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Pivot Tables

Pivoting data can sometimes help clarify relationships and connections.

Full documentation on a variety of related pivot methods:

https://pandas.pydata.org/docs/user_guide/reshaping.html (https://pandas.pydata.org/docs/user_guide/reshaping.html)

Data

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

```
In [2]: df = pd.read_csv('Sales_Funnel_CRM.csv')
```

In [3]:

Out[3]:

	Account Number	Company	Contact	Account Manager	Product	Licenses	Sale Price	Status
0	2123398	Google	Larry Pager	Edward Thorp	Analytics	150	2100000	Presented
1	2123398	Google	Larry Pager	Edward Thorp	Prediction	150	700000	Presented
2	2123398	Google	Larry Pager	Edward Thorp	Tracking	300	350000	Under Review
3	2192650	вово	Larry Pager	Edward Thorp	Analytics	150	2450000	Lost
4	420496	IKEA	Elon Tusk	Edward Thorp	Analytics	300	4550000	Won
5	636685	Tesla Inc.	Elon Tusk	Edward Thorp	Analytics	300	2800000	Under Review
6	636685	Tesla Inc.	Elon Tusk	Edward Thorp	Prediction	150	700000	Presented
7	1216870	Microsoft	Will Grates	Edward Thorp	Tracking	300	350000	Under Review
8	2200450	Walmart	Will Grates	Edward Thorp	Analytics	150	2450000	Lost
9	405886	Apple	Cindy Phoner	Claude Shannon	Analytics	300	4550000	Won
10	470248	Exxon Mobile	Cindy Phoner	Claude Shannon	Analytics	150	2100000	Presented
11	698032	ATT	Cindy Phoner	Claude Shannon	Tracking	150	350000	Under Review
12	698032	ATT	Cindy Phoner	Claude Shannon	Prediction	150	700000	Presented
13	902797	CVS Health	Emma Gordian	Claude Shannon	Tracking	450	490000	Won
14	2046943	Salesforce	Emma Gordian	Claude Shannon	Analytics	750	7000000	Won
15	2169499	Cisco	Emma Gordian	Claude Shannon	Analytics	300	4550000	Lost
16	2169499	Cisco	Emma Gordian	Claude Shannon	GPS Positioning	300	350000	Presented

The pivot() method

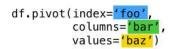
The pivot method reshapes data based on column values and reassignment of the index. Keep in mind, it doesn't always make sense to pivot data. In our machine learning lessons, we will see that our data doesn't need to be pivoted. Pivot methods are mainly for data analysis, visualization, and exploration.

Here is an image showing the idea behind a pivot() call:

Pivot

df





bar	A	В	С
foo			
one	1	2	3

In [4]: help(pd.pivot)

Help on function pivot in module pandas.core.reshape.pivot:

pivot(data:'DataFrame', index=None, columns=None, values=None) -> 'DataFra
me'

Return reshaped DataFrame organized by given index / column values.

Reshape data (produce a "pivot" table) based on column values. Uses unique values from specified `index` / `columns` to form axes of the resulting DataFrame. This function does not support data aggregation, multiple values will result in a MultiIndex in the columns. See the :ref:`User Guide <reshaping>` for more on reshaping.

Parameters

data: DataFrame

index : str or object, optional

Column to use to make new frame's index. If None, uses existing index.

columns : str or object

Column to use to make new frame's columns.

values : str, object or a list of the previous, optional Column(s) to use for populating new frame's values. If not specified, all remaining columns will be used and the result will have hierarchically indexed columns.

.. versionchanged:: 0.23.0
Also accept list of column names.

Returns

DataFrame

Returns reshaped DataFrame.

Raises

ValueError:

When there are any `index`, `columns` combinations with multiple values. `DataFrame.pivot_table` when you need to aggregate.

See Also

DataFrame.pivot_table : Generalization of pivot that can handle
 duplicate values for one index/column pair.

DataFrame.unstack: Pivot based on the index values instead of a column.

Notes

For finer-tuned control, see hierarchical indexing documentation along with the related stack/unstack methods.

