King Saud University College of Computer and Information Science Computer Science Department 2nd Semester - 1445

Optimal Portfolio Allocation CSC311 project implementation report

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Brute Force Pseudocode

```
function findOptimalAllocationBF(assets, totalInvestment, riskTolerance):
   optimalPortfolio = null
   maxReturn = 0
   portfolio = Portfolio(0, 0, 0)
   for i = 0 to 100:
                                      // 0% to 100%
       for j = 0 to (100 - i):
                                     // 0% to 100 - i%
           remaining = 100 - i - j // remaining%
           portfolio = Portfolio(i, j, remaining)
           if ((i / 100.0) * totalInvestment <= assets[0].units</pre>
               && (j / 100.0) * totalInvestment <= assets[1].units</pre>
               && (remaining / 100.0) * totalInvestment <= assets[2].units):
               portfolio = Portfolio(i, j, remaining)
               portfolio.calculatePortfolioEfficiency(assets)
               if (portfolio.risk <= riskTolerance № portfolio.expectedReturn > maxReturn):
                   maxReturn = portfolio.expectedReturn
                   optimalPortfolio = portfolio
   return optimalPortfolio
```

Explanation

Our code consists of two classes: Asset and Portfolio. The Asset class represents individual assets and stores information related to each asset. Also there is the Portfolio class which represents a client's investment portfolio and contains an allocation array for his assets, expected return, and risk.

In the main method, the code reads asset information from the file; each line of the file corresponds to an asset, and the information is split and stored in the assets list. Note that the last two lines of our input file are the total investment amount and the risk tolerance values.

- 1. We initialize variables optimalPortfolio and maxReturn to store the best portfolio found so far. Create a Portfolio object to represent different allocation combinations. To allocate, Use nested loops to iterate through all possible asset allocation percentages; the outer loop iterates from 0% to 100%, which is the allocation percentage for the first asset, and the inner loop iterates from 0% to (100 i)% second asset allocation percentage, then it calculates the remaining percentage for the third asset after allocating percentages for the first two assets.
- 2. Create a Portfolio object with the current allocation percentages, then validate:
 - Check if the units allocated to each asset don't exceed the available units.
 - If the allocations are valid:

- Calculate the portfolio's return and risk using the calculatePortfolioEfficiency function.
- Check if the portfolio's risk is within the specified tolerance and if its expected return is higher than the best return found so far.
- 3. If the portfolio meets the risk tolerance & has a higher expected return than the current maximum, update maxReturn and optimalPortfolio with the current portfolio's allocation values.
- 4. After all allocations are checked, return the optimal portfolio found, which represents the best(best in highest return context) combination of assets within the risk tolerance. If no valid allocation is found, optimalPortfolio remains null.

Sample run on the provided cases

```
Optimal Allocation:

AAPL: 300 units (30% of investment)
GOOGL: 500 units (50% of investment)
MSFT: 200 units (20% of investment)
Expected Portfolio Return: 0.0630
Portfolio Risk Level: 0.0240
macbook@Lameess-MacBook-Air Algorithms-
```

Source Code

```
public class InvestmentFirm {
    private static class Asset {
        String id;
        double expectedReturn;
        double individualrisk;
        double units;

    public Asset(String id, double expectedReturn, double risk, double units) {
            this.id = id;
            this.expectedReturn = expectedReturn;
            this.individualrisk = risk;
            this.units = units;
        }
    }

// Portfolio class representing each client
private static class Portfolio {
        double() allocation; // Percentage how much of each asset in total investment
        double expectedReturn;
        double risk;

public Portfolio(int asset1, int asset2, int asset3) {
        allocation = new double[3];
        allocation[0] = asset1 / 100.0;
        allocation[1] = asset2 / 100.0;
        allocation[2] = asset3 / 100.0;
    }
}
```

```
Scanner scanner = new Scanner(file);
   while (scanner.hasNextLine()) {
       String line = scanner.nextLine();
Portfolio optimalPortfolio = findOptimalAllocationBF(assets, totalInvestment, riskTolerance);
```

