

Day26_Data Analysis_Python_vs_SQL_Comparison

June 23, 2025

Introduction

This notebook is a practical guide to **Data Analysis using Python and SQL**, based on real-world queries performed on a dataset containing travel and customer behavior data.

It demonstrates how to:

- Load and explore data using **pandas** in Python
- Perform SQL-like operations such as **SELECT**, **WHERE**, **GROUP BY**, **ORDER BY**, and **JOIN** using Python code
- Compare each Python operation with its **equivalent SQL query**
- Apply core data analysis techniques like filtering, grouping, aggregating, and joining data

Whether you're a beginner learning pandas or someone transitioning from SQL to Python (or vice versa), this notebook will help you understand how the two languages mirror each other for data analysis tasks.

Most Common `groupby()` Functions in Pandas

Function	Description	Example Code
<code>mean()</code>	Average of the values	<code>df.groupby('col')['val'].mean()</code>
<code>sum()</code>	Sum of the values	<code>df.groupby('col')['val'].sum()</code>
<code>count()</code>	Count of non-null values	<code>df.groupby('col')['val'].count()</code>
<code>size()</code>	Count of total rows (includes NaNs)	<code>df.groupby('col').size()</code>
<code>min()</code>	Minimum value	<code>df.groupby('col')['val'].min()</code>
<code>max()</code>	Maximum value	<code>df.groupby('col')['val'].max()</code>
<code>median()</code>	Median (middle) value	<code>df.groupby('col')['val'].median()</code>
<code>std()</code>	Standard deviation	<code>df.groupby('col')['val'].std()</code>
<code>var()</code>	Variance	<code>df.groupby('col')['val'].var()</code>
<code>nunique()</code>	Number of unique values	<code>df.groupby('col')['val'].nunique()</code>
<code>unique()</code>	List/array of unique values	<code>df.groupby('col')['val'].unique()</code>
<code>first()</code>	First non-null value in group	<code>df.groupby('col')['val'].first()</code>
<code>last()</code>	Last non-null value in group	<code>df.groupby('col')['val'].last()</code>
<code>describe()</code>	Summary stats (count, mean, std, min, max, etc.)	<code>df.groupby('col')['val'].describe()</code>
<code>apply(func)</code>	Apply a custom function	<code>df.groupby('col')['val'].apply(lambda x: x.max() - x.min())</code>

Function	Description	Example Code
agg()	Apply multiple functions at once	<code>df.groupby('col')['val'].agg(['mean', 'max', 'min'])</code>

```
[2]: import pandas as pd
```

```
[3]: # Load dataset
df = pd.read_csv(r'C:\Users\aksha\OneDrive\Desktop\SQL_DATA_EXPORT\dataset1.
↳csv')
```

SQL Equivalent:

```
SELECT * FROM dataset_1;
```

1 Display specific columns

SQL Equivalent:

```
SELECT weather, temperature FROM dataset_1;
```

Python:

```
[5]: df[['weather', 'temperature']]
```

```
[5]:      weather  temperature
0      Sunny           55
1      Sunny           80
2      Sunny           80
3      Sunny           80
4      Sunny           80
...      ...           ...
12679  Rainy           55
12680  Rainy           55
12681  Snowy           30
12682  Snowy           30
12683  Sunny           80
```

```
[12684 rows x 2 columns]
```

