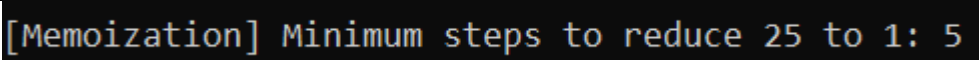
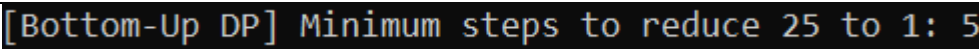
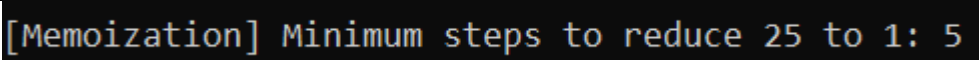
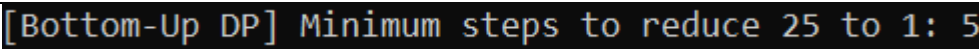
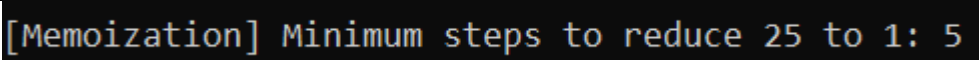
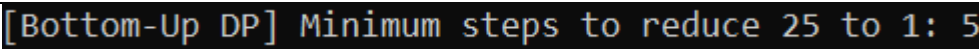


