

Tutorial: Market Simulator

Outline

1. (Review) Install Python and some libraries
2. Download Template File
3. Create a 'market simulator' that builds a portfolio, analyze it, computes expected return.
 1. Create an analyzer:
 - Edit **the analysis.py** file
 2. Create a market simulator on your own
 - Your **Simulator** will use functions from analysis.py which is **[Project 1] a warm-up project.**

Installation:

Step 1: Install your python platform

a) Install Anaconda

Step 2 (later) : Install Market Simulator Templates.

It needs SciPy — so:

Note: The **Anaconda python distribution** includes

* NumPy, Pandas, SciPy, Matplotlib, and Python,

and over 250 more packages available via a

simple “conda install <packagename>”

It also has an IDE.

Instructor got 2.7, and the anaconda distribution of python

To get the appropriate software you'll need:

python (scripting 'programming' language)

sci.py (numerical routines),

num.py (matrices, linear algebra), and

matplotlib (enables generating plots of data)

Installing Python (2.7) via Anaconda:

Anaconda instruction site including lots of libraries with python.

<https://docs.continuum.io/anaconda/install>

Mac Installation:

1) Instruction that the instructor used:

a) installed anaconda (got required packages)

[https://www.continuum.io/downloads\(2.7\)](https://www.continuum.io/downloads(2.7))

includes, sci.py, num.py, and matplotlib

Fundamentals

- **Read Data:** Read Stock Data from a CSV File and input it into a pandas DataFrame
 - [Pandas.DataFrame](#)
 - [Pands.read_csv](#)
- **Select Subsets of Data:** Select desired rows and columns
 - [Indexing and slicing data](#)
 - Gotchas: [Label-based slicing convention](#)
- **Generate Useful Plots:** Visual data by generating plots
 - [Plotting](#)
 - [Pandas.DataFrame.Plot](#)
 - [Matplot.pyplot.plot](#)

- Scrape S&P 500 ticker list and industry sectors from list of S&P 500 companies on Wikipedia (code provided).
 - https://en.wikipedia.org/wiki/List_of_S%26P_500_companies
- Download **daily close data** for each industry sector from Yahoo finance
 - using pandas DataReader.
- Build a sample Portfolio (in lecture by hand):
- Look at measures of the performance of a portfolio (project 1). We will use the first measure for project 1.
 - **Sharp ratio (in class)**
 - Treynor ratio
 - Jensen's alpha

Goal

- Go from RAW data (adjusted close prices in a .csv file) all the way to visualization

First Something Familiar: Weather Data

- .csv Comma Separated Values of weather conditions from **Oct 2009 to Aug 2017**
- Town of Cary, North Carolina
 - Temperature, pressure, humidity, ... lets see
 - Import as “text data”
- Next ... stock data.

https://catalog.data.gov/dataset?res_format=CSV&tags=weather

Comma Separated Values (.CSV)

- CSV File
- Header Files
- Lines/Rows of Dates
- Each Element is separated by columns
- Shift-ctrl-down

