Solving Linear Equations, FFT and Sorting on Distributed-Memory Architectures

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Solving System of Linear Equations

- Solving Ax = b as a benchmark to rank world's top supercomputers (https://www.top500.org/).
- We will discuss the classical two-step approach:
 - 1) Gaussian Elimination;
 - 2) Back Substitution.
- We assume a 2D block data distribution on $\sqrt{P} \times \sqrt{P}$ procs.
 - \Box Proc P_{ij} holds matrix block A_{ij}
 - Diagonal processor P_{ii} holds
 vector block b_i and computes x_i

P_{00} A_{00}, b_0 $ \rightarrow x_0$	P ₀₁ A ₀₁	$P_{02} \\ A_{02}$
$\begin{array}{c} P_{10} \\ A_{10} \end{array}$	P_{11} A_{11}, b_1 $\rightarrow x_1$	P ₁₂ A ₁₂
P ₂₀ A ₂₀	P ₂₁ A ₂₁	P_{22} A_{22}, b_2 $\rightarrow x_2$

A Simple Example

$$\begin{bmatrix}
2x_0 + 4x_1 + 6x_2 = 8 \\
3x_0 + 9x_1 + 15x_2 = 21 \\
5x_0 + 12x_1 + 7x_2 = 2
\end{bmatrix}$$

$$\begin{bmatrix}
2 & 4 & 6 \\
3 & 9 & 15 \\
5 & 12 & 7
\end{bmatrix}
\begin{bmatrix}
x_0 \\
x_1 \\
x_2
\end{bmatrix} = \begin{bmatrix}
8 \\
21 \\
2
\end{bmatrix}$$

System of linear equations

*Matrix form
$$Ax = b$$*

A Simple Example

Step 1: Gaussian Elimination

- Reduce Ax = b to Ux = y (row echelon form), where U is an upper triangular matrix (with 1 on diagonal).
- Achieved by dividing a row by a non-zero number and subtracting a row by a multiple of another row.

$$\begin{bmatrix} 2 & 4 & 6 & 8 \\ 3 & 9 & 15 & 21 \\ 5 & 12 & 7 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 3 & 4 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 2 \end{bmatrix}$$

$$A \qquad b \qquad U \qquad y$$

A lower triangular matrix *L* records intermediate steps of this process

$$\begin{bmatrix} 2 & 0 & 0 \\ 3 & 3 & 0 \\ 5 & 2 & -12 \end{bmatrix}$$

A Simple Example

Step 2: Back Substitution

□ Solve for x in Ux = y in reverse order from x_{n-1} to x_0

$$\begin{bmatrix} 1 & 2 & 3 & | & 4 \\ 0 & 1 & 2 & | & 3 \\ 0 & 0 & 1 & | & 2 \end{bmatrix} \rightarrow \begin{cases} x_0 + 2x_1 + 3x_2 = 4 \\ x_1 + 2x_2 = 3 \\ x_2 = 2 \end{cases} \rightarrow \begin{cases} x_0 = 0 \\ x_1 = -1 \\ x_2 = 2 \end{cases}$$

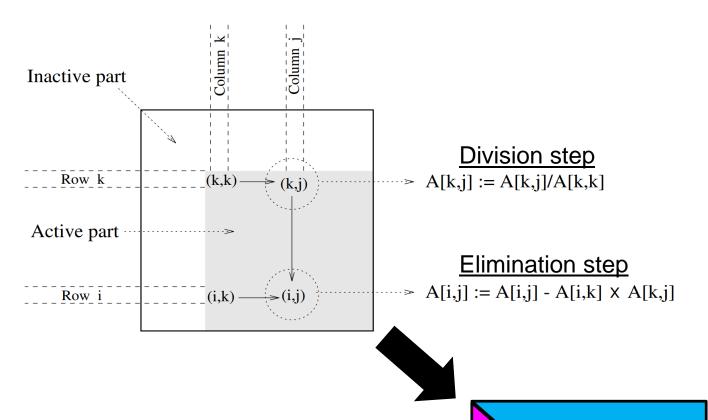
For an $n \times n$ matrix A, Step 1 (Gaussian Elimination) is more expensive taking $O(n^3)$ time, and Step 2 (Back Substitution) takes only $O(n^2)$ time.

