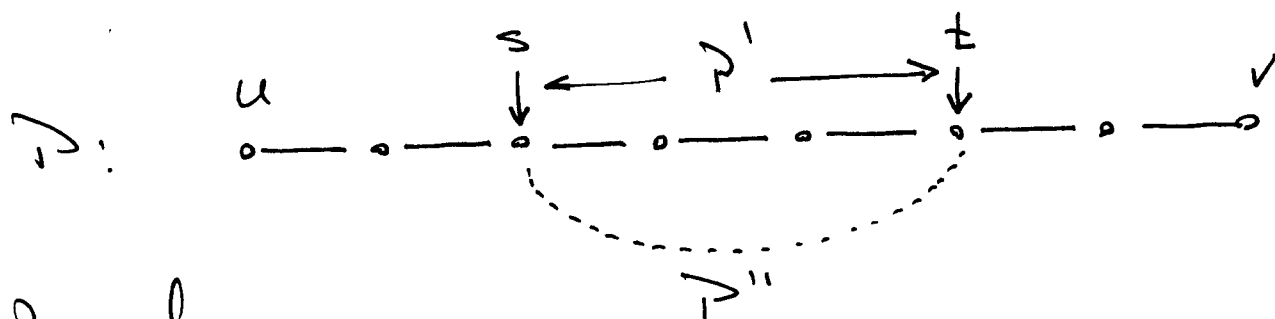


continue: shortest & longest Paths

Thm

Any subsequence of a shortest path is also a shortest path.



Proof.

Let P be a shortest $u-v$ path in G .
 Let s, t be any two intermediate vertices along P . Let P' be the subpath of P joining s to t . must show P' is a shortest $s-t$ path.

Assume, to get a contradiction, that P' is not a shortest s - t Path.

Then there must exist a shorter s - t Path in G , call it P'' .

Let P''' be the u - v Path obtained by concatenating 3 Paths

- u to s along P : P_1
- s to t along P'' : P''
- t to v along P : P_2

Then

$$\begin{aligned} \text{length}(P''') &= \text{length}(P_1) + \text{length}(P'') + \text{length}(P_2) \\ &< \text{length}(P_1) + \text{length}(P') + \text{length}(P_2) \\ &= \text{length}(P), \end{aligned}$$

Contradicting that P is a shortest u - v Path.

Therefore, no such shorter s - t Path P'' exists, and hence P' is a shortest s - t Path, as claimed. ■

what about longest Paths?

Ex. P. 7 of Dyn. Prog. handout.

A sub Path of a longest Path is not necessarily a longest Path.

so

Problem 1: shortest Path satisfies Principle of optimality, and

Problem 2: longest Path does not.

Summary of Dyn. Prog. Procedure

1. characterize the structure of an opt. solution as consisting of optimal sub-instance solutions.
2. Recursively define an optimal soln. in terms of opt. subinstance solns, i.e. write a recurrence formula for value of opt. solution.
3. compute value of opt. solution in a bottom-up fashion, i.e. fill in a table giving values of sub-instance solns.
4. construct an opt. soln. recursively by backtracking through table.

Problem : Matrix Chain multiplication

Defn a matrix chain is a Product

$$A_1 \cdot A_2 \cdot A_3 \cdot A_4 \cdots A_{n-1} \cdot A_n$$

$$\begin{array}{ccccccc} \uparrow & \uparrow & \uparrow & & \uparrow & \uparrow & \\ P_0 \times P_1 & P_1 \times P_2 & P_2 \times P_3 & & P_{n-2} \times P_{n-1} & P_{n-1} \times P_n & \end{array}$$

that is well defined, i.e. # col. in one matrix is eq. to # rows in next matrix.

Ex $n=3$: $A B C = A \cdot (B \cdot C) = (A \cdot B) C$

$$\begin{array}{ccccccc} & \nearrow & \uparrow & \uparrow & \uparrow & & \\ & P \times q & q \times r & r \times s & P \times q & q \times s & P \times r \quad r \times s \end{array}$$

$$[\text{cost of } A \cdot B] = P \cdot q \cdot r \quad \text{real no. mults.}$$

$$[\text{cost of } B \cdot C] = q \cdot r \cdot s \quad " \quad " \quad "$$

$$[\text{cost of } A(BC)] = Pqs + qrs \quad (1)$$

$$[\text{cost of } (AB)C] = Pqr + prs \quad (2)$$

Let $P=10, q=100, r=10, s=100$

(1) gives : 200,000

(2) gives : 20,000

Problem given a matrix chain of size n , find its optimal Parenthesization

Exercise

- write the $\leq$ Paren. S. in case $n=4$
- determine # of Paren. S. if $n=5$, then write them.

Exercise.

Let $P(n) = \#$ of Paren. s. of a matrix chain of len. n .

show $P(n)$ satisfies

$$P(n) = \begin{cases} 1 & n=1 \\ \sum_{k=1}^{n-1} P(k) \cdot P(n-k) & n \geq 2 \end{cases}$$

check:

n	1	2	3	4	5	...
$P(n)$	1	1	2	5	?	

Defn $C_n = P(n+1)$ are called the Catalan numbers.

Fact. $P(n) = \frac{1}{n} \cdot \binom{2n-2}{n-1} = \frac{1}{n} \binom{2(n-1)}{n-1}$

Exercise (hard) Prove this

By Stirling's formula

$$T(n) = \Theta\left(\frac{4^n}{n^{3/2}}\right)$$

↑
exponential.

so Brute force is intractable.

Dyn. Prog. approach

sub-problem: optimally parenthesize

$$A_i \dots A_j$$

for $1 \leq i \leq j \leq n$.

Any Paren. of $A_i \dots A_j$ splits into two chains

$$(A_i \dots A_k) \cdot (A_{k+1} \dots A_j)$$

for some k : $i \leq k < j$

claim

If $A_1 \dots A_j$ is optimally Paren., then
 so are both factors $A_1 \dots A_k$ and
 $A_{k+1} \dots A_j$

$$\begin{array}{c}
 \underbrace{A_1 \dots A_j}_{\text{if opt.}} = \overbrace{(A_1 \dots A_k) \cdot (A_{k+1} \dots A_j)}^{P_{1..k} \times P_{k+1..j}} \\
 \underbrace{\hspace{10em}}_{\text{opt.}} \quad \underbrace{\hspace{10em}}_{\text{opt.}} \\
 \text{then } \nearrow
 \end{array}$$

Proof.

Assume, to get a $\cdot X \cdot$, that $A_1 \dots A_k$
 is not optimally Parenthesized. Then
 we can Replace the Paren. of $A_1 \dots A_k$
 by a better one, yielding a Paren.
 of $A_1 \dots A_j$ better than optimal $\cdot X \cdot$.
 $\therefore A_1 \dots A_k$ must be opt. Parenthesized.
 Similarly for $A_{k+1} \dots A_j$. ~~■~~

define (for $1 \leq i \leq j \leq n$)

$$m[i, j] = \min \text{ cost of a Parenthetization} \\ \text{of } A_i \dots A_j$$

goal find $m[1, n]$

we see from Prev. Proof:

$$m[i, j] = m[i, k] + m[k+1, j] + p_{i-1} p_k p_j$$

we don't know what k is, so

$$m[i, j] = \begin{cases} 0 & i = j \\ \min_{i \leq k < j} (m[i, k] + m[k+1, j] + p_{i-1} p_k p_j) & i < j \end{cases}$$