2 View first 10 rows

SQL Equivalent:

```
SELECT * FROM dataset_1 LIMIT 10;
```

Python:

```
[6]: df.head(10)
```

```

[6]:      destination  passanger weather  temperature  time  \
0  No Urgent Place      Alone   Sunny           55   2PM
1  No Urgent Place  Friend(s)   Sunny           80  10AM
2  No Urgent Place  Friend(s)   Sunny           80  10AM
3  No Urgent Place  Friend(s)   Sunny           80   2PM
4  No Urgent Place  Friend(s)   Sunny           80   2PM
5  No Urgent Place  Friend(s)   Sunny           80   6PM
6  No Urgent Place  Friend(s)   Sunny           55   2PM
7  No Urgent Place    Kid(s)   Sunny           80  10AM
8  No Urgent Place    Kid(s)   Sunny           80  10AM
9  No Urgent Place    Kid(s)   Sunny           80  10AM

      coupon expiration  gender age  maritalStatus  ...  \
0      Restaurant(<20)      1d Female  21  Unmarried partner  ...
1      Coffee House      2h Female  21  Unmarried partner  ...
2  Carry out & Take away      2h Female  21  Unmarried partner  ...
3      Coffee House      2h Female  21  Unmarried partner  ...
4      Coffee House      1d Female  21  Unmarried partner  ...
5      Restaurant(<20)      2h Female  21  Unmarried partner  ...
6  Carry out & Take away      1d Female  21  Unmarried partner  ...
7      Restaurant(<20)      2h Female  21  Unmarried partner  ...
8  Carry out & Take away      2h Female  21  Unmarried partner  ...
9      Bar      1d Female  21  Unmarried partner  ...

      CarryAway RestaurantLessThan20 Restaurant20To50 toCoupon_GEQ5min  \
0      NaN      4~8      1~3      1
1      NaN      4~8      1~3      1
2      NaN      4~8      1~3      1
3      NaN      4~8      1~3      1
4      NaN      4~8      1~3      1
5      NaN      4~8      1~3      1
6      NaN      4~8      1~3      1
7      NaN      4~8      1~3      1
8      NaN      4~8      1~3      1
9      NaN      4~8      1~3      1

      toCoupon_GEQ15min toCoupon_GEQ25min direction_same direction_opp  Y  \
0      0      0      0      0      1  1
1      0      0      0      0      1  0
2      1      0      0      0      1  1
3      1      0      0      0      1  0
4      1      0      0      0      1  0
5      1      0      0      0      1  1
6      1      0      0      0      1  1
7      1      0      0      0      1  1
8      1      0      0      0      1  1
9      1      0      0      0      1  0

```

	row_count
0	1
1	2
2	3
3	4
4	5
5	6
6	7
7	8
8	9
9	10

[10 rows x 27 columns]

3 Unique values in a column

SQL Equivalent:

```
SELECT DISTINCT passanger FROM dataset_1;
```

Python:

```
[7]: df['passanger'].unique()
```

```
[7]: array(['Alone', 'Friend(s)', 'Kid(s)', 'Partner'], dtype=object)
```

4 Filter rows based on condition

SQL Equivalent:

```
SELECT * FROM dataset_1 WHERE destination = 'Home';
```

Python:

```
[8]: df[df['destination'] == 'Home']
```

```
[8]:
```

	destination	passanger	weather	temperature	time	coupon \
13	Home	Alone	Sunny	55	6PM	Bar
14	Home	Alone	Sunny	55	6PM	Restaurant(20-50)
15	Home	Alone	Sunny	80	6PM	Coffee House
35	Home	Alone	Sunny	55	6PM	Bar
36	Home	Alone	Sunny	55	6PM	Restaurant(20-50)
...
12675	Home	Alone	Snowy	30	10PM	Coffee House
12676	Home	Alone	Sunny	80	6PM	Restaurant(20-50)
12677	Home	Partner	Sunny	30	6PM	Restaurant(<20)
12678	Home	Partner	Sunny	30	10PM	Restaurant(<20)
12679	Home	Partner	Rainy	55	6PM	Carry out & Take away

	expiration	gender	age	maritalStatus	...	CarryAway	\
13	1d	Female	21	Unmarried partner	...	NaN	
14	1d	Female	21	Unmarried partner	...	NaN	
15	2h	Female	21	Unmarried partner	...	NaN	
35	1d	Male	21	Single	...	4~8	
36	1d	Male	21	Single	...	4~8	
...	
12675	2h	Male	26	Single	...	1~3	
12676	1d	Male	26	Single	...	1~3	
12677	1d	Male	26	Single	...	1~3	
12678	2h	Male	26	Single	...	1~3	
12679	1d	Male	26	Single	...	1~3	

	RestaurantLessThan20	Restaurant20To50	toCoupon_GEQ5min	\
13	4~8	1~3	1	
14	4~8	1~3	1	
15	4~8	1~3	1	
35	4~8	less1	1	
36	4~8	less1	1	
...	
12675	4~8	1~3	1	
12676	4~8	1~3	1	
12677	4~8	1~3	1	
12678	4~8	1~3	1	
12679	4~8	1~3	1	

	toCoupon_GEQ15min	toCoupon_GEQ25min	direction_same	direction_opp	Y	\
13	0	0	1	0	1	
14	1	0	0	1	1	
15	0	0	0	1	0	
35	0	0	1	0	1	
36	1	0	0	1	0	
...	
12675	1	0	0	1	0	
12676	0	0	1	0	1	
12677	1	1	0	1	1	
12678	1	0	1	0	0	
12679	0	0	1	0	1	

	row_count
13	14
14	15
15	16
35	36
36	37
...	...

```

12675    12676
12676    12677
12677    12678
12678    12679
12679    12680

