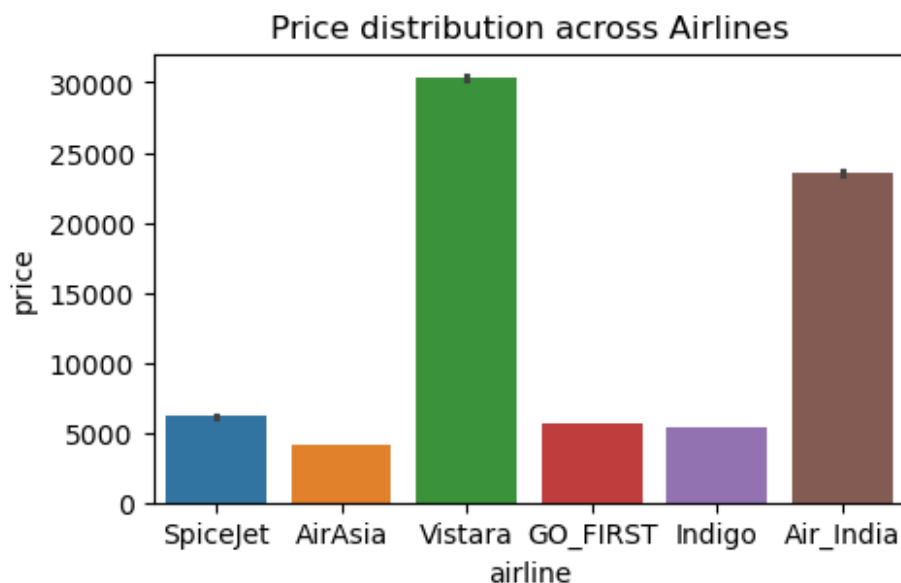
```
data[data.duplicated()]
```

```
Empty DataFrame
```

```
Columns: [index, airline, flight, source_city, departure_time, stops,
arrival_time, destination_city, class, duration, days_left, price]
Index: []
```

```
# 1)price distribution across Airlines
```

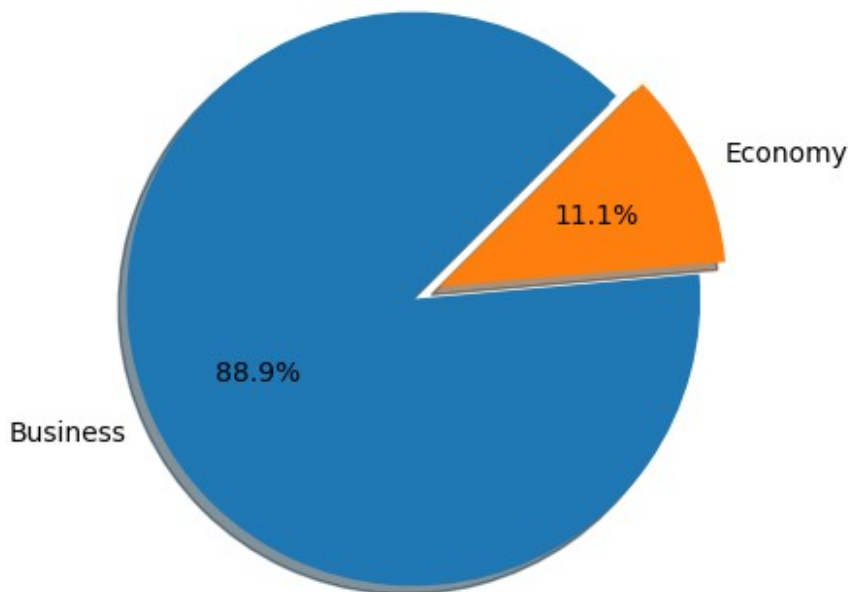
```
plt.figure(figsize=(5, 3))
sns.barplot(data=data, x="airline", y="price", hue="airline")
plt.title("Price distribution across Airlines")
plt.show()
```



```
# 2)Average Price: Economy vs. Business Class
Average_price = data.groupby('class')['price'].mean().reset_index()
values = Average_price['price']
labels = Average_price['class']
explode = [0.1, 0]

plt.pie(values,labels=labels,autopct='%1.1f%%',startangle=45,shadow=True,explode=explode)
plt.title("Average Price: Economy vs Business Class")
plt.show()
```

