You're expected to work on the discussion problems before coming to the lab. Discussion session is not meant to be a lecture. TA will guide the discussion and correct your solutions if needed. We may or may not release 'official' solutions. If you're better prepared for discussion, you will learn more. TAs will record names of the students who actively engage in discussion and report them to the instructor; they are also allowed to give some extra points to those students at their discretion. The instructor will factor in participation in final grade.

1. (Basic) Recall that the The Fibonacci Sequence is 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, \cdots . Formally, f(0) = 0, f(1) = 1 and f(i) = f(i-1) + f(i-2) for all $i \geq 2$, where f(i) denotes the *i*th Fibonacci number.

We saw the following naive pseudo-code/implementation of the function returns the n the Fibonacci number:

```
int F(n)
  if n == 0 return 0
  if n == 1 return 1
  return F(n-2) + F(n-1)
```

However, the running time is exponential in n. We can improve the running time to O(n) using DP. There are typically two equivalent ways to implement a DP approach, i) bottom-up method and ii) top-down with memoization. Give a pseudo-code of each implementation.

```
bottom up:
int F(n)
  Array A[0 ... n]
  A[0] = 0, A[1] = 1
  for i = 2; i \le n; i++
     A[i] = A[i-1] + A[i-2]
  return A[i]
top down:
int F(n)
  Array A[0 ... n]
  A[0] = 0, A[1] = 1
  A[2] = A[3] = \dots = A[n] = - infinity
  return Aux-F(A, n)
int Aux-F(n)
  if A[n] >= 0 return A[n]
  A[n] = Aux-F(A, n-1) + Aux-F(A, n-2)
  return A[n]
```

(Note: If you're asked to give a dynamic programming and decided to use a bottom-up method, you can describe it as follows: We set up a DP table a[0], a[1], ..., a[n] where a[i] denotes the *i*th Fibonacci number. Using the given recurrence, we fill out the table in the order of a[0], a[1], ..., a[n], and return a[n]. If you're asked to analyze the running time, you can say: there're O(n) table entries, and computing each table entry takes O(1) time, so the running time is O(n).

- 2. (Basic) Consider a sequence of integers defined by the following recurrence: $f(0) = 0, f(1) = 1, \text{ and } f(i) = f(i-1) + f(\lfloor i/2 \rfloor) \text{ for } i \geq 2.$ We would like to compute f(n) using a bottom-up DP method. Give a pseudo-code.
- 3. (Basic) Consider a sequence of integers defined by the following recurrence: f(0) = 0, f(1) = 1, and f(i) = f(i-1) + f(n-2) + f(n-3) + ... + f(1) + f(0) for $i \ge 2$. We would like to compute f(n) using a bottom-up DP method. Describe the DP and analyze its running time.
- 4. (Intermediate) Consider a modification of the rod-cutting problem (see CLRS Ch 15.1 page 360) in which, in addition to a price p_i for each rod, each cut incurs a fixed cost of c. The revenue associated with a solution is now the sum of the prices of the pieces minus the costs of making the cuts. For simplicity, we will be only interested in computing the optimum.
 - (a) Recall that if c = 0, then we would have the same recursion we had for the original rod-cutting problem: $r_n = \max_{1 \le i \le n} (p_i + r_{n-i})$. How would you change this recursion?
 - (b) Give a dynamic-programming algorithm to solve this modified problem.
- 5. (Advanced) Limited-Rod-Cutting (again, see CLRS Ch 15.1 page 360). Let's consider another variant of the rod-cutting problem. The input is exactly the same, $p_1, p_2, \dots p_n$. The only difference is that we're allowed to make at most 10 cuts. In other words, you want to maximize your revenue by cutting a rod of length n into at most 11 pieces. Our goal is to give a $O(n^2)$ time algorithm for this task. As before, let's focus on computing the optimum.
 - (a) Give a simple $O(n^{10})$ time algorithm.

- (b) We need to extend subproblems. Let $r_{i,k}$ denote the max revenue you can get from a rod of length i by making at most k cuts. We will be interested in computing $r_{n,10}$. Give a recurrence for computing $r_{i,k}$.
- (c) Give a DP. You can simply state how the DP table is defined, and in which order you will compute the DP entries. What is the number of DP entries? How much time do you need to compute each entry? What is the running time of your dp?
- (d) In the above, we only computed $r_{n,10}$. How do you find a solution that gives the revenue?

We compute s[i, k], which indicates the length of the first piece in an optimal solution for length i with at most k cuts. Then,

```
k = 10
while(n >=0 && k >=0)
  output s[n,k]
  n = n - s[n, k]
  k = k-1
```

- 6. (Basic) Matrix Chain Multiplication. Suppose we are given four matrices, A_1, A_2, A_3, A_4 with $p_0 = 3, p_1 = 2, p_2 = 3, p_3 = 2, p_4 = 3$. Fill out the DP table of entries m[i, j], where $1 \le i \le j \le 4$, and return m[i, j]; see the lecture slides or textbook for the definition of m[i, j]. Also show all s[i, j], where $1 \le i \le j \le 4$ and show an optimal full parenthesization using them.
- 7. (Basic) We would like to find an LCS of two sequences $\langle 1, 0, 0, 1, 0, 1, 0, 1 \rangle$ and $\langle 0, 1, 0, 1, 1, 0, 1, 1, 0 \rangle$ using the DP we learned. Fill out the DP table as shown on page 395 of the textbook.
- 8. (Advanced) Give an algorithm for computing an LCS of three given sequences, X, Y, and Z; for simplicity, assume that all the tree sequences are of length n. You have to define subproblems, give a recursion, and explain in which order the DP table is filled out. What is the running time of your algorithm?