# Review of pandas DataFrames

PANDAS FOUNDATIONS



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## pandas DataFrames

• Example: DataFrame of Apple Stock data

Date		Open	High	Low	Close	Volume	Adj Close
2014-09-1	16	99.80	101.26	98.89	100.86	66818200	100.86
2014-09-1	15	102.81	103.05	101.44	101.63	61216500	101.63
2014-09-1	2	101.21	102.19	101.08	101.66	62626100	101.66

#### Indexes and columns

```
import pandas as pd
type(AAPL)
pandas.core.frame.DataFrame
AAPL.shape
 (8514, 6)
AAPL.columns
Index(['Open', 'High', 'Low', 'Close', 'Volume', 'Adj Close'], dtype='object')
type(AAPL.columns)
pandas.indexes.base.Index
```



#### Indexes and columns

AAPL.index

type(AAPL.index)

pandas.tseries.index.DatetimeIndex



## Slicing

```
AAPL.iloc[:5,:]
```

	0pen	High	Low	Close	Volume	Adj Close
Date						
2014-09-16	99.80	101.26	98.89	100.86	66818200	100.86
2014-09-15	102.81	103.05	101.44	101.63	61216500	101.63
2014-09-12	101.21	102.19	101.08	101.66	62626100	101.66
2014-09-11	100.41	101.44	99.62	101.43	62353100	101.43
2014-09-10	98.01	101.11	97.76	101.00	100741900	101.00

#### AAPL.iloc[-5:,:]

		0pen	High	Low	Close	Volume	Adj Close
Date							
1980-	-12-18	26.63	26.75	26.63	26.63	18362400	0.41
1980-	-12-17	25.87	26.00	25.87	25.87	21610400	0.40
1980-	-12-16	25.37	25.37	25.25	25.25	26432000	0.39
1980-	-12-15	27.38	27.38	27.25	27.25	43971200	0.42
1980-	-12-12	28.75	28.87	28.75	28.75	117258400	0.45



## head()

#### AAPL.head(5)

	0pen	High	Low	Close	Volume	Adj Close
Date						
2014-09-16	99.80	101.26	98.89	100.86	66818200	100.86
2014-09-15	102.81	103.05	101.44	101.63	61216500	101.63
2014-09-12	101.21	102.19	101.08	101.66	62626100	101.66
2014-09-11	100.41	101.44	99.62	101.43	62353100	101.43
2014-09-10	98.01	101.11	97.76	101.00	100741900	101.00

#### AAPL.head(2)

	0pen	High	Low	Close	Volume	Adj Close
Date						
2014-09-16	99.80	101.26	98.89	100.86	66818200	100.86
2014-09-15	102.81	103.05	101.44	101.63	61216500	101.63



## tail()

#### AAPL.tail()

		0pen	High	Low	Close	Volume	Adj Close
ı	Date						
ı	1980-12-18	26.63	26.75	26.63	26.63	18362400	0.41
ı	1980-12-17	25.87	26.00	25.87	25.87	21610400	0.40
ı	1980-12-16	25.37	25.37	25.25	25.25	26432000	0.39
ı	1980-12-15	27.38	27.38	27.25	27.25	43971200	0.42
ı	1980-12-12	28.75	28.87	28.75	28.75	117258400	0.45

#### AAPL.tail(3)

	0pen	High	Low	Close	Volume	Adj Close
Date						
1980-12-16	25.37	25.37	25.25	25.25	26432000	0.39
1980-12-15	27.38	27.38	27.25	27.25	43971200	0.42
1980-12-12	28.75	28.87	28.75	28.75	117258400	0.45



#### info()

```
AAPL.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 8514 entries, 2014-09-16 to 1980-12-12
Data columns (total 6 columns):
           8514 non-null float64
0pen
High
     8514 non-null float64
     8514 non-null float64
Low
Close 8514 non-null float64
Volume 8514 non-null int64
Adj Close 8514 non-null float64
dtypes: float64(5), int64(1)
memory usage: 465.6 KB
```



#### Broadcasting

• Assigning scalar value to column slice broadcasts value to each row.

```
import numpy as np
AAPL.iloc[::3, -1] = np.nan
```

AAPL.head(6)

