

Algorithm Class Assignment 3 Dynamic Programming

Question 1

Explain Why The Problem Is (Or Not) Good For Dynamic Programming

The problem is good for dynamic programming because for calculating the biggest total preference value, we will need to consider a lot of probabilities and there are conditions that recalculate the result gained before. By implementing dynamic programming, we can reduce the redundant steps to improve efficiency.

Design & Implement The Algorithm For The Problem

Design & Implement:

```

1  #include <iostream>
2  using namespace std;
3
4  // find the bigger integer between the two
5  int Maximum(int a, int b) {
6      if (a > b) {
7          return a;
8      }
9
10     else {
11         return b;
12     }
13 }
14
15 // bottom up method dynamic programming
16 // find the maximum value of preference that can be accomplished under the bud
17 int Preference(int budget, int item_price[], int item_preference[], int items)
18     int i;
19     int j;
20     int K[items + 1][budget + 1];
21
22     for(i = 0; i <= items; i++) {
23         for(j = 0; j <= budget; j++) {
24             if (i == 0 || j == 0) {
25                 K[i][j] = 0;
26             }
27
28             else if (item_price[i - 1] <= j) {
29                 K[i][j] = Maximum(item_preference[i - 1] + K[i - 1][j - item_p
30             }
31
32             else {
33                 K[i][j] = K[i - 1][j];
34             }
35         }
36     }
37
38     return K[items][budget];
39 }
40
41 int main() {
42     // sample item preference list
43     int item_preference[] = {80, 28, 70, 55, 43};
44     // sample item price list
45     int item_price[] = {40, 20, 60, 75, 59};
46     // sample budget
47     int budget = 150;
48     // amount of items
49     int items = sizeof (item_preference) / sizeof (item_preference[0]);
50
51     cout << Preference(budget, item_price, item_preference, items);
52
53     return 0;
54 }

```

Analyze The Complexity Of The Algorithm

Time Complexity:

- time complexity for the implementation is $O(n * m)$

'n' to be the amount of the items

'm' to be the total budget

Space Complexity:

- space complexity for the implementation is $O(n * m)$

'n * m' is the size of a two-dimensional array

Question 2

Explain Why The Problem Is (Or Not) Good For Dynamic Programming

Dynamic programming is good for this question is because the result of the problems is related. By using brute force, we will spend a lot of time wasting on same calculation, which dynamic programming resolve this problem.

Design & Implement The Algorithm For The Problem

Design & Implement:

```

1  #include <iostream>
2  using namespace std;
3
4  // find the larger value
5  int Maximum(int a, int b) {
6      if (a > b) {
7          return a;
8      } else {
9          return b;
10     }
11 }
12
13 // find the maximum earnings for the government
14 // accomplished by the limited shore range
15 // bottom up method dynamic programming
16 int Earnings(int length, int section[], int price[], int amount) {
17     int i;
18     int j;
19     int S[amount + 1][length + 1];
20
21     for(i = 0; i <= amount; i++) {
22         for(j = 0; j <= length; j++) {
23             if (i == 0 || j == 0) {
24                 S[i][j] = 0;
25             }
26
27             else if (section[i - 1] <= j) {
28                 S[i][j] = Maximum(price[i - 1] +
29                                 S[i - 1][j - section[i - 1]],
30                                 S[i - 1][j]);
31             }
32
33             else {
34                 S[i][j] = S[i - 1][j];
35             }
36         }
37     }
38
39     return S[amount][length];
40 }
41
42 int main() {
43     // price list for different shore section
44     int price[] = {15, 25, 10, 20, 5};
45     // partition for different shore section
46     int section[] = {27, 22, 19, 15, 31};
47     // total length of the shore
48     int length = 50;
49     // amount of the partitions
50     int amount = sizeof (price) / sizeof (price[0]);
51
52     cout << Earnings(length, section, price, amount);
53
54     return 0;
55 }

```

Analyze The Complexity Of The Algorithm

Time Complexity:

- time complexity for the implementation is $O(n * m)$

'n' to be the total partitions we have split

'm' to be the length of the shore

Space Complexity:

