## Lecture 23: Fast Fourier Transform II

Recap:

Def. Let  $\{a_j\}_{j=0}^n$  and  $\{b_k\}_{k=0}^n$  be two sequences of real numbers. We define their convolution a\*b as the ren sequence  $\{c_k\}_{k=0}^{n+m}$  given by:

$$C_{\ell} = (a * b)_{\ell} = \sum_{j=0}^{\ell} a_j b_{\ell-j}$$

## The Strategy

Input: Coefficient vectors

Fird god droice
of x... x mem f ()
using Dac to get
Elhemolog(n-m) time

Offet: Coefficient ver axb of r(x)=pbs.g(x)

Interplate

endrators

(ompute evaluations  $X_j \mapsto p(x_j)$ ,  $X_j \mapsto q(x_j)$  for j=0...n-m

· # evaluations

Discrete Fourier Trensform: Given {2,3;=0} specifying a poly p(x)= \( \frac{5}{2} \alpha\_{1} \times^{2} \times  $W_n \leftarrow p(\omega_n \leftarrow) \forall k=0,...,n-1$ where who is the primitive not root of unity.

FFT for computing DFT: Input: as...an.i; Assume n is power of 2 If n=1:

Return[ao]

Rild deven: = (ao,0,...9n-2), add: = (a,0,...9n)

Recurse Fecen= FFT(a,m), Fodd: FFT(add) }27(n/2) If n=1: For KE & O. J. .. n/2-13: Set F[k]: Feen [k] + wn k. Fold[k]

Set F[k+2] = Feen[k] + wn k+1/2 · Fold[k]

M = O(nlog n): ! Return F

Claim: FFT computes DFT correctly for degree 2-1 poly Proof: Induction on l. Buse case l=0 is obvious. Kecall  $p(x) = a_0 + a_1 x + a_2 x^2 \dots$  $\int_{\text{even}}^{\infty} (x) = a_0 + a_2 x + a_4 x^2 - \cdots = \sum_{k=0}^{\frac{n}{2}-1} a_k x^k$   $P \cap C$  $P_{odd}(x) = a_1 + a_3x + a_6x^2 = \sum_{k=0}^{\infty} a_{2k+} \times k$  $\rho(x) = P_{even}(x^2) + x - P_{old}(x^2)$  $p(x) = p_{even}(x^2) - x p_{och}(x^2)$ 

Inductive stepi

$$F[K] = F_{een}[K] + w_{2e+1}^{k} \cdot F_{odd}[K]$$

$$= f_{even}(w_{2e}^{k}) + w_{2e+1}^{k} \cdot F_{odd}[w_{2e}^{k}]$$

$$= \rho(w_{2e+1}^{k})$$

Obs: 
$$w_{N_2} = \exp\left(\frac{2\pi i}{N_2}k\right)$$

$$= \exp\left(\frac{2\pi i}{N_1}ik\right) = \exp\left(\frac{2\pi i}{N_1}ik\right)$$

## Interpolation

Fact: Any degree-n polynomial p is uniquely determined by its evaluations on any set of n+1 distinct points.

$$\begin{bmatrix}
\rho(x_0) \\
\rho(x_1)
\\
\rho(x_2)
\end{bmatrix} = \begin{bmatrix}
1 & x_0 & x_0^2 & \dots & x_0^n \\
1 & x_1 & x_1^2 & \dots & x_1^n \\
1 & x_2 & x_2^2 & \dots & x_2^n
\end{bmatrix}$$

$$\begin{bmatrix}
C_0 \\
C_1 \\
C_2 \\
C_3
\end{bmatrix}$$

$$\begin{bmatrix}
C_0 \\
C_1 \\
C_2
\end{bmatrix}$$

$$\begin{bmatrix}
C_1 \\
C_2 \\
C_2
\end{bmatrix}$$

$$\begin{bmatrix}
C_1 \\
C_2 \\
C_3 \\
C_4
\end{bmatrix}$$

Fact: the meetrix above is invertible iff x ≠ X.

We want to compute  $F=Wc \Rightarrow c=W'F$ Lemma: W'= \(\frac{1}{n+1}\) W, where \(\vec{W}\) is entry vise complex conjugate of \(\vec{W}\). Corollary: We can compute (=WF in O(nlogn)
again using FFT.

Proof: Goal is to verify the W·W=I Diagonal Entries: Fix ke{0,...n} (n+1 W·W) (k, k)= n+1 = w(k, j)- W(j, k)= n+1 = (with with with it) = 1 Off-Diagonal Entries: Fix  $k \neq l$  in  $\{0,...,n\}$ .  $\left(\frac{1}{n+1} \overline{W} \cdot W\right) \left(k,l\right) = \frac{1}{n+1} \sum_{j=0}^{n} \overline{W_{n+1}^{jk}} W_{n+1}^{j+1}$  $= \frac{1}{n+1} \sum_{j=0}^{n} \left( \mathcal{W}_{n+1}^{l-k} \right)^{j}$ another primitive rat of unity.  $=\frac{1}{n+1}\sum_{j=0}^{n}\left(\omega_{n-1}^{*}\right)^{j}$ 

Convolution:

Input: a, b \( \mathbb{R}^n \); Assume n is power of 2

Compute \( F=DFT(a) \), G=DFT(b)

Compute \( H=F \cdot G \in C^n \) entrywise

Return InverseDFT(H)

Inverse DFT:

Input:  $F_0$ ...  $F_{n-1}$  Assume n: 2Compute  $y = FFT(F_0 \cdot ... F_{n-1})$ Return  $\sqrt[n]{n}$ 

Bonus! Revisiting Sketching Median Trick: Geal: Unknown I nont to Estimate. Let X r.v. sortsying E[X]=Y & Var(x)=CY Wont: Design a new estimator X based on copies of X s.t.  $\forall \varepsilon, S$ ,  $\Pr[(1-\varepsilon)Y = \hat{X} \leq (1+\varepsilon)Y \geq 1-S]$ Noive Approach:  $X = + \sum_{i=1}^{n} X_i$ , where  $X_i$  is indep. copy of X. Vor(x)=+Var(x)

Median-of-means Estimator: Let  $\widetilde{X}_{i} = + \sum_{i=1}^{n} X_{i,j}$  for copies  $X_{i,j}$  of  $X_{i,j}$ For j=1,..., L (Set T= 1000 L=0(log(1/8)) Theorem: Let  $\hat{x} = median(\hat{x}_1, ..., \hat{x}_c)$ , then  $P_{\ell}(\hat{x} - Y | SEY)$ Proof: Define Z; =1[|x;-Y|=EY] By earlier-analysis, if T= 1000 then Pr[Z=1]=9 Pr[[x-Y]>EY]=Pr[[x-Y]>EY for more than half of i 

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