# Select rows from a DataFrame based on values in a column in pandas

Ask Question



How to select rows from a DataFrame based on values in some column in pandas?

1272

In SQL I would use:



select \* from table where colume\_name = some\_value.

773

I tried to look at pandas documentation but did not immediately find the answer.

python

pandas

dataframe

edited Apr 17 '16 at 21:49



Thanos 1,475 1 7 25

asked Jun 12 '13 at 17:42



👯 szli

**7,527** 6 23 33

#### Check here:

github.com/debaonline4u/Python\_Pro gramming/tree/master/... – debaonline4u Jul 10 '18 at 19:49 /

df.query and pd.eval seem like good fits for this use case. For information on the pd.eval() family of functions, their features and use cases, please visit <a href="Dynamic Expression Evaluation in pandas using\_pd.eval">Dynamic Expression Evaluation in pandas using\_pd.eval</a>(). — coldspeed Dec 16 '18 at 4:54

This is a Comparison with SQL: pandas.pydata.org/pandas-docs/stable/comparison with sql.html where you can run pandas as SQL. – i\_th Jan 5 at 8:07

# 14 Answers



To select rows whose column value equals a scalar, some\_value, use

2480

== :

```
To select rows whose column value is in an iterable, <code>some_values</code>, use <code>isin</code>:
```

```
df.loc[df['column_name'].isin(some
```

Combine multiple conditions with &:

```
df.loc[(df['column_name'] >= A) &
```

Note the parentheses. Due to Python's <u>operator precedence rules</u>, & binds more tightly than <= and >= . Thus, the parentheses in the last example are necessary. Without the parentheses

```
df['column_name'] >= A & df['colum
```

is parsed as

```
df['column_name'] >= (A & df['colu
```

which results in a <u>Truth value of a Series is ambiguous error</u>.

To select rows whose column value does not equal some\_value, use !=:

```
df.loc[df['column_name'] != some_v
```

is in returns a boolean Series, so to select rows whose value is not in some\_values , negate the boolean Series using  $\sim$ :

```
df.loc[~df['column_name'].isin(som
```

For example,

```
import pandas as pd
import numpy as np
df = pd.DataFrame({'A': 'foo bar f
                   'B': 'one one t
                   'C': np.arange(
print(df)
             B C
                    D
# 0
    foo
           one 0
                    0
# 1 bar
           one 1
                    2
# 2
    foo
           two 2
# 3
         three 3
                    6
    bar
# 4
    foo
           two 4
                    8
# 5
           two 5 10
    bar
# 6
    foo
           one
                6
                   12
    foo three
                   14
print(df.loc[df['A'] == 'foo'])
```

vields

```
2 foo two 2 4
4 foo two 4 8
6 foo one 6 12
7 foo three 7 14
```

If you have multiple values you want to include, put them in a list (or more generally, any iterable) and use isin:

```
print(df.loc[df['B'].isin(['one','
```

#### yields

```
Α
           В
              C
                  D
  foo
  bar
                  2
1
         one
              1
3
  bar
       three
              3
                  6
6
  foo
         one
              6
                 12
  foo
              7
       three
                 14
```

Note, however, that if you wish to do this many times, it is more efficient to make an index first, and then use df.loc:

```
df = df.set_index(['B'])
print(df.loc['one'])
```

#### yields

```
\begin{array}{cccccc} & A & C & D \\ B & & & \\ one & foo & \emptyset & \emptyset \\ one & bar & 1 & 2 \\ one & foo & 6 & 12 \end{array}
```

or, to include multiple values from the index use df.index.isin:

```
df.loc[df.index.isin(['one','two']
```

## yields

	Α	С	D
В			
one	foo	0	0
one	bar	1	2
two	foo	2	4
two	foo	4	8
two	bar	5	10
one	foo	6	12

edited Jan 18 at 2:47

```
answered Jun 12 '13 at 17:44

unutbu
```

```
df[df['colume name']==some value]
also works. But my first attempt,
df.where(df['colume_name']==some_
value) does not work... not sure why...
- szli Jun 12 '13 at 18:12
```

