

# An Efficient and Flexible Parallel FFT Implementation Based on FFTW

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# Discrete Fast Fourier Transform

## Task of 3d-DFT

Consider a three-dimensional dataset of  $n_0 \times n_1 \times n_2$  complex numbers  $\hat{g}_{k_0 k_1 k_2} \in \mathbb{C}$ . Compute

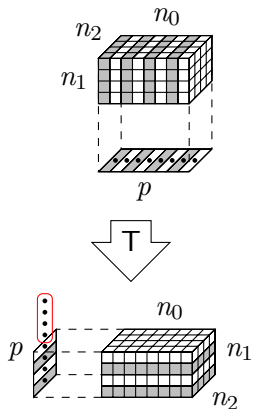
$$\begin{aligned} g_{l_0 l_1 l_2} &:= \sum_{k_0=0}^{n_0-1} \sum_{k_1=0}^{n_1-1} \sum_{k_2=0}^{n_2-1} \hat{g}_{k_0 k_1 k_2} e^{-2\pi i(k_2 \frac{l_2}{n_2} + k_1 \frac{l_1}{n_1} + k_0 \frac{l_0}{n_0})} \\ &= \sum_{k_0=0}^{n_0-1} \left( \sum_{k_1=0}^{n_1-1} \sum_{k_2=0}^{n_2-1} \hat{g}_{k_0 k_1 k_2} e^{-2\pi i(k_2 \frac{l_2}{n_2} + k_1 \frac{l_1}{n_1})} \right) e^{-2\pi i k_0 \frac{l_0}{n_0}} \end{aligned}$$

for all  $l_t = 0, \dots, n_t - 1$  ( $t = 0, 1, 2$ ).

**Realized by 3d-FFT** ( $n_0 = n_1 = n_2 = n$ )

$\Rightarrow \mathcal{O}(n^3 \log n)$  instead of  $\mathcal{O}(n^6)$

# One-Dimensional Data Distribution



$p$  - number of processors

## Features of FFTW [Frigo, Johnson]

- open source
- easy interface
- communicator
- arbitrary size
- $d$ -dim. FFT
- in place FFT
- high performance
- many transforms
- adjust blocksize
- adjust planning
- collect wisdom

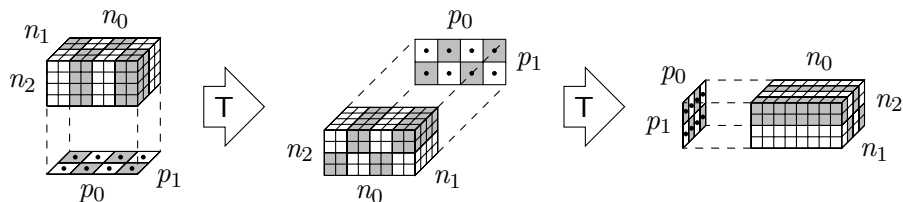
## Maximum Number of Processors $p_{\max}^{1D}$ ( $n_0 = n_1 = n_2 = n$ )

$$p_{\max}^{1D} = n$$

FFTW combines portable performance and good usability, but is not scalable enough.

# Two-Dimensional Data Distribution

[Ding, Eleftheriou et al. 03, Plimpton, Pekurovsky - P3DFFT]



$p_0 \times p_1$  - size of processor grid

**Maximum Number of Processors**  $p_{\max}^{2D}$   
( $n_0 = n_1 = n_2 = n$ )

$$p_{\max}^{2D} = n^2$$

$n$	$p_{\max}^{1D} = n$	$p_{\max}^{2D} = n^2$
64	64	4096
128	128	16384
256	256	65536
512	512	262144
1024	1024	1048576

# Algorithms Supported by FFTW3.3alpha1

## Aim

Implement a new parallel FFT software library (PFFT) based on FFTW and the highly scalable two-dimensional data distribution.

## 1d-FFT Combined with Local Transposition

$$\begin{array}{lcl} \hat{n}_0 \times \hat{n}_1 \times \hat{n}_2 & \begin{array}{c} \text{FFT2} \\ \rightarrow \\ 012 \end{array} & \hat{n}_0 \times \hat{n}_1 \times n_2 \\ \hat{n}_0 \times \hat{n}_1 \times \hat{n}_2 & \begin{array}{c} \text{FFT2} \\ \rightarrow \\ 102 \end{array} & \hat{n}_1 \times \hat{n}_0 \times n_2 \\ \hat{n}_0 \times \hat{n}_1 \times \hat{n}_2 & \begin{array}{c} \text{FFT2} \\ \rightarrow \\ 021 \end{array} & \hat{n}_0 \times n_2 \times \hat{n}_1 \\ (\hat{n}_0 \times \hat{n}_1) \times \hat{n}_2 & \begin{array}{c} \text{FFT2} \\ \rightarrow \\ 201 \end{array} & n_2 \times (\hat{n}_0 \times \hat{n}_1) \end{array}$$

