

Outline: Dynamic Programming II (of 4)

- review of big ideas & examples so far
- bottom-up implementation
- longest common subsequence
- parent pointers for guesses

Reading: CLRS 15Summary:

- \* DP  $\approx$  "controlled brute force"
- \* DP  $\approx$  guessing + recursion + memoization
- \* DP  $\approx$  dividing into reasonable # subproblems whose solutions relate — acyclicly — usually via guessing parts of solution
- \* time = # subproblems  $\cdot$  time/subproblem  
 treating recursive calls as  $O(1)$   
 (usually mainly guessing)
  - essentially an amortization
  - count each subproblem only once; after first time, costs  $O(1)$  via memoization

Examples: Fibonacci      Shortest Paths      Crazy Eights

subprobs.:  $\text{fib}(k)$   
 $0 \leq k \leq n$

$\delta_k(s,t) \forall s, k < n$   
 $= \text{min. path } s \rightarrow t$   
 using  $\leq k$  edges

$\text{trick}(i)$   
 $= \text{longest trick}$   
 from card  $i$

#subprobs.:  $\Theta(n)$

$\Theta(V^2)$

$\Theta(n)$

guessing: none

edge from  $s_n$  if any

next card  $j$

#choices: 1

$\text{deg}(s)$

$n-i$

relation:  $= \text{fib}(k-1)$   
 $+ \text{fib}(k-2)$

$= \min \{ \delta_{k-1}(s,t) \}$   
 $\cup \{ w(s,v) + \delta_{k-1}(v,t) \}$   
 $| v \in \text{Adj}[s] \}$

$= 1 + \max(\text{trick}(j))$   
 for  $i < j < n$  if  
 $\text{match}(c[i], c[j])$

time/subpr.:  $\Theta(1)$

$\Theta(1 + \text{deg}(s))$

$\Theta(n-i)$

DP time:  $\Theta(n)$

$\Theta(VE)$

$\Theta(n^2)$

orig. prob:  $\text{fib}(n)$

$\delta_{n-1}(s,t)$

$\max \{ \text{trick}(i) \}$   
 $| 0 \leq i < n \}$

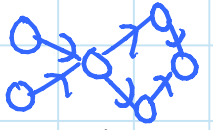
extra time:  $\Theta(1)$

$\Theta(1)$

$\Theta(n)$

## Bottom-up implementation of DP: alternative to recursion

- subproblem dependencies form DAG
- imagine topological sort
- iterate through subproblems in that order
- ⇒ when solving a subproblem, have already solved all dependencies
- often just: "solve smaller subproblems first"



e.g. Fibonacci:

for  $k$  in range( $n+1$ ):  $fib[k] = \dots$

Shortest Paths:

for  $k$  in range( $n$ ): for  $v$  in  $V$ :  $d[k, v, t] = \dots$

Crazy Eights:

for  $i$  in reversed(range( $n$ )):  $trick[i] = \dots$

- no recursion or memoized tests
- ⇒ faster in practice

- building DP table of solutions to all subprobs.
- can often optimize space:
  - Fibonacci: **PSG**
  - Shortest Paths: re-use same table  $\forall k$

# Longest common subsequence: (LCS)

A.K.A. edit distance, diff, CVS/SVN, spellchecking, DNA comparison, plagiarism detection, etc.

given two strings/sequences  $x$  &  $y$ ,  
find longest common subsequence  $LCS(x,y)$   
sequential but not necessarily contiguous

- e.g.: HIEROGLYPHOLOGY } HELLO  
vs. MICHAELANGELO
- equivalent to "edit distance" (unit costs):  
# character insertions/deletions to transform  $x \rightarrow y$   
(everything except the matches)

- brute force: try all  $2^{|x|}$  subsequences of  $x$   
 $\Rightarrow \Theta(2^{|x|} \cdot |y|)$  time
- instead: DP on two sequences simultaneously

\* useful subproblems for strings/sequences  $x$ :

- suffixes  $x[i:]$
- prefixes  $x[:i]$
- substrings  $x[i:j]$

}  $\Theta(|x|)$  ← cheaper  $\Rightarrow$  use if possible  
}  $\Theta(|x|^2)$

- idea: combine such subproblems for  $x$  &  $y$   
(suffixes or prefixes work)

