# Types of Pandas Joins and How to use them in Python

Here are different types of pandas joins and how to use them in Python. Inner, Outer, Right, and Left joins are explained with examples from Amazon and Meta.



Pandas is an important Python tool to do data analysis and manipulation. Its typical use is working with data frames while analyzing and manipulating data. While working on different data frames, you can combine them using three different functions in four different ways.

# How many Methods for Joining Data Frames are there in Pandas?

There are three types of join functions in Python.

- 1. **Merge**: It merges two data frames on columns or indices.
- 2. **Join**: The join() function will join two data frames together on a key column or index.

3. Concat: The concat() function bonds two data frames across the rows or columns.

It sounds rather similar, so what is the difference between these three approaches?

Merge() allows you to perform more flexible table joins because it provides you more combinations, yet concat() is less structured. Join() combines data frames on the index but not on columns, yet merge() gives you a chance to specify the column you want to join on.

In our examples, we will use merge() to show you how different types of joins in python work.

# What's the Point of Joins in Pandas?



In simple terms, pandas joins in python are used to combine two data frames. When doing that, you have to specify the type of join. The defined pandas join type specifies how the data frames will be joined.

Now, let's look at the types of python pandas joins with the merge function.

# **Types of Pandas Joins in Python?**

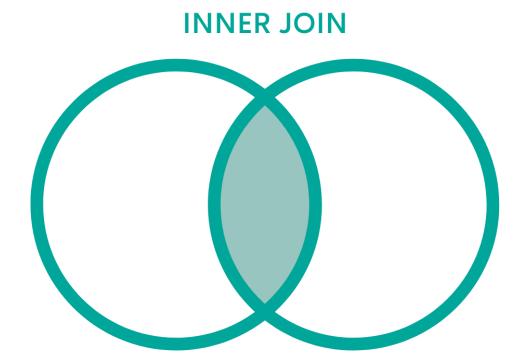
There are four main types of pandas joins in python, which we will explain in this article.

- Inner
- Outer
- Left
- Right

Here is the <u>official page</u> of the Python Pandas "merge" function, which we will use to join two data frames.

Now, let's get started with Inner Join.

# **Inner Join in Python**



# What is Inner Join in Python?

Inner join in pandas is used to merge two data frames at the intersection. It returns the mutual rows of both.

# **Inner Join Example**

Now, let's look at the example from the platform. Our Inner Join question from the Meta(Facebook).

Meta developed a new programming language called Hack. To measure its popularity, they ran a survey with their employees. Due to an error location, data was not collected, but your supervisor demands a report showing the average popularity of Hach by office location. Now the aim is to find the average popularity of the Hack per office location. Output has to contain the location along with the average popularity.



Link to the question: https://platform.stratascratch.com/coding/10061-popularity-of-hack

### Data

We have two data frames. Our first data frame is **facebook\_employees**. The table has the following columns.

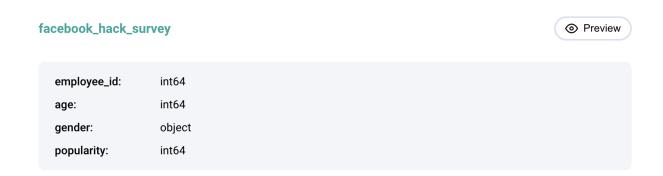


The data preview is shown below.

-

id	location	age	gender	is_senior
0	USA	24	М	FALSE
1	USA	31	F	TRUE
2	USA	29	F	FALSE
3	USA	33	М	FALSE
4	USA	36	F	TRUE

The second data frame is **facebook\_hack\_survey**, and it has the following columns.



Also, the data preview is shown below.

employee_id	age	gender	popularity
0	24	М	6
1	31	F	4
2	29	F	0
3	33	М	7
4	36	F	6

# **Solution Approach**

- 1. Let's load the libraries.
- 2. Now, we have to merge two data frames to find the popularity of the location.
- 3. Since the question wants us to show popularity and location, we will group by two columns, and then we will use the mean() function to find the average and reset\_index() to remove indexes that the groupby() function creates.