```

[3237 rows x 27 columns]

5 Order by a column

SQL Equivalent:

```
SELECT * FROM dataset_1 ORDER BY coupon;
```

Python:

```
[9]: df.sort_values('coupon')
```

```
[9]:
```

	destination	passanger	weather	temperature	time	coupon \
11702	Home	Partner	Sunny	30	10PM	Bar
9930	No Urgent Place	Alone	Snowy	30	2PM	Bar
10632	Home	Alone	Rainy	55	6PM	Bar
7997	No Urgent Place	Friend(s)	Rainy	55	10PM	Bar
11166	Work	Alone	Snowy	30	7AM	Bar
...
10476	Home	Alone	Sunny	80	6PM	Restaurant(<20)
5447	Home	Alone	Sunny	80	10PM	Restaurant(<20)
10478	Home	Alone	Snowy	30	10PM	Restaurant(<20)
5440	No Urgent Place	Alone	Sunny	80	2PM	Restaurant(<20)
0	No Urgent Place	Alone	Sunny	55	2PM	Restaurant(<20)

	expiration	gender	age	maritalStatus	...	CarryAway \
11702	2h	Female	50plus	Married partner	...	4~8
9930	1d	Female	21	Single	...	gt8
10632	1d	Male	21	Single	...	gt8
7997	2h	Male	26	Unmarried partner	...	4~8
11166	1d	Female	41	Married partner	...	gt8
...
10476	1d	Female	31	Unmarried partner	...	1~3
5447	2h	Female	50plus	Single	...	less1
10478	2h	Female	31	Unmarried partner	...	1~3
5440	2h	Female	50plus	Single	...	less1
0	1d	Female	21	Unmarried partner	...	NaN

	RestaurantLessThan20	Restaurant20To50	toCoupon_GEQ5min \
11702	1~3	less1	1
9930	gt8	4~8	1
10632	less1	less1	1

7997	never	1~3	1
11166	1~3	less1	1
...
10476	1~3	less1	1
5447	less1	never	1
10478	1~3	less1	1
5440	less1	never	1
0	4~8	1~3	1

	toCoupon_GEQ15min	toCoupon_GEQ25min	direction_same	direction_opp	Y	\
11702	1	1	0	1	0	
9930	0	0	0	1	0	
10632	1	1	0	1	0	
7997	1	0	0	1	1	
11166	1	1	0	1	0	
...	
10476	0	0	1	0	1	
5447	0	0	1	0	0	
10478	1	1	0	1	0	
5440	1	0	0	1	0	
0	0	0	0	1	1	

	row_count
11702	11703
9930	9931
10632	10633
7997	7998
11166	11167
...	...
10476	10477
5447	5448
10478	10479
5440	5441
0	1

[12684 rows x 27 columns]

6 Rename a column

SQL Equivalent:

```
SELECT destination AS Destination FROM dataset_1;
```

Python:

```
[10]: df.rename(columns={'destination': 'Destination'}, inplace=True)
```

7 Group by with count

SQL Equivalent:

```
SELECT occupation, COUNT(*) AS Count FROM dataset_1 GROUP BY occupation;
```

Python:

```
[11]: df.groupby('occupation').size()
```

```
[11]: occupation
Architecture & Engineering      175
Arts Design Entertainment Sports & Media  629
Building & Grounds Cleaning & Maintenance    44
Business & Financial            544
Community & Social Services      241
Computer & Mathematical        1408
Construction & Extraction       154
Education&Training&Library      943
Farming Fishing & Forestry       43
Food Preparation & Serving Related  298
Healthcare Practitioners & Technical  244
Healthcare Support              242
Installation Maintenance & Repair  133
Legal                          219
Life Physical Social Science     170
Management                    838
Office & Administrative Support    639
Personal Care & Service           175
Production Occupations          110
Protective Service              175
Retired                        495
Sales & Related                 1093
Student                       1584
Transportation & Material Moving   218
Unemployed                    1870
dtype: int64
```

```
[12]: df.groupby('occupation').size().to_frame('Count').reset_index()
```

```
[12]:
```

	occupation	Count
0	Architecture & Engineering	175
1	Arts Design Entertainment Sports & Media	629
2	Building & Grounds Cleaning & Maintenance	44
3	Business & Financial	544
4	Community & Social Services	241
5	Computer & Mathematical	1408
6	Construction & Extraction	154
7	Education&Training&Library	943

