# Final Project

Due: Dec 10, 2015

Choose one of the following two topics to implement for your final project.

## • Topic 1: Gamma Scalping Trading Strategy (100 points)

Design a C++ or a Python program to implement the Gamma scalping trading strategy involving one underlying stock and two at-the-money (ATM) options (one Call and one Put) as described in the posted PPT lecture file.

Let the Gamma scalping testing period be from start date  $t_0$  to end date  $t_N$ . At the start date  $t_0$ , an initial capital in the amount of \$1,000,000.00 is deposited in a trading account. For simplicity, assume the trading account does not earn interest from the cash in the account, and additional cash can be borrowed from bank at risk free interest rate if needed.

The implemented strategy is expected to contain the following three main components.

## 1. Entry/Exit Signal Generator

- At  $t_0$ , instead of gamma scalping right away, we start observing the moving average of volatility to see when the underlying volatility is about to explode, so that we can find profitable trading periods to do gamma scalping. We automate such a process by implementing an Entry/Exit Signal Generator.
- Pick an integer w > 0 to be the length of time window over which the moving average of volatility is computed. Let  $MA_w(\sigma_t)$  denote the average volatility at time t over periods t-1,  $t-2, \ldots, t-w$ .
- Trade entry/exit signal generator.
  - \* Pick two positive parameters: lower threshold  $\sigma_{entry}$  and higher threshold  $\sigma_{exit}$  for the signal generator.  $\sigma_{entry} < \sigma_{exit}$ .
  - \* Entry/exit signal rules:
    - · Entry signal: If the moving average of volatility  $MA(\sigma_t)$  goes down to the lower threshold, namely,  $MA(\sigma_t) < \sigma_{entry}$ , then the signal generator will output an entry signal. It indicates the volatility is in the low range and may subsequently go up, and so a gamma scalping trade is initiated in this period  $t_{entry}$ .
    - · Exit signal: After a gamma scalping position is opened, if the moving average of volatility reaches up to the higher threshold, namely,  $MA(\sigma_t) > \sigma_{exit}$ , or expiration date is no more than E days later (E is a parameter which can be set to a number between 2 to 20), the a trade exit signal is generated. All the positions of this gamma scalp trade are closed immediately (namely, close the underlying positions and sell call and put) to lock our profits. After that, we wait until the next entry signal.
  - \* The entry/exit signal generator might use different sources of volatilities to calculate  $\sigma_t$ . We test two sources:
    - $\cdot$  Historical volatility: sample standard deviation of last m underlying stock prices (or, log-returns):

$$\sigma_t = \sigma_m(S_t)$$

· The average of implied volatility of at-the-money call and put options:

$$\sigma_t = (IV_t(Call_{ATM}) + IV_t(Put_{ATM}))/2$$

- **Note**: Between  $t_0$  and  $t_N$ , the gamma scalping trade may be carried out for multiple times, depending on the choice of the entry/exit signal thresholds.

