

Lecture 4: Discrete Fourier Transform

Rafael Oliveira

University of Waterloo
Cheriton School of Computer Science

rafael.oliveira.teaching@gmail.com

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Overview

- Polynomial Evaluation & Interpolation
- Basics for DFT
- Fast Fourier Transform (FFT)
- Conclusion
- Acknowledgements

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Representing Polynomials

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1 As a list of coefficients

$$p_0 + p_1 x + \dots + p_d x^d$$

$$p(x) \leftrightarrow (p_0, p_1, \dots, p_d)$$

2 As evaluations at $d + 1$ points

distinct

$$(\alpha_0, y_0), \dots, (\alpha_d, y_d)$$

$$\det(V) = \prod_{i > j} (\alpha_i - \alpha_j) \neq 0$$

where $y_i := p(\alpha_i)$

- Proof of equivalence:

$$\underbrace{\begin{pmatrix} 1 & \alpha_0 & \alpha_0^2 & \dots & \alpha_0^d \\ 1 & \alpha_1 & \alpha_1^2 & \dots & \alpha_1^d \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \alpha_d & \dots & \dots & \alpha_d^d \end{pmatrix}}_{\text{invertible}} \begin{pmatrix} p_0 \\ p_1 \\ \vdots \\ p_d \end{pmatrix} = \begin{pmatrix} p(\alpha_0) \\ p(\alpha_1) \\ \vdots \\ p(\alpha_d) \end{pmatrix}$$

Multiplying Polynomials via Interpolation

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 - ① Compute $2d + 1$ evaluations of p, q (why $2d + 1$?)

$$p(x) \cdot q(x) \leftarrow \text{degree} \leq 2d$$



$2d + 1$ evaluations
to uniquely determine it

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 - ② Compute the products $\gamma_i := p(\alpha_i) \cdot q(\alpha_i)$ ←
 - ③ Use Lagrange's interpolation polynomials

$$L_i(x) = \prod_{j \neq i} \frac{x - \alpha_j}{\alpha_i - \alpha_j} \quad \}$$

$$L_i(\alpha_i) = 1$$

$$L_i(\alpha_j) = 0 \quad \text{if } j \neq i$$

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$$L_i(x) = \prod_{j \neq i} \frac{x - \alpha_j}{\alpha_i - \alpha_j}$$

- ④ From previous slide's equivalence, we know

$$p(x) \cdot q(x) = \sum_{i=0}^{2d} \gamma_i L_i(x)$$

$$p(\alpha_i) \cdot q(\alpha_i) = \gamma_i \cdot \underbrace{L_i(\alpha_i)}_1$$

they evaluate to same values at $2d+1$ pts.

Multiplying Polynomials via Interpolation

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- We saw the analysis of this in Ostrowski's non-scalar model

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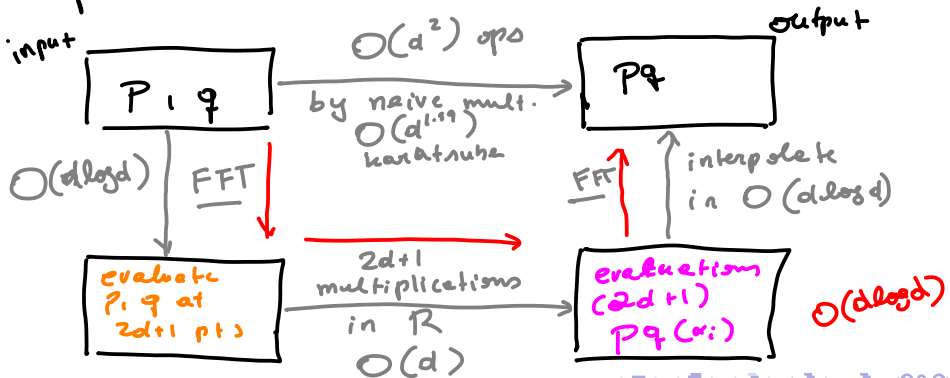
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$$p(x) \cdot q(x) = \sum_{i=0}^{2d} \gamma_i L_i(x)$$

- We saw the analysis of this in Ostrowski's non-scalar model
- To get a fast algorithm in the scalar model (where we count all ring operations) we will use *Fast Fourier Transform*