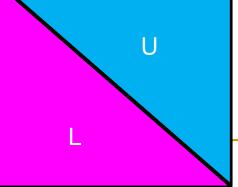
Pseudocodes of Two Steps

```
procedure GAUSSIAN_ELIMINATION (A, b, y)
         begin
3.
            for k := 0 to n-1 do
                                              /* Outer loop */
4.
             begin
5.
               for j := k + 1 to n - 1 do
                   A[k,j] := A[k,j]/A[k,k]; /* Division step */
6.
7.
               y[k] := b[k]/A[k,k];
               A[k, k] := 1;
9.
               for i := k + 1 to n - 1 do
10.
               begin
11.
                   for j := k + 1 to n - 1 do
12.
                      A[i,j] := A[i,j] - A[i,k] \times A[k,j]; /* Elimination step */
13.
                   b[i] := b[i] - A[i, k] \times y[k];
14.
                   A[i, k] := 0;
15.
                               /* Line 9 */
                endfor:
16.
             endfor:
                         /* Line 3 */
17.
         end GAUSSIAN_ELIMINATION
```

```
procedure BACK_SUBSTITUTION (U, x, y)
2.
          begin
3.
             for k := n - 1 downto 0 do /* Main loop */
4.
                begin
5.
                   x[k] := y[k];
6.
                   for i := k - 1 downto 0 do
7.
                      y[i] := y[i] - x[k] \times U[i, k];
8.
                endfor:
9.
         end BACK_SUBSTITUTION
```

Illustration of Gaussian Elimination



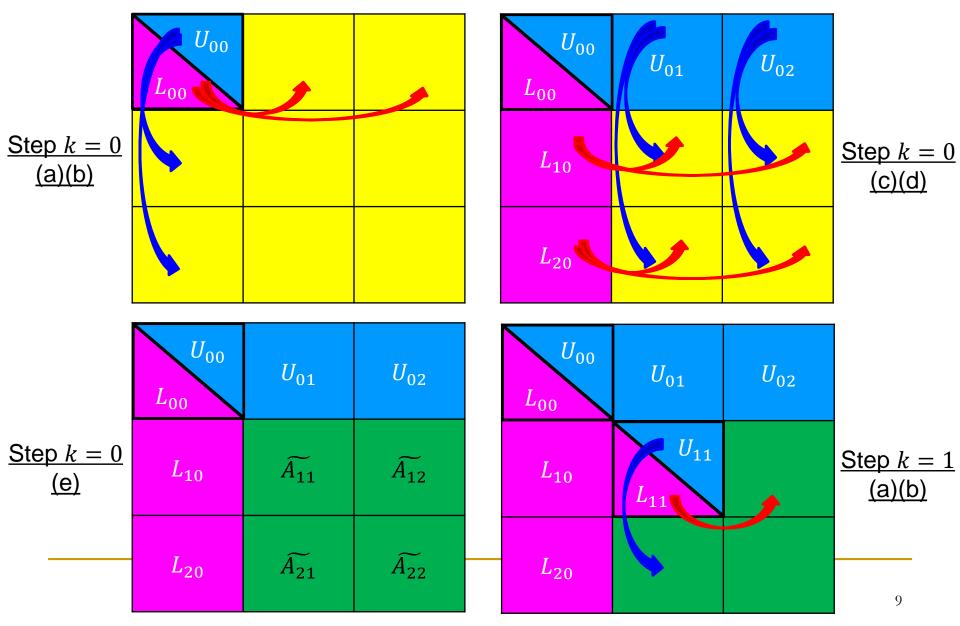


Gaussian Elimination in 2D Blocks

- For step $k = 0, 1, ..., \sqrt{P} 1$:
 - Processor P_{kk} performs local division and elimination on matrix block A_{kk} (resulting in L_{kk} and U_{kk});
 - Processor P_{kk} broadcasts L_{kk} to all processors P_{kj} (j > k), and U_{kk} to all processors P_{ik} (i > k);
 - Processors P_{kj} and P_{ik} (i, j > k) perform local division and elimination (resulting in U_{ki} and L_{ik} , respectively);
 - Processors P_{ki} broadcasts U_{ki} down to other processors in same column, and processors P_{ik} broadcasts L_{ik} to other processors to the right in same row;
 - Processors P_{ij} (i,j > k) perform local division and elimination (resulting in modified block $\widetilde{A_{i,i}}$).