# Algorithms Supported by FFTW3.3alpha1

## Transposition of One-Dimensional Distributed Data

$$N_0 \times \frac{N_1}{P} \xrightarrow{\mathbf{T}} \frac{N_0}{P} \times N_1$$

Group two of the three dimensions to use FFTWs matrix transposition on two-dimensional decomposed data, e.g.

$$N_0 = n_2, N_1 = n_0 \times \frac{n_1}{p_1}, P = p_0.$$

## Transposition of Two-Dimensional Distributed Data

$$n_2 \times \left( \frac{n_0}{p_0} \times \frac{n_1}{p_1} \right) \xrightarrow{\mathbf{T}} \frac{n_2}{p_0} \times \left( n_0 \times \frac{n_1}{p_1} \right)$$

# Two-Dimensional Distributed FFT Based on FFTW

## PFFT Forward Transform

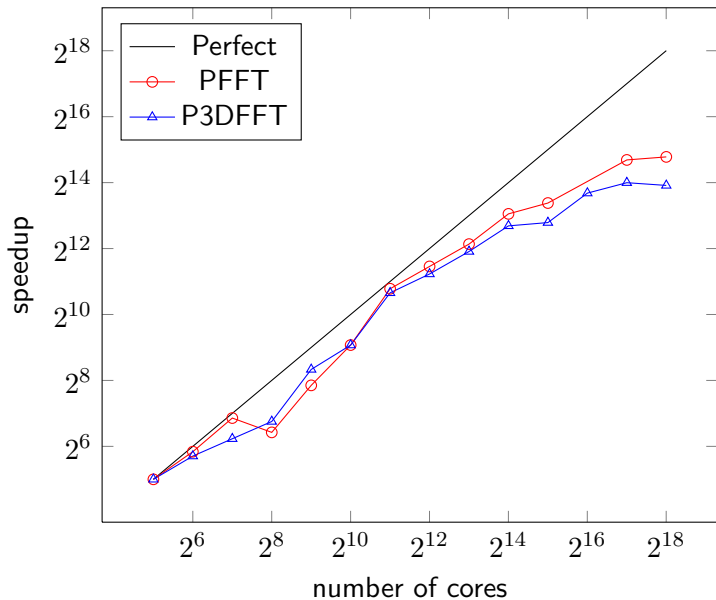
$$\begin{array}{rcc}
 \frac{\hat{n}_0}{p_0} \times \frac{\hat{n}_1}{p_1} \times \hat{n}_2 & \xrightarrow{\text{FFT2}} & n_2 \times \frac{\hat{n}_0}{p_0} \times \frac{\hat{n}_1}{p_1} \xrightarrow{\text{T}} \\
 \frac{n_2}{p_1} \times \frac{\hat{n}_0}{p_0} \times \hat{n}_1 & \xrightarrow{\text{FFT2}} & n_1 \times \frac{n_2}{p_1} \times \frac{\hat{n}_0}{p_0} \xrightarrow{\text{T}} \\
 \frac{n_1}{p_0} \times \frac{n_2}{p_1} \times \hat{n}_0 & \xrightarrow{\text{FFT2}} & \frac{n_2}{p_1} \times \frac{n_1}{p_0} \times n_0
 \end{array}$$

