

Dynamic Programming

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Dynamic Programming

- Dynamic Programming, like the divide-and-conquer method, solves problems by combining the solutions to sub-problems.
- Pure divide-and-conquer:
 - divides problems into independent sub-problems,
 - solves the sub-problem recursively, and then,
 - combines their solutions to solve the original sub-problem.
- Dynamic programming in contrast is used when the sub-problems are not independent, that is sub-problems share sub-problems.
- It is typically applied to optimization problems.



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Example: Matrix Chain Multiplication

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Matrix Multiplication

```
MATRIX-MULTIPLY(A, B)
1 if columns[A] ≠ rows[B]
2   then error “incompatible dimensions”
3 else for i ← 1 to rows[A]
4     do for j ← 1 to columns[B]
5       do C[i, j] ← 0
6         for k ← 1 to columns[A]
7           do C[i, j] ← C[i, j] + A[i, k] · B[k, j]
8   return C
```

- Cost of multiplying $A[p][q] \times B[q][r]$ is $p \cdot q \cdot r$
- What is the cost of multiplying three matrices A , B , and C of sizes 10×100 , 100×5 , and 5×50 ?
- How to find the best way of multiplying?



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Matrix Chain Multiplication

- Given a chain $A_1, A_2, A_3, \dots, A_n$ of n matrices, such that A_i has dimension $p_{i-1} \times p_i$, find the sequence of multiplication that will result in minimum number of scalar multiplication.
 - Recursive Cost Function Catalan numbers:

$$(A_1 \cdot (A_2 \cdot (A_3 \cdot A_4)))$$

$$((A_1 \cdot A_2) \cdot (A_3 \cdot A_4))$$

$$((A_1 \cdot (A_2 \cdot A_3)) \cdot A_4)$$

$$p(n) = \begin{cases} 1 & \text{if } n = 1 \\ \sum_{k=1}^{n-1} p(k) \cdot P(n-k), & \text{if } n > 1 \end{cases}$$

$$P(n+1) = \Omega\left(\frac{4^n}{n^2}\right)$$



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Observations

- Existence of Optimal Substructure: In an optimum sequence of decisions, each subsequence must also be optimum.
- $(A_1 A_2 A_3) \cdot (A_4 \cdot A_5 \cdot A_6)$
 - Total cost is $C(1..3) + C(4..6) +$ cost of multiplying the two final matrices.
- Recursive Solution Possible: If $m[i,j]$ is the optimum cost of multiplying all matrices between i^{th} and j^{th} matrices $\dots(A_i A_{i+1} \dots A_j) \dots$
 - if $i=j$ then $m[i,j] = 0$
 - otherwise,

$$m[i, j] = \min_{i \leq k < j} \{m[i, k] + m[k+1, j] + p_{i-1} \cdot p_k \cdot p_j\}$$



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Recursive Solution

RECURSIVE-MATRIX-CHAIN(p, i, j)

```
1 if  $i = j$ 
2   then return 0
3  $m[i, j] \leftarrow \infty$ 
4 for  $k \leftarrow i$  to  $j - 1$ 
5   do  $q \leftarrow$  RECURSIVE-MATRIX-CHAIN( $p, i, k$ )
        + RECURSIVE-MATRIX-CHAIN( $p, k + 1, j$ ) +  $p_{i-1}p_kp_j$ 
6     if  $q < m[i, j]$ 
7       then  $m[i, j] \leftarrow q$ 
8 return  $m[i, j]$ 
```

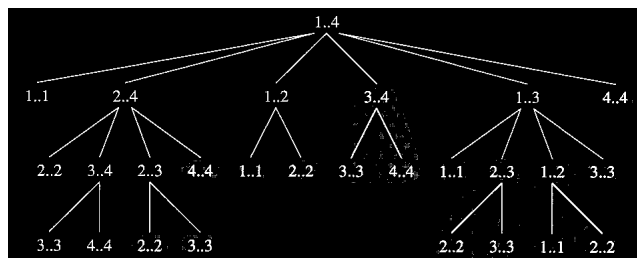
- Running time is exponential $O(2^n)$!



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Observation-2



- Existence of Overlapping Sub-problem: the same sub-sequence is part of many super sequences.
- For a string of limited size, the actual number of subproblems are quite small. $O(n^2)$ only!



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Dynamic Programming Solution: A Bottom up approach

- Compute the optimum cost for multiplying all matrix chains of size 2.
- Store them in a matrix $m[i,j]$, when $i-j$ spans two matrices.
- Use the above values to compute optimum cost for multiplying all matrix chains of size 3.
- Then size 4 .. Up to size n .



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Algorithm

```
MATRIX-CHAIN-ORDER( $p$ )
1   $n \leftarrow \text{length}[p] - 1$ 
2  for  $i \leftarrow 1$  to  $n$ 
3      do  $m[i, i] \leftarrow 0$ 
4  for  $l \leftarrow 2$  to  $n$ 
5      do for  $i \leftarrow 1$  to  $n - l + 1$ 
6          do  $j \leftarrow i + l - 1$ 
7               $m[i, j] \leftarrow \infty$ 
8              for  $k \leftarrow i$  to  $j - 1$ 
9                  do  $q \leftarrow m[i, k] + m[k + 1, j] + p_{i-1}p_kp_j$ 
10                     if  $q < m[i, j]$ 
11                         then  $m[i, j] \leftarrow q$ 
12                             $s[i, j] \leftarrow k$ 
13  return  $m$  and  $s$ 
```