Next slide illustrates one step for P = 9 processors $_{s}$

Gaussian Elimination in 2D Blocks

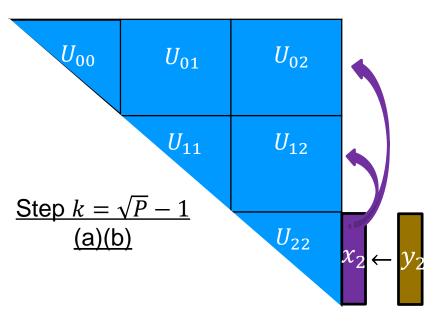


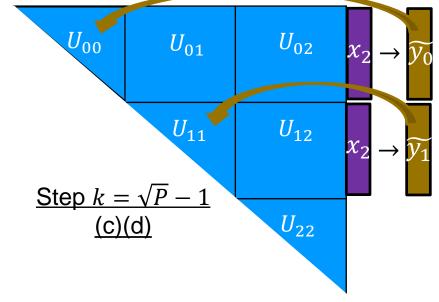
Back Substitution in 2D Blocks

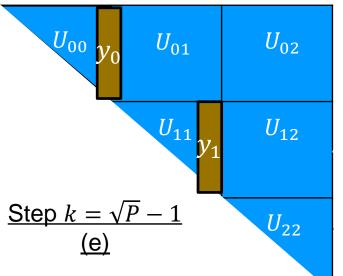
- For step $k = \sqrt{P} 1, \sqrt{P} 2, ..., 0$:
 - Processor P_{kk} performs local back substitution on matrix block U_{kk} and y_k (resulting in x_k);
 - Processor P_{kk} broadcasts x_k to all processors P_{jk} (j < k);
 - Processors P_{jk} (j < k) computes $\widetilde{y_j} \leftarrow U_{jk} \times x_k$;
 - Processors P_{jk} (j < k) sends \widetilde{y}_j to processors $P_{j,j}$;
 - e) Processors $P_{j,j}$ (j < k) updates $y_j \leftarrow y_j \widetilde{y}_j$.

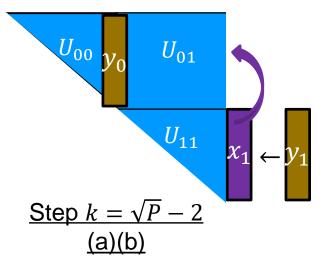
Next slide illustrates one step for P = 9 processors

Back Substitution in 2D Blocks



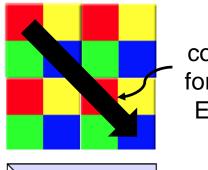




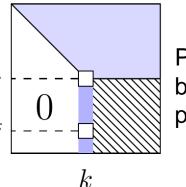


Solving Ax = b in Practice

- 2D Block-cyclic distribution is used (for better load balance);
- Partial pivoting is used during Gaussian Elimination (for numerical stability);
- LU factorization of matrix A (for reusing left side).



Order of computation for Gaussian Elimination



Partial pivoting by row permutation

These and other optimizations are implemented in the software package High-Performance Linpack (HPL) (http://www.netlib.org/benchmark/hpl/).

Fast Fourier Transform (FFT)

■ **Discrete Fourier Transform (DFT):** converts a discrete signal of N samples from time domain $(x_0, x_1, ..., x_{N-1})$ to frequency domain $(y_0, y_1, ..., y_{N-1})$, or vice versa:

$$y_j = \sum_{k=0}^{N-1} w_N^{jk} \cdot x_k$$
, $\forall j = 0,1,...N-1$

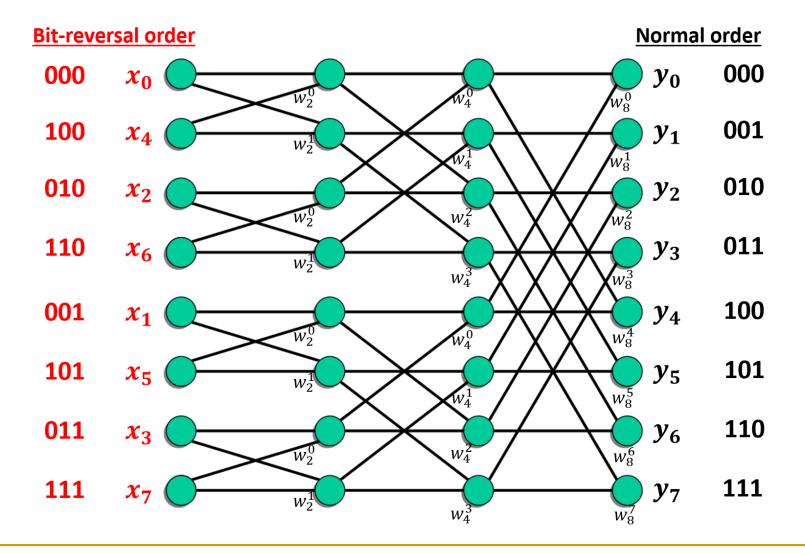
where $w_N = e^{-\frac{2\pi i}{N}}$ is the *N*-th root of unity, called the *twiddle factor*.

• Fast Fourier Transform (FFT): any algorithm that computes DFT in $O(n\log n)$ time instead of the naïve $O(n^2)$ complexity.

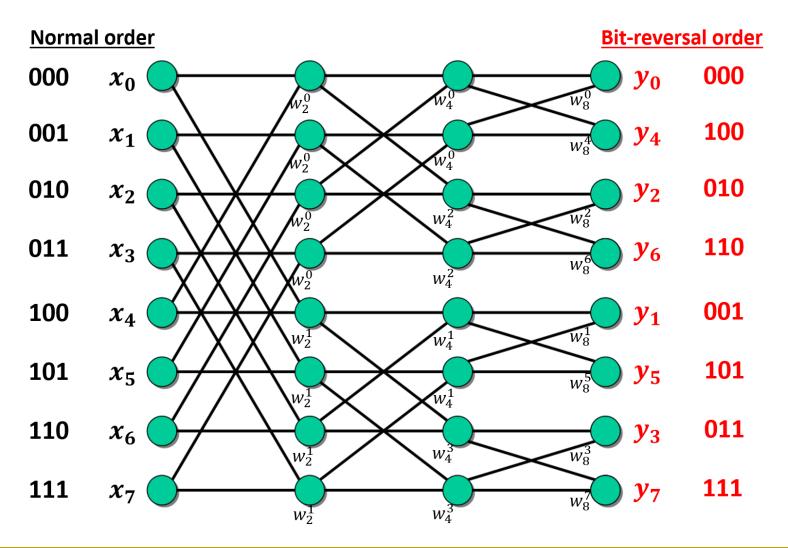
Fast Fourier Transform (FFT)