Examples

```
1
    one
          В
                2
                     У
2
          C
    one
                3
                     Z
3
    two
          Α
                4
                     q
4
    two
          В
                5
                     W
5
          C
                     +
    two
                6
>>> df.pivot(index='foo', columns='bar', values='baz')
foo
one
    1
         2
              3
         5
     4
              6
two
>>> df.pivot(index='foo', columns='bar')['baz']
         В
har
foo
one
     1
         2
              3
         5
two
>>> df.pivot(index='foo', columns='bar', values=['baz', 'zoo'])
                 Z00
                 A B C
bar
         В
            C
foo
one
      1
         2
             3
                       Z
                 Х
                    У
two
         5
                 q
A ValueError is raised if there are any duplicates.
>>> df = pd.DataFrame({"foo": ['one', 'one', 'two', 'two'],
                         "bar": ['A', 'A', 'B', 'C'],
"baz": [1, 2, 3, 4]})
. . .
>>> df
   foo bar
             baz
  one
         Α
               1
1
         Α
               2
  one
2
   two
         В
               3
3
         C
               4
   two
Notice that the first two rows are the same for our `index`
and `columns` arguments.
>>> df.pivot(index='foo', columns='bar', values='baz')
Traceback (most recent call last):
ValueError: Index contains duplicate entries, cannot reshape
```

Note: Common Point of Confusion: Students often just randomly pass in index,column, and value choices in an attempt to see the changes. This often just leads to formatting errors. You should first go through this checklist BEFORE running a pivot():

- · What question are you trying to answer?
- What would a dataframe that answers the question look like? Does it need a pivot()
- What you want the resulting pivot to look like? Do you need all the original columns?

In [5]:

Out[5]:

	Account Number	Company	Contact	Account Manager	Product	Licenses	Sale Price	Status
0	2123398	Google	Larry Pager	Edward Thorp	Analytics	150	2100000	Presented
1	2123398	Google	Larry Pager	Edward Thorp	Prediction	150	700000	Presented
2	2123398	Google	Larry Pager	Edward Thorp	Tracking	300	350000	Under Review
3	2192650	вово	Larry Pager	Edward Thorp	Analytics	150	2450000	Lost
4	420496	IKEA	Elon Tusk	Edward Thorp	Analytics	300	4550000	Won
5	636685	Tesla Inc.	Elon Tusk	Edward Thorp	Analytics	300	2800000	Under Review
6	636685	Tesla Inc.	Elon Tusk	Edward Thorp	Prediction	150	700000	Presented
7	1216870	Microsoft	Will Grates	Edward Thorp	Tracking	300	350000	Under Review
8	2200450	Walmart	Will Grates	Edward Thorp	Analytics	150	2450000	Lost
9	405886	Apple	Cindy Phoner	Claude Shannon	Analytics	300	4550000	Won
10	470248	Exxon Mobile	Cindy Phoner	Claude Shannon	Analytics	150	2100000	Presented
11	698032	ATT	Cindy Phoner	Claude Shannon	Tracking	150	350000	Under Review
12	698032	ATT	Cindy Phoner	Claude Shannon	Prediction	150	700000	Presented
13	902797	CVS Health	Emma Gordian	Claude Shannon	Tracking	450	490000	Won
14	2046943	Salesforce	Emma Gordian	Claude Shannon	Analytics	750	7000000	Won
15	2169499	Cisco	Emma Gordian	Claude Shannon	Analytics	300	4550000	Lost
16	2169499	Cisco	Emma Gordian	Claude Shannon	GPS Positioning	300	350000	Presented

Imagine we wanted to know, how many licenses of each product type did Google purchase? Currently the way the data is formatted is hard to read. Let's pivot it so this is clearer, we will take a subset of the data for the question at hand.

^{**} What type of question does a pivot help answer?**

Out[6]:		Company	Product	Licenses
	0	Google	Analytics	150
	1	Google	Prediction	150
	2	Google	Tracking	300
	3	вово	Analytics	150
	4	IKEA	Analytics	300
	5	Tesla Inc.	Analytics	300
	6	Tesla Inc.	Prediction	150
	7	Microsoft	Tracking	300
	8	Walmart	Analytics	150
	9	Apple	Analytics	300
	10	Exxon Mobile	Analytics	150
	11	ATT	Tracking	150
	12	ATT	Prediction	150
	13	CVS Health	Tracking	450
	14	Salesforce	Analytics	750
	15	Cisco	Analytics	300
	16	Cisco	GPS Positioning	300

In [7]: pd.pivot(data=licenses,index='Company',columns='Product',values='Licenses')

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Product	Analytics	GPS Positioning	Prediction	Tracking
Company				
Google	150.0	NaN	150.0	300.0
ATT	NaN	NaN	150.0	150.0
Apple	300.0	NaN	NaN	NaN
вово	150.0	NaN	NaN	NaN
CVS Health	NaN	NaN	NaN	450.0
Cisco	300.0	300.0	NaN	NaN
Exxon Mobile	150.0	NaN	NaN	NaN
IKEA	300.0	NaN	NaN	NaN
Microsoft	NaN	NaN	NaN	300.0
Salesforce	750.0	NaN	NaN	NaN
Tesla Inc.	300.0	NaN	150.0	NaN
Walmart	150.0	NaN	NaN	NaN

The pivot_table() method

Similar to the pivot() method, the pivot_table() can add aggregation functions to a pivot call.