#### 2. Portfolio Rebalancing Engine

- Select one trading parameter,  $\bar{\Delta}$ , which represents the threshold for rebalancing the total Delta position of the gamma scalping trade to 0. For example,  $\bar{\Delta} = 0.15$ .
- After a trade entry signal is generated in period  $t_{entry}$ , the gamma scalping trade is opened by purchasing K shares of ATM straddles: namely, simultaneously purchasing K shares of Call and K shares of Put which have the same strike price equaling the current underlying stock price, and the identical expiration date T.
- Compute the Deltas of 1 share of call  $(\Delta_{Call}(t))$  and 1 share of put  $(\Delta_{Put}(t))$  at time t (using Black-Scholes implied volatility function).
  - \* If  $t = t_{entry}$ : If  $-\bar{\Delta} \leq (\Delta_{Call}(t_{entry}) + \Delta_{Put}(t_{entry})) \leq \bar{\Delta}$ , then do nothing and set  $\Delta_{stock}(t_{entry}) = 0$ . If otherwise, long  $(\Delta_{stock}(t_{entry})^*100^*\text{K}$  shares) of the underlying stocks to make the overall position delta neutral, where  $\Delta_{stock}(t_{entry}) \equiv -1 * (\Delta_{Call}(t_{entry}) + \Delta_{Put}(t_{entry}))$  (If the sign of  $\Delta_{stock}(t_{entry})$  is negative, then the trading action is to short). Such a portfolio is gamma positive.
  - \* If  $t > t_{entry}$ : for each passing period t, we calculate the portfolio Delta  $\Delta_t = \Delta_{Call}(t) + \Delta_{Put}(t) + \Delta_{stock}(t-1)$ .
    - · If  $|\Delta_t| \leq \bar{\Delta}$ , do nothing and set  $\Delta_{stock}(t) = \Delta_{stock}(t-1)$ .
    - · If  $|\Delta_t| > \bar{\Delta}$ , then buy/sell  $K*100*\Delta_t$  shares of stock to make the portfolio Delta  $\Delta_t$  equal 0. The position of the underlying stock is updated to  $\Delta_{stock}(t) = (-1*\Delta_t + \Delta_{stock}(t-1))$ . (Note: one share of option contract corresponds to 100 shares of the underlying stock.)
- We keep adjusting the number of shares of the stocks in the gamma scalping trade portfolio until the exit signal or expiration date, while recording all of the long/short actions and the corresponding trading gain/loss in each period t.
- Transactions costs:  $TC_{cost} = 0.001$  of the transaction dollar amount. Note that the value of 1 share of Call/Put is equal to  $100^*$ (Price per share of Call/Put).
- Note: When selecting the expiration date of the Call and the Put in forming the straddle position, one can choose the options with the smallest time to expiration (T) among all available options with  $T_{min} \leq T \leq T_{max}$ , where  $T_{min}$  and  $T_{max}$  are pre-specified parameters representing the range of expiration days, say, 60 days to 120 days, respectively. When using ATM implied volatility as signal generator, follow the same rules when selecting option to calculate implied volatility.
- Note: After  $t_{entry}$ , when calculating implied volatility of ATM calls and puts for signal generator, the ATM straddle position we already purchased on  $t_{entry}$  might not be ATM anymore during the trading period, because  $S_t$  keeps changing and ATM option is the option with strike price closest to current underlying price. So for each day after  $t_{entry}$ , we need to find which option is at the money.

## 3. Backtesting and Performance Evaluator

During the entire testing period from  $t_0$  to  $t_N$ , we calculate the annualized daily return of gamma scalping strategy  $R_t$  and evaluate the performance of current parameter setting by the following criteria:

- Total profit/loss
- Average daily return (annualized)
- Sharpe Ratio
- Maximum Drawdowns (MDD)

### **Data and Requirements**

- Data

Same format as midterm. An extra underlying (SPY) is provided.

- Input

There are three ways to pre-specify input parameters: given at the beginning of the program, prompt user to input, read a configuration file. The input parameters are:

- \* General setting: choices of GOOG or SPY,  $t_0$ ,  $t_N$ , K (initially set to 1).
- \* Signal Generator:  $\sigma_{entry}, \sigma_{exit}, w, m$ , choices of historical or implied volatility
- \* Portfolio Rebalancing Engine:  $\bar{\Delta}$ ,  $T_{min}$ ,  $T_{max}$

### - Output

Write all the output in a table form into a csv file "result.csv". The first column of the csv file is the date ranging from  $t_0$  to  $t_N$ .

The header of the table shall contain date, underlying price,  $\sigma_t$ ,  $MA_k(\sigma_t)$ , signal, strike price, time to expiration, call price, put price, call IV, put IV, call Delta, put Delta, call Delta+put Delta, call Gamma, put Gamma, stock shares, Position Delta, Action, Realized profit, Position Total Value, Return, Total Wealth. If ATM implied volatility is selected as signal generator, include ATM call and put price in the header. The row index is from  $t_0$  to  $t_N$ . During the non-trading period, only need to present signal related data.

Output the performance statistics to a csv file "performance.csv". It contains the total profit/loss, average daily return, Sharpe ratio and maximum drawdowns.

### • Pseudo Code

- Divide the data into two periods: in-sample testing period from 0 to  $T_{test}$  and out-of-sample testing period from  $T_{test} + 1$  to  $T_{end}$ .
- Loop: for time t from 0 to  $T_{test}$ 
  - \* In each period t, check the signal for open/exit a gamma scalping trade.
    - $\cdot$  If entry signal appears, open a gamma scalping trade and do the rebalancing in the subsequent periods.
    - · if exit signal appears, close all the options and stocks positions.
  - \* Compute the profit and loss from the rebalancing and record the return in time t.
- End of loop.
- Vary trading input parameters to obtain the best performance of this trading strategy in time period from 0 to  $T_{test}$ .
- Apply the trading parameters to time period  $T_{test} + 1$  to  $T_{end}$  to obtain the out-of-sample performance of the trading strategy with the best parameter set obtained using data in time period from 0 to  $T_{test}$ .

