

The Discrete Fourier Transform

The continuous case: any smooth, periodic, complex-valued function $f: [0, 2\pi) \rightarrow \mathbb{C}$ can be written in terms of its Fourier series:

$$f(\theta) = \sum_{n=-\infty}^{\infty} \alpha_n e^{in\theta}$$

$$\alpha_n = \frac{1}{2\pi} \int_0^{2\pi} f(\theta) e^{-in\theta} d\theta \quad (\text{by the orthogonality of } e^{in\theta}, \dots)$$

The "smoother" f is, the faster the coefficients α_n decay.

Let's imagine a discretization of the formula for α_n using the trapezoidal rule:

$$\begin{aligned} \alpha_n &\approx \frac{1}{2\pi} \frac{1}{N} \sum_{k=1}^N f(\theta_k) e^{-in\theta_k}, & \theta_k &= \frac{(k-1)2\pi}{N} \\ &= \frac{1}{2\pi} \frac{1}{N} \sum_{k=1}^N f(\theta_k) e^{-in(k-1)2\pi/N} \end{aligned}$$

With this in mind, we now define the

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Discrete Fourier Transform as:

$$\hat{f}_k = \sum_{j=1}^N f_j e^{-2\pi i(k-1)(j-1)/N} \quad k=1, \dots, N.$$

One interpretation of \hat{f}_k is as ~~an~~ ^{computing an} approximation

to α_k , the true Fourier series coefficient of f . (properly scaled)

(f_j are merely values, they can be assumed to be equispaced samples of a function on some interval).

$\hat{f}_k \approx \alpha_k$ if $N \gtrsim 2k$. (Nyquist sampling for periodic functions).

The Inverse DFT is:

$$f_j = \sum_{k=1}^N \hat{f}_k e^{2\pi i(k-1)(j-1)/N}$$

Let F = DFT matrix: $F_{jk} = e^{-2\pi i(j-1)(k-1)/N}$.

$$F^{**} F^* = N \cdot I \quad (\text{I.e. } F^{-1} = \frac{1}{N} F^*)$$

F^* = IDFT matrix

complex conjugate transpose.

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$$(F^* F)_{jk} = \sum_{n=1}^N e^{2\pi i(j-1)(n-1)/N} e^{-2\pi i(n-1)(k-1)/N}$$

$$= \sum_{n=1}^N e^{\frac{2\pi i(n-1)(j-1-k+1)}{N}}$$

$$= \sum_{n=1}^N e^{2\pi i(n-1)(j-k)/N}$$

$$\text{If } j=k, \Rightarrow \sum e^0 = N$$

$$\text{If } j \neq k, \Rightarrow \sum e^{2\pi i(n-1)(j-k)/N} = e^{-2\pi i(j-k)/N} \underbrace{\sum_{n=1}^N e^{2\pi i n/N}}_{=0}$$

HW exercise.

The direct application of F or F^* requires $\mathcal{O}(n^2)$ calculation since they are dense.

Alternative forms:

$$\hat{f}_k = \sum_{l=-\frac{N}{2}+1}^{\frac{N}{2}} f_l e^{-2\pi i l k / N}$$

(N even, slightly more convenient)

$$\text{or } \hat{f}_k = \sum_{l=-\frac{(N-1)}{2}}^{\frac{N-1}{2}} f_l e^{-2\pi i l k / N}$$

(N odd).

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For simplicity, let's look at the form

$$\hat{f}_k = \sum_{l=0}^{N-1} f_l e^{-2\pi i k l / N}, \quad k=0, \dots, N-1$$

from now ~~on~~ on.

Can we compute this sum, for all k , faster than $O(N^2)$? Observe that:

$$\hat{f}_k = \sum_{l=0}^{N-1} f_l e^{-2\pi i k l / N} = \sum_{\text{even } l} f_l \omega_N^{-k l} + \sum_{\text{odd } l} f_l \omega_N^{-k l}$$

when $\omega_N = e^{2\pi i / N}$

$$= \sum_{l=0}^{N/2} f_{2l} \omega_N^{-k(2l)} + \sum_{l=0}^{N/2} f_{2l+1} \omega_N^{-k(2l+1)}$$

$$= \sum_{l=0}^{N/2} f_{2l} \underbrace{\omega_{N/2}^{-k l}}_{\substack{\downarrow \\ e^{-2\pi i k l / N/2}}} + \omega_N^{-k} \sum_{l=0}^{N/2} f_{2l+1} \underbrace{\omega_{N/2}^{-k l}}$$

This is the sum of two DFTs, each of size $N/2$.