Fast Fourier Transform Idea

If we can evaluate poly of deg d and interpolate poly of deg d fast, then can multiply fast



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Roots of Unity

- \mathbb{F} a field, $n \in \mathbb{N}$

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- \mathbb{F} a field, $n \in \mathbb{N}$
- We say $\omega \in \mathbb{F}$ is a primitive n^{th} *root of unity* (n -PRU) if
 - 1 $\omega^n = 1$
 - 2 $\omega^k \neq 1$ for $1 \leq k < n$

$$x^n - 1$$

ω root

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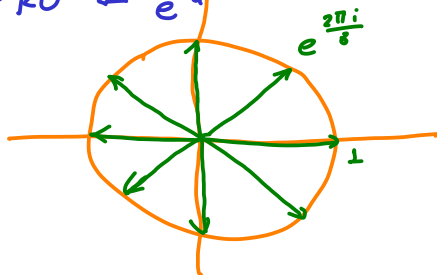
- Examples:

- 1 $\mathbb{F} = \mathbb{C}$ and $\omega = e^{2\pi i/n}$

$$= \cos\left(\frac{2\pi}{n}\right) + i \sin\left(\frac{2\pi}{n}\right)$$

4-PRU $\leftarrow e^{\frac{2\pi i}{4}}$

8-PRU



Roots of Unity

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 - 1 $\omega^n = 1$
 - 2 $\omega^k \neq 1$ for $1 \leq k < n$
- Examples:
 - 1 $\mathbb{F} = \mathbb{C}$ and $\omega = e^{2\pi i/n}$
 - 2 If p is a prime and $\mathbb{F} = \mathbb{Z}_p$, then \mathbb{F} has a $(p-1)$ -PRU. How many such roots does \mathbb{Z}_p have?

in \mathbb{Z}_p we know that $a^{p-1} = 1 \quad \forall a \in \mathbb{Z}_p^*$

$P(x) = x^{p-1} - 1$ has exactly $p-1$ roots
over \mathbb{Z}_p

$$P(x) = \underbrace{(x^{\frac{p-1}{2}} - 1)}_{\leq \frac{p-1}{2} \text{ roots over } \mathbb{Z}_p} \underbrace{(x^{\frac{p-1}{2}} + 1)}_{\text{has } \frac{p-1}{2} \text{ roots over } \mathbb{Z}_p}$$

$$\alpha^{\frac{p-1}{2}} \equiv -1$$

$$\alpha^{p-1} \equiv 1 \quad \therefore \alpha \text{ is } (p-1)\text{-PRU}$$

$$= \frac{p-1}{2} \text{ roots over } \mathbb{Z}_p$$

$$\text{has } \frac{p-1}{2} \text{ roots over } \mathbb{Z}_p$$
$$\alpha \text{ root}$$

More on Roots of Unity

Vandermonde Matrix

- Given elements u_0, \dots, u_d

$(d+1) \times (d+1)$ matrix

$$V(u_0, \dots, u_d) = \begin{pmatrix} u_0^0 & u_0^1 & \cdots & u_0^d \\ u_1^0 & u_1^1 & \cdots & u_1^d \\ \vdots & \vdots & \vdots & \vdots \\ u_d^0 & u_d^1 & \cdots & u_d^d \end{pmatrix}$$

Vandermonde Matrix

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- From previous lecture, maps coefficients to evaluations:

$$\begin{pmatrix} u_0^0 & u_0^1 & \cdots & u_0^d \\ u_1^0 & u_1^1 & \cdots & u_1^d \\ \vdots & \vdots & \ddots & \vdots \\ u_d^0 & u_d^1 & \cdots & u_d^d \end{pmatrix} \begin{pmatrix} p_0 \\ p_1 \\ \vdots \\ p_d \end{pmatrix} = \begin{pmatrix} p(u_0) \\ p(u_1) \\ \vdots \\ p(u_d) \end{pmatrix}$$

Vandermonde Matrix

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- For interpolation, we can choose our evaluation points. Choosing roots of unity give rise to the FFT!

