

02B-Pandas

September 11, 2017

1 Getting to know your data with Pandas

1.1 Pandas

Pandas is the Python Data Analysis Library.

Pandas is an extremely versatile tool for manipulating datasets.

It also produces high quality plots with matplotlib, and integrates nicely with other libraries that expect NumPy arrays.

The most important tool provided by Pandas is the **data frame**.

A data frame is a table in which each row and column is given a label.

Pandas DataFrames are documented at:

<http://pandas.pydata.org/pandas-docs/dev/generated/pandas.DataFrame.html>

1.2 Getting started

1.3 Fetching, storing and retrieving your data

For demonstration purposes, we'll use a library built-in to Pandas that fetches data from standard online sources, such as Yahoo! Finance.

More information on what types of data you can fetch is at:
http://pandas.pydata.org/pandas-docs/stable/remote_data.html

```
In [202]: stocks = 'YELP'
          data_source = 'yahoo'
          start = datetime(2015,1,1)
          end = datetime(2015,12,31)

          yahoo_stocks = web.DataReader(stocks, data_source, start, end)

          yahoo_stocks.head()
          yahoo_stocks.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 252 entries, 2015-01-02 to 2015-12-31
Data columns (total 6 columns):
Open                252 non-null float64
High                252 non-null float64
Low                 252 non-null float64
```

```
Close          252 non-null float64
Adj Close      252 non-null float64
Volume         252 non-null int64
dtypes: float64(5), int64(1)
memory usage: 13.8 KB
```

1.3.1 Reading data from a .csv file

```
In [203]: yahoo_stocks.to_csv('yahoo_data.csv')
          print(open('yahoo_data.csv').read())
```

```
Date,Open,High,Low,Close,Adj Close,Volume
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2015-10-20,22.58,23.120001000000002,22.129998999999998,22.52,22.52,1663800
2015-10-21,22.52,22.75,22.190001000000002,22.389999,22.389999,1340800
2015-10-22,22.51,23.0,22.07,22.52,22.52,2341200
2015-10-23,22.799999,22.889999,22.110001,22.559998999999998,22.559998999999998,3188700
2015-10-26,22.6,24.799999,22.32,24.43,24.43,6890700
2015-10-27,24.299999,24.299999,22.01,22.9,22.9,6616200
2015-10-28,21.639999,22.809998999999998,21.43,22.07,22.07,9403700
2015-10-29,23.26,24.200001,22.41,22.950001,22.950001,9740500
2015-10-30,23.110001,23.15,22.0,22.25,22.25,5010000
2015-11-02,22.26,23.85,22.209999,23.799999,23.799999,6459600
2015-11-03,23.67,24.450001,23.6,24.120001000000002,24.120001000000002,2482100
2015-11-04,24.030001000000002,24.6,23.879998999999998,24.42,24.42,1704600
2015-11-05,24.5,25.559998999999998,24.32,25.049999,25.049999,2794300
2015-11-06,25.049999,25.530001000000002,24.5,25.5,25.5,2095400
2015-11-09,25.34,25.65,24.629998999999998,24.959999,24.959999,1567200
2015-11-10,24.85,25.360001,24.59,25.16,25.16,1267900
2015-11-11,25.1,25.309998999999998,24.540001,24.98,24.98,1364400
2015-11-12,25.18,27.219998999999998,24.9,25.9,25.9,5213300
2015-11-13,26.719998999999998,27.49,26.120001000000002,27.1,27.1,4976000
2015-11-16,27.0,27.59,26.469998999999998,27.440001000000002,27.440001000000002,3066700
2015-11-17,27.34,27.610001,26.860001,27.540001,27.540001,2018000
2015-11-18,27.540001,28.83,27.309998999999998,28.23,28.23,3091600
2015-11-19,28.190001000000002,28.690001000000002,27.91,28.059998999999998,28.059998999999998,148
2015-11-20,28.1,31.25,28.049999,31.209999,31.209999,6697500
2015-11-23,30.58,30.809998999999998,29.15,29.860001,29.860001,4029900
2015-11-24,29.459999,30.629998999999998,29.450001,30.01,30.01,2584500

```

2015-11-25,29.790001,30.540001,29.709999,30.51,30.51,1287100
2015-11-27,30.5,30.6,29.610001,30.18,30.18,1058900
2015-11-30,30.110001,30.719998999999998,29.77,30.129998999999998,30.129998999999998,2015600
2015-12-01,30.110001,30.459999,29.799999,30.309998999999998,30.309998999999998,1886000
2015-12-02,30.299999,32.470001,30.290001,31.389999,31.389999,4650300
2015-12-03,31.389999,32.240002000000004,30.48,30.629998999999998,30.629998999999998,2698900
2015-12-04,30.530001000000002,30.860001,29.32,30.450001,30.450001,2313800
2015-12-07,30.379998999999998,30.639999,29.629998999999998,30.040001,30.040001,1362300
2015-12-08,29.809998999999998,31.379998999999998,29.5,30.92,30.92,1830200
2015-12-09,30.98,31.139999,29.26,30.0,30.0,2238500
2015-12-10,30.110001,31.299999,29.99,30.83,30.83,1251800
2015-12-11,30.690001000000002,30.75,29.6,29.65,29.65,1415000
2015-12-14,29.6,29.889999,28.85,29.58,29.58,2328600
2015-12-15,29.68,30.0,26.459999,26.870001000000002,26.870001000000002,5759200
2015-12-16,26.889999,28.24,26.26,28.030001000000002,28.030001000000002,2992100
2015-12-17,28.139999,28.32,27.190001000000002,27.42,27.42,1483900
2015-12-18,27.309998999999998,27.91,26.9,27.17,27.17,1299800
2015-12-21,27.17,27.360001,26.030001000000002,26.25,26.25,1947600
2015-12-22,26.25,28.700001,26.15,27.93,27.93,2952700
2015-12-23,27.950001,28.42,27.440001000000002,28.15,28.15,1001000
2015-12-24,28.27,28.59,27.9,28.4,28.4,587400
2015-12-28,28.120001000000002,28.379998999999998,27.77,27.879998999999998,27.879998999999998,100
2015-12-29,27.950001,28.540001,27.74,28.48,28.48,1103900
2015-12-30,28.58,28.780001000000002,28.17,28.25,28.25,1068000
2015-12-31,28.1,28.969998999999998,28.02,28.799999,28.799999,1295600

```