# Coding

1. Let's import the NumPy and Pandas libraries first to manipulate the data and use the statistical methods with it.

```
import pandas as pd
import numpy as np
```

If you want to know how to import pandas as pd in python and its importance for doing data science, check out our article "How to Import Pandas as pd in Python".

2. Now, question asks us to return to a location with popularity. We have the location in the first data frame and the popularity in our second data frame. So to draw popularity and location together, let's merge two data frames using the inner join on id. The age and gender columns are in common, yet the id column has a different name in both data frames. That's why we matched the left\_on argument with id and the right\_on argument with employee\_id.

We want to find the popularity of the Hack per office location. So the location and the popularity should match, that's why we need the intersection, so we will use inner join.

Selecting the right python join type is crucial to get the correct answer. In this case, the left and inner join will return the same result. They will both return 14 rows, which are the commons of both tables.

Yet, the right join will return the whole right data frame, which contains 17 rows, and for the rest, there will be NA assigned on the left data frame.

Below is the info table of three data frames to see the information of the rows of the first, the second, and the merged data frames.

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14 entries, 0 to 13
Data columns (total 5 columns):
     Column Non-Null Count Dtype
 #
               _____
___
     ____
                                 ____
                               int64
objec
                14 non-null
 0
     id
     location 14 non-null
 1
                                 object
   age
 2 age 14 non-null
3 gender 14 non-null
                                int64
                                 object
     is_senior 14 non-null
                                 bool
dtypes: bool(1), int64(2), object(2)
memory usage: 590.0+ bytes
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 17 entries, 0 to 16
Data columns (total 4 columns):
     Column Non-Null Count Dtype
 #
     ____
                 -----
   employee id 17 non-null
 0
                                   int64
   age 17 non-null gender 17 non-null
 1
                                   int64
 2
                                  object
    popularity 17 non-null
 3
                                   int64
dtypes: int64(3), object(1)
memory usage: 672.0+ bytes
<class 'pandas.core.frame.DataFrame'>
Int64Index: 14 entries, 0 to 13
Data columns (total 9 columns):
 #
     Column Non-Null Count
                                   Dtype
___
     _____
                  _____
                                   ____
                 14 non-null
 0
    id
                                   int64
1 location 14 non-null
2 age_x 14 non-null
3 gender_x 14 non-null
4 is_senior 14 non-null
5 employee_id 14 non-null
                                   object
                                   int64
                                   object
                                   bool
                                   int64
                 14 non-null
 6
    age y
                                   int64
     gender_y 14 non-null popularity 14 non-null
 7
                                   object
                                   int64
dtypes: bool(1), int64(5), object(3)
memory usage: 1022.0+ bytes
```

OK, let's get back to writing the answer using the inner join.

```
import pandas as pd
import numpy as np
merged = pd.merge(facebook_employees,facebook_hack_survey, left_on = 'id',
right_on = 'employee_id', how = 'inner')
```

3. The question wants us to return the average popularity based on the location, so let's use the groupby() function with mean() and reset indexes that groupby() creates using the reset index() method.