- When you use df.where(condition) , the condition has to have the same shape as df . - unutbu Jun 12 '13 at 18:19
- FYI: If you want to select a row based upon two (or more) labels (either requiring both or either), see stackoverflow.com/questions/317563 40/... – Shane Aug 1 '15 at 0:18
- What about the negative "isnotin" does that exist? - BlackHat Mar 24 '16 at 6:13
- @BlackHat: isin returns a boolean mask. To find rows not in some\_iterable, negate the boolean mask using ~ (a tilde). That df.loc[~df['column\_name'].isi n(some values)] - unutbu Mar 24 '16 at 10:27 🎤



# tl;dr

The pandas equivalent to 212



```
select * from table where column_na
```

is

```
table[table.column_name == some_vai
```

Multiple conditions:

```
table[(table.column_name == some_va
```

or

```
table.query('column_name == some_va
```

# Code example

```
import pandas as pd
# Create data set
d = {'foo':[100, 111, 222],
     'bar':[333, 444, 555]}
df = pd.DataFrame(d)
```

```
# 0 333 100
# 1 444 111
# 2 555 222

# Output only the row(s) in df when
df[df.foo == 222]

# Shows:
# bar foo
# 2 555 222
```

In the above code it is the line

df[df.foo == 222] that gives the
rows based on the column value, 222
in this case.

Multiple conditions are also possible:

```
df[(df.foo == 222) | (df.bar == 444
# bar foo
# 1 444 111
# 2 555 222
```

But at that point I would recommend using the <u>query</u> function, since it's less verbose and yields the same result:

```
df.query('foo == 222 | bar == 444')
```

edited Jun 28 '18 at 15:30

answered Jul 8 '15 at 15:17



- 3 I really like the approach here. Thanks for having added it. It seems a bit more elegant than the accepted answer which is still ok but this is great thanks. – kiltannen Apr 22 '18 at 5:21
- 1 query is the only answer here that is compatible with method chaining. It seems like it's the pandas analog to filter in dplyr. – Berk U. Apr 23 '18 at 17:26
- 2 Hi, in your third example (multiple columns) I think you need square brackets [ not round brackets ( on the outside. user2739472 Jun 28 '18 at 12:40
- 1 at first I thought that | was for AND, but of course it is OR-operator... – O95 Nov 7 '18 at 9:32

I like query a lot as it is very readable. It is worth noting that it also works for multi-index dataframes where one can also query on different index levels (see the answer <a href="here">here</a>). — Cleb Nov 25 '18 at 15:09

151





- 2. Positional indexing
- +500
- 3. Label indexing

4. API

For each base type, we can keep things simple by restricting ourselves to the pandas API or we can venture outside the API, usually into <code>numpy</code>, and speed things up.

I'll show you examples of each and guide you as to when to use certain techniques.

#### Setup

The first thing we'll need is to identify a condition that will act as our criterion for selecting rows. The OP offers up column\_name == some\_value . We'll start there and include some other common use cases.

Borrowing from @unutbu:

Assume our criterion is column 'A' = 'foo'

1.

Boolean indexing requires finding the true value of each row's 'A' column being equal to 'foo', then using those truth values to identify which rows to keep. Typically, we'd name this series, an array of truth values, mask. We'll do so here as well.

```
mask = df['A'] == 'foo'
```

We can then use this mask to slice or index the data frame

df[mask]

```
A B C D
0 foo one 0 0
2 foo two 2 4
4 foo two 4 8
6 foo one 6 12
7 foo three 7 14
```

This is one of the simplest ways to accomplish this task and if

consider an alternative way of creating the  $\ensuremath{\mathsf{mask}}$  .

