

Defn

let G be a graph, $u, v \in V(G)$.

a $u-v$ Path is a sequence of alternating vertices & edges

$$P: u = x_0, e_1, x_1, e_2, x_3, \dots, e_k, x_k = v$$

where the e 's are no repeated edges,
& no repeated vertices (unless $u=v$, called a closed path.)

$$\text{length}(P) = k = \# \text{ edges}$$

notation

$d(u, v)$ = length of a min-length $u-v$ Path

$l(u, v)$ = length of a max-length (longest) $u-v$ Path.

Problem 1

Given G , $u, v \in V(G)$: determine $d(u, v)$ and a Path of length $d(u, v)$.

Problem 2

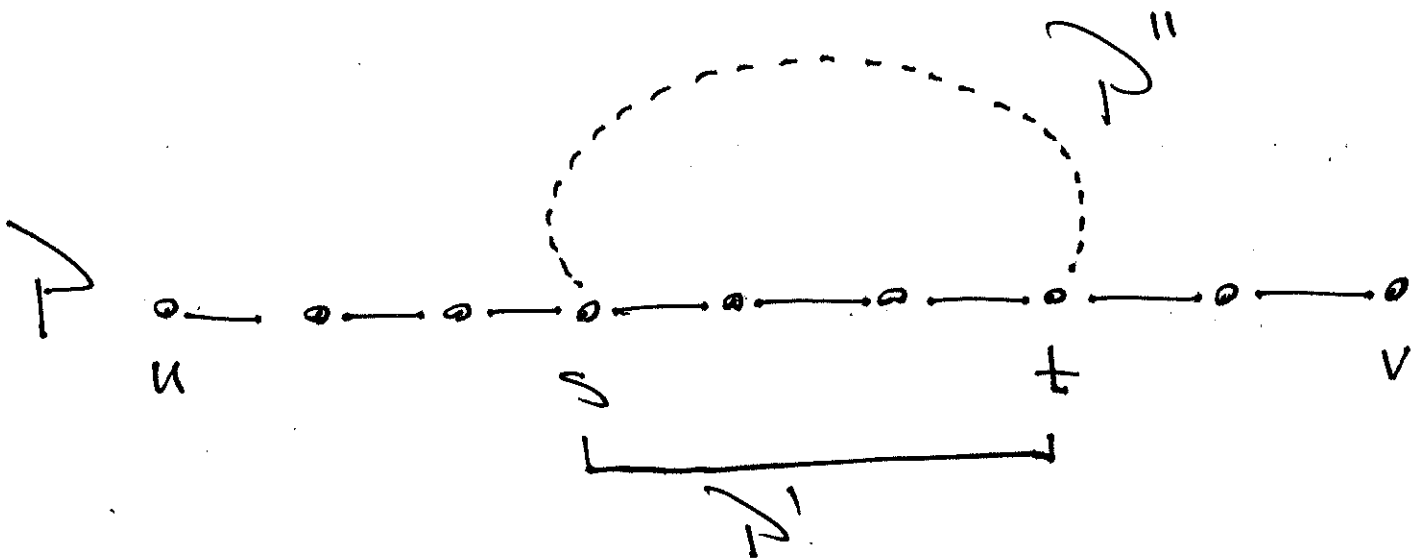
Given G , $u, v \in V(G)$: determine $l(u, v)$ and a Path of length $l(u, v)$.

Claim

Any subsequence of a shortest path is also a shortest path.
(note: converse is false!)

Proof.

let \mathcal{P} be a shortest $u-v$ path in G . let s, t be two intermediate vertices on \mathcal{P}



let \mathcal{P}' be sub-seg. from s to t .
must show \mathcal{P}' is a shortest $s-t$ path.

Assume, to get a X_i , that P^1 is not a shortest s - t Path, Then there is a shorter one, call it P'' i.e. $\text{len}(P'') < \text{len}(P^1)$.

Then consider sequence

$$P_{us} + P''_{st} + P_{tv} \dots$$

(may not be a Path, but is a walk.)

we have

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$$\begin{aligned} & \text{len}(P_{us} + P''_{st} + P_{tv}) \\ &= \text{len}(P_{us}) + \text{len}(P''_{st}) + \text{len}(P_{tv}) \\ &< \text{len}(P_{us}) + \text{len}(P'_{st}) + \text{len}(P_{tv}) \\ &= \text{len}(P_{us} + P'_{st} + P_{tv}) \\ &= \text{len}(P) \end{aligned}$$