In [8]: |df

Out[8]:

	Account Number	Company	Contact	Account Manager	Product	Licenses	Sale Price	Status
0	2123398	Google	Larry Pager	Edward Thorp	Analytics	150	2100000	Presented
1	2123398	Google	Larry Pager	Edward Thorp	Prediction	150	700000	Presented
2	2123398	Google	Larry Pager	Edward Thorp	Tracking	300	350000	Under Review
3	2192650	вово	Larry Pager	Edward Thorp	Analytics	150	2450000	Lost
4	420496	IKEA	Elon Tusk	Edward Thorp	Analytics	300	4550000	Won
5	636685	Tesla Inc.	Elon Tusk	Edward Thorp	Analytics	300	2800000	Under Review
6	636685	Tesla Inc.	Elon Tusk	Edward Thorp	Prediction	150	700000	Presented
7	1216870	Microsoft	Will Grates	Edward Thorp	Tracking	300	350000	Under Review
8	2200450	Walmart	Will Grates	Edward Thorp	Analytics	150	2450000	Lost
9	405886	Apple	Cindy Phoner	Claude Shannon	Analytics	300	4550000	Won
10	470248	Exxon Mobile	Cindy Phoner	Claude Shannon	Analytics	150	2100000	Presented
11	698032	ATT	Cindy Phoner	Claude Shannon	Tracking	150	350000	Under Review
12	698032	ATT	Cindy Phoner	Claude Shannon	Prediction	150	700000	Presented
13	902797	CVS Health	Emma Gordian	Claude Shannon	Tracking	450	490000	Won
14	2046943	Salesforce	Emma Gordian	Claude Shannon	Analytics	750	7000000	Won
15	2169499	Cisco	Emma Gordian	Claude Shannon	Analytics	300	4550000	Lost
16	2169499	Cisco	Emma Gordian	Claude Shannon	GPS Positioning	300	350000	Presented

In [9]: # Notice Account Number sum() doesn't make sense to keep/use
pd.pivot_table(df,index="Company",aggfunc='sum')

Out[9]:

Account Number Licenses Sale Price

Company			
Google	6370194	600	3150000
ATT	1396064	300	1050000
Apple	405886	300	4550000
вово	2192650	150	2450000
CVS Health	902797	450	490000
Cisco	4338998	600	4900000
Exxon Mobile	470248	150	2100000
IKEA	420496	300	4550000
Microsoft	1216870	300	350000
Salesforce	2046943	750	7000000
Tesla Inc.	1273370	450	3500000
Walmart	2200450	150	2450000

In [10]: # Either grab the columns
pd.pivot_table(df,index="Company",aggfunc='sum')[['Licenses','Sale Price']]

Out[10]:

Licenses Sale Price

Company		
Google	600	3150000
ATT	300	1050000
Apple	300	4550000
вово	150	2450000
CVS Health	450	490000
Cisco	600	4900000
Exxon Mobile	150	2100000
IKEA	300	4550000
Microsoft	300	350000
Salesforce	750	7000000
Tesla Inc.	450	3500000
Walmart	150	2450000

```
In [11]: # Or state them as wanted values
pd.pivot_table(df,index="Company",aggfunc='sum',values=['Licenses','Sale Pr
```

Out[11]:

L	icens	es	Sal	e P	rice

Company		
Google	600	3150000
ATT	300	1050000
Apple	300	4550000
вово	150	2450000
CVS Health	450	490000
Cisco	600	4900000
Exxon Mobile	150	2100000
IKEA	300	4550000
Microsoft	300	350000
Salesforce	750	7000000
Tesla Inc.	450	3500000
Walmart	150	2450000

In [12]: df.groupby('Company').sum()[['Licenses','Sale Price']]

Out[12]:

Licenses Sale Price

Company		
Google	600	3150000
ATT	300	1050000
Apple	300	4550000
вово	150	2450000
CVS Health	450	490000
Cisco	600	4900000
Exxon Mobile	150	2100000
IKEA	300	4550000
Microsoft	300	350000
Salesforce	750	7000000
Tesla Inc.	450	3500000
Walmart	150	2450000

pd.pivot_table(df,index=["Account Manager","Contact"],values=['Sale Price'] In [13]:

Out[13]: Sale Price

Account Manager	Contact	
Claude Shannon	Cindy Phoner	7700000
	Emma Gordian	12390000
	Elon Tusk	8050000
Edward Thorp	Larry Pager	5600000
	Will Grates	2800000

Columns are optional - they provide an additional way to segment the actual values you care about. The aggregation functions are applied to the values you list.