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Vandermonde with Roots of Unity

- Given d , Take ω to be a $(d + 1)$ -PRU, and let

$$V(\omega) = V(\omega^0, \omega, \dots, \omega^d) = \begin{pmatrix} 1 & 1 & \dots & 1 \\ 1 & \omega^1 & \dots & \omega^d \\ 1 & \omega^2 & \dots & \omega^{2d} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^d & \dots & \omega^{d^2} \end{pmatrix} \begin{matrix} \omega^0 \\ \omega \\ \omega^2 \\ \vdots \\ \omega^d \end{matrix}$$

$1, \omega, \dots, \omega^d$
all $d+1$
roots of unity

Vandermonde with Roots of Unity

$$x^{d+1} - 1 = (x-1)(x^d + x^{d-1} + \dots + x + 1)$$

ω^{a-b} root \uparrow

- Given d , Take ω to be a $(d+1)$ -PRU, and let ω^{-1} also $(d+1)$ -PRU

$$V(\omega) = V(\omega^0, \omega, \dots, \omega^d) = \begin{pmatrix} 1 & 1 & \dots & 1 \\ 1 & \omega^1 & \dots & \omega^d \\ 1 & \omega^2 & \dots & \omega^{2d} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^d & \dots & \omega^{d^2} \end{pmatrix}$$

$V(\omega^{-1})$

- Lemma:** $V(\omega) \cdot V(\omega^{-1}) = (d+1) \cdot I$

$$W_{ee} = \sum_{i=0}^d \underbrace{\omega^{ie}}_{i\text{th entry of } V(\omega)} \cdot \underbrace{\omega^{-ie}}_{i\text{th entry of } V(\omega^{-1})} = \sum_{i=0}^d 1 = d+1$$

$(\omega^{a-b})^{d+1} = 1$ $a \neq b$
 $\omega^{a-b} \neq 1$

$$W_{ab} = \sum_{i=0}^d \omega^{ia} \cdot \omega^{-ib} = \sum_{i=0}^d \omega^{(a-b) \cdot i} = 0$$

Proof of lemma

upshot of Lemma

if use PRU then

$$\underbrace{V(\omega^{-1})}_{\text{evaluation}} = (\mathcal{D}+1) \cdot \underbrace{V(\omega)^{-1}}_{\text{interpolation}}$$

\Rightarrow interpolation (inverse of Vandermonde)

is equivalent to

evaluation (matrix-vector multiplication by Vandermonde matrix)

Discrete Fourier Transform (DFT)

- Given ω a $(d + 1)$ -PRU, the DFT is given by $DFT(\omega) : \mathbb{F}^{d+1} \rightarrow \mathbb{F}^{d+1}$

$$DFT(\omega) \begin{pmatrix} p_0 \\ p_1 \\ \vdots \\ p_d \end{pmatrix} = \begin{pmatrix} p(1) \\ p(\omega) \\ \vdots \\ p(\omega^d) \end{pmatrix}$$

$$V(\omega) \begin{pmatrix} p_0 \\ \vdots \\ p_d \end{pmatrix} = \begin{pmatrix} p(1) \\ \vdots \\ p(\omega^d) \end{pmatrix}$$

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- Fast Fourier Transform is a way of computing the DFT in $O(n \log n)$ operations!

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- Fast Fourier Transform is a way of computing the DFT in $O(n \log n)$ operations!
- $DFT(\omega)$ computes polynomial evaluation
- $DFT(\omega^{-1})$ computes polynomial interpolation

Fast Fourier Transform

- Assume that $d = 2^k$
- Idea:
 - $p(x) = p_{\text{even}}(x^2) + \underline{x} \cdot \underline{p_{\text{odd}}(x^2)}$

$$P(x) = \underline{p_0} + \underline{p_1 x} + \underline{p_2 x^2} + \underline{p_3 x^3} + \underline{p_4 x^4}$$

$$\left. \begin{array}{l} \text{degree} \\ \leq \frac{d}{2} \end{array} \right\} \begin{cases} P_{\text{even}}(y) = p_0 + p_2 y + p_4 y^2 \\ P_{\text{odd}}(y) = p_1 + p_3 y \end{cases}$$

Fast Fourier Transform

- Assume that $d = 2^k$

- Idea:

- $p(x) = p_{\text{even}}(x^2) + x \cdot p_{\text{odd}}(x^2)$

- Reduced problem of

evaluating $p(x)$ at $1, \omega, \dots, \omega^d$
into $d+1$ values

evaluating $p_{\text{even}}(x), p_{\text{odd}}(x)$ (of degree $d/2$) at $1, \omega^2, \omega^4, \omega^6, \dots, \omega^{2d}$
 $d+1$ values

Fast Fourier Transform

- Assume that $d = 2^k$

$$\omega^d = 1$$

ω d -PRU

ω^2 $\frac{d}{2}$ -PRU

- Idea:

- $p(x) = p_{\text{even}}(x^2) + x \cdot p_{\text{odd}}(x^2)$
- Reduced problem of

evaluating $p(x)$ at $1, \omega, \dots, \omega^{d-1}$

into

evaluating $p_{\text{even}}(x), p_{\text{odd}}(x)$ (of degree $d/2$) at $1, \omega^2, \omega^4, \omega^6, \dots, \omega^{2(d-1)}$

- are we evaluating it at d points?

$$\left(\omega^{d/2}\right)^2 = \omega^d = 1$$

$$\rightarrow \left(\omega^{d/2+1}\right)^2 = \omega^{d+2} = \omega^2$$

\vdots

$$\rightarrow \left(\omega^{d-1}\right)^2 = \omega^{2d-2} = \left(\omega^2\right)^{\frac{d}{2}-1}$$

ω^2

$\Rightarrow \frac{d}{2}$

then
evaluated
it already

$\left(\omega^2\right)^k$

ω^2

even powers

not d

points!

they are
actually

$\frac{d}{2}$ points!

Fast Fourier Transform

- Assume that $d = 2^k$
- Idea:
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 - Reduced problem of

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evaluating $p_{\text{even}}(x), p_{\text{odd}}(x)$ (of degree $d/2$) at $1, \omega^2, \omega^4, \omega^6, \dots, \omega^{2d}$

- are we evaluating it at d points? *no $d/2$ pts*
- need to combine them back

$$\text{d evaluations} \quad \boxed{p(\omega^t)} = p_{\text{even}}(\omega^{2t}) + \omega^t \cdot p_{\text{odd}}(\omega^{2t})$$

Two more operations to get $p(\omega^t)$ plus d operations to get $1, \omega, \dots, \omega^d$

d evaluations of p

*$\frac{d}{2}$ evaluations of p_{even}
 $\frac{d}{2}$ evaluations of p_{odd}*

Fast Fourier Transform

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Two more operations to get $p(\omega^t)$ plus d operations to get $1, \omega, \dots, \omega^d$

- Recurrence:

$$T(d) = 2T(d/2) + 3d$$

operations
eval $p(\omega^i)$

p_{even}
 p_{odd}

put everything
together

Fast Fourier Transform

- Assume that $d = 2^k$
- Idea:
 - $p(x) = p_{\text{even}}(x^2) + x \cdot p_{\text{odd}}(x^2)$
 - Reduced problem of

evaluating $p(x)$ at $1, \omega, \dots, \omega^d$

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Two more operations to get $p(\omega^t)$ plus d operations to get $1, \omega, \dots, \omega^d$

- Recurrence:

$$T(d) = 2T(d/2) + 3d$$

- Master's theorem gives us $O(n \log n)$

Fast Polynomial Multiplication

- Now that we know how to evaluate and interpolate polynomials fast, can use it to multiply faster!
- Given p, q of degree $< 2^{k-1}$, take ω to be a 2^k -PRU.

