# **Fast Fourier Transform**

Přemysl Šůcha

Based on the texts: Ananth Grama, Anshul Gupta, George Karypis, and Vipin Kumar ``Introduction to Parallel Computing'', Addison Wesley, 2003, and Paul Heckbert ``Fourier Transforms and the Fast Fourier Transform (FFT) Algorithm'', 1 Carnegie Mellon School of Computer Science, 1998.

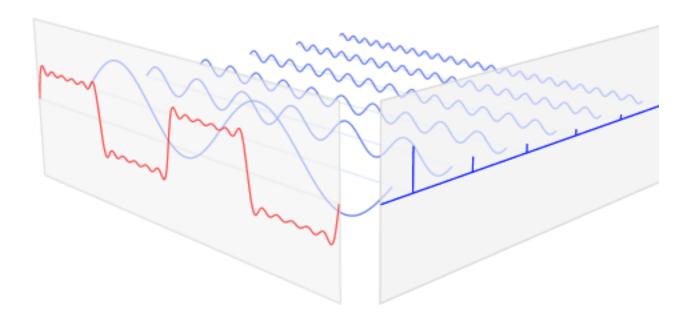
# **Topic Overview**

- Introduction to Fast Fourier Transform
- Binary-Exchange algorithm
- Transpose algorithm

# **Introduction to Fast Fourier Transform**

- The discrete Fourier transform (DFT) plays an important role in many applications, including digital signal processing, image filtering, solutions to linear partial differential equations, convolution ...
- The DFT is a linear transformation that maps n regularly sampled points from a cycle of a periodic signal onto an equal number of points representing the frequency spectrum of the signal.
- In 1965, Cooley and Tukey devised an algorithm to compute the DFT of an n-point series in O(n log n) operations. Its variations are referred to as the Fast Fourier Transform (FFT).

### **Fourier Transform**



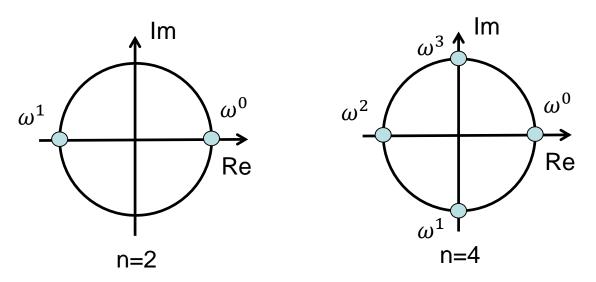
Transformation of a signal (red) onto an equal number of points representing the **frequency spectrum** (blue).

 Consider a sequence X = <X [0], X [1], ..., X [n - 1]> of length n. The discrete Fourier transform of the sequence X is the sequence Y = <Y [0], Y [1], ..., Y [n - 1]>, where

$$Y[l] = \sum_{k=0}^{\infty} X[k] \omega^{kl}, 0 \le l < n$$

•  $\omega$  is the *n*-th root of unity in the complex plane; that is  $\omega = e^{2\pi\sqrt{-1}/n}$ .

- The powers of  $\omega$  used in an FFT computation are also known as **twiddle factors**.
- Note:  $\omega = e^{2\pi\sqrt{-1}/n} = \cos\left(\frac{2\pi}{n}\right) + i\sin\left(\frac{2\pi}{n}\right)$



• Power of roots of unity are **periodic** with period *n*.

- The sequential complexity of computing the entire sequence Y of length n is O(n<sup>2</sup>).
- The **fast Fourier transform** algorithm reduces this complexity to O(*n* log *n*).
- The FFT algorithm is based on the idea that permits an *n*-point DFT computation to be **split into two (***n***/2)-point DFT** computations.

# Two-point DFT (n=2)

- For *n*=2 the twiddle factor is  $\omega = e^{2\pi\sqrt{-1}/n} = e^{-\pi i} = -1^{(*)}$ .
- Then DFT is:

$$Y[l] = \sum_{k=0}^{n-1} X[k] (-1)^{kl} = X[0](-1)^{l0} + X[1](-1)^{l1} = X[0] + X[1](-1)^{l}$$

SO

$$Y[0] = X[0] + X[1] \qquad and \qquad Y[1] = X[0] - X[1]$$

 $^{(*}e^{i\theta} = \cos\theta + i\sin\theta$ 

### Four-point DFT (n=4)

- For *n*=4 the twiddle factor is  $\omega = e^{-i\pi/2} = -i$ .
- Then DFT is:

$$Y[l] = \sum_{k=0}^{n-1} X[k] (-i)^{kl} =$$
$$= X[0] + X[1](-i)^{l} + X[2](-1)^{l} + X[3]i^{l}$$

