# **Financial Information Systems Final Project Questions**

# **Part A Questions**

### **Function Creation**

Write a function `runMovingAverageAndBB` that:

- Takes five inputs: (a) a DataFrame with price data for one instrument, (b) `fastWindow` and `slowWindow` for the Moving Average Crossover, and (c) `bbWindow` and `stdevBand` for Bollinger Bands.
- Runs the Moving Average Crossover and Bollinger Bands strategies on the input data and combines the results in a single DataFrame with daily signals for both strategies.

## **Data Loading**

Load the file `PricesProjectA.csv`, which contains daily closing prices from 1999 to 2018 for:

- Six equities: AAPL, AMZN, ATT, GE, INTC, SPY
- Two fixed income instruments: FBNDX, LEHM
- Two commodities: GOLD, SILVER
- Two currency pairs: AUD, EUR

### **Instrument Selection**

Select 8 instruments for your portfolio as follows:

- Five of the six equities, one fixed income instrument, one commodity, and one currency pair.
- For each selected instrument, choose one of the following strategies: Moving Average Crossover, Bollinger Bands, or a benchmark that tracks daily returns.

### **Parameter Specification**

Specify the parameters ('fastWindow', 'slowWindow', 'bbWindow', 'stdevBand') for each strategy and list these in a Markdown cell for reproducibility.

### **Parameter Code Block**

Include five lines of code at the top of the notebook with chosen parameter values for easy replication, e.g.,

## **Run Strategies**

Run `runMovingAverageAndBB` on all 12 instruments using the chosen parameters, and combine the results into a master DataFrame with 5 columns per instrument.

# **Strategy Rationale**

Provide reasoning for your chosen parameters and strategies for each instrument, focusing on achieving a high Sharpe ratio and balanced risk-return.

### **Subset Data**

Create a DataFrame `myStrategies` with daily returns of only the 8 selected strategies for further analysis.

#### **Correlation Matrix**

Calculate and analyze the correlation matrix for `myStrategies`, noting any low correlations that indicate good diversification.

# **Equal-Weighted Portfolio Sharpe**

Calculate the annualized Sharpe ratio of an equal-weighted portfolio using the 'myStrategies' DataFrame, comparing it to each individual strategy's performance.

# **Mean-Variance Optimization (MVO)**

Run a 5,000-step Monte Carlo simulation to find the optimal weights for maximizing the portfolio's Sharpe ratio.

# Risk/Return Plot

Plot a risk-return scatter plot of all simulated portfolios, highlighting the portfolio with the highest Sharpe ratio.

## **Part B Questions**

## **Data Loading for Extended Period**

Load the new data file, which contains extended closing price data for all the same instruments, covering the period from 1999 through 2022.

### **Parameter Code Block for Part B**

Copy the same five lines of parameter code from Part A for consistency and reproducibility.

# **Run Strategies for Full Period**

Re-run your chosen portfolio strategies with the same parameters over the full period from 1999 through 2022, and create a DataFrame with daily returns through the extended date range.

# **Portfolio Performance Evaluation**

Using the MVO weights from Part A, calculate performance statistics of the maximum-Sharpe portfolio for two periods: in-sample (1999-2018) and out-of-sample (2019-2022). Compare the in-sample Sharpe with the out-of-sample Sharpe.

# **Passive Benchmark Comparison**

Construct an equal-weighted portfolio of the benchmark versions (base returns) of your chosen instruments for the out-of-sample period and calculate its summary statistics.

# **Performance Commentary**

Discuss how your active portfolio performed out-of-sample compared to the passive benchmark. Note whether it outperformed the benchmark and any insights or lessons learned from the results.