Module 2: Exploratory Data Analysis

Part 2: Two Dimensional Data

The Data Frame

```
df = pd.DataFrame( <data>, <index>, <column_names> )
```

CSV

```
df = pd.read_csv( <file> )
df = pd.read_csv( <url> )
```

Excel

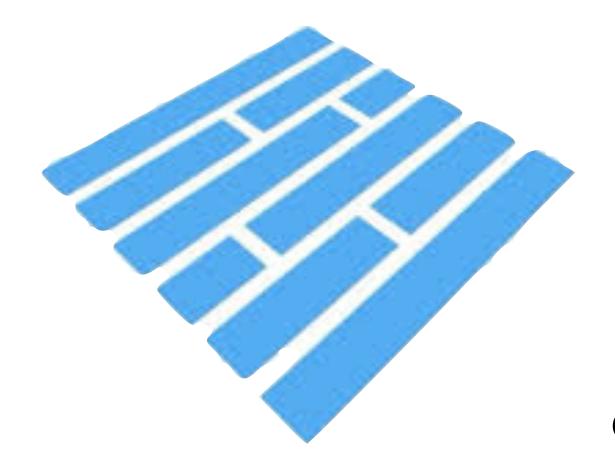
```
df = pd.read_excel( <file>, sheetname=<sheetname> )
```

```
"changed": "2015-04-0300:00:00",
"created": "2014-04-10 00:00:00",
"dnssec": "False",
"expires": "2016-04-10 00:00:00",
"isMalicious":0,
"url": "nuteczki.com"
```

JS0N

From a Database

```
df = pd.read_sql( <query>, <connection> )
```



Parquet

```
df = pd.read_parquet( <file> )
```

HTML

```
df = pd.read_html( <source> )
```

XML

```
import requests
user_agent_url = 'http://www.user-agents.org/allagents.xml'
xml_data = requests.get(user_agent_url).content
```

http://www.austintaylor.io/lxml/python/pandas/xml/dataframe/2016/07/08/convert-xml-to-pandas-dataframe/



```
import xml.etree.ElementTree as ET
class XML2DataFrame:
    def init (self, xml data):
        self.root = ET.XML(xml data)
    def parse root(self, root):
        return [self.parse_element(child) for child in iter(root)]
    def parse_element(self, element, parsed=None):
        if parsed is None:
            parsed = dict()
        for key in element.keys():
            parsed[key] = element.attrib.get(key)
        if element.text:
            parsed[element.tag] = element.text
        for child in list(element):
            self.parse element(child, parsed)
        return parsed
    def process data(self):
        structure_data = self.parse_root(self.root)
        return pd.DataFrame(structure_data)
xml2df = XML2DataFrame(xml data)
xml_dataframe = xml2df.process_data()
```

Log Files...

logdf = pd.read_table('../data/mysql.log', names=['raw'])

	raw
0	070823 21:00:32 1 Connect root@local
1	070823 21:00:48 1 Query show tables
2	070823 21:00:56 1 Query select * f
3	070917 16:29:01 21 Query select * f
4	070917 16:29:12 21 Query select * f

```
logdf = pd.read_table('../data/mysql.log', names=['raw'])
logdf['raw'].str.extract('(\d{6}\s\d{2}:\d{2}:\d{2})\s+(\d+)\s(\S+)
\s(.+)', expand=False)
```

	0	1	2	3
0	070823 21:00:32	1	Connect	root@localhost on test1
1	070823 21:00:48	1	Query	show tables
2	070823 21:00:56	1	Query	select * from category
3	070917 16:29:01	21	Query	select * from location
4	070917 16:29:12	21	Query	select * from location where id = 1 LIMIT 1

```
logdf = pd.read_table('../data/mysql.log', names=['raw'])
logdf['raw'].str.extract('(?P<date>\d{6}\s\d{2}:\d{2}:\d{2})\s+(?P<PID>\d+)\s(?P<Action>\S+)\s(?P<Query>.+)', expand=False)
```

	Date	PID	Action	Query
0	070823 21:00:32	1	Connect	root@localhost on test1
1	070823 21:00:48	1	Query	show tables
2	070823 21:00:56	1	Query	select * from category
3	070917 16:29:01	21	Query	select * from location
4	070917 16:29:12	21	Query	select * from location where id = 1 LIMIT 1



```
195.154.46.135 - - [25/Oct/2015:04:11:25 +0100] "GET /linux/doing-pxe-without-dhcp-control HTTP/1.1" 200 24323 "http://howto.basjes.nl/" "Mozilla/5.0 (Windows NT 5.1; rv:35.0) Gecko/20100101 Firefox/35.0"
```

```
195.154.46.135 - - [25/Oct/2015:04:11:25 +0100] "GET /linux/
doing-pxe-without-dhcp-control HTTP/1.1" 200 24323 "http://
howto.basjes.nl/" "Mozilla/5.0 (Windows NT 5.1; rv:35.0)
Gecko/20100101 Firefox/35.0"

import apache_log_parser
line_parser = apache_log_parser.make_parser("%h %l %u %t
\"%r\" %>s %b \"%{Referer}i\" \"%{User-agent}i\"")
```

```
import apache log parser
line parser = apache log parser.make parser("%h %l %u %t
\"%r\" %>s %b \"%{Referer}i\" \"%{User-agent}i\")
server log = open("../data/hackers-access.httpd", "r")
parsed server data = []
for line in server log:
    data = \{\}
    data = line parser(line)
    parsed server data.append( data )
server df = pd.DataFrame( parsed server data
```

