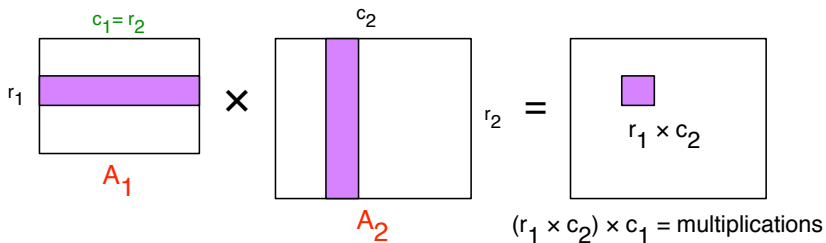


Strassen's Algorithm

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Matrix Multiplication



If $r_1 = c_1 = r_2 = c_2 = N$, this standard approach takes $\Theta(N^3)$:

- ▶ For every row \vec{r} (N of them)
- ▶ For every column \vec{c} (N of them)
- ▶ Take their inner product: $r \cdot c$ using N multiplications

Can we multiply faster than $\Theta(N^3)$?

For simplicity, assume $N = 2^n$ for some n . The multiplication is:

$$\begin{array}{|c|c|} \hline A_{11} & A_{12} \\ \hline A_{21} & A_{22} \\ \hline \end{array} \times \begin{array}{|c|c|} \hline B_{11} & B_{12} \\ \hline B_{21} & B_{22} \\ \hline \end{array} = \begin{array}{|c|c|} \hline C_{11} & C_{12} \\ \hline C_{21} & C_{22} \\ \hline \end{array}$$

$N = 2^n$ (height of first matrix), $N = 2^n$ (width of first matrix)

▶ $C_{11} = A_{11}B_{11} + A_{12}B_{21}$

▶ $C_{12} = A_{11}B_{12} + A_{12}B_{22}$

▶ $C_{21} = A_{21}B_{11} + A_{22}B_{21}$

▶ $C_{22} = A_{21}B_{12} + A_{22}B_{22}$

Uses 8 multiplications

Strassen's Algorithm

$$\begin{array}{|c|c|} \hline A_{11} & A_{12} \\ \hline A_{21} & A_{22} \\ \hline \end{array} \times \begin{array}{|c|c|} \hline B_{11} & B_{12} \\ \hline B_{21} & B_{22} \\ \hline \end{array} = \begin{array}{|c|c|} \hline C_{11} & C_{12} \\ \hline C_{21} & C_{22} \\ \hline \end{array}$$

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

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

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

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

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

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

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

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

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

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

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

Uses only 7 multiplications!

Since the submatrix multiplications are the expensive operations, we save a lot by eliminating one of them.

Apply the above idea recursively to perform the 7 matrix multiplications contained in P_1, \dots, P_7 .

Need to show how much savings this results in overall.

Recurrence

$$T(N) = T(2^n) = \underbrace{7T(2^n/2)}_{\text{recursive } \times} + \underbrace{c4^n}_{\text{additions}}$$

Solving the recurrence:

$$\frac{T(2^n)}{7^n} = \frac{7T(2^{n-1})}{7^n} + \frac{c4^n}{7^n} = \frac{T(2^{n-1})}{7^{n-1}} + \frac{c4^n}{7^n}$$

The **red** term is same as the left-hand side but with $n - 1$, so we can recursively expand:

$$\frac{T(2^n)}{7^n} = \gamma + \sum_{i=1}^n \frac{c4^i}{7^i} = \gamma + c \sum_{i=1}^n \left(\frac{4}{7}\right)^i \leq \alpha \quad \text{for some constants } \alpha, \gamma$$

So:

$$T(2^n) \leq 7^n \alpha = \alpha 2^{n \log_2(7)} = \alpha N^{2.807\dots} = O(N^{2.807\dots})$$

Summary

- ▶ Strassen first to show matrix multiplication can be done faster than $O(N^3)$ time.
- ▶ Strassen's algorithm gives a performance improvement for large-ish N , depending on the architecture, e.g. $N > 100$ or $N > 1000$.
- ▶ Strassen's algorithm isn't optimal though! Over the years it's been improved:

Authors	Year	Runtime
Strassen	1969	$O(N^{2.807})$
⋮		
Coppersmith & Winograd	1990	$O(N^{2.3754})$
Stothers	2010	$O(N^{2.3736})$
Williams	2011	$O(N^{2.3727})$

- ▶ Conjecture: an $O(N^2)$ algorithm exists.

Karatsuba's Algorithm for Integer Multiplication

Integer Multiplication

$$\begin{array}{r} 10101110 \\ \times 01011101 \\ \hline 10101110 \\ 10101110 \\ 10101110 \\ 10101110 \\ 10101110 \\ \hline 11111100110110 \end{array} \left. \vphantom{\begin{array}{r} 10101110 \\ \times 01011101 \\ \hline 10101110 \\ 10101110 \\ 10101110 \\ 10101110 \\ 10101110 \\ \hline 11111100110110 \end{array}} \right\} \begin{array}{l} n \text{ numbers of } n \text{ bits each} \\ O(n^2)\text{-time} \end{array}$$

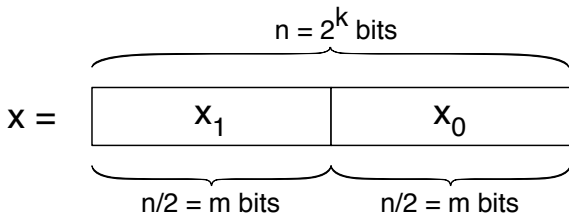
Start similar to Strassen's algorithm, breaking the items into blocks ($m = n/2$):

- ▶ $x = x_1 2^m + x_0$
- ▶ $y = y_1 2^m + y_0$

Then:

$$xy = (x_1 2^m + x_0)(y_1 2^m + y_0) = x_1 y_1 2^{2m} + (x_1 y_0 + x_0 y_1) 2^m + x_0 y_0$$

Breaking x and y into blocks



$x_1 2^m$ can be computed via “shift right by m ”

So this multiplication only costs $O(n)$ operations.

4 Multiplications \rightarrow 3 Multiplications

$$xy = x_1y_12^{2m} + (x_1y_0 + x_0y_1)2^m + x_0y_0$$

We can write two multiplications as one, plus some subtractions:

$$x_1y_0 + x_0y_1 = (x_1 + x_0)(y_1 + y_0) - x_1y_1 - x_0y_0$$

But what we need to subtract is exactly what we need for the original multiplication!

- ▶ $p_0 = x_0y_0$
- ▶ $p_1 = x_1y_1$
- ▶ $p_2 = (x_1 + x_0)(y_1 + y_0) - p_1 - p_0$

$$xy = p_12^{2m} + p_22^m + p_0$$

Analysis

Assume $n = 2^k$ for some k (this is the common case when the integers are stored in computer words):

$$T(2^k) = 3T(2^{k-1}) + c2^k$$

$$\frac{T(2^k)}{3^k} = \frac{T(2^{k-1})}{3^{k-1}} + \frac{c2^k}{3^k}$$

$$= \gamma + c \sum_{i=1}^k \frac{2^i}{3^i}$$

$$\leq \beta \quad \text{for some constants } \gamma, \beta$$

(γ handles the constant work for the base case.) So:

$$T(2^k) \leq \beta 3^k = \beta (2^k)^{\log_2(3)} = \beta n^{\log_2(3)} = O(n^{1.58\dots})$$