Dynamic Programming

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

D*ynamic 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 independent.
  - 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

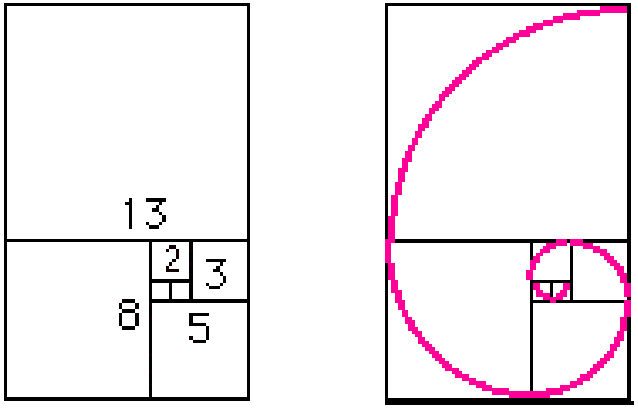
1. Define subproblems
2. Guess part of the solution
3. Relate subproblem solutions
4. Recurse + memoize or Build a DP bottom-up table.
5. Solve original problem

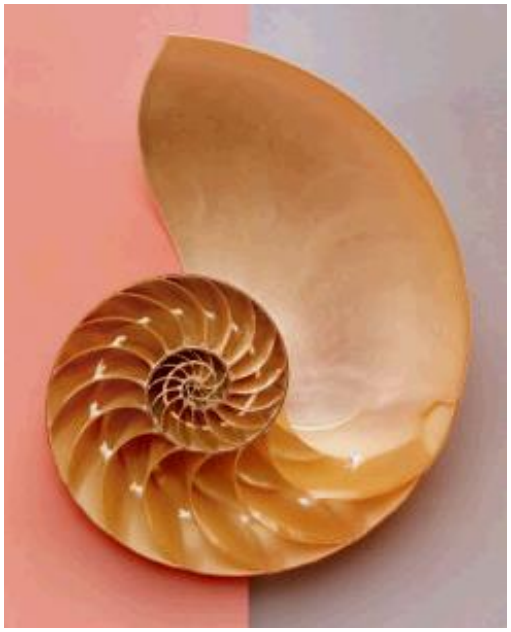
DP Examples

- 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,...

