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6.006 Introduction to Algorithms Spring 2008

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Lecture 19: Dynamic Programming I: Memoization, Fibonacci, Crazy Eights, Guessing

Lecture Overview

- Fibonacci Warmup
- Memoization and subproblems
- Shortest Paths
- Crazy Eights
- Guessing Viewpoint

Readings

CLRS 15

Dynamic Programming (DP)

Big idea: :hard yet simple

- Powerful algorithmic design technique
- Large class of seemingly exponential problems have a polynomial solution ("only") via DP
- Particularly for optimization problems (min / max) (e.g., shortest paths)
- * DP \approx "controlled brute force"
- * DP \approx recursion + re-use

Fibonacci Numbers

$$F_1 = F_2 = 1;$$
 $F_n = F_{n-1} + F_{n-2}$

Naive Algorithm

follow recursive definition

$$\begin{array}{l} \underline{\operatorname{fib}}(n): \\ & \text{if } n \leq 2: \text{ return } 1 \\ & \text{else return } \operatorname{fib}(n-1) + \operatorname{fib}(n-2) \\ \implies T(n) = T(n-1) + T(n-2) + O(1) \approx \phi^n \\ & \geq 2T(n-2) + O(1) \geq 2^{n/2} \\ & \underline{\operatorname{EXPONENTIAL}} - \underline{\operatorname{BAD}!} \end{array}$$

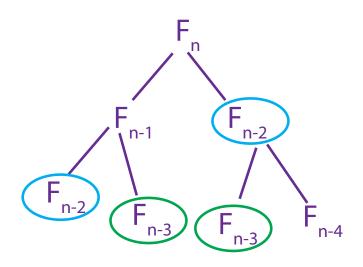


Figure 1: Naive Fibonacci Algorithm

Simple Idea

memoize

$$\begin{split} \text{memo} &= \{ \ \} \\ \text{fib}(n): \\ & \text{if } n \text{ in memo: return memo}[n] \\ & \text{else: if } n \leq 2: f = 1 \\ & \text{else: } f = \text{fib}(n-1) + \underbrace{\text{fib}(n-2)}_{\text{free}} \\ & \text{memo}[n] = f \\ & \text{return } f \\ T(n) &= T(n-1) + O(1) = O(n) \end{split}$$

[Side Note: There is also an $O(\lg n)$ - time algorithm for Fibonacci, via different techniques]

* DP \approx recursion + memoization

• remember (memoize) previously solved "subproblems" that make up problem

- in Fibonacci, subproblems are F_0, F_1, \cdots, F_n

- if subproblem already solved, re-use solution
- * \implies time = \sharp of subproblems \cdot time/subproblem
 - - in fib: \sharp of subproblems is O(n) and time/subproblem is O(1) giving us a total time of O(n).

Shortest Paths

- Recursive formulation: $\delta(s,t) = \min\{w(s,v) + \delta(v,t) | (s,v) \in E\}$
- does this work with memoization? no, cycles \implies infinite loops (see Figure 2).

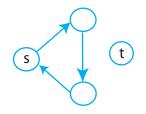


Figure 2: Shortest Paths

- in some sense necessary for neg-weight cycles
- works for directed acyclic graphs in O(V + E)(recursion effectively DFS/topological sort)
- trick for shortest paths: removing cyclic dependency.

$$- \delta_{k}(s,t) = \text{shortest path using} \leq k \text{ edges}$$

$$= \min\{\delta_{k-1}(s,t)\} \cup \{w(s,v) + \delta_{k-1}(v,t) \mid (s,v) \in E\}$$

$$\dots \text{ except } \delta_{k}(t,t) = \phi, \ \delta_{\phi}(s,t) = \infty \text{ if } s \neq t$$

$$- \delta(s,t) = \delta_{n-1}(s,t) \text{ assuming no negative cycles}$$

$$\implies \text{ time} = \underbrace{\sharp \text{ subproblems}}_{O(n^{3}) \text{ for } s,t,k\cdots \text{ really } O(n^{2})} \underbrace{\bullet}_{O(n)\cdots \text{ really } \text{ degV}}_{O(n)\cdots \text{ really } \text{ degV}}$$

$$= O(V \cdot \sum_{V} \text{ deg}(V)) = O(VE)$$

* Subproblem dependency should be acyclic.

Crazy Eights Puzzle

- given a sequence of cards $c[\phi], c[1], \cdots, c[n-1]$ e.g., $7\heartsuit, 6\heartsuit, 7\diamondsuit, 3\diamondsuit, 8\clubsuit, J\bigstar$
- find longest left-to-right "trick" (subsequence)

 $c[i_1], c[i_2], \cdots c[i_k] \quad (i_1 < i_2 < \cdots i_k)$ where $c[i_j] \& c[i_{j+1}]$ "<u>match</u>" for all jhave some suit or rank or one has rank 8

• recursive formulation:

$$\begin{aligned} \operatorname{trick}(i) &= \operatorname{length} \text{ of best trick starting at } c[i] \\ &= 1 + \max(\operatorname{trick}(j) \operatorname{for} j \operatorname{in range}(i+1,n) \operatorname{if match} (c[i], c[j])) \\ \operatorname{best} &= \max(\operatorname{trick}(i) \operatorname{for} i \operatorname{in range}(n)) \end{aligned}$$

• memoize: trick(i) depends only on trick(>i)

$$\implies \text{time} = \underbrace{\sharp \text{subproblems}}_{O(n)} \cdot \underbrace{\text{time/subproblem}}_{O(n)}$$
$$= O(n^2) \quad \text{(to find actual trick, trace through max's)}$$

"Guessing" Viewpoint

- what is the first card in best trick? guess!
 i.e., try all possibilities & take best result
 only O(n) choices
- what is next card in best trick from i? guess!
 - if you pretend you knew, solution becomes easy (using other subproblems)
 - actually pay factor of O(n) to try all
- * use only small # choices/guesses per subproblem

 $\operatorname{poly}(n){\sim}O(1)$