

Comparison with SQL

Since many potential pandas users have some familiarity with [SQL](#), this page is meant to provide some examples of how various SQL operations would be performed using pandas.

If you're new to pandas, you might want to first read through [10 Minutes to pandas](#) to familiarize yourself with the library.

As is customary we import pandas and numpy as follows:

```
In [1]: import pandas as pd
In [2]: import numpy as np
```

Most of the examples will utilize the `tips` dataset found within pandas tests. We'll read the data into a DataFrame called `tips` and assume we have a database table of the same name and structure.

```
In [3]: url = 'https://raw.githubusercontent.com/pandas-dev/pandas/master/pandas/tests/data/tips.csv'
In [4]: tips = pd.read_csv(url)
In [5]: tips.head()
Out[5]:
```

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4

SELECT

In SQL, selection is done using a comma-separated list of columns you'd like to select (or `*` to select all columns):

```
SELECT total_bill, tip, smoker, time
FROM tips
LIMIT 5;
```

With pandas, column selection is done by passing a list of column names to your DataFrame:

```
In [6]: tips[['total_bill', 'tip', 'smoker', 'time']].head(5)
Out[6]:
```

	total_bill	tip	smoker	time
0	16.99	1.01	No	Dinner
1	10.34	1.66	No	Dinner
2	21.01	3.50	No	Dinner

3	23.68	3.31	No	Dinner
4	24.59	3.61	No	Dinner

Calling the DataFrame without the list of column names would display all columns (akin to SQL*).

WHERE

Filtering in SQL is done via a WHERE clause.

```
SELECT *
FROM tips
WHERE time = 'Dinner'
LIMIT 5;
```

DataFrames can be filtered in multiple ways; the most intuitive of which is using [boolean indexing](#).

```
In [7]: tips[tips['time'] == 'Dinner'].head(5)
Out[7]:
```

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4

The above statement is simply passing a series of True/False objects to the DataFrame, returning all rows with True.

```
In [8]: is_dinner = tips['time'] == 'Dinner'

In [9]: is_dinner.value_counts()
Out[9]:
```

True	176
False	68

Name: time, dtype: int64

```
In [10]: tips[is_dinner].head(5)
Out[10]:
```

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4

Just like SQL's OR and AND, multiple conditions can be passed to a DataFrame using | (OR) and & (AND).

```
-- tips of more than $5.00 at Dinner meals
SELECT *
FROM tips
WHERE time = 'Dinner' AND tip > 5.00;
```

```
# tips of more than $5.00 at Dinner meals
In [11]: tips[(tips['time'] == 'Dinner') & (tips['tip'] > 5.00)]
Out[11]:
```

	total_bill	tip	sex	smoker	day	time	size
23	39.42	7.58	Male	No	Sat	Dinner	4
44	30.40	5.60	Male	No	Sun	Dinner	4
47	32.40	6.00	Male	No	Sun	Dinner	4
52	34.81	5.20	Female	No	Sun	Dinner	4
59	48.27	6.73	Male	No	Sat	Dinner	4
116	29.93	5.07	Male	No	Sun	Dinner	4
155	29.85	5.14	Female	No	Sun	Dinner	5
170	50.81	10.00	Male	Yes	Sat	Dinner	3
172	7.25	5.15	Male	Yes	Sun	Dinner	2
181	23.33	5.65	Male	Yes	Sun	Dinner	2
183	23.17	6.50	Male	Yes	Sun	Dinner	4
211	25.89	5.16	Male	Yes	Sat	Dinner	4
212	48.33	9.00	Male	No	Sat	Dinner	4
214	28.17	6.50	Female	Yes	Sat	Dinner	3
239	29.03	5.92	Male	No	Sat	Dinner	3

```
-- tips by parties of at least 5 diners OR bill total was more than $45
SELECT *
FROM tips
WHERE size >= 5 OR total_bill > 45;
```

```
# tips by parties of at least 5 diners OR bill total was more than $45
In [12]: tips[(tips['size'] >= 5) | (tips['total_bill'] > 45)]
Out[12]:
```

	total_bill	tip	sex	smoker	day	time	size
59	48.27	6.73	Male	No	Sat	Dinner	4
125	29.80	4.20	Female	No	Thur	Lunch	6
141	34.30	6.70	Male	No	Thur	Lunch	6
142	41.19	5.00	Male	No	Thur	Lunch	5
143	27.05	5.00	Female	No	Thur	Lunch	6
155	29.85	5.14	Female	No	Sun	Dinner	5
156	48.17	5.00	Male	No	Sun	Dinner	6
170	50.81	10.00	Male	Yes	Sat	Dinner	3
182	45.35	3.50	Male	Yes	Sun	Dinner	3
185	20.69	5.00	Male	No	Sun	Dinner	5
187	30.46	2.00	Male	Yes	Sun	Dinner	5
212	48.33	9.00	Male	No	Sat	Dinner	4
216	28.15	3.00	Male	Yes	Sat	Dinner	5

