

Fast Fourier Transform

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Recall we found that the exponential polynomial interpolating a function f at $x_j = \frac{2\pi j}{N}$ is given by:

$$P = \sum_{k=0}^{N-1} c_k E_k, \quad c_k = (f, E_k)_N$$

$$c_k = \frac{1}{N} \sum_{j=0}^{N-1} f(x_j) (\lambda^k)^j, \quad \lambda = e^{-2i\pi/N}$$

Thus computing P requires $O(N^2)$ computations

(N operations to compute each of the $N c_k$).

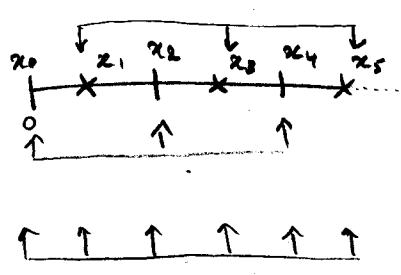
The FFT is based on a tremendous simplification that allows to compute P in $O(N \log N)$ operations.

For example: $N = 2^{20} \approx 1M$

$$N^2 = 2^{40} \approx 1G, \quad N \log_2 N \approx 20M$$

Basically factor of a million improvement! (in this case)

Here is the basic fact that is used by FFT:



$q_0 \equiv$ trig interpolant at odd nodes
 $q_0(x_{2j}) = f(x_{2j+1}) \quad 0 \leq j \leq n-1$
 $q_e \equiv$ trig interpolant at even nodes
 $q_e(x_{2j}) = f(x_{2j}) \quad 0 \leq j \leq n-1$
 $P \equiv$ trig interpolant at all nodes
 $P(x_j) = f(x_j) \quad 0 \leq j \leq 2n-1$

$$P(x) = \frac{1}{2} (1 + e^{inx}) q_e(x) + \frac{1}{2} (1 - e^{inx}) q_0(x - \frac{\pi}{n})$$

Proof:

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$$\left. \begin{array}{l} q_0, q_e \text{ have degree } \leq n-1 \\ (e^{in})^n \text{ has degree } n \end{array} \right\} \Rightarrow p \text{ has degree } \leq 2n-1$$

We need to show p interpolates f at the $2n$ nodes

$$x_0, x_1, \dots, x_{2n-1}; \quad x_j = \frac{\pi j}{n}, \quad 0 \leq j \leq 2n-1$$

$$P(x_j) = \frac{1}{2} [1 + E_n(x_j)] q_e(x_j) + \frac{1}{2} [1 - E_n(x_j)] q_0(x_j - \frac{\pi}{n})$$

$$E_n(x_j) = \exp\left[i n \frac{\pi j}{n}\right] = e^{i j \pi} = \begin{cases} +1 & j \text{ even} \\ -1 & j \text{ odd} \end{cases} \quad 0 \leq j \leq 2n-1$$

thus for j even:

$$P(x_j) = q_e(x_j) = f(x_j)$$

— for j odd:

$$P(x_j) = q_0(x_j - \frac{\pi}{n}) = q_0(x_{j-1}) = f(x_j) \quad \underline{QED}$$

Now this relation is also useful in practice since we can get coeff of P from those of q_e and q_0 easily.

Theorem
let

$$q_e = \sum_{j=0}^{n-1} \alpha_j E_j$$

$$q_0 = \sum_{j=0}^{n-1} \beta_j E_j$$

$$P = \sum_{j=0}^{2n-1} \gamma_j E_j$$

Then for $0 \leq j \leq n-1$:

$$\gamma_j = \frac{1}{2} \alpha_j + \frac{1}{2} e^{-i j \pi / n} \beta_j$$

$$\gamma_{j+n} = \frac{1}{2} \alpha_j - \frac{1}{2} e^{-i j \pi / n} \beta_j$$

Proof:

$$\begin{aligned}
 q_0\left(x - \frac{\pi}{n}\right) &= \sum_{j=0}^{n-1} \beta_j E_j\left(x - \frac{\pi}{n}\right) \\
 &= \sum_{j=0}^{n-1} \beta_j e^{ij\left(x - \frac{\pi}{n}\right)} = \sum_{j=0}^{n-1} \beta_j e^{-i\pi j/n} E_j(x)
 \end{aligned}$$