#### 2.

Positional indexing has its use cases, but this isn't one of them. In order to identify where to slice, we first need to perform the same boolean analysis we did above. This leaves us performing one extra step to accomplish the same task.

```
mask = df['A'] == 'foo'
pos = np.flatnonzero(mask)
df.iloc[pos]
           В С
0
   foo
          one 0
                  0
2
   foo
          two
                  4
         two 4
   foo
         one 6 12
6
  foo
   foo three 7
```

#### 3

Label indexing can be very handy, but in this case, we are again doing more work for no benefit

```
df.set_index('A', append=True, droj
           В
             C
  foo
         one 0
                  4
2
   foo
         two 2
4
  foo
         two 4
                 8
         one 6 12
  foo
  foo three 7 14
```

#### 4.

pd.DataFrame.query is a very elegant/intuitive way to perform this task. But is often slower. However, if you pay attention to the timings below, for large data, the query is very efficient. More so than the standard approach and of similar magnitude as my best suggestion.

```
df.query('A == "foo"')

    A     B     C     D
0 foo     one    0     0
2 foo     two    2     4
4 foo     two    4     8
6 foo     one    6     12
7 foo     three    7     14
```

My preference is to use the Boolean mask

Actual improvements can be made by modifying how we create our Boolean mask.

forgo the overhead of creating another pd.Series

```
mask = df['A'].values == 'foo'
```

I'll show more complete time tests at the end, but just take a look at the performance gains we get using the sample data frame. First, we look at the difference in creating the mask

```
%timeit mask = df['A'].values == ''
%timeit mask = df['A'] == 'foo'
5.84 \mu s \pm 195 ns per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s \pm 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 166 \mu s + 4.45 \mu s per loop (mean \pm : 1
```

Evaluating the mask with the numpy array is ~ 30 times faster. This is partly due to numpy evaluation often being faster. It is also partly due to the lack of overhead necessary to build an index and a corresponding pd.Series object.

Next, we'll look at the timing for slicing with one mask versus the other.

```
mask = df['A'].values == 'foo'
%timeit df[mask]
mask = df['A'] == 'foo'
%timeit df[mask]

219 μs ± 12.3 μs per loop (mean ± :
239 μs ± 7.03 μs per loop (mean ± :
```

The performance gains aren't as pronounced. We'll see if this holds up over more robust testing.

#### mask alternative 2

We could have reconstructed the data frame as well. There is a big caveat when reconstructing a dataframe—you must take care of the dtypes when doing so!

Instead of df[mask] we will do this

```
pd.DataFrame(df.values[mask], df.i
```

If the data frame is of mixed type, which our example is, then when we get df.values the resulting array is of dtype object and consequently, all columns of the new data frame will be of dtype object. Thus requiring the astype(df.dtypes) and killing any potential performance gains.

```
216 \mus \pm 10.4 \mus per loop (mean \pm : 1.43 ms \pm 39.6 \mus per loop (mean \pm
```

However, if the data frame is not of mixed type, this is a very useful way to do it.

#### Given

```
np.random.seed([3,1415])
d1 = pd.DataFrame(np.random.randin)
          C D
                Е
    Α
       В
 0
   0
       2
          7
             3
                8
 1
   7
          6
             8
 2
          0
             4
                3
          2
   3
       6
 5
                9
   5
          7
             5
       3
 6
   8
          6
             1
   6
       2
          6
             6
                5
 8
   2 8 7
             5 8
%%timeit
 mask = d1['A'].values == 7
d1[mask]
 179 \mus \pm 8.73 \mus per loop (mean \pm !
Versus
%timeit
mask = d1['A'].values == 7
 pd.DataFrame(d1.values[mask], d1.ii
```

We cut the time in half.

## mask alternative 3

@unutbu also shows us how to use pd.Series.isin to account for each element of df['A'] being in a set of values. This evaluates to the same thing if our set of values is a set of one value, namely 'foo'. But it also generalizes to include larger sets of values if needed. Turns out, this is still pretty fast even though it is a more general solution. The only real loss is in intuitiveness for those not familiar with the concept.