But $k=0, \dots, N-1$

what happens for $k > \frac{N}{2}-1$

$$w_{N/2}^{kl} = e^{-2\pi i k l / N/2}$$

Let $k = \frac{N}{2} + j, \quad j \geq 0, \dots$

$$= e^{-2\pi i (\frac{N}{2} + j) \cdot l / N/2}$$

$$= \underbrace{e^{-2\pi i l}}_{=1} \cdot e^{-2\pi i j l / N/2} = e^{-2\pi i j l / N/2}$$

We say that $k = \frac{N}{2} + j$ aliases to the ~~freq~~ frequency j (or mode j).

$$\text{So } F_N \vec{f} = \begin{pmatrix} F_{N/2} & W_N^* F_{N/2} \\ F_{N/2} & -W_N^* F_{N/2} \end{pmatrix} \begin{pmatrix} \vec{f}_{\text{even}} \\ \vec{f}_{\text{odd}} \end{pmatrix} \quad W_N = \begin{pmatrix} W_N^0 & & \\ & \dots & \\ & & W_N^{N/2-1} \end{pmatrix}$$

since $e^{-2\pi i (\frac{N}{2} + j) / N} = -e^{-2\pi i j / N}$

$$= \begin{pmatrix} F_{N/2} & W_N F_{N/2} \\ F_{N/2} & -W_N F_{N/2} \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ 0 & 0 & 0 & 0 & 1 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 1 & 0 & 0 & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & 1 & 0 & \dots \end{pmatrix} \vec{f}$$

↑ diagonal matrix

extract even and odd parts

The cost for this calculation is

$$\underbrace{2 \cdot \mathcal{O}\left(\left(\frac{N}{2}\right)^2\right)}_{\text{the DFTs}} + \underbrace{2 \cdot \frac{N}{2}}_{\text{scaling by } W_N} + \underbrace{2 \cdot N}_{\text{Addition } (W_N^{-k})}$$

$$= \mathcal{O}\left(\frac{N^2}{2}\right) + \mathcal{O}(N)$$

decreased by factor of two.

Continuing to split each of the $F_{N/2}$ s we eventually arrive at a cost of $\mathcal{O}(N \log_2 N)$

there are \log_2 splittings that occur.

(In fact, the cost is basically $\underbrace{5N \log_2 N}_{\text{Very fast.}}$.)

Note This algorithm was exact ← consequence of the algebraic properties of $e^{-2\pi ijk/N}$. The only source of error in the computation is numerical round-off, but $K(F) = 1$, so this is very minimal.

Computing convolution with the FFT.

Type 2: ~~per~~

In the continuous case, $f * g$ is given by

$$\begin{aligned}(f * g)(x) &= \int f(x-t) g(t) dt \\ &= \int f(t) g(x-t) dt\end{aligned}$$

the Fourier transform of a convolution is given by the product of the Fourier transforms:

$$\mathcal{F}(f * g) = \hat{f} \cdot \hat{g}$$

Discretely, the N -point cyclic convolution is given

by

$$h_n = f * g$$

$$h_n = \sum_{l=0}^{N-1} f_{n-l} g_l = \sum_{l=0}^{N-1} f_l g_{n-l}$$

where f, g are periodic sequences.

The DFT of $\mathcal{F}(f * g) = \mathcal{D}(f) \cdot \mathcal{D}(g)$

$$\text{i.e. } \hat{h}_n = \hat{f}_n \cdot \hat{g}_n$$

But the point is that this actual convolution
can be computed fast using the FFT.

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$$\text{since } h = \frac{1}{N} \mathcal{D}^{-1} \mathcal{D}(f * g)$$

$$= \frac{1}{N} \mathcal{D}^{-1}(\hat{f} \cdot \hat{g})$$

↑ ↑
compute using FFT, $O(N \log N)$

$$* = f * g \text{ by IFFT, } O(N \log N).$$

Exact statement, again.

Fun things with the FFT

Clenshaw-Curtis quadrature: Expand function f in Chebyshev series ^{cosine series} and then integrate. Direct cost: $\mathcal{O}(n^2)$ if f is sampled at n points.