```
import pandas as pd
import numpy as np

merged = pd.merge(facebook_employees,facebook_hack_survey, left_on = 'id',
    right_on = 'employee_id', how = 'inner')
    result = merged.groupby(['location'])['popularity'].mean().reset_index()
```

# Output

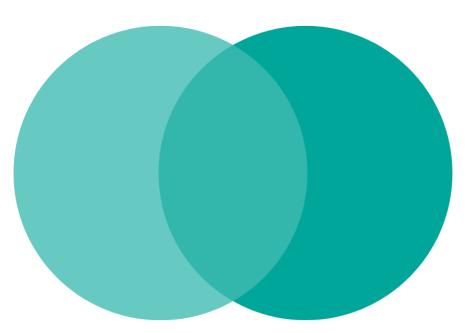
Here is the output, the average popularity based on the locations.

location	popularity
India	7.5
Switzerland	1
UK	4.333
USA	4.6

By the way, if you want to learn more about Pandas, here are <u>Pandas Interview Questions</u> for Data Science.

# **Outer Join in Python**





# What is Outer Join in Python?

It might also be called Full Outer Join and returns all rows from both DataFrames. It will match all rows that exist in both data frames. The rows found only in one of the two DataFrames will get the NA value.

# **Outer Join Example**

Now, let's look at an example of Outer join from the platform.

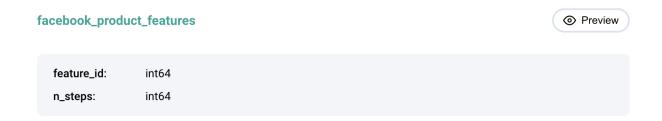
An app has product features to help guide users through a marketing funnel. Each funnel has steps as a guide to complete the funnel. Meta asks us to find the average percentage of completion for each feature.

# 

Link to the question: <a href="https://platform.stratascratch.com/coding/9792-user-feature-completion">https://platform.stratascratch.com/coding/9792-user-feature-completion</a>

# Data

We have two data frames. Our first data frame is **facebook\_product\_features**. The data frame has the following columns.



Also, the data preview is shown below.

feature_id	n_steps
0	5
1	7
2	3

The second data frame is **facebook\_product\_features\_realizations** with the following columns.

## facebook\_product\_features\_realizations



feature\_id: int64 user\_id: int64 step\_reached: int64

timestamp: datetime64[ns]

# The data preview is shown below.

feature_id	user_id	step_reached	timestamp
0	0	1	2019-03-11 17:15:00
0	0	2	2019-03-11 17:22:00
0	0	3	2019-03-11 17:25:00
0	0	4	2019-03-11 17:27:00
0	1	1	2019-03-11 19:51:00

# **Solution Approach**

- 1. Load the pandas library.
- 2. Group by the feature\_id and user id, and calculate the max step reached.
- 3. Merge two data frames on feature\_id using the outer join and fill NAs with zero.
- 4. Calculate the share of completion by dividing the step reached with n\_step times 100 to find the percentage.
- 5. Group the data frame by feature\_id and select the share of completion, calculate the mean, reset the index, and save the results to frame.

## Coding

1. Let's import the pandas library first to manipulate the data.

# import pandas as pd

2. Now here is the time to find the maximum step by grouping by the feature\_id and user\_id first. Then select the step reached and use the max() function afterward. After that, we will reset the index that the groupby() function creates.

```
import pandas as pd

max_step = facebook_product_features_realizations.groupby(["feature_id",
    "user_id"])[
        "step_reached"].max().reset_index()
```

3. Next, we have to calculate the share of completion. We will divide the step reached by n\_steps and multiply by 100. So we have to select n\_steps from the first data frame and step\_reached from the second data frame. We will combine them on feature\_id using the outer join because we need all values from both data frames to do the math. The non-matching values will be NA, so we will replace these values with zero after merging.

```
import pandas as pd

max_step = facebook_product_features_realizations.groupby(["feature_id",
    "user_id"])[
        "step_reached"].max().reset_index()

df = pd.merge(facebook_product_features, max_step, how='outer',
        on='feature_id').fillna(0)
```

4. At this stage, we will calculate the share of completion by dividing the step reached by the number of steps and multiplying by 100.

```
import pandas as pd

max_step = facebook_product_features_realizations.groupby(["feature_id",
    "user_id"])[
        "step_reached"].max().reset_index()

df = pd.merge(facebook_product_features, max_step, how='outer',
        on='feature_id').fillna(0)

df["share_of_completion"] = (df["step_reached"] / df["n_steps"])*100
```

5. Now, we will group by the data frame by feature\_id, select the share of completion and calculate the mean. Then we will assign these results to the column avg\_share\_of\_completion and reset the index.