```

In [204]: df = pd.read_csv('yahoo_data.csv')
df

```

```

Out[204]:
   Date      Open      High      Low      Close  Adj Close  \
0  2015-01-02  55.459999  55.599998  54.240002  55.150002  55.150002
1  2015-01-05  54.540001  54.950001  52.330002  52.529999  52.529999
2  2015-01-06  52.549999  53.930000  50.750000  52.439999  52.439999
3  2015-01-07  53.320000  53.750000  51.759998  52.209999  52.209999
4  2015-01-08  52.590000  54.139999  51.759998  53.830002  53.830002
5  2015-01-09  55.959999  56.990002  54.720001  56.070000  56.070000
6  2015-01-12  56.000000  56.060001  53.430000  54.020000  54.020000
7  2015-01-13  54.470001  54.799999  52.520000  53.180000  53.180000
8  2015-01-14  52.799999  53.680000  51.459999  52.200001  52.200001
9  2015-01-15  53.000000  53.610001  50.029999  50.119999  50.119999
10 2015-01-16  50.180000  51.490002  50.029999  51.389999  51.389999
11 2015-01-20  51.650002  51.779999  50.689999  51.410000  51.410000
12 2015-01-21  51.200001  53.500000  51.200001  53.410000  53.410000
13 2015-01-22  53.869999  55.279999  53.119999  54.799999  54.799999
14 2015-01-23  54.660000  55.639999  54.299999  55.189999  55.189999
15 2015-01-26  55.119999  55.790001  54.830002  55.410000  55.410000

```

16	2015-01-27	56.060001	56.160000	54.570000	55.630001	55.630001
17	2015-01-28	56.150002	56.150002	52.919998	53.000000	53.000000
18	2015-01-29	52.849998	53.310001	51.410000	52.930000	52.930000
19	2015-01-30	52.590000	53.419998	52.049999	52.470001	52.470001
20	2015-02-02	52.939999	53.500000	51.209999	53.470001	53.470001
21	2015-02-03	53.830002	55.930000	53.410000	55.779999	55.779999
22	2015-02-04	55.529999	57.070000	55.250000	56.740002	56.740002
23	2015-02-05	57.599998	57.700001	56.080002	57.470001	57.470001
24	2015-02-06	47.700001	48.169998	44.860001	45.110001	45.110001
25	2015-02-09	44.910000	45.040001	42.099998	42.169998	42.169998
26	2015-02-10	43.830002	45.549999	43.310001	44.660000	44.660000
27	2015-02-11	45.389999	46.430000	44.810001	46.180000	46.180000
28	2015-02-12	46.450001	47.840000	45.950001	47.630001	47.630001
29	2015-02-13	48.509998	49.049999	47.220001	47.529999	47.529999
..
222	2015-11-18	27.540001	28.830000	27.309999	28.230000	28.230000
223	2015-11-19	28.190001	28.690001	27.910000	28.059999	28.059999
224	2015-11-20	28.100000	31.250000	28.049999	31.209999	31.209999
225	2015-11-23	30.580000	30.809999	29.150000	29.860001	29.860001
226	2015-11-24	29.459999	30.629999	29.450001	30.010000	30.010000
227	2015-11-25	29.790001	30.540001	29.709999	30.510000	30.510000
228	2015-11-27	30.500000	30.600000	29.610001	30.180000	30.180000
229	2015-11-30	30.110001	30.719999	29.770000	30.129999	30.129999
230	2015-12-01	30.110001	30.459999	29.799999	30.309999	30.309999
231	2015-12-02	30.299999	32.470001	30.290001	31.389999	31.389999
232	2015-12-03	31.389999	32.240002	30.480000	30.629999	30.629999
233	2015-12-04	30.530001	30.860001	29.320000	30.450001	30.450001
234	2015-12-07	30.379999	30.639999	29.629999	30.040001	30.040001
235	2015-12-08	29.809999	31.379999	29.500000	30.920000	30.920000
236	2015-12-09	30.980000	31.139999	29.260000	30.000000	30.000000
237	2015-12-10	30.110001	31.299999	29.990000	30.830000	30.830000
238	2015-12-11	30.690001	30.750000	29.600000	29.650000	29.650000
239	2015-12-14	29.600000	29.889999	28.850000	29.580000	29.580000
240	2015-12-15	29.680000	30.000000	26.459999	26.870001	26.870001
241	2015-12-16	26.889999	28.240000	26.260000	28.030001	28.030001
242	2015-12-17	28.139999	28.320000	27.190001	27.420000	27.420000
243	2015-12-18	27.309999	27.910000	26.900000	27.170000	27.170000
244	2015-12-21	27.170000	27.360001	26.030001	26.250000	26.250000
245	2015-12-22	26.250000	28.700001	26.150000	27.930000	27.930000
246	2015-12-23	27.950001	28.420000	27.440001	28.150000	28.150000
247	2015-12-24	28.270000	28.590000	27.900000	28.400000	28.400000
248	2015-12-28	28.120001	28.379999	27.770000	27.879999	27.879999
249	2015-12-29	27.950001	28.540001	27.740000	28.480000	28.480000
250	2015-12-30	28.580000	28.780001	28.170000	28.250000	28.250000
251	2015-12-31	28.100000	28.969999	28.020000	28.799999	28.799999

Volume
0 1664500

1	2023000
2	3762800
3	1548200
4	2015300
5	6222600
6	2405100
7	1952100
8	1854600
9	2647800
10	2183300
11	1227600
12	3248100
13	2295400
14	1629000
15	1450300
16	2410400
17	2013100
18	1844100
19	1875400
20	2105500
21	2876400
22	2498600
23	4657300
24	25137400
25	13079300
26	11267700
27	6359400
28	4375000
29	4713100
..	...
222	3091600
223	1487500
224	6697500
225	4029900
226	2584500
227	1287100
228	1058900
229	2015600
230	1886000
231	4650300
232	2698900
233	2313800
234	1362300
235	1830200
236	2238500
237	1251800
238	1415000
239	2328600

```

240    5759200
241    2992100
242    1483900
243    1299800
244    1947600
245    2952700
246    1001000
247     587400
248    1004500
249    1103900
250    1068000
251    1295600

```

```
[252 rows x 7 columns]
```

The number of rows in the DataFrame:

```
In [205]: len(df)
```

```
Out[205]: 252
```

1.4 Working with data columns

The columns or "features" in your data

```
In [206]: df.columns
```

```
Out[206]: Index(['Date', 'Open', 'High', 'Low', 'Close', 'Adj Close', 'Volume'], dtype='object')
```

Selecting a single column from your data

```
In [207]: df['Open']
```

```

Out[207]: 0      55.459999
          1      54.540001
          2      52.549999
          3      53.320000
          4      52.590000
          5      55.959999
          6      56.000000
          7      54.470001
          8      52.799999
          9      53.000000
         10      50.180000
         11      51.650002
         12      51.200001
         13      53.869999
         14      54.660000
         15      55.119999

```

16	56.060001
17	56.150002
18	52.849998
19	52.590000
20	52.939999
21	53.830002
22	55.529999
23	57.599998
24	47.700001
25	44.910000
26	43.830002
27	45.389999
28	46.450001
29	48.509998
...	
222	27.540001
223	28.190001
224	28.100000
225	30.580000
226	29.459999
227	29.790001
228	30.500000
229	30.110001
230	30.110001
231	30.299999
232	31.389999
233	30.530001
234	30.379999
235	29.809999
236	30.980000
237	30.110001
238	30.690001
239	29.600000
240	29.680000
241	26.889999
242	28.139999
243	27.309999
244	27.170000
245	26.250000
246	27.950001
247	28.270000
248	28.120001
249	27.950001
250	28.580000
251	28.100000

Name: Open, Length: 252, dtype: float64

Another way of selecting a single column from your data

```
In [208]: df.Open
```

```
Out[208]: 0      55.459999
          1      54.540001
          2      52.549999
          3      53.320000
          4      52.590000
          5      55.959999
          6      56.000000
          7      54.470001
          8      52.799999
          9      53.000000
         10      50.180000
         11      51.650002
         12      51.200001
         13      53.869999
         14      54.660000
         15      55.119999
         16      56.060001
         17      56.150002
         18      52.849998
         19      52.590000
         20      52.939999
         21      53.830002
         22      55.529999
         23      57.599998
         24      47.700001
         25      44.910000
         26      43.830002
         27      45.389999
         28      46.450001
         29      48.509998
          ...
        222      27.540001
        223      28.190001
        224      28.100000
        225      30.580000
        226      29.459999
        227      29.790001
        228      30.500000
        229      30.110001
        230      30.110001
        231      30.299999
        232      31.389999
        233      30.530001
        234      30.379999
        235      29.809999
        236      30.980000
```