- FFT has numerous applications in science and engineering, and is considered to be one of the most important algorithms in the 20th century!
- We consider the classical radix-2 algorithm and study two parallel implementations:
 - Binary-Exchange Algorithm;
 - 2) Transpose Algorithm.
- For simplicity, we assume that both N and P (number of processors) are powers-of-2, i.e., $N = 2^r$, $P = 2^d$ and $d \le r$.

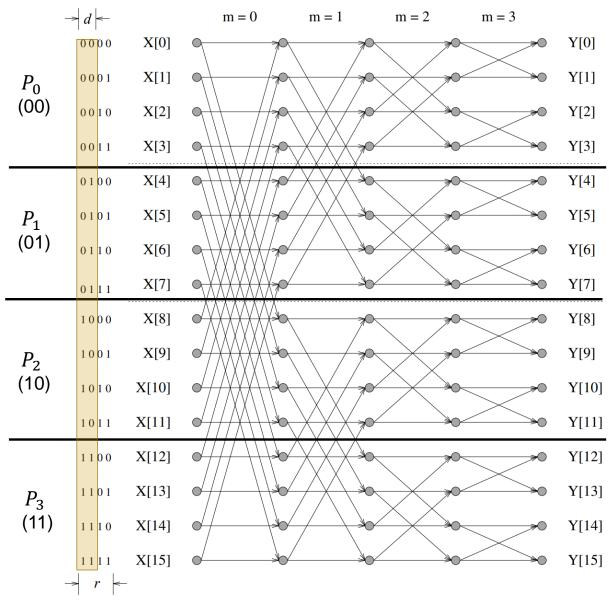
Radix-2 FFT (Decimation in Time)

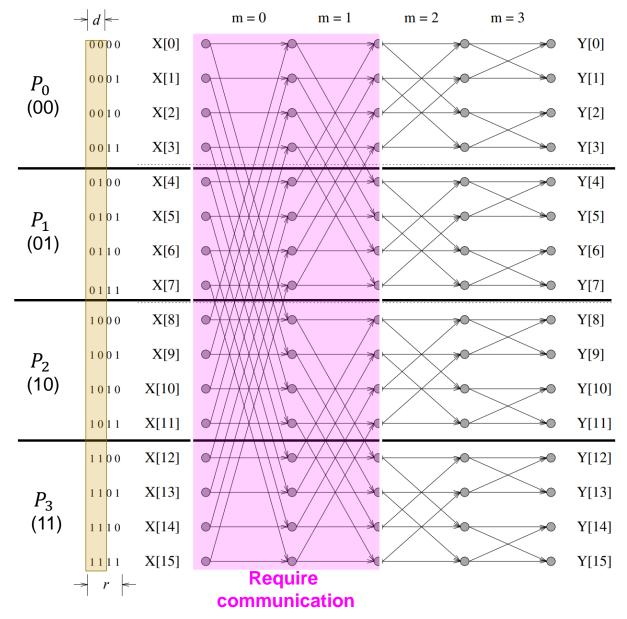


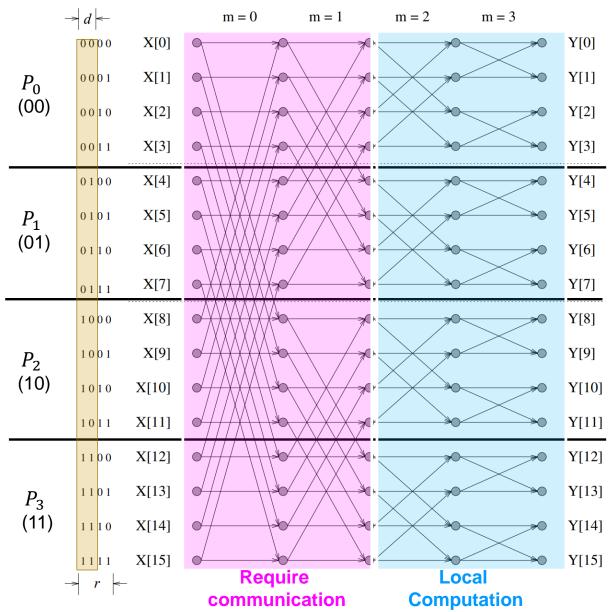
Radix-2 FFT (Decimation in Frequency)



- We assume 1D block data distribution for the DIF implementation of the radix-2 algorithm:
 - Each processor holds N/P consecutive input values in x and responsible for computing the corresponding output values in y.
- The algorithm works in two phases:
 - □ Phase 1 (w/ communication): step m = 0, 1, ..., d 1:
 - Each processor P_i exchanges N/P numbers with processor $P_{i'}$, where P_i and $P_{i'}$ only differ in the m-th bit (from left);
 - b) Each processor P_i does N/P local complex computations.
 - □ Phase 2 (local computation): step m = d, ..., r 1:
 - Each processor P_i does N/P local complex computations.