Average Price: Economy vs Business Class



```
# 3) Busiest Source & Destination Cities
#A)Top Source Cities

source_counts = data.groupby("destination_city")
["flight"].count().reset_index()
source_counts.rename(columns={"flight": "flight_count"}, inplace=True)

plt.figure(figsize=(6,3))
sns.barplot(x="destination_city", y="flight_count",
data=source_counts, palette="viridis")
plt.xticks(rotation=45)
plt.title("Busiest Destination Cities")
plt.show()
```

```
C:\Users\david\AppData\Local\Temp\ipykernel_7160\2846909472.py:8:  
FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

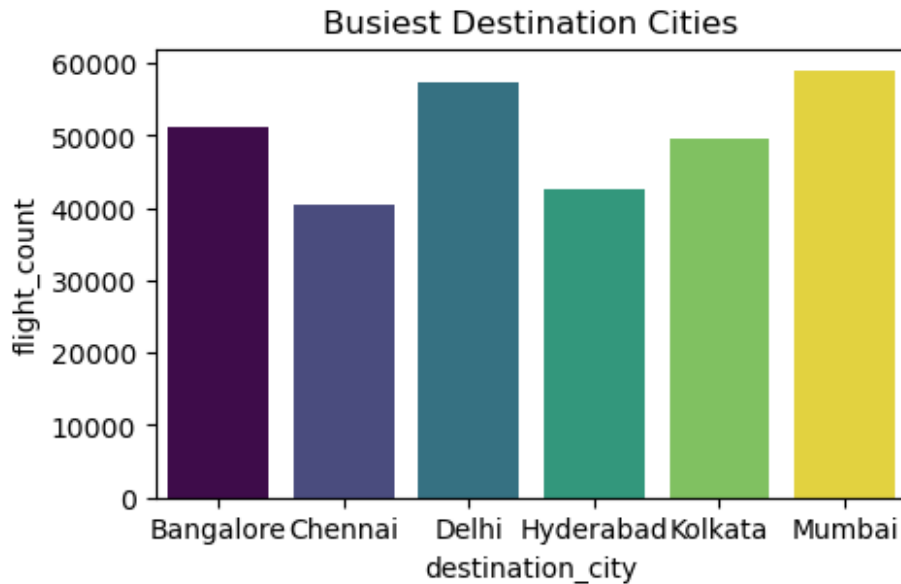
```
sns.barplot(x="destination_city", y="flight_count",  
data=source_counts, palette="viridis")
```



```
#3) Busiest Source & Destination Cities
```

```
#B) Top Destination Cities
```

```
dest_counts = data.groupby("destination_city")  
["flight"].count().reset_index()  
dest_counts.rename(columns={"flight": "flight_count"}, inplace=True)  
print(dest_counts.columns)  
  
Index(['destination_city', 'flight_count'], dtype='object')  
  
plt.figure(figsize=(5,3))  
sns.barplot(x="destination_city",y="flight_count",hue=dest_counts.index,  
x, legend=False, data=dest_counts,palette="viridis")  
plt.title("Busiest Destination Cities")  
plt.show()
```



4) Days Left vs. Ticket Price Trend

```
avg_price_by_days = data.groupby("days_left")["price"].mean()
```

--- Bar Plot ---

