Python For Data Science Cheat Sheet

Pandas Basics

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Pandas

The **Pandas** library is built on NumPy and provides easy-to-use data structures and data analysis tools for the Python programming language.

Use the following import convention:

>>> import pandas as pd

Pandas Data Structures

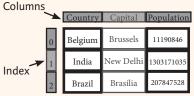
Series

A one-dimensional labeled array capable of holding any data type



>>> s = pd.Series([3, -5, 7, 4], index=['a', 'b', 'c', 'd'])

DataFrame



A two-dimensional labeled data structure with columns of potentially different types

```
>>> data = {'Country': ['Belgium', 'India', 'Brazil'],
           'Capital': ['Brussels', 'New Delhi', 'Brasília'],
           'Population': [11190846, 1303171035, 207847528]}
>>> df = pd.DataFrame(data,
                     columns=['Country', 'Capital', 'Population'])
```

Asking For Help

>>> help(pd.Series.loc)

Selection

Also see NumPy Arrays

Getting

```
>>> s['b']
  -5
>>> df[1:]
   Country
              Capital Population
  1 India New Delhi 1303171035
  2 Brazil
             Brasília 207847528
```

Get one element

Get subset of a DataFrame

Selecting, Boolean Indexing & Setting

By Position

```
>>> df.iloc[[0],[0]]
 'Belgium'
>>> df.iat([0],[0])
 'Belgium'
```

Select single value by row & column

By Label

```
>>> df.loc[[0], ['Country']]
 'Belgium'
>>> df.at([0], ['Country'])
  'Belgium'
```

Select single value by row & column labels

By Label/Position

. . 16 ! [0]

>>> df.ix[2]
Country Brazil
Capital Brasília Population 207847528
>>> df.ix[:,'Capital']
0 Brussels
1 New Delhi
2 Brasília
>>> df.ix[1,'Capital']

Select single row of subset of rows

Select a single column of subset of columns

Select rows and columns

Boolean Indexing

'New Delhi'

>>>	s[~(s > 1)]
>>>	s[(s < -1) (s > 2)]
>>>	df[df['Population']>120000
l _	·

Series s where value is not >1 s where value is <-1 or >2

Setting

>>> s['a'] = 6

b NaN 5.0 С

>>> s + s3

Dropping

Sort & Rank

>>> df.rank()

>>> df.shape

>>> df.index

>>> df.info()

>>> df.count()

Summary

>>> df.sum()

>>> df.cumsum()

>>> df.describe()

Applying Functions

>>> f = lambda x: x*2

Internal Data Alignment

>>> df.mean()

>>> df.apply(f)

Data Alignment

10.0

>>> df.applymap(f)

>>> df.median()

>>> df.min()/df.max()

>>> df.idxmin()/df.idxmax(

>>> df.columns

>>> df.sort index()

Basic Information

>>> df.sort values(by='Country')

>>> s.drop(['a', 'c'])

Drop values from rows (axis=0)

Sort by labels along an axis

Assign ranks to entries

Sort by the values along an axis

>>> df.drop('Country', axis=1) Drop values from columns(axis=1)

Retrieving Series/DataFrame Information

(rows,columns)

Describe index

Info on DataFrame

Describe DataFrame columns

Cummulative sum of values

Minimum/maximum values

Minimum/Maximum index value

Number of non-NA values

Sum of values

Summary statistics Mean of values

Median of values

Apply function element-wise

Apply function

7.0 d

Arithmetic Operations with Fill Methods

You can also do the internal data alignment yourself with the help of the fill methods:

NA values are introduced in the indices that don't overlap:

>>> s3 = pd.Series([7, -2, 3], index=['a', 'c', 'd'])

```
>>> s.add(s3, fill value=0)
 a 10.0
 b
      -5.0
     5.0
 С
 d
     7.0
>>> s.sub(s3, fill value=2)
>>> s.div(s3, fill value=4)
>>> s.mul(s3, fill value=3)
```

000001 Use filter to adjust DataFrame

Set index a of Series s to 6

Read and Write to CSV

```
>>> pd.read csv('file.csv', header=None, nrows=5)
>>> df.to csv('myDataFrame.csv')
```

Read and Write to Excel

```
>>> pd.read excel('file.xlsx')
>>> pd.to excel('dir/myDataFrame.xlsx', sheet name='Sheet1')
 Read multiple sheets from the same file
```

>>>	xlsx	= pd.ExcelFile('file.xls')
>>>	df =	<pre>pd.read_excel(xlsx, 'Sheet1')</pre>

Read and Write to SQL Query or Database Table

>>> from sqlalchemy import create_engine					
>>> engine = create_engine('sqlite:///:memory:')					
>>> pd.read_sql("SELECT * FROM my_table;", engine)					
>>> pd.read_sql_table('my_table', engine)					
>>> pd.read_sql_query("SELECT * FROM my_table;", engine)					
read sql() is a convenience wrapper around read sql table() and					

read sql query()

>>> pd.to sql('myDf', engine)

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Pandas

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Reshaping Data

Pivot

>>> df3= df2.pivot(index='Date', columns='Type', values='Value')

Spread rows into columns

	Date	Туре	Value			
0	2016-03-01	a	11.432	Туре	a	
1	2016-03-02	ь	13.031	Date		Г
2	2016-03-01	С	20.784	2016-03-01	11.432	N
3	2016-03-03	a	99.906	2016-03-02	1.303	13
4	2016-03-02	a	1.303	2016-03-03	99.906	N
5	2016-03-03	с	20.784			