# LCS DP:

- subproblem  $c(i, j) = |\text{LCS}(x[i:], y[j:])|$   
for  $0 \leq i, j < n$
- $\Rightarrow \Theta(n^2)$  subproblems
- original problem  $\approx c[0, 0]$  (length ~ find seq. later)
- idea: either  $x[i] = y[j]$  part of LCS  
or not  $\Rightarrow$  either  $x[i]$  or  $y[j]$  (or both)  
not in LCS (with anyone)

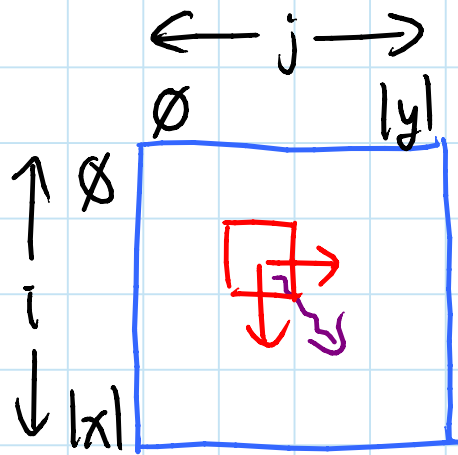
- guess: drop  $x[i]$  or  $y[j]$ ? (2 choices)

- relation among subproblems:  
if  $x[i] = y[j]$ :  $c(i, j) = 1 + c(i+1, j+1)$   
(otherwise  $x[i]$  or  $y[j]$  unused ~ can't help)
- else:  $c(i, j) = \max \left\{ \underbrace{c(i+1, j)}_{x[i] \text{ out}}, \underbrace{c(i, j+1)}_{y[j] \text{ out}} \right\}$
- base cases:  $c(|x|, j) = c(i, |y|) = 0$

$\Rightarrow \Theta(1)$  time per subproblem

$\Rightarrow \Theta(n^2)$  total time for DP

- DP table:



$\rightarrow$  if  $x[i] \neq y[j]$   
 $\searrow$  if  $x[i] = y[j]$

[ linear space via antidiagonal order  $\nearrow$  ]

- recursive DP:

```
def LCS(x, y):
```

```
    seen = {}
```

```
    def c(i, j):
```

```
        if i ≥ len(x) or j ≥ len(y): return ∅
```

```
        if (i, j) not in seen:
```

```
            if x[i] == y[j]:
```

```
                seen[i, j] = 1 + c(i+1, j+1)
```

```
            else:
```

```
                seen[i, j] = max(c(i+1, j), c(i, j+1))
```

```
        return seen[i, j]
```

```
    return c(0, 0)
```

- bottom-up DP:

```
def LCS(x, y):
```

```
    c = {}
```

```
    for i in range(len(x)):
```

```
        c[i, len(y)] = ∅
```

```
    for j in range(len(y)):
```

```
        c[len(x), j] = ∅
```

```
    for i in reversed(range(len(x))):
```

```
        for j in reversed(range(len(y))):
```

```
            if x[i] == y[j]:
```

```
                c[i, j] = 1 + c[i+1, j+1]
```

```
            else:
```

```
                c[i, j] = max(c[i+1, j], c[i, j+1])
```

```
    return c[0, 0]
```

## Recovering LCS: [material covered in recitation]

- to get LCS, not just its length, store parent pointers (like shortest paths) to remember correct choices for guesses:

if  $x[i] == y[j]$ :  
 $c[i, j] = 1 + c[i+1, j+1]$   
 $\text{parent}[i, j] = (i+1, j+1)$

else:

if  $c[i+1, j] > c[i, j+1]$ :

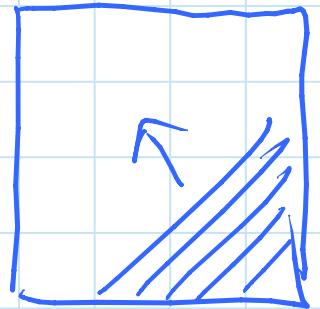
$c[i, j] = c[i+1, j]$

$\text{parent}[i, j] = (i+1, j)$

else:

$c[i, j] = c[i, j+1]$

$\text{parent}[i, j] = (i, j+1)$



- ... and follow them at the end:

$\text{lcs} = []$

$\text{here} = (0, 0)$

while  $c[\text{here}]$ :

if  $x[i] == y[j]$ :

$\text{lcs.append}(x[i])$

$\text{here} = \text{parent}[\text{here}]$