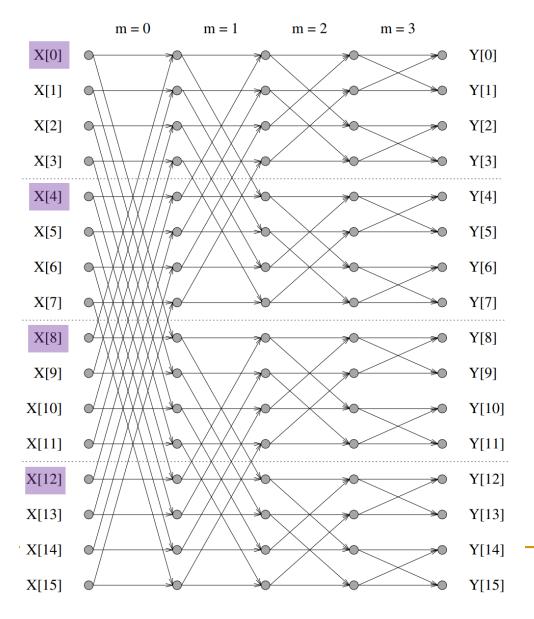


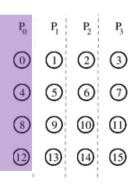


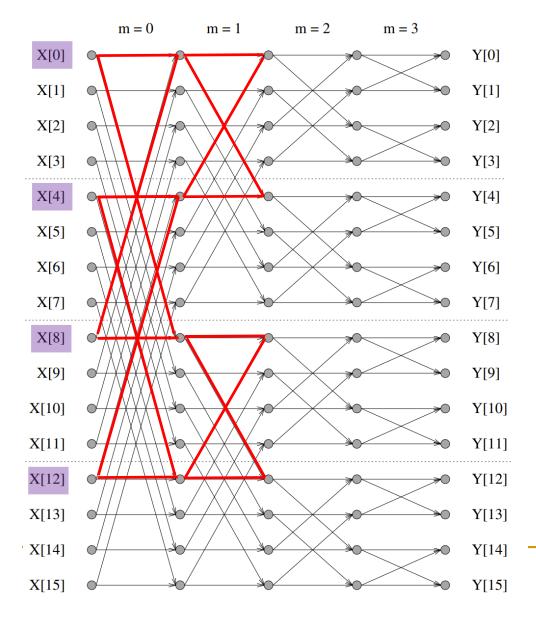


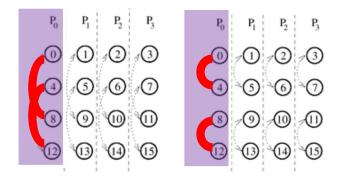
We assume:

- □ The input data (vector x) is arranged in a $\sqrt{N} \times \sqrt{N}$ matrix (in row-major order);
- The matrix is distributed in 1D block-cyclic (column) manner to the processors.
- The algorithm works in three steps:
 - Each processor does \sqrt{N}/P local FFTs for each column of the matrix it holds;
 - Perform a global matrix transpose through an all-toall personalized communication (MPI_Alltoall);
 - Each processor again does \sqrt{N}/P local FFTs for each column of the new matrix it holds.

