# **Pandas Basics Cheat Sheet**



Use the following import convention: >>> import pandas as pd

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

# **Pandas Data Structures**

## Series

## A one-dimensional

labeled array a capable of holding any

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

## **Data Frame**

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

>>> data = {'Country': ['Belgium', 'India', 'Brazil'], 'Country': ['Belgium', 'India', 'Brazil'],

'Capital': ['Brussels', 'New Delhi', 'Brasília'],

'Population': [11190846, 1303171035n207847 >>> df = pd.DataFrame(data,

columns=['Country', 'Capital', 'Population'])

# **Dropping**

>>> s.drop(['a', 'c']) >>> df.drop('Country', axis=1)

Drop values from rows (axis=0) Drop values from columns(axis=1)

# Sort & Rank

>>> df.sort index() >>> df.sort\_values(by='Country') >>> df.rank()

Sort by labels along an axis Sort by the values along an axis Assign ranks to entries

Median of values

# Retrieving Series/ **DataFrame Information**

>>> df shane (rows.columns) >>> df index Describe index >>> df.columns Describe DataFrame columns >>> df infn() Info on DataFrame >>> df count() Number of non-NA values

## Summarv

>>> df median(

>>> df.sum() Sum of values >>> df.cumsum() Cummulative sum of values >>> df.min()/df.max() Minimum/maximum values >>> df.idxmin()/df.idxmax() Minimum/Maximum index value >>> df.describe() Summary statistics >>> df.mean() Mean of values

# Selection

Also see NumPy Arrays

# Getting

>>> s['b'] Get one element >>> df[1:] Get subset of a DataFrame Population Country Capital New Delhi 1303171035

## Selecting, Boolean Indexing & Setting

By Position

Select single value by row & >>> df.iloc[[0],[0]] 'Belgium' >>> df.iat([0],[0])

By Label

Select single value by row & >>> df.loc[[0], ['Country']] 'Belgium' >>> df.at([0], ['Country']) 'Belgium'

By Label/Position

>>> df.ix[2] Country Capital Brasília Population 207847528 Select a single column of >>> df.ix[:,'Ca 0 Brussels 1 New Delhi 2 Brasília >>> df.ix[1,'Capital']
'New Delhi' Select rows and columns

**Boolean Indexing** 

Series s where value is not >1 >>> s[~(s > 1)] >>> s[(s < -1) | (s > 2)] s where value is <-1 or >2 >>> df[df['Population']>1200000000] Use filter to adjust DataFrame

Setting

Set index a of Series s to 6 >>> s['a'] = 6

# **Asking For Help**

>>> help(pd.Series.loc)

# **Applying Functions**

>>> f = lambda x: x\*2 Apply function >>> df.apply(f) >>> df.applymap(f) Apply function element-wise

# **Data Alignment**

## **Internal Data Alignment**

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

>>> s3 = pd.Series([7, -2, 3], index=['a', 'c', 'd']) >>> s + s3 a 10.0 b NaN

# **Arithmetic Operations with Fill Methods**

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

>>> s.add(s3, fill\_value=0) a 10.0 **b -5.0** >>> s.sub(s3, fill value=2) >>> s.div(s3, fill\_value=4)

1/0

# Read and Write to CSV

>>> pd.read csv('file.csv', header=None, nrows=5) >>> df.to\_csv('mvDataFrame.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 = pd.read excel(xlsx, 'Sheet1')

# 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)

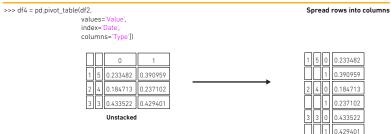
# **Pandas Cheat Sheet**

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# **Pandas Data Structures**

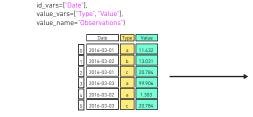
## **Pivot** >>> df3= df2.pivot(index='Date Spread rows into columns values='Value') Date Type Value 2016-03-0 Date 2016-03-2016-03-01 2016-03-03 2016-03-03 2016-03-03

# **Pivot Table**



# Melt

>>> pd.melt(df2,



Gather columns into rows

Stacked

	Date	Variable	Observations
0	2016-03-01	Type	a
1	2016-03-02	Туре	b
2	2016-03-01	Туре	С
3	2016-03-03	Туре	a
4	2016-03-02	Туре	a
5	2016-03-03	Type	С
6	2016-03-01	Value	11.432
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

# Advanced Indexing

## Also see NumPy Arrays

### Selecting >>> df3.loc[:,(df3>1).any()] Select cols with any vals >1 >>> df3.loc[:.(df3>1).all()] Select cols with vals > 1 >>> df3.loc[:.df3.isnull().anv()] Select cols with NaN >>> df3.loc[:.df3.notnull().all()] Select cols without NaN

## Indexing With isin

>>> df[(df.Country.isin(df2.Type))] Find same elements >>> df3.filter(items="a"."b"]) Filter on values >>> df.select(lambda x: not x%5) Select specific elements

>>> s where(s > 1)

Subset the data

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

Query DataFrame

# Setting/Resetting Index

ntry') Set the inde		
ex() Reset the inde		
dex=str, Rename DataFrame	>>> df = df.rename(index=str,	
lumns={"Country":"cntry",		

Forward Filling

4 3

>>> s3 = s.reindex(range(5)

# Reindexing

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

## Forward Filling

>>> df.reindex(range(4),

method='ffill')

0 3 Country Capital Population 0 Belgium Brussels 11190846 1 3 1 India New Delhi 1303171035 2 3 2 Brazil Brasília 207847528 3 3 207847528

# 3 Brazil Brasília 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() >>> df2.duplicated('Type')

>>> df2.drop\_duplicates('Type', keep='last')

>>> df.index.duplicated()

Return unique values Check duplicates Drop duplicates Drop duplicates

# **Grouping Data**

## Aggregation

>>> df2.groupby(by=['Date','Type']).mean()

>>> df4.groupby(level=0).sum()

>>> df4.groupby(level=0).agg({'a':lambda x:sum(x)/len(x), 'b': np.sum})

## Transformation

>>> customSum = lambda x: (x+x%2)

>>> df4.groupby(level=0).transform(customSum)

# Missing Data

>>> df.dropna() >>> df3 fillna(df3 mean()) >>> df2.replace("a", "f")

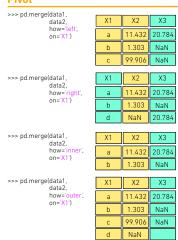
Drop NaN value Fill NaN values with a predetermined value Replace values with others

# **Combining Data**

# data1 X1 X2 11.432 1.303

data2 NaN

# **Pivot**



# Join

>>> 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,

>>> 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**

>>> import matplotlib.pyplot as plt

>>> s.plot() >>> plt.show() >>> df2.plot() >>> plt.show()