- space complexity for the implementation is $O(n * m)$

'n * m' is the size of a two-dimensional array

Question 3

Textbook 3.4

Binomial Theorem Using One-Dimensional Array:

```
1 // Reference For nCk:
2 //
3 // nCk = n! / [k! * (n - k)!]
4 //      = [n * (n - 1) * ... * (n - k + 1)] / k!
5
6 int binomial_theorem_with_1D_array (int n, int k) {
7     index i;
8     int result[k + 1];
9
10    for (i = 1; i <= k; i++) {
11        result[i] = ((n + 1 - i) * result[i - 1]) / i;
12    }
13
14    int answer = result[k];
15
16    return answer;
17 }
```

Textbook 3.5

D1: (START)

	1	2	3	4	5	6	7
1	0	4	INF	INF	INF	10	INF
2	3	0	INF	18	INF	INF	INF
3	INF	6	0	INF	INF	INF	INF
4	INF	5	15	0	2	19	5

5	INF	INF	12	1	0	INF	INF
6	INF	INF	INF	INF	INF	0	10
7	INF	INF	INF	8	INF	INF	0

P1: (START)

	1	2	3	4	5	6	7
1	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0
7	0	0	0	0	0	0	0

D2:

	1	2	3	4	5	6	7
1	0	4	INF	INF	INF	10	INF
2	3	0	INF	18	INF	13	INF
3	INF	6	0	INF	INF	INF	INF
4	INF	5	15	0	2	19	5
5	INF	INF	12	1	0	0	INF
6	INF	INF	INF	INF	INF	0	10
7	INF	INF	INF	8	INF	INF	0

P2:

	1	2	3	4	5	6	7
1	0	0	0	0	0	0	0
2	0	0	0	0	0	1	0
3	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0

7	0	0	0	0	0	0	0
---	---	---	---	---	---	---	---

D3:

	1	2	3	4	5	6	7
1	0	4	INF	22	INF	10	INF
2	3	0	INF	18	INF	13	INF
3	9	6	0	24	INF	19	INF
4	8	5	15	0	2	18	5
5	INF	INF	12	1	0	INF	INF
6	INF	INF	INF	INF	INF	0	10
7	INF	INF	INF	8	INF	INF	0

P3:

	1	2	3	4	5	6	7
1	0	0	0	2	0	0	0
2	0	0	0	0	0	1	0
3	2	0	0	2	0	2	0
4	2	0	0	0	0	2	0
5	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0
7	0	0	0	0	0	0	0

D4:

	1	2	3	4	5	6	7
1	0	4	INF	22	INF	10	INF
2	3	0	INF	18	INF	13	INF
3	9	6	0	24	INF	19	INF
4	8	5	15	0	2	18	5
5	21	18	12	1	0	31	INF
6	INF	INF	INF	INF	INF	0	10
7	INF	INF	INF	8	INF	INF	0

P4:

	1	2	3	4	5	6	7
1	0	0	0	2	0	0	0
2	0	0	0	0	0	1	0
3	2	0	0	2	0	2	0
4	2	0	0	0	0	2	0
5	3	3	0	0	0	3	0
6	0	0	0	0	0	0	0
7	0	0	0	0	0	0	0

D5:

	1	2	3	4	5	6	7
1	0	4	37	22	24	10	27
2	3	0	33	18	20	13	23
3	9	6	0	24	26	19	29
4	8	5	15	0	2	18	5
5	9	6	12	1	0	19	6
6	INF	INF	INF	INF	INF	0	10
7	16	13	23	8	10	26	0

P5:

	1	2	3	4	5	6	7
1	0	0	4	2	4	0	4
2	0	0	4	0	4	1	4
3	2	0	0	2	4	2	4
4	2	0	0	0	0	2	0
5	4	4	0	0	0	4	4
6	0	0	0	0	0	0	0
7	4	4	4	0	4	4	0

D6:

--	--	--	--	--	--	--	--

	1	2	3	4	5	6	7
1	0	4	36	22	24	10	27
2	3	0	32	18	20	13	23
3	9	6	0	24	26	19	29
4	8	5	14	0	2	18	5
5	9	6	12	1	0	19	6
6	INF	INF	INF	INF	INF	0	10
7	16	13	22	8	10	26	0

P6:

	1	2	3	4	5	6	7
1	0	0	5	2	4	0	4
2	0	0	5	0	4	1	4
3	2	0	0	2	4	2	4
4	2	0	5	0	0	2	0
5	4	4	0	0	0	4	4
6	0	0	0	0	0	0	0
7	4	4	5	0	4	4	0

D7:

	1	2	3	4	5	6	7
1	0	4	36	22	24	10	20
2	3	0	32	18	20	13	23
3	9	6	0	24	26	19	29
4	8	5	14	0	2	18	5
5	9	6	12	1	0	19	6
6	INF	INF	INF	INF	INF	0	10
7	16	13	22	8	0	26	0

P7:

	1	2	3	4	5	6	7
1	0	0	5	2	4	0	6

2	0	0	5	0	4	1	4
3	2	0	0	2	4	2	4
4	2	0	5	0	0	2	0
5	4	4	0	0	0	4	4
6	0	0	0	0	0	0	0
7	4	4	5	0	4	4	0

D8: (RESULT)

	1	2	3	4	5	6	7
1	0	4	36	22	24	10	20
2	3	0	32	18	20	13	23
3	9	6	0	24	26	19	29
4	8	5	14	0	2	18	5
5	9	6	12	1	0	19	6
6	26	23	32	18	20	0	10
7	16	13	22	8	10	26	0

P8: (RESULT)

	1	2	3	4	5	6	7
1	0	0	5	2	4	0	6
2	0	0	5	0	4	1	4
3	2	0	0	2	4	2	4
4	2	0	5	0	0	2	0
5	4	4	0	0	0	4	4
6	7	7	7	7	7	0	0
7	4	4	5	0	4	4	0

Textbook 3.6

P:

	1	2	3	4	5	6	7
1	0	0	5	2	4	0	6

2	0	0	5	0	4	1	4
3	2	0	0	2	4	2	4
4	2	0	5	0	0	2	0
5	4	4	0	0	0	4	4
6	7	7	7	7	7	0	0
7	4	4	5	0	4	4	0

Operation:

Find the shortest path from v7 to v3:

- path(7, 3)
- $P[7][3] = 5$
- path(7, $P[7][3]$)
 - path(7, 5)
 - $P[7][5] = 4$
 - path(7, $P[7][4]$)
 - path(7, 4)
 - $P[7][4] = 0$
 - mid point “v4”
 - path($P[7][5]$, 5)
 - path(4, 5)
 - $P[4][5] = 0$
- mid point “v5”
- path(5, 3)
- $P[5][3] = 0$

Result:

v7 -> v4 -> v5 -> v3

Textbook 3.13

Implementation:

```

1  Zero Interval:
2
3  M[1][1] = 0
4  M[2][2] = 0
5  M[3][3] = 0
6  M[4][4] = 0
7  M[5][5] = 0
8  M[6][6] = 0
9
10 One Interval:
11
12 M[1][2] = M[1][1] + M[2][2] + d0 * d1 * d2 = 200
13 M[2][3] = M[2][2] + M[3][3] + d1 * d2 * d3 = 400
14 M[3][4] = M[3][3] + M[4][4] + d2 * d3 * d4 = 200
15 M[4][5] = M[4][4] + M[5][5] + d3 * d4 * d5 = 2000
16
17 Two Interval:
18
19 M[1][3] = M[1][2] + M[3][3] + d0 * d2 * d3 = 1200
20 M[2][4] = M[2][2] + M[3][4] + d1 * d2 * d4 = 240
21 M[3][5] = M[3][4] + M[5][5] + d2 * d4 * d5 = 700
22
23 Three Interval:
24
25 M[1][4] = M[1][1] + M[2][4] + d0 * d1 * d4 = 320
26 M[2][5] = M[2][4] + M[5][5] + d1 * d4 * d5 = 400
27
28 Four Interval:
29
30 M[1][5] = M[1][4] + M[5][5] + d0 * d4 * d5 = 1320
31

```

Result:

	1	2	3	4	5
1	0	200	1200	320	1320
2		0	400	240	640
3			0	200	700
4				0	2000
5					0

Order & Multiply times:

Order: ((A1(A2(A3A4)))A5)

Multiply Times: 1320

tags: Algorithm Class