Dynamic Programming

Invented by American mathematician Richard Bellman in the 1950s to solve optimization problems and later assimilated by CS. "Programming" here means "planning"

Dynamic Programming is a powerful algorithm design technique for solving problems that

- Appear to be exponential but have a poly solution with DP
- In many cases are optimization problems (min/max)
- Defined by or formulated as recurrences with overlapping subproblems
- Optimal solution to a problem contains optimal solutions to subproblems.

Dynamic Programming

- Like divide and conquer, DP solves problems by combining solutions to subproblems.
- Unlike divide and conquer, subproblems are not unique.
 - Subproblems may share subsubproblems,
 - However, solution to one subproblem may not affect the solutions to other subproblems of the same problem.
- Key: Determine structure of optimal solutions

5 Steps to DP

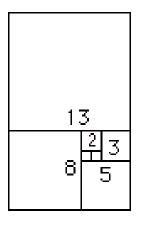
- 1. Define subproblems
- 2. Guess part of the solution
- 3. Relate subproblem solutions
- 4. Recurse + memoize or Build a DP bottom-up table.
- 5. Solve original problem

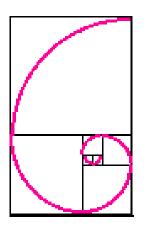
DP Examples

- Fibonacci
- Binomial Coefficients
- Longest Common Subsequence
- Longest Increasing Subsequence
- Knapsack
- Shortest Path
- Chain Matrix Multiplication
- Edit Distance
- Rod Cutting
- Optimal BST

Fibonacci Sequence

• 0,1,1,2,3,5,8,13,21,34,...





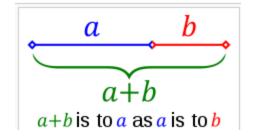




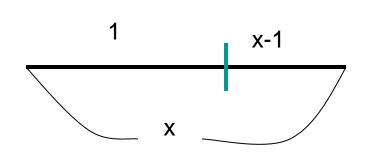
Fibonacci Number and Golden Ratio

0,1,1,2,3,5,8,13,21,34,...

$$\begin{cases} f_n = 0 & \text{if } n = 0 \\ f_n = 1 & \text{if } n = 1 \\ f_n = f_{n-1} + f_{n-2} & \text{if } n \ge 2 \end{cases}$$



$$\lim_{n\to\infty} \frac{f_n}{f_{n-1}} = \frac{1+\sqrt{5}}{2} = \text{Golden Ratio} = \phi = 1.61803..$$



$$\frac{x}{1} = \frac{1}{x-1}$$

$$x^2 - x - 1 = 0$$

$$x = \frac{1+\sqrt{5}}{2}$$

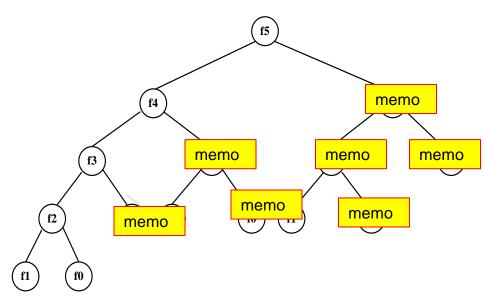
Naive Recursive Algorithm

```
fib (n) {
    if (n = 0) {
        return 0;
    } else if (n = 1) {
        return 1;
    } else {
        return fib(n-1) + fib(n-2);
    }
}
```

- Solved by a <u>recursive</u> program
- Much replicated computation is done.
- Running time Θ(φⁿ) exponential

Memoized DP Algorithm

```
memo = { }
fib (n) {
    if (n in memo) { return memo[n] }
    if (n <= 1) {
        f = n;
    } else {
        f = fib(n-1) + fib(n-2);
    }
    memo[n] = f;
    return f
}</pre>
```



- fib(k) only recurses the first time called only n nonmemoized calles
- Memorized calls "free" Θ(1).
- Time = #subproblems * time/subproblem
 = n * ⊕(1)
- Running time ⊕(n) linear

Bottom-up DP Algorithm

```
fib = { }
fib[0] = 0;
fib[1] = 1;
for k = 2 to n
fib[k] = fib[k-1] + fib[k-2];
return fib[n]
```

- Same as memoized DP with recursion "unrolled" into iteration.
- Practically faster since no recursion
- Analysis is more obvious
- Running time ⊕(n) linear

A Basic Idea of Dynamic Programming

- DP = recursion + memoization
 - Memoize = remember and reuse solutions to subproblems
 - Botton-Up Method stores all values in a table

Binomial Coefficient/Combinations C(n, k)

- the number of ways can you select k lottery balls out of n
- the number of birth orders possible in a family of n children where k are sons
- the number of acyclic paths connecting 2 corners of an k×(n-k) grid
- the coefficient of the a^kb^{n-k} term in the polynomial expansion of (a + b)ⁿ

$$C(n,k) \equiv \binom{n}{k} \equiv \frac{n!}{k!(n-k)!}$$

Example: Combinations

While easy to define, a binomial coefficient is difficult to compute 6 number lottery with 49 balls → 49!/6!43!