20.784 3.031 NaN NaN 20.784

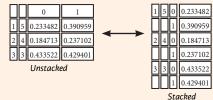
Pivot Table

>>> df4 = pd.pivot table(df2, values='Value' index='Date', columns='Type']

Spread rows into columns

Stack / Unstack

>>> stacked = df5.stack() Pivot a level of column labels >>> stacked.unstack() Pivot a level of index labels



Gather columns into rows >>> pd.melt(df2, id vars=["Date"], value_vars=["Type", "Value"], value name="Observations")

	Date	Type	Value	1		Date	Variable	Observations
		турс		!	0	2016-03-01	Type	a
0	2016-03-01	a	11.432		1	2016-03-02	Type	ь
1	2016-03-02	b	13.031		2	2016-03-01	Туре	С
2	2016-03-01	С	20.784]	3	2016-03-03	Туре	a
Ē	2016-03-03	a	99,906	→	4	2016-03-02	Type	a
믬		a			5	2016-03-03	Type	С
4	2016-03-02	a	1.303		6	2016-03-01	Value	11.432
5	2016-03-03	с	20.784		7	2016-03-02	Value	13.031
					8	2016-03-01	Value	20.784
					9	2016-03-03	Value	99.906
					10	2016-03-02	Value	1.303
					11	2016-03-03	Value	20.784

Iteration

Melt

(Column-index, Series) pairs >>> df.iteritems() (Row-index, Series) pairs >>> df.iterrows()

Advanced Indexing

Selecting

>>> df3.loc[:,(df3>1).any()] >>> df3.loc[:,(df3>1).all()] >>> df3.loc[:,df3.isnull().any()] >>> df3.loc[:,df3.notnull().all()]

Indexing With isin >>> df[(df.Country.isin(df2.Type))]

>>> df3.filter(items="a","b"]) >>> df.select(lambda x: not x%5) Where

>>> s.where(s > 0) Query

>>> df6.query('second > first')

Also see NumPy Arrays

Select cols with NaN Select cols without NaN Find same elements

Select cols with vals > 1

Select cols with any vals >1

Filter on values Select specific elements

Subset the data

Query DataFrame

Backward Filling

Setting/Resetting Index

<pre>>>> df.set_index('Country') >>> df4 = df.reset_index() >>> df = df.rename(index=str,</pre>	Set the index Reset the index Rename DataFrame
--	--

Reindexing

>>> s2 = s.reindex(['a','c','d','e','b'])

Forward Filling

		9				
>>>	df.reind	ex(range(4)		>>>	s3 =	s.reindex(range(5),
		method='	ffill')			method='bfill')
	Country	Capital	Population	0	3	
0	Belgium	Brussels	11190846	1	3	
1	India	New Delhi	1303171035	2	3	
2	Brazil	Brasília	207847528	3	3	
3	Brazil	Brasília	207847528	4	3	

MultiIndexing

```
>>> arrays = [np.array([1,2,3]),
              np.array([5,4,3])]
>>> df5 = pd.DataFrame(np.random.rand(3, 2), index=arrays)
>>> tuples = list(zip(*arrays))
>>> index = pd.MultiIndex.from tuples(tuples,
                                      names=['first', 'second'])
>>> df6 = pd.DataFrame(np.random.rand(3, 2), index=index)
>>> df2.set index(["Date", "Type"])
```

Duplicate Data

>>>	s3.unique()	Return unique values
>>>	df2.duplicated('Type')	Check duplicates
>>>	<pre>df2.drop_duplicates('Type', keep='last')</pre>	Drop duplicates
>>>	df.index.duplicated()	Check index duplicates
	>>> >>>	>>> s3.unique() >>> df2.duplicated('Type') >>> df2.drop_duplicates('Type', keep='last') >>> df.index.duplicated()

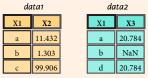
Grouping Data

	Aggregation
ı	>>> df2.groupby(by=['Date','Type']).mean()
ı	>>> df4.groupby(level=0).sum()
ı	>>> df4.groupby(level=0).agg({'a':lambda x:sum(x)/len(x),
ı	'b': np.sum})
	Transformation
ı	>>> customSum = lambda x: (x+x%2)
ı	>>> df4.groupby(level=0).transform(customSum)

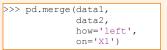
Missing Data

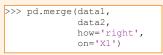
>>> df.dropna() >>> df3.fillna(df3.mean()) >>> df2.replace("a", "f")	Drop NaN values Fill NaN values with a predetermined value Replace values with others
--	---

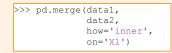
Combining Data



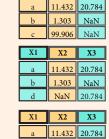
Merge







>>> pd.merge(data1,
data2,
how='outer',
on='X1')



X2 Х3



1.303

Oin

```
>>> data1.join(data2, how='right')
```

Concatenate

Vertical >>> s.append(s2)

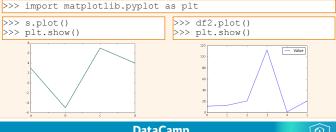
```
Horizontal/Vertical
>>> pd.concat([s,s2],axis=1, keys=['One','Two'])
>>> pd.concat([data1, data2], axis=1, join='inner')
```

Dates

```
>>> df2['Date'] = pd.to datetime(df2['Date'])
>>> df2['Date']= pd.date_range('2000-1-1',
                               periods=6,
                               freq='M')
>>> dates = [datetime(2012,5,1), datetime(2012,5,2)]
>>> index = pd.DatetimeIndex(dates)
>>> index = pd.date range(datetime(2012,2,1), end, freq='BM')
```

Visualization

Also see Matplotlib



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