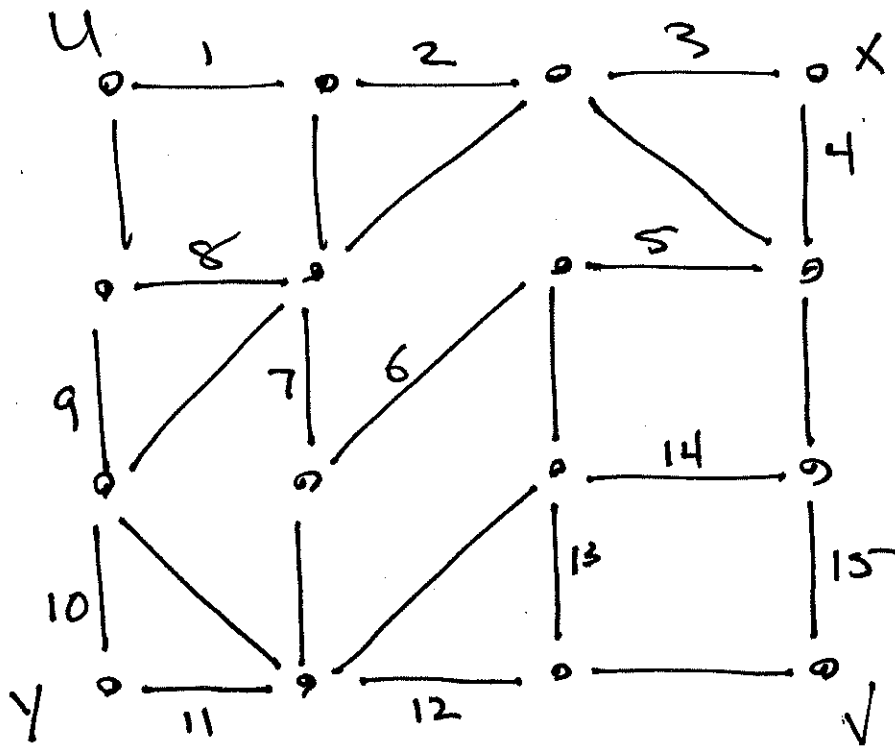
contradicting that P is a shortest $u-v$ path. So this \times shows no such path P'' can exist. $\therefore P'$ is a shortest $s-t$ path. \blacksquare

So Problem 1 "exhibits optimal sub-structure."

Problem 2 does not!

$$d(u, v) = 5$$

Ex.



$$l(u, v) = 15 \quad \text{but} \quad l(x, y) \geq 14$$

so sub-seq. of a longest path is not necessarily a longest path,

Procedure for designing a Dyn. Programming algorithm.

1. characterize the optimal sub-structure, i.e. show that solving entails making a seq. of choices, each leaving a sub-instance.
2. Recursively define value of optimal soln in terms of optimal sub-instance solutions (write a recursive formula)
3. compute value of an optimal soln in a bottom-up fashion i.e. fill in a table.

4. Construct an optimal solution by backtracking through the table.

Problem Matrix. Chain multiplication

$$A_1 A_2 A_3 \dots A_n$$

start with : $A \cdot B$

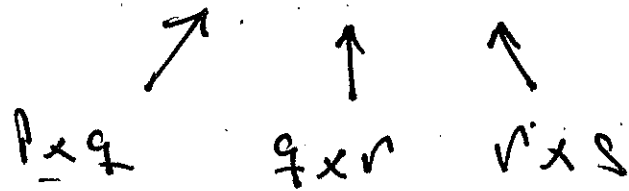
for $\begin{pmatrix} 1 \leq i \leq p \\ 1 \leq j \leq r \end{pmatrix}$ $p \times q \quad q \times r$

$$(A \cdot B)_{ij} = \sum_{k=1}^q a_{ik} \cdot b_{kj} \quad \left(\begin{matrix} q \text{ basis} \\ \text{ops.} \end{matrix} \right)$$

basic operation : real multiplication

$$\text{Total cost} = pqr$$

What is cost of $A \cdot B \cdot C$



$$(1) A(BC) : Pqs + qrs$$

or

$$(2) (AB)C : Pqr + Prs$$

Ex. $P = 10, q = 100, r = 10, S = 100$

Then

$$(1) \text{ here cost } 200,000$$

$$(2) \text{ " " } 20,000$$

Exercise

write the 5 different
 Parenthesizations of: ABCD

Exercise

let $P(n) = \#$ of Parenthesizations
 of $A_1 A_2 \dots A_n$. Show that

$P(n)$ satisfies

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

Hard exercise! show

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

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note: $C_n = \rightarrow(n+1)$ is
called Catalan Numbers.

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

Problem:

Given $n \geq 1$ and a matrix-
chain product $A_1 A_2 \dots A_n$,
where

A_i has size $p_{i-1} \times p_i$ ($1 \leq i \leq n$)

determine a Parenthesization
that minimizes total # of
basic ops.

Actually 2 problems

- o value at opt. soln, i.e. min
Poss. # of real multiplications
- o opt. soln itself! a Parenthesization
that achieves min.

Typical Subinstance

$$A_i \cdots A_j$$

where $1 \leq i \leq j \leq n$

Note: any opt. Paren. of
 $A_i \cdots A_j$ has a first
 split point k :

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

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

claim:

\Rightarrow If $A_i \dots A_j$ is optimally parenthesized, then so are both $A_i \dots A_k$ and $A_{k+1} \dots A_j$

$$\underbrace{A_i \dots A_j}_{\text{if optimal}} = \underbrace{(A_i \dots A_k)}_{\text{optimal}} \cdot \underbrace{(A_{k+1} \dots A_j)}_{\text{optimal}}$$

then both optimal.

Proof.

Assume, to get a $\cdot X$, that $(A_i \dots A_k)$ is not actually parenthesized. Then there is a better paren. than the one inherited from the opt. paren. of $A_i \dots A_j$. Substitute that better parenth. of $(A_i \dots A_k)$ to get a better paren. of $A_i \dots A_j$, which is impossible. \blacksquare

Exercise show converse
is false, i.e. Concatenation
of Opt. Parns. is not necessarily
Optimal.

Let $m[i, j] = \min \#$ of real multiplications
needed to compute
 $A_i \cdots A_j$

where $1 \leq i \leq j \leq n$ or

Goal is to find $m[1, n]$.

note: if $i = j$, then initialize

$$m[i, i] = 0$$

By Proof ...

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