```
import pandas as pd

max_step = facebook_product_features_realizations.groupby(["feature_id",
    "user_id"])[
        "step_reached"].max().reset_index()

df = pd.merge(facebook_product_features, max_step, how='outer',
        on='feature_id').fillna(0)

df["share_of_completion"] = (df["step_reached"] / df["n_steps"])*100

result =
    df.groupby("feature_id")["share_of_completion"].mean().to_frame("avg_share_of_completion").reset_index()
```

# Output

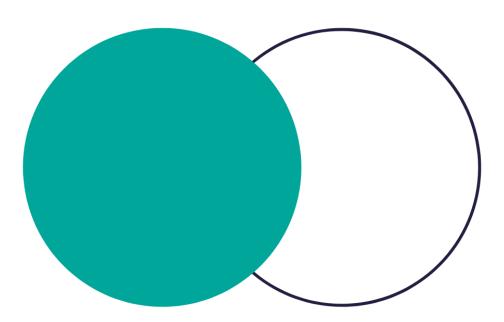
Here is the expected output.

feature_id	avg_share_of_completion
0	80
1	76.19
2	0

If you want to enhance your Python skills, here are Python Interview Questions and Answers.

# **Left Outer Join in Python**

# **LEFT OUTER JOIN**



# What is Left Outer Join in Python?

The left outer join returns all rows from the left data frame, which will be merged with the corresponding rows from the right data frame. It fills the unmatched rows with NA.

# **Left Outer Join Example**

This time, our coding question is from Amazon.

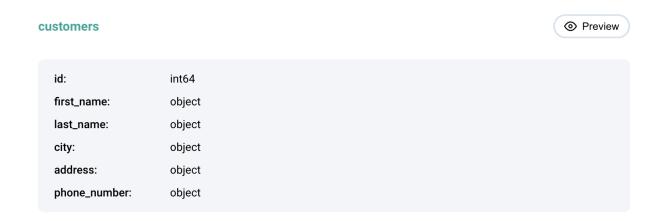
Amazon asks us to sort records based on the customer's first name and the order details in ascending order.

# Customer Details Interview Question Date: April 2019 Amazon Easy Interview Questions ID 9891 Chapter 68 ✓ Find the details of each customer regardless of whether the customer made an order. Output the customer's first name, last name, and the city along with the order details. You may have duplicate rows in your results due to a customer ordering several of the same items. Sort records based on the customer's first name and the order details in ascending order. DataFrames: customers, orders Expected Output Type: pandas.DataFrame

Link to the question: https://platform.stratascratch.com/coding/9891-customer-details

### **Data**

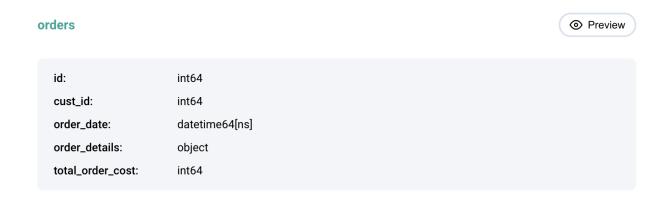
We have two data frames. The first data frame is **customers**. The data frame has the following columns.



The data preview is shown below.

id	first_name	last_name	city	address	phone_number
8	John	Joseph	San Francisco		928-386-8164
7	Jill	Michael	Austin		813-297-0692
4	William	Daniel	Denver		813-368-1200
5	Henry	Jackson	Miami		808-601-7513
13	Emma	Isaac	Miami		808-690-5201

Our second data frame is **orders**. The data frame has the following columns.



Also, the data preview is shown below.

id	cust_id	order_date	order_details	total_order_cost
1	3	2019-03-04 00:00:00	Coat	100
2	3	2019-03-01 00:00:00	Shoes	80
3	3	2019-03-07 00:00:00	Skirt	30
4	7	2019-02-01 00:00:00	Coat	25
5	7	2019-03-10 00:00:00	Shoes	80

# **Solution Approach**

- 1. Let's load the libraries.
- 2. Merge the data frames from the left on id and cust id.
- 3. Sort values by the first names and order details and show first\_name, last\_name, city, and order\_details in the output.

# Coding

1. Now first, let's import pandas and NumPy libraries.

```
import pandas as pd
import numpy as np
```

2. Here, we will merge both data frames using the left join because the output should contain the sorted records based on the customer's first name and the order details. We need the list of all customers.

```
import pandas as pd
import numpy as np

merged = pd.merge(customers, orders, left_on = 'id', right_on = 'cust_id',
how = 'left')
```

3. It is time to sort values according to the first name and order details and select the first name, last name, city, and order details.