```
plt.figure(figsize=(15,6))
sns.barplot(x=avg_price_by_days.index,
y=avg_price_by_days.values,dodge=False, palette="Set2")
plt.title("Average Ticket Price by Days Left (Bar Plot)", fontsize=14,
fontweight="bold")
plt.xlabel("Days Left (Binned)")
plt.ylabel("Average Ticket Price (INR)")
plt.show()
```

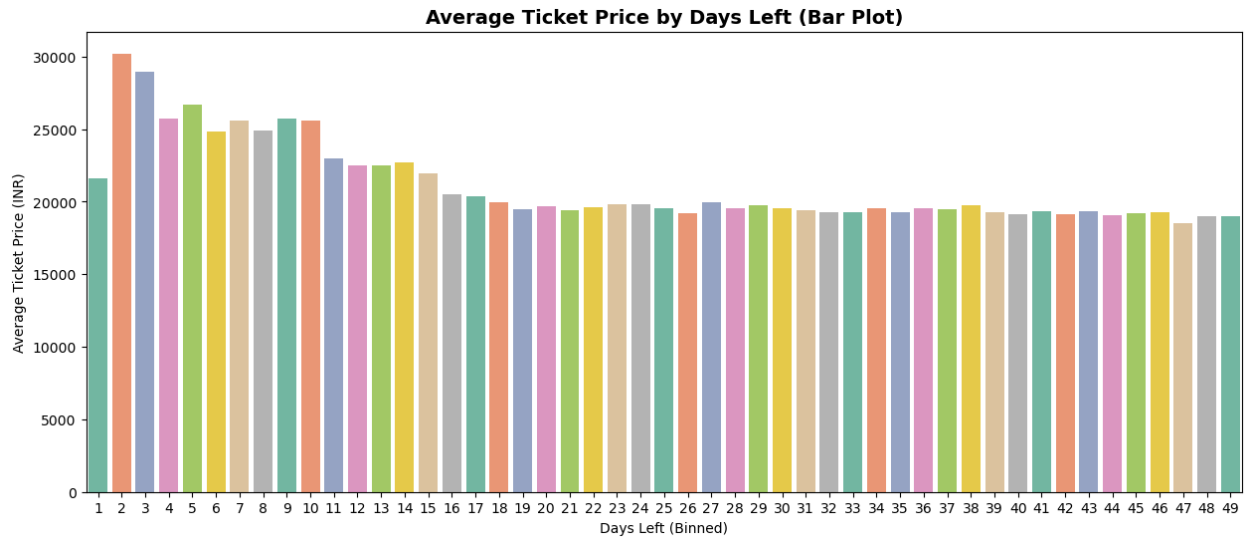
```
print("Average Ticket Price by Days Left Bin:\n")
```

```
#print(avg_price_by_days)
```

C:\Users\david\AppData\Local\Temp\ipykernel_25960\2776151662.py:5:
FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

```
sns.barplot(x=avg_price_by_days.index,
y=avg_price_by_days.values,dodge=False, palette="Set2")
```



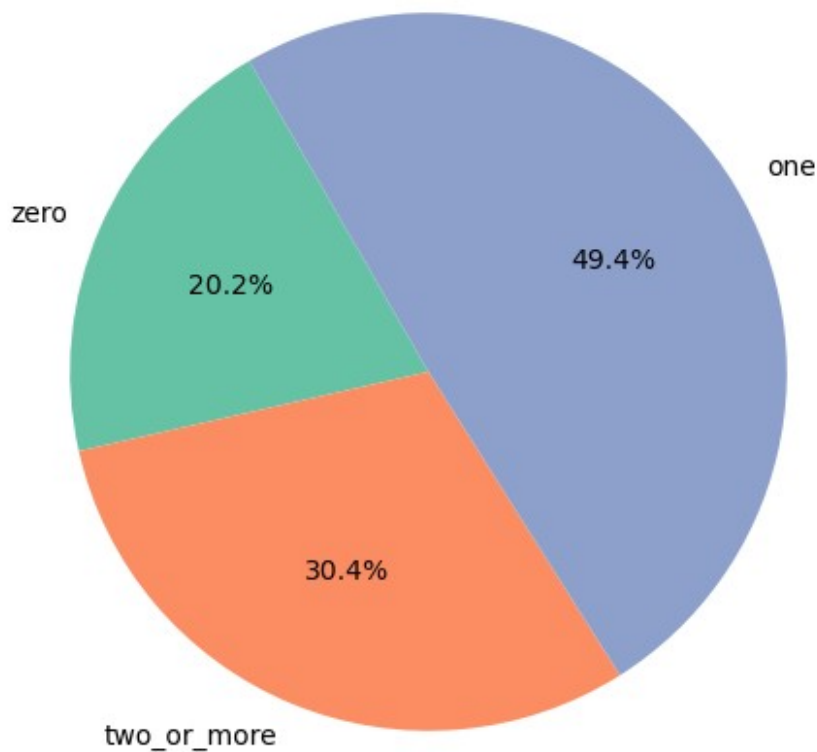
Average Ticket Price by Days Left Bin:

5) Stops vs. Price Comparison