```

237    30.110001
238    30.690001
239    29.600000
240    29.680000
241    26.889999
242    28.139999
243    27.309999
244    27.170000
245    26.250000
246    27.950001
247    28.270000
248    28.120001
249    27.950001
250    28.580000
251    28.100000
Name: Open, Length: 252, dtype: float64

```

```
In [209]: df[['Open', 'Close']].head()
```

```

Out[209]:
      Open    Close
0  55.459999  55.150002
1  54.540001  52.529999
2  52.549999  52.439999
3  53.320000  52.209999
4  52.590000  53.830002

```

```
In [210]: df.Date.head(10)
```

```

Out[210]:
0    2015-01-02
1    2015-01-05
2    2015-01-06
3    2015-01-07
4    2015-01-08
5    2015-01-09
6    2015-01-12
7    2015-01-13
8    2015-01-14
9    2015-01-15
Name: Date, dtype: object

```

```
In [211]: df.Date.tail(10)
```

```

Out[211]:
242    2015-12-17
243    2015-12-18
244    2015-12-21
245    2015-12-22
246    2015-12-23
247    2015-12-24
248    2015-12-28

```

```
249    2015-12-29
250    2015-12-30
251    2015-12-31
Name: Date, dtype: object
```

Changing the column names:

```
In [212]: new_column_names = [x.lower().replace(' ', '_') for x in df.columns]
df.columns = new_column_names
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 252 entries, 0 to 251
Data columns (total 7 columns):
date           252 non-null object
open           252 non-null float64
high           252 non-null float64
low            252 non-null float64
close          252 non-null float64
adj_close      252 non-null float64
volume         252 non-null int64
dtypes: float64(5), int64(1), object(1)
memory usage: 13.9+ KB
```

Now **all** columns can be accessed using the **dot** notation:

```
In [213]: df.adj_close.head()

Out[213]: 0    55.150002
          1    52.529999
          2    52.439999
          3    52.209999
          4    53.830002
          Name: adj_close, dtype: float64
```

1.5 Data Frame methods

A DataFrame object has many useful methods.

```
In [214]: df.mean()

Out[214]: open           3.728766e+01
          high           3.805464e+01
          low            3.656373e+01
          close          3.729917e+01
          adj_close      3.729917e+01
          volume         3.491051e+06
          dtype: float64
```

```
In [215]: df.std()
```

```
Out[215]: open          1.128093e+01  
         high          1.138111e+01  
         low           1.113097e+01  
         close         1.125233e+01  
         adj_close     1.125233e+01  
         volume        4.145546e+06  
         dtype: float64
```

```
In [216]: df.median()
```

```
Out[216]: open          3.796500e+01  
         high          3.871500e+01  
         low           3.637500e+01  
         close         3.783500e+01  
         adj_close     3.783500e+01  
         volume        2.354050e+06  
         dtype: float64
```

```
In [217]: df.open.mean()
```

```
Out[217]: 37.28765869841269
```

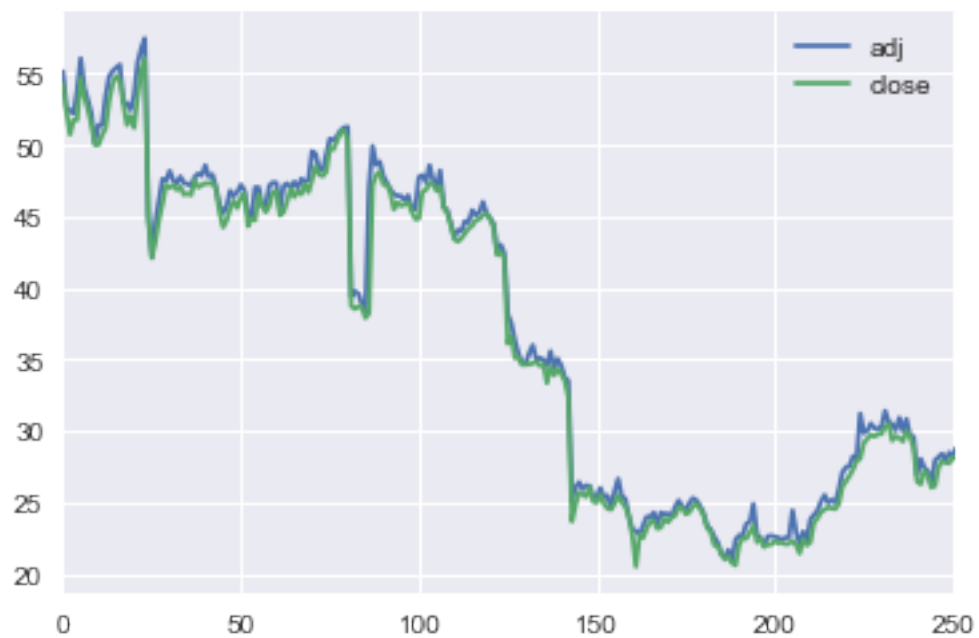
```
In [218]: df.high.mean()
```

```
Out[218]: 38.05464295238094
```

1.5.1 Plotting methods

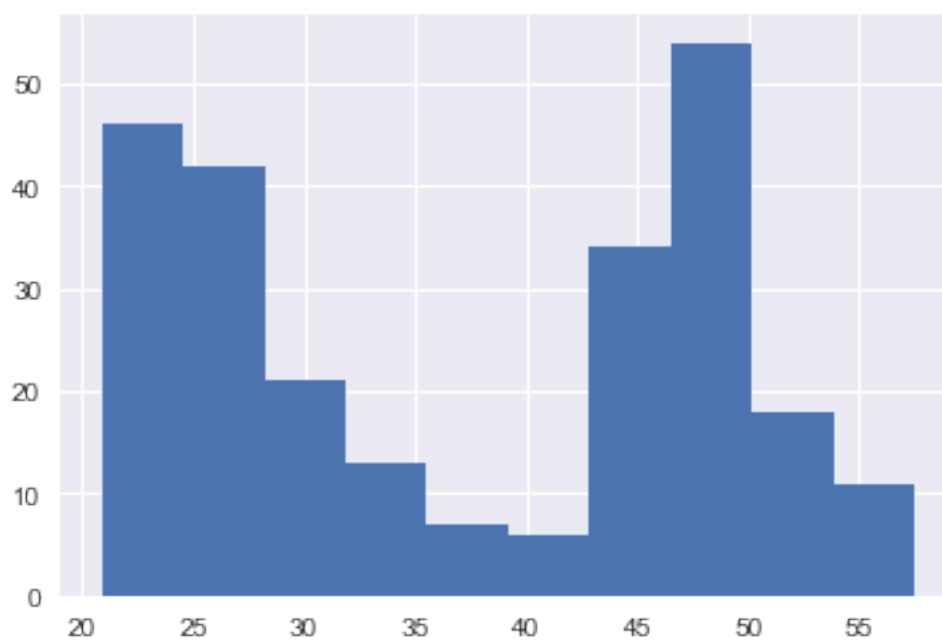
```
In [219]: df.adj_close.plot(label='adj')  
         df.low.plot(label='close')  
         plt.legend(loc='best')
```

```
Out[219]: <matplotlib.legend.Legend at 0x1156e9e80>
```



```
In [220]: df.adj_close.hist()
```

```
Out[220]: <matplotlib.axes._subplots.AxesSubplot at 0x115979cf8>
```



1.5.2 Bulk Operations

Methods like `sum()` and `std()` work on entire columns.

We can run our own functions across all values in a column (or row) using `apply()`.

