RADIX-2 FAST FOURIER TRANSFORM

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1 Discrete Fourier transform

The *discrete*, or *finite*, *Fourier transform* (DFT) of a (complex) vector \mathbf{x} with N elements $(x_0, x_1, ..., x_{N-1}) = \{x_n\}$ is another vector \mathbf{X} with N elements $(X_0, X_1, ..., X_{N-1}) = \{X_k\}$,

$$X_k = \sum_{n=0}^{N-1} x_n e^{-\frac{2\pi i}{N}kn} = \sum_{n=0}^{N-1} \omega^{kn} x_n,$$
(1)

where $i = \sqrt{-1}$, k = 0, 1, ..., N - 1, and we introduce the notation

$$\omega = e^{-\frac{2\pi i}{N}}. (2)$$

The discrete Fourier transform can be expressed with matrix-vector notation:

$$\mathbf{X} = \mathbf{F}_N \, \mathbf{x},\tag{3}$$

where the Fourier matrix F has the elements

$$(F_N)_{kn} = \omega^{kn}, \qquad k, n = 0, 1, ..., N - 1.$$
 (4)

$$F_{N} = \begin{pmatrix} \omega^{0} & \omega^{0} & \omega^{0} & \dots & \omega^{0} \\ \omega^{0} & \omega^{1} & \omega^{2} & \dots & \omega^{N-1} \\ \omega^{0} & \omega^{2} & \omega^{4} & \dots & \omega^{2(N-1)} \\ \omega^{0} & \omega^{3} & \omega^{6} & \dots & \omega^{3(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \omega^{0} & \omega^{N-1} & \omega^{2(N-1)} & \dots & \omega^{(N-1)^{2}} \end{pmatrix}.$$
 (5)

In matlab the DFT can be codes, for example, as shown in Listing 1.

```
function X = mynaivedft(x)

% MYNAIVEDFT - naive implementation of the discrete Fourier transform
    np = length(x);
    omega = exp(-2*pi*1i/np);
    n = 0:np-1;
    k = n';
    F = omega.^(k*n);
    X = F*x;
end
```

Listing 1: Naive MATLAB implementation of the discrete Fourier transform

Direct application of the definition Eq. (1) shown Listing 1 in requires N multiplications and N additions for each of the N components of \mathbf{X} for a total of $2N^2$ floating-point operations. This does not include the generation of the matrix F.

2 Radix-2 algorithm

Radix-2 algorithm is a member of the family of so called Fast Fourier transform (FFT) algorithms. It computes separately the DFTs of the even-indexed inputs $(x_0, x_2, ..., x_{N-2})$ and of the odd-indexed inputs $(x_1, x_3, ..., x_{N-1})$, and then combines those two results to produce the DFT of the whole sequence. This idea can then be performed recursively to reduce the overall runtime from $O(N^2)$ to $O(N \log N)$. Radix-2 algorithm requires that N

is a power of two; since the number of sample points N can usually be chosen freely by the application, this is often not an important restriction.

To derive the algorithm, lets rearrange the DFT of \mathbf{x} , Eq. (1), into two parts: a sum over the even-numbered indices and a sum over the odd-numbered indices:

$$X_k = \sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N}(2m)k} + \sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N}(2m+1)k}.$$
 (6)

One can factor a common multiplier $e^{-\frac{2\pi i}{N}k}$ out of the second sum.

$$X_{k} = \sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{N}(2m)k} + e^{-\frac{2\pi i}{N}k} \sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{N}(2m)k}.$$
 (7)

The two sums in Eq. (7) are the DFT of the even-indexed part and the DFT of odd-indexed part of x_n . Denote the DFT of the even-indexed inputs by E_k and the DFT of the odd-indexed inputs by O_k and we obtain:

$$X_{k} = \underbrace{\sum_{m=0}^{N/2-1} x_{2m} e^{-\frac{2\pi i}{(N/2)} m k}}_{\text{DFT of even-indexed part}} + e^{-\frac{2\pi i}{N} k} \underbrace{\sum_{m=0}^{N/2-1} x_{2m+1} e^{-\frac{2\pi i}{(N/2)} m k}}_{\text{DFT of odd-indexed part}} = E_{k} + e^{-\frac{2\pi i}{N} k} O_{k}.$$
(8)

As the functions of k E_k and O_k are periodic with the period N/2:

$$E_{k+\frac{N}{2}} = E_k \tag{9}$$

and

$$O_{k+\frac{N}{2}} = O_k. (10)$$

Therefore, we can rewrite Eq. (8) as

$$X_{k} = \begin{cases} E_{k} + e^{-\frac{2\pi i}{N}k} O_{k} & \text{for } 0 \le k < N/2, \\ E_{k-N/2} + e^{-\frac{2\pi i}{N}k} O_{k-N/2} & \text{for } N/2 \le k < N, \end{cases}$$
(11)

where we used the periodicity of O_k and E_k to translate the index k.