87  $\mu$ s  $\pm$  5.12  $\mu$ s per loop (mean  $\pm$  st

```
mask = df['A'].isin(['foo'])
df[mask]
            В
               C
                    D
   foo
          one
2
   foo
          two
               2
                    4
4
   foo
          two
                4
                    8
```

However, as before, we can utilize numpy to improve performance while sacrificing virtually nothing. We'll use np.in1d

```
mask = np.in1d(df['A'].values, ['foundf[mask]]
```

	Α	В	C	D
0	foo	one	0	0
2	foo	two	2	4
4	foo	two	4	8
6	foo	one	6	12
7	foo	three	7	14

#### **Timing**

I'll include other concepts mentioned in other posts as well for reference. Code Below

Each Column in this table represents a different length data frame over which we test each function. Each column shows relative time taken, with the fastest function given a base index of 1.0.

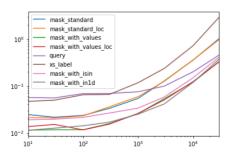
#### res.div(res.min())

		10	
10000	30000		
mask_stan	dard	2.156872	1.8
2.981326	3.131151		
mask_stan	dard_loc	1.879035	1.
2.998112	2.990103		
mask with	values	1.010166	1.0
1.007824	1.016919		
mask with	values loc	1.196843	1.3
1.037020	1.000000		
query		4.997304	4.
1.680447	1.398190		
xs label		4.124597	4.2
6.032809	8.950255		
mask with	isin	1.674055	1.6
1.253554	1.264760		
mask with	in1d	1.000000	1.0
1.000000	1.144175		

You'll notice that fastest times seem to be shared between

mask\_with\_values and
mask\_with\_in1d

#### res.T.plot(loglog=True)



```
def mask standard(df):
    mask = df['A'] == 'foo'
    return df[mask]
def mask_standard_loc(df):
    mask = df['A'] == 'foo'
    return df.loc[mask]
def mask_with_values(df):
    mask = df['A'].values == 'foo'
    return df[mask]
def mask_with_values_loc(df):
    mask = df['A'].values == 'foo'
    return df.loc[mask]
def query(df):
    return df.query('A == "foo"')
def xs_label(df):
    return df.set_index('A', append
def mask_with_isin(df):
    mask = df['A'].isin(['foo'])
    return df[mask]
def mask_with_in1d(df):
    mask = np.in1d(df['A'].values,
    return df[mask]
```

#### **Testing**

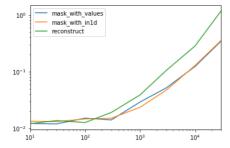
# **Special Timing**

Looking at the special case when we have a single non-object dtype for the entire data frame. Code Below

```
spec.div(spec.min())

10 30
10000 30000
mask_with_values 1.009030 1.00000
1.000000 1.000000
mask_with_inld 1.104638 1.09453
1.040043 1.027100
reconstruct 1.000000 1.14283
2.294913 3.406735
```

Turns out, reconstruction isn't worth it past a few hundred rows.



#### **Functions**

```
np.random.seed([3,1415])
d1 = pd.DataFrame(np.random.randint
def mask_with_values(df):
    mask = df['A'].values == 'foo'
    return df[mask]
def mask_with_in1d(df):
    mask = np.in1d(df['A'].values,
    return df[mask]
def reconstruct(df):
    v = df.values
    mask = np.in1d(df['A'].values,
    return pd.DataFrame(v[mask], d
spec = pd.DataFrame(
    index=['mask_with_values', 'mas
    columns=[10, 30, 100, 300, 1000
    dtype=float
)
```

#### **Testing**

```
for j in spec.columns:
    d = pd.concat([df] * j, ignore_
    for i in spec.index:
        stmt = '{}(d)'.format(i)
        setp = 'from __main__ impore
        spec.at[i, j] = timeit(stm*)
```

#### edited Dec 18 '18 at 15:08



Prakash Pazhanisamy 876 1 9 23

answered Sep 11 '17 at 22:14



piRSquared 160k 24 159 303

3 Fantastic answer! 2 questions though, i) how would .iloc(numpy.where(..)) compare in this scheme? ii) would you expect the rankings to be the same when using multiple conditions? – posdef Mar 6 '18 at 13:49