```
plt.figure(figsize=(6,6))
plt.pie(
    avg_price_by_stops.values,
    labels=avg_price_by_stops.index,
    autopct="%1.1f%%",
    startangle=120,
    colors=sns.color_palette("Set2")
)
plt.title("Average Ticket Price by Stops (Pie Chart)", fontsize=14,
fontweight="bold")
plt.show()

print("Average Price by Stops:\n")
print(avg_price_by_stops)
```


Average Ticket Price by Stops (Pie Chart)



Average Price by Stops:

```
stops
zero      9375.938535
two_or_more 14113.450775
one      22900.992482
Name: price, dtype: float64
```

6)Duration Impact on Ticket Pricing

1. Create bins for flight durations

```
bins = [0, 2, 4, 6, 8, 12] # adjust according to your dataset
labels = ["0-2h", "2-4h", "4-6h", "6-8h", "8-12h"]
```

Create the duration_bin column

```
data["duration_bin"] = pd.cut(data["duration"], bins=bins,
labels=labels)
```

```
avg_price_by_duration = data.groupby("duration_bin")
["price"].mean().dropna()
```

```

# --- Pie Plot ---
plt.figure(figsize=(7,7))
plt.pie(
    avg_price_by_duration.values,
    labels=avg_price_by_duration.index,
    autopct="%1.1f%%",
    startangle=120,
    colors=sns.color_palette("Set2")
)
plt.title("Average Ticket Price Share by Flight Duration",
    fontsize=14, fontweight="bold")
plt.show()

```

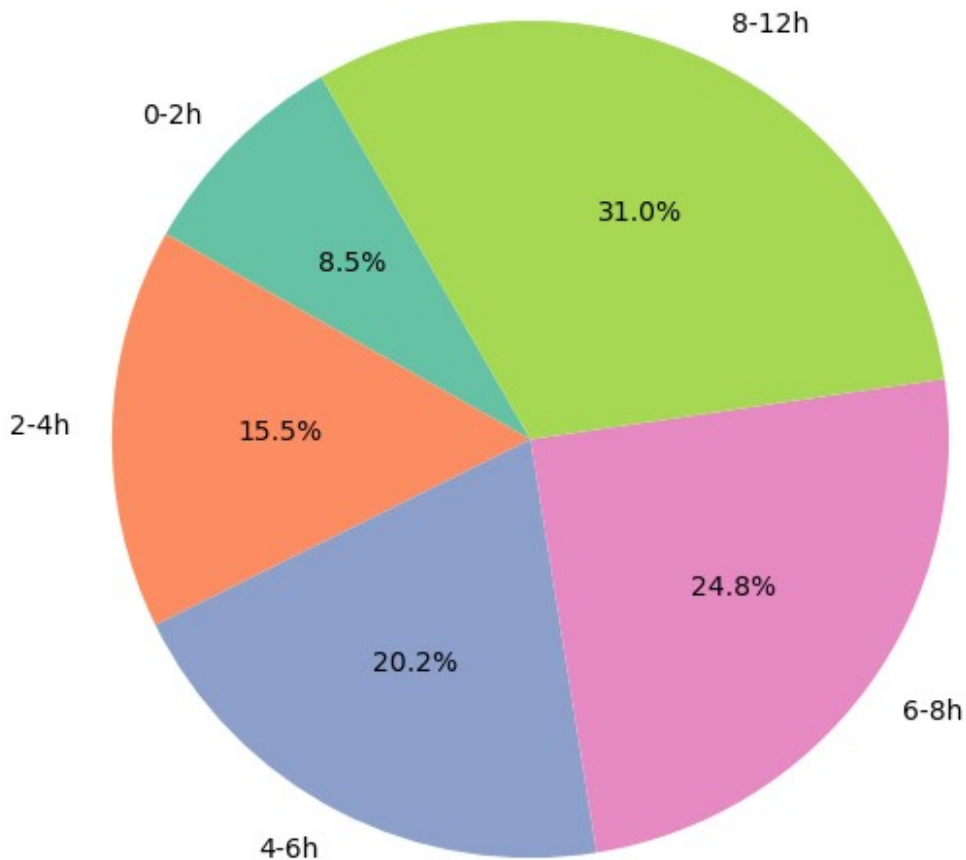
C:\Users\david\AppData\Local\Temp\ipykernel_25816\1675222407.py:10:
FutureWarning: The default of observed=False is deprecated and will be
changed to True in a future version of pandas. Pass observed=False to
retain current behavior or observed=True to adopt the future default
and silence this warning.

