

# **An Introduction To Dynamic Programming**

S.M Ahsanul Kabir Kowshid  
Student ID : 1505102

Waqar Hassan Khan  
Student ID : 1505107



Department of Computer Science and Engineering  
Bangladesh University of Engineering and Technology  
(BUET)

Dhaka 1000

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# 1 Introduction

*Dynamic programming* is a method that in general solves optimization problems that involve making a sequence of decisions by determining, for each decision, subproblems that can be solved in like fashion, such that an optimal solution of the original problem can be found from optimal solutions of subproblems.

This method is based on Bellman’s Principle of Optimality, which he phrased as follows

*An optimal policy has the property that whatever the initial state and initial decision are, the remaining decisions must constitute an optimal policy with regard to the state resulting from the first decision.*

More succinctly, this principle asserts that “optimal policies have optimal sub-policies.” That the principle is valid follows from the observation that, if a policy has a sub-policy that is not optimal, then replacement of the subpolicy by an optimal sub-policy would improve the original policy. The principle of optimality is also known as the “optimal substructure” property.

So the main properties of **DP** is as follows

- Overlapping Subproblems
- Optimal Substructure

## 1.1 Overlapping Subproblems

The subproblems of a problem may overlap while solving using **Divide and Conquer**, that is we may end solving the same thing over and over again. In DP we keep track of the solved sub-problems and avoid solving them more than once. Once solved it is stored to use for farther use.



3. Formulate state relationship - find a relation between previous states to reach the current state.
4. Do tabulation (or add memoization)

**Tabulation** Bottom Up Dynamic Programming (starting from  $dp[0]$  we move up)

**Memoization** Top Down Dynamic Programming (starting from  $dp[n]$  we use recursion here)

### 3 Some Solved Problems

#### 3.1 Calculating $C(n, r)$

Calculating combination or  $C(n, r)$  using dynamic programming is really easy. We can see the optimal substructure property from the figure below,  $C(n, r) = C(n - 1, r - 1) + C(n - 1, r)$ . Here we can either take an object then we have  $C(n - 1, r - 1)$  ways to choose or we left out that object then we have  $C(n, r - 1)$  ways to choose

We can also see overlapping sub-problems here in the figure marked red. The

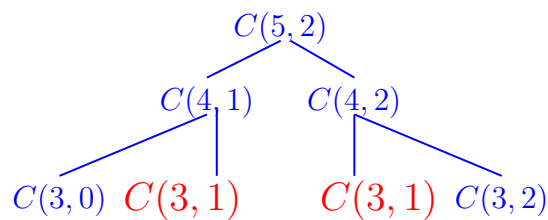


Figure 2: dynamic programming properties of  $C(n, r)$

#### 3.2 $C(n, r)$

```

1  /**from dust i have come, dust i will be***/
2
3  #include<bits/stdc++.h>
4
5  typedef long long int ll;

```

```

6
7 #define dbg printf("in\n")
8 #define nl printf("\n");
9 #define N 100
10
11 using namespace std;
12
13 ll dp[N][N];
14
15 ll nCr(int n, int r)
16 {
17     if(r==1) return n;
18     else if(r==n) return 1;
19
20     if(dp[n][r]!=-1) return dp[n][r];
21
22     return dp[n][r]=nCr(n-1,r)+nCr(n-1,r-1);
23 }
24
25 int main()
26 {
27     memset(dp,-1,sizeof(dp));
28     cout<<nCr(5,2); nl
29     cout<<nCr(9,6);
30
31     return 0;
32 }

```

## 4 Conclusion

Dynamic Programming has many classic problems and other problems need intuition and practice to solve.