In [14]: pd.pivot_table(df,index=["Account Manager","Contact"],values=["Sale Price"]

Out[14]: sum

		:			Sale Price
	Product	Analytics	GPS Positioning	Prediction	Tracking
Account Manager	Contact				
Claude Shannon	Cindy Phoner	6650000.0	NaN	700000.0	350000.0
	Emma Gordian	11550000.0	350000.0	NaN	490000.0
	Elon Tusk	7350000.0	NaN	700000.0	NaN
Edward Thorp	Larry Pager	4550000.0	NaN	700000.0	350000.0
	Will Grates	2450000.0	NaN	NaN	350000.0

pd.pivot_table(df,index=["Account Manager","Contact"],values=["Sale Price"]

Out[15]: sum

> Sale Price Product Analytics GPS Positioning Prediction Tracking **Account Manager** Contact 6650000 **Cindy Phoner** 0 700000 350000 **Claude Shannon Emma Gordian** 11550000 350000 490000 7350000 0 700000 0 **Elon Tusk Edward Thorp Larry Pager** 4550000 0 700000 350000 Will Grates

2450000

0

0

350000

Out[16]: sum

Sale Price

	Product	Analytics	GPS Positioning	Prediction	Tracking	Analytics	GPS Positioning	Predic
Account Manager	Contact							
Claude	Cindy Phoner	6650000	0	700000	350000	3325000	0	70
Shannon	Emma Gordian	11550000	350000	0	490000	5775000	350000	
	Elon Tusk	7350000	0	700000	0	3675000	0	70
Edward Thorp	Larry Pager	4550000	0	700000	350000	2275000	0	70
	Will Grates	2450000	0	0	350000	2450000	0	
4								•

Out[17]:

Licenses

	Product	Analytics	GPS Positioning	Prediction	Tracking	Analytics	GPS Positioning	Predic
Account Manager	Contact							
Claude	Cindy Phoner	450	0	150	150	6650000	0	70
Shannon	Emma Gordian	1050	300	0	450	11550000	350000	
	Elon Tusk	600	0	150	0	7350000	0	70
Edward Thorp	Larry Pager	300	0	150	300	4550000	0	70
	Will Grates	150	0	0	300	2450000	0	
4								>

sum

Out[18]:

			Licenses	Sale Price
Account Manager	Contact	Product		
		Analytics	450	6650000
	Cindy Phoner	Prediction	150	700000
Claude Shannon		Tracking	150	350000
Claude Shannon		Analytics	1050	11550000
	Emma Gordian	GPS Positioning	300	350000
		Tracking	450	490000
	Elon Tusk	Analytics	600	7350000
	Eloli Tusk	Prediction	150	700000
		Analytics	300	4550000
Edward Thorp	Larry Pager	Prediction	150	700000
		Tracking	300	350000
	Will Grates	Analytics	150	2450000
	will Grates	Tracking	300	350000

Out[19]:

sum

Licenses Sale Price

			Liberioes	Ouic i iloc
Account Manager	Contact	Product		
		Analytics	450	6650000
	Cindy Phoner	Prediction	150	700000
Claude Shannon		Tracking	150	350000
Claude Shannon		Analytics	1050	11550000
	Emma Gordian	GPS Positioning	300	350000
		Tracking	450	490000
	Elon Tusk	Analytics	600	7350000
		Prediction	150	700000
		Analytics	300	4550000
Edward Thorp	Larry Pager	Prediction	150	700000
		Tracking	300	350000
	Will Grates	Analytics	150	2450000
	will Grates	Tracking	300	350000
All			4500	36540000

Out[20]:

sum

Sale Price

Account Manager	Status	
	Lost	4550000
Claude Shannon	Presented	3150000
Claude Shannon	Under Review	350000
	Won	12040000
	Lost	4900000
Edward Thorn	Presented	3500000
Edward Thorp	Under Review	3500000
	Won	4550000
All		36540000