SO

$$\begin{split} Y[0] &= X[0] + X[1] + X[2] + X[3], \\ Y[1] &= X[0] - iX[1] - X[2] + iX[3], \\ Y[2] &= X[0] - X[1] + X[2] - X[3], \\ Y[3] &= X[0] + iX[1] - X[2] - iX[3]. \end{split}$$

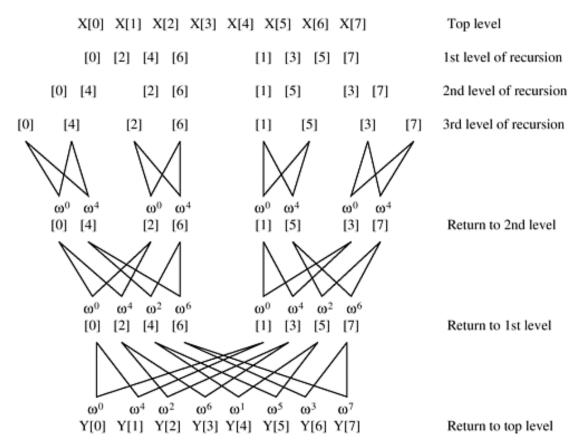
# Four-point DFT (n=4)

• To compute Y faster, we can precompute common subexpressions:

$$\begin{split} Y[0] &= (X[0] + X[2]) + (X[1] + X[3]), \\ Y[1] &= (X[0] - X[2]) - i(X[1] - X[3]), \\ Y[2] &= (X[0] + X[2]) - (X[1] + X[3]), \\ Y[3] &= (X[0] - X[2]) + i(X[1] - X[3]). \end{split}$$

 Pre-computation of brackets in two-point DFT can save a lot of addition operations.

#### A recursive unordered FFT



If *n* is a power of two (e.g. 8 in the figure above), each of these DFT **computations can be divided similarly into smaller computations** in a **recursive manner**. This leads to the recursive FFT algorithm. 11

 This FFT algorithm is called the radix-2 algorithm because at each level of recursion, the input sequence is split into two equal halves.

```
1. procedure R_FFT(X, Y, n, w)
2. if (n = 1) then Y[0] := X[0] else
3. begin
4. R_FFT(<X[0], X[2], ..., X[n - 2]>, <Q[0], Q[1], ..., Q[n/2]>, n/2, w<sup>2</sup>);
5. R_FFT(<X[1], X[3], ..., X[n - 1]>, <T[0], T[1], ..., T[n/2]>, n/2, w<sup>2</sup>);
6. for i := 0 to n - 1 do
7. Y[i] :=Q[i mod (n/2)] + w<sup>i</sup> T[i mod (n/2)];
8. end R_FFT
```

- The maximum **number of levels of recursion is log** *n* for an initial sequence of length *n*.
- The total number of **arithmetic operations** (line 7) at each level is O(*n*).
- The overall sequential complexity of the algorithm is O(n log n).
- The serial FFT algorithm can also be cast in an iterative form.
- An iterative FFT algorithm is derived by casting each level of recursion, starting with the deepest level, as an iteration.

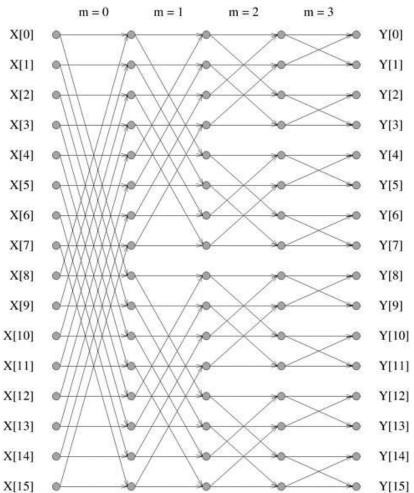
### **Cooley-Tukey algorithm**

The outer loop (line 5) is executed log *n* times for an n-point FFT, and the inner loop (line 8) is executed *n* times during each iteration of the outer loop.

```
procedure ITERATIVE FFT(X, Y, n)
1.
2.
   begin
      r := \log n;
3.
4.
      for i := 0 to n - 1 do R[i] := X[i];
      5.
6.
          begin
             for i := 0 to n - 1 do S[i]:= R[i];
7.
             for i := 0 to n - 1 do /* Inner loop */
8.
9.
                 begin
                    /* Let (b0b1 ··· br -1) be the binary representation of i */
10.
11.
                    j := (b0...bm-1 0 bm+1...br -1);
                    k := (b0...bm-1 \ 1 \ bm+1...br \ -1);
12.
                    R[i] := S[i] + S[k] \times \omega^{(bm, bm-1, b0, 0, ..., 0)};
13.
             endfor; /* Inner loop */
14.
      endfor;
                 /* Outer loop */
15.
      for i := 0 to n - 1 do Y[i] := R[i];
16.
17. end ITERATIVE FFT
```

