Problem statement:

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Input: A sequence (chain) of (A_1, A_2, ..., A_n) of matrices, where A_i is of order p_{i-1} \times p_i.
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Output: full parenthesization (ordering) for the product $A_1 \times A_2 \times \cdots \times A_n$ that minimizes the number of (scalar) multiplications.

- Counting the total number of orderings
 - 1. Define $P(n) = \mbox{the number of orderings for a chain of } n \mbox{ matrices}$
 - 2. Then for $n \geq 2$,

$$P(n) = \sum_{k=1}^{n-1} P(k)P(n-k)$$

- and P(1) = 1
- 3. It can be shown that $P(n) = \Omega(2^n)$
- ► A Brute-force solution by exhaustive search for determining the optimal ordering is infeasible!

DP - Step 1: characterize the structure of an optimal ordering

An optimal ordering of the product $A_1A_2\cdots A_n$ splits the product between A_k and A_{k+1} for some k:

$$A_1 A_2 \cdots A_n = A_1 \cdots A_k \times A_{k+1} \cdots A_n$$

and we first compute $A_1 \cdots A_k$ and $A_{k+1} \cdots A_n$, and then multiply them together.

- ▶ Key observation: the ordering of $A_1 \cdots A_k$ within this ("global") optimal ordering must be an optimal ordering of (sub-product) $A_1 \cdots A_k$. ¹
- ▶ Similar observation holds for $A_{k+1} \cdots A_n$
- Thus, an optimal ("global") solution contains within it the optimal ("local") solutions to subproblems. = the optimal substructure property

¹Why? simply argue by contradiction: If there were a less costly way to order the product $A_1 \cdots A_k$, substituting that ordering within this (global) optimal ordering would produce another ordering of $A_1 A_2 \cdots A_n$, whose cost would be less than the optimum, a contradiction!

DP - Step 2: recursively define the value of an optimal solution

Define

 $m[i,j] = \mathsf{min.}$ number of multip. needed to compute $A_i \cdots A_j$.

- ▶ By the definition, $m[1,n] = \text{the cheapest way for the product } A_1A_2\cdots A_n.$
- ightharpoonup m[i,j] can be defined recursively:
 - if i = j, m[i, i] = 0.
 - $\qquad \text{if } i < j \text{, } m[i,j] = m[i, \textcolor{red}{k}] + m[\textcolor{red}{k} + 1, j] + p_{i-1}p_{\textcolor{red}{k}}p_j \text{ for some } \textcolor{red}{k}$

DP - Step 2: recursively define the value of an optimal solution

▶ Thus, for $1 \le i < j \le n$,

$$m[i,j] = \left\{ \begin{array}{ll} 0 & \text{if } i = j \\ \min_{i \leq k < j} \{m[i,k] + m[k+1,j] + p_{i-1}p_kp_j\} & \text{if } i < j \end{array} \right.$$

▶ In addition, to construct an optimal ordering, we keep track

$$s[i,j]=k_{st}=$$
 the value s.t. $m[i,j]$ attains the minimum

DP – Step 3: compute the value of an optimal solution in a bottom-up approach

- ▶ Compute m[i,j] and s[i,j] in a bottom-up approach. (pseudocode next page)
- ▶ Cost: $T(n) = \Theta(n^3)$ since
 - 1. compute roughly $n^2/2$ entries of m-table
 - 2. for each entry of m-table, it finds the minimum of fewer than n expressions.

```
matrix-chain-order(p)
create m[1...n,1...n] and s[1...n,1...n] and n = length(p)-1
for i = 1 to n
 m[i.i] = 0
endfor
for d = 2 to n
  for i = 1 to n-d+1
     j = i + d - 1
     //compute m[i,j]=min_k{...}
     m[i,j] = +infty
     for k = i to j-1
        q = m[i,k] + m[k+1,j] + p[i-1]*p[k]*p[j]
        if q < m[i,j]
           m[i,j] = q
           s[i,j] = k // track k such that min. is attained.
        endif
     endfor
  endfor
endfor
return m and s
```

DP - Step 4: construct an optimal solution from computed information

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Example: Let p = [30 \ 35 \ 15 \ 5 \ 10 \ 20 \ 25]
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matrix-chain-order(p) generates the m-array and s-array:

By s-array, an optimal parenthesization/ordering is given by

$$(A_1 (A_2 A_3)) ((A_4 A_5) A_6)$$