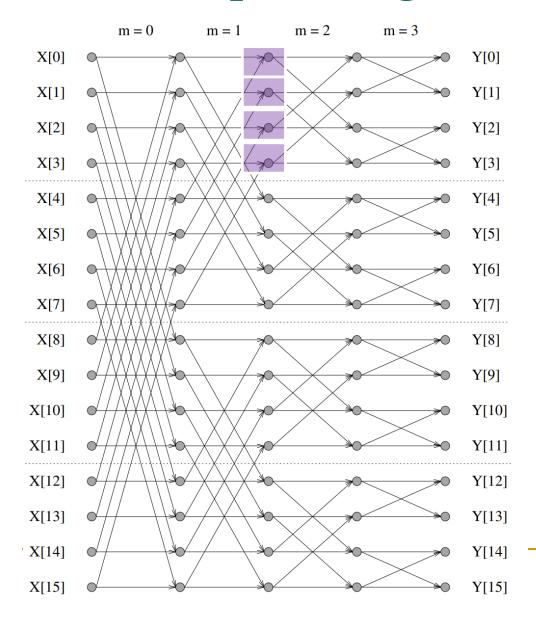


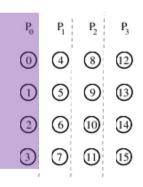






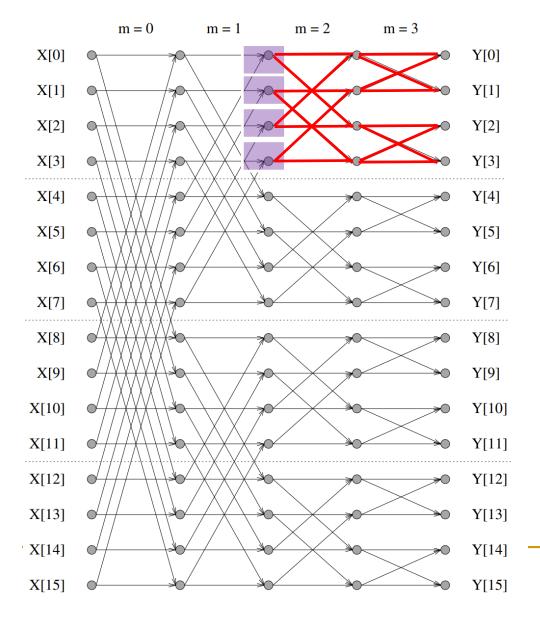
Step 1: Local FFT for each column (on same processor)

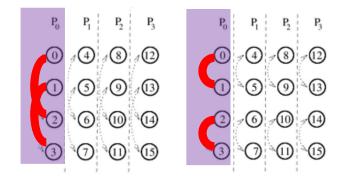




Step 1: Local FFT for each column (on same processor)

<u>Step 2</u>: Transpose matrix (all-to-all personalized comm)





Step 1: Local FFT for each column (on same processor)

Step 2: Transpose matrix (all-to-all personalized comm)

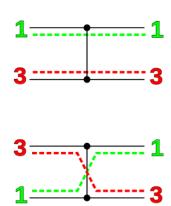
Step 3: Local FFT for each column (on same processor)

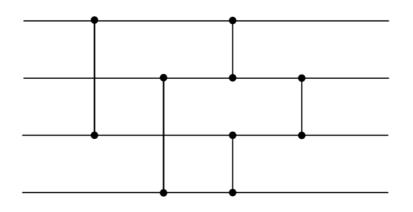
Sorting Algorithms

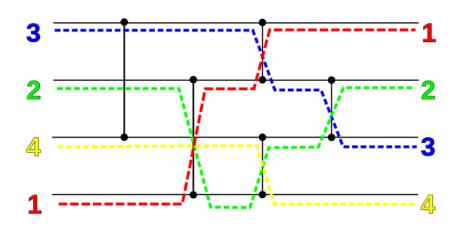
- A sorting algorithm arranges a list of n numbers or items in ascending or descending order --- a very important procedure in scientific computing.
- We have seen previously quicksort and mergesort in shared-memory architectures.
- In this class, we discuss three other sorting algorithms and their parallel implementations:
 - 1) Bitonic Sort;
 - 2) Brick Sort;
 - 3) Shear Sort.

Sorting Networks

 A sorting network sorts a fixed number of input values using a network of comparators connected by wires.



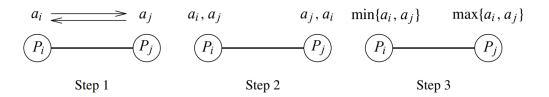




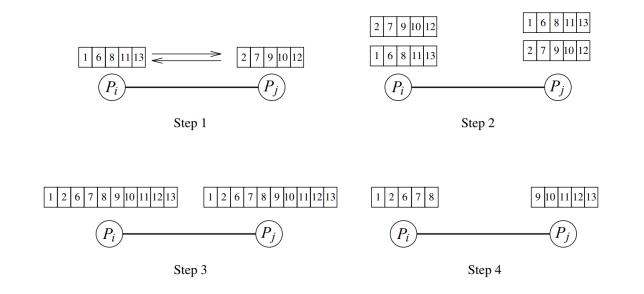
Why is this sorting network able to sort any four input values in ascending order?

Basic Comparison Operations in Parallel

Compare-Exchange (one element per process)



Compare-Split (multiple elements per process)

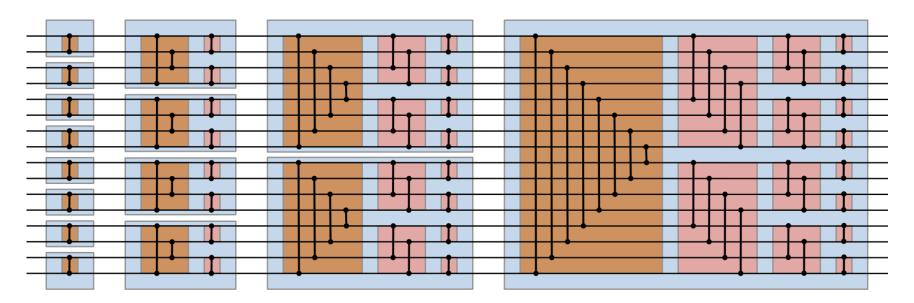


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(1) Bitonic Sort

Bitonic Sorting Network

 Constructed recursively using divide-and-conquer for any number of inputs n that is a power of 2 (read the reference materials on details of its construction).