14

### **Cooley-Tukey algorithm**

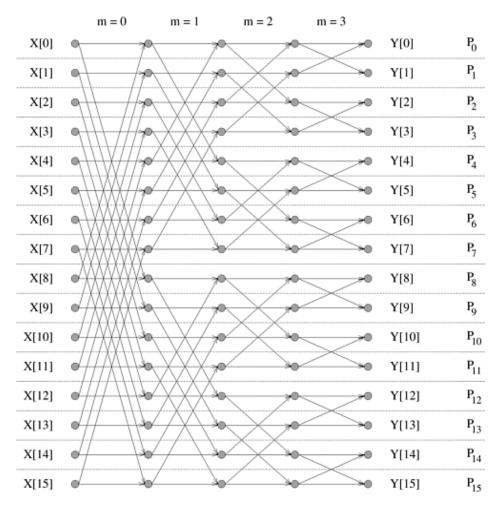


The pattern of combination of elements of the input and the intermediate sequences during a **16-point** unordered **FFT** computation. 15

# **Binary-Exchange algorithm**

- The decomposition for the parallel algorithm is induced by **partitioning the input or the output vector**.
- We first consider a simple mapping in which one task is assigned to each process.
- Each task starts with one element of the input vector and computes the corresponding element of the output.
   Process *i* (0 ≤ *i* < *n*) initially stores X [i] and finally generates Y [i].
- In each of the log *n* iterations of the outer loop, process
   *P<sub>i</sub>* updates the value of *R[i]* by executing line 12 of
   Cooley-Tukey algorithm.
- All *n* updates are performed in parallel.

### **16-point FFT on 16 processes**



Parallel mapping where one task is assigned to each process.

# **Binary-Exchange algorithm**

- To perform the updates, process P<sub>i</sub> requires an element of S from a process whose label differs from *i* at only one bit.
- Parallel FFT computation maps naturally onto a hypercube with a one-to-one mapping of processes to nodes.
- In each of the log *n* iterations of this algorithm, every process performs one complex multiplication and addition, and exchanges one complex number with another process.
- Now consider a mapping in which the *n* are mapped onto *p* processes.

### **Binary-Exchange algorithm**

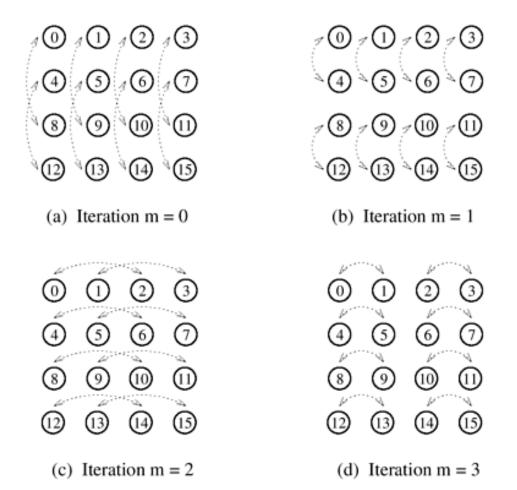
- For the sake of simplicity, let us assume that both n and p are powers of two, i.e., n = 2<sup>r</sup> and p = 2<sup>d</sup>.
- Elements with indices differing at their d (= 2) most significant bits are mapped onto different processes, i.e. there is no interaction during the last r d iterations.
- Each interaction operation exchanges n/p words of data. The time spent in communication in the entire algorithm is  $t_s \log p + t_w(n/p) \log p$ .
- The parallel run time of the algorithm on a *p*-node hypercube network is

$$T_P = t_c \frac{n}{p} \log n + t_s \log p + t_w \frac{n}{p} \log p.$$

# **Transpose Algorithm**

- The binary-exchange algorithm yields good performance on parallel computers with sufficiently high communication bandwidth with respect to the processing speed of the CPUs.
- The simplest transpose algorithm requires a single transpose operation over a two-dimensional array; we call this algorithm the two-dimensional transpose algorithm.
- Assume that  $\sqrt{n}$  is a power of 2, and that the input sequence of size *n* is arranged in a  $\sqrt{n} \times \sqrt{n}$  two-dimensional square array.

#### **Two-dimensional transpose**

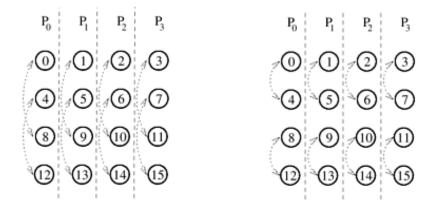


The pattern of combination of elements in a **16-point FFT** when the data are arranged in a **4 x 4 two-dimensional square array**. <sup>21</sup>

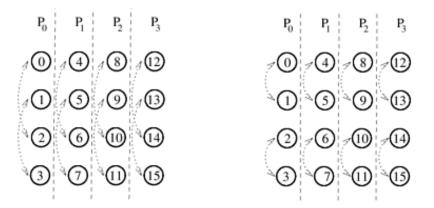
# **Transpose Algorithm**

- The FFT computation in each column can proceed independently for  $\log \sqrt{n}$  iterations without any column requiring data from any other column.
- Similarly, in the remaining  $\log \sqrt{n}$  iterations, computation proceeds independently in each row without any row requiring data from any other row.