```
import pandas as pd
import numpy as np
```

```
merged = pd.merge(customers, orders, left_on = 'id', right_on = 'cust_id',
how = 'left')
result =
merged[['first_name','last_name','city','order_details']].sort_values(['first_name','order_details'])
```

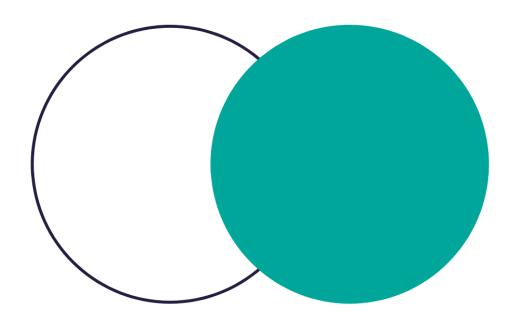
# Output

Here is the expected output.

EmmaIsaacMiamiEvaLucasArizonaCoatEvaLucasArizonaShirtsEvaLucasArizonaSlipperFaridaJosephSan FranciscoCoatFaridaJosephSan FranciscoShirtsFaridaJosephSan FranciscoShoesFaridaJosephSan FranciscoSkirtFrankJacobMiamiShoesHenryJacksonMiamiShoesJackAidenArizonaCoatJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinDressesJillMichaelAustinDressesJillMichaelAustinDresses	first_name	last_name	city	order_details
EvaLucasArizonaShirtsEvaLucasArizonaSlipperFaridaJosephSan FranciscoCoatFaridaJosephSan FranciscoShirtsFaridaJosephSan FranciscoSkirtFaridaJosephSan FranciscoSkirtFrankJacobMiamiShoesHenryJacksonMiamiShoesJackAidenArizonaJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinDresses	Emma	Isaac	Miami	
EvaLucasArizonaSlipperFaridaJosephSan FranciscoCoatFaridaJosephSan FranciscoShirtsFaridaJosephSan FranciscoSkirtFaridaJosephSan FranciscoSkirtFrankJacobMiamiShoesHenryJacksonMiamiShoesJackAidenArizonaJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinDresses	Eva	Lucas	Arizona	Coat
Farida Joseph San Francisco Coat Farida Joseph San Francisco Shirts Farida Joseph San Francisco Shoes Farida Joseph San Francisco Skirt Frank Jacob Miami Henry Jackson Miami Shoes  Jack Aiden Arizona  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Dresses	Eva	Lucas	Arizona	Shirts
Farida Joseph San Francisco Shirts Farida Joseph San Francisco Shoes Farida Joseph San Francisco Skirt Frank Jacob Miami Henry Jackson Miami Shoes  Jack Aiden Arizona  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Dresses	Eva	Lucas	Arizona	Slipper
Farida Joseph San Francisco Shoes Farida Joseph San Francisco Skirt  Frank Jacob Miami  Henry Jackson Miami Shoes  Jack Aiden Arizona  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Coat	Farida	Joseph	San Francisco	Coat
Farida Joseph San Francisco Skirt  Frank Jacob Miami  Henry Jackson Miami Shoes  Jack Aiden Arizona  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Dresses	Farida	Joseph	San Francisco	Shirts
Frank Jacob Miami  Henry Jackson Miami Shoes  Jack Aiden Arizona  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Dresses	Farida	Joseph	San Francisco	Shoes
Henry Jackson Miami Shoes  Jack Aiden Arizona  Jill Michael Austin Coat  Jill Michael Austin Coat  Jill Michael Austin Dresses	Farida	Joseph	San Francisco	Skirt
JackAidenArizonaJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinDresses	Frank	Jacob	Miami	
JillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinDresses	Henry	Jackson	Miami	Shoes
JillMichaelAustinCoatJillMichaelAustinCoatJillMichaelAustinDresses	Jack	Aiden	Arizona	
JillMichaelAustinCoatJillMichaelAustinDresses	Jill	Michael	Austin	Coat
Jill Michael Austin Dresses	Jill	Michael	Austin	Coat
	Jill	Michael	Austin	Coat
Jill Michael Austin Shoes	Jill	Michael	Austin	Dresses
	Jill	Michael	Austin	Shoes

# **Right Outer Join in Python**

# **RIGHT OUTER JOIN**



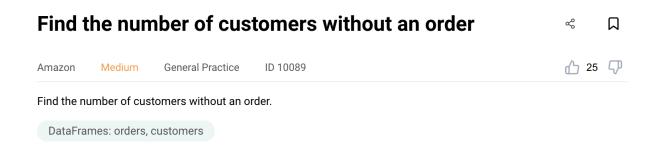
# What is Right Outer Join in Python?

The right outer join returns all rows from the right data frame and the remaining data from the left. The data which does not correspond to the right data frame will have NAs assigned.

# **Right Outer Join Example**

This question is from Amazon.

Amazon asks us to find the number of customers without an order.

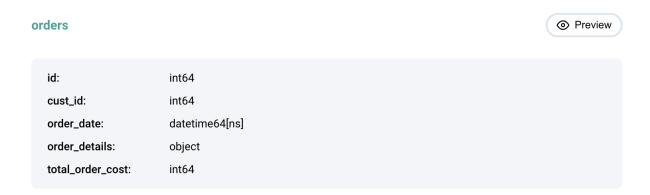


# Link to the question:

https://platform.stratascratch.com/coding/10089-find-the-number-of-customers-without-an-order

## Data

We have two data frames. The **orders** data frame has the following columns.



Here is the preview of the data.

id	cust_id	order_date	order_details	total_order_cost
1	3	2019-03-04 00:00:00	Coat	100
2	3	2019-03-01 00:00:00	Shoes	80
3	3	2019-03-07 00:00:00	Skirt	30
4	7	2019-02-01 00:00:00	Coat	25
5	7	2019-03-10 00:00:00	Shoes	80

The second data frame is **customers**. The data frame has the following columns.

customers 

© Preview

id:	int64		
first_name:	object		
last_name:	object		
city:	object		
address:	object		
phone_number:	object		

# Also, here is the preview.

id	first_name	last_name	city	address	phone_number
8	John	Joseph	San Francisco		928-386-8164
7	Jill	Michael	Austin		813-297-0692
4	William	Daniel	Denver		813-368-1200
5	Henry	Jackson	Miami		808-601-7513
13	Emma	Isaac	Miami		808-690-5201

# **Solution Approach**

- 1. First, let's load the libraries.
- 2. Merge two data frames from the right.
- 3. Find the customers without an order by using the is\_null() method.
- 4. Find the number of customers without an order by using the len() method.

# Coding

1. Let's import the NumPy and pandas libraries first to manipulate the data and use the statistical methods with it.