	A	B	C	D	E	F	G	H	
1	date	temperaturemin	temperaturemax	precipitation	snowfall	snowdepth	avgwindspeed	fastest2minwinddir	fas
2	6/25/12	72	93	0	0	0	6.49	40	
3	7/1/12	75	102.9	0	0	0	4.92	20	
4	7/6/12	71.1	100	0	0	0	3.8	20	
5	7/9/12	73	96.1	0.23	0	0	3.36	180	
6	7/11/12	68	80.1	0.45	0	0	4.03	90	
7	7/21/12	71.1	93	1.09	0	0	7.38	200	
8	7/25/12	70	90	0	0	0	4.03	240	
9	7/27/12	73.9	99	0.14	0	0	6.93	20	
10	7/29/12	66.9	91.9	0	0	0	2.01	100	
11	8/2/12	72	93	0.05	0	0	5.82	180	
12	8/6/12	73.9	93	0.49	0	0	6.26	230	
13	8/10/12	72	87.1	0.13	0	0	8.95	220	
14	8/14/12	66	91	0	0	0	5.59	230	
15	8/24/12	68	77	0.02	0	0	4.7	80	
16	8/26/12	64.9	84	0	0	0	3.13	50	
17	8/28/12	73	87.1	0	0	0	7.61	210	
18	8/30/12	68	89.1	0	0	0	4.47	170	
19	9/2/12	72	88	1.85	0	0	3.58	340	
20	9/14/12	63	80.1	0	0	0	1.57	50	
21	9/25/12	48	77	0	0	0	3.13	210	
22	10/2/12	66.9	88	0.12	0	0	7.61	210	
23	10/7/12	50	64.9	0.23	0	0	6.71	40	
24	10/14/12	44.1	78.1	0	0	0	3.8	230	
25	10/16/12	44.1	70	0	0	0	2.01	280	

What is in a Historical **Stock Data File**?

- a) # of employees
- b) Date/Time
- c) Company Name
- d) Price of the Stock
- e) Company's Hometown

What is in a Historical Stock Data File?

- a) # of employees
- b) Date/Time**
- c) Company Name (does not change over time)
- d) Price of the Stock**
- e) Company's Hometown (does not change over time)

Comma Separated Values (.CSV)

- Stock Data from Yahoo Finance
- CSV file pulled by panda's (later) DataReader()

```
1 Date,Open,High,Low,Close,Volume,Adj Close
2 2012-09-12,57.01,57.54,56.68,56.91,2362700,56.91
3 2012-09-11,56.15,56.73,56.14,56.68,2118300,56.68
4 2012-09-10,57.01,57.07,56.02,56.12,2772700,56.12
5 2012-09-07,56.60,57.20,56.38,57.19,3011800,57.19
6 2012-09-06,55.08,56.32,54.69,56.26,3304200,56.26
7 2012-09-05,54.55,54.82,54.34,54.67,2660700,54.67
8 2012-09-04,54.50,55.06,54.26,54.63,2299900,54.63
9 2012-08-31,54.41,54.90,54.15,54.51,2368400,54.51
10 2012-08-30,54.01,54.24,53.59,54.09,2233600,54.09
11 2012-08-29,53.86,54.68,53.75,54.49,2772800,54.49
12 2012-08-28,54.00,54.29,53.79,53.83,2065600,53.83
13 2012-08-27,54.79,54.85,53.94,54.15,1736300,54.15
```

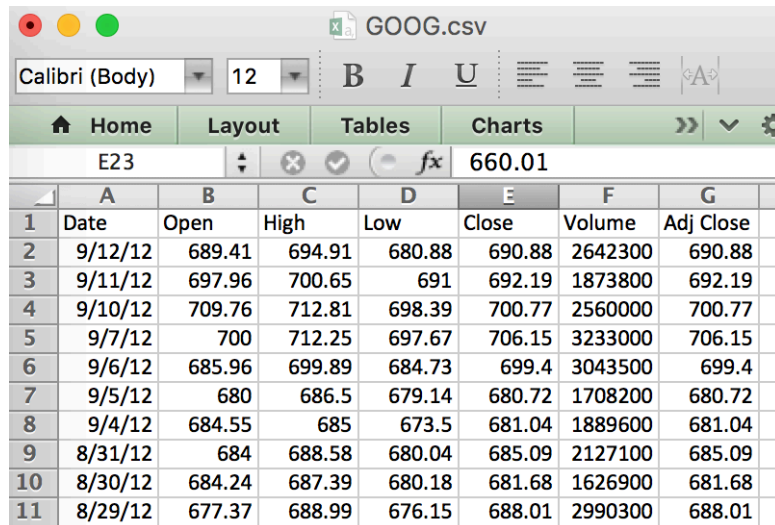
	A1							
	Date	Open	High	Low	Close	Volume	Adj Close	
1	Date	Open	High	Low	Close	Volume	Adj Close	
2	9/12/12	57.01	57.54	56.68	56.91	2362700	56.91	
3	9/11/12	56.15	56.73	56.14	56.68	2118300	56.68	
4	9/10/12	57.01	57.07	56.02	56.12	2772700	56.12	
5	9/7/12	56.6	57.2	56.38	57.19	3011800	57.19	
6	9/6/12	55.08	56.32	54.69	56.26	3304200	56.26	
7	9/5/12	54.55	54.82	54.34	54.67	2660700	54.67	