```
In [221]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 252 entries, 0 to 251
Data columns (total 7 columns):
date                252 non-null object
open                252 non-null float64
high                252 non-null float64
low                 252 non-null float64
close               252 non-null float64
adj_close           252 non-null float64
volume              252 non-null int64
dtypes: float64(5), int64(1), object(1)
memory usage: 13.9+ KB
```

```
In [222]: df.date.head()
```

```
Out[222]: 0    2015-01-02
          1    2015-01-05
          2    2015-01-06
          3    2015-01-07
          4    2015-01-08
          Name: date, dtype: object
```

The **values** property of the column returns a list of values for the column. Inspecting the first value reveals that these are strings with a particular format.

```
In [223]: first_date = df.date.values[0]
          first_date
```

```
Out[223]: '2015-01-02'
```

```
In [224]: datetime.strptime(first_date, "%Y-%m-%d")
```

```
Out[224]: datetime.datetime(2015, 1, 2, 0, 0)
```

```
In [225]: df.date = df.date.apply(lambda d: datetime.strptime(d, "%Y-%m-%d"))
          df.date.head()
```

```
Out[225]: 0    2015-01-02
          1    2015-01-05
          2    2015-01-06
          3    2015-01-07
          4    2015-01-08
          Name: date, dtype: datetime64[ns]
```

Each row in a DataFrame is associated with an index, which is a label that uniquely identifies a row.

The row indices so far have been auto-generated by pandas, and are simply integers starting from 0.

From now on we will use dates instead of integers for indices -- the benefits of this will show later.

Overwriting the index is as easy as assigning to the index property of the DataFrame.

```
In [226]: df.index = df.date
          df.head()
```

```
Out[226]:
```

	date	open	high	low	close	adj_close	\
	date						
	2015-01-02	2015-01-02	55.459999	55.599998	54.240002	55.150002	55.150002
	2015-01-05	2015-01-05	54.540001	54.950001	52.330002	52.529999	52.529999
	2015-01-06	2015-01-06	52.549999	53.930000	50.750000	52.439999	52.439999
	2015-01-07	2015-01-07	53.320000	53.750000	51.759998	52.209999	52.209999
	2015-01-08	2015-01-08	52.590000	54.139999	51.759998	53.830002	53.830002

	volume	
	date	
	2015-01-02	1664500
	2015-01-05	2023000
	2015-01-06	3762800
	2015-01-07	1548200
	2015-01-08	2015300

Now that we have made an index based on date, we can drop the original date column.

```
In [227]: df = df.drop(['date'],axis=1)
          df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 252 entries, 2015-01-02 to 2015-12-31
Data columns (total 6 columns):
open          252 non-null float64
high          252 non-null float64
low           252 non-null float64
close         252 non-null float64
adj_close     252 non-null float64
volume        252 non-null int64
dtypes: float64(5), int64(1)
memory usage: 13.8 KB
```

1.5.3 Accessing rows of the DataFrame

So far we've seen how to access a column of the DataFrame. To access a row we use a different notation.

To access a row by its index value, use the `.loc()` method.

```
In [228]: df.loc[datetime(2015,1,23,0,0)]
```

```
Out[228]: open          5.466000e+01
           high          5.564000e+01
           low           5.430000e+01
           close         5.519000e+01
           adj_close     5.519000e+01
           volume        1.629000e+06
           Name: 2015-01-23 00:00:00, dtype: float64
```

To access a row by its sequence number (ie, like an array index), use `.iloc()` ('Integer Location')

```
In [229]: df.iloc[0,:]
```

```
Out[229]: open          5.546000e+01
           high          5.560000e+01
           low           5.424000e+01
           close         5.515000e+01
           adj_close     5.515000e+01
           volume        1.664500e+06
           Name: 2015-01-02 00:00:00, dtype: float64
```

To iterate over the rows, use `.iterrows()`

```
In [230]: num_positive_days = 0
           for idx, row in df.iterrows():
               if row.close > row.open:
                   num_positive_days += 1

           print("The total number of positive-gain days is {}".format(num_positive_days))
```

The total number of positive-gain days is 126.

1.6 Filtering

It is very easy to select interesting rows from the data.

All these operations below return a new DataFrame, which itself can be treated the same way as all DataFrames we have seen so far.

```
In [231]: tmp_high = df.high > 55
           tmp_high.head()
```

```
Out[231]: date
           2015-01-02    True
           2015-01-05   False
           2015-01-06   False
           2015-01-07   False
           2015-01-08   False
           Name: high, dtype: bool
```

Summing a Boolean array is the same as counting the number of True values.

```
In [232]: sum(tmp_high)
```

```
Out[232]: 11
```

Now, let's select only the rows of df1 that correspond to tmp_high

```
In [233]: df[tmp_high]
```

```
Out[233]:
```

	open	high	low	close	adj_close	volume
date						
2015-01-02	55.459999	55.599998	54.240002	55.150002	55.150002	1664500
2015-01-09	55.959999	56.990002	54.720001	56.070000	56.070000	6222600
2015-01-12	56.000000	56.060001	53.430000	54.020000	54.020000	2405100
2015-01-22	53.869999	55.279999	53.119999	54.799999	54.799999	2295400
2015-01-23	54.660000	55.639999	54.299999	55.189999	55.189999	1629000
2015-01-26	55.119999	55.790001	54.830002	55.410000	55.410000	1450300
2015-01-27	56.060001	56.160000	54.570000	55.630001	55.630001	2410400
2015-01-28	56.150002	56.150002	52.919998	53.000000	53.000000	2013100
2015-02-03	53.830002	55.930000	53.410000	55.779999	55.779999	2876400
2015-02-04	55.529999	57.070000	55.250000	56.740002	56.740002	2498600
2015-02-05	57.599998	57.700001	56.080002	57.470001	57.470001	4657300

Putting it all together, we have the following commonly-used patterns:

```
In [234]: positive_days = df[df.close > df.open]
positive_days.head()
```

```
Out[234]:
```

	open	high	low	close	adj_close	volume
date						
2015-01-08	52.590000	54.139999	51.759998	53.830002	53.830002	2015300
2015-01-09	55.959999	56.990002	54.720001	56.070000	56.070000	6222600
2015-01-16	50.180000	51.490002	50.029999	51.389999	51.389999	2183300
2015-01-21	51.200001	53.500000	51.200001	53.410000	53.410000	3248100
2015-01-22	53.869999	55.279999	53.119999	54.799999	54.799999	2295400

```
In [235]: very_positive_days = df[df.close-df.open > 4]
very_positive_days.head()
```

```
Out[235]:
```

	open	high	low	close	adj_close	volume
date						
2015-05-07	38.220001	48.73	38.220001	47.009998	47.009998	33831600

1.7 Creating new columns

To create a new column, simply assign values to it. Think of the columns as a dictionary:

```
In [236]: df['profit'] = (df.open < df.close)
df.head()
```

```
Out[236]:
```

	open	high	low	close	adj_close	volume	\
date							
2015-01-02	55.459999	55.599998	54.240002	55.150002	55.150002	1664500	
2015-01-05	54.540001	54.950001	52.330002	52.529999	52.529999	2023000	
2015-01-06	52.549999	53.930000	50.750000	52.439999	52.439999	3762800	
2015-01-07	53.320000	53.750000	51.759998	52.209999	52.209999	1548200	
2015-01-08	52.590000	54.139999	51.759998	53.830002	53.830002	2015300	

	profit
date	
2015-01-02	False
2015-01-05	False
2015-01-06	False
2015-01-07	False
2015-01-08	True