	0pen	High	Low	Close	Volume	Adj Close
Date						
2014-09-16	99.80	101.26	98.89	100.86	66818200	NaN
2014-09-15	102.81	103.05	101.44	101.63	61216500	101.63
2014-09-12	101.21	102.19	101.08	101.66	62626100	101.66
2014-09-11	100.41	101.44	99.62	101.43	62353100	NaN
2014-09-10	98.01	101.11	97.76	101.00	100741900	101.00
2014-09-09	99.08	103.08	96.14	97.99	189560600	97.99
2014-09-08	99.30	99.31	98.05	98.36	46277800	NaN

#### Broadcasting

AAPL.info()

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 8514 entries, 2014-09-16 to 1980-12-12
Data columns (total 6 columns):
            8514 non-null float64
0pen
            8514 non-null float64
High
            8514 non-null float64
Low
Close
            8514 non-null float64
Volume 8514 non-null int64
Adj Close 5676 non-null float64
dtypes: float64(5), int64(1)
memory usage: 465.6 KB
```



#### Series

```
low = AAPL['Low']
type(low)
pandas.core.series.Series
low.head()
Date
2014-09-16
               98.89
2014-09-15
              101.44
2014-09-12
              101.08
2014-09-11
              99.62
2014-09-10
               97.76
Name: Low, dtype: float64
lows = low.values
type(lows)
```

DataCamp

numpy.ndarray

# Let's practice!

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# Building DataFrames from scratch

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#### DataFrames from CSV files

```
import pandas as pd
users = pd.read_csv('datasets/users.csv', index_col=0)
print(users)
```

```
weekday city visitors signups
       Austin
                    139
   Sun
        Dallas
                   237
                             12
   Sun
        Austin
                    326
                              3
   Mon
       Dallas
                              5
   Mon
                    456
```



#### DataFrames from dict (1)

```
weekdaycity visitors signups0Sun Austin13971Sun Dallas237122Mon Austin32633Mon Dallas4565
```

#### DataFrames from dict (2)

```
import pandas as pd

cities = ['Austin', 'Dallas', 'Austin', 'Dallas']

signups = [7, 12, 3, 5]

visitors = [139, 237, 326, 456]

weekdays = ['Sun', 'Sun', 'Mon', 'Mon']

list_labels = ['city', 'signups', 'visitors', 'weekday']

list_cols = [cities, signups, visitors, weekdays]

zipped = list(zip(list_labels, list_cols))
```

#### DataFrames from dict (3)

```
print(zipped)
```

```
[('city', ['Austin', 'Dallas', 'Austin', 'Dallas']),
('signups', [7, 12, 3, 5]),
('visitors', [139, 237, 326, 456]),
('weekday', ['Sun', 'Sun', 'Mon', 'Mon'])]
```

```
data = dict(zipped)
users = pd.DataFrame(data)
print(users)
```

```
weekdaycityvisitorssignups0SunAustin13971SunDallas237122MonAustin32633MonDallas4565
```

#### Broadcasting

```
users['fees'] = 0 # Broadcasts to entire column
print(users)
```

```
city signups visitors weekday fees
Austin
                                    0
                    139
                            Sun
Dallas
                                    0
       12
                    237
                            Sun
             3
Austin
                    326
                            Mon
                                    0
Dallas
                    456
                            Mon
                                    0
```

## Broadcasting with a dict

```
import pandas as pd
heights = [ 59.0, 65.2, 62.9, 65.4, 63.7, 65.7, 64.1 ]
data = {'height': heights, 'sex': 'M'}
results = pd.DataFrame(data)
print(results)
```

```
height sex

0 59.0 M

1 65.2 M

2 62.9 M

3 65.4 M

4 63.7 M

5 65.7 M

6 64.1 M
```

#### Index and columns

```
results.columns = ['height (in)', 'sex']
results.index = ['A', 'B', 'C', 'D', 'E', 'F', 'G']
print(results)
```

```
height (in) sex

A 59.0 M

B 65.2 M

C 62.9 M

D 65.4 M

E 63.7 M

F 65.7 M

G 64.1 M
```



# Let's practice!

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# Importing & exporting data

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#### Original CSV file

Dataset: Sunspot observations collected from SILSO

```
1818,01,01,1818.004, -1,1
1818,01,02,1818.007, -1,1
1818,01,03,1818.010, -1,1
1818,01,04,1818.012, -1,1
1818,01,05,1818.015, -1,1
1818,01,06,1818.018, -1,1
```

<sup>&</sup>lt;sup>1</sup> Source: SILSO, Daily total sunspot number (http://www.sidc.be/silso/infossntotdaily)



#### Datasets from CSV files