Stock Data Files

- **Date**
- **Open** – price stock opens at in the morning, it is first price in the day.
- **High** – highest price in the day
- **Low** – lowest price in the day
- **Close** – closing price at 4 PM.
- **Volume** – how many shares traded all together on that day.
- **Adjusted Close** – accounts for splits/and dividends – encapsulates the increase in value if you hold stock for a long time (later).

GOOG.csv (from Yahoo).

- Newer dates on top, older descending.

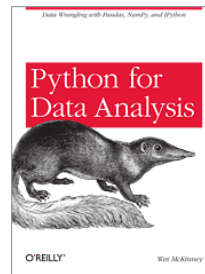


	A	B	C	D	E	F	G
1	Date	Open	High	Low	Close	Volume	Adj Close
2	9/12/12	689.41	694.91	680.88	690.88	2642300	690.88
3	9/11/12	697.96	700.65	691	692.19	1873800	692.19
4	9/10/12	709.76	712.81	698.39	700.77	2560000	700.77
5	9/7/12	700	712.25	697.67	706.15	3233000	706.15
6	9/6/12	685.96	699.89	684.73	699.4	3043500	699.4
7	9/5/12	680	686.5	679.14	680.72	1708200	680.72
8	9/4/12	684.55	685	673.5	681.04	1889600	681.04
9	8/31/12	684	688.58	680.04	685.09	2127100	685.09
10	8/30/12	684.24	687.39	680.18	681.68	1626900	681.68
11	8/29/12	677.37	688.99	676.15	688.01	2990300	688.01

- Adjusted Close – adjusts / accounts for stocks splits and dividend payments.
- On the Current Day – Adjusted Close and Close are always the same.
- Previous Days:
 - But as we go back in time start they to differ they are not always the same.
 - Actual Return is not captured by the closing price, need to use adjusted close on historical data.

Pandas: Included in Anaconda

- [https://en.wikipedia.org/wiki/Pandas_\(software\)](https://en.wikipedia.org/wiki/Pandas_(software))
- Developed by Wes McKinney while at AQR Capital Management to analyze financial data
 - Open Source.
 - Numerical Tables and **Time Series**
 - **A Key Element : Data Frames**
 - **Slicing**
 - Panel Data



Store Portfolio in a Panda Data Frame

- Want: <Symbols> vs Time
- Includes a set of equities (ownership)
 - Exchange Traded Fund (ETF)
 - SPY 500
 - Tracks the index S&P 500 Index.
 - Russell 1000
 - AAPL – apple
 - GOOG – Google
 - Other: securities (government)
- NaN
- <https://en.wikipedia.org/wiki/Google>
 - Initial public offering (IPO) - August 19, 2004.

symbols

→

	SPY	AAPL	GOOG	GLD
2010-01-04				
2010-01-05				
2010-01-06				
2010-01-07				
2010-01-08				
2010-01-11				
2010-01-12				
2010-01-13				
2010-01-14				
2010-01-15				
2010-01-19				
2010-01-20				
2010-01-21				
2010-01-22				
2010-01-25				