```
In [237]: for idx, row in df.iterrows():
            if row.close > row.open:
                df.loc[idx, 'gain'] = 'negative'
            elif (row.open - row.close) < 1:
                df.loc[idx, 'gain'] = 'small_gain'
            elif (row.open - row.close) < 6:
                df.loc[idx, 'gain'] = 'medium_gain'
            else:
                df.loc[idx, 'gain'] = 'large_gain'
df.head()
```

```
Out[237]:
```

	open	high	low	close	adj_close	volume	\
date							
2015-01-02	55.459999	55.599998	54.240002	55.150002	55.150002	1664500	
2015-01-05	54.540001	54.950001	52.330002	52.529999	52.529999	2023000	
2015-01-06	52.549999	53.930000	50.750000	52.439999	52.439999	3762800	
2015-01-07	53.320000	53.750000	51.759998	52.209999	52.209999	1548200	
2015-01-08	52.590000	54.139999	51.759998	53.830002	53.830002	2015300	

	profit	gain
date		
2015-01-02	False	small_gain
2015-01-05	False	medium_gain
2015-01-06	False	small_gain
2015-01-07	False	medium_gain
2015-01-08	True	negative

Here is another, more "functional", way to accomplish the same thing.
 Define a function that classifies rows, and apply it to each row.

```
In [238]: def namerow(row):
            if row.close > row.open:
                return 'negative'
```

```

elif (row.open - row.close) < 1:
    return 'small_gain'
elif (row.open - row.close) < 6:
    return 'medium_gain'
else:
    return 'large_gain'

```

```
df['test_column'] = df.apply(namerow, axis = 1)
```

In [239]: df.head()

```

Out[239]:
           open      high      low      close  adj_close  volume  \
date
2015-01-02  55.459999  55.599998  54.240002  55.150002  55.150002  1664500
2015-01-05  54.540001  54.950001  52.330002  52.529999  52.529999  2023000
2015-01-06  52.549999  53.930000  50.750000  52.439999  52.439999  3762800
2015-01-07  53.320000  53.750000  51.759998  52.209999  52.209999  1548200
2015-01-08  52.590000  54.139999  51.759998  53.830002  53.830002  2015300

           profit      gain  test_column
date
2015-01-02   False  small_gain  small_gain
2015-01-05   False  medium_gain  medium_gain
2015-01-06   False  small_gain  small_gain
2015-01-07   False  medium_gain  medium_gain
2015-01-08    True   negative   negative

```

OK, point made, let's get rid of that extraneous test_column:

In [240]: df.drop('test_column', axis = 1)

```

Out[240]:
           open      high      low      close  adj_close  volume  \
date
2015-01-02  55.459999  55.599998  54.240002  55.150002  55.150002  1664500
2015-01-05  54.540001  54.950001  52.330002  52.529999  52.529999  2023000
2015-01-06  52.549999  53.930000  50.750000  52.439999  52.439999  3762800
2015-01-07  53.320000  53.750000  51.759998  52.209999  52.209999  1548200
2015-01-08  52.590000  54.139999  51.759998  53.830002  53.830002  2015300
2015-01-09  55.959999  56.990002  54.720001  56.070000  56.070000  6222600
2015-01-12  56.000000  56.060001  53.430000  54.020000  54.020000  2405100
2015-01-13  54.470001  54.799999  52.520000  53.180000  53.180000  1952100
2015-01-14  52.799999  53.680000  51.459999  52.200001  52.200001  1854600
2015-01-15  53.000000  53.610001  50.029999  50.119999  50.119999  2647800
2015-01-16  50.180000  51.490002  50.029999  51.389999  51.389999  2183300
2015-01-20  51.650002  51.779999  50.689999  51.410000  51.410000  1227600
2015-01-21  51.200001  53.500000  51.200001  53.410000  53.410000  3248100
2015-01-22  53.869999  55.279999  53.119999  54.799999  54.799999  2295400
2015-01-23  54.660000  55.639999  54.299999  55.189999  55.189999  1629000
2015-01-26  55.119999  55.790001  54.830002  55.410000  55.410000  1450300

```

2015-01-27	56.060001	56.160000	54.570000	55.630001	55.630001	2410400
2015-01-28	56.150002	56.150002	52.919998	53.000000	53.000000	2013100
2015-01-29	52.849998	53.310001	51.410000	52.930000	52.930000	1844100
2015-01-30	52.590000	53.419998	52.049999	52.470001	52.470001	1875400
2015-02-02	52.939999	53.500000	51.209999	53.470001	53.470001	2105500
2015-02-03	53.830002	55.930000	53.410000	55.779999	55.779999	2876400
2015-02-04	55.529999	57.070000	55.250000	56.740002	56.740002	2498600
2015-02-05	57.599998	57.700001	56.080002	57.470001	57.470001	4657300
2015-02-06	47.700001	48.169998	44.860001	45.110001	45.110001	25137400
2015-02-09	44.910000	45.040001	42.099998	42.169998	42.169998	13079300
2015-02-10	43.830002	45.549999	43.310001	44.660000	44.660000	11267700
2015-02-11	45.389999	46.430000	44.810001	46.180000	46.180000	6359400
2015-02-12	46.450001	47.840000	45.950001	47.630001	47.630001	4375000
2015-02-13	48.509998	49.049999	47.220001	47.529999	47.529999	4713100
...
2015-11-18	27.540001	28.830000	27.309999	28.230000	28.230000	3091600
2015-11-19	28.190001	28.690001	27.910000	28.059999	28.059999	1487500
2015-11-20	28.100000	31.250000	28.049999	31.209999	31.209999	6697500
2015-11-23	30.580000	30.809999	29.150000	29.860001	29.860001	4029900
2015-11-24	29.459999	30.629999	29.450001	30.010000	30.010000	2584500
2015-11-25	29.790001	30.540001	29.709999	30.510000	30.510000	1287100
2015-11-27	30.500000	30.600000	29.610001	30.180000	30.180000	1058900
2015-11-30	30.110001	30.719999	29.770000	30.129999	30.129999	2015600
2015-12-01	30.110001	30.459999	29.799999	30.309999	30.309999	1886000
2015-12-02	30.299999	32.470001	30.290001	31.389999	31.389999	4650300
2015-12-03	31.389999	32.240002	30.480000	30.629999	30.629999	2698900
2015-12-04	30.530001	30.860001	29.320000	30.450001	30.450001	2313800
2015-12-07	30.379999	30.639999	29.629999	30.040001	30.040001	1362300
2015-12-08	29.809999	31.379999	29.500000	30.920000	30.920000	1830200
2015-12-09	30.980000	31.139999	29.260000	30.000000	30.000000	2238500
2015-12-10	30.110001	31.299999	29.990000	30.830000	30.830000	1251800
2015-12-11	30.690001	30.750000	29.600000	29.650000	29.650000	1415000
2015-12-14	29.600000	29.889999	28.850000	29.580000	29.580000	2328600
2015-12-15	29.680000	30.000000	26.459999	26.870001	26.870001	5759200
2015-12-16	26.889999	28.240000	26.260000	28.030001	28.030001	2992100
2015-12-17	28.139999	28.320000	27.190001	27.420000	27.420000	1483900
2015-12-18	27.309999	27.910000	26.900000	27.170000	27.170000	1299800
2015-12-21	27.170000	27.360001	26.030001	26.250000	26.250000	1947600
2015-12-22	26.250000	28.700001	26.150000	27.930000	27.930000	2952700
2015-12-23	27.950001	28.420000	27.440001	28.150000	28.150000	1001000
2015-12-24	28.270000	28.590000	27.900000	28.400000	28.400000	587400
2015-12-28	28.120001	28.379999	27.770000	27.879999	27.879999	1004500
2015-12-29	27.950001	28.540001	27.740000	28.480000	28.480000	1103900
2015-12-30	28.580000	28.780001	28.170000	28.250000	28.250000	1068000
2015-12-31	28.100000	28.969999	28.020000	28.799999	28.799999	1295600

	profit	gain
date		

2015-01-02	False	small_gain
2015-01-05	False	medium_gain
2015-01-06	False	small_gain
2015-01-07	False	medium_gain
2015-01-08	True	negative
2015-01-09	True	negative
2015-01-12	False	medium_gain
2015-01-13	False	medium_gain
2015-01-14	False	small_gain
2015-01-15	False	medium_gain
2015-01-16	True	negative
2015-01-20	False	small_gain
2015-01-21	True	negative
2015-01-22	True	negative
2015-01-23	True	negative
2015-01-26	True	negative
2015-01-27	False	small_gain
2015-01-28	False	medium_gain
2015-01-29	True	negative
2015-01-30	False	small_gain
2015-02-02	True	negative
2015-02-03	True	negative
2015-02-04	True	negative
2015-02-05	False	small_gain
2015-02-06	False	medium_gain
2015-02-09	False	medium_gain
2015-02-10	True	negative
2015-02-11	True	negative
2015-02-12	True	negative
2015-02-13	False	small_gain
...
2015-11-18	True	negative
2015-11-19	False	small_gain
2015-11-20	True	negative
2015-11-23	False	small_gain
2015-11-24	True	negative
2015-11-25	True	negative
2015-11-27	False	small_gain
2015-11-30	True	negative
2015-12-01	True	negative
2015-12-02	True	negative
2015-12-03	False	small_gain
2015-12-04	False	small_gain
2015-12-07	False	small_gain
2015-12-08	True	negative
2015-12-09	False	small_gain
2015-12-10	True	negative
2015-12-11	False	medium_gain

2015-12-14	False	small_gain
2015-12-15	False	medium_gain
2015-12-16	True	negative
2015-12-17	False	small_gain
2015-12-18	False	small_gain
2015-12-21	False	small_gain
2015-12-22	True	negative
2015-12-23	True	negative
2015-12-24	True	negative
2015-12-28	False	small_gain
2015-12-29	True	negative
2015-12-30	False	small_gain
2015-12-31	True	negative

[252 rows x 8 columns]

1.8 Grouping

An **extremely** powerful DataFrame method is `groupby()`.

This is entirely analogous to `GROUP BY` in SQL.

It will group the rows of a DataFrame by the values in one (or more) columns, and let you iterate through each group.

Here we will look at the average gain among the categories of gains (negative, small, medium and large) we defined above and stored in column `gain`.