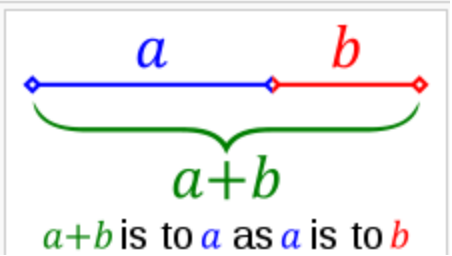




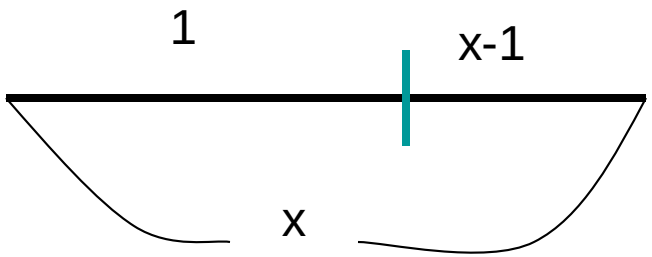
Fibonacci Number and Golden Ratio

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

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



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



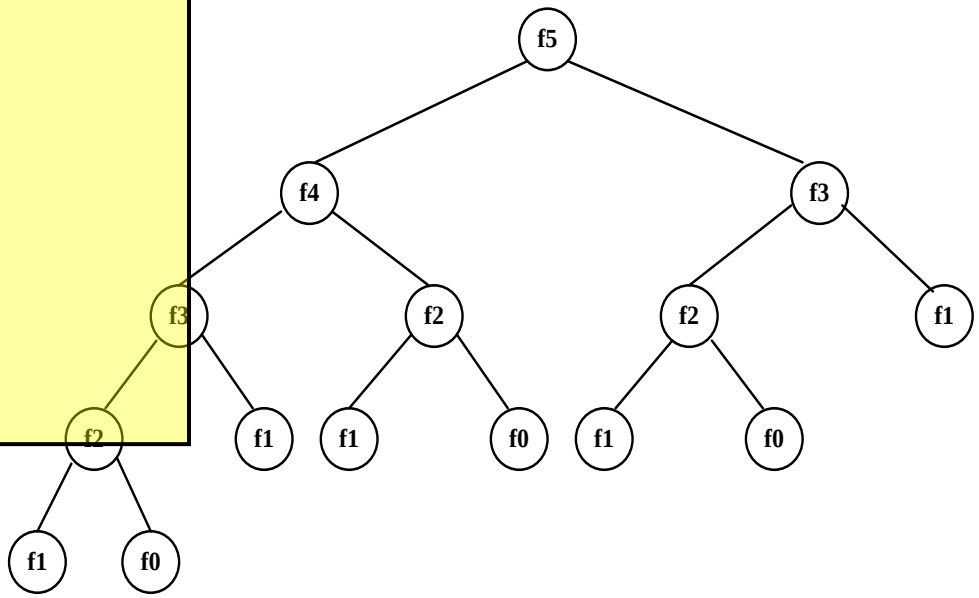
$$\frac{x}{1} = \frac{1}{x-1}$$

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

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

Naive Recursive Algorithm

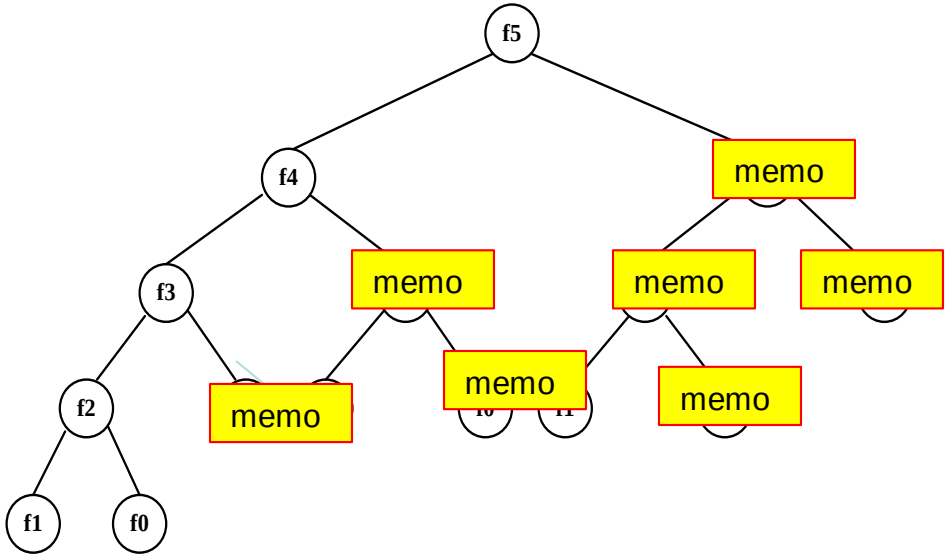
```
fib (n) {
  if (n = 0) {
      return 0;
  } else if (n = 1) {
      return 1;
  } else {
      return fib(n-1) + fib(n-2);
  }
}
```