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- Given p, q of degree $< 2^{k-1}$, take ω to be a 2^k -PRU.
- $c(x) = p(x)q(x)$ has degree $< 2^k$ so it is uniquely determined by the evaluations $c(1), \dots, c(\omega^{2^k-1})$

all roots of unity
evaluations

Fast Polynomial Multiplication

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- $c(x) = p(x)q(x)$ has degree $< 2^k$ so it is uniquely determined by the evaluations $c(1), \dots, c(\omega^{2^k-1})$
- use FFT to evaluate $p(1), p(\omega), \dots, p(\omega^{2^k-1})$ and $q(1), q(\omega), \dots, q(\omega^{2^k-1})$

$O(d \log d)$ operations

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- use FFT to evaluate $p(1), p(\omega), \dots, p(\omega^{2^k-1})$ and $q(1), q(\omega), \dots, q(\omega^{2^k-1})$
- Multiply $c(\omega^t) = p(\omega^t) \cdot q(\omega^t)$

d operations

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- $c(x) = p(x)q(x)$ has degree $< 2^k$ so it is uniquely determined by the evaluations $c(1), \dots, c(\omega^{2^k-1})$
- use FFT to evaluate $p(1), p(\omega), \dots, p(\omega^{2^k-1})$ and $q(1), q(\omega), \dots, q(\omega^{2^k-1})$
- Multiply $c(\omega^t) = p(\omega^t) \cdot q(\omega^t)$
- use FFT to interpolate $c(1), \dots, c(\omega^{2^k-1})$

$O(d \log d)$ operations

Fast Polynomial Multiplication

- Now that we know how to evaluate and interpolate polynomials fast, can use it to multiply faster!
- Given p, q of degree $< 2^k - 1$, take ω to be a 2^k -PRU.
- $c(x) = p(x)q(x)$ has degree $< 2^k$ so it is uniquely determined by the evaluations $c(1), \dots, c(\omega^{2^k-1})$
- use FFT to evaluate $p(1), p(\omega), \dots, p(\omega^{2^k-1})$ and $q(1), q(\omega), \dots, q(\omega^{2^k-1})$ $O(d \log d)$
- Multiply $c(\omega^t) = p(\omega^t) \cdot q(\omega^t)$ $O(d)$
- use FFT to interpolate $c(1), \dots, c(\omega^{2^k-1})$ $O(d \log d)$
- running time: ~~$O(n \log n)$~~
 $O(d \log d)$

Fast Polynomial Multiplication

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- Given p, q of degree $< 2^{k-1}$, take ω to be a 2^k -PRU.
- $c(x) = p(x)q(x)$ has degree $< 2^k$ so it is uniquely determined by the evaluations $c(1), \dots, c(\omega^{2^k-1})$
- use FFT to evaluate $p(1), p(\omega), \dots, p(\omega^{2^k-1})$ and $q(1), q(\omega), \dots, q(\omega^{2^k-1})$
- Multiply $c(\omega^t) = p(\omega^t) \cdot q(\omega^t)$
- use FFT to interpolate $c(1), \dots, c(\omega^{2^k-1})$
- running time: $O(n \log n)$
- **Remark:** FFT can be (and is) used in fast integer multiplication - starting from the works of Strassen and Schönhage in 1971 up to the recent work of Harvey and van der Hoeven in 2019.

Read more on Anne's notes

- Polynomial Evaluation & Interpolation
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- Fast Fourier Transform (FFT)
- **Conclusion**
- Acknowledgements

Conclusion

In today's lecture, we learned

- Discrete Fourier Transform for evaluation and interpolation
- Properties of roots of unity
- Fast Fourier Transform
- Multiplying polynomials in $O(n \log n)$ operations

Acknowledgement

- Based largely on Arne's notes

`https://cs.uwaterloo.ca/~r5olivei/courses/
2021-winter-cs487/lec4-ref.pdf`

$$T(n) = \alpha T\left(\frac{n}{\beta}\right) + f(n)$$

Master's theorem

α, β, f gives you
upper bound on running time
of $T(n)$

$$f(n) = 3n \quad \alpha = 2 \quad \beta = 2$$

$$T(n) = 2T\left(\frac{n}{2}\right) + 3n$$

$$= 2\left(2T\left(\frac{n}{4}\right) + 3\frac{n}{2}\right) + 3n$$

$$= 4 \cdot T\left(\frac{n}{4}\right) + 2 \cdot (3n)$$

$$\dots = 2^k \cdot T\left(\frac{n}{2^k}\right) + k \cdot (3n) \quad \begin{array}{l} \text{after} \\ k \\ \text{steps} \end{array}$$

$$k = \log n$$

$$= n \cdot \overbrace{T(1)}^{\text{const.}} + 3n \cdot \log n$$

$$= O(n \log n)$$