```
In [241]: gain_groups = df.groupby('gain')
```

Essentially, `gain_groups` behaves like a dictionary * whose keys are the unique values found in the `gain` column, and * whose values are DataFrames that contain only the rows having the corresponding unique values.

```
In [242]: for gain, gain_data in gain_groups:
           print(gain)
           print(gain_data.head())
           print('=====')
```

```
medium_gain
      open      high      low      close  adj_close  volume  \
date
2015-01-05  54.540001  54.950001  52.330002  52.529999  52.529999  2023000
2015-01-07  53.320000  53.750000  51.759998  52.209999  52.209999  1548200
2015-01-12  56.000000  56.060001  53.430000  54.020000  54.020000  2405100
2015-01-13  54.470001  54.799999  52.520000  53.180000  53.180000  1952100
2015-01-15  53.000000  53.610001  50.029999  50.119999  50.119999  2647800

      profit      gain  test_column
date
2015-01-05  False  medium_gain  medium_gain
2015-01-07  False  medium_gain  medium_gain
```

```

2015-01-12    False  medium_gain  medium_gain
2015-01-13    False  medium_gain  medium_gain
2015-01-15    False  medium_gain  medium_gain

```

```
=====
```

```
negative
```

	open	high	low	close	adj_close	volume	\
date							
2015-01-08	52.590000	54.139999	51.759998	53.830002	53.830002	2015300	
2015-01-09	55.959999	56.990002	54.720001	56.070000	56.070000	6222600	
2015-01-16	50.180000	51.490002	50.029999	51.389999	51.389999	2183300	
2015-01-21	51.200001	53.500000	51.200001	53.410000	53.410000	3248100	
2015-01-22	53.869999	55.279999	53.119999	54.799999	54.799999	2295400	

	profit	gain	test_column
date			
2015-01-08	True	negative	negative
2015-01-09	True	negative	negative
2015-01-16	True	negative	negative
2015-01-21	True	negative	negative
2015-01-22	True	negative	negative

```
=====
```

```
small_gain
```

	open	high	low	close	adj_close	volume	\
date							
2015-01-02	55.459999	55.599998	54.240002	55.150002	55.150002	1664500	
2015-01-06	52.549999	53.930000	50.750000	52.439999	52.439999	3762800	
2015-01-14	52.799999	53.680000	51.459999	52.200001	52.200001	1854600	
2015-01-20	51.650002	51.779999	50.689999	51.410000	51.410000	1227600	
2015-01-27	56.060001	56.160000	54.570000	55.630001	55.630001	2410400	

	profit	gain	test_column
date			
2015-01-02	False	small_gain	small_gain
2015-01-06	False	small_gain	small_gain
2015-01-14	False	small_gain	small_gain
2015-01-20	False	small_gain	small_gain
2015-01-27	False	small_gain	small_gain

```
=====
```

```

In [243]: for gain, gain_data in df.groupby("gain"):
           print('The average closing value for the {} group is {}'.format(gain,
                                                                              gain_data.close.mean()))

```

```
The average closing value for the medium_gain group is 41.008888629629624
```