```
import pandas as pd
import numpy as np
```

2. Now, we will merge these two data frames from the right to find the number of customers without an order. After merging two data frames from the right, the customer's order data column will be null if there's no order. And we can find the customers who haven't had any orders in the next step.

```
import pandas as pd
import numpy as np

merged =
pd.merge(orders,customers,left_on='cust_id',right_on='id',how='right')
```

3. Here we will use the isnull() function to find the customer ids that don't have any orders.

```
import pandas as pd
import numpy as np

merged =
pd.merge(orders,customers,left_on='cust_id',right_on='id',how='right')
null_cust = merged[merged['cust_id'].isnull()]
```

4. By using the len() function, we find the number of customers that have not had any orders.

```
import pandas as pd
import numpy as np

merged =
pd.merge(orders,customers,left_on='cust_id',right_on='id',how='right')
null_cust = merged[merged['cust_id'].isnull()]
result = len(null_cust)
```

# **Output**

Here is the expected output.

If you want to discover the join in SQL too, here are <u>Different Types of SQL JOINs</u>.

# Conclusion

In this article, you learned about four pandas joins in python through the interview questions by the companies like Meta and Amazon. These questions showed you how to use the joins in python and, more specifically, where to use them while doing data manipulation step by step.

Practicing similar interview questions will keep you ready for interviews. You should turn it into a habit. So join the StrataScratch community and sign up today to help us find your dream job.