```
import pandas as pd
filepath = 'ISSN_D_tot.csv'
sunspots = pd.read_csv(filepath)
sunspots.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 71921 entries, 0 to 71920
Data columns (total 6 columns):
           71921 non-null int64
1818
01
           71921 non-null int64
01.1
           71921 non-null int64
           71921 non-null float64
1818.004
 -1
           71921 non-null int64
           71921 non-null int64
dtypes: float64(1), int64(5)
memory usage: 3.3 MB
```



#### Datasets from CSV files

```
sunspots.iloc[10:20, :]
```

```
1818
          01
              01.1
                     1818.004
10
    1818
                 12
                     1818.034
    1818
                     1818.037
                                 22 1
                 13
    1818
                     1818.040
12
    1818
                     1818.042
13
                     1818.045
    1818
14
                 16
15
    1818
                     1818.048
                                 46 1
    1818
                     1818.051
16
                                 59
    1818
                     1818.053
                                 63
18
    1818
                     1818.056
                 20
    1818
                     1818.059
19
                 21
```



#### **Problems**

- CSV file has no column headers
  - Columns 0-2: Gregorian date (year, month, day)
  - Column 3: Date as fraction as year
  - Column 4: Daily total sunspot number
  - Column 5: Definitive/provisional indicator (1 or 0)
- Missing values in column 4: indicated by -1
- Dates representation inconvenient

#### Using header keyword

```
sunspots = pd.read_csv(filepath, header=None)
sunspots.iloc[10:20, :]
```

```
4 5
    1818
                 1818.031
                 1818.034
    1818
            12
                 1818.037
    1818
            13
                           22
                1818.040
13
    1818
         1 14
    1818
            15
                1818.042
    1818
                1818.045
15
            16
    1818
                 1818.048
16
                           46
            18
                1818.051
    1818
                           59 1
                 1818.053
    1818
18
            19
                           63 1
    1818
            20
                 1818.056
```



#### Using names keyword

```
month
                day dec_date sunspots definite
   year
10 1818
                     1818.031
                                     -1
   1818
                    1818.034
                                    -1
12 1818
                    1818.037
                                    22
13 1818
                    1818.040
                                     -1
  1818
                    1818.042
14
                                     -1
15 1818
                     1818.045
                                     -1
16 1818
                     1818.048
                                     46
17 1818
                    1818.051
                                     59
  1818
                     1818.053
                                     63
                    1818.056
   1818
                 20
                                     -1
```



## Using na\_values keyword (1)

```
dec_date sunspots definite
        month
                day
   year
  1818
                     1818.031
                                     -1
   1818
                 12
                     1818.034
                                     -1
12 1818
                     1818.037
                                     22
                     1818.040
13 1818
                                     -1
                     1818.042
14 1818
                                     -1
                     1818.045
   1818
                                     -1
16 1818
                     1818.048
                                     46
                     1818.051
   1818
                                     59
                     1818.053
                                     63
   1818
  1818
                 20 1818.056
                                     -1
```



## Using na\_values keyword (2)

```
dec_date sunspots definite
        month
                day
   year
  1818
                     1818.031
                                    NaN
   1818
                 12
                     1818.034
                                    NaN
12 1818
                     1818.037
                                   22.0
                     1818.040
                                    NaN
13 1818
                     1818.042
                                    NaN
14 1818
                     1818.045
                                    NaN
   1818
16 1818
                     1818.048
                                   46.0
                                   59.0
   1818
                     1818.051
                     1818.053
                                   63.0
   1818
  1818
                 20 1818.056
                                    NaN
```



## Using na\_values keyword (3)

```
day dec_date sunspots definite
        month
   year
  1818
                     1818.031
                                    NaN
   1818
                 12
                     1818.034
                                    NaN
12 1818
                     1818.037
                                   22.0
                     1818.040
                                    NaN
13 1818
                    1818.042
                                    NaN
14 1818
                                    NaN
   1818
                     1818.045
16 1818
                     1818.048
                                   46.0
                                   59.0
   1818
                     1818.051
   1818
                    1818.053
                                   63.0
  1818
                 20 1818.056
                                    NaN
```



#### Using parse\_dates keyword

