

# Dynamic Programming

## Dynamic Programming Defined

Dynamic programming amounts to breaking down an optimization problem into simpler sub-problems, and storing the solution to each sub-problem so that each sub-problem is only solved once. Dynamic Programming is mainly used when solutions of same sub-problems are needed again and again (overlapping sub-problem).

The term **Programming** is not related to coding but it is from literature, and means **filling tables**

**Sub-problems** are smaller versions of the original problem. In fact, sub-problems often look like a reworded version of the original problem. If formulated correctly, sub-problems build on each other in order to obtain the solution to the original problem.

DP is a useful technique for optimization problems, those problems that seek the maximum or minimum solution given certain constraints, because it looks through all possible sub-problems and never recomputes the solution to any sub-problem. This guarantees correctness and efficiency, which we cannot say of most techniques used to solve or approximate algorithms.

General explanation:

Dynamic Programming = Recursion + memoization (caching subproblem solutions)

Two main properties of a problem that suggest that the given problem can be solved using Dynamic programming:

1. Overlapping Subproblems
2. Optimal Substructure

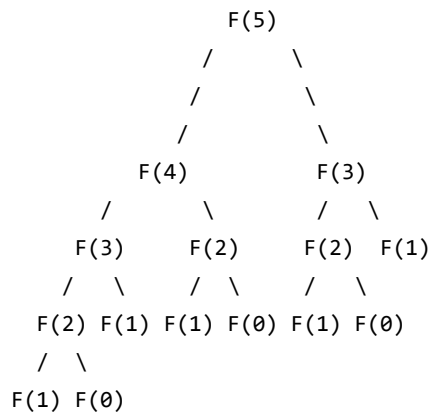
There are two main techniques of storing solution of sub-problems which are *not mutually exclusive*:

- Tabulation (Bottom Up)
- Memoization (Top down)

Bottom Up: If we start our transition from our base state i.e  $dp[0]$  and follow our state transition relation to reach our destination state  $dp[n]$ , we call it Bottom Up approach. Reverse is true for Top down approach.

Example:

fibonacci(5) - with general recursive code



The tree above represents each computation that must be made in order to find the Fibonacci value for  $n = 5$ . Notice how the sub-problem for  $n = 2$  is solved thrice. For a relatively small example ( $n = 5$ ), that's a lot of repeated , and wasted, computation!

What if, instead of calculating the Fibonacci value for  $n = 2$  three times, we created an algorithm that calculates it once, stores its value, and accesses the stored Fibonacci value for every subsequent occurrence of  $n = 2$ ? That's exactly what memoization does.

## Popular Dynamic Programming questions

- Fibonacci Sequence ✓
- 0-1 Knapsack problem ✓
- Longest common subsequence problem ✓
- Egg Dropping Puzzle
- String Edit Distance ✓
- Cutting a Rod
- Longest Increasing Subsequence

## How to solve a Dynamic Programming Problem

**Step 1:** Identify the sub-problem in words.

If you can identify a sub-problem that builds upon previous sub-problems to solve the problem at hand, then y

**Step 2:** Write out the sub-problem as a recurring mathematical decision.

If it is difficult to encode your sub-problem from Step 1 in math, then it may be the wrong sub-problem!

There are two questions that I ask myself every time I try to find a recurrence:

- What decision do I make at every step?
- If my algorithm is at step  $i$ , what information would it need to decide what to do in step  $i+1$ ?

(And sometimes: If my algorithm is at step  $i$ , what information did it need to decide what to do in step  $i-1$ ?)

**Step 3:** Solve the original problem using Steps 1 and 2.

Since the sub-problem we found in Step 1 is the maximum value schedule for punchcards  $i$  through  $n$  such that  $t$  we can write out the solution to the original problem as the maximum value schedule for punchcards 1 through  $n$ . Since Steps 1 and 2 go hand in hand, the original problem can also be written as  $OPT(1)$

**Step 4:** Determine the dimensions of the memoization array and the direction in which it should be filled.

**Step 5:** Code it!

## Performance analysis of dynamic algorithm

Generally, a dynamic program's runtime is composed of the following features:

Pre-processing

How many times the for loop runs

How much time it takes the recurrence to run in one for loop iteration

Post-processing

Overall, runtime takes the following form:

Pre-processing + Loop \* Recurrence + Post-processing

online articles:

- [Topcode Notes on DP](#)
- [Medium article on DP](#)
- [Top 20 DP questions](#)