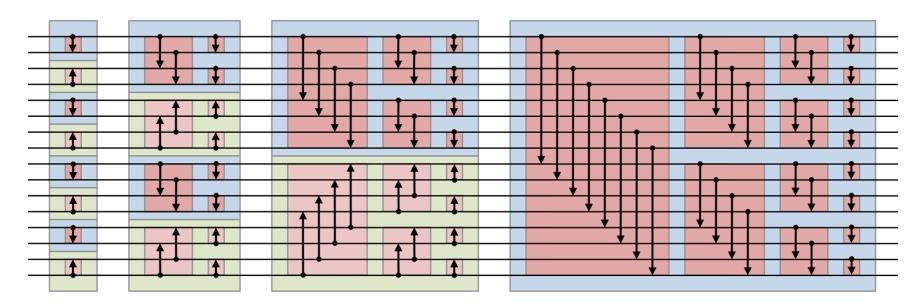


Standard form: the larger number of a comparator always goes down.

(1) Bitonic Sort

Bitonic Sorting Network

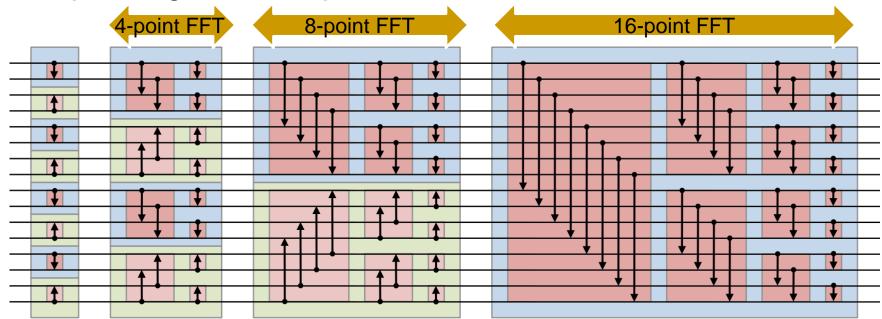
 Constructed recursively using divide-and-conquer for any number of inputs n that is a power of 2 (read the reference materials on details of its construction).



Non-standard form: the larger number of a comparator goes to point of arrow.

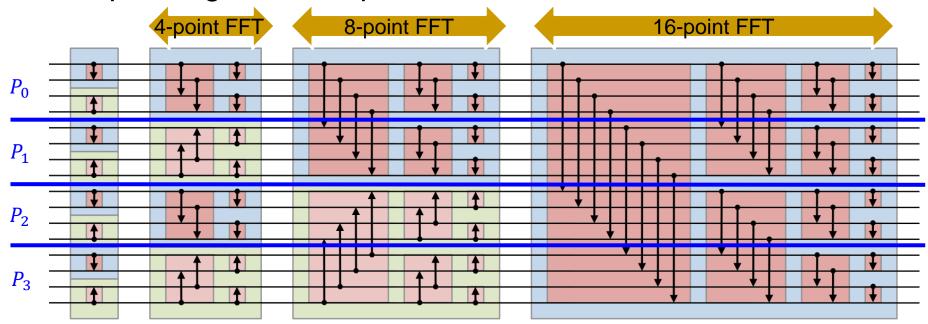
(1) Bitonic Sort in Parallel

Observe that a bitonic sorting network (non-standard) has multiple stages of computations with FFT-like structures.



(1) Bitonic Sort in Parallel

Observe that a bitonic sorting network (non-standard) has multiple stages of computations with FFT-like structures.



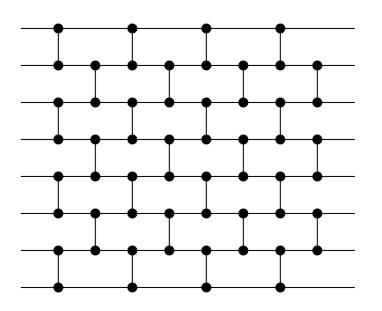
Parallel algorithms similar to Binary-Exchange for FFT can be implemented for Bitonic Sort.

 1D Block distribution of data, and comparisons across two processors can be done via compare-exchange operations.

(2) Brick Sort

Odd-Even Sorting Network

 Constructed by n levels of comparators connected in a "brick-like" pattern, with alternating odd and even sorting steps (it's also called Odd-Even Transposition Sort).