```
year_month_day dec_date sunspots definite
10
      1818-01-11 1818.031
                                 NaN
                                 NaN
      1818-01-12 1818.034
      1818-01-13 1818.037
                                22.0
12
13
      1818-01-14 1818.040
                                 NaN
14
                                 NaN
      1818-01-15 1818.042
15
      1818-01-16 1818.045
                                 NaN
16
                                46.0
      1818-01-17 1818.048
17
      1818-01-18 1818.051
                                59.0
18
      1818-01-19 1818.053
                                63.0
19
      1818-01-20 1818.056
                                 NaN
```



#### Inspecting DataFrame

sunspots.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 71922 entries, 0 to 71921
Data columns (total 4 columns):
year_month_day 71922 non-null datetime64[ns]
dec_date
          71922 non-null float64
                 68675 non-null float64
sunspots
definite 71922 non-null int64
dtypes: datetime64[ns](1), float64(2), int64(1)
memory usage: 2.2 MB
```



#### Using dates as index

```
sunspots.index = sunspots['year_month_day']
sunspots.index.name = 'date'
sunspots.info()
```



## Trimming redundant columns

```
cols = ['sunspots', 'definite']
sunspots = sunspots[cols]
sunspots.iloc[10:20, :]
```

```
sunspots definite
date
1818-01-11
                 NaN
1818-01-12
                 NaN
1818-01-13
                22.0
                 NaN
1818-01-14
1818-01-15
                 NaN
1818-01-16
                 NaN
1818-01-17
                46.0
1818-01-18
                59.0
                63.0
1818-01-19
1818-01-20
                 NaN
```



#### Writing files

```
out_csv = 'sunspots.csv'
sunspots.to_csv(out_csv)

out_tsv = 'sunspots.tsv'
sunspots.to_csv(out_tsv, sep='\t')

out_xlsx = 'sunspots.xlsx'
sunspots.to_excel(out_xlsx)
```

# Let's practice!

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# Plotting with pandas

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#### **AAPL** stock data

	adj_close	close	high	low	open	volume
date						
2000-03-01	31.68	130.31	132.06	118.50	118.56	38478000
2000-03-02	29.66	122.00	127.94	120.69	127.00	11136800
2000-03-03	31.12	128.00	128.23	120.00	124.87	11565200
2000-03-06	30.56	125.69	129.13	125.00	126.00	7520000
2000-03-07	29.87	122.87	127.44	121.12	126.44	9767600
2000-03-08	29.66	122.00	123.94	118.56	122.87	9690800

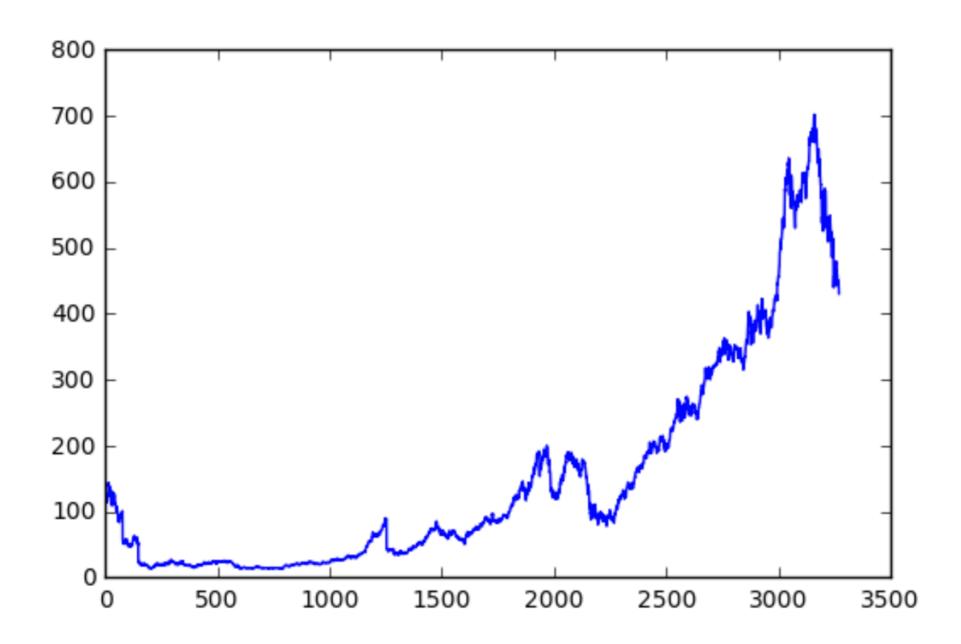


#### Plotting arrays (matplotlib)

```
close_arr = aapl['close'].values
type(close_arr)
numpy.ndarray
plt.plot(close_arr)
[<matplotlib.lines.Line2D at 0x115550358>]
plt.show()
```