Dynamic Programming Pseudocode

```
function findOptimalAllocationDP(assets, totalInvestment, riskTolerance):
   optimalPortfolio = null
   maxReturn = 0
   dp[101][101][101]
   for i from 0 to 100:
       for j from 0 to (100 - i):
           remaining = 100 - i - j
           if ((i / 100.0) * totalInvestment <= assets[0].units &&
                    (j / 100.0) * totalInvestment <= assets[1].units &&
                    (remaining / 100.0) * totalInvestment <= assets[2].units):</pre>
                portfolio = Portfolio(i, j, remaining)
                if i > 0 and dp[i-1][j][remaining] is not null:
                   portfolio.expectedReturn += dp[i - 1][j][remaining].expectedReturn
                   portfolio.risk += dp[i - 1][j][remaining].risk
                if j > 0 and dp[i][j-1][remaining] is not null:
                    portfolio.expectedReturn += dp[i][j - 1][remaining].expectedReturn
                   portfolio.risk += dp[i][j - 1][remaining].risk
                if remaining > 0 and dp[i][j][remaining-1] is not null:
                   portfolio.expectedReturn += dp[i][j][remaining - 1].expectedReturn
                    portfolio.risk += dp[i][j][remaining - 1].risk
                portfolio.calculatePortfolioEfficiency(assets)
                if portfolio.risk <= riskTolerance and portfolio.expectedReturn > maxReturn:
                   maxReturn = portfolio.expectedReturn
                   optimalPortfolio = portfolio
                dp[i][j][remaining] = portfolio
   return optimalPortfolio
```

Explanation

dp here is a 3D array used for memoization. It stores precalculated portfolio information for different combinations of asset percentages. (Size: 101x101x101 to represent 0% to 100% each asset. A Portfolio object is created with the current weight distribution (i, j, remaining) but the algorithm checks for previously calculated portfolios in the dp array for similar percentage combinations, and if a valid portfolio exists, it adds its expected return and risk to the current portfolio instead of recalculating. This makes use of previously computed values to avoid redundant calculations. Then use calculatePortfolioEfficiency function to calculate the current portfolio's expected return and risk. & If the current portfolio's risk falls within the tolerance and its expected return is higher than maxReturn, maxReturn gets updated. optimalPortfolio is replaced with the current portfolio.

The final step is the crucial step of dynamic programming. The algorithm stores the current portfolio information in the corresponding dp array slot for future reference. This allows the algorithm to avoid recalculating the same portfolio multiple times. Finally, the function returns the optimalPortfolio containing the best asset allocation found.

Recurrence relation

```
F(ij,k) = \begin{cases} & \text{null} & \text{if } (i/100.0) \times \text{totalInvestment} \times \text{assets}[0].\text{units} \vee (j/100.0) \times \text{totalInvestment} \times \text{assets}[1].\text{units} \vee (k/100.0) \times \text{totalInvestment} \times \text{assets}[2].\text{units} \\ & \text{Portfolio}(i, j, k) + \text{dp}[i-1][j][k] + \text{dp}[i][j-1][k] + \text{dp}[i][j][k-1] \end{cases} \quad \text{otherwise}
```

Sample run on the provided cases

```
Optimal Allocation:

AAPL: 300 units (30% of investment)
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Portfolio Risk Level: 0.0240
macbook@Lameess-MacBook-Air Algorithms-
```