time

↓

Warm-up: Reading into a Data frame

- Interactively
 - Import pandas
 - Rename it to pd
- Read it in.
- First column is index helping you to access rows.
- SPY, AAPL, GOOG, GLD

```
{ingrid:632} python
Python 2.7.11 |Anaconda 4.1.0 (x86_64)| (default, Jun 15 2016, 16:09:16)
[GCC 4.2.1 (Based on Apple Inc. build 5658) (LLVM build 2336.11.00)] on darwin
Type "help", "copyright", "credits" or "license" for more information.
Anaconda is brought to you by Continuum Analytics.
Please check out: http://continuum.io/thanks and https://anaconda.org
>>> import pandas as pd
>>> df = pd.read_csv("data/AAPL.csv")
>>> print df.head()
   Date      Open      High      Low      Close      Volume  Adj Close
0  2012-09-12  666.85  669.90  656.00  669.79  25410600    669.79
1  2012-09-11  665.11  670.10  656.50  660.59  17987400    660.59
2  2012-09-10  680.45  683.29  662.10  662.74  17428500    662.74
3  2012-09-07  678.05  682.48  675.77  680.44  11773800    680.44
4  2012-09-06  673.17  678.29  670.80  676.27  13971300    676.27
>>> print df
   Date      Open      High      Low      Close      Volume  Adj Close
0  2012-09-12  666.85  669.90  656.00  669.79  25410600    669.79
1  2012-09-11  665.11  670.10  656.50  660.59  17987400    660.59
2  2012-09-10  680.45  683.29  662.10  662.74  17428500    662.74
3  2012-09-07  678.05  682.48  675.77  680.44  11773800    680.44
4  2012-09-06  673.17  678.29  670.80  676.27  13971300    676.27
5  2012-09-05  675.57  676.35  669.60  670.23  12013400    670.23
6  2012-09-04  665.76  675.14  664.50  674.97  13139000    674.97
7  2012-08-31  667.25  668.60  657.25  665.24  12082900    665.24
8  2012-08-30  670.64  671.55  662.85  663.87  10810700    663.87
9  2012-08-29  675.25  677.67  672.60  673.47  7243100    673.47
10 2012-08-28  674.98  676.10  670.67  674.80  9550600    674.80
11 2012-08-27  679.99  680.87  673.54  675.68  15250300    675.68
12 2012-08-24  659.51  669.48  655.55  663.22  15619300    663.22
13 2012-08-23  666.11  669.90  661.15  662.63  15004600    662.63
14 2012-08-22  654.42  669.00  648.11  668.87  20190100    668.87
15 2012-08-21  670.82  674.88  650.33  656.06  29025700    656.06
16 2012-08-20  650.01  665.15  649.90  665.15  21906600    665.15
17 2012-08-17  640.00  648.10  638.81  648.11  15812000    648.11
```

Exercises

Exercise 1.

- Read in the entire CSV file in a function
 - Print it out.

Exercise 2.

- Read in the entire file in a function
 - Print out a selection of file
 - Top 5 lines : .head()
 - Bottom 5 lines: .tail()

def -- Make it a function

```
1 import pandas as pd
2
3
4 def test_run():
5     df = pd.read_csv("data/AAPL.csv")
6     print df #print entire dataframe
7
8
9 if __name__ == "__main__":
10     test_run()
```

- simple-frame.py
 - Entire frame
 - **Try: printing - `df.head()`, `df.tail()`**
 - **Question:** Print last 5 lines?
-
- Only print top 5 line of data frame
 - print `df.head()`
 - Only print bottom 5 lines of data frame
 - print `df.tail()`
- Print out a subset of columns, and/or rows:**
- **Slicing:** Only print rows between index 10, 20 (not inclusive)
 - print `df[10:21]`
 - print `df[:21]`
 - print `df[['Date', 'High']].values[5]`

Computation on CVS File

- From the file, find out maximum closing price.
 1. Read the file into a data frame
 - **Now** - SPY.csv
 - **Later** – any symbol.
 2. Process the Column 'Close'
 3. Use pandas function .max() to return max.