$$P(x) = \frac{1}{2} (1 + E_n(x)) p(x) + \frac{1}{2} (1 - E_n(x)) q\left(x - \frac{\pi}{n}\right)$$

$$\begin{aligned}
 P &= \frac{1}{2} \sum_{j=0}^{n-1} \left[(1 + E_n) \alpha_j E_j + (1 - E_n) \beta_j e^{-i\pi j/n} E_j \right] \\
 &= \frac{1}{2} \sum_{j=0}^{n-1} \left[\alpha_j + \beta_j e^{-i\pi j/n} \right] E_j + \left[\alpha_j - \beta_j e^{-i\pi j/n} \right] E_{j+n}
 \end{aligned}$$

(since $E_j E_n = E_{j+n}$)Q.E.D.

The relation between odd and even interpolants can be made more general by using the following operators: (linear)

$$L_n f = \text{trig poly interpolat nodes } x_j = \frac{2\pi j}{n} \quad 0 \leq j \leq n-1$$

$$(T_h f)(x) = f(x+h) \quad (\text{translation})$$

Of course we have:

$$L_n f = \sum_{k=0}^{n-1} (f, E_k)_n E_k$$

And in our formula:

$$p = L_{2n} f$$

$$q_e = L_n f$$

$$q^o = L_n T_{\pi/n} f$$

Thus:

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$$L_{2n} f = \frac{1}{2} (1 + E_n) L_n f + \frac{1}{2} (1 - E_n) T_{-\pi/n} L_n T_{\pi/n} f \quad (*)$$

We now want to design an algorithm to compute $L_N f$ with $N = 2^m$.

Notation, interp on 2^n nodes

$$P_k^{(m)} = L_{2^n} T_{\frac{2k\pi}{N}} f \quad \begin{array}{l} 0 \leq n \leq m \\ 0 \leq k \leq 2^{m-n} - 1 \end{array}$$

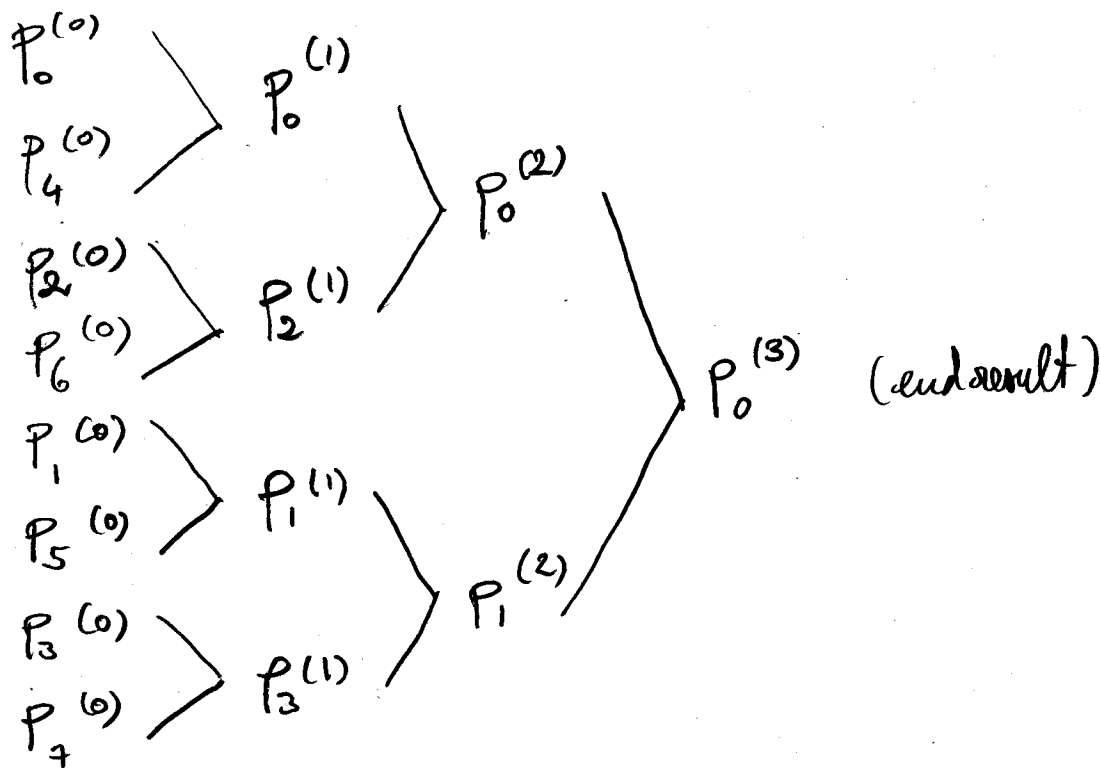
↑
starting at node k

= trig interp of degree $2^n - 1$ s.t.

