

TO DEVELOP A DYNAMIC PROGRAMMING SOLUTION,  
WE CONSIDER THE GENERAL SUBPROBLEM OF  
FINDING AN OPTIMAL PARENTHESIZATION OF  
 $A_i \dots A_j$  WHERE  $1 \leq i \leq j \leq n$ .

OBSEVE THAT AN OPTIMAL PARENTHESIZATION  
SPLITS  $A_i \dots A_j$  INTO

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

FOR SOME  $k$  ( $i \leq k < j$ ).

NOTE ALSO THAT IF  $A_i \dots A_j$  IS OPTIMALLY  
PARENTHESIZED, THEN SO ARE BOTH  $(A_i \dots A_k)$   
AND  $(A_{k+1} \dots A_j)$ .

PROOF: IF THE PARENTHESIZATION OF  $(A_i \dots A_k)$   
IS NOT OPTIMAL, THEN WE CAN REPLACE IT  
WITH AN OPTIMAL ONE YIELDING A PARENTHESIZATION  
OF  $A_i \dots A_k$  WITH FEWER SCALAR MULTIPLICATIONS.

THIS CONTRADICTS THAT OUR ORIGINAL PARENTHESIZATION  
OF  $A_i \dots A_k$  WAS OPTIMAL.

11

THEREFORE THE PRINCIPLE OF OPTIMALITY IS SATISFIED  
IN THIS PROBLEM.

$$\begin{array}{c} \text{optimal} & \text{optimal} \\ (A_i \dots A_k) \cdot (A_{k+1} \dots A_j) & \text{optimal} \\ (P_{i-1} \times P_k) & (P_k \times P_j) \end{array}$$

LET  $m[i, j]$  DENOTE THE MINIMUM NUMBER OF SCALAR MULTIPLICATIONS NECESSARY TO COMPUTE  $A_i \dots A_j$ . IF  $i = j$  THEN  $m[i, i] = 0$  SINCE THE PRODUCT CONSISTS OF JUST ONE MATRIX.

IF  $i \leq k < j$ , AND  $k$  IS THE SPLIT POSITION OF AN OPTIMAL PARENTHESIZATION THEN

$$m[i, j] = m[i, k] + m[k+1, j] + P_{i-1} P_k P_j$$

SINCE WE DON'T KNOW  $k$ , WE DEFINE IN GENERAL

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

Ex.  $n=5$ ;  $P_0=10, P_1=20, P_2=30, P_3=10, P_4=40, P_5=10$

TABLE M

	1	2	3	4	5
1	0	6000	8000	12000	13000
2	x	0	6000	14000	12000
3	x	x	0	12000	7000
4	x	x	x	0	4000
5	x	x	x	x	0

For instance, to compute  $m[2,5]$ :

$$m[2,5] = \min \begin{cases} m[2,2] + m[3,5] + P_1 P_2 P_5 = 13000 & (k=2) \\ m[2,3] + m[4,5] + P_1 P_3 P_5 = 12000 & (k=3) \\ m[2,4] + m[5,5] + P_1 P_4 P_5 = 22000 & (k=4) \end{cases}$$

Observe that to compute  $m[2,5]$  one needs entries  $m[2,2..4]$  and  $m[3..5,5]$ . Thus to fill the table, first initialize the main diagonal to 0, then successively fill each off diagonal above the main.

$$\text{i.e. } m[i,i] = 0 \quad (1 \leq i \leq n)$$

$$\text{then } m[i, i+1] \quad (1 \leq i \leq n-1)$$

$$m[i, i+2] \quad (1 \leq i \leq n-2)$$

and in general

$$m[i, i+l] \quad (1 \leq i \leq n-l)$$

for  $l = 1, 2, \dots, n-1$

SEE P. 336 FOR PSEUDO-CODE

FROM THIS TABLE WE CAN RE-CONSTRUCT THE VALUES  $K$ , WHICH GIVE THE SPLIT POINTS FOR EACH SUBPROBLEM.