Compute Max Closing Price get_max_close(symbol)

```
import pandas as pd

def get_max_close(symbol):
    """Return the maximum closing value for stock indicated by symbol.

    Note: Data for stock is stored in file: data/<symbol>.csv
    """
    df = pd.read_csv("data/{}.csv".format(symbol)) # read in data
    return df['Close'].max() # compute & return max

def test_run():
    """Function called by Test Run."""
    for symbol in ['AAPL', 'IBM']:
        print "Max close"
        print symbol, get_max_close(symbol)

if __name__ == "__main__": # if run standalone
    test_run()
```

Exercises

- Calculate the **mean volume**.
- Calculate the **max** adjusted close.
- **Challenge:** Return date(s) when :
 - closing price is different from the adjusted price?
 - IBM

1b-meanvolume-quiz.py

Plotting matplotlib

```
import pandas as pd
import matplotlib.pyplot as plt

def test_run():
    df = pd.read_csv("data/AAPL.csv") # read in data
    print df ['Adj Close']
    df ['Adj Close'].plot()
    plt.show() # must be called to show plots

if __name__ == "__main__": # if run standalone
    test_run()
```

2a-1column-plots.py

http://matplotlib.org/users/pyplot_tutorial.html#working-with-text

Plot 2 Columns in a single Plot

```
import pandas as pd
import matplotlib.pyplot as plt

def test_run():
    df = pd.read_csv("data/AAPL.csv") # read in data
    #print df ['Adj Close']
    df [['Close', 'Adj Close']].plot() # double square brackets.
    plt.show() # must be called to show plots

if __name__ == "__main__": # if run standalone
    test_run()
```

2b-2column-plots.py

Coming UP.

- Restrict Data Ranges (e.g., specific date range)?
(join)
- Drop Missing Data Rows
- Join Data Incrementally, column by column

Want to get a frame with Closing date of Different Stocks.

```
{ingrid:503} python 3b-multiple.py
Adj Close    GOOG    IBM    GLD
2010-01-22   104.34  550.01  119.61  107.17
2010-01-25   104.87  540.00  120.20  107.48
2010-01-26   104.43  542.42  119.85  107.56
```

Only on trading days ...

How many days were US Stocks Traded in 2014 (over an entire year)

- a) 365
- b) 260
- c) 252

How many days were US Stocks Traded in 2014 (over an entire year)

- a) 365
- b) 260 (52x5) But there are also holidays ...
- c) **252**

Steps: Building a DataFrame

1. DF1 = First build a data frame by **specifying the date range**.
 - Includes weekend dates (markets are not open).
2. DF2 = SPY = Load in SPY data (adjusted close) into a separate data frame (all data and prices).
 - Only trading days (market open) in DF2.
3. **Join DF2 and DF 1** – join so that **only dates that are present** in 'both' frames (it eliminates the weekends in Data frame 1).
4. Additional Joins with other 'symbol' that we want to add, IBM, GOOG.

Steps 0-2: Specifying the Data Range

- **Step 0:**
- **Step 1: Create a list of data time index objects**
 - `dates = pd.date_range(start_date, end_date)`
 - Check it out (print).
 - List of data time index objects
 - `Dates[0]` (dates with time stamp)
 - `Dates[1]`
- **Step 2. Index it by dates** instead of integer by specifying index and setting it to 'dates'
 - `index = dates`.
 - NOTE seen the default of integers already ...

3a-simple-join.py

Step 3: Combine the data frames with Joining Frames

a) `df2`: Create SPY date frame w/ SPY data

b) Combine date frames via join.

- `df1`: Empty date frame with a **date range**
- `df2_SPY` Populated date frame (only trading days)
- Join: **left** join
 - `df1.join(df2_SPY)`
 - Only SPY row are retained.
- ? No values from SPY??