# Plotting arrays (matplotlib)



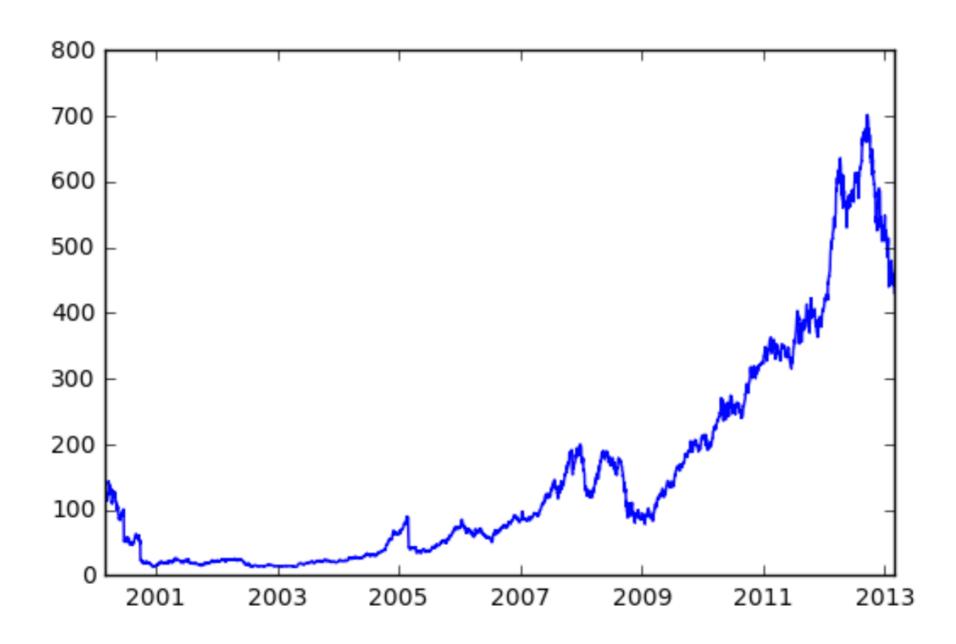


#### Plotting Series (matplotlib)

```
close_series = aapl['close']
type(close_series)
pandas.core.series.Series
plt.plot(close_series)
[<matplotlib.lines.Line2D at 0x11801cd30>]
plt.show()
```