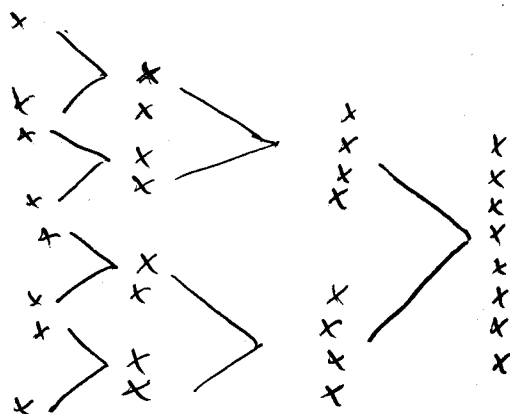
$$P_k^{(m)} \left(\frac{2\pi j}{2^n} \right) = f \left(\frac{2\pi k}{N} + \frac{2\pi j}{2^n} \right), \quad 0 \leq j \leq 2^n - 1$$

Applying (*),

$$P_k^{(m+1)}(x) = \frac{1}{2} (1 + e^{i2^n x}) P_k^{(m)}(x) + \frac{1}{2} (1 - e^{i2^n x}) P_{k+2^{m-n-1}}^{(m)} \left(x - \frac{\pi}{2^n} \right)$$



2^0 pts 2^1 pts 2^2 pts 2^3 pts.
 2^2 sep 2^1 sep. 2^0 sep 0 sep



How do we get an $N \log N$ operation count?

Let $R(n)$ be the # of multiplication needed to compute coeff in interp polynomial for points $\frac{2\pi j}{n}$, $0 \leq j \leq n-1$.

Then clearly:

$$\underbrace{R(2n)}_{\text{interp poly for } 2n \text{ pts.}} \leq \underbrace{2R(n)}_{\text{compute interp poly on even \& odd}} + \underbrace{2n}_{\text{mult to convert even \& odd into true ones.}}$$

We shall show that $R(2^m) \leq m2^m$ by induction. (100)

$m=0$: no multiplications are involved because constant interpolant

$$\Rightarrow R(2^0) = 0$$

Assume $R(2^m) \leq m2^m$ holds:

$$\begin{aligned} R(2^{m+1}) &= R(2 \cdot 2^m) \leq 2R(2^m) + 2 \cdot 2^m \\ &\leq 2m2^m + 2^{m+1} = (m+1)2^{m+1} \end{aligned}$$

So total number of operations is $O(N \log_2 N) = O(m2^m)$

in Matlab: `fft` and `ifft`

`fft w3` (fastest Fourier transform in the net)

not restricted to powers of 2, but N should have many prime factors

$\triangle$ careful with normalization, e.g. Matlab uses:

$$X_k = \sum_{j=0}^{N-1} x_j e^{-2i\pi k j / N} = N(x, E_k)_N$$

$$\begin{aligned} x_j &= \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{2i\pi k j / N} \quad (\text{inverse fast Fourier transform}) \\ &= (x, \overline{E_j})_N \end{aligned}$$

Convolution using FFT

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Convolution has many uses in e.g. signal processing.

$$h = g * f = f * g$$

$$h(x) = \int_{-\infty}^{\infty} g(y) f(x-y) dy = \int_{-\infty}^{\infty} f(y) g(x-y) dy$$

h = output signal

g = input signal

f = filter "impulse response"

$$g \rightarrow \boxed{f} \rightarrow h = f * g$$

In the discrete case we can think of h, g as sets of N samples:

$$h_n = \sum_{m=0}^{N-1} g_m f_{n-m} \quad \text{= discrete convolution}$$

Or in matrix form:

$$\begin{bmatrix} h_0 \\ h_1 \\ h_2 \\ \vdots \\ h_{N-1} \end{bmatrix} = \begin{bmatrix} f_0 & f_{-1} & f_{-2} & \dots & f_{1-N} \\ f_1 & f_0 & f_{-1} & \dots & f_{2-N} \\ f_2 & f_1 & f_0 & \dots & f_{3-N} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ f_{N-1} & \dots & \dots & \dots & f_0 \end{bmatrix} \begin{bmatrix} g_0 \\ g_1 \\ \vdots \\ g_{N-1} \end{bmatrix} \quad (\star)$$

$$\underline{h} = T_N \underline{g} \quad \text{where } T_N = \text{Toeplitz matrix}$$

naïve implementation would cost $O(N^2)$

Convolution can be done efficiently using FFT.

Here is the key result we need:

$$\begin{aligned} (\text{ifft}(F * G))_n &= \frac{1}{N} \sum_{k=0}^{N-1} F_k G_k e^{\frac{2i\pi kn}{N}} \\ &= \frac{1}{N} \sum_{k=0}^{N-1} \left(\sum_{j=0}^{N-1} f_j e^{-2i\pi kj/N} \right) \left(\sum_{j'=0}^{N-1} g_{j'} e^{-2i\pi kj'/N} \right) e^{2i\pi kn/N} \\ &= \sum_{j=0}^{N-1} f_j \sum_{j'=0}^{N-1} g_{j'} \underbrace{\frac{1}{N} \sum_{k=0}^{N-1} e^{\frac{2i\pi k}{N}(n-j-j')}}_{(E_{n-j}, E_{j'})_N} \end{aligned}$$

Now note we have:

$$(E_{n-j}, E_{j'})_N = \begin{cases} 1 & \text{if } n-j-j' \text{ is divisible by } N \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} \text{ie. } n-j-j' &= 0 \pmod N \\ j' &= n-j \pmod N \end{aligned}$$

Thus:

$$\boxed{(\text{ifft}(F * G))_n = \sum_{j=0}^{N-1} f_j g_{(n-j) \pmod N}}$$

So we can evaluate a slightly different convolution efficiently using FFT:

1. $F = \text{fft}(f); \quad G = \text{fft}(g); \quad \mathcal{O}(N \log N)$
2. $H = F * G \quad \mathcal{O}(N)$
3. $h = \text{ifft}(H) \quad \mathcal{O}(N \log N)$

total $\mathcal{O}(N \log N)$ operations

In matrix vector product for n this becomes.

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$$\begin{bmatrix} h_0 \\ h_1 \\ \vdots \\ h_{N-1} \end{bmatrix} = \underbrace{\begin{bmatrix} f & f^{-1} & f^{-2} & \dots & f^{1-N} \\ f_1 & f_0 & f^{-1} & \dots & f^{2-N} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ f^{1-N} & \dots & \dots & \dots & f \end{bmatrix}}_{C_N} \begin{bmatrix} g_0 \\ g_1 \\ \vdots \\ g_{N-1} \end{bmatrix}$$

$C_N =$ circulant matrix.

So the FFT gives us a tool to compute matrix vector products of circulant matrices in $O(N \log N)$ operations!

How do we use it to compute the matrix product $(*)$?

The trick is to formulate $(*)$ as a $u-v$ prod with larger matrix that is circulant.

$$\text{Let } B_N = \begin{bmatrix} 0 & f_{N-1} & f_{N-2} & \dots & f_2 & f_1 \\ f_{1-N} & 0 & f_{N-1} & \dots & f_2 & f_1 \\ f_{2-N} & f_{1-N} & 0 & \dots & f_3 & f_2 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ f_{1-N} & f_{2-N} & \dots & f_{1-N} & 0 & \end{bmatrix} \in \mathbb{R}^{N \times N}$$

It is easy to check that $C = \begin{bmatrix} T_N & B_N \\ B_N & T_N \end{bmatrix}$ is a circulant matrix

so we can efficiently evaluate:

in $O(2N \log(2N))$
 $= O(N \log N)$ operations.

$$\begin{bmatrix} T_N g \\ B_N g \end{bmatrix} = \underbrace{\begin{bmatrix} T_N & B_N \\ B_N & T_N \end{bmatrix}}_C \begin{bmatrix} g \\ 0 \end{bmatrix}$$

$\leftarrow$ zero padding

throw away