→ Accelerate via the FFT.

$$\int_{-1}^1 f(x) dx = \int_0^\pi f(\cos\theta) \sin\theta d\theta \quad (\text{let } x = \cos\theta)$$

$$\text{If } f(\cos\theta) = \frac{a_0}{2} + \sum_{k=1}^{\infty} a_k \cos k\theta, \text{ then we can}$$

do this integral exactly:

$$\int_0^\pi \sin\theta d\theta = 2$$

$$\int_0^\pi \cos k\theta \sin\theta d\theta = \begin{cases} 0 & k \text{ odd} \\ \frac{2}{1-k^2} & k \text{ even} \end{cases}$$

$$\Rightarrow \int_0^\pi f(\cos\theta) \sin\theta d\theta = \int_0^\pi \sin\theta \left(\frac{a_0}{2} + \sum_{k=1}^{\infty} a_k \cos k\theta \right) d\theta$$

$$= a_0 + \sum_{k=1}^{\infty} \frac{2a_{2k}}{1-4k^2}$$

only even terms.

This computation requires that we compute the coefficients a_0, a_1, \dots

$$a_n = \frac{1}{2\pi} \int_0^{2\pi} f(\cos\theta) \cos n\theta \, d\theta$$

$$a_n = \frac{1}{\pi} \int_0^{\pi} f(\cos\theta) \cos n\theta \, d\theta$$

How can we compute these terms using the FFT?

One option:

$f(\cos\theta) = \frac{a_0}{2} + \sum_{n=1}^{\infty} a_n \cos n\theta$ is the cosine series for $f(\cos\theta)$

Alternatively, f also admits a standard Fourier series:

$$f(\cos\theta) = \sum_{-\infty}^{\infty} \beta_n e^{in\theta}$$

Equating terms, we see that

$$\beta_0 = \frac{a_0}{2}$$

~~$\beta_n e^{in\theta} + \beta_{-n} e^{-in\theta} = \beta_n \cos n\theta + \beta_{-n} \sin n\theta$~~

~~$+ \beta_n \cos n\theta + \beta_{-n} \sin n\theta$~~

~~$= (\beta_n + \beta_{-n}) \cos n\theta + (\beta_n - \beta_{-n}) \sin n\theta$~~
 ~~$= a_n$~~ ~~$= 0$~~

Recall, $\cos n\theta = \frac{e^{int} + e^{-int}}{2}$

$\Rightarrow f(\cos\theta) = \frac{a_0}{2} + \sum_{n \neq 0} \frac{a_n}{2} e^{int}$

$\Rightarrow \text{Re}(\beta_n) = \frac{a_n}{2}$

We only require that $\text{Im}(\beta_n) = -\text{Im}(\beta_{-n})$

Since if $\varphi(\theta)$ is real, then

$\varphi(\theta) = \sum_{-\infty}^{\infty} \alpha_n e^{int}$

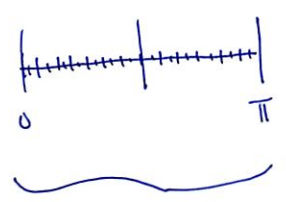
$\overline{\varphi(\theta)} = \overline{\sum_{-\infty}^{\infty} \alpha_n e^{int}} = \sum_{-\infty}^{\infty} \overline{\alpha_n} e^{-int} = \sum_{-\infty}^{\infty} \overline{\alpha_{-n}} e^{int}$

$\Rightarrow \alpha_n = \overline{\alpha_{-n}} \checkmark$

Therefore by computing the FFT of $f(\cos\theta)$ to calculate β_n 's, we can compute these integrals.

Advantages: Nested quadrature:

$\int_0^{\pi} f = \int_0^{\pi/2} f + \int_{\pi/2}^{\pi} f$



splitting the interval does not require the eval of f at all new points.

Option 2 for computing the coefficients a_n :

The discrete cosine transform: (DCT)

Very similar to the DFT, but the identities are slightly more complicated:

$$\hat{X}_k = \sum_{\substack{l=0 \\ l \neq 0}}^{N-1} X_l \cos\left(\frac{\pi k}{N} \left(l + \frac{1}{2}\right)\right) \quad \text{for } k=0, \dots, N-1$$

Other forms:

$$= X_0 + (-1)^k X_{N-1} + 2 \sum_{l=1}^{N-2} X_l \cos\left(\frac{\pi k l}{N-1}\right)$$

clearly the trapezoidal rule on $[0, \pi]$

Connection with Chebyshev polynomials

$$f(x) = \sum_{n=0}^{\infty} \alpha_n T_n(x) \quad \text{under } x = \cos \theta$$

$$= \sum_{n=0}^{\infty} \alpha_n \cos(n \arccos x)$$

$$f(\cos \theta) = \sum \alpha_n \cos(n\theta)$$

← FFT techniques can be used to compute α_n .