_	III [43].						
er	request_first_line	request_header_referer	request_header_user_agent		request_method	request_url	
	GET /linux/doing-pxe-without-dhcp-control HTTP		Mozilla/5.0 (Windows NT 5.1; rv:35.0) Gecko/20		GET	/linux/doing-pxe-v	
	GET /join_form HTTP/1.0		Mozilla/5.0 (Windows NT 5.1; rv:35.0) Gecko/20		GET	/join_form	
	POST /join_form HTTP/1.1	http://howto.basjes.nl/join_form	Mozilla/5.0 (Windows NT 5.1; rv:35.0) Gecko/20	1.1	POST	/join_form	
	GET /join_form HTTP/1.0	http://howto.basjes.nl/	Mozilla/5.0 (Windows NT 6.3; WOW64; rv:34.0) G	1.0	GET	/join_form	
	POST /join_form HTTP/1.1	http://howto.basjes.nl/join_form	Mozilla/5.0 (Windows NT 6.3; WOW64; rv:34.0) G	1.1	POST	/join_form	
	GET /acl_users/credentials_cookie_auth/require	http://howto.basjes.nl/join_form	Mozilla/5.0 (Windows NT 6.3; WOW64; rv:34.0) G	1.1	GET	/acl_users/creden	

Windows Event Logs

pip install python-evtx

```
import Evtx.Evtx as evtx

xml = "<Events>"
with evtx.Evtx("window_event_log.evtx") as log:
    for record in log.records():
        xml += record.xml()
xml += "</Events>"
```

Nested Data?



Nested Data?

```
{"time": 1084443427.311224,
"timestamp": "2004-05-13T10:17:07.311224",
"IP": {
  "version": 4,
  "ttl": 128,
  "proto": 6,
  "options": [],
  "len": 48,
  "dst": "65.208.228.223",
  "frag": 0,
  "flags": 2, "src": "145.254.160.237",
  "chksum": 37355
  "Ethernet": { "src": "00:00:01:00:00", "type": 2048, "dst": "fe:ff:20:00:01:00"},
```

Nested Data

```
pd.read_json( 'nested_data.json')
```

pd.read_json('nested_data.json')

	DNS	Ethernet	IP	TCP	UDP	time	timestamp
0	NaN	{'type': 2048, 'dst': 'fe:ff:20:00:01:00', 'sr	{'dst': '65.208.228.223', 'len': 48, 'version'	('window': 8760, 'chksum': 49932, 'sport': 337	NaN	1.084443e+09	2004-05-13 10:17:07.311224
1	NaN	{'type': 2048, 'dst': '00:00:01:00:00:00', 'sr	{'dst': '145.254.160.237', 'len': 48, 'version	('window': 5840, 'chksum': 23516, 'sport': 80,	NaN	1.084443e+09	2004-05-13 10:17:08.222534
2	NaN	{'type': 2048, 'dst': 'fe:ff:20:00:01:00', 'sr	{'dst': '65.208.228.223', 'len': 40, 'version'	('window': 9660, 'chksum': 31076, 'sport': 337	NaN	1.084443e+09	2004-05-13 10:17:08.222534
3	NaN	{'type': 2048, 'dst': 'fe:ff:20:00:01:00', 'sr	{'dst': '65.208.228.223', 'len': 519, 'version	('window': 9660, 'chksum': 43352, 'sport': 337	NaN	1.084443e+09	2004-05-13 10:17:08.222534
4	NaN	{'type': 2048, 'dst': '00:00:01:00:00:00', 'sr	{'dst': '145.254.160.237', 'len': 40, 'version	('window': 6432, 'chksum': 33825, 'sport': 80,	NaN	1.084443e+09	2004-05-13 10:17:08.783340

Nested Data

```
from pandas.io.json import json normalize
import json
import pandas as pd
with open ('nested.json') as data file:
    pcap data = json.load(data file)
df = pd.DataFrame( json_normalize(pcap_data) )
```

df = pd.DataFrame(json_normalize(pcap_data))

•••	TCP.seq	TCP.sport	TCP.urgptr	TCP.window	ι
	951057939.0	3372.0	0.0	8760.0	
	290218379.0	80.0	0.0	5840.0	
	951057940.0	3372.0	0.0	9660.0	
	951057940.0	3372.0	0.0	9660.0	
	290218380.0	80.0	0.0	6432.0	

ElasticSearch & Splunk





ElasticSearch & Splunk

pip install huntlib

```
e = ElasticDF(
    url="https://localhost:9200",
    ssl=True,
    username="myuser",
    password="mypass"
)

df = e.search_df(
    lucene="proto:tcp AND port:80",
    index="myindex-*",
    days=1,
    normalize=False
)
```

```
s = SplunkDF(
  host="<splunk_ip>",
  username="myuser",
  password="mypass"
)
df = s.search_df(
  spl='search index=win_events src="10.9.*.*"',
  start_time=datetime.now()-timedelta(days=2),
  end_time=datetime.now()
)
```

Manipulating a DataFrame

series = df['column1']

Returns a **series**

df['ip'].value_counts().head()

Two Ways of Accessing Columns



```
series = df['column1']
```

series = df.column1

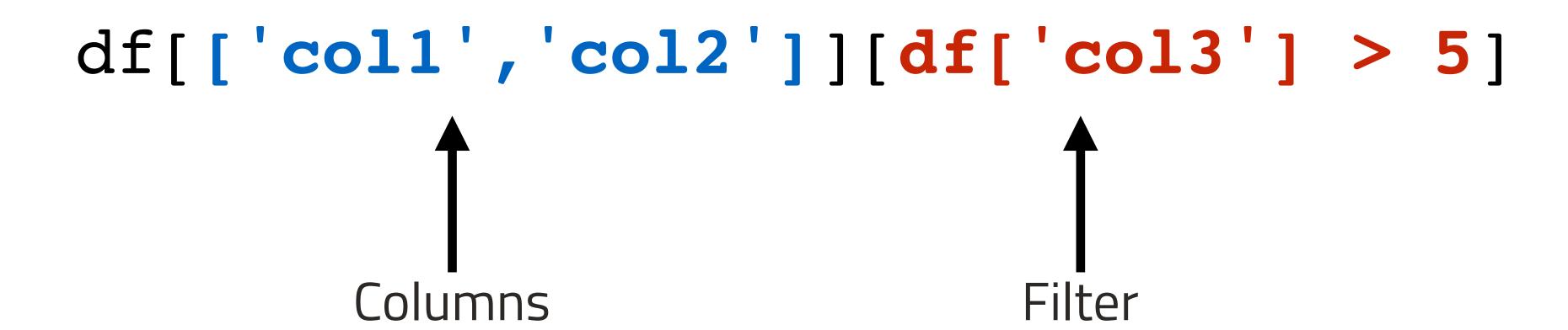
Don't use the dots!