# Plotting Series (matplotlib)

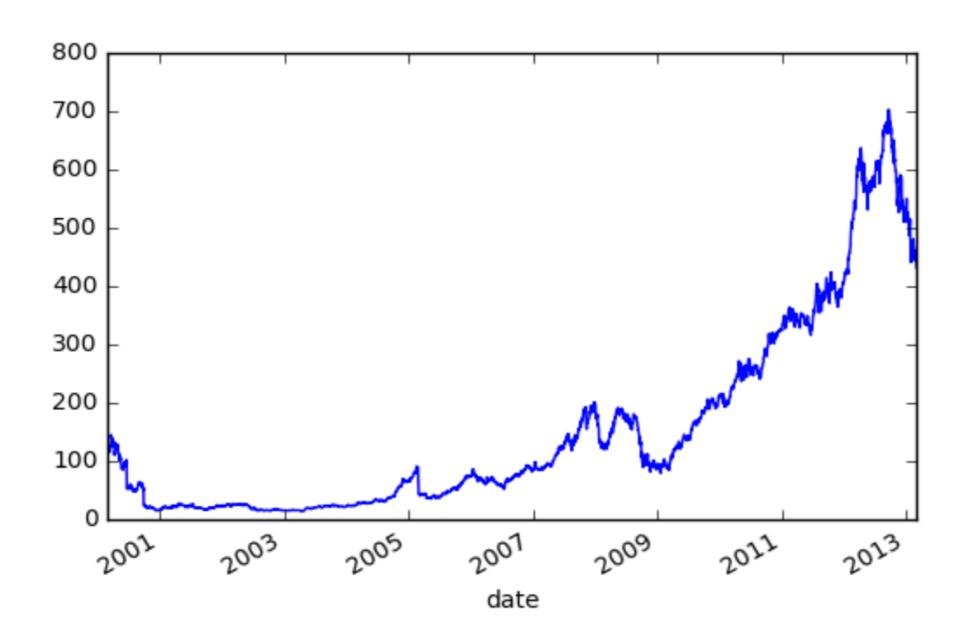




# Plotting Series (pandas)

```
close_series.plot() # plots Series directly
plt.show()
```

# Plotting Series (pandas)





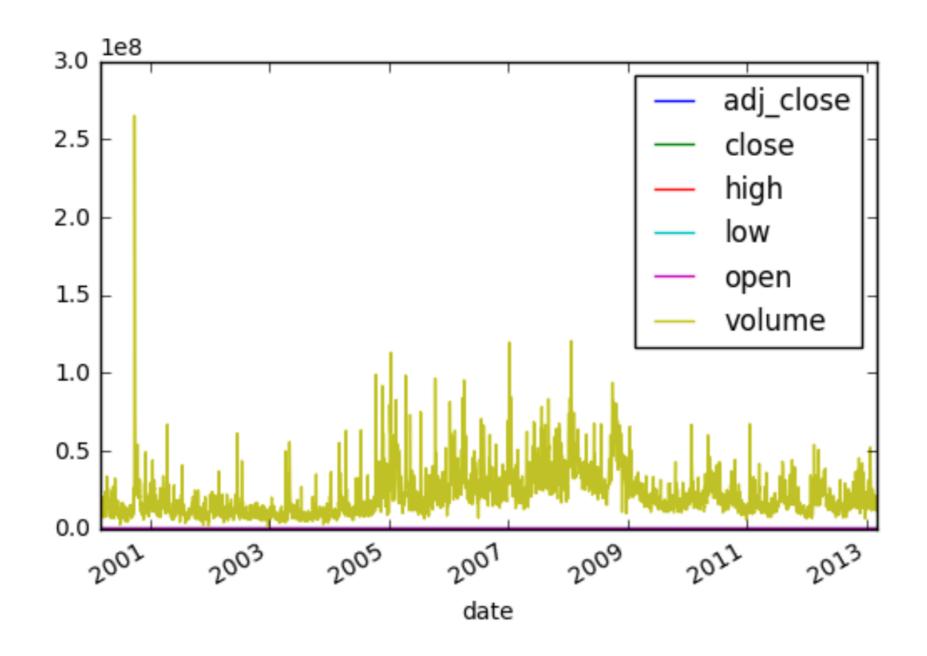
#### Plotting DataFrames (pandas)

```
aapl.plot() # plots all Series at once
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x118039b38>

plt.show()

### Plotting DataFrames (pandas)





#### Plotting DataFrames (matplotlib)

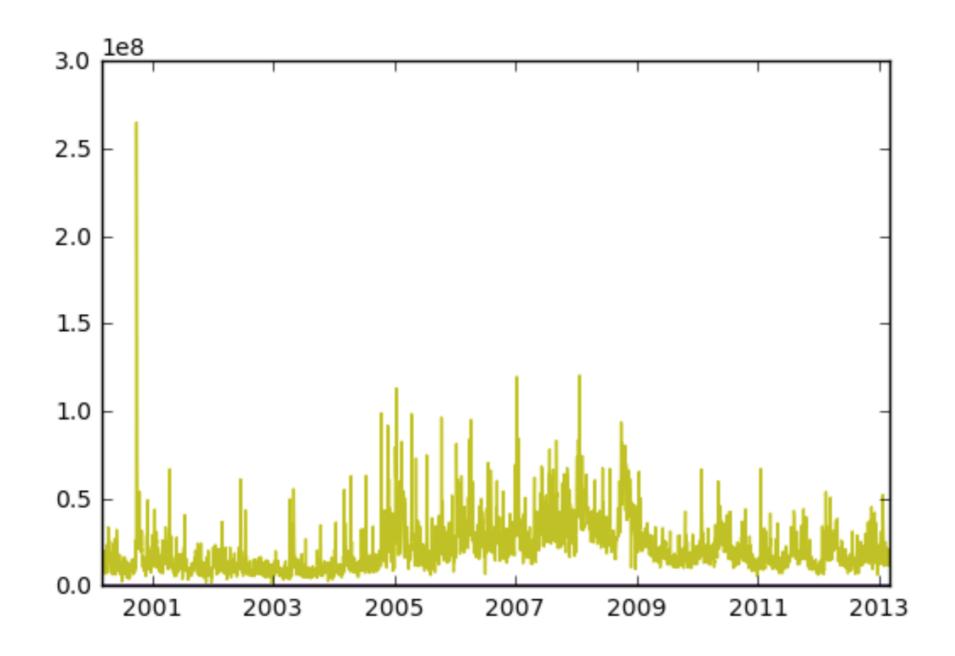
```
plt.plot(aapl) # plots all columns at once
```

