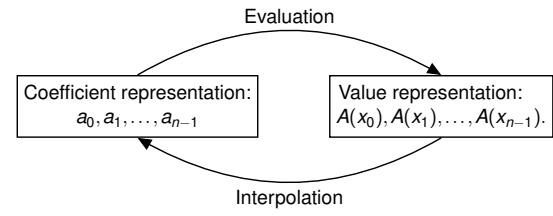


Recall: Multiplying polynomials, coefficient/value representation

Evaluation: $O(n \log n)$ if choose $1, \omega, \omega^2, \dots, \omega^{n-1}$.

CS170 - Lecture 5
Sanjam Garg
UC Berkeley



Interpolation: From points $A(x_0), \dots, A(x_{n-1})$ to coefficients..
We will see this today!

Polynomial Evaluation and Matrices

Evaluation: Compute $A(\cdot)$ from a_i 's:

$$\begin{bmatrix} A(x_0) \\ A(x_1) \\ \vdots \\ A(x_{n-1}) \end{bmatrix} = \begin{bmatrix} 1 & x_0 & x_0^2 & \dots & x_0^{n-1} \\ 1 & x_1 & x_1^2 & \dots & x_1^{n-1} \\ \vdots & \vdots & \ddots & & \vdots \\ 1 & x_{n-1} & x_{n-1}^2 & \dots & x_{n-1}^{n-1} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

Interpolation (going back to coefficient matrix).

How?

Compute inverse of matrix above.

Multiply. $O(n^2)$!

This sounds expensive!!

Also, computing inverse not even easy.

Using roots of unity

FFT: ω is complex n th root of unity
and matrix is ...

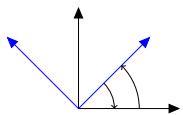
$$M_n(\omega) = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \dots & \omega^{(n-1)} \\ 1 & \omega^2 & \omega^4 & \dots & \omega^{2(n-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^j & \omega^{2j} & \dots & \omega^{j(n-1)} \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & \omega^{(n-1)} & \omega^{2(n-1)} & \dots & \omega^{(n-1)(n-1)} \end{bmatrix}$$

Compute inverse of $M_n(\omega)$?

Geometry and FFT.

Rows are orthogonal.

Multiply by $M_n(\omega)$: project point onto each row (and scaled.)
Rigid Rotation (and scaling.)!



Reverse Rotation is inverse operation.

Scaling: for rotation, axis should have length 1, FFT length n .

Algebraically.

Inversion formula: $(M_n(\omega))^{-1} = \frac{1}{n} M_n(\omega^{-1})$.

$$C = M_n(\omega) \times M_n(\omega^{-1})? \quad i \begin{array}{|c|} \hline \textcolor{blue}{\rule{0.5cm}{0.5pt}} \\ \hline \end{array} \times \begin{array}{|c|} \hline \textcolor{blue}{\rule{0.5cm}{0.5pt}} \\ \hline \end{array}^j = \begin{array}{|c|} \hline \textcolor{blue}{\rule{0.5cm}{0.5pt}} \\ \hline \end{array}^{(i,j)}$$

Recall: $\omega = e^{2\pi i / n}$.

$$c_{ij} = \sum_k \omega^{ik} \omega^{-kj} = \sum_k \omega^{(ik-kj)} = \sum_k \omega^{k(i-j)} = \sum_k r^k, \quad r = \omega^{(i-j)}$$

Case $i=j$: $r = \omega^0 = 1$ and $c_{ii} = n$.

Case $i \neq j$:

$$c_{ij} = 1 + r + r^2 + \dots + r^{n-1} = \frac{1-r^n}{1-r}$$

$$r^n = (\omega^{(i-j)})^n = (\omega^n)^{(i-j)} = 1^{(i-j)} \implies c_{ij} = 0.$$

For C – diagonals are n and the off-diagonals are 0.

Divide by n get identity!

Inversion formula: $M_n(\omega)^{-1} = \frac{1}{n} M_n(\omega^{-1})$.

Computing inverse.

Multiplying polynomials?

Evaluation: $O(n \log n)$ if choose $1, \omega, \omega^2, \dots, \omega^{n-1}$.

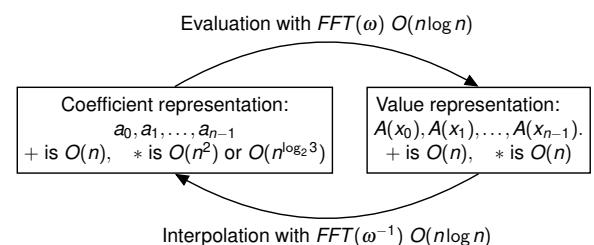
FFT works with points with basic root of unity: ω or ω^{-1}
 $1, \omega^{-1}, \omega^{-2}, \dots, \omega^{-(n-1)}$.

ω^{-1} is a primitive n th root of unity!

Evaluation: $\text{FFT}(a, \omega)$.

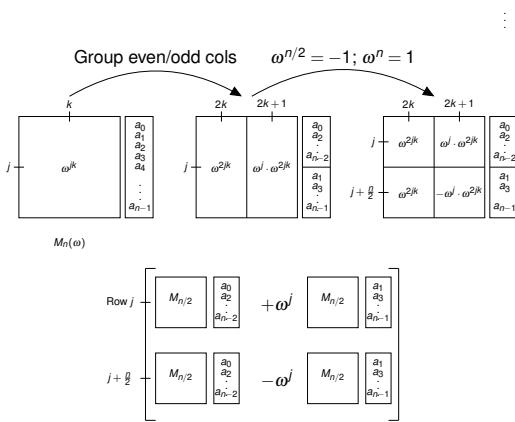
Interpolation: $\frac{1}{n} \text{FFT}(a, \omega^{-1})$.