df = df[['column1','column2','column3']]

Returns a DataFrame

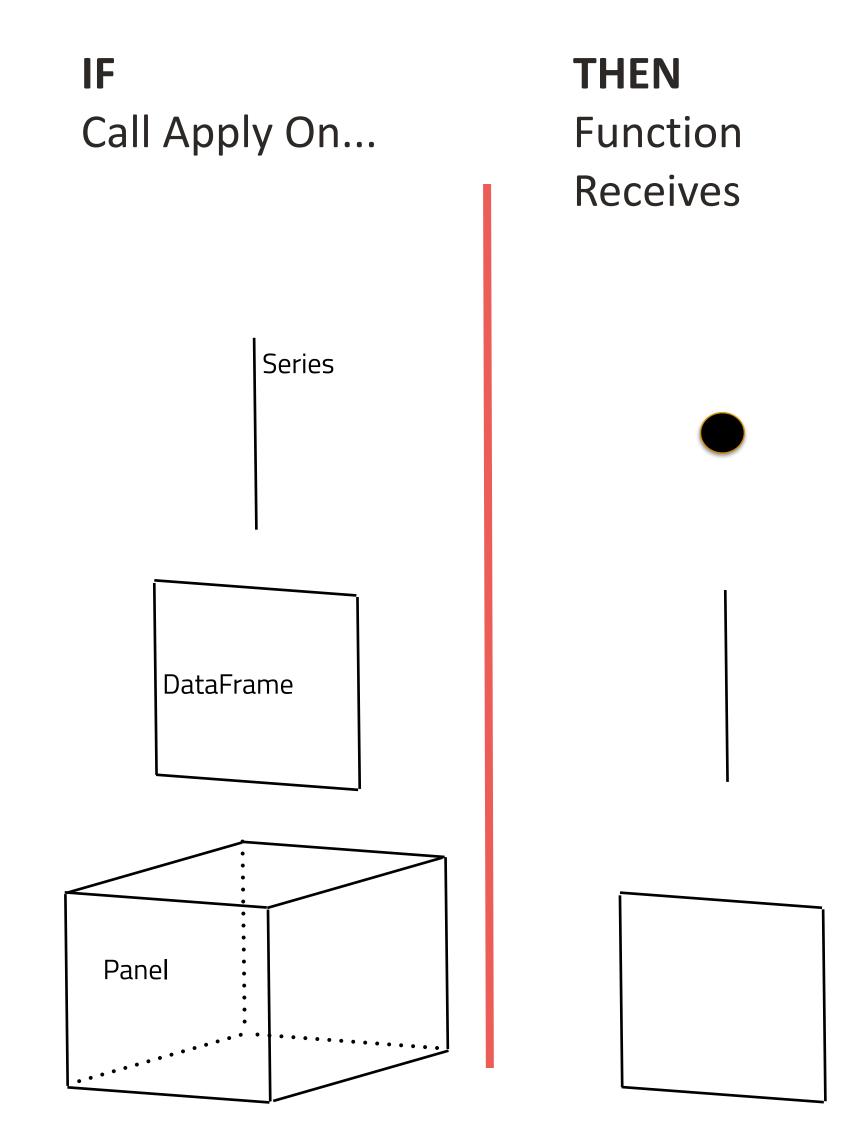
Filtering a DataFrame

```
df[<boolean condition>]
```

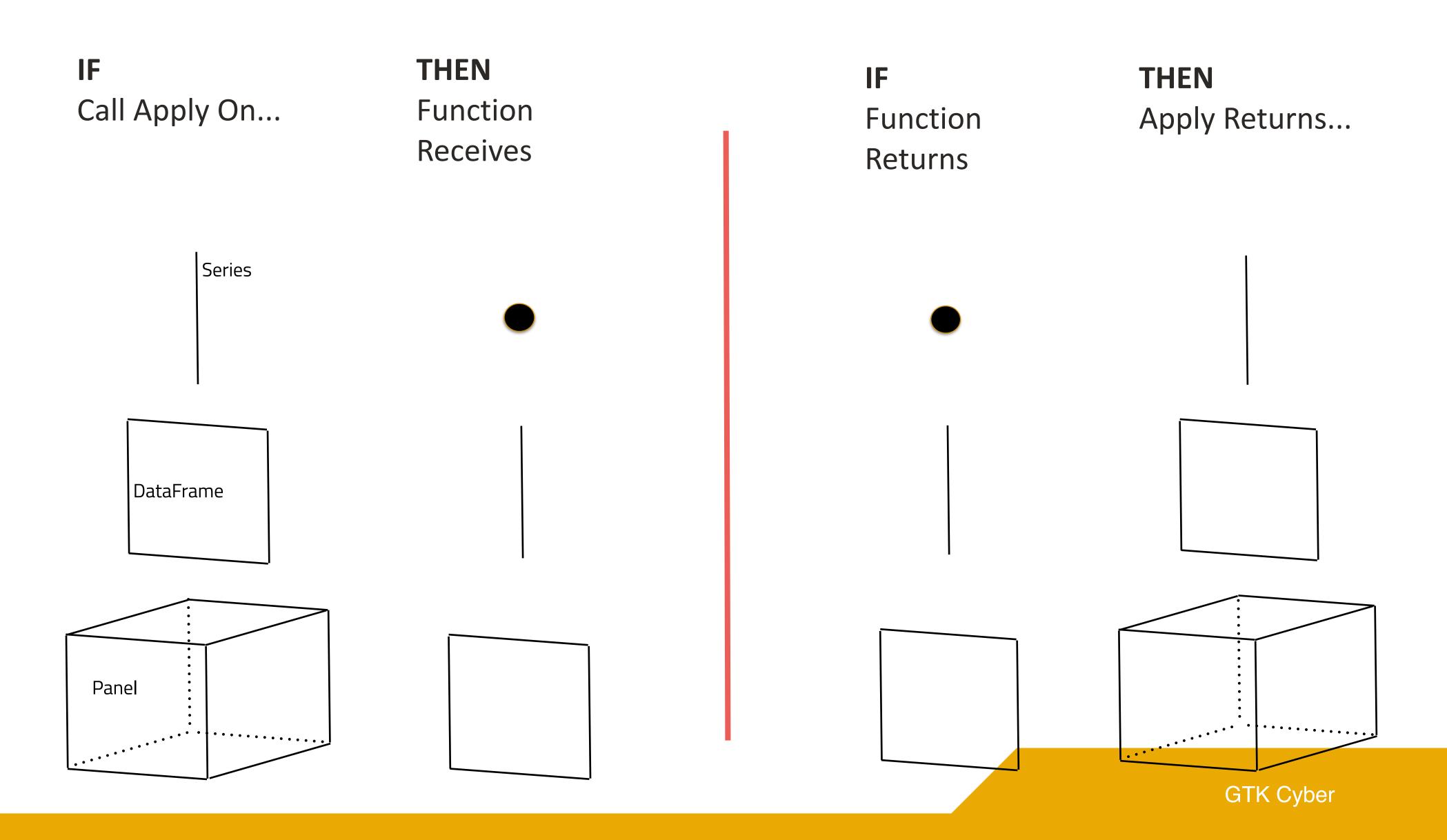


```
data.loc[<index>]
data.loc[<list of indexes>]
  data.sample(<n>)
```

