Dynamic Programming

- Computing Fibonacci Numbers
- Introduction to Dynamic Programming
- Optimal Substructure
- Memoization

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Algorithm Design Paradigms

Greedy Algorithms

- Build up a solution incrementally
- Myopically and locally optimizing some local criterion

Divide and Conquer

- Break up a problem into (independent) sub-problems
- Solve each sub-problem independently
- Combine solution to sub-problems to form solution to original problem
- Dynamic programming = planning over time
 - More general and powerful than divide and conquer
 - Break up a problem into (in)(dependent) sub-problems
 - Generally there is a sequence of problems
 - Identify the optimal substructure: when optimal solution to a problem is made up of optimal solution to smaller subproblems
 - Build up solution to larger and larger subproblems
 - Identify redundancy and repetitions
 - Use memoization or build up memo on the run

 $0, 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89 \dots$

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

For $n \ge 8$ $F_n > 2^{n/2}$

▷ Prove it by induction

Implementing the recursive definition of F_n

Algorithm Recursive F_n computation function FIB1(n) if n = 0 then return 0 else if n = 1 then return 1 else return FIB1(n - 1) + FIB1(n - 2)

It's correctness follows from the definition

How much time it takes to compute F_n ?

Recursive F_n computation

Algorithm Recursive *F_n* computation

function FIB1(n) if n = 0 then return 0 else if n = 1 then return 1 else return FIB1(n - 1) + FIB1(n - 2)

Let T(n) be the number of operations on input n

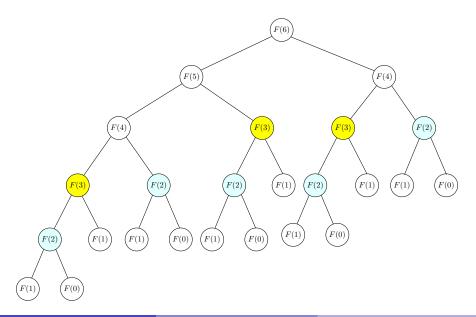
$$T(n) = \begin{cases} 1 & \text{if } n = 0 \\ 2 & \text{if } n = 1 \\ T(n-1) + T(n-2) + 3 & \text{if } n \ge 2 \end{cases} \qquad F_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ F_{n-1} + F_{n-2} & \text{if } n \ge 2 \end{cases}$$

For $n \ge 8$, $T(n) > F_n \ge 2^{n/2}$

exponential in n

Problem is unnecessarily repeated recursive calls

Recursive F_n computation



Algorithm Recursive *F_n* computation

function FIB1(n) if n = 0 then return 0 else if n = 1 then return 1 else return FIB1(n - 1) + FIB1(n - 2)

For
$$n \geq 8$$
, $T(n) > F_n \geq 2^{n/2}$

Problem is unnecessarily repeated recursive calls

Memoization: Save results of subproblems in a memo

Use the memo when needed instead of recomputing

Algorithm *F_n* computation with memoization

```
F[0...n] \leftarrow \operatorname{NEGONES}(n+1)
F[0] \leftarrow 0
F[1] \leftarrow 1
function FIB2(n)
if F[n-1] = -1 then
F[n-1] \leftarrow \operatorname{FIB2}(n-1) \qquad \triangleright \text{ Call FIB2 function only if } F[n-1] = -1
if F[n-2] = -1 then
F[n-2] \leftarrow \operatorname{FIB2}(n-2)
return F[n-1] + F[n-2]
```

F_n computation with Memoization

Algorithm Compute F_n with memo

- $F[0...n] \leftarrow \text{NEGONES}(n+1)$ $F[0] \leftarrow 0$ $F[1] \leftarrow 1$ function Fib2(n)
 - if F[n-1] = -1 then $F[n-1] \leftarrow FIB2(n-1)$ if F[n-2] = -1 then $F[n-2] \leftarrow FIB2(n-2)$ return F[n-1] + F[n-2]

Let $T_2(n)$ be runtime of FIB2(n)

- Count number of calls
- Only calls if $F[\cdot] = -1$
- Total calls n+1
- O(1) operations per call

 $T_2(n) = O(n)$

▷ Compare with $T(n) = O(2^n)$

F_n computation Bottom Up Approach

Algorithm Bottom-Up F_n Computation

```
F[0...n] \leftarrow \text{NEGONES}(n+1)
F[0] \leftarrow 0
F[1] \leftarrow 1
for i = 2 to n do
F[i] \leftarrow F[i-1] + F[i-2]
return F[n]
```

- No recursion overhead
- Analyze time needed to fill up memo
- Total number of updates to memo is n+1
- Total runtime $T_3(n) = O(n)$

▷ Compare with $T(n) = O(2^n)$