



Real Python

Lesson #09

Transforming Data with Pandas

March 2019



Introduction to Pandas

Exploring Data with Pandas

Data Cleaning Basics


Data Aggregation

Combining Data with Pandas

Transforming Data with Pandas

Working with String in Pandas

Working with missing and duplicate data

- 
- The background of the slide is a photograph of a library. In the foreground, there is a large, light-colored wooden table with several blue upholstered chairs around it. In the background, there are tall wooden bookshelves filled with books. A small white metal cart is visible on the left side of the bookshelves.
- Transforming data with Pandas
 - a. Map
 - b. Apply
 - c. Applymap
 - d. Melt

Update from repository

```
git clone https://github.com/ivanovitchm/datascience_one_2019_1
```

Or

```
git pull
```



Dataset

^

745

World Happiness Report

Happiness scored according to economic production, social support, etc.



Sustainable Development Solutions Network • updated 2 years ago (Version 2)

[Data](#)[Kernels \(421\)](#)[Discussion \(6\)](#)[Activity](#)[Download \(29 KB\)](#)[New Kernel](#)

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economics, social sciences, emotion

Description

Context

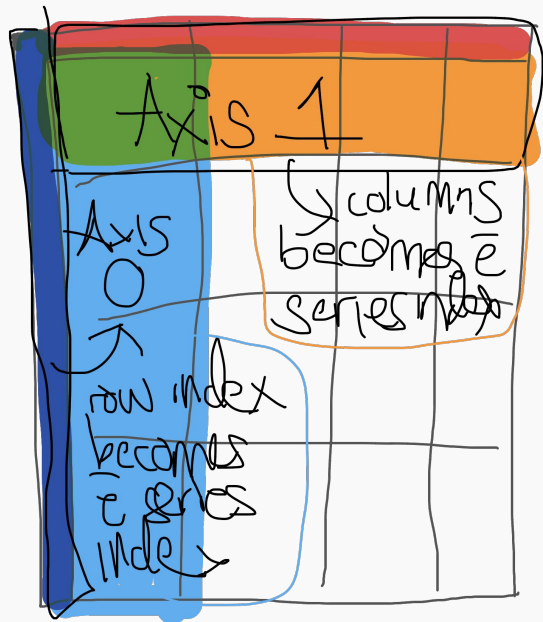
The World Happiness Report is a landmark survey of the state of global happiness. The first report was published in 2012, the second in 2013, the third in 2015, and the fourth in the 2016 Update. The World Happiness 2017, which ranks 155 countries by their happiness levels, was released at the United Nations at an event celebrating International Day of Happiness on March 20th. The report continues to gain global recognition as governments, organizations and civil society increasingly use happiness indicators to inform their policy-making decisions. Leading experts across fields – economics, psychology, survey analysis, national statistics, health, public policy and



Which of these factors contribute the most to the happiness score?

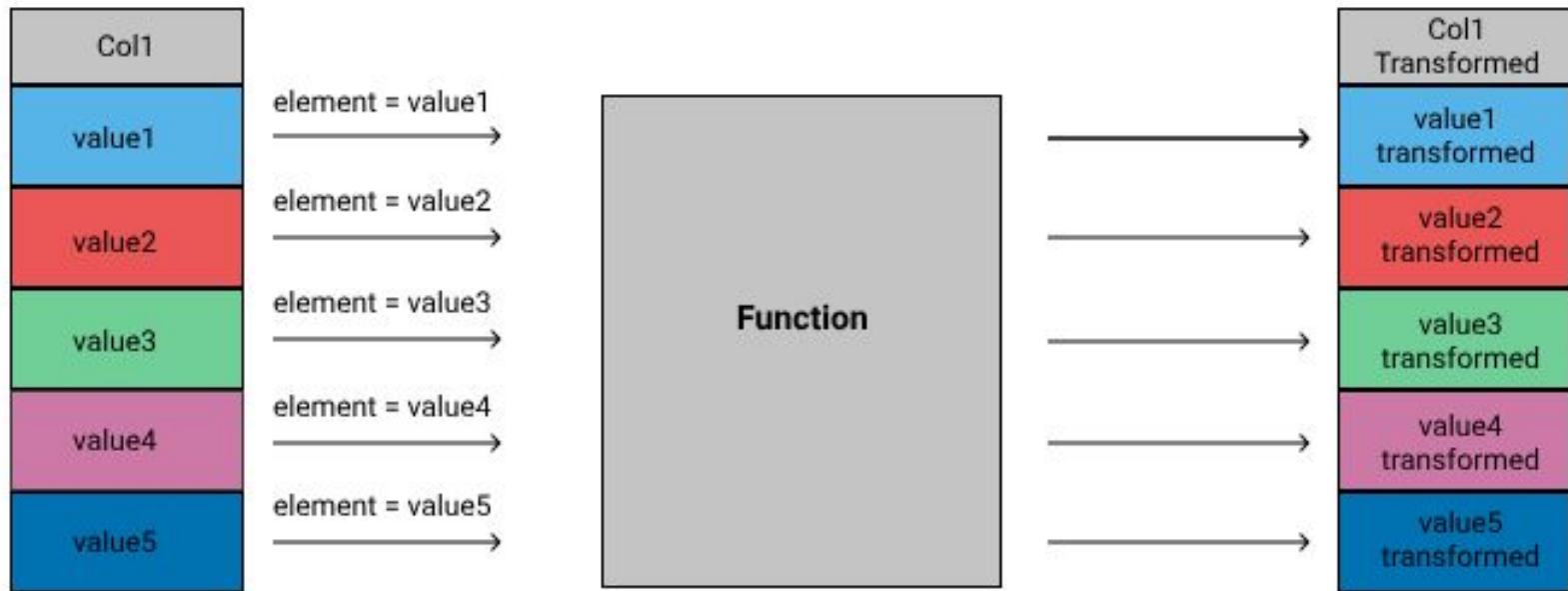


	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy	Family	Health	Freedom	Trust	Generosity
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811