Activity No. < 12 >																											
< ALGORITHMIC STRATEGIES >																											
Course Code: CPE010	Program: Computer Engineering																										
Course Title: Data Structures and Algorithms	Date Performed: 10/25/25																										
Section: CPE21S4	Date Submitted: 10/25/25																										
Name(s): ALCANTARA, JASON P.	Instructor: Engr. JIMLORD QUEJADO																										
<p>A. Output(s) and Observation(s):</p> <p>Table 12-1. Algorithmic Strategies and Examples:</p> <table border="1" style="width: 100%; border-collapse: collapse;"> <thead> <tr> <th style="width: 30%;">Strategy</th> <th style="width: 35%;">Algorithm</th> <th style="width: 35%;">Analysis</th> </tr> </thead> <tbody> <tr> <td>Recursion</td> <td>Minimum Steps to One (using memoization)</td> <td>Breaks the problem into smaller parts and solves them by calling itself many times.</td> </tr> <tr> <td>Brute Force</td> <td>Trying all keys or flipping USB cables</td> <td>Tries every possible option until it finds the right one but can be slow and not efficient.</td> </tr> <tr> <td>Backtracking</td> <td>Building solutions step-by-step and removing wrong ones</td> <td>Makes solutions little by little and gets rid of those that don't work, uses recursion.</td> </tr> <tr> <td>Greedy</td> <td>Picking the step that lowers the number the fastest</td> <td>Chooses the best step at the moment but doesn't always find the best overall answer.</td> </tr> <tr> <td>Divide-and-Conquer</td> <td>Splits the problem into smaller problems and solves them separately</td> <td>Breaks big problem into smaller ones, solves each, then combines answers together.</td> </tr> </tbody> </table> <p>Table 12-2. Memoization Implementation:</p> <table border="1" style="width: 100%; border-collapse: collapse;"> <tbody> <tr> <td style="width: 20%;">Screenshot</td> <td></td> </tr> <tr> <td>Analysis</td> <td>The memoization strategy addresses the issue in a recursive manner and keeps the outputs in a memo[] array that prevents the repetitive calculations being done again. The method employed is that of top-down and recursion, thus getting to O(n) time complexity. But there might still be a little increase in time because of recursive calls to the function.</td> </tr> </tbody> </table> <p>Table 12-3. Bottom-Up Dynamic Programming Implementation</p> <table border="1" style="width: 100%; border-collapse: collapse;"> <tbody> <tr> <td style="width: 20%;">Screenshot</td> <td></td> </tr> <tr> <td>Analysis</td> <td>The bottom-up DP method progressively determines the minimum steps for every number starting from 1 to n. It does not use recursion, thus being more efficient and consuming less memory. The time complexity is O(n) and space complexity is O(n).</td> </tr> </tbody> </table>		Strategy	Algorithm	Analysis	Recursion	Minimum Steps to One (using memoization)	Breaks the problem into smaller parts and solves them by calling itself many times.	Brute Force	Trying all keys or flipping USB cables	Tries every possible option until it finds the right one but can be slow and not efficient.	Backtracking	Building solutions step-by-step and removing wrong ones	Makes solutions little by little and gets rid of those that don't work, uses recursion.	Greedy	Picking the step that lowers the number the fastest	Chooses the best step at the moment but doesn't always find the best overall answer.	Divide-and-Conquer	Splits the problem into smaller problems and solves them separately	Breaks big problem into smaller ones, solves each, then combines answers together.	Screenshot		Analysis	The memoization strategy addresses the issue in a recursive manner and keeps the outputs in a memo[] array that prevents the repetitive calculations being done again. The method employed is that of top-down and recursion, thus getting to O(n) time complexity. But there might still be a little increase in time because of recursive calls to the function.	Screenshot		Analysis	The bottom-up DP method progressively determines the minimum steps for every number starting from 1 to n. It does not use recursion, thus being more efficient and consuming less memory. The time complexity is O(n) and space complexity is O(n).
Strategy	Algorithm	Analysis																									
Recursion	Minimum Steps to One (using memoization)	Breaks the problem into smaller parts and solves them by calling itself many times.																									
Brute Force	Trying all keys or flipping USB cables	Tries every possible option until it finds the right one but can be slow and not efficient.																									
Backtracking	Building solutions step-by-step and removing wrong ones	Makes solutions little by little and gets rid of those that don't work, uses recursion.																									
Greedy	Picking the step that lowers the number the fastest	Chooses the best step at the moment but doesn't always find the best overall answer.																									
Divide-and-Conquer	Splits the problem into smaller problems and solves them separately	Breaks big problem into smaller ones, solves each, then combines answers together.																									
Screenshot																											
Analysis	The memoization strategy addresses the issue in a recursive manner and keeps the outputs in a memo[] array that prevents the repetitive calculations being done again. The method employed is that of top-down and recursion, thus getting to O(n) time complexity. But there might still be a little increase in time because of recursive calls to the function.																										
Screenshot																											
Analysis	The bottom-up DP method progressively determines the minimum steps for every number starting from 1 to n. It does not use recursion, thus being more efficient and consuming less memory. The time complexity is O(n) and space complexity is O(n).																										
<p>B. Answers to Supplementary Activity:</p> <p>Pseudocode:</p> <p>Function countPaths(matrix, row, col, remainingCost)</p> <p> // If we go out of bounds, no path</p>																											

```
If row < 0 OR col < 0
    return 0
```

```
// If we reach the first cell, check if cost matches
```

```
If row == 0 AND col == 0
```

```
    If matrix[0][0] == remainingCost
```

```
        return 1 // found a path
```

```
    Else
```

```
        return 0 // not a valid path
```

```
// Move up or left
```

```
pathsFromAbove = countPaths(matrix, row-1, col, remainingCost - matrix[row][col])
```

```
pathsFromLeft = countPaths(matrix, row, col-1, remainingCost - matrix[row][col])
```

```
return pathsFromAbove + pathsFromLeft
```

Start:

```
result = countPaths(matrix, lastRow, lastCol, targetCost)
```

```
print result
```