```

    avg_price_by_duration = data.groupby("duration_bin")
    ["price"].mean().dropna()

```

Average Ticket Price Share by Flight Duration



SQL Queries

```
!pip install pandasql
```

```
Requirement already satisfied: pandasql in c:\users\david\anaconda3\lib\site-packages (0.7.3)  
Requirement already satisfied: numpy in c:\users\david\anaconda3\lib\site-packages (from pandasql) (2.1.3)  
Requirement already satisfied: pandas in c:\users\david\anaconda3\lib\site-packages (from pandasql) (2.2.3)  
Requirement already satisfied: sqlalchemy in c:\users\david\anaconda3\lib\site-packages (from pandasql) (2.0.39)  
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\david\anaconda3\lib\site-packages (from pandas->pandasql) (2.9.0.post0)  
Requirement already satisfied: pytz>=2020.1 in c:\users\david\
```

```

anaconda3\lib\site-packages (from pandas->pandasql) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in c:\users\david\
anaconda3\lib\site-packages (from pandas->pandasql) (2025.2)
Requirement already satisfied: six>=1.5 in c:\users\david\anaconda3\
lib\site-packages (from python-dateutil>=2.8.2->pandas->pandasql)
(1.17.0)
Requirement already satisfied: greenlet!=0.4.17 in c:\users\david\
anaconda3\lib\site-packages (from sqlalchemy->pandasql) (3.1.1)
Requirement already satisfied: typing-extensions>=4.6.0 in c:\users\
david\anaconda3\lib\site-packages (from sqlalchemy->pandasql) (4.12.2)

```

```

import pandas as pd
import pandasql as ps

```

```

# Load CSV into Pandas DataFrame

```

```

df = pd.read_csv(r"C:\Users\david\Downloads\
airlines_flights_data.csv")
df

```

	index	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]

#-- 1. Find the average ticket price per airline.

```
import pandasql as ps
```

```
query = """
SELECT airline, AVG(price) as avg_price
FROM df
WHERE source_city = 'Delhi'
GROUP BY airline
ORDER BY avg_price DESC
LIMIT 5;
"""
```

```
result = ps.sqldf(query, locals())
print(result)
```

	airline	avg_price
0	Vistara	28938.500342
1	Air_India	21899.890758
2	SpiceJet	6084.649762
3	GO_FIRST	5742.409679
4	Indigo	5386.570134

#-- 2. List the top 5 most expensive routes (source to destination).

```
import pandasql as ps
```

```
query= """
SELECT
    source_city,
    destination_city,
    AVG(price) AS Average_price
FROM
    df
GROUP BY
    source_city, destination_city
ORDER BY
    Average_price DESC
LIMIT 5;
"""
```

```
result = ps.sqldf(query, locals())
print(result)
```

	source_city	destination_city	Average_price
0	Chennai	Bangalore	25081.850454
1	Kolkata	Chennai	23660.361040
2	Bangalore	Kolkata	23500.061229
3	Bangalore	Chennai	23321.850078
4	Mumbai	Bangalore	23147.873807

#3.Which airline offers the lowest average ticket price for Business class?

```
import pandas as pd
import pandasql as ps
query = """
SELECT airline, AVG(price) AS avg_Business_price
FROM df
WHERE class = 'Business'
GROUP BY airline
ORDER BY avg_Business_price
LIMIT 1;
"""
result = ps.sqldf(query, locals())
print(result)
```

	airline	avg_Business_price
0	Air_India	47131.039212

#4.Find routes with more than 50 flights and their average ticket prices.