Using the following property of the *twiddle* factor $e^{-2\pi i k/N}$,

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

we can rewrite X_k as:

$$\begin{array}{rcl} X_k & = & E_k + e^{-\frac{2\pi i}{N}k} O_k, \\ X_{k+\frac{N}{2}} & = & E_k - e^{-\frac{2\pi i}{N}k} O_k. \end{array}$$

This result, expressing the DFT of length N recursively in terms of two DFTs of size N/2, is the core of the radix-2 fast Fourier transform.

```
function X = myradix2dft(x)
   % MYRADIX2DFT radix-2 discrete Fourier transform
       np = length(x); % must be a power of two
       if np == 1
            X = x;
5
       else
            xe = x(1:2:end);
            xo = x(2:2:end);
            xe = myradix2dft(xe);
            xo = myradix2dft(xo);
10
            omega = exp(-2*pi*1i/np);
11
            k = (0:(np/2-1))';
12
            w = omega.^k;
13
            xo = w.*xo;
            X = [xe+xo; xe-xo];
15
       end
   end
17
```

Listing 2: MATLAB implementation of radix-2 discrete Fourier transform

3 The number of floating point operations

The DFT of length N is expressed in terms of two DFTs of length N/2, then four DFTs of length N/4, then eight DFTs of length N/8, and so on until we reach N DFTs of length one. An DFT of length one is just the number itself. If $N = 2^p$, the number of steps in the recursion is $p = \log_2 N$. There is O(N) work at each step, independent of the step number, so the total amount of work is $O(Np) = O(N \log_2 N)$.

4 Inverse DFT

The Fourier matrix F_N has the explicit inverse:

$$\left(\mathbf{F}_{N}^{-1}\right)_{kn} = \frac{1}{N}\omega^{-kn}, \qquad k, n = 0, 1, \dots, N - 1,$$
 (12)

or

$$F_{N}^{-1} = \frac{1}{N} \begin{pmatrix} \omega^{0} & \omega^{0} & \omega^{0} & \dots & \omega^{0} \\ \omega^{0} & \omega^{-1} & \omega^{-2} & \dots & \omega^{-(N-1)} \\ \omega^{0} & \omega^{-2} & \omega^{-4} & \dots & \omega^{-2(N-1)} \\ \omega^{0} & \omega^{-3} & \omega^{-6} & \dots & \omega^{-3(N-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \omega^{0} & \omega^{-(N-1)} & \omega^{-2(N-1)} & \dots & \omega^{-(N-1)^{2}} \end{pmatrix}.$$
(13)

Eq. (1) can be inverted as following:

$$x_k = \frac{1}{N} \sum_{n=0}^{N-1} X_n \, \omega^{-k \, n} = \frac{1}{N} \sum_{n=0}^{N-1} X_n \, e^{\frac{2\pi i}{N} k \, n}. \tag{14}$$

To prove that indeed,

$$F_N F_N^{-1} = F_N^{-1} F_N = I, (15)$$

where I is the identity matrix, notice that

$$1 + \omega + \omega^2 + \omega^3 + \dots + \omega^{N-1} = 0, \tag{16}$$

since the sum on the left of Eq. (16) is an N-terms geometric progression with the start value $\omega^0 = 1$ and the common ratio ω . The value of the sum is

$$\frac{1-\omega^N}{1-\omega} = 0\tag{17}$$

since

$$\omega^{N} = \left(e^{-\frac{2\pi i}{N}}\right)^{N} = e^{-2\pi i} = 1,\tag{18}$$

and thus the numerator in Eq. (17) is zero whereas the denominator is not.

Similarly, we can show that

$$1 + \omega^2 + \omega^4 + \omega^6 + \dots + \omega^{2(N-1)} = 0, \tag{19}$$

$$1 + \omega^3 + \omega^6 + \omega^9 + \dots + \omega^{3(N-1)} = 0, \tag{20}$$

and in general

$$1 + \omega^k + \omega^{2k} + \omega^{3k} + \dots + \omega^{(N-1)k} = 0, \qquad k = 1, \dots, N-1 \text{ and } k = -N+1, \dots, -1.$$
 (21)

However, when k = 0 the sum in Eq. (21)

$$1 + \omega^k + \omega^{2k} + \omega^{3k} + \dots + \omega^{(N-1)k} = 1 + 1 + 1 + \dots + 1 = N.$$
 (22)

Summarizing Eqs. (21), (22):

$$\sum_{l=0}^{N-1} \omega^{lk} = N \,\delta_{k0} = \begin{cases} 0, & k = 1, \dots, N-1 \text{ and } k = -N+1, \dots, -1\\ N, & k = 0, \end{cases}$$
 (23)

where δ_{mn} is the Kronecker symbol.

Finally,

$$\left(\mathbf{F}_{N}\,\mathbf{F}_{N}^{-1}\right)_{kp} = \sum_{l=0}^{N-1} \left(\mathbf{F}_{N}\right)_{kl} \left(\mathbf{F}_{N}^{-1}\right)_{lp} = \frac{1}{N} \sum_{l=0}^{N-1} \omega^{kl} \omega^{-lp} = \frac{1}{N} \sum_{l=0}^{N-1} \omega^{l(k-p)} = \delta_{kp}. \tag{24}$$