Lecture 7: Dynamic Programming I

Overview

- Memoization & subproblems; bottom up
- Fibonacci
- Shortest paths

Dynamic Programming (DP)

- Powerful algorithmic design technique
- Large class of seemingly exponential problems have a polynomial solution ("only") via DP
- Particularly for optimization problems (min / max)

DP ~ «careful brute force»
DP ~ subproblems + «re-use»

Fibonacci Numbers

Problem

$$F_1 = F_2 = 1;$$

 $F_n = F_{n-1} + F_{n-2}$

Goal: compute Fn

Solution: Naive Algorithm:

```
fib(n):

if n \le 2: f = 1

else: f = fib(n - 1) + fib(n - 2)

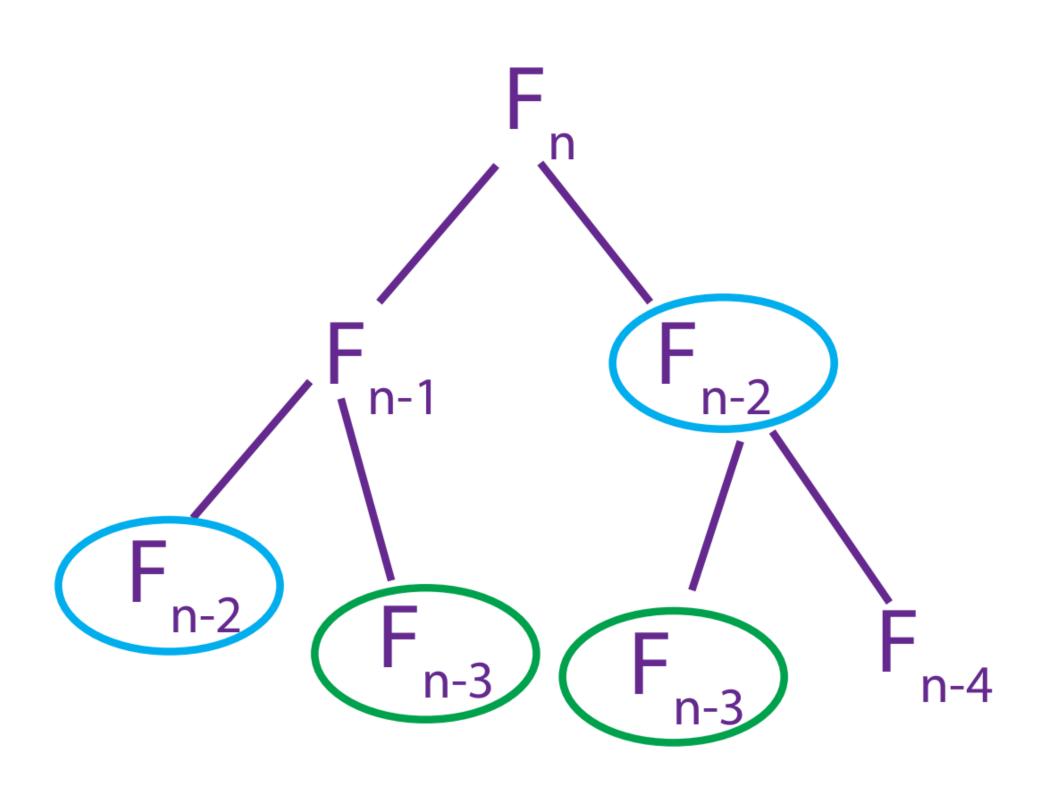
return f

T(n) = T(n-1) + T(n-2) + O(1)

\ge F_n \approx \Phi^n

\ge 2T(n-2) + O(1) \ge 2^{n/2}
```

Fibonacci Numbers Memoized Algorithm



Solution: Memoized Algorithm:

```
memo = \{\}
fib(n):
if n in memo: return memo[n]
else: if n \le 2: f = 1
else: f = fib(n - 1) + fib(n - 2)
memo[n] = f
return f
```

Fibonacci Numbers Memoized Algorithm

- fib(k) only recurses first time called, ∀k
- only n nonmemoized calls: k = n, n-1,...,1
- memoized calls free (Θ(1) time)
- Θ(1) time per call (ignoring recursion)

Solution: Memoized Algorithm:

```
memo = \{\}
fib(n):
if n in memo: return memo[n]
else: if n \le 2: f = 1
else: f = fib(n - 1) + fib(n - 2)
memo[n] = f
return f
```

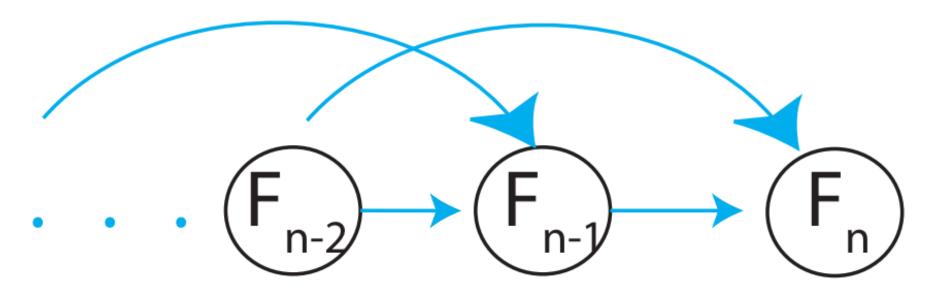
Memoized DP Algorithm

DP ~ recursion + memoization

- memoize (remember) & re-use solutions to subproblems that help solve problem
- in Fibonacci, subproblems are F1, F2, . . . , F
 - time = # of subproblems · time/subproblem
- Fibonacci: # of subproblems is n, and time/subproblem is $\Theta(1) = \Theta(n)$

Fibonacci Numbers Bottom-up DP Algorithm

- exactly the same computation as memoized DP (recursion "unrolled")
- in general: topological sort of subproblem dependency DAG



- practically faster: no recursion
- can save space: just remember
 last 2 fibs

Solution: Memoized Algorithm:

```
fib = {}

for k in [1, 2, . . . , n]:

  if k \le 2: f = 1

  else: f = fib[k - 1] + fib[k - 2]

  fib[k] = f

return fib[n]
```