data.apply(<function>)



data.apply(<function>)



Dataframe Demo

```
#Data1
{"first_name":{"0":"Robert","1":"Steve","2":"Anne","3":"Alice"},
"last_name":{"0":"Hernandez","1":"Smith","2":"Raps","3":"Muller"},
"birthday":{"0":"5\\/3\\/67","1":"8\\/4\\/84","2":"9\\/13\\/91","3":"4\\/15\\/75"}
}
```

df1 = pd.read json('../data/data1.json')

```
#Data2
{"0":{"first name":"Robert","last name":"Hernandez","birthday":"5\\/3\
\/67"},
"1":{"first name":"Steve","last name":"Smith","birthday":"8\\/4\\/
84"},
"2":{"first name":"Anne","last name":"Raps","birthday":"9\\/13\\/91"},
"3":{"first name":"Alice","last name":"Muller","birthday":"4\\/15\\/
75"}}
df2 = pd.read json('../data/data2.json', orient='index')
```

```
#Data3
{"first name": "Robert", "last name": "Hernandez", "birthday": "5\\/3\\/
67"},
{"first name": "Steve", "last name": "Smith", "birthday": "8\\/4\\/84"},
{"first name": "Anne", "last name": "Raps", "birthday": "9\\/13\\/91"},
{"first name": "Alice", "last name": "Muller", "birthday": "4\\/15\\/75"}]
 df3 = pd.read json('../data/data3.json',
 orient='columns')
```

```
#Data4
{"columns":["first_name","last_name","birthday"],
"index":[0,1,2,3],
"data":[["Robert","Hernandez","5\\/3\\/67"],["Steve","Smith","8\\/4\\/84"],["Anne","Raps","9\\/13\\/91"],["Alice","Muller","4\\/15\\/75"]]}
df4 = pd.read json('../data/data4.json', orient='split')
```

```
#Write the functions
def get os(x):
    user agent = parse(x)
    return user agent.os.family
def get browser(x):
    user agent = parse(x)
    return user agent.browser.family
```

```
#Apply the functions to the dataframe
server_df['os'] =
server_df['request_header_user_agent'].apply( get_os )
server_df['browser'] =
server_df['request_header_user_agent'].apply( get_browser )
```

```
#Apply the functions to the dataframe
server df['os'] =
server df['request header user agent'].apply( get_os )
server df['browser'] =
server df['request header user agent'].apply( get_browser )
#Get the top 10 values
server df['os'].value counts().head(10)
```

```
#Get the top 10 values server_df['os'].value_counts().head(10)
```

Windows 7	2041
Windows Vista	500
Windows XP	423
Windows 8.1	221
Linux	125
Mac OS X	80
Chrome OS	60
Ubuntu	6

GTK Cyber

```
bots = pd.read_csv('../data/dailybots.csv')
gov_bots = bots[['botfam', 'hosts']][bots['industry'] ==
"Government/Politics"]
gov_bots.groupby('botfam', as_index=False).sum()
```

Questions?

1	2	3
4	5	6
7	8	9

data.T

1	4	7
2	5	8
3	6	9

```
#Gets you the sum of columns
data.sum( axis=0 )

#Gets you the sum of the rows
data.sum( axis=1 )
```

```
DataFrame.drop(labels,
axis=0,
level=None,
inplace=False,
errors='raise')¶
```