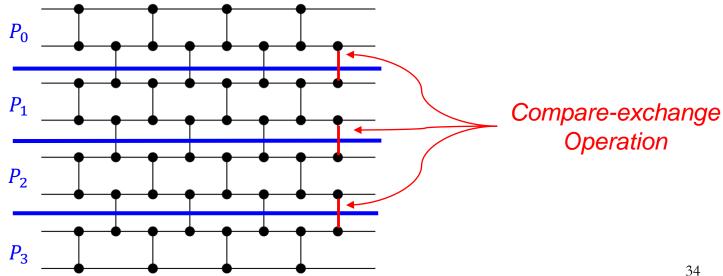
```
Unsorted array: 2, 1, 4, 9, 5, 3, 6, 10
Step 1(odd): 2
                                                                10
Step 2(even): 1
                                                                10
Step 3(odd): 1
                                                                10
Step 4(even): 1
                                                                10
Step 5(odd): 1
                                                                10
Step 6(even): 1
                                                                10
Step 7(odd): 1
                                                                10
Step 8(even): 1
                                                                10
Sorted array: 1, 2, 3, 4, 5, 6, 9, 10
```

Figure source: https://www.geeksforgeeks.org/odd-even-transposition-sort-brick-sort-using-pthreads/

(2) Brick Sort in Parallel

Parallel Implementation 1:

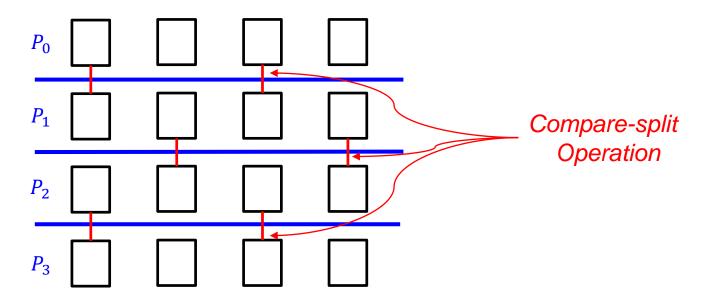
- 1D Block distribution;
- Follow exactly the same pattern of an n-step oddeven sorting network;
- Only boundary elements of a processor need communication (done via compare-exchange).



(2) Brick Sort in Parallel

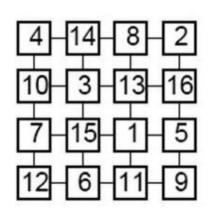
Parallel Implementation 2:

- 1D Block distribution;
- □ Follow the same pattern again, but treating blocks of elements together \rightarrow *P* steps instead of *n*;
- Entire block of elements of a processor need to be communicated and compared (done via compare-split).



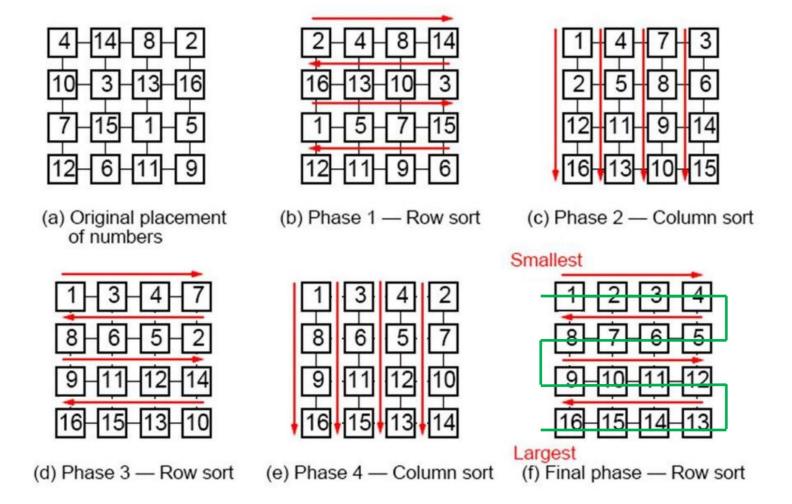
(3) Shear Sort

- Algorithm works in 2D:
 - □ A list of n numbers initially arranged in a $\sqrt{n} \times \sqrt{n}$ matrix;
 - □ For step $k = 1, 2, ..., \lg n + 1$:
 - If k is odd, sort odd rows in ascending order and sort even rows in descending order;
 - If k is even, sort all columns in ascending order.
 - The final list will be sorted in 2D snake-like order.



Original placement of numbers

Shear Sort



Shear Sort in Parallel

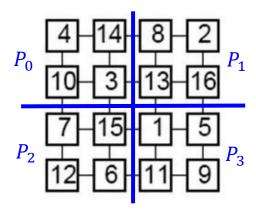
1D Block Implementation:

- Row-sorting can be done *locally* on each processor;
- Column-sorting can be done across processors with *Brick Sort* for each column.

P_0 4 14 8 2 P_1 10 3 13 16 P_2 7 15 1 5 P_3 12 6 11 9

2D Block Implementation:

 Both row-sorting and columnsorting can be done with *Brick Sort* across multiple processor groups (rows and columns).



References

- A. Grama, A. Gupta, G. Karypis, V. Kumar. Introduction to Parallel Computing (2nd Edition). Addison-Wesley Professional.
- H. Casanova, A. Legrand, Y. Robert. Parallel Algorithms. Chapman & Hall/CRC.
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 - https://computing.llnl.gov/tutorials/parallel_comp/