A MORE EFFICIENT APPROACH IS TO STORE  $K$  VALUES IN A PARALLEL TABLE  $S[i,j]$  AS WE CONSTRUCT  $m[i,j]$ . (P.336).

Ex. Same as above. We see  $S[2,5] = 3$ , AND

TABLE 3

	1	2	3	4	5
1	x	1	1	3	3
2	x	x	2	3	3
3	x	x	x	3	3
4	x	x	x	x	4
5	x	x	x	x	x

From this table we can construct the optimal parenthesization:

$$(A_1(A_2A_3))(A_4A_5)$$

## All Pairs Shortest Paths (APSP) (25,2)

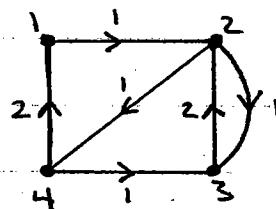
Consider a Directed Graph in which a weight (cost) is assigned to each directed Edge.

We write  $G = (V, E)$  where  $V$  is the vertex set and  $E$  is the set of directed edges.

The adjacency matrix of  $G$  is defined as  $W = (w_{ij})$  where

$$w_{ij} = \begin{cases} 0 & \text{if } i = j \\ \text{weight of dir. edge } (i,j) & \text{if } i \neq j, (i,j) \in E \\ \infty & \text{if } i \neq j, (i,j) \notin E \end{cases}$$

Ex.



$$V = \{1, 2, 3, 4\}$$

$$E = \{(1,2), (2,4), (4,1), (4,3), (3,2), (2,3)\}$$

$$W = \begin{pmatrix} 0 & 1 & \infty & \infty \\ \infty & 0 & 1 & 1 \\ \infty & 2 & 0 & \infty \\ 2 & \infty & 1 & 0 \end{pmatrix}$$

THE WEIGHT OF A DIRECTED  $i$ - $j$  PATH ( $i, j \in V$ ) IS THE SUM OF THE WEIGHTS OF EACH OF ITS DIRECTED EDGES.

Problem: (APP)

FOR EACH PAIR  $(i, j) \in V \times V$ , DETERMINE AN  $i$ - $j$  PATH OF MINIMUM WEIGHT. (ALSO CALLED A SHORTEST PATH.)

AGAIN THERE ARE REALLY TWO PROBLEMS

- DETERMINE THE MINIMUM PATH WEIGHTS FOR EACH  $(i, j)$
- DETERMINE SHORTEST  $i$ - $j$  PATHS.

WE CONCENTRATE ON THE FIRST PROBLEM, LEAVING THE SECOND AS AN EXERCISE.

### Floyd-Warshall Algorithm

AN INTERMEDIATE VERTEX OF A DIRECTED PATH  $P = (v_1, v_2, \dots, v_e)$  IS ANY VERTEX OTHER THAN  $v_1$  OR  $v_e$ , i.e. ONE OF THE VERTICES  $\{v_2, \dots, v_{e-1}\}$ .

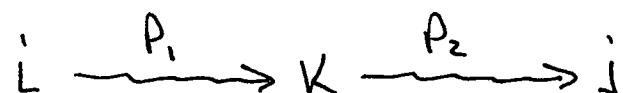
LET  $G = (V, E)$  BE A DIRECTED GRAPH  
WITH  $V = \{1, 2, \dots, n\}$ . DEFINE SUBSETS  
 $V_K$  OF  $V$  AS FOLLOWS

$$V_K = \begin{cases} \emptyset & K=0 \\ \{1, 2, \dots, K\} & 1 \leq K \leq n \end{cases}$$

LET  $(i, j) \in V \times V$  AND  $1 \leq k \leq n$ . LET  $P$   
DENOTE A MINIMUM WEIGHT PATH AMONGST  
ALL  $i-j$  PATHS WITH INTERMEDIATE VERTEXES  
IN  $V_K$ .