NULL checking is done using the `notnull()` and `isnull()` methods.

```
In [13]: frame = pd.DataFrame({'col1': ['A', 'B', np.NaN, 'C', 'D'],
.....:                        'col2': ['F', np.NaN, 'G', 'H', 'I']})
.....:

In [14]: frame
Out[14]:
```

	col1	col2
0	A	F
1	B	NaN
2	NaN	G
3	C	H
4	D	I

Assume we have a table of the same structure as our DataFrame above. We can see only the records where col2 IS NULL with the following query:

```
SELECT *
FROM frame
WHERE col2 IS NULL;
```

```
In [15]: frame[frame['col2'].isnull()]
Out[15]:
   col1 col2
1     B  NaN
```

Getting items where col1 IS NOT NULL can be done with `notnull()`.

```
SELECT *
FROM frame
WHERE col1 IS NOT NULL;
```

```
In [16]: frame[frame['col1'].notnull()]
Out[16]:
   col1 col2
0     A    F
1     B  NaN
3     C    H
4     D    I
```

GROUP BY

In pandas, SQL's GROUP BY operations are performed using the similarly named `groupby()` method. `groupby()` typically refers to a process where we'd like to split a dataset into groups, apply some function (typically aggregation), and then combine the groups together.

A common SQL operation would be getting the count of records in each group throughout a dataset. For instance, a query getting us the number of tips left by sex:

```
SELECT sex, count(*)
FROM tips
GROUP BY sex;
/*
Female      87
Male       157
*/
```

The pandas equivalent would be:

```
In [17]: tips.groupby('sex').size()
Out[17]:
sex
```

```
Female      87
Male       157
dtype: int64
```

Notice that in the pandas code we used `size()` and not `count()`. This is because `count()` applies the function to each column, returning the number of not null records within each.

```
In [18]: tips.groupby('sex').count()
Out[18]:
```

	total_bill	tip	smoker	day	time	size
sex						
Female	87	87	87	87	87	87
Male	157	157	157	157	157	157

Alternatively, we could have applied the `count()` method to an individual column:

```
In [19]: tips.groupby('sex')['total_bill'].count()
Out[19]:
```

sex	
Female	87
Male	157

Name: total_bill, dtype: int64

Multiple functions can also be applied at once. For instance, say we'd like to see how tip amount ~~differs~~ by day of the week - `agg()` allows you to pass a dictionary to your grouped DataFrame, indicating which functions to apply to specific columns.

```
SELECT day, AVG(tip), COUNT(*)
FROM tips
GROUP BY day;
/*
Fri    2.734737    19
Sat    2.993103    87
Sun    3.255132    76
Thur   2.771452    62
*/
```

```
In [20]: tips.groupby('day').agg({'tip': np.mean, 'day': np.size})
Out[20]:
```

	tip	day
day		
Fri	2.734737	19
Sat	2.993103	87
Sun	3.255132	76
Thur	2.771452	62

Grouping by more than one column is done by passing a list of columns to the `groupby()` method.

```
SELECT smoker, day, COUNT(*), AVG(tip)
FROM tips
GROUP BY smoker, day;
/*
```

```

smoker day
No      Fri      4  2.812500
        Sat     45  3.102889
        Sun     57  3.167895
        Thur    45  2.673778
Yes     Fri     15  2.714000
        Sat     42  2.875476
        Sun     19  3.516842
        Thur    17  3.030000
*/

```

```
In [21]: tips.groupby(['smoker', 'day']).agg({'tip': [np.size, np.mean]})
```

```

Out[21]:
           tip
smoker day size      mean
No      Fri    4.0  2.812500
        Sat   45.0  3.102889
        Sun   57.0  3.167895
        Thur  45.0  2.673778
Yes     Fri   15.0  2.714000
        Sat   42.0  2.875476
        Sun   19.0  3.516842
        Thur  17.0  3.030000

```

JOIN

JOINS can be performed with `join()` or `merge()`. By default, `join()` will join the DataFrames on their indices. Each method has parameters allowing you to specify the type of join to perform (LEFT, RIGHT, INNER, FULL) or the columns to join on (column names or indices).

```

In [22]: df1 = pd.DataFrame({'key': ['A', 'B', 'C', 'D'],
.....:                      'value': np.random.randn(4)})
.....:

In [23]: df2 = pd.DataFrame({'key': ['B', 'D', 'D', 'E'],
.....:                      'value': np.random.randn(4)})
.....:

```

Assume we have two database tables of the same name and structure as our DataFrames.

