

2 recursive calls

$\frac{N}{2}$ values

$k=0, \dots, \frac{N}{2}-1$

3. for $k=0, \dots, N-1$,
 output $P(e^{-\frac{2\pi i k}{N}}) = P_1(e^{-\frac{4\pi i k}{N}}) e^{-\frac{2\pi i k}{N}} + P_2(e^{-\frac{4\pi i k}{N}})$

Annotations:
 - P_1 and P_2 are circled in green.
 - P_1 is labeled "known from step 2".
 - P_2 is labeled "known from step 2".
 - A green arrow points to the output with the label "N values".

note: if $k > \frac{N}{2}$,

$$e^{-\frac{4\pi i k}{N}} = e^{-\frac{4\pi i}{N} (k - \frac{N}{2})}$$
 because $e^{+\frac{4\pi i}{N} \cdot \frac{N}{2}} = e^{2\pi i} = 1$

$$\Rightarrow T(N) = 2T\left(\frac{N}{2}\right) + O(N)$$

$$\Rightarrow T(N) = \boxed{O(N \log N)}$$

Rank - above algm for Problem A is called Fast Fourier Transform (FFT)

Why? it computes $P(e^{-\frac{2\pi i k}{N}}) = \sum_{j=0}^{N-1} a_j e^{-\frac{2\pi i k j}{N}}$
 $\forall k=0, \dots, N-1$

$(P(x) = \sum_{j=0}^{N-1} a_j x^j)$

(similar to Fourier transform)

$$\hat{f}(x) = \int_{-\infty}^{\infty} f(t) e^{-2\pi i x t} dt$$

Algm for Problem B:

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approach 1:

Similar, $O(N \log N)$ time

approach 2:

inverse-FFT "equiv." to FFT

(inverse Fourier transform

$$f(t) = \int_{-\infty}^{\infty} \hat{f}(x) e^{2\pi i t x} dx$$

$$\widehat{f \circ g} = \hat{f} \cdot \hat{g}$$

convol.

\Rightarrow polynomial mult / convolution
in $O(n \log n)$ time

Remark - precision issues ...

$$(e^{ix} = \cos x + i \sin x)$$

Matrix Multiplication

Given $n \times n$ matrices A, B ,
compute $C = AB$

e.g.

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} 2 & 3 & 4 \\ 5 & 6 & 7 \\ 8 & 9 & 10 \end{pmatrix} = \begin{pmatrix} 36 & 42 & x \\ x & x & x \\ x & x & x \end{pmatrix}$$

Obvious algm (by def'n):

for $i = 1$ to n

for $j = 1$ to n

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$$

$O(n^3)$ time

better?

(lots of appl'ns: inverse / det / $Ax = b$...)

Strassen's Alg'm ('69)

divide $A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$

$$B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$$

where each A_{ij}, B_{ij} is $\frac{n}{2} \times \frac{n}{2}$

1st idea -

$$AB = \begin{pmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{pmatrix}$$

$$\Rightarrow T(n) = 8 T\left(\frac{n}{2}\right) + O(n^2)$$

Master
thm \Rightarrow

$$O(n^{\log_2 8}) = O(n^3)$$

not better ☹

more clever idea -

$$\begin{aligned}
 C_1 &= A_{11}(B_{11} - B_{21}) & C_2 &= (A_{11} + A_{12}) B_{21} \\
 C_3 &= A_{22}(B_{22} - B_{12}) & C_4 &= (A_{21} + A_{22}) B_{22} \quad B_{12} \\
 & & & \text{(Hypo fixed)} \\
 C_5 &= (A_{11} + A_{22})(B_{21} + B_{12}) \\
 C_6 &= (A_{12} - A_{22})(B_{21} + B_{22}) \\
 C_7 &= (A_{21} - A_{11})(B_{11} + B_{12})
 \end{aligned}$$

$$AB = \left(\begin{array}{c|c} C_1 + C_2 & C_5 + C_6 + C_3 - C_2 \\ \hline C_5 + C_7 + C_1 - C_4 & C_3 + C_4 \end{array} \right)$$

$$\Rightarrow T(n) = 7T\left(\frac{n}{2}\right) + O(n^2)$$

$$\Rightarrow O(n^{\log_2 7}) \leq \boxed{O(n^{2.81})}$$

better?

3-way D&C: $T(n) = 23T\left(\frac{n}{3}\right) + O(n^2)$
 $\Rightarrow O(n^{2.85})$ worse!

Pan '78: $T(n) = \frac{143640}{3} T\left(\frac{n}{70}\right) + O(n^2)$
 $\Rightarrow O(n^{2.795})$

- 179: $O(n^{2.781})$
- 180: $O(n^{2.780})$ ←
- '81: $O(n^{2.522})$
- Strassen '86: $O(n^{2.479})$
- 1986: $O(n^{2.376})$

Strassen '86: $O(n^{2.717})$

Coppersmith - Winograd '90: $O(n^{2.376})$

Stothers '10

$O(n^{2.373})$

Vassilevska W. '12:

Le Gall '14:

$O(n^{2.37287})$

Alman, Vassilevska W. '21:

$O(n^{2.37286})$