## PFFT Backward Transform

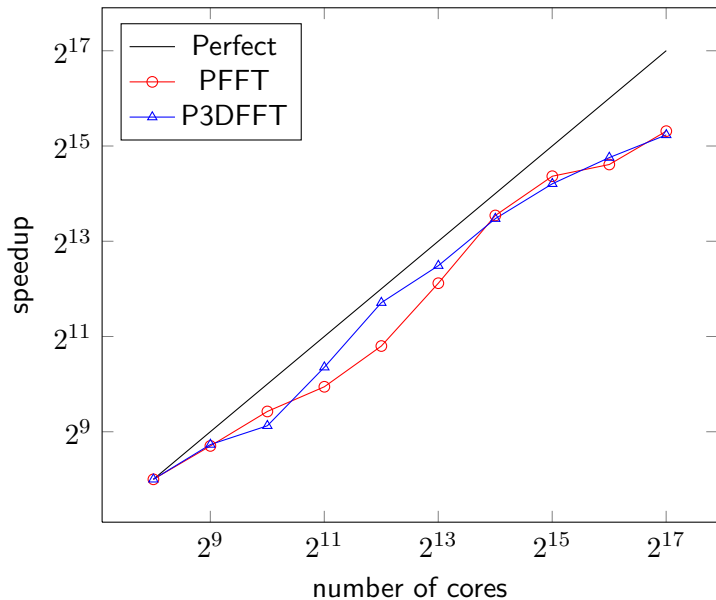
$$\begin{array}{rcc}
 \frac{n_2}{p_1} \times \frac{n_1}{p_0} \times n_0 & \xrightarrow{\text{FFT2}} & \hat{n}_0 \times \frac{n_2}{p_1} \times \frac{n_1}{p_0} \xrightarrow{\text{T}} \\
 \frac{\hat{n}_0}{p_0} \times \frac{n_2}{p_1} \times n_1 & \xrightarrow{\text{FFT2}} & \hat{n}_1 \times \frac{\hat{n}_0}{p_0} \times \frac{n_2}{p_1} \xrightarrow{\text{T}} \\
 \frac{\hat{n}_1}{p_1} \times \frac{\hat{n}_0}{p_0} \times n_2 & \xrightarrow{\text{FFT2}} & \frac{\hat{n}_0}{p_0} \times \frac{\hat{n}_1}{p_1} \times \hat{n}_2
 \end{array}$$



# Scaling FFT of Size $512^3$ on BlueGene/P



# Scaling FFT of Size $1024^3$ on BlueGene/P



# Comparison of PFFT and P3DFFT

## P3DFFT Unique Features [Pekurovsky]

- r2c FFT
- Fortran interface

## Common Features

- open source
- high scalability
- portability
- multiple precisions
- C interface
- ghost cell support

## PFFT Unique Features

- c2c FFT
- completely in place FFT
- FFTW like interface
- basic, advanced and guru interface
- adjustable blocksize
- separate communicator
- accumulated wisdom
- change of planning effort without recompilation
- $d$ -dimensional parallel FFT
- truncated FFT support

# Non-Equispaced Discrete Fourier Transform

## Task of 3d-DFT and 3d-NDFT

For  $\hat{f}_{k_0 k_1 k_2} \in \mathbb{C}$  compute

$$f_{l_0 l_1 l_2} := \sum_{k_0=0}^{N-1} \sum_{k_1=0}^{N-1} \sum_{k_2=0}^{N-1} \hat{f}_{k_0 k_1 k_2} e^{-2\pi i(k_2 \frac{l_2}{N} + k_1 \frac{l_1}{N} + k_0 \frac{l_0}{N})} \quad (\text{DFT})$$

for all  $0 \leq l_t < N$  ( $\Rightarrow 0 \leq \frac{l_t}{N} < 1$ ),  $t = 0, 1, 2$ , and compute

$$f_j := \sum_{k_0=0}^{N-1} \sum_{k_1=0}^{N-1} \sum_{k_2=0}^{N-1} \hat{f}_{k_0 k_1 k_2} e^{-2\pi i(k_2 x_j^{(2)} + k_1 x_j^{(1)} + k_0 x_j^{(0)})} \quad (\text{NDFT})$$

for  $x_j^{(t)} \in [0, 1)$  ( $t = 0, 1, 2$ ),  $j = 1, \dots, M$ .

**Realized by 3d-NFFT [NFFT software library]**

$\Rightarrow \mathcal{O}(N^3 \log N + \log^3(\frac{1}{\epsilon})M)$  instead of  $\mathcal{O}(N^3 M)$

## Matrix-Vector-Notation

$$\mathbf{f} = \mathbf{A}\hat{\mathbf{f}},$$

where  $\mathbf{f} = (f_j)_j$ ,  $\mathbf{A} = (e^{-2\pi i \mathbf{k} \cdot \mathbf{x}_j})_{\mathbf{k}, j}$ ,  $\hat{\mathbf{f}} = (\hat{f}_{k_0 k_1 k_2})_{\mathbf{k}}$

## Approximation [Dutt, Rokhlin 93, Beylkin 95, Steidl 96, ...]

$$\mathbf{f} = \mathbf{A}\hat{\mathbf{f}} \approx \mathbf{BFD}\hat{\mathbf{f}},$$

where

- $\mathbf{D} \in \mathbb{C}^{N^3 \times N^3}$  diagonal matrix,
- $\mathbf{F} \in \mathbb{C}^{n^3 \times N^3}$  truncated Fourier matrix ( $n \geq N$ )
- $\mathbf{B} \in \mathbb{C}^{M \times n^3}$  sparse matrix

# Summary

