



Korea Institute of  
Science and Technology Information

# **My Internship Experience: Development and Research Projects**

[ 2024.04.22 – 2024.07.19 ]

Presenter: Chanyoung Ahn

## Overview of projects

- 1 Lecture Materials: developed three lecture materials on MPI
- 2 Git Website: documentation for PaScaL\_TDMA library
- 3 Benchmark: performance benchmark of large sparse matrix using three libraries

# 1 Lecture Materials

Developed three lecture materials on MPI

\* MPI 병렬컴퓨팅교육 운영 2회 (강지훈, 권오경) : 상반기 1회 완료

파이썬을 이용한 MPI 병렬프로그래밍

한국과학기술정보연구원(KISTI) | 강지훈

Part 1 (~29 pp.)

Contents

- MPI 병렬처리 이해
  - MPI 소개 및 프로그래밍 모델
  - MPI 프로세스와 커뮤니케이터
- mpi4py 패키지 소개
  - mpi4py 패키지
  - 집합점 통신 (MPI Collective)
- 파이썬 MPI 프로그래밍
  - 루프 병렬화
  - 영역 분할 (Domain Decomposition)

1 | 피란 무엇인가

파이썬을 이용한 MPI 병렬프로그래밍

Wrote up 6+ jupyter notebook materials for MPI Workshop

Python으로 실습하는 병렬컴퓨팅

MPI Parallelization with Python

Supercomputing Center, KISTI | Oh-Kyoung Kwon

MPI Parallelization

Data decomposition (domain decomposition)

- Each core has its own data
  - Each core communicates necessary data
  - Calculation is automatically divided

Python으로 실습하는 병렬컴퓨팅

8 April 2024

Oh-Kyoung Kwon  
Supercomputing Center, KISTI

2 | It fits for MPI parallelization  
Programmer should do everything: data decomposition

Python으로 실습하는 병렬컴퓨팅

Wrote up a jupyter notebook material for MPI Workshop

NN w/ MPI

NN w/o MPI

3 | Segmentation model: [8, 9, 1, 2, 6]

MNIST segmentation NN w/ MPI

Converted NN w/ MPI example to Python from C code

# Overview of projects

- 1** Lecture Materials: developed three lecture materials on MPI
- 2** Git Website: documentation for PaScaL\_TDMA library
- 3** Benchmark: performance benchmark of large sparse matrix using three libraries

