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6.006 Introduction to Algorithms Spring 2008

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Lecture 21: Dynamic Programming III: Text Justification, Parenthesization, Knapsack, Pseudopolynomial Time, Tetris Training

Lecture Overview

- Text Justification
- Parenthesization
- \bullet Knapsack
- Pseudopolynomial Time
- Tetris Training

Readings

CLRS 15

Review:

- * DP is all about subproblems & guessing
- * 5 easy steps:
 - (a) define subproblems: count # subprobs.
 - (b) guess (part of solution): count # choices
 - (c) relate subprob. solutions: compute time/subprob.
 - (d) recurse + memoize <u>OR</u> build DP table bottom up: time = time/subprob. ×# subprobs (check subproblems related acyclically)
 - (e) check original problem = a subproblem or solvable from DP table (\Longrightarrow extra time)
- * for sequences, good subproblems are often prefixes OR suffixes OR substrings

Text Justification:

Split text into "good lines"

- obvious (MS Word/Open Office) algorithm: put as many words fit on first line, repeat
- but this can make very bad lines

```
blah blah blah blah

∴ b l a h vs. blah blah

reallylongword reallylongword
```

Figure 1: Good vs. Bad Justification

• define $\underline{\text{badness}}(i,j)$ for line of words [i:j] e.g.,

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 \left\{ \begin{array}{l} \mbox{if total length} > \mbox{page width} \\ \mbox{(page width - total length)}^3 \mbox{ else} \end{array} \right.
```

- \bullet goal: split words into lines to min \sum badness
- 1. subproblem = min badness for suffix words[i:] $\implies \sharp$ subproblems = $\Theta(n)$ where $n = \sharp$ words
- 2. guessing = where to end first line, say i:j $\implies \sharp \text{ choices} = n i = O(n)$
- 3. relation:
 - DP[i] = min(badness(i, j) + DP[j] for j in range(i + 1, n + 1))
 - $DP[n] = \phi$ \implies time per subproblem = O(n)
- 4. total time = $O(n^2)$
- 5. solution = $DP[\phi]$ (& use parent pointers to recover split)

Parenthesization:

Optimal evaluation of associative expression - e.g., multiplying rectangular matrices



Figure 2: Evaluation of an Expression

```
 \begin{array}{l} 2. \  \, \mathrm{guessing} = \mathrm{outermost} \,\, \mathrm{multiplication} \,\, \underbrace{\hspace{1cm} \big( \dots \big) \big( \dots \big)}_{\uparrow_k} \\ \Longrightarrow \, \sharp \,\, \mathrm{choices} = O(n) \\ 1. \,\, \mathrm{subproblems} = \frac{\mathrm{prefixes} \,\, \& \,\, \mathrm{suffixes?} \,\, \mathrm{NO}}{= \,\, \mathrm{cost} \,\, \mathrm{of} \,\, \mathrm{substring} \,\, A[i:j]} \\ \Longrightarrow \, \sharp \,\, \mathrm{subproblems} = \Theta(n^2) \\ \end{array}
```

- 3. Relation:
 - $DP[i, j] = min(DP[i, k] + DP[k, j] + cost of multiplying <math>(A[i] \cdots A[k-1])$ by $(A[k] \cdots A[j-1])$ for k in range(i+1, j)
 - $DP[i, i+1] = \phi$ \implies cost per subproblem = O(n)
- 4. total time = $O(n^3)$
- 5. solution = DP[0, n] (& use parent pointers to recover parens.)

Knapsack:

Knapsack of size S you want to pack

- item i has integer size s_i & real value v_i
- goal: choose subset of items of maximum total value subject to total size $\leq S$

First Attempt:

- 1. subproblem = value for suffix i: WRONG
- 2. guessing = whether to include item $i \implies \sharp$ choices = 2
- 3. relation:

- $DP[i] = \max(DP[i+1], v_i + DP[i+1] \text{ if } s_i \leq S?!)$
- not enough information to know whether item *i* fits how much space is left? GUESS!
- 1. subproblem = value for suffix i: $\underline{\text{given knapsack of size } X}$ $\implies \sharp \text{ subproblems} = O(nS)$
- 3. relation:
 - $DP[i, X] = \max(DP[i+1, X], v_i + DP[i+1, X-s_i] \text{ if } s_i \le X)$
 - $DP[n, X] = \phi$ \implies time per subproblem = O(1)
- 4. total time = O(nS)
- 5. solution = $DP[\phi, S]$ (& use parent pointers to recover subset) AMAZING: effectively trying all possible subsets!

Knapsack is in fact NP-complete! \implies suspect no <u>polynomial-time</u> algorithm (polynomial in length of input).

What gives?

- here input = $\langle S, s_0, \dots, s_{n-1}, v_0, \dots, v_{n-1} \rangle$
- length in binary: $O(\lg S + \lg s_0 + \cdots) \approx O(n \lg \ldots)$
- so O(nS) is not "polynomial-time"
- O(nS) still pretty good if S is small
- "pseudopolynomial time": polynomial in length of input & integers in the input

Remember:
polynomial - GOOD
exponential - BAD
pseudopoly - SO SO



Figure 3: Tetris

Tetris Training:

- given sequence of n Tetris pieces & a board of small width w
- must choose orientation & x coordinate for each
- then must drop piece till it hits something
- full rows do not clear without these artificialities WE DON'T KNOW! (but: if w large then NP-complete)
- goal: survive i.e., stay within height h

[material below covered in recitation]

First Attempt:

- 1. subproblem = survive in suffix i:? WRONG
- 2. guessing = how to drop piece $i \implies \sharp \text{ choices} = O(w)$
- 3. relation: DP[i] = DP[i+1]?! not enough information! What do we need to know about prefix: i?
 - 1. subproblem = survive? in suffix i: given initial column occupancies $h_0, h_1, \cdots, h_{w-1}$ $\implies \sharp \text{ subproblems} = O(n \cdot h^w)$
 - 3. relation: $DP[i, h] = \max(DP[i, m])$ for valid moves m of piece i in h) \implies time per subproblem = O(w)
 - 4. total time = $O(nwh^w)$
 - 5. solution = $DP[\phi, \overline{\phi}]$ (& use parent pointers to recover moves)