- dfSPY is indexed by integers by default, change index to dates by index_col
 - index_col="Date"

- Multiple Stocks from a list
 - symbols = ['GOOG', 'IBM', 'GLD']
 - For loop iterating through symbols

```
pd_read_csv("data/{}.csv".format(symbol),
            index_col='Date',
            parse_dates=True,
            Usecols=['Date',Adj Close'],
            na_values=['nan'])
```
 - ... overlap of Adj Close column
 - Rename the column to stock symbol instead.

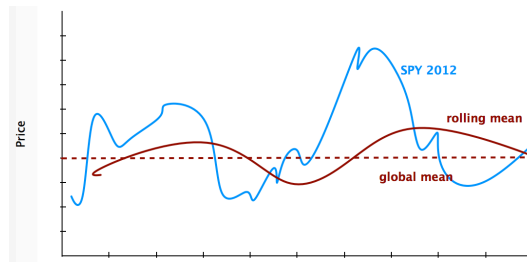
Exercise:

- Utility Functions to read in data no NaNs.

```
3 import os
4 import pandas as pd
5
6 def symbol_to_path(symbol, base_dir="data"):
7     """Return CSV file path given ticker symbol."""
8     return os.path.join(base_dir, "{}.csv".format(str(symbol)))
9
10
11 def get_data(symbols, dates):
12     """Read stock data (adjusted close) for given symbols from CSV files."""
13     df = pd.DataFrame(index=dates)
14     if 'SPY' not in symbols: # add SPY for reference, if absent
15         symbols.insert(0, 'SPY')
16
17     for symbol in symbols:
18         # TODO: Read and join data for each symbol
19
20     return df
21
```

Re-Cap: Last Week

- Worked on board ... on code.
- Compute / Code financial statistics in pandas and numpy:
 - Global Statistics
 - Mean
 - Median
 - Standard Deviations
 - Rolling Statistics
 - Rolling mean
 - Representation of underlying value of a stock
 - Rolling standard deviation
 - deviate from the mean (buy and sell signal)



- **Bollinger Bands**
 - Upper band :
 - $\text{rolling mean} + 2 * \text{rolling StdDev}$
 - Lower band :
 - $\text{rolling mean} - 2 * \text{rolling StdDev}$



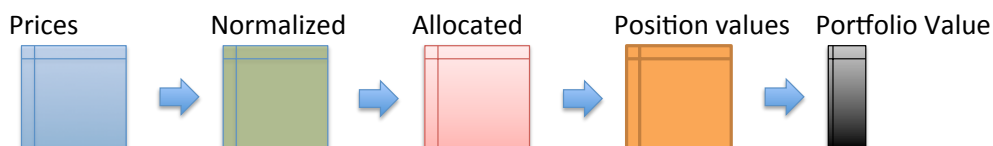
https://en.wikipedia.org/wiki/Bollinger_Bands

Get the Daily Total Value of the Portfolio

- **Step 1:** Prices Data Frame index by dates
- **Step 2:** Normalize by First Row
 - $\text{Normed} = \text{prices} / \text{prices}[0]$
- **Step 3:** Multiply by allocation (a vector)
 - $\text{Allocated} = \text{Normed} * \text{allocs}$
- **Step 4:** Position values = worth each day
 - $\text{Pos_vals} = \text{Allocated} * \text{start_val}$
- **Step 5:** Daily Total Value of Portfolio
 - $\text{Port_val} = \text{Pos_vals}.\text{sum}(\text{axis} = 1)$

Given:

```
start_val = $1,000,000
start_date = 2011-01-01
end_date = 2011-12-31
symbols = ['SPY', 'XOM',
           'GOOG', 'GLD']
allocs = [0.4, 0.4, 0.1, 0.1]
```



