

Strassen's Matrix Multiplication

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[partially based on slides by Prof. Welch]

Matrix Multiplication

Consider two $n \times n$ matrices A and B

Recall that the matrix product $C = AB$ of two $n \times n$ matrices is defined as the $n \times n$ matrix that has the coefficient

$$c_{kl} = \sum_m a_{km} b_{ml}$$

in row k and column l , where the sum ranges over the integers from 1 to n ; the scalar product of the k^{th} row of A with the l^{th} column of B .

The straightforward algorithm uses $O(n^3)$ scalar operations.

Can we do better?

Idea: Use Divide and Conquer

The divide and conquer paradigm is important general technique for designing algorithms. In general, it follows the steps:

- divide the problem into subproblems
- recursively solve the subproblems
- combine solutions to subproblems to get solution to original problem

Divide-and-Conquer

Let write the product $A B = C$ as follows:

$$\begin{array}{|c|c|} \hline A & A \\ \hline A & A \\ \hline \end{array} \times \begin{array}{|c|c|} \hline B & B \\ \hline B & B \\ \hline \end{array} = \begin{array}{|c|c|} \hline A & A \\ \hline A & A \\ \hline \end{array}$$

- Divide matrices A and B into four submatrices each
- We have 8 smaller matrix multiplications and 4 additions. Is it faster?

Divide-and-Conquer

Let us investigate this recursive version of the matrix multiplication.

Since we divide A , B and C into 4 submatrices each, we can compute the resulting matrix C by

- 8 matrix multiplications on the submatrices of A and B ,
- plus $\Theta(n^2)$ scalar operations

Divide-and-Conquer

- Running time of recursive version of straightforward algorithm is

$$T(n) = 8T(n/2) + \Theta(n^2) \text{ and } T(2) = \Theta(1)$$

where $T(n)$ is running time on an $n \times n$ matrix

- Master theorem gives us:

$$T(n) = \Theta(n^3)$$

- Can we do fewer recursive calls (fewer multiplications of the $n/2 \times n/2$ submatrices)?

Strassen's Matrix Multiplication

$$A \times B = C$$

A	A
A	A

 ×

B	B
B	B

 =

	C	
	C	C

$$P_1 = (A_{11} + A_{22})(B_{11} + B_{22})$$

$$P_2 = (A_{21} + A_{22}) * B_{11}$$

$$P_3 = A_{11} * (B_{12} - B_{22})$$

$$P_4 = A_{22} * (B_{21} - B_{11})$$

$$P_5 = (A_{11} + A_{12}) * B_{22}$$

$$P_6 = (A_{21} - A_{11}) * (B_{11} + B_{12})$$

$$P_7 = (A_{12} - A_{22}) * (B_{21} + B_{22})$$

$$C_{11} = P_1 + P_4 - P_5 + P_7$$

$$C_{12} = P_3 + P_5$$

$$C_{21} = P_2 + P_4$$

$$C_{22} = P_1 + P_3 - P_2 + P_6$$

Strassen's Matrix Multiplication

- Strassen found a way to get all the required information with only 7 matrix multiplications, instead of 8.
- Recurrence for new algorithm is
 - $T(n) = 7T(n/2) + \Theta(n^2)$

Solving the Recurrence Relation

Applying the Master Theorem to

$$T(n) = a T(n/b) + f(n)$$

with $a=7$, $b=2$, and $f(n)=\Theta(n^2)$.

Since $f(n) = O(n^{\log_b(a)-\epsilon}) = O(n^{\log_2(7)-\epsilon})$,

case a) applies and we get

$$T(n) = \Theta(n^{\log_b(a)}) = \Theta(n^{\log_2(7)}) = O(n^{2.81}).$$

Discussion of Strassen's Algorithm

- Not always practical
 - constant factor is larger than for naïve method
 - specially designed methods are better on sparse matrices
 - issues of numerical (in)stability
 - recursion uses lots of space
- Not the fastest known method
 - Fastest known is $O(n^{2.3727})$ [Winograd-Coppersmith algorithm improved by V. Williams]
 - Best known lower bound is $\Omega(n^2)$