```
The average closing value for the negative group is 37.213571404761886
```

```
The average closing value for the small_gain group is 36.39636363636365
```

1.9 Other Pandas Classes

A DataFrame is essentially an annotated 2-D array.

Pandas also has annotated versions of 1-D and 3-D arrays.

A 1-D array in Pandas is called a Series.

A 3-D array in Pandas is called a Panel.

To use these, read the documentation!

1.10 Comparing multiple stocks

As a last task, we will use the experience we obtained so far -- and learn some new things -- in order to compare the performance of different stocks we obtained from Yahoo finance.

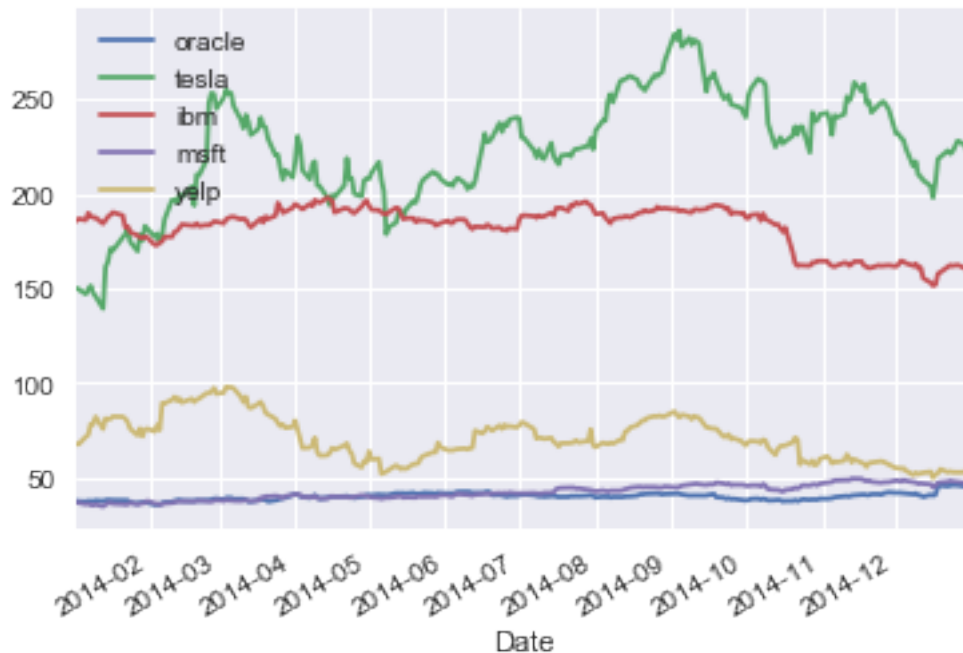
```
In [244]: stocks = ['ORCL', 'TSLA', 'IBM', 'YELP', 'MSFT']
          attr = 'Close'
          df = web.DataReader(stocks,
                              data_source,
                              start=datetime(2014, 1, 1),
                              end=datetime(2014, 12, 31))[attr]

          df.head()
```

```
Out [244]:
```

	IBM	MSFT	ORCL	TSLA	YELP
Date					
2014-12-31	160.440002	46.450001	44.970001	222.410004	54.730000
2014-12-30	160.050003	47.020000	45.340000	222.229996	54.240002
2014-12-29	160.509995	47.450001	45.610001	225.710007	53.009998
2014-12-26	162.339996	47.880001	46.099998	227.820007	52.939999
2014-12-24	161.820007	48.139999	46.230000	222.259995	53.000000

```
In [245]: df.ORCL.plot(label = 'oracle')
          df.TSLA.plot(label = 'tesla')
          df.IBM.plot(label = 'ibm')
          df.MSFT.plot(label = 'msft')
          df.YELP.plot(label = 'yelp')
          _ = plt.legend(loc='best')
```



Next, we will calculate returns over a period of length T , defined as:

$$r(t) = \frac{f(t) - f(t - T)}{f(t)}$$

The returns can be computed with a simple DataFrame method `pct_change()`. Note that for the first T timesteps, this value is not defined (of course):

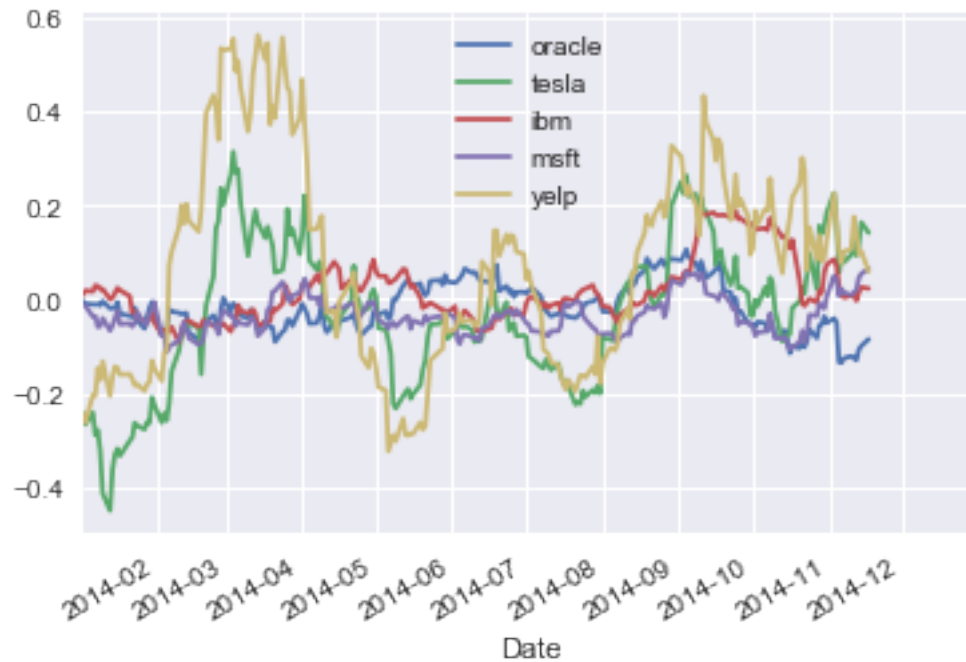
```
In [246]: rets = df.pct_change(30)
          rets.iloc[25:35]
```

```
Out [246]:
```

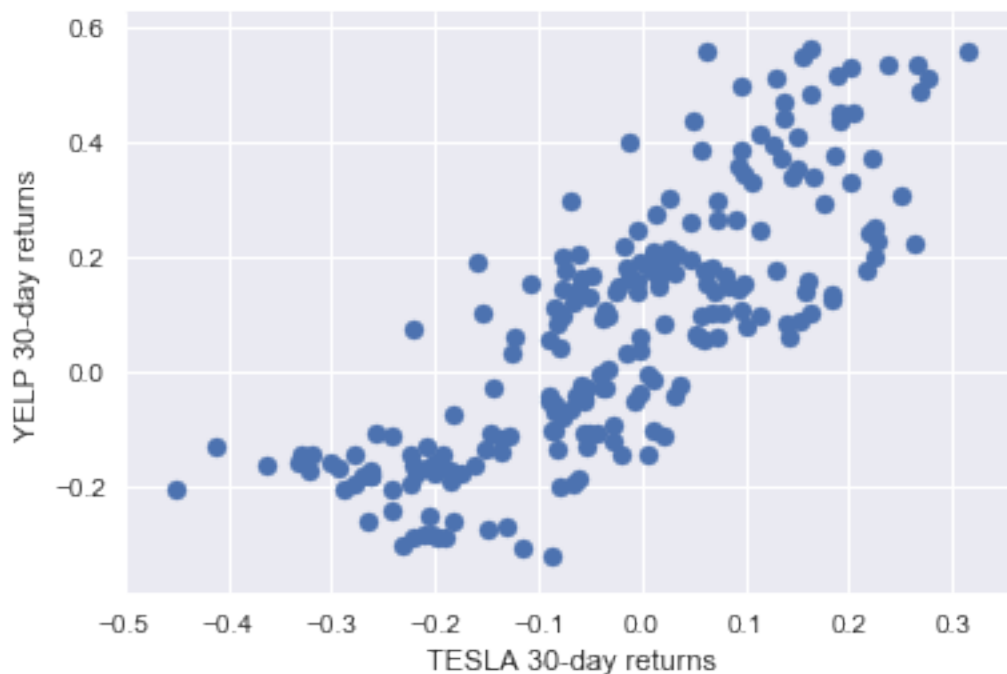
	IBM	MSFT	ORCL	TSLA	YELP
Date					
2014-11-24	NaN	NaN	NaN	NaN	NaN
2014-11-21	NaN	NaN	NaN	NaN	NaN
2014-11-20	NaN	NaN	NaN	NaN	NaN
2014-11-19	NaN	NaN	NaN	NaN	NaN
2014-11-18	NaN	NaN	NaN	NaN	NaN
2014-11-17	0.023186	0.064801	-0.084723	0.141945	0.059748
2014-11-14	0.025679	0.054445	-0.099250	0.164019	0.102876
2014-11-13	0.014205	0.045522	-0.107213	0.115148	0.096020
2014-11-12	-0.002587	0.018797	-0.129284	0.093407	0.144314
2014-11-11	0.009146	0.015164	-0.124594	0.129668	0.175660

Now we'll plot the timeseries of the returns of the different stocks. Notice that the NaN values are gracefully dropped by the plotting function.

