Week 7 Solution

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1. You are given two implementations for finding the nth Fibonacci number(F)
Fibonacci numbers are defined by
F(n) = F(n-1) + F(n-2)
with F(0) = 0 and F(1) = 1
The two implementations are
1. Approach 1
int fib(int n)
 if (n \le 1)
   return n;
 return fib(n-1) + fib(n-2);
}
2. Approach 2
int fib(int n)
/* array to store fibonacci numbers. */
int f[n+1];
int i;
f[0] = 0;
f[1] = 1;
for (i = 2; i \le n; i++) {
 f[i] = f[i-1] + f[i-2];
return f[n];
Which of the two algorithms has better time complexity?
    1. Approach 1
    2. Approach 2
    3. Both have same time complexity
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Explanation: Approach 2 is linear time while approach 1 is exponential time

2. Consider the problem of matrix chain multiplication.

Answer: 2

Let $\mathbf{p_0}, \mathbf{p_1}, \mathbf{p_2}, \mathbf{p_n}$ be the dimension of the matrices

$$A_1 \ A_2 \ \ A_n$$
 such that dimension of A_i is $p_{i\text{-}1} \ x \ p_i$

Given the structure of optimal solution as

$$Ai...j = (A_i A_k)(A_{k+1} A_i)$$

For
$$1 \le i \le j \le n$$
, Let $m[i,j]$

denote the minimum number of multiplications needed to compute Ai...j . The optimum cost can be described by which of the following recursive definition

- A. $m[i,j] = \min_{i \le k \le i} (m[i,k] + m[k+1,j] + p_{i-1} p_k p_i)$
- B. $m[i,j] = min_{i \le k \le j}(m[i,k] + m[k+1,j] + p_{i-1} p_k p_j)$
- C. $m[i,j] = min_{i \le k < j}(m[i,k] + m[k,j] + p_{i-1} p_k p_j)$
- $D. \quad m[i,j] = min_{i \leq k < j} (m[i,k-1] + m[k+1,j] + p_{i-1} p_k p_j)$

Answer: A

3. Consider the following 4 matrices

A:5x4

B:4x6

C:6x2

D:2x7

Using the recursive definition for matrix chain multiplication, compute the values for x and y in the table below

	1	2 3		4
1	0	х	у	
2	×	0	48	104
3	у	48	0	84
4		104	84	0

A.
$$x = 88$$
 and $y = 120$

B.
$$x = 116$$
 and $y = 88$

C.
$$x = 120$$
 and $y = 116$

D.
$$x = 120$$
 and $y = 88$

Answer: D

4. What is the total number of scalar multiplications required to multiply the 4 matrices defined in question above?

- A. 36
- B. 168
- C. 158
- D. 104

Answer: C

Explanation: Answer can be verified by filling the table using the recursive definition.

5. int fun(int n){ T[0]=T[1]=2; T[2]=2*T[0]*T[1];

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for (int i=3;i< n;i++)
  T[i]=T[i-1]+2*T[i-1]*T[i-2];
return T[n]
}
Which of the following corresponds to the space and time complexity for the above code?
   A. O(n) & O(n)
   B. O(n) \& O(n^2)
   C. O(n^2) & O(n)
    D. None of the above
Answer: A
6. int fun(int n){
T[0]=T[1]=2;
T[2]=2*T[0]*T[1];
for (int i=3;i< n;i++)
  T[i]=T[i-1]+2*T[i-1]*T[i-2];
return T[n]
}
If T[0]=T[1]=2, then for n>1 the recurrence relation for the above code can be given by:
T[0]=T[1]=2.
   A. i=1n-12*T[i]*T[i-1]
   B. i=1n-12*T[i-1]*T[i-2]
   C. 2*T[i]*T[i-1]
i=1n2*T[i]*T[i-1]
7. Given n types of coin denominations of values V_1 < V_2 < ... < V_n (integers).
Let V_1=1, so that for any amount of money M change can always be made.
If T(j) indicates the minimum number of coins requires to make a change for the amount of
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money equal to j and coin V_k was the last denomination added to the solution.

Then which recurrence best describes T(j)?