## • Bonus Problem (20 points)

- Collect minute-by-minute option prices and underlying stock prices for GOOG and SPY between 9:30 to 16:00 from 12/1/2015 to 12/10/2015. (This part can be done by Python script in HW5).
- Run your implementation of the Gamma Scalping Strategy over the minute-by-minute data and report results.

## • Deliverables

A final report (in original Word or latex format, do not submit pdf file) is expected and it shall contain the following parts.

- Problem description.
- Details on the implementation framework and code structure description
- Use data prior to 2011-8-31 as in-sample testing data and data from 2011-9-1 to 2011-12-31 as out-sample testing dataset.
- The performance of the trading strategy under the best set of trading parameters over the insample testing period and how the best trading parameter set is obtained.

— Use tables to illustrate the percentage of trades being winning/losing ones, the average profit/loss per trade, the annualized rate of return and the standard deviation of the daily return, and the Sharpe ratio for both the in-sample testing period and the out-of-sample testing period separately. Use figures to show the cumulative growth rate of account value (with starting value being \$1 and the distribution of the daily returns of the account value.

Submit all the program codes/documentations as ONE zip file containing the final report and a sub folder containing all the codes and input configuration file. In each folder, write a readme file to record the compiler and library you use and any other necessary information to run your codes, i.e. how to fill in your parameter configuration file. Use relative path when reading data file.

Also document the estimated time to run your program if you search all combinations of parameters to find the best trading parameters by brute force. If the running time is more than one hour, a relatively small test case of a parameter domain shall be provided which contains the best trading parameter set.

## • Topic 2: Multi-factor Portfolio Trading Strategy (100 points)

Consider the problem of investing a total amount of \$10,000,000 in a large universe of stocks. The Chinese stocks traded on the Shanghai Stock Exchange and Shenzhen Stock Exchange (over 2000 stocks) are chosen to be the entire universe.

Suppose a subset of 1000 (or 1500) stocks are selected from the entire universe. At every rebalancing time point, a small group of 100 to 150 stocks are selected out of the 1000 (or 1500) stocks during the in-sample backtesting period based on the following criterion for long position.

## - Stock selection criteria

- \* Criterion 1: The market capitalization needs to be no less than 500,000,000 (RMB)
- \* Criterion 2: The average daily trading volume over the past 15 business-days needs to be no less than 1,000,000.
- \* Criterion 3: Computing M-score based on a group of factors for each stock i which passes Criterion 1 and 2. Let  $C^i(t)$  denote the price of stock i in period t. The group of factors are described below.
  - · Factor  $F_1$ : Price to Book ratio (PB)
  - · Factor  $F_2$ : Price to CashFlow ratio (PCF)
  - · Factor  $F_3$ : Price to Earning ratio (PE)
  - · Factor  $F_4$ : Price to Sales ratio: PS
  - · Factor  $F_5$ : n-period momentum factor (PM):  $PM^i(t,) = \ln \frac{C^i(t-1)}{C^i(t-n)}$ ; (n = 5)
  - · Factor  $F_6$ : m-period reversion factor (PRev):  $PR^i(t) = \ln \frac{C^i(t-m)}{C^i(t-1)}$ ; (m = 20)
  - · Factor  $F_7$ : L-period log-return volatility (Vol):  $Vol^i(t) = \sigma_L^i(r(t-1))$  where  $\sigma_L^i(r(t-1))$  denotes the annualized standard deviation of the log-return of stock i over time window [t-L,t-1].

Choose a weight vector  $w \equiv (w_1, w_2, \dots, w_7)$ . For example, w = (0, 0, 0, 0, 0.5, 0.5, 0) or w = (1/7, 1/7, 1/7, 1/7, 1/7, 1/7). M-score for stock i is calculated as:

$$M_w^i(t) \equiv w_1 \cdot F_1^i(t) + w_2 \cdot F_2^i(t) + \dots + w_7 \cdot F_7^i(t).$$