For performance of pd.Series.isin, note it does use np.in1d under the hood in a specific scenario, uses khash in others, and implicitly applies a trade-off between cost of hashing versus performance in specific situations. This answer has more detail — inp. Jun 17 '18 at 19:08



I find the syntax of the previous answers to be redundant and difficult to remember. Pandas introduced the query() method in v0.13 and I much prefer it. For your question, you could do df.query('col == val')

#### Reproduced from

http://pandas.pydata.org/pandas-docs/version/0.17.0/indexing.html#indexing-query

```
In [167]: n = 10
In [168]: df = pd.DataFrame(np.rane)
In [169]: df
Out[169]:
                   h
          а
   0.687704
            0.582314
                      0.281645
  0.250846 0.610021 0.420121
1
  0.624328 0.401816 0.932146
3
   0.011763 0.022921 0.244186
  0.590198 0.325680 0.890392
  0.598892 0.296424 0.007312
  0.634625 0.803069 0.123872
6
7
  0.924168 0.325076 0.303746
  0.116822 0.364564 0.454607
8
  0.986142 0.751953 0.561512
# pure python
In [170]: df[(df.a < df.b) & (df.b</pre>
Out[170]:
                    b
   0.011763 0.022921 0.244186
  0.116822 0.364564 0.454607
# auerv
In [171]: df.query('(a < b) & (b <</pre>
Out[171]:
   0.011763
            0.022921 0.244186
  0.116822 0.364564 0.454607
```

You can also access variables in the environment by prepending an @.

```
exclude = ('red', 'orange')
df.query('color not in @exclude')
```

answered Feb 9 '16 at 1:36



- 1 You only need package numexpr installed. MERose Mar 13 '16 at 9:16
- 3 In my case I needed quotation because val is a string. df.query('col == "val"') – smerlung Aug 10 '17 at 18:34

```
In [76]: df.iloc[np.where(df.A.val)
Out[76]:
            В
     Α
0
   foo
          one
               0
                   0
2
   foo
          two
               2
                    4
4
                   8
   foo
          two
               4
   foo
          one
              6 12
       three
              7
   foo
```

#### Timing comparisons:

```
In [68]: %timeit df.iloc[np.where(
1000 loops, best of 3: 380 μs per
In [69]: %timeit df.loc[df['A'] ==
1000 loops, best of 3: 745 μs per
In [71]: %timeit df.loc[df['A'].is:
1000 loops, best of 3: 562 μs per
In [72]: %timeit df[df.A=='foo']
1000 loops, best of 3: 796 μs per
In [74]: %timeit df.query('(A=="foo')
1000 loops, best of 3: 1.71 ms per
```

# edited Oct 3 '17 at 16:17 Brian Burns 7,082 5 46 48

answered Jul 5 '17 at 16:34



**3,997** 11 24



Here is a simple example

15 from pandas import DataFrame

```
# Create data set
d = {'Revenue':[100,111,222],
     'Cost':[333,444,555]}
df = DataFrame(d)
# mask = Return True when the value
mask = df['Revenue'] == 111
print mask
# Result:
# 0
       False
# 1
        True
# 2
       False
# Name: Revenue, dtype: bool
# Select * FROM df WHERE Revenue =
df[mask]
# Result:
     Cost
             Revenue
# 1 444
             111
```

answered Jun 13 '13 at 11:49





I just tried editing this, but I wasn't logged in, so I'm not sure where my edit went. I was trying to incorporate multiple selection. So I think a better answer is:

For a single value, the most straightforward (human readable) is probably:

```
df.loc[df['column_name'] == some_va
```

For lists of values you can also use:

```
df.loc[df['column_name'].isin(some_
```