$\implies O(n \log n)$ time for multiplying degree n polynomials.



Interpolation: From points $A(x_0), \dots, A(x_{n-1})$ to “function”.

FFT: a closer look.



Definitive Algorithm: FFT

FFT: "M(ω)a"

Idea:

" $M(\omega)a$ " computed from...
" $M(\omega^2)a_o$ " and " $M(\omega^2)a_e$ ".

FFT(a, ω):

if $\omega = 1$ return a

$(s_0, s_1, \dots, s_{n/2-1}) = \text{FFT}((a_0, a_2, \dots, a_{n-2}), \omega^2)$
 $(s'_0, s'_1, \dots, s'_{n/2-1}) = \text{FFT}((a_1, a_3, \dots, a_{n-1}), \omega^2)$

for $j = 0$ to $n/2 - 1$:

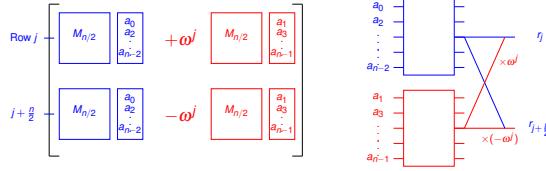
$r_j = s_j + \omega^j s'_j$

$r_{j+n/2} = s_j - \omega^j s'_j$

return $(r_0, r_1, \dots, r_{n-1})$

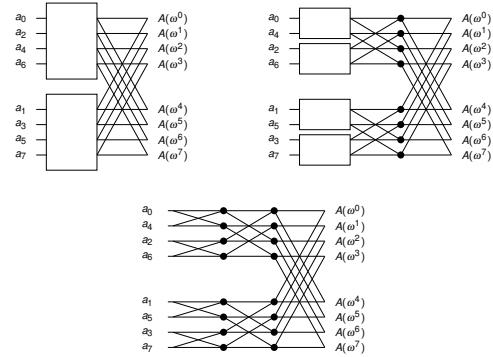
Runtime: $T(n) = 2T(n/2) + O(n)$

Unfolding FFT.



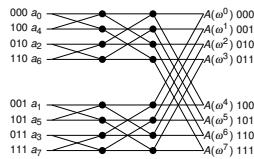
Butterfly switches!

Expanding FFT...

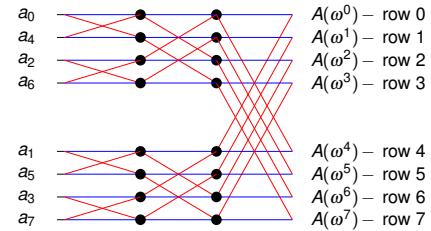


Edges from lower half of FFT have multipliers!

Order on Left



FFT Network.



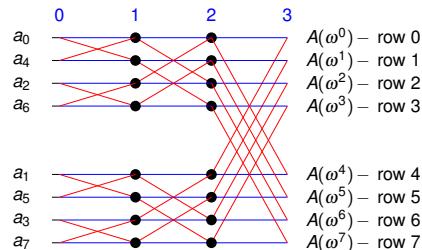
log N - levels.

N - rows.

In level i :

Row r node is connected to row r node in level $i+1$.
Row r node connected to row $r \pm 2^i$ node in level $i+1$

FFT Network.



Row r node connected to row $r \pm 2^i$ node in level $i+1$

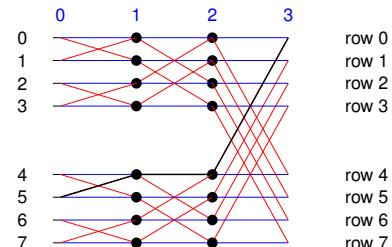
When is it $r + 2^i$?

(A) When $[r/2^i]$ is odd.

(B) When $[r/2^i]$ is even.

(B). Red edges flip bit!

Unique Paths.



Route from input $i = 101$ to output $j = 000$?

Flip first bit. Red (cross) edge.

Keep second bit. Blue (straight) edge.

Flip third bit. Red (cross) edge.

$$A(x) = \sum_{i=0}^d a_i x^i = A_L(x) + x^{d/2} A_H(x), A_L(x) := \sum_{i=0}^{d/2} a_i x^i, A_H(x) = \sum_{i=1}^{d/2} a_{i+d/2} x^i$$

$$B(x) = \sum_{i=0}^d b_i x^i = B_L(x) + x^{d/2} B_H(x), B_L(x) := \sum_{i=0}^{d/2} b_i x^i, B_H(x) = \sum_{i=1}^{d/2} b_{i+d/2} x^i$$

The product $A(x)B(x)$ is

$$A_L(x)B_L(x) + x^{d/2}(A_L(x)B_H(x) + A_H(x)B_L(x)) + x^d A_H(x)B_H(x)$$

Compute ...

$$A_L(x)B_L(x), \quad A_H(x)B_H(x), \quad (A_L(x) + A_H(x))(B_L(x) + B_H(x))$$

and recurse

Time is $O(d^{\log_2 3})$

FFT does better. (But this is useful to see)

Definitive FFT algorithm and code.