NOW OBSERVE THAT WE HAVE TWO ALTERNATIVES

- $K$  IS NOT AN INTERMEDIATE VERTEX OF  $P$ .  
IN THIS CASE  $P$  IS ALSO OF MINIMUM WEIGHT  
AMONGST ALL  $i-j$  PATHS WITH INTERMEDIATE  
VERTEXES IN  $V_{K-1}$ .
- $K$  IS AN INTERMEDIATE VERTEX OF  $P$ . WE  
CAN DECOMPOSE  $P$  INTO SUBPATHS  $P_1$  AND  $P_2$ :



NOTE VERTEX  $K$  IS NOT INTERMEDIATE TO  
EITHER  $P_1$  OR  $P_2$ .

Thus  $P_1$  was minimum weight amongst all  $i-k$  paths with intermediate vertices in  $V_{k-1}$ , and likewise  $P_2$  was minimum weight amongst all  $k-i$  paths with intermediate vertices in  $V_{k-1}$ .

These observations show ADP exhibits optimal substructure, necessary for dynamic programming.

Let  $d_{ij}^{(k)}$  denote the weight of a minimum weight  $i-j$  path with all intermediate vertices in  $V_k$ .

When  $k=0$ , such a path has no intermediate vertices, hence at most one edge. Thus  $d_{ii}^{(0)} = w_{ii}$ .

The above observations show that for  $1 \leq k \leq n$  we have

$$d_{ij}^{(k)} = \min(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}).$$

Let  $D^{(k)}$  denote the matrix  $(d_{ij}^{(k)})$ . Then we seek  $D^{(n)}$  given  $D^{(0)} = W$ .

## Floyd-Warshall (W)

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1.)  $n \leftarrow \text{Rowel}[W]$ 
2.)  $D^{(0)} \leftarrow W$ 
3.) for  $k \leftarrow 1$  to  $n$ 
4.)   for  $i \leftarrow 1$  to  $n$ 
5.)     for  $j \leftarrow 1$  to  $n$ 
6.)        $d_{ij}^{(k)} \leftarrow \min(d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)})$ 
7.) return  $D^{(n)}$ 

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Since (6) takes time  $O(1)$ , Floyd-Warshall runs in time  $\Theta(n^3)$ .

NOTE The above algorithm also uses memory  $n^3$ . It is possible to accomplish this with just  $n^2$  memory (Exercise.)

To construct shortest paths we could use  $D = D^{(n)}$  to determine the predecessor matrix  $\Pi = (\pi_{ij})$ , where

$\pi_{ij} = \text{Predecessor of } i \text{ along a shortest } i-j \text{ path}$

Alternatively we could determine intermediate predecessor matrices  $\Pi^{(k)} = (\pi_{ij}^{(k)})$  ( $0 \leq k \leq n$ )

$\pi_{ij}^{(k)}$  = Predecessor of  $i$  along a shortest  $i-j$  path amongst those with intermediate vertices in  $V_k$ .

(SEE P. 632 FOR DETAILS.)

Exercise

- RUN FLOYD-WARSHALL ON THE WEIGHTED DIGRAPH  
IN PRECEDING EXAMPLE.
- WRITE AN ALGORITHM TO DETERMINE  $\overline{\Pi}$  FROM  
 $D = D^{(n)}$
- ALTER FLOYD-WARSHALL TO BUILD  $\overline{\Pi}^{(k)}$  ( $0 \leq k \leq n$ )  
AS YOU GO.
- WRITE AN ALGORITHM TO PRINT A SHORTEST  $i-j$   
PATH GIVEN  $\overline{\Pi} = \overline{\Pi}^{(n)}$ .

READ

- LONGEST COMMON SUBSEQUENCE (15.4)
- OPTIMAL BINARY SEARCH TREES (15.5)