```
<matplotlib.lines.Line2D at 0x1156290f0>,
<matplotlib.lines.Line2D at 0x1156525f8>,
<matplotlib.lines.Line2D at 0x1156527f0>,
<matplotlib.lines.Line2D at 0x1156529e8>,
<matplotlib.lines.Line2D at 0x115652be0>,
<matplotlib.lines.Line2D at 0x115652dd8>
```

```
plt.show()
```



# Plotting DataFrames (matplotlib)





#### Fixing scales

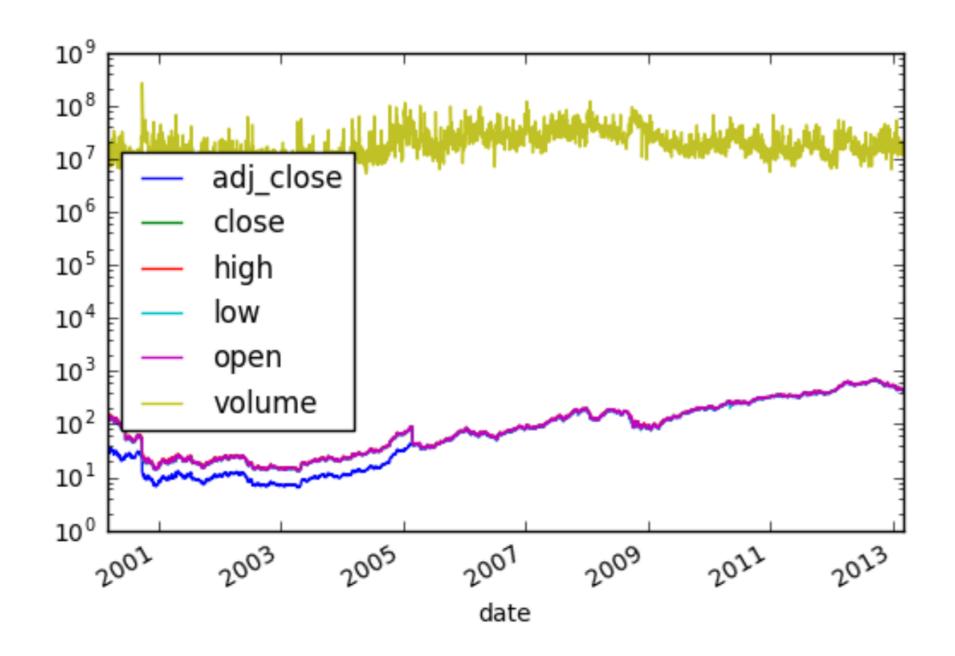
```
aapl.plot()
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x118afe048>

```
plt.yscale('log') # logarithmic scale on vertical axis
plt.show()
```