Daily Return on the portfolio value

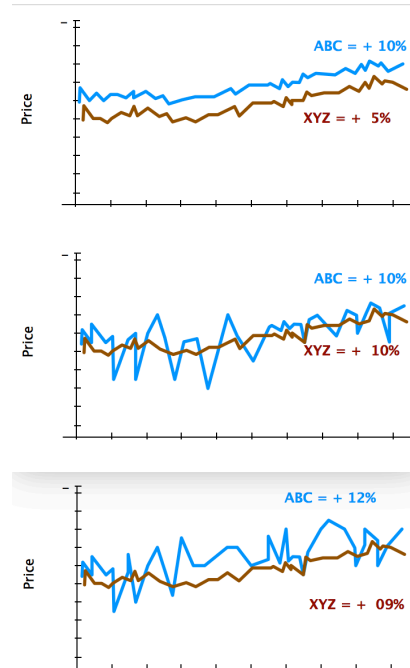
- Daily return[t] = (prices[t]/prices[t-1]) - 1
 - Now on port_val (instead of prices).
 - Observation: 1st value is always 0
 - daily_rets = daily_rets[1:]

Statistics on the Portfolio

- Cumulative Returns
 - Form beginning to end (last value/initialial val) - 1
 - cum_ret = (port_val[-1]/port_val[0]) - 1
- Average Daily Returns
 - daily_rets.mean()
- Standard Deviation of Daily Return
 - Daily_rets.std()
- Sharpe Ratio

Sharpe Ratio

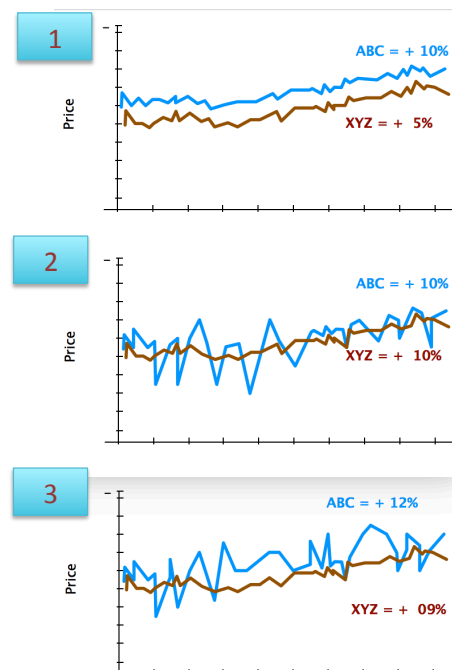
- Considers our return in the context of risk
- Risk is volatile (standard deviation)
- Adjust our return in return for the risk
- Volatility
 - Measured by standard deviation



Which is better?

Sharpe Ratio

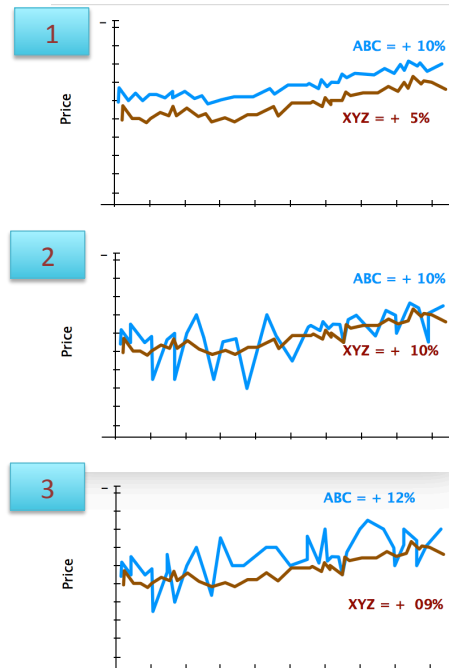
- Considers our return in the context of risk
- Risk is volatile (standard deviation)
- Adjust our return in return for the risk
- Volatility
 - Measured by standard deviation



Which is better?

Sharpe Ratio

1. Higher Returns is Better
2. Less Volatility/Less Risk is Better
3. Not Enough Information
 - Returns: $ABC > XYZ$
 - Volatility $ABC > XYZ$
 - **ABC is higher returns, but more risk**



Which is better?

