

Algorithm

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Dynamic Programming

Guess

Recursion

Memoization

Dynamic Programming

- General and powerful design technique of algorithm
- Do **brute force**, but be **careful**.
- Do finding the shortest paths in some DAG.
- Solve subproblems and do recycling.
- Time complexity?

$$\text{time} = (\# \text{ sub problems}) \times [(\text{time}) / (\text{subproblem})]$$

Today

- **Fibonacci**
- **Shortest Path**
- **Guess**
- **DAG**

Fibonacci numbers

- **Sequence**
 - $F_1 = F_2 = 1$
 - $F_n = F_{n-1} + F_{n-2}$
- **Goal: Compute F_n**

Fibonacci numbers

- **Naive recursive algorithm:**

```
fib(n):  
    if n <= 2: f = 1  
    else f = fib(n-1) + fib(n-2)  
    return f
```

- **Time complexity**

$$T(n) = T(n-1) + T(n-2) + \Theta(1) \geq \Theta(2^{n/2})$$

Fibonacci numbers

- Memoized DP Alg.:

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

Fibonacci numbers

- **Look at the memorized DP's efficiency.**

Draw a tree

or

fib(k) only recurses the first time it's called for all k

- memoized calls cost $\Theta(1)$
- # nonmemoized calls is n
- nonrecursive work per call: $\Theta(1)$
- ▶ time = $\Theta(n)$

Note. Memoized DP here is not the best alg. for computing Fib!

Fibonacci numbers

- **Bottom-up DP Alg.:**

```
fib = {}  
for k in range(1, n+1):  
    if k <= 2: f = 1  
    else f = fib(k-1) + fib(k-2)  
    fib[k] = f  
return fib[n]
```

- **exactly same computation as the memorized version.**
- **topological sort of subproblems dependency DAG.**
- **can often save space.**

Shortest Path

- $\delta(s, v)$ for all v
- $\delta(s, v) = \min_{(u, v) \in E} (\underbrace{\delta(s, u)}_{\text{recursive call}} + w(u, v))$
- Is this good algorithm?
- Subproblem dependencies should be acyclic.
- Time?
 - $\Theta(V^3)$
 - but actually? $\Theta(VE)$ ► Bellman-Ford