# Fixing scales

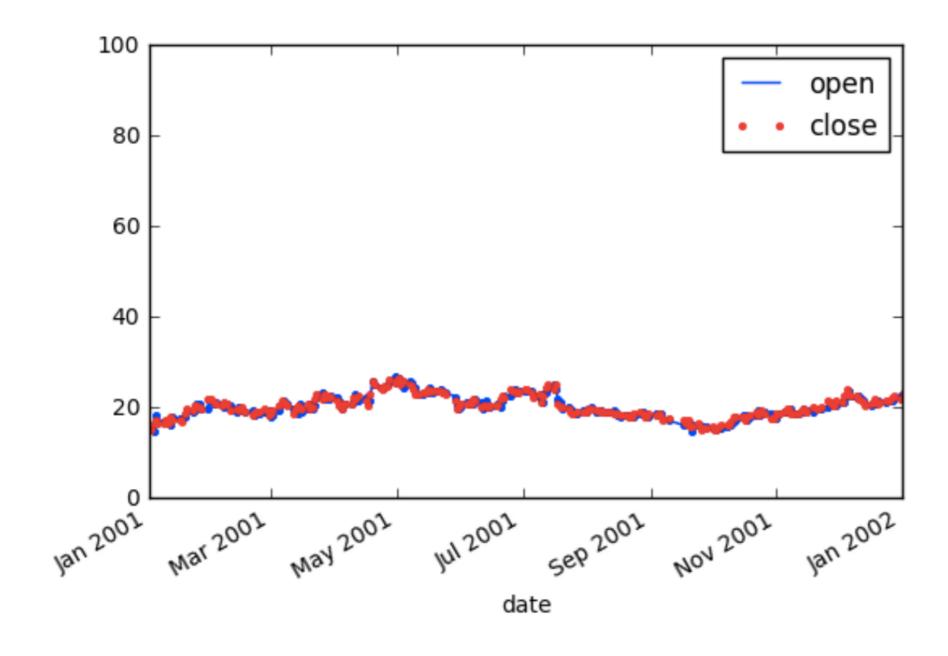


## **Customizing plots**

```
aapl['open'].plot(color='b', style='.-', legend=True)
<matplotlib.axes._subplots.AxesSubplot at 0x11a17db38>
aapl['close'].plot(color='r', style='.', legend=True)
<matplotlib.axes._subplots.AxesSubplot at 0x11a17db38>
plt.axis(('2001', '2002', 0, 100))
'2001', '2002', 0, 100)
plt.show()
```

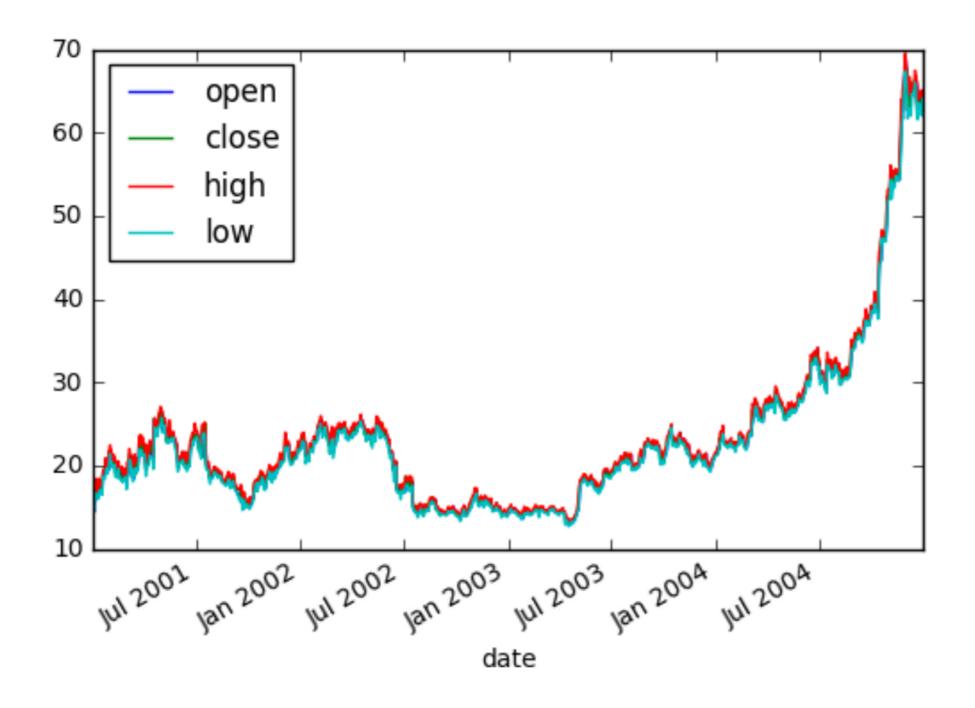


# **Customizing plots**





## Saving plots



#### Saving plots

<matplotlib.axes.\_subplots.AxesSubplot at 0x11ab42978>

```
plt.savefig('aapl.png')
plt.savefig('aapl.jpg')
plt.savefig('aapl.pdf')
plt.show()
```

# Let's practice!

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