Now let's go over the various types of JOINS.

INNER JOIN

```

SELECT *
FROM df1
INNER JOIN df2
ON df1.key = df2.key;

```

```
# merge performs an INNER JOIN by default
In [24]: pd.merge(df1, df2, on='key')
Out[24]:
```

	key	value_x	value_y
0	B	-0.318214	0.543581
1	D	2.169960	-0.426067
2	D	2.169960	1.138079

`merge()` also offers parameters for cases when you'd like to join one DataFrame's column with another DataFrame's index.

```
In [25]: indexed_df2 = df2.set_index('key')

In [26]: pd.merge(df1, indexed_df2, left_on='key', right_index=True)
Out[26]:
```

	key	value_x	value_y
1	B	-0.318214	0.543581
3	D	2.169960	-0.426067
3	D	2.169960	1.138079

LEFT OUTER JOIN

```
-- show all records from df1
SELECT *
FROM df1
LEFT OUTER JOIN df2
ON df1.key = df2.key;
```

```
# show all records from df1
In [27]: pd.merge(df1, df2, on='key', how='left')
Out[27]:
```

	key	value_x	value_y
0	A	0.116174	NaN
1	B	-0.318214	0.543581
2	C	0.285261	NaN
3	D	2.169960	-0.426067
4	D	2.169960	1.138079

RIGHT JOIN

```
-- show all records from df2
SELECT *
FROM df1
RIGHT OUTER JOIN df2
ON df1.key = df2.key;
```

```
# show all records from df2
In [28]: pd.merge(df1, df2, on='key', how='right')
Out[28]:
```

	key	value_x	value_y
--	-----	---------	---------

```

0   B -0.318214  0.543581
1   D  2.169960 -0.426067
2   D  2.169960  1.138079
3   E         NaN  0.086073

```

FULL JOIN

pandas also allows for FULL JOINS, which display both sides of the dataset, whether or not the joined columns find a match. As of writing, FULL JOINS are not supported in all RDBMS (MySQL).

```

-- show all records from both tables
SELECT *
FROM df1
FULL OUTER JOIN df2
ON df1.key = df2.key;

```

```

# show all records from both frames
In [29]: pd.merge(df1, df2, on='key', how='outer')
Out[29]:
   key  value_x  value_y
0   A  0.116174      NaN
1   B -0.318214  0.543581
2   C  0.285261      NaN
3   D  2.169960 -0.426067
4   D  2.169960  1.138079
5   E         NaN  0.086073

```

UNION

UNION ALL can be performed using `concat()`.

```

In [30]: df1 = pd.DataFrame({'city': ['Chicago', 'San Francisco', 'New York City'],
.....:                      'rank': range(1, 4)})
.....:

In [31]: df2 = pd.DataFrame({'city': ['Chicago', 'Boston', 'Los Angeles'],
.....:                      'rank': [1, 4, 5]})
.....:

```

```

SELECT city, rank
FROM df1
UNION ALL
SELECT city, rank
FROM df2;
/*
      city  rank
Chicago    1
San Francisco  2
New York City  3
Chicago    1
Boston      4

```



```

    Los Angeles      5
*/

```

```

In [32]: pd.concat([df1, df2])
Out[32]:

```

	city	rank
0	Chicago	1
1	San Francisco	2
2	New York City	3
0	Chicago	1
1	Boston	4
2	Los Angeles	5

SQL's UNION is similar to UNION ALL, however UNION will remove duplicate rows.

```

SELECT city, rank
FROM df1
UNION
SELECT city, rank
FROM df2;
-- notice that there is only one Chicago record this time
/*
      city  rank
San Francisco  2
New York City  3
      Boston   4
Los Angeles   5
*/

```

In pandas, you can use `concat()` in conjunction with `drop_duplicates()`.

```

In [33]: pd.concat([df1, df2]).drop_duplicates()
Out[33]:

```

	city	rank
0	Chicago	1
1	San Francisco	2
2	New York City	3
1	Boston	4
2	Los Angeles	5

Pandas equivalents for some SQL analytic and aggregate functions

Top N rows with offset

```

-- MySQL
SELECT * FROM tips
ORDER BY tip DESC
LIMIT 10 OFFSET 5;