### - Portfolio construction

- \* Choose a rebalance frequency to be every U periods. U can be set to a number between 1 and 20. This means while the M-scores are computed for all the periods from  $t = 0, 1, 2, \dots, T_N$ , one can choose to rebalance the portfolio only at time points:  $0, U, 2 * U, 3 * U, \dots$  to avoid incurring excessive transaction costs.
- \* At each time point t, select the best K stocks with highest M-score. Initially, choose K = 100.
- \* Put 1% of total account value at time t into each of the 100 stocks.
- \* Transaction cost is  $TC_{cost} = 0.001$  of total dollar amount of transactions.
- Backtesting and Performance Evaluator During the entire testing period from  $t_0$  to  $t_N$ , calculate the annualized daily return of this factor-based portfolio strategy  $R_t$  and evaluate the performance of a given set of trading parameters by the following criteria:
  - \* Total profit/loss
  - \* Average daily return (annualized)
  - \* Sharpe Ratio
  - \* Maximum Drawdowns (MDD)

### - Data and Implementation Expectation

\* Ticker universe and the fundamental data are provided as csv files in the Resources on T-Squares. Note: some of the tickers in the fundamental dataset have the suffix ".SH" which correspond to the security tickers ends with suffix ".ss" in the ticker universe file.

- \* You are expected to download all the price data from Yahoo finance and save them in ./data folder in your project directory.
- \* Your implementation shall contain at least the following classes.
  - 1. SecurityData Class: contain all the relevant time series data for a given universe of tickers.
  - 2. Strategy Class: generate buy/sell signals for a universe of securities and their corresponding data
  - 3. Portfolio Class: generate portfolio of securities based on the buy/sell signals generated in Strategy Class.

## \* Strategy input parameters:

- · Score weight vector  $w = (w_1, w_2, \dots, w_7)$  for computing the M-score. This can be determined by running cross-sectional regression of daily (or weekly) return over the 7 factors.
- · n value for PM, m value for PR and L value for Vol.
- $\cdot$  Frequency U at which the portfolio is rebalanced.

#### - Pseudo Code

- \* Divide the data into two periods: in-sample testing period from 0 to  $T_{test}$  and out-of-sample testing period from  $T_{test} + 1$  to  $T_{end}$ .
- \* Loop: for time t from 0 to  $T_{test}$ 
  - · In each period t, apply the stock selection criteria to get a subset of  $N_t$  securities ( $N_t$  can vary with t).
  - · Compute the M-score for the  $N_t$  securities.
  - · Form the portfolio based on the strategy rule if t = 0. If t > 0 and t is a rebalancing time point, rebalance the portfolio based on the strategy rule (namely, buy stocks which become members of the top 100 M-score group and sell stocks which fall out of the top 100 M-score group.
  - $\cdot$  Compute the profit and loss from the rebalancing and record the return in time t.
- \* End of loop.
- \* Vary trading input parameters to obtain the best performance of this trading strategy in time period from 0 to  $T_{test}$ .
- \* Apply the trading parameters to time period  $T_{test} + 1$  to  $T_{end}$  to obtain the out-of-sample performance of the trading strategy with the best parameter set obtained using data in time period from 0 to  $T_{test}$ .

## - Bonus Problem (20 points)

- \* At every rebalance time t, use the values of 000300.ss as market index in the past 100 days to compute the beta of each stock with respect to this market index using the CAPM model.
- \* Calculate the total beta of the 100 stocks with the highest M-scores at time t, denoted by  $\beta_{port}(t)$ . Add a short position of  $-1*\beta_{port}(t)$  shares of 000300.ss to the portfolio of 100 stocks so that the overall portfolio has a 0-beta with respect to 000300.ss. Report the performance of this new portfolio and compare it with the original portfolio.

## - Deliverables

A final report (submit original Word or latex files, donnot submit pdf file) shall be submitted. It shall contain the following parts.

- \* Problem description.
- \* Details on the implementation framework and code structure description
- \* Use data prior to 2014-10-31 as in-sample testing data and data from 2014-11-1 to 2015-7-31 as out-sample-test dataset.
- \* The performance of the trading strategy under the best set of **strategy input parameters** and how the best trading parameter set is obtained.

\* Use tables to illustrate the percentage of trades being winning/losing ones, the average profit/loss per trade, the annualized rate of return and the standard deviation of the daily return, and the Sharpe ratio for both in-sample testing period and the out-of-sample testing period separately. Use figures to show the cumulative growth rate of account value (with starting value being \$1 and the distribution of the daily returns of the account value.

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Also document the estimated time to run your program if you search all combinations of parameters to find the best trading parameters by brute force. If the running time is more than one hour, provide a relatively small test case of a parameter domain which contains the best trading parameter set.