Merging Data Sets

Series 1

Index	Value	
0	6	
1	4	
2	2	
3	3	

Series 1

Index	Value	
0	6	
1	4	
2	2	
3	3	

Series 2

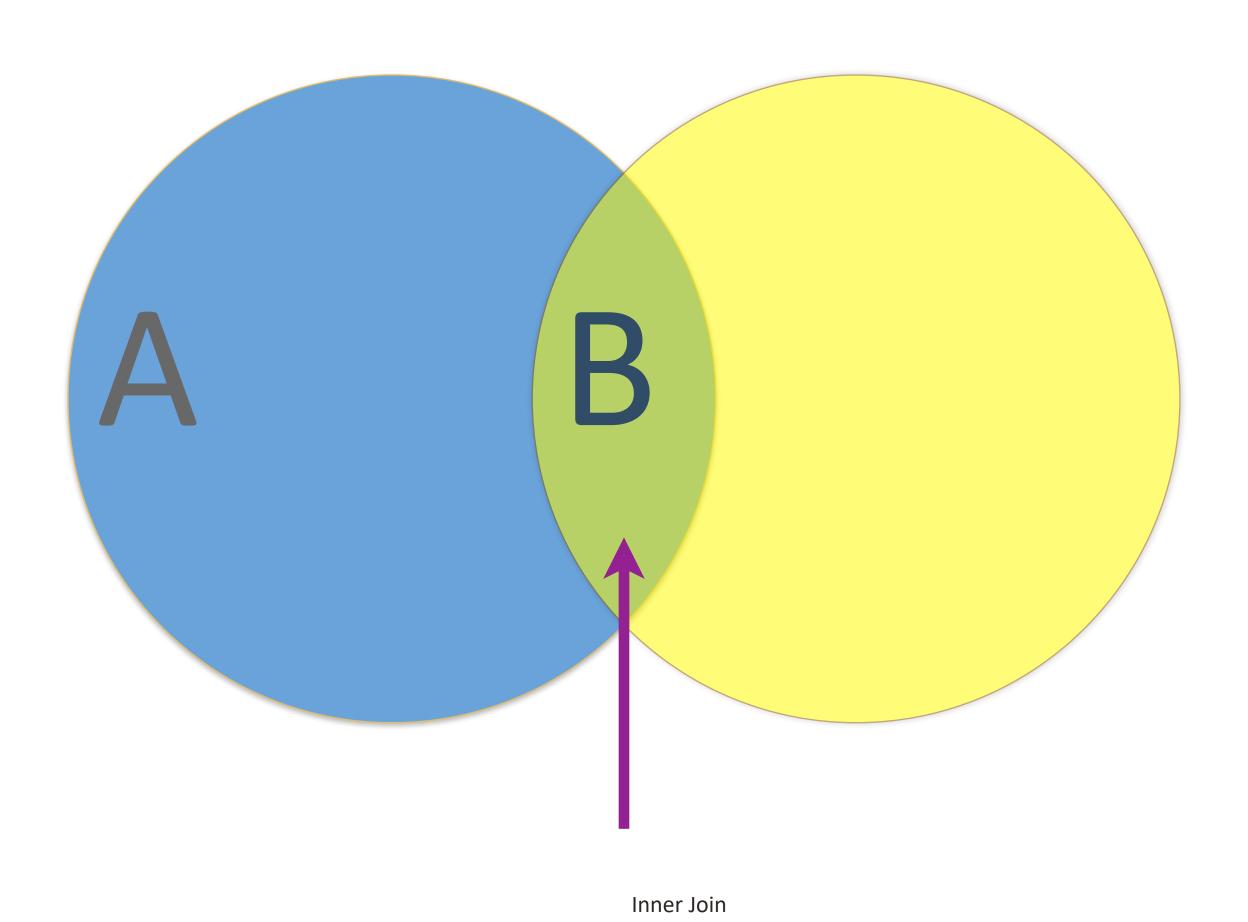
Index	Value	
0	7	
1	5	
2	3	
3	4	

combinedSeries

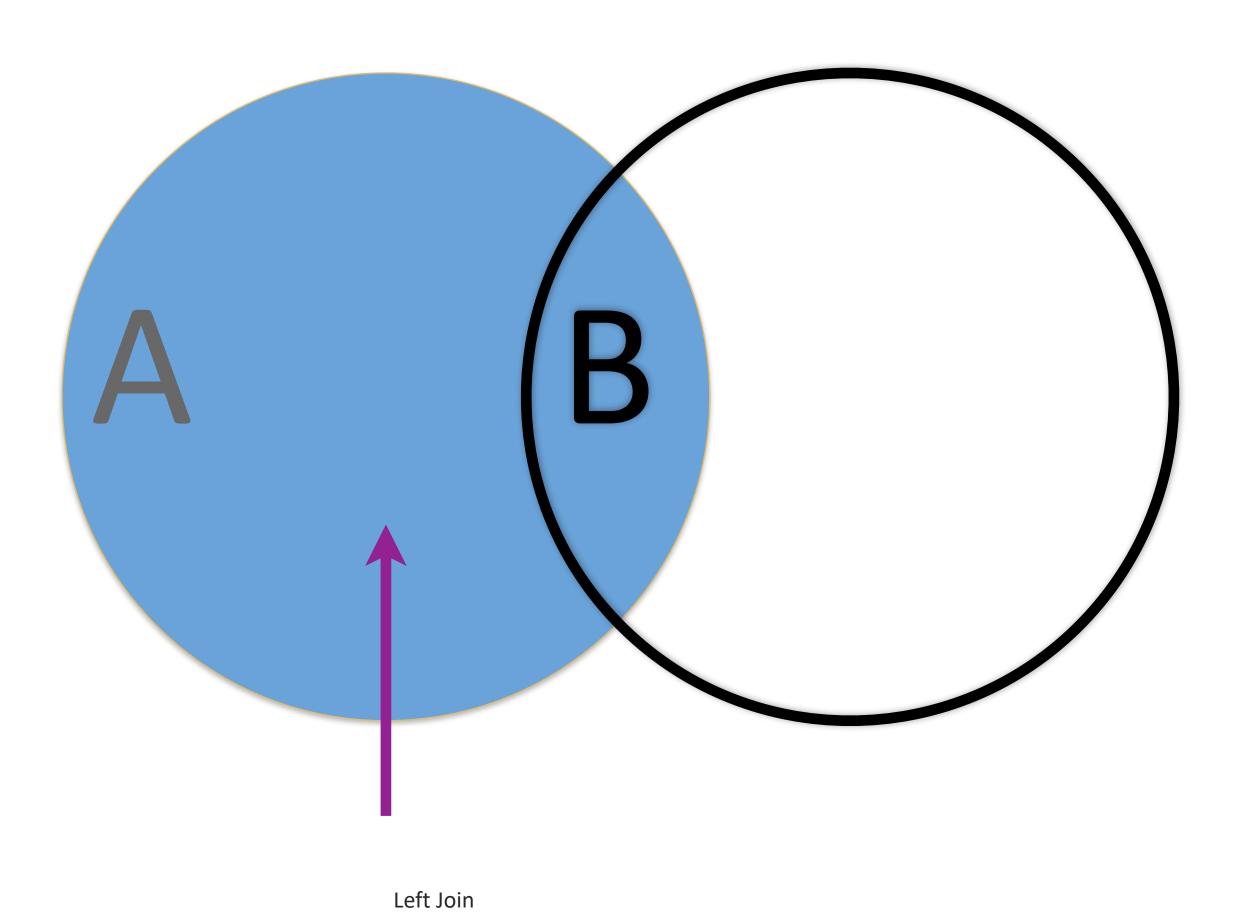
Index	Value
0	6
1	4
2	2
3	3
4	7
5	5
6	3
7	4