```
import pandasql as ps
query = """
SELECT source_city, destination_city, COUNT(*) as
flight_count,AVG(price) as avg_price
FROM df
GROUP BY source_city, destination_city
HAVING COUNT(*)> 50;
"""
result = ps.sqldf(query, locals())
print(result)
```

	source_city	destination_city	flight_count	avg_price
0	Bangalore	Chennai	6410	23321.850078
1	Bangalore	Delhi	13756	17723.313972
2	Bangalore	Hyderabad	8928	21226.121192
3	Bangalore	Kolkata	10028	23500.061229
4	Bangalore	Mumbai	12939	23128.618672
5	Chennai	Bangalore	6493	25081.850454
6	Chennai	Delhi	9783	18981.863948
7	Chennai	Hyderabad	6103	21591.345404
8	Chennai	Kolkata	6983	22669.932407

9	Chennai	Mumbai	9338	22765.849647
10	Delhi	Bangalore	14012	17880.216315
11	Delhi	Chennai	10780	19369.881354
12	Delhi	Hyderabad	9328	17347.288379
13	Delhi	Kolkata	11934	20566.409418
14	Delhi	Mumbai	15289	19355.829812
15	Hyderabad	Bangalore	7854	21347.177998
16	Hyderabad	Chennai	6395	21848.065989
17	Hyderabad	Delhi	8506	17243.945685
18	Hyderabad	Kolkata	7987	20823.893201
19	Hyderabad	Mumbai	10064	20080.865759
20	Kolkata	Bangalore	9824	22744.808428
21	Kolkata	Chennai	6653	23660.361040
22	Kolkata	Delhi	10506	19422.354559
23	Kolkata	Hyderabad	7897	21500.011397
24	Kolkata	Mumbai	11467	22078.883579
25	Mumbai	Bangalore	12885	23147.873807
26	Mumbai	Chennai	10130	22781.899112
27	Mumbai	Delhi	14809	18725.320008
28	Mumbai	Hyderabad	10470	21004.046705
29	Mumbai	Kolkata	12602	22379.146723

#5. Compare average price difference between Economy and Business class per airline.

```
import pandasql as ps
query = """
SELECT
    airline,
    MAX(CASE WHEN class = 'Business' THEN avg_price END) -
    MAX(CASE WHEN class = 'Economy' THEN avg_price END) AS price_diff
FROM (
    SELECT
        airline,
        class,
        AVG(price) AS avg_price
    FROM df
    WHERE class IN ('Economy', 'Business')
    GROUP BY airline, class
) t
GROUP BY airline
ORDER BY price_diff DESC;

"""
result = ps.sqldf(query, locals())
print(result)
```

	airline	price_diff
0	Vistara	47670.084132
1	Air_India	39817.357044

2	SpiceJet	NaN
3	Indigo	NaN
4	GO_FIRST	NaN
5	AirAsia	NaN

#6. Which cities have the most incoming flights?

```
import pandasql as ps
query = """
SELECT destination_city as citty, COUNT(*) as incoming_flights
FROM df
GROUP BY destination_city
ORDER BY incoming_flights DESC;
"""
result = ps.sqldf(query, locals())
print(result)
```

	citty	incoming_flights
0	Mumbai	59097
1	Delhi	57360
2	Bangalore	51068
3	Kolkata	49534
4	Hyderabad	42726
5	Chennai	40368

#7.Which cities have the most outgoing flights?

```
import pandasql as ps
query = """
SELECT source_city as city, COUNT(*) as total_outgoing_flights
FROM df
GROUP BY source_city
ORDER BY total_outgoing_flights DESC;
"""
resul = ps.sqldf(query, locals())
print(result)
```

	citty	incoming_flights
0	Mumbai	59097
1	Delhi	57360
2	Bangalore	51068
3	Kolkata	49534
4	Hyderabad	42726
5	Chennai	40368

#8.Find the busiest route (most number of flights).