8	Farming Fishing & Forestry	43
9	Food Preparation & Serving Related	298
10	Healthcare Practitioners & Technical	244
11	Healthcare Support	242
12	Installation Maintenance & Repair	133
13	Legal	219
14	Life Physical Social Science	170
15	Management	838
16	Office & Administrative Support	639
17	Personal Care & Service	175
18	Production Occupations	110
19	Protective Service	175
20	Retired	495
21	Sales & Related	1093
22	Student	1584
23	Transportation & Material Moving	218
24	Unemployed	1870

8 Group by with average

SQL Equivalent:

```
SELECT weather, AVG(temperature) AS avg_temp FROM dataset_1 GROUP BY weather;
```

Python:

```
[13]: df.groupby('weather')['temperature'].mean().to_frame('avg_temp').reset_index()
```

```
[13]:   weather  avg_temp
0   Rainy   55.000000
1   Snowy   30.000000
2   Sunny   68.946271
```

Group by with count of temperature entries SQL Equivalent:

```
SELECT weather, COUNT(temperature) AS Count_temp FROM dataset_1 GROUP BY weather;
```

Python:

```
[14]: df.groupby('weather')['temperature'].size().to_frame('Count_temp').reset_index()
```

```
[14]:   weather  Count_temp
0   Rainy         1210
1   Snowy         1405
2   Sunny         1069
```

9 Group by with count of distinct values

SQL Equivalent:

```
SELECT weather, COUNT(DISTINCT temperature) AS count_distinct_temp FROM dataset_1 GROUP BY weather;
```

Python:

```
[15]: df.groupby('weather')['temperature'].nunique().to_frame('count_distinct_temp').  
      ↪reset_index()
```

```
[15]:   weather  count_distinct_temp  
0    Rainy                      1  
1    Snowy                      1  
2    Sunny                      3
```

10 Group by with sum

SQL Equivalent:

```
SELECT weather, SUM(temperature) AS sum_temp FROM dataset_1 GROUP BY weather;
```

Python:

```
[16]: df.groupby('weather')['temperature'].sum().to_frame('sum_temp').reset_index()
```

```
[16]:   weather  sum_temp  
0    Rainy    66550  
1    Snowy    42150  
2    Sunny   694220
```

11 Group by with min and max

SQL Equivalent:

```
SELECT weather, MIN(temperature) AS min_temp FROM dataset_1 GROUP BY weather;  
SELECT weather, MAX(temperature) AS max_temp FROM dataset_1 GROUP BY weather;
```

Python:

```
[18]: df.groupby('weather')['temperature'].min().to_frame('min_temp').reset_index()
```

```
[18]:   weather  min_temp  
0    Rainy         55  
1    Snowy         30  
2    Sunny         30
```

```
[19]: df.groupby('weather')['temperature'].max().to_frame('max_temp').reset_index()
```

```
[19]:   weather  max_temp  
0    Rainy         80  
1    Snowy         30  
2    Sunny         80
```

12 Group by with HAVING clause logic

SQL Equivalent:

```
SELECT occupation FROM dataset_1 GROUP BY occupation HAVING occupation = 'Student';
```

Python:

```
[20]: df.groupby('occupation').filter(lambda x: x['occupation'].iloc[0] == 'Student').  
      ↳groupby('occupation').size()
```

```
[20]: occupation  
      Student      1584  
      dtype: int64
```

13 Union + Drop duplicates

SQL Equivalent:

```
SELECT DISTINCT destination FROM (  
    SELECT * FROM dataset_1  
    UNION  
    SELECT * FROM table_to_union  
);
```

If you're running this notebook yourself, make sure to define or load `df1` and `df2` from appropriate CSVs or simulated data before using these examples.

Python:

```
pd.concat([df, df1])['destination'].drop_duplicates()
```

14 INNER JOIN

SQL Equivalent:

```
SELECT a.destination, a.time, b.part_of_day  
FROM dataset_1 a  
INNER JOIN table_to_join b ON a.time = b.time;
```

Python:

```
pd.merge(df, df2[['time', 'part_of_day']], on='time', how='inner')[['destination', 'time', 'part_of_day']]
```

If you're running this notebook yourself, make sure to define or load `df1` and `df2` from appropriate CSVs or simulated data before using these examples.