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Example

matrix	dimension
A_1	30×35
A_2	35×15
A_3	15×5
A_4	5×10
A_5	10×20
A_6	20×25

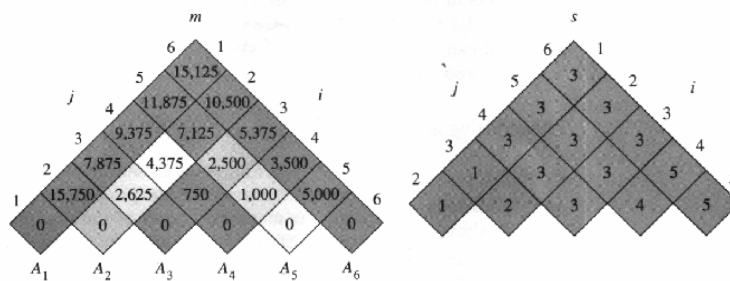


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Example

matrix	dimension
A_1	30×35
A_2	35×15
A_3	15×5
A_4	5×10
A_5	10×20
A_6	20×25



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Constructing the Optimal Solution

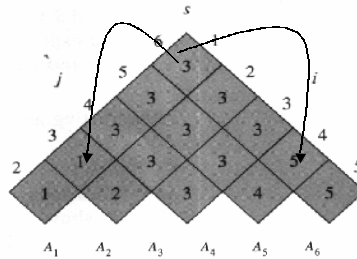
MATRIX-CHAIN-MULTIPLY(A, s, i, j)

```

1  if  $j > i$ 
2    then  $X \leftarrow$  MATRIX-CHAIN-MULTIPLY( $A, s, i, s[i, j]$ )
3          $Y \leftarrow$  MATRIX-CHAIN-MULTIPLY( $A, s, s[i, j] + 1, j$ )
4         return MATRIX-MULTIPLY( $X, Y$ )
5  else return  $A_i$ 

```

In the example of Figure 16.1, the call MATRIX-CHAIN-MULTIPLY($A, s, 1, 6$) computes the matrix-chain product according to the parenthesization $((A_1(A_2A_3))((A_4A_5)A_6))$. (16.3)



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Complexity of Algorithm

MATRIX-CHAIN-ORDER(p)

```

1   $n \leftarrow$  length( $p$ ) - 1
2  for  $i \leftarrow 1$  to  $n$ 
3    do  $m[i, i] \leftarrow 0$ 
4  for  $l \leftarrow 2$  to  $n$ 
5    do for  $i \leftarrow 1$  to  $n - l + 1$ 
6       do  $j \leftarrow i + l - 1$ 
7           $m[i, j] \leftarrow \infty$ 
8          for  $k \leftarrow i$  to  $j - 1$ 
9             do  $q \leftarrow m[i, k] + m[k + 1, j] + p_{i-1}p_kp_j$ 
10              if  $q < m[i, j]$ 
11                 then  $m[i, j] \leftarrow q$ 
12                     $s[i, j] \leftarrow k$ 
13  return  $m$  and  $s$ 

```

- Running time?
- Space?



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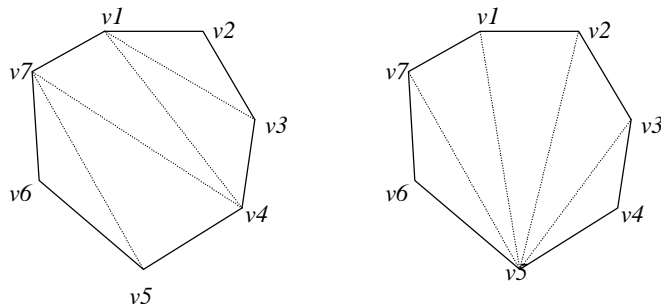
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Example: Optimal Polygon Triangulation

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Polygon Triangulation

- We are given a convex polygon $P = \langle v_0, v_1, \dots, v_{n-1} \rangle$ and a weight function w defined on triangles formed by sides and chords of P . The problem is to find a triangulation that minimizes the sum of the weights of the triangles in the triangulation.



$$w(\Delta v_i v_j v_k) = |v_i v_j| + |v_j v_k| + |v_k v_i|$$

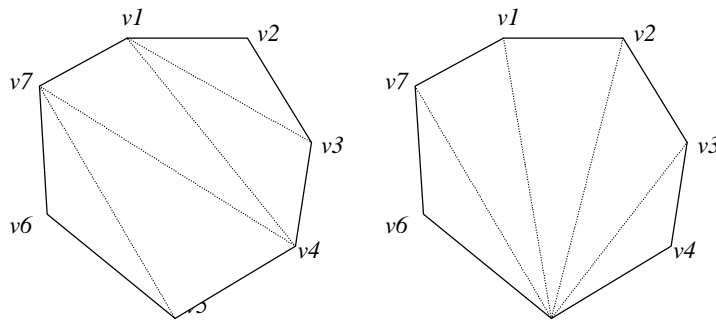


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Observations

- Optimum substructure:
- Overlapping subproblems:



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Dynamic programming Solution

- For all degenerated polygon of size 2, $\langle v_{i-1}, v_i \rangle$ cost = zero.
- For all polygons of size 3 the cost is

$$w(\Delta v_i v_j v_k) = |v_i v_j| + |v_j v_k| + |v_k v_i|$$

- For all polygons of size 4 or more try all division point k and pick the best:

$$t[i, j] = \min_{i \leq k \leq j-1} \{t[i, k] + t[k+1, j] + w(\Delta v_{i-1} v_k v_j)\}$$



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Final Exam

- 4 Questions Total:
 - 1 True-False
 - 1 Overall Concept
 - 1 Tree & Graph
 - 1 String Matching & Dynamic Programming
- Open Book 60 min.
- Project Due/Demo on “Exam Day”
 - Quiz Today or extra 5% goes to project?
 - End Term (20%)
 - Last Assignment (~5%)



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