Sharpe Ratio

- Adjusts return for risk
- A quantities way to assess portfolio
 - 1. ABC is better because same volatility but higher returns
 - 2. same returns but XYZ has lower risk so XYZ is better
 - 3. Sharpe Ratio may give as a number to determine which is better
- Sharpe ratio also considers (comparative)
 - Risk free rate of returns
 - Bank account or treasure note
 - Lately risk free return is 0, bank interest rate is 0, or or close to 0

Which Formula is Best?

- R_p : Portfolio Return
- R_f : Risk Free Rate of Return
- σ_p : Standard Deviation of Portfolio Return

a) $R_p - R_f + \sigma_p$

b) $R_f / R_f - \sigma_p$

c) $(R_p - R_f) / \sigma_p$

General Form of the Sharpe Ratio

Computing Sharpe Ratio

- SR (expected value)
= $E [R_p - R_f] / \text{std}[R_p - R_f]$
- Expected value \rightarrow mean over time:
= $\text{mean}(\text{daily_rets} - \text{daily_rf}) / \text{std}(\text{daily_rets} - \text{daily_rf})$
- What is the risk free rate?
 - LIBOR (London Inter Bank Offer Rate)
 - Interest Rate 3 months Treasury Bill
 - 0%! Short Cut.

Computing Sharpe Ratio

- SR (expected value)
= $E [R_p - R_f] / \text{std}[R_p - R_f]$
- Expected value \rightarrow mean over time:
= $\text{mean}(\text{daily_rets} - \text{daily_rf}) / \text{std}(\text{daily_rets} - \text{daily_rf})$
- Risk Free Rate not given on a daily bases
 - LIBOR
 - Annual/6 month bases
 - Short Cut –
 - Convert Annual to a daily amount
 - Example:
 - Annual Rate: 0.1 per year Risk Free RATE
 - Total Value at end of year: $1.0 * 0.1$
 - What is the Interest Rate per Day:
 - » $\text{Daily_RF} = \text{SQRT}_{252}(1.0 + 0.1) - 1 \rightarrow 0.0$ (approximation)
 - Constant Standard Deviation of a Constant

Sample Frequency

- SR can vary depending on how frequently we sample the data (need an adjustment factor to convert between different sampling)
 - Annual (initial vision of SR)
 - Monthly
 - Daily

$$SR_{\text{annualized}} = k * SR$$

$$k = \text{sqrt} (\# \text{ samples per year})$$

Sample Frequency

- SR can vary depending on how frequently we sample the data (need an adjustment factor to convert between different sampling)
 - Annual (initial vision of SR)
 - Monthly
 - Daily

Daily $k = \sqrt{252}$
Weekly $k = \sqrt{52}$
Monthly $k = \sqrt{12}$

$$SR_{\text{annualized}} = k * SR$$

$$k = \sqrt{\text{\# samples per year}}$$

ReCap: Sharpe Ratio for Daily Returns

- SR
= $\sqrt{252}$
* $(\text{mean}(\text{daily_rets} - \text{daily_rf}) / \text{std}(\text{daily_rets} - \text{daily_rf}))$

Quiz: What is the Sharpe Ratio

- Given:
 - 60 days of data
 - Average daily return = 0.001 (10 bases points)
 - Dailyrisk free return = 0.0002 (2 bases points)
 - Std daily return = 0.001
- What is the Sharpe Ratio?

Quiz: What is the Sharpe Ratio

- Given:
 - 60 days of data
 - Average daily return = 0.001 (10 bases points)
 - Dailyrisk free return = 0.0002 (2 bases points)
 - Std daily return = 0.001
- What is the Sharpe Ratio?
- $= \text{Sqrt}(252) * \text{mean}(R_p - R_f) / \text{Std}(R_p)$
 - $= \text{Sqrt}(252) * (10 - 2) / 10 = 12.7$

Python

- `sharpe_ratio =`
`np.sqrt(samples_per_year)`
`* np.mean(daily_rets - daily_rf) / std_daily_ret`