- `Series.map()`
- `Series.apply()`
- `DataFrame.applymap()`
- `DataFrame.apply()`
- `Pandas.melt()`

Apply a Function Element-wise: **map**, **apply**




```
def label(element):  
    if element > 1:  
        return 'High'  
    else:  
        return 'Low'
```

Series.map()
Series.apply()

```
happiness2015['Economy Impact'] = happiness2015['Economy'].map(label)  
happiness2015['Economy Impact'] = happiness2015['Economy'].apply(label)
```

	Economy	Economy Impact
0	1.39651	High
1	1.30232	High
2	1.32548	High
3	1.45900	High
4	1.32629	High

```
def label(element, x):  
    if element > x:  
        return 'High'  
    else:  
        return 'Low'  
economy_map = happiness2015['Economy'].map(label, x = .8)
```



TypeError: map() got an unexpected keyword argument 'x'

```
def label(element,x):  
    if element > x:  
        return 'High'  
    else:  
        return 'Low'  
economy_impact_apply = happiness2015['Economy'].apply(label,x=0.8)
```



```
def label(element):  
    if element > 1:  
        return 'High'  
    else:  
        return 'Low'
```

Apply a Function **Element-wise** to Multiple Columns Using Applymap Method

```
happiness2015['Economy Impact'] = happiness2015['Economy'].apply(label)  
happiness2015['Health Impact'] = happiness2015['Health'].apply(label)  
happiness2015['Family Impact'] = happiness2015['Family'].apply(label)
```



Dataframe.applymap()

```
factors = ['Economy', 'Family', 'Health', 'Freedom', 'Trust', 'Generosity']  
factors_impact = happiness2015[factors].applymap(label)
```

	Economy	Family	Health	Freedom	Trust	Generosity
0	High	High	Low	Low	Low	Low
1	High	High	Low	Low	Low	Low
2	High	High	Low	Low	Low	Low
3	High	High	Low	Low	Low	Low
4	High	High	Low	Low	Low	Low

```
def label(element):  
    if element > 1:  
        return 'High'  
    else:  
        return 'Low'
```

Dataframe.apply()

```
factors = ['Economy', 'Family', 'Health', 'Freedom', 'Trust', 'Generosity']  
factors_impact = happiness2015[factors].apply(label)
```



ValueError: ('The truth value of a Series is ambiguous.

factors_impact

	Economy	Family	Health	Freedom	Trust	Generosity
0	High	High	Low	Low	Low	Low
1	High	High	Low	Low	Low	Low
2	High	High	Low	Low	Low	Low
3	High	High	Low	Low	Low	Low
4	High	High	Low	Low	Low	Low

Apply Functions
along an Axis
using the Apply
Method

factors_impact.apply(pd.value_counts)

	Economy	Family	Health	Freedom	Trust	Generosity
High	66	89	2	NaN	NaN	NaN
Low	92	69	156	158.0	158.0	158.0

`factors_impact.apply(pd.value_counts)`

`factors_impact`
`['Economy'].`
`value_counts()`

`factors_impact`
`['Family'].`
`value_counts()`

`factors_impact`
`['Health'].`
`value_counts()`

`factors_impact`
`['Freedom'].`
`value_counts()`

`factors_impact`
`['Trust'].`
`value_counts()`

`factors_impact`
`['Generosity'].`
`value_counts()`

	Economy	Family	Health	Freedom	Trust	Generosity
High	66	89	2	NaN	NaN	NaN
Low	92	69	156	158.0	158.0	158.0

```
def v_counts(col):  
    num = col.value_counts()  
    den = col.size  
    return num/den  
factors_impact.apply(v_counts)
```

	Economy	Family	Health	Freedom	Trust	Generosity
High	0.417722	0.563291	0.012658	NaN	NaN	NaN
Low	0.582278	0.436709	0.987342	1.0	1.0	1.0

	Country	Happiness Score	Economy	Family	Health
0	Switzerland	7.587	1.39651	1.34951	0.94143
1	Iceland	7.561	1.30232	1.40223	0.94784

```
pd.melt(happy_two, id_vars=['Country'], value_vars=['Economy', 'Family', 'Health'])
```

	Country	variable	value
0	Switzerland	Economy	1.39651
1	Iceland	Economy	1.30232
2	Switzerland	Family	1.34951
3	Iceland	Family	1.40223
4	Switzerland	Health	0.94143
5	Iceland	Health	0.94784

Reshaping Data with
the Melt Function
#tidydata

Lesson #09

1 - Transforming Data With Pandas