49! = 608,281,864,034,267,560,872,252,163,321,295,376,887,552,831,379,210,2 40.000,000

could try to get fancy by canceling terms from numerator & denominator

can still can end up with individual terms that exceed integer limits

A computationally easier approach makes use of the following recursive relationship

$$\binom{n}{k} \equiv \binom{n-1}{k-1} + \binom{n-1}{k}$$

e.g., to select 6 lottery balls out of 49, partition into:

selections that include 1 (must select 5 out of remaining 48)

selections that don't include 1 (must select 6 out of remaining 48)

Recursive Binomial Coefficient

could use straight divide & conquer to compute based on this relation

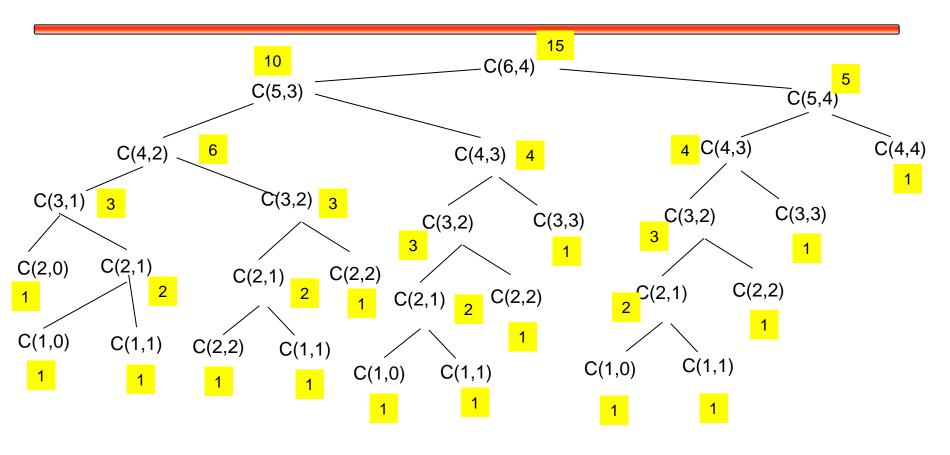
```
/** using divide-and-conquer
  * Calculates n choose k
  * n the total number to choose from (n > 0)
  * k the number to choose (0 <= k <= n)
  * */
int Binomial(int n, int k) {
  if (k == 0 || n == k) {
    return 1;
  }
  else {
    return Binomial(n-1, k-1) + Binomial(n-1, k);
  }
}</pre>
```

however, this will take a <u>long</u> time or exceed memory due to redundant work

```
\begin{pmatrix}
49 \\
6
\end{pmatrix}

\begin{pmatrix}
48 \\
5
\end{pmatrix}
+
\begin{pmatrix}
47 \\
4
\end{pmatrix}
+
\begin{pmatrix}
47 \\
5
\end{pmatrix}
+
\begin{pmatrix}
47 \\
6
\end{pmatrix}
```

Binomial Coefficient Tree



$$T(n, k) = T(n-1, k-1) + T(n-1, k) + 1$$

$$T(k,0) = 1$$
, $T(k,k) = 1$

Binomial Coefficient Recurrence

$$T(n, k) = T(n-1,k-1) + T(n-1, k) + 1$$

$$T(n,0) = 1, T(n,n) = 1$$

$$T(n, k) = T(n-1,k-1) + T(n-1, k) + 1 < 2 T(n-1) + 1$$
.....
$$O(2^n)$$

	K = 0	1	2	3	4	5	6
N = 0	1						
1	1	1					
2	1		1				
3	1			1			
4	1				1		
5	1					1	
6	1						1

	K = 0	1	2	3	4	5	6
N = 0	1						
1	1	1					
2	1	1+1=2	1				
3	1			1			
4	1				1		
5	1					1	
6	1						1

	K = 0	1	2	3	4	5	6
N = 0	1						
1	1	1					
2	1	2	1				
3	1	3		1			
4	1				1		
5	1					1	
6	1						1

	K = 0	1	2	3	4	5	6
N = 0	1						
1	1	1					
2	1 (2	1				
3	1	3	3	1			
4	1				1		
5	1					1	
6	1						1

	K = 0	1	2	3	4	5	6
N = 0	1						
1	1	1					
2	1	2	1				
3 (1 (3	3	1			
4	1	4			1		
5	1					1	
6	1						1

	K = 0	1	2	3	4	5	6
N = 0	1						
1	1	1					
2	1	2	1				
3	1	3	3	1			
4	1	4	6	4	1		
5	1	5	10	10	5	1	
6	1	6	15	20	15	6	1

DP Algorithm Binomial Coefficient

```
Binomial (n,k) 

// Computes C(n,k) by DP 

// Input: A pair of nonnegative integers n \ge k \ge 0 

// Output: the value of C(n,k) 

for i \leftarrow 0 to n do 

    for j \leftarrow 0 to min(i, k) do 

        if j = 0 or j = i 

        C[i, j] \leftarrow 1 

        else C[i, j] \leftarrow C[i-1, j-1] + C[i-1, j] 

Return C[n.k]
```

Running time: $\Theta(nk)$

Space efficiency: $\Theta(nk)$

```
Recurrence: C(n,k) = C(n-1,k) + C(n-1,k-1) for n > k > 0

C(n,0) = 1, C(n,n) = 1 for n \ge 0
```