- A. $Min_k\{T(j-V_k)\}+1$
- B. $Min_k\{T(j-V_k)\}$
- C. $T(j-V_k)+1$
- D. None of the above

Answer: A. Explanation:

if the coin denomination k was the last denomination added to the solution then the optimal way to finish the solution with that one is optimally make change for the amount of money $j-v_k$ and then add one extra coin of value V_k ?

8. Consider the following set of denominations {1,3,4,5} in a currency. What is the number of coins returned by the greedy strategy to make change for 7 rupees?

Also state what is the number of coins in an optimal solution?

- A. 3, 2
- B. 2, 3
- C. 3, 3
- D. None of the above

Answer: A. Explanation:

Using the greedy strategy we'll get one coin of denomination 5 and 2 coins of denomination 1. The best solution would be to pick one coin of denomination 3 and one coin of denomination 4

- 9. We use dynamic programming approach when
- A. The solution has optimal substructure
- B. The problem can be has divided into subproblems which are overlapping.
- C. The problem can be has divided into subproblems and a global solution can be achieved by making locally optimal solutions
- D. A & C
- E. A & B

Answer: E

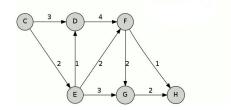
Explanation: DP needs optimal substructure and overlap between subproblems.

- 10. We use Greedy approach when
- A. The solution has optimal substructure
- B. The problem can be has divided into subproblems which are overlapping.
- C. The problem can be has divided into subproblems and a global solution can be achieved by making locally optimal solutions
- D. A & C
- E. A & B

Answer: D

Explanation: Greedy needs optimal substructure and global solution should be obtainable by making locally optimal solutions.

11. Find length of Shortest path from C to H?

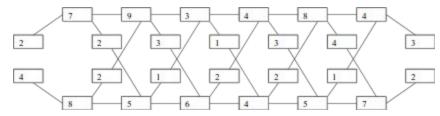


- A. 8
- B. 5
- C. 7
- D. 4

Answer: B

Explanation: C->E->F->H

12.



Given above assembly line with 6 stations, find fastest time to get through entire factory

- A. 38
- B. 36
- C. 39
- D. None of the above

Answer: A Explanation:

i \ j	1	2	3	4	5	6	exit
$f_1[j]$	1: $2+7=9$	<u>1:</u> 9 +9 = <u>18</u>	1: $18 + 3 = 21$	<u>1:</u> 20 +4 = <u>24</u>	<u>1</u> : 24 +8 = <u>32</u>	1: 32 +4 = 36	35+3 = 38
	8 80	2: 12+2+9 = 21	<u>2:</u> 16+1+3 = <u>20</u>	2: 22+2+4=28	2: 25+2+8 = 35	$\underline{2}$: 30+1+4 = $\underline{35}$	13.7 — 33%
$f_2[j]$	2:4+8=12	2: 12+ 5 = 17	$\underline{2}$: 16 +6 = $\underline{22}$	2: 22 +4 = 26	$\underline{2}$: 25 +5 = $\underline{30}$	$\underline{2}$: 30 +7 = $\underline{37}$	37+2 == 39
1236	711	1: 9+2+5 = 16	1: 18+3+6 = 27	1: 20+1+4 = 25	1: 24+3+5 = 32	1: 32+4+7 = 43	777

13. A naive way to calculate the nth Fibonacci number is to use the definition of Fibonacci number

$$F(n) = F(n-1) + F(n-2), F(0) = F(1) = 1$$

We know that this algorithm is exponential, because we do a lot of repetitive computation. A DP solution to the above problem would give us a linear time algorithm. The number of calculations done by naive method in computing F(4) are ?

Answer: 9 Explanation:

$$F(4) = F(3) + F(2)$$

$$F(3) = F(2) + F(1)$$

$$F(2) = F(1) + F(0)$$

$$F(4) = 8$$
 calls in calls + 1 call to $F(4)$

- 14. For the coin change problem, does greedy algorithm design strategy always result in the right answer?
 - A. Yes
 - B. No

Answer: B

Explanation: Optimality of greedy strategy is not guaranteed for arbitrary basic coins.

For example:

If basic coins were (1,3,4)

To get coins worth 6

Greedy would give (4,1,1)

while optimal would be (3,3)