Joins

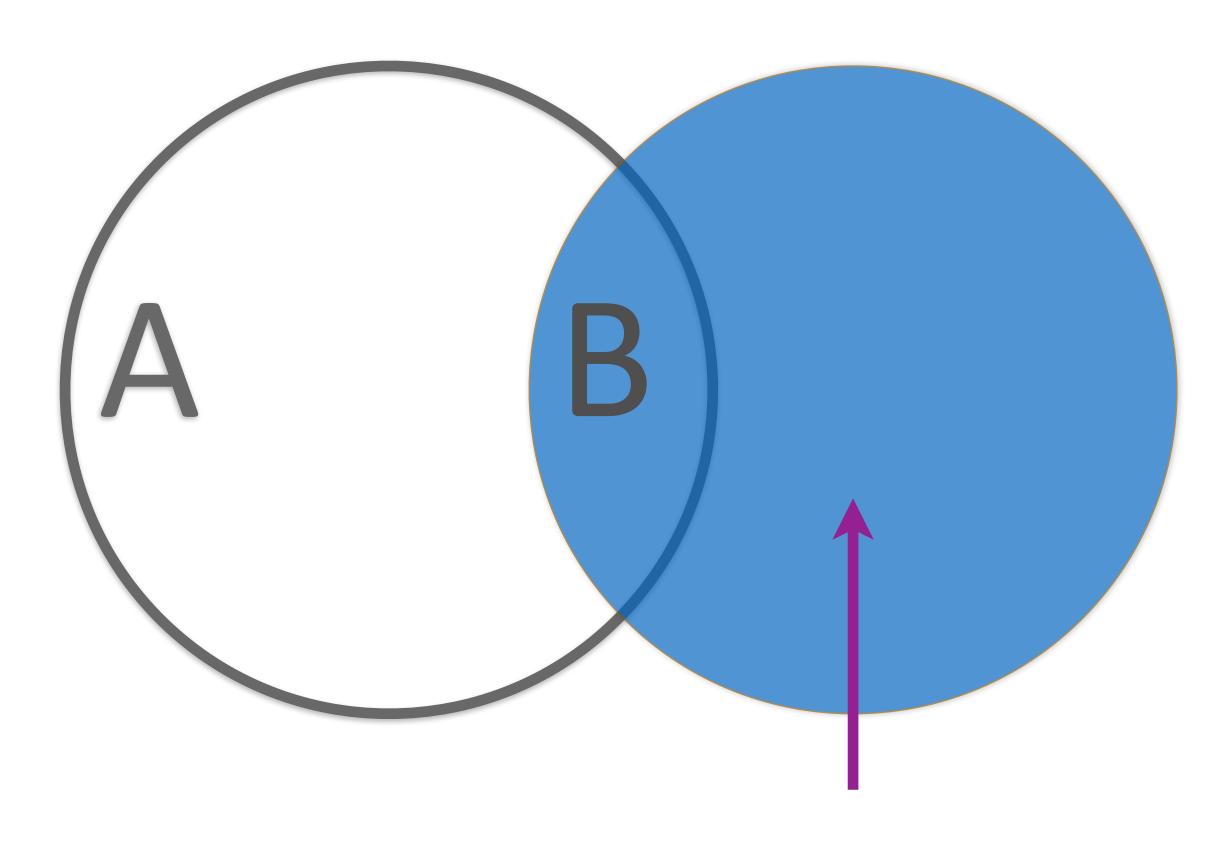
Inner Join (Intersection)



Left Join



Right Join



```
pd.merge(leftData, rightData)
```

pd.merge(leftData, rightData, how="<join type>")

Option	SQL Equivalent	Description
inner	INNER JOIN	Intersection
left	LEFT OUTER JOIN	Returns items in Set A, but not in Set B
right	RIGHT OUTER JOIN	Returns items in Set B, but not in Set A
outer	FULL OUTER JOIN	Returns the union of both sets

```
pd.merge( leftData, rightData,
    how="<join type>",
    on=<field list> )
```

date	src_ip	dst_ip	port
2018-06-21	192.168.20.2	10.10.4.1	80
2018-06-21	192.168.20.1	10.10.4.2	443
2018-06-21	192.168.20.2	10.10.5.1	80
2018-06-22	192.168.20.2	10.10.4.1	80

df.groupby(field or list of fields)

date	src_ip	dst_ip	port
2018-06-21	192.168.20.2	10.10.4.1	80
2018-06-21	192.168.20.1	10.10.4.2	443
2018-06-21	192.168.20.2	10.10.5.1	80
2018-06-22	192.168.20.2	10.10.4.1	80

df.groupby('src_ip')['port'].count()

src_ip	count
192.168.20.1	1
192.168.20.2	3

date	src_ip	dst_ip	port
2018-06-21	192.168.20.2	10.10.4.1	80
2018-06-21	192.168.20.1	10.10.4.2	443
2018-06-21	192.168.20.2	10.10.5.1	80
2018-06-22	192.168.20.2	10.10.4.1	80

df.groupby(['date','src_ip'])['port'].count()