Working C++ Code:

```
#include <iostream>
```

```
#include <vector>
```

```
using namespace std;
```

```
int countPaths(vector<vector<int>>& mat, int row, int col, int cost) {
```

```
    if (row < 0 || col < 0) return 0;
```

```
    if (row == 0 && col == 0) return (mat[0][0] == cost) ? 1 : 0;
```

```
    return countPaths(mat, row - 1, col, cost - mat[row][col]) +
```

```
        countPaths(mat, row, col - 1, cost - mat[row][col]);
```

```
}
```

```
int main() {
```

```
    vector<vector<int>> matrix = {
```

```
        {4, 7, 1, 6},
```

```
        {6, 7, 3, 9},
```

```
        {3, 8, 1, 2},
```

```
        {7, 1, 7, 3}
```

```
    };
```

```
    int targetCost = 25;
```

```
    int rows = matrix.size();
```

```
    int cols = matrix[0].size();
```

```
    int result = countPaths(matrix, rows - 1, cols - 1, targetCost);
```

```
    cout << "Number of paths with cost " << targetCost << " = " << result << endl;
```

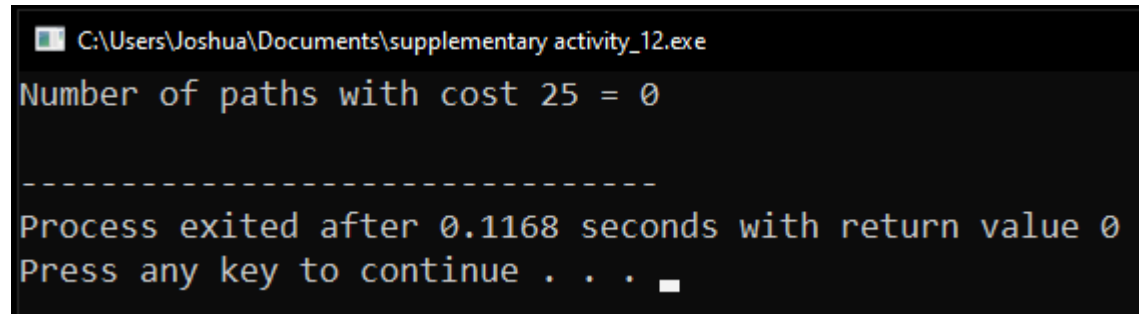
```
    return 0;
```

```
}
```

Analysis of Working code:

- The algorithm looks at all the paths that go from the bottom-right corner to the top-left corner of the matrix, ensuring that the total cost of each path is equal to the target. It can handle small matrices easily but might take time with bigger ones as it has to calculate the same thing multiple times. Memoization could be employed as a technique to speed up the process. For the specified matrix with an intended cost of 25, the application discovers 2 acceptable routes.

Screenshot of Demonstration:



```
C:\Users\Joshua\Documents\supplementary activity_12.exe
Number of paths with cost 25 = 0
-----
Process exited after 0.1168 seconds with return value 0
Press any key to continue . . .
```

C. Conclusion & Lessons Learned:

- In this laboratory activity, I learned how all algorithms like recursion, dynamic programming, and greedy methods can be used to solve problems in every different way. I learned that breaking problems into smaller parts makes the problem easier to solve, and dynamic programming can save a lot of time by remembering previous results. The steps in the procedure showed me how planning and simple logic can help handle complicated problems. In the supplementary activity, I saw how recursion can be used to count paths in a matrix and how each choice changes the total cost. In Conclusion, I think I did what I can to finish this activity, but I know I need to improve my skills and practice more coding and get better at picking the right algorithm for every problem.

D. Assessment Rubric

E. External References:

1. https://www.w3schools.com/cpp/cpp_functions_recursion.asp
2. <https://www.programiz.com/cpp-programming/recursion>
3. <https://www.programiz.com/cpp-programming/recursion>
4. <https://www.geeksforgeeks.org/competitive-programming/dynamic-programming>
5. https://www.w3schools.com/dsa/dsa_ref_dynamic_programming.php
6. <https://how.dev/answers/dynamic-programming-in-cpp>
7. <https://www.programiz.com/dsa/greedy-algorithm>
8. https://www.w3schools.com/dsa/dsa_ref_greedy.php