```
import pandasql as ps
query = """
SELECT source_city,destination_city, COUNT(*) as total_flights
FROM df
```



```
GROUP BY source_city,destination_city
ORDER BY total_flights DESC
LIMIT 1;
"""
```

```
result = ps.sqldf(query, locals())
print(result)
```

```
source_city destination_city total_flights
0         Delhi             Mumbai      15289
```

#9. List the top 3 airlines with cheapest average prices for each route.

```
import pandas as pd
```

```
# Make route column
```

```
df["route"] = df["source_city"] + " → " + df["destination_city"]
```

```
# Step 1: Average price per airline per route
```

```
avg_prices = (
    df.groupby(["route", "airline"], as_index=False)["price"]
    .mean()
    .rename(columns={"price": "avg_price"})
)
```

```
# Step 2: Rank airlines within each route
```

```
avg_prices["rank"] = avg_prices.groupby("route")
["avg_price"].rank(method="first")
```

```
# Step 3: Keep only top 3 cheapest per route
```

```
top3 = avg_prices[avg_prices["rank"] <= 3].sort_values(["route",
"avg_price"])
```

```
top3
```

	route	airline	avg_price	rank
0	Bangalore → Chennai	AirAsia	2073.043478	1.0
3	Bangalore → Chennai	Indigo	2363.326241	2.0
4	Bangalore → Chennai	SpiceJet	2613.310345	3.0
6	Bangalore → Delhi	AirAsia	4807.092426	1.0
8	Bangalore → Delhi	GO_FIRST	5524.702628	2.0
...
168	Mumbai → Hyderabad	GO_FIRST	4603.866889	2.0
169	Mumbai → Hyderabad	Indigo	5870.954610	3.0
171	Mumbai → Kolkata	AirAsia	3977.937365	1.0
173	Mumbai → Kolkata	GO_FIRST	6106.502609	2.0
175	Mumbai → Kolkata	SpiceJet	7065.210689	3.0

```
[90 rows x 4 columns]
```

#10. Find flights with duration more than 5 hours but priced below average.

```
import pandasql as ps
query = """
SELECT *FROM df
WHERE duration >5
AND price < (SELECT AVG(price) FROM df);
"""
result = ps.sqldf(query, locals())
print(result)
```

	index	airline	flight	source_city	departure_time
stops \					
0	18	AirAsia	I5-747	Delhi	Evening
one					
1	19	AirAsia	I5-747	Delhi	Evening
one					
2	20	G0_FIRST	G8-266	Delhi	Early_Morning
one					
3	21	G0_FIRST	G8-101	Delhi	Early_Morning
one					
4	22	G0_FIRST	G8-103	Delhi	Evening
one					
...
...					
169086	206663	Vistara	UK-826	Chennai	Afternoon
one					
169087	206664	Vistara	UK-822	Chennai	Morning
one					
169088	206665	Vistara	UK-824	Chennai	Night
one					
169089	258681	Air_India	AI-776	Bangalore	Early_Morning
two_or_more					
169090	258806	Air_India	AI-776	Bangalore	Early_Morning
two_or_more					

	arrival_time	destination_city	class	duration	days_left
price \					
0	Early_Morning	Mumbai	Economy	12.25	1
5949					
1	Morning	Mumbai	Economy	16.33	1
5949					
2	Evening	Mumbai	Economy	11.75	1
5954					
3	Night	Mumbai	Economy	14.50	1
5954					
4	Morning	Mumbai	Economy	15.67	1
5954					
...

```

...
169086      Morning      Hyderabad      Economy      20.58      49
8640
169087      Morning      Hyderabad      Economy      23.33      49
8640
169088      Night        Hyderabad      Economy      24.42      49
8640
169089      Afternoon     Hyderabad      Business      6.67      45
12000
169090      Afternoon     Hyderabad      Business      6.67      47
12000

```

```

                                route
0      Delhi → Mumbai
1      Delhi → Mumbai
2      Delhi → Mumbai
3      Delhi → Mumbai
4      Delhi → Mumbai
...
169086      Chennai → Hyderabad
169087      Chennai → Hyderabad
169088      Chennai → Hyderabad
169089      Bangalore → Hyderabad
169090      Bangalore → Hyderabad

```

[169091 rows x 13 columns]

#11. Which flight class shows the highest average price variation (std deviation)?

