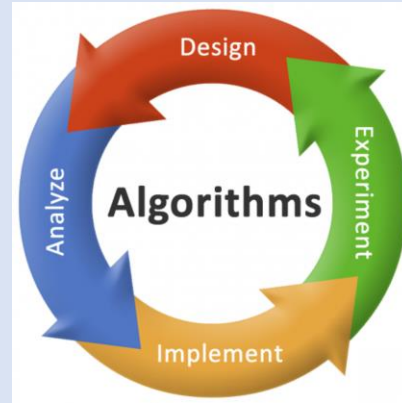


Dynamic Programming

Matrix Chain Multiplication

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The Problem

- Our input is a sequence of n matrices $\langle A_1, A_2, \dots, A_n \rangle$
- The desired output is a multiplication setup $A_1 * A_2 * A_3 * \dots * A_n$
 $((A_1 * A_2) * (A_3 * A_4) * A_5)$
- A product of matrices is fully parenthesized if it is either: one matrix, product of two fully parenthesized matrix products, or surrounded by parentheses
- 5 ways we can parenthesize the above product:
 1. $((A_1 A_2) A_3) A_4$
 2. $((A_1 A_2) (A_3 A_4))$
 3. $(A_1 (A_2 A_3)) A_4$
 4. $(A_1 ((A_2 A_3) A_4))$
 5. $(A_1 (A_2 (A_3 A_4)))$

Multiplying Matrices

Matrix-Multiplication(A,B)

if A.cols not equal B.rows

 return error

for i = 1 to A.rows //p

 for j = 1 to B.cols //r

 c[i,j] = 0

 for k = 1 to A.cols //q

 c[i,j] = c[i,j] + A[i,k] * B[k,j]

return C

Number of Scalar Multiplications is $p * r * q$

So why is this even relative?

- Every multiplication can be costly.
- Type equation here.
- $\langle A_1, A_2, \dots, A_n \rangle$
 - 10 x 100, 100 x 5, 5 x 50
 - $((A_1 A_2) A_3) \rightarrow 5000$ scalar multiplications
 - $(A_1 (A_2 A_3)) \rightarrow 75000$ scalar multiplications
- The objective of this is to MINIMIZE the number of multiplications.

The Matrix-Chain Multiplication

- We are given a chain of n matrices $\langle A_1, A_2, \dots, A_n \rangle$ where $i = 1, 2, \dots, n$ matrix A_i has dimension $p_{i-1} \times p_i$
- The output is a fully parenthesize product of A_1, A_2, \dots, A_n in a way that MINIMIZES the number of scalar multiplications.

Dynamic Programming

- Step 1: Optimal parenthesization of $A_i \dots A_j$ which splits the product between A_k and A_{k+1} must contain within it optimal parenthesization of $A_i \dots A_k$ and $A_{k+1} \dots A_j$
- To optimally parenthesize $A_i \dots A_j$
 - Examine ALL candidates k , $k = i, i + 1, \dots, j - 1$ for splitting
 - Take optimal parenthesization $A_i \dots A_k$ and $A_{k+1} \dots A_j$

Dynamic Programming cont.

- Step 2 is to recursively define an optimal solution
- $m[i, j]$ = minimum of scalar multiplications needed to compute $A_i A_{i+1} \dots A_j$
- $m[i, j] = m[i, k] + m[k + 1, j] + P_{i-1} P_k P_j$
- Table S will store the optimal splitting

Dynamic Programming cont.

- Step 3 Computing the Optimal Answer

MATRIX-CHAIN-ORDER(P)

n = P.length - 1

for i = 1 to n

 m[i,i] = 0

for l = 2 to n

 for i = 1 to n - l + 1

 j = i + l - 1

 m[i,j] = ∞

 for k = i to j - 1

 q = m[i, k] + m[k + 1, j] + $P_{i-1}P_kP_j$

 if q < m[i,j]

 m[i,j] = q

 s[i,j] = k

return m,s

Dynamic Programming cont.

- Step 4 Display the Solution

```
PRINT-OPTIMAL-PARENTHESIS(s,i,j)
```

```
if i == j
```

```
    print Ai
```

```
else
```

```
    print “(”
```

```
    PRINT-OPTIMAL-PARENTHESIS(s,i,s[i,j])
```

```
    PRINT-OPTIMAL-PARENTHESIS(s,s[i,j] + 1,j)
```

```
    print “)”
```

Example