- Solved by a recursive program
- Much replicated computation is done.
- Running time $\Theta(\phi^n)$ - 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
}
```



- fib(k) only recurses the first time called only n nonmemoized calles
- Memorized calls “free” $\Theta(1)$.
- Time = #subproblems * time/subproblem
  = n * $\Theta(1)$
- Running time $\Theta(n)$ - 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 $\Theta(n)$ - 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 way to select a group of 4 from 6 students
- the number of acyclic paths connecting 2 corners of an k×(n-k) grid
- the coefficient of the $a^k b^{n-k}$ term in the polynomial expansion of $(a + b)^n$

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

Binomial Coefficient/Combinations C(n, k)

– The number of way to select a group of 4 from 6 students

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

$$C(6,4) \equiv \binom{6}{4} \equiv \frac{6!}{4!(6-4)!} = 15$$

C(6,4) number of different groups of 4 students selected from 6

{a, b, c, d}	{a, b, e, f}	{c, d, e, f}
{a, b, c, e}	{a, c, d, e}	{b, d, e, f}
{a, b, c, f}	{a, c, d, f}	{b, c, e, f}
{a, b, d, e}	{a, c, e, f}	{b, c, d, e}
{a, b, f, d}	{a, d, e, f}	{b, c, d, f}

Binomial Coefficient/Combinations C(n, k)

– The number of way to select a group of 4 from 6 students

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

$$C(6,4) \equiv \binom{6}{4} \equiv \frac{6!}{4!(6-4)!} = 15$$

$$C(6,4) = \mathbf{C(5,3)} + \mathbf{C(5,4)}$$

Six students = {a, b, c, d, e, f}

Groups of 4:

{a, b, c, d}	{a, b, e, f}	{c, d, e, f}
{a, b, c, e}	{a, c, d, e}	{b, d, e, f}
{a, b, c, f}	{a, c, d, f}	{b, c, e, f}
{a, b, d, e}	{a, c, e, f}	{b, c, d, e}
{a, b, f, d}	{a, d, e, f}	{b, c, d, f}

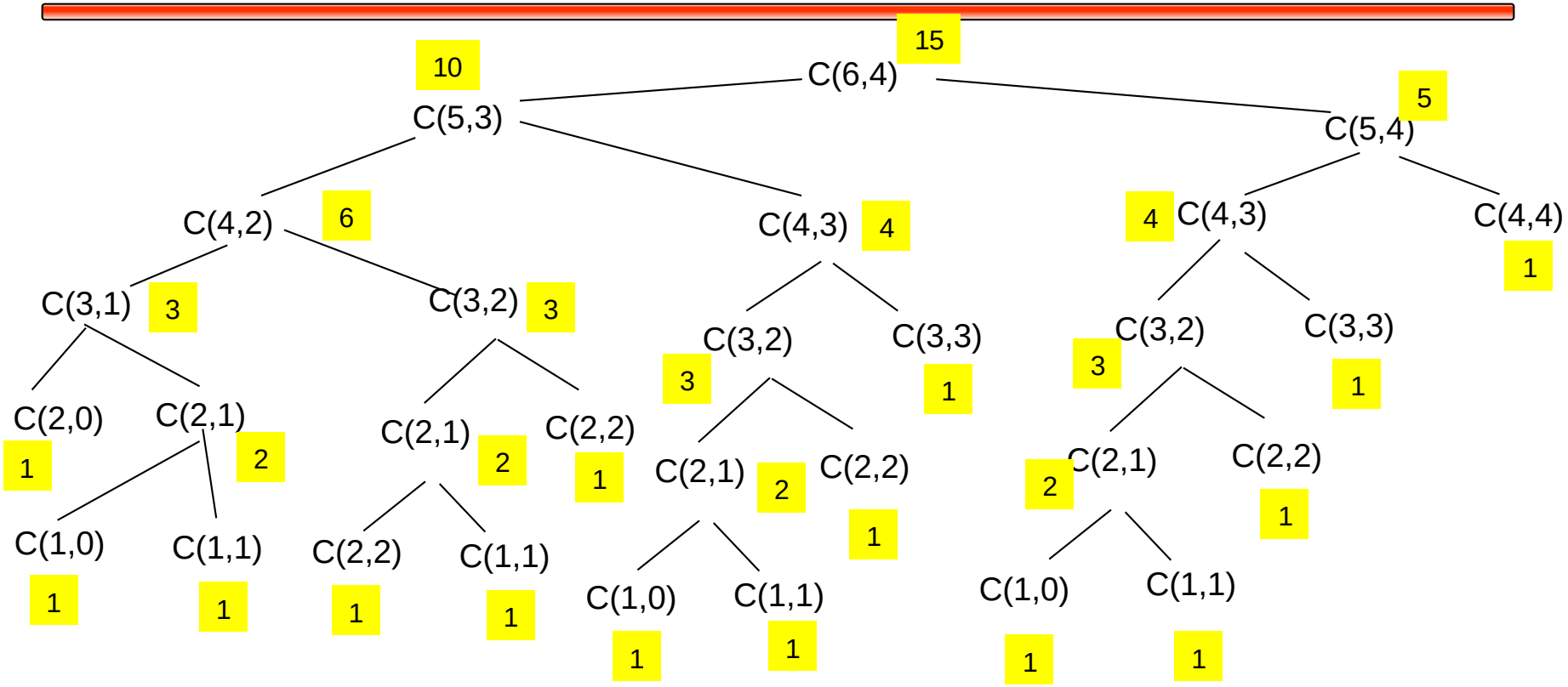
Recursive Relationship

A computationally easier approach makes use of the following recursive relationship

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

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

Binomial Coefficient Tree



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

$T(j,0) = 1, \quad T(j,j) = 1$

Example: Combinations

The number of ways to select 6 lottery balls from 49

6 number lottery with 49 balls → 49!/(6!43!) = 13,983,816

49! =608,281,864,034,267,560,872,252,163,321,295,376,887,552,831,379,210,240,000,000,000

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

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

$$\binom{49}{6} = \binom{48}{5} + \binom{48}{6}$$

$$\binom{47}{4} + \binom{47}{5} + \binom{47}{5} + \binom{47}{6}$$

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)

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

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

Recursive Combination

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 Combinationl(int n, int k) {
    if (k == 0 || n == k) {
        return 1;
    }
    else {
        return Combination(n-1, k-1) + Combinationl(n-1, k);
    }
}
```

however, this will take a long time or exceed memory due to redundant work

Recurrence

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

$T(j,0) = 1, \quad T(j,j) = 1$

$T(n, k) = T(n-1,k-1) + T(n-1, k) + 1 < 2\ T(n-1) + 1$

. . . .

$O(2^n)$

Computing a Combination by DP

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

$C(j,0) = 1, \quad C(j,j) = 1$ for $j \geq 0$

	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

Computing by DP

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

	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

Computing by DP

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

	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

Computing by DP

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	K = 0	1	2	3	4	5	6
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Computing by DP

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

Computing by DP

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

	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 for Combinations

```
CombDPl(n,k)
// Computes C(n,k) by DP
// Input: A pair of nonnegative integers n≥k≥0
// Output: the value of C(n,k)
for i ← 0 to n do
      for j ← 0 to min(i, k) do
            if j = 0 or j = i
                  C[i, j] ← 1
            else
                  C[i, j] ← C[i-1, j-1] + C[i-1, j]

Return C[n.k]
```

Running time: $\Theta(nk)$ k is bounded by n so in the worst case $\Theta(n^2)$

Space efficiency: $\Theta(nk)$ or $\Theta(n^2)$