15 Filtering rows by value

SQL Equivalent:

```
SELECT destination, passenger FROM dataset_1 WHERE passenger = 'Alone';
```

Python:

```
[21]: df[df['passanger'] == 'Alone'][['Destination', 'passanger']]
```

```
[21]:
```

	Destination	passanger
0	No Urgent Place	Alone
13	Home	Alone
14	Home	Alone
15	Home	Alone
16	Work	Alone
...
12676	Home	Alone
12680	Work	Alone
12681	Work	Alone
12682	Work	Alone
12683	Work	Alone

[7305 rows x 2 columns]

16 Filtering rows by prefix

SQL Equivalent:

```
SELECT * FROM dataset_1 WHERE weather LIKE 'Sun%';
```

Python:

```
[22]: df[df['weather'].str.startswith('Sun')]
```

```
[22]:
```

	Destination	passanger	weather	temperature	time	\
0	No Urgent Place	Alone	Sunny	55	2PM	
1	No Urgent Place	Friend(s)	Sunny	80	10AM	
2	No Urgent Place	Friend(s)	Sunny	80	10AM	
3	No Urgent Place	Friend(s)	Sunny	80	2PM	
4	No Urgent Place	Friend(s)	Sunny	80	2PM	
...	
12673	Home	Alone	Sunny	30	6PM	
12676	Home	Alone	Sunny	80	6PM	
12677	Home	Partner	Sunny	30	6PM	
12678	Home	Partner	Sunny	30	10PM	
12683	Work	Alone	Sunny	80	7AM	

	coupon	expiration	gender	age	maritalStatus	...	\
0	Restaurant(<20)	1d	Female	21	Unmarried partner	...	
1	Coffee House	2h	Female	21	Unmarried partner	...	
2	Carry out & Take away	2h	Female	21	Unmarried partner	...	
3	Coffee House	2h	Female	21	Unmarried partner	...	
4	Coffee House	1d	Female	21	Unmarried partner	...	

...
12673	Carry out & Take away	1d	Male	26	Single	...
12676	Restaurant(20-50)	1d	Male	26	Single	...
12677	Restaurant(<20)	1d	Male	26	Single	...
12678	Restaurant(<20)	2h	Male	26	Single	...
12683	Restaurant(20-50)	2h	Male	26	Single	...

	CarryAway	RestaurantLessThan20	Restaurant20To50	toCoupon_GEQ5min	\
0	NaN	4~8	1~3	1	
1	NaN	4~8	1~3	1	
2	NaN	4~8	1~3	1	
3	NaN	4~8	1~3	1	
4	NaN	4~8	1~3	1	
...	
12673	1~3	4~8	1~3	1	
12676	1~3	4~8	1~3	1	
12677	1~3	4~8	1~3	1	
12678	1~3	4~8	1~3	1	
12683	1~3	4~8	1~3	1	

	toCoupon_GEQ15min	toCoupon_GEQ25min	direction_same	direction_opp	Y	\
0	0	0	0	1	1	
1	0	0	0	1	0	
2	1	0	0	1	1	
3	1	0	0	1	0	
4	1	0	0	1	0	
...	
12673	0	0	0	1	0	
12676	0	0	1	0	1	
12677	1	1	0	1	1	
12678	1	0	1	0	0	
12683	0	0	1	0	0	

	row_count
0	1
1	2
2	3
3	4
4	5
...	...
12673	12674
12676	12677
12677	12678
12678	12679
12683	12684

[10069 rows x 27 columns]

17 Filter values within a range

SQL Equivalent:

```
SELECT DISTINCT temperature FROM dataset_1 WHERE temperature BETWEEN 29 AND 75;
```

Python:

```
[23]: df[(df['temperature'] >= 29) & (df['temperature'] <= 75)]['temperature'].  
      ↪unique()
```

```
[23]: array([55, 30], dtype=int64)
```

18 Filter rows with specific values

SQL Equivalent:

```
SELECT occupation FROM dataset_1 WHERE occupation IN ('Sales & Related', 'Management');
```

Python:

```
[24]: df[df['occupation'].isin(['Sales & Related', 'Management'])][['occupation']]
```

```
[24]:
```

	occupation
193	Sales & Related
194	Sales & Related
195	Sales & Related
196	Sales & Related
197	Sales & Related
...	...
12679	Sales & Related
12680	Sales & Related
12681	Sales & Related
12682	Sales & Related
12683	Sales & Related

[1931 rows x 1 columns]

Conclusion

In this notebook, we explored how to perform essential data analysis tasks using both **Python** (**pandas**) and **SQL** side-by-side. From filtering and sorting to grouping and aggregating data, we covered a wide range of operations commonly used in real-world data workflows.

This comparative approach not only strengthens your Python and SQL skills but also helps you transition smoothly between the two, depending on the data environment you're working with.

Keep practicing these techniques on different datasets to build a strong foundation in data analysis. Remember — clean, well-understood data is the first step toward building powerful insights and models.

Happy Learning!