For example,

```
import pandas as pd
import numpy as np
df = pd.DataFrame({'A': 'foo bar foo
               'B': 'one one two th
               'C': np.arange(8),
print(df)
             в с
      Α
# 0
    foo
            one 0
                    a
# 1
    bar
           one 1
                    2
           two 2
# 2
    foo
                    4
# 3
    bar
          three 3
# 4
    foo
            two 4
                    8
            two 5 10
# 5
    bar
           one 6 12
    foo
# 7 foo three 7 14
print(df.loc[df['A'] == 'foo'])
```

yields

```
В
               C
                    D
0
   foo
           one
                0
                    0
2
   foo
           two
                    4
   foo
           two 4
6
               6
                   12
   foo
           one
7
   foo
        three
                7
```

If you have multiple criteria you want to select against, you can put them in a list and use 'isin':

```
print(df.loc[df['B'].isin(['one','
```

yields

```
В С
                    D
  foo
          one 0
1
  bar
                   2
          one
              1
3
  bar
        three
               3
                   6
6
  foo
          one
              6
                  12
       three
              7
                 14
  foo
```

Note, however, that if you wish to do this many times, it is more efficient to make A the index first, and then use

```
yields
```

```
A B C D foo one 0 0 foo two 2 4 foo two 4 8 foo one 6 12 foo three 7 14
```

answered Jan 25 '15 at 23:27





If you finding rows based on some integer in a column, then

7

```
df.loc[df['column_name'] == 2017]
```

If you are finding value based on string

```
df.loc[df['column_name'] == 'string
```

If based on both

```
df.loc[(df['column_name'] == 'stri
```

answered Nov 16 '18 at 7:26

prateek singh

91 1 6





OUTPUT:

A B C D

0 foo one 0 0

2 foo two 2 4

4 foo two 4 8

6 foo one 6 12

7 foo three 7 14

answered Mar 6 '16 at 6:02



user15051990 **710** 7 17

5 How is this any different from imolit's answer? – MERose Mar 13 '16 at 9:15



To append to this famous question (though a bit too late): You can also

specified column having a particular value. E.g.

#### Run this gives:

```
Original dataframe:
             В
     Α
   foo
           one
1
   bar
          one
   foo
           two
   bar
        three
   foo
           two
   bar
           two
6
   foo
          one
   foo
        three
Sub dataframe where B is two:
     Α
          R
   foo
        two
1
   foo
        two
   bar
        two
```

answered Nov 18 '16 at 12:10



TuanDT 1,190 6 22



For selecting only specific columns out of multiple columns for a given value in pandas:



select col\_name1, col\_name2 from ta

# Options:

```
df.loc[df['column_name'] == some_va

or

df.query['column_name' == 'some_va'
```

edited Jun 22 '18 at 7:44



answered Dec 7 '17 at 10:39



SP001 69 1

4

those whose column's value is NOT any of a list of values, here's how to flip around unutbu's answer for a list of values above:

```
df.loc[~df['column_name'].isin(some
```

(To not include a single value, of course, you just use the regular not equals operator, != .)

#### Example:

#### gives us

```
В
   foo
           one
   bar
           one
   foo
           two
3
        three
   bar
   foo
           two
   bar
           two
   foo
          one
   foo
       three
```

To subset to just those rows that AREN'T one or three in column B:

```
df.loc[~df['B'].isin(['one', 'three
```

# yields

```
A B
2 foo two
4 foo two
5 bar two
```

answered Nov 12 '15 at 20:03





You can also use .apply:



df.apply(lambda row: row[df['B'].ig



It actually works row-wise (i.e., applies the function to each row).

#### The output is

	Α	ВС		D
0	foo	one	0	0
1	bar	one	1	2
3	bar	three	3	6
6	foo	ana	6	1 2

The results is the same as using as mentioned by @unutbu

df[[df['B'].isin(['one','three'])]]

answered Dec 7 '18 at 17:38



Vahidn





df.loc[df['column\_name'] == some\_va



answered Feb 10 at 19:36



John Nero

# protected by jezrael Feb 24 '18 at 18:33

Thank you for your interest in this question. Because it has attracted lowquality or spam answers that had to be removed, posting an answer now requires 10 reputation on this site (the association bonus does not count).

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