Multi-Index

```
date src_ip
2018-06-21 192.168.20.1 1
192.168.20.2 2
2018-06-22 192.168.20.2 1
```

date	src_ip	dst_ip	port
2018-06-21	192.168.20.2	10.10.4.1	80
2018-06-21	192.168.20.1	10.10.4.2	443
2018-06-21	192.168.20.2	10.10.5.1	80
2018-06-22	192.168.20.2	10.10.4.1	80

```
df.groupby(['date','src_ip'], as_index=False)
['port'].count()
```

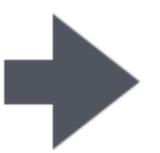
	date	src_ip	port
0	2018-06-21	192.168.20.1	1
1	2018-06-21	192.168.20.2	2
2	2018-06-22	192.168.20.2	1

Pivot

df

	foo	bar	baz	Z00
0	one	Α	1	Х
1	one	В	2	у
2	one	С	3	Z
3	two	Α	4	q
4	two	В	5	W

two



bar	A	В	С
foo			
one	1	2	3
two	4	5	6

columns='bar',

values='baz')

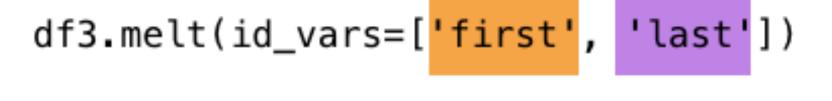
df.pivot(index='foo',

6

Melt

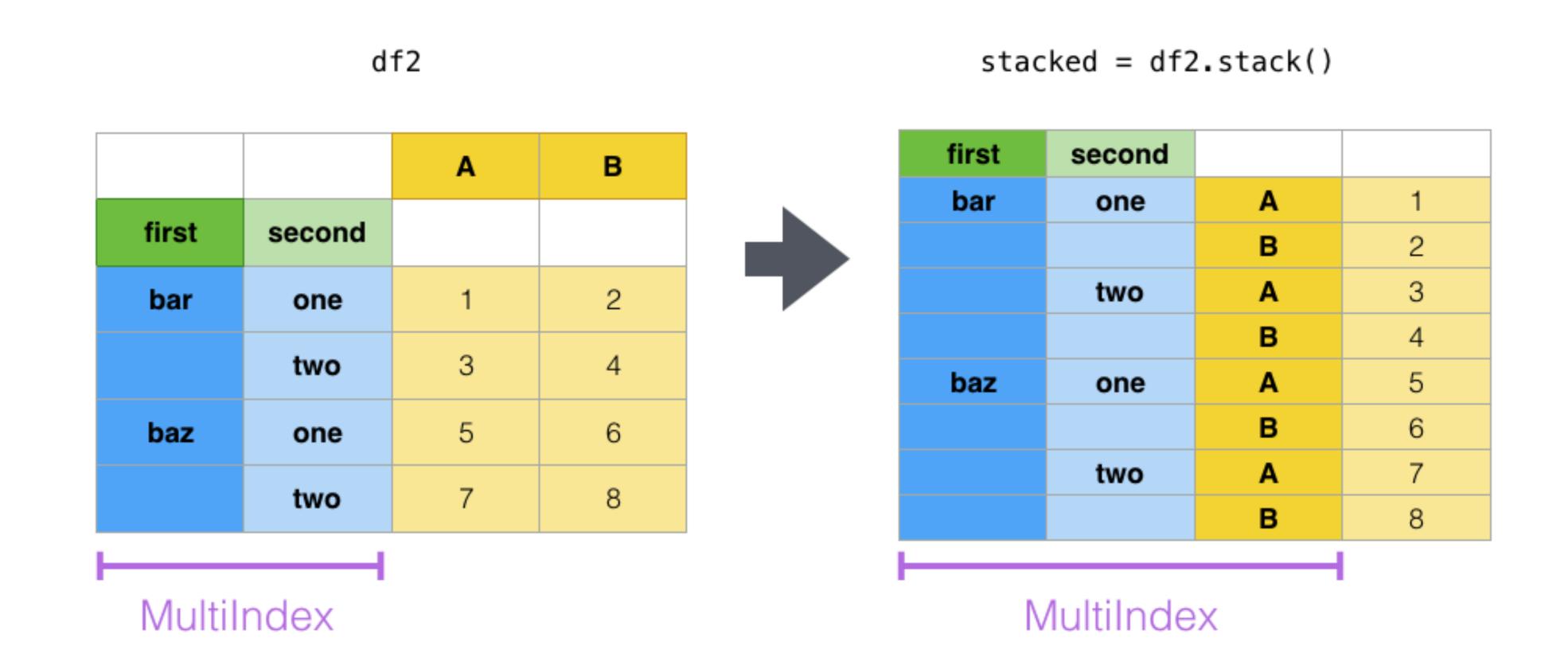
df3

	first	last	height	weight
0	John	Doe	5.5	130
1	Mary	Во	6.0	150

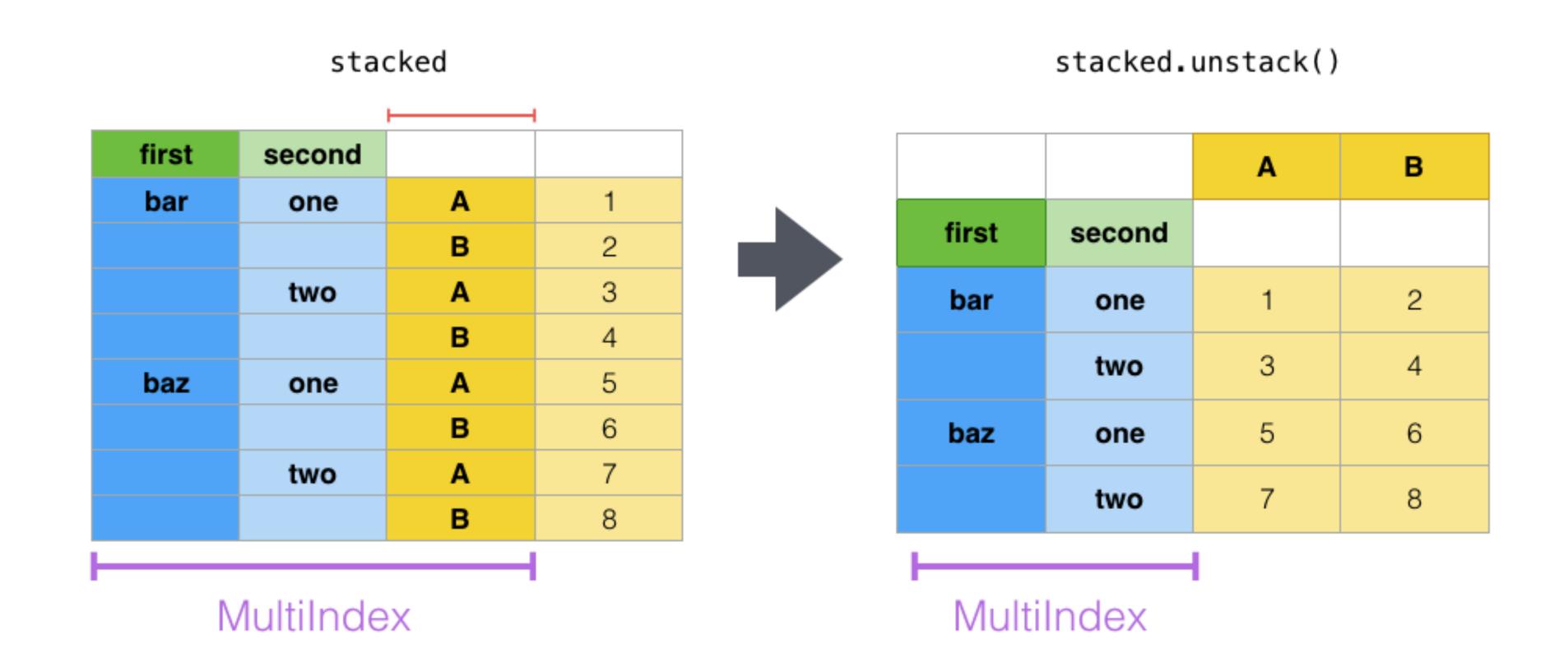


	first	last	variable	value
0	John	Doe	height	5.5
1	Mary	Во	height	6.0
2	John	Doe	weight	130
3	Mary	Во	weight	150

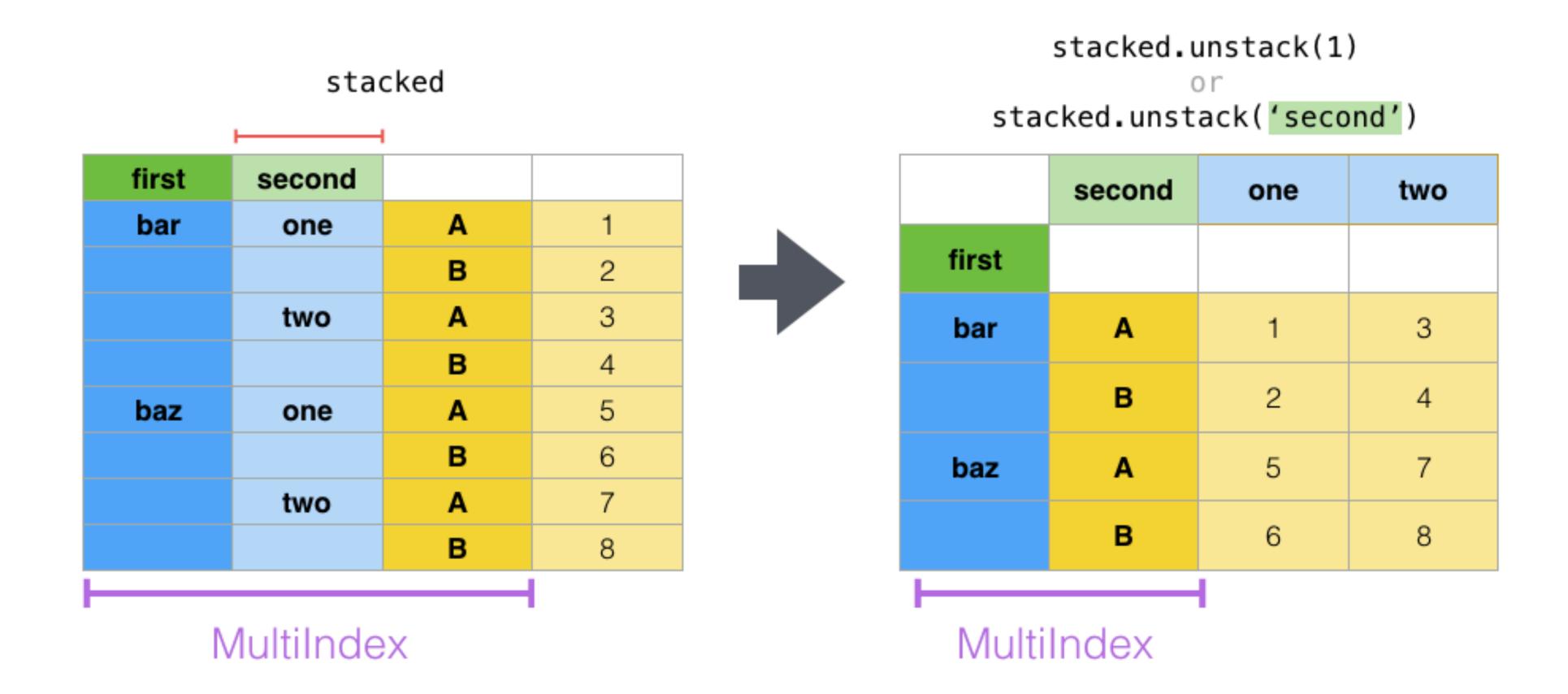
Stack



Unstack



Unstack(1)



```
Series.dropna()
Series.fillna(value="<something>")
```

Group Exercise

In your groups, for your use case, discuss:

- 1. What data you will need?
- 2. How will you gather it?
- 3. What form is it in?
- 4. What will you have to do in order to transform it?