Source Code

```
import java.io.*;
import java.util.*;

public class InvestmentFirm {
    private static class Asset {
        String id;
        double expectedReturn;
        double individualrisk;
        double units;

        public Asset(String id, double expectedReturn, double risk, double units) {
            this.id = id;
            this.expectedReturn = expectedReturn;
            this.individualrisk = risk;
            this.units = units;
        }
    }
}

// Portfolio class representing each client
private static class Fortfolio {
```

```
File file = new File("Example1.txt");
```

```
Portfolio optimalPortfolio = findOptimalAllocationDP(assets, totalInvestment,
```

```
if (i > 0&&(dp[i-1][j][remaining])!=null) {
if (j > 0&&(dp[i][j-1][remaining])!=null) {
if (remaining > 0&&(dp[i][j][remaining-1])!=null) {
   portfolio.expectedReturn += dp[i][j][remaining - 1].expectedReturn;
portfolio.calculatePortfolioEfficiency(assets); //calculate portfolio return and
```

```
dp[i][j][remaining] = portfolio; // Store the optimal portfolio for current
asset percentages

}

return optimalPortfolio;
}
```

Algorithm	Best Case Complexit y	Example	Worst Case Complexity	Example
Brute Force Allocation	if n=3 O(100^n-1) When we have N assets: O(1)	When considering 3 assets as in implementation: The best case is 100^2 Going through 2 loops giving each asset a different precent every time and the remaining will be assigned to the 3 rd asset. When we have N assets: The best case is that we have only 1 asset to allocate the algorithm will give it 100% directly therefore; O(1)	O(100^n-1) n=3 When we have N assets: O(100^n-1)	When considering 3 assets as in our implementation: It'll be like the best case O(100^2) When we have N assets: The worst case will be O(100^n-1) Let n be number of assets Since every percent given to an asset is derived from the asset before it we will iterate through a new loop to allocate every asset and the remaining will go to the last asset; basically if we have 8 assets we'll have 7 loops and the last asset will take the remaining percent
Dynamic Programmin Allocation	ng O(1)	The best-case scenario occurs when the risk tolerance is exceeded by the very first portfolio that satisfies the constraints. In this case, the algorithm would terminate early after checking only a few portfolios. The best-case time complexity would be O(1), as it finds the optimal portfolio quickly without fully exploring all possible allocations	O(100^n-1) N bieng num of assets In our code it is n=3 Therefore; O(100^3-1)	The worst-case scenario happens when the algorithm needs to explore all possible allocations within the given constraints to find the optimal portfolio. This occurs when the risk tolerance is very low, and the algorithm has to iterate through all possible combinations of asset allocations. In this case, the worst-case time complexity is $O(100^n-1)$, where $n = num$ aof assets in our code for simplicty we assumed only 3 assets but if there were 10 assets this means we will have 9 loops

Challenges

Understanding how to solve the problem using brute force was tough since we had to deal with inconsistent amounts of assets, which made things tricky. To handle this, we thought about splitng the final allocation into a sum of percentages of each asset. Another challenge we faced was figuring out the best and worst cases and coming up with different situations to test the solution.

The main challenge we faced lay in considering many aspects to make a decision all at once, so we fixated on some details while failing to consider others. We overcame this challenge with time, when we became thoroughly familiar with the project.

Another challenge we faced was in teamwork where communication and collaboration can be challenging and team members may feel unmotivated or disconnected from the project, leading to a lack of participation and effort.

Overall, solving this problem needed careful planning and a good understanding of how to approach it step by step.

Peer Evaluation

Part 1: Team Work					
Criteria	Student1	Student2	Student3	Student4	
Work division: Contributed equally to the work	1	1	1	-	
Student succeeds in smoothly forming /joining group within time	1	1	1	-	
Peer evaluation: Level of commitments (Interactivity with other team members), and professional behavior towards team & TA	1	1	1	-	
Project Discussion: Accurate answers, understanding of the presented work, good listeners to questions					
Time management: Attending on time, being ready to start the demo, good time management in discussion and demo.					
Total/3					

Part 2: Functional Requirements		
	Criteria	Evaluation
	Overall quality of the code implementation (organization,	
General	clearness, design,)	
	Complete report with well organized sections	
	Algorithm description (pseudo-code with explanation)	
Brute Force	Time and space complexity	
	Implementation correctness + sample run	
	Algorithm description (pseudo-code with explanation)	
Dynamic programming	Time and space complexity	
	Implementation correctness + sample run	
	Total/2.5	