```
In [247]: rets.ORCL.plot(label = 'oracle')
rets.TSLA.plot(label = 'tesla')
rets.IBM.plot(label = 'ibm')
rets.MSFT.plot(label = 'msft')
rets.YELP.plot(label = 'yelp')
_ = plt.legend(loc='best')
```



```
In [248]: plt.scatter(rets.TSLA, rets.YELP)
plt.xlabel('TESLA 30-day returns')
_ = plt.ylabel('YELP 30-day returns')
```



There appears to be some (fairly strong) correlation between the movement of TSLA and YELP stocks. Let's measure this.

The correlation coefficient between variables X and Y is defined as follows:

$$\text{Corr}(X, Y) = \frac{E[(X - \mu_X)(Y - \mu_Y)]}{\sigma_X \sigma_Y}$$

Pandas provides a DataFrame method to compute the correlation coefficient of all pairs of columns: `corr()`.

```
In [249]: rets.corr()
```

```
Out[249]:
```

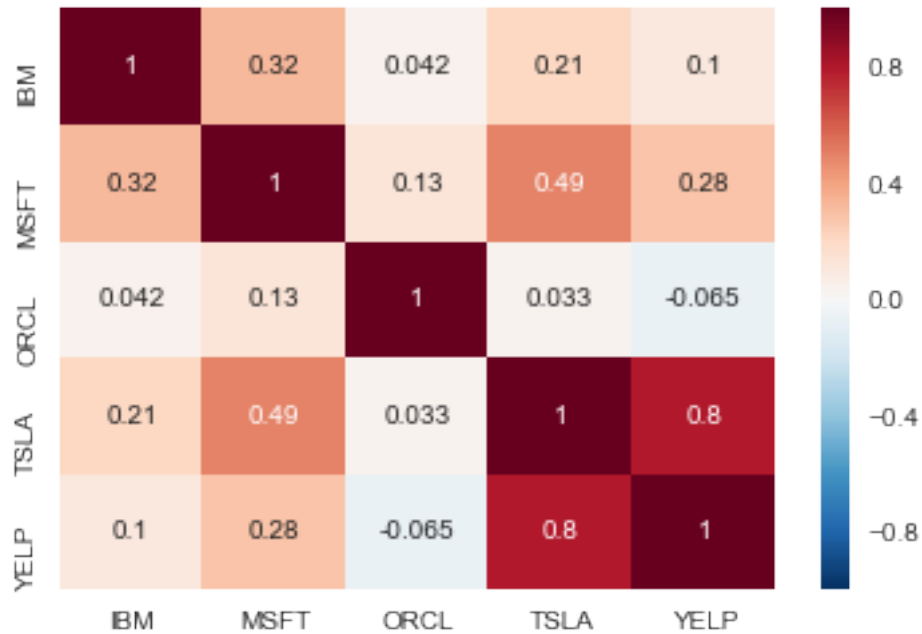
	IBM	MSFT	ORCL	TSLA	YELP
IBM	1.000000	0.321583	0.042213	0.208735	0.103837
MSFT	0.321583	1.000000	0.130515	0.492674	0.282827
ORCL	0.042213	0.130515	1.000000	0.032724	-0.065211
TSLA	0.208735	0.492674	0.032724	1.000000	0.800936
YELP	0.103837	0.282827	-0.065211	0.800936	1.000000

It takes a bit of time to examine that table and draw conclusions.

To speed that process up it helps to visualize the table.

We will learn more about visualization later, but for now this is a simple example.

```
In [250]: _ = sns.heatmap(rets.corr(), annot=True)
```



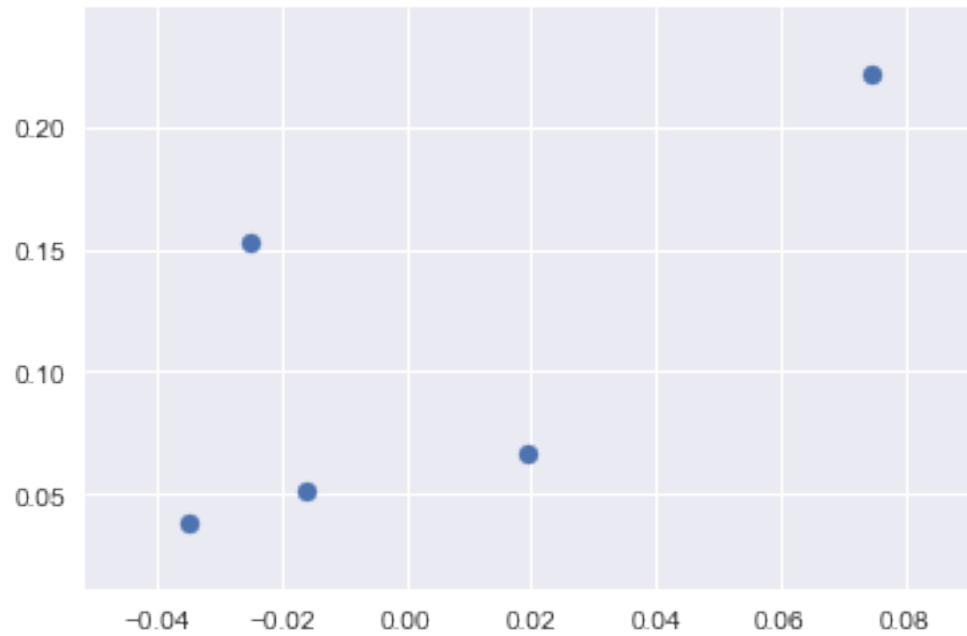
Finally, it is important to know that the plotting performed by Pandas is just a layer on top of matplotlib (i.e., the plt package).

So Panda's plots can (and should) be replaced or improved by using additional functions from matplotlib.

For example, suppose we want to know both the returns as well as the standard deviation of the returns of a stock (i.e., its risk).

Here is visualization of the result of such an analysis, and we construct the plot using only functions from matplotlib.

```
In [251]: _ = plt.scatter(rets.mean(), rets.std())
# plt.xlabel('Expected returns')
# plt.ylabel('Standard Deviation (Risk)')
# for label, x, y in zip(rets.columns, rets.mean(), rets.std()):
#     plt.annotate(
#         label,
#         xy = (x, y), xytext = (20, -20),
#         textcoords = 'offset points', ha = 'right', va = 'bottom',
#         bbox = dict(boxstyle = 'round,pad=0.5', fc = 'yellow', alpha = 0.5),
#         arrowprops = dict(arrowstyle = '->', connectionstyle = 'arc3,rad=0'))
```



To understand what these functions are doing, (especially the `annotate` function), you will need to consult the online documentation for `matplotlib`. Just use Google to find it.