#### **Two-dimensional (2D) transpose**



(a) Steps in phase 1 of the transpose algorithm (before transpose)



(b) Steps in phase 3 of the transpose algorithm (after transpose)

The 2D transpose algorithm for a **16-point FFT** on **four processes**. 23

## Transpose Algorithm (n>p)

- The 2D array of data is striped into blocks, and one block of  $\sqrt{n}/p$  rows is assigned to each process.
- In the **first and third phases** of the algorithm, each process computes  $\sqrt{n}/p$  FFTs of  $\sqrt{n}$  each.
- The **second phase** transposes the  $\sqrt{n} \times \sqrt{n}$  matrix (all-to-all personalized communication).
- The parallel run time of the transpose algorithm on a hypercube is:

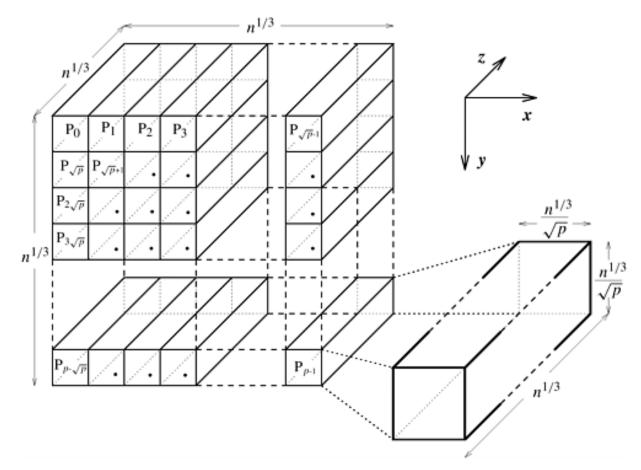
$$T_P = 2t_c \frac{\sqrt{n}}{p} \sqrt{n} \log \sqrt{n} + t_s(p-1) + t_w \frac{n}{p}$$
$$= t_c \frac{n}{p} \log n + t_s(p-1) + t_w \frac{n}{p}$$

#### **Three-dimensional transpose algorithm**

• As an extension of the two-dimensional transpose algorithm, consider the *n* data points to be arranged in an  $n^{1/3} \ge n^{1/3} \ge n^{1/3}$  three-dimensional array mapped onto a logical  $\sqrt{p} \ge \sqrt{p}$  two-dimensional mesh of processes.

• Each process has 
$$(\frac{n^{\frac{1}{3}}}{\sqrt{p}})(\frac{n^{\frac{1}{3}}}{\sqrt{p}})n^{\frac{1}{3}}$$
 elements of data.

#### **Three-dimensional transpose algorithm**



Data distribution in the three-dimensional transpose algorithm for an n-point FFT on p processes ( $\sqrt{p} \le n^{1/3}$ )

# **Three-dimensional transpose algorithm**

- This algorithm can be divided into the following five phases:
  - In the first phase, n<sup>1/3</sup>-point FFTs are computed on all the rows along the z-axis.
  - 2. In the second phase, all the  $n^{1/3}$  cross-sections of size  $n^{1/3}$  x  $n^{1/3}$  along the *y-z* plane are transposed.
  - 3. In the third phase,  $n^{1/3}$ -point FFTs are computed on all the rows of the modified array **along the** *z*-axis.
  - 4. In the fourth phase, each of the  $n^{1/3} \ge n^{1/3}$  cross-sections along the *x-z* plane is transposed.
  - 5. In the fifth and final phase, *n*<sup>1/3</sup>-point FFTs of all the **rows** along the *z*-axis are computed again.

#### **Binary-Exchange vs. Transpose Algorithm**

• Parallel runtime of the **transpose** algorithm

( $T_P = t_c \frac{n}{p} \log n + t_s(p-1) + t_w \frac{n}{p}$ ) has a much higher overhead than the **binary-exchange** algorithm ( $T_P = t_c \frac{n}{p} \log n + t_s \log p + t_w \frac{n}{p} \log p$ .) due to the **message startup** time  $t_s$ , but has a lower overhead due to **perword transfer** time  $t_w$ .

- If the latency t<sub>s</sub> is very low, then the transpose algorithm may be the algorithm of choice.
- The binary-exchange algorithm may perform better on a parallel computer with a high communication bandwidth but a significant startup time.