```

```
In [34]: tips.nlargest(10+5, columns='tip').tail(10)
Out[34]:
```

	total_bill	tip	sex	smoker	day	time	size
183	23.17	6.50	Male	Yes	Sun	Dinner	4
214	28.17	6.50	Female	Yes	Sat	Dinner	3
47	32.40	6.00	Male	No	Sun	Dinner	4
239	29.03	5.92	Male	No	Sat	Dinner	3
88	24.71	5.85	Male	No	Thur	Lunch	2
181	23.33	5.65	Male	Yes	Sun	Dinner	2
44	30.40	5.60	Male	No	Sun	Dinner	4
52	34.81	5.20	Female	No	Sun	Dinner	4
85	34.83	5.17	Female	No	Thur	Lunch	4
211	25.89	5.16	Male	Yes	Sat	Dinner	4

Top N rows per group

```
-- Oracle's ROW_NUMBER() analytic function
SELECT * FROM (
  SELECT
    t.*,
    ROW_NUMBER() OVER(PARTITION BY day ORDER BY total_bill DESC) AS rn
  FROM tips t
)
WHERE rn < 3
ORDER BY day, rn;
```

```
In [35]: (tips.assign(rn=tips.sort_values(['total_bill'], ascending=False)
.....:               .groupby(['day'])
.....:               .cumcount() + 1)
.....:   .query('rn < 3')
.....:   .sort_values(['day', 'rn'])
.....: )
Out[35]:
```

	total_bill	tip	sex	smoker	day	time	size	rn
95	40.17	4.73	Male	Yes	Fri	Dinner	4	1
90	28.97	3.00	Male	Yes	Fri	Dinner	2	2
170	50.81	10.00	Male	Yes	Sat	Dinner	3	1
212	48.33	9.00	Male	No	Sat	Dinner	4	2
156	48.17	5.00	Male	No	Sun	Dinner	6	1
182	45.35	3.50	Male	Yes	Sun	Dinner	3	2
197	43.11	5.00	Female	Yes	Thur	Lunch	4	1
142	41.19	5.00	Male	No	Thur	Lunch	5	2

the same using `rank(method='first')` function

```
In [36]: (tips.assign(rnk=tips.groupby(['day'])['total_bill']
.....:               .rank(method='first', ascending=False))
.....:   .query('rnk < 3')
.....:   .sort_values(['day', 'rnk'])
.....: )
Out[36]:
```

	total_bill	tip	sex	smoker	day	time	size	rnk
95	40.17	4.73	Male	Yes	Fri	Dinner	4	1.0
90	28.97	3.00	Male	Yes	Fri	Dinner	2	2.0

170	50.81	10.00	Male	Yes	Sat	Dinner	3	1.0
212	48.33	9.00	Male	No	Sat	Dinner	4	2.0
156	48.17	5.00	Male	No	Sun	Dinner	6	1.0
182	45.35	3.50	Male	Yes	Sun	Dinner	3	2.0
197	43.11	5.00	Female	Yes	Thur	Lunch	4	1.0
142	41.19	5.00	Male	No	Thur	Lunch	5	2.0

```
-- Oracle's RANK() analytic function
SELECT * FROM (
  SELECT
    t.*,
    RANK() OVER(PARTITION BY sex ORDER BY tip) AS rnk
  FROM tips t
  WHERE tip < 2
)
WHERE rnk < 3
ORDER BY sex, rnk;
```

Let's find tips with (rank < 3) per gender group for (tips < 2). Notice that when using `rank(method='min')` function `rnk_min` remains the same for the same `tip` (as Oracle's RANK() function)

```
In [37]: (tips[tips['tip'] < 2]
.....:      .assign(rnk_min=tips.groupby(['sex'])['tip']
.....:              .rank(method='min'))
.....:      .query('rnk_min < 3')
.....:      .sort_values(['sex', 'rnk_min'])
.....: )
.....:
Out[37]:
```

	total_bill	tip	sex	smoker	day	time	size	rnk_min
67	3.07	1.00	Female	Yes	Sat	Dinner	1	1.0
92	5.75	1.00	Female	Yes	Fri	Dinner	2	1.0
111	7.25	1.00	Female	No	Sat	Dinner	1	1.0
236	12.60	1.00	Male	Yes	Sat	Dinner	2	1.0
237	32.83	1.17	Male	Yes	Sat	Dinner	2	2.0

UPDATE

```
UPDATE tips
SET tip = tip*2
WHERE tip < 2;
```

```
In [38]: tips.loc[tips['tip'] < 2, 'tip'] *= 2
```

DELETE

```
DELETE FROM tips
WHERE tip > 9;
```

In pandas we select the rows that should remain, instead of deleting them

```
In [39]: tips = tips.loc[tips['tip'] <= 9]
```

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