```

import pandasql as ps
query = """
SELECT
    class, AVG(price) AS price_std_dev
FROM df
GROUP BY class
ORDER BY price_std_dev DESC;
"""
result = ps.sqldf(query, locals())
print(result)

```

```

    class  price_std_dev
0  Business    52540.081124
1   Economy     6572.342383

```

#12. Find how ticket prices vary with number of days left before departure.

```

import pandasql as ps
query="""

```

```

SELECT days_left, AVG(price) as avg_price
FROM df
GROUP BY days_left
ORDER BY avg_price DESC;
"""

```

```

result = ps.sqlldf(query, locals())
print(result)

```

	days_left	avg_price
0	2	30211.299801
1	3	28976.083569
2	5	26679.773368
3	4	25730.905653
4	9	25726.246072
5	7	25588.367351
6	10	25572.819134
7	8	24895.883995
8	6	24856.493902
9	11	22990.656070
10	14	22678.002363
11	12	22505.803322
12	13	22498.885384
13	15	21952.540852
14	1	21591.867151
15	16	20503.546237
16	17	20386.353949
17	18	19987.445168
18	27	19950.866195
19	23	19840.913451
20	24	19803.908896
21	29	19744.653119
22	38	19734.912316
23	20	19699.983390
24	22	19590.667385
25	25	19571.641791
26	30	19567.580834
27	34	19562.008266
28	28	19534.986047
29	36	19517.688444
30	19	19507.677375
31	37	19506.306516
32	21	19430.494058
33	31	19392.706612
34	41	19347.440460
35	43	19340.528894
36	33	19306.271739
37	46	19305.351623
38	39	19262.095556
39	32	19258.135308
40	35	19255.652996

41	26	19238.290278
42	45	19199.876307
43	42	19154.261659
44	40	19144.972439
45	44	19049.080174
46	48	18998.126851
47	49	18992.971888
48	47	18553.272038

#13. Which time of day (Morning, Evening, etc.) has the highest average ticket price?

```
import pandasql as ps
query = """
SELECT arrival_time,departure_time, AVG(price) as avg_price
FROM df
GROUP BY arrival_time,departure_time
ORDER BY avg_price DESC
LIMIT 1;
"""
result = ps.sqldf(query, locals())
print(result)
```

	arrival_time	departure_time	avg_price
0	Evening	Night	31425.824194

#14. Find Price trends by airline and travel class

```
import pandasql as ps
query= """
SELECT airline, class, AVG(price) as avg_price
FROM df
GROUP BY airline,class
ORDER BY airline,avg_price DESC;
"""
result = ps.sqldf(query, locals())
print(result)
```

	airline	class	avg_price
0	AirAsia	Economy	4091.072742
1	Air_India	Business	47131.039212
2	Air_India	Economy	7313.682169
3	GO_FIRST	Economy	5652.007595
4	Indigo	Economy	5324.216303
5	SpiceJet	Economy	6179.278881
6	Vistara	Business	55477.027777
7	Vistara	Economy	7806.943645

#15. Compare the airlines average prices for non-stop vs. 1-stop vs. 2+ stops flights.

```

import pandasql as ps
query="""
SELECT airline, stops as stop_type, AVG(price) as avg_price
FROM df
GROUP BY airline, stops
ORDER BY airline, avg_price DESC;
"""

result = ps.sqldf(query, locals())
print(result)

```

	airline	stop_type	avg_price
0	AirAsia	two_or_more	4432.956367
1	AirAsia	one	4096.963741
2	AirAsia	zero	3747.960970
3	Air_India	one	24805.782885
4	Air_India	zero	14403.191918
5	Air_India	two_or_more	13771.177062
6	GO_FIRST	two_or_more	7107.708642
7	GO_FIRST	one	5972.272755
8	GO_FIRST	zero	3526.924915
9	Indigo	two_or_more	7834.838753
10	Indigo	one	5733.028878
11	Indigo	zero	4023.049037
12	SpiceJet	one	6789.364636
13	SpiceJet	zero	4556.430950
14	Vistara	one	32353.149720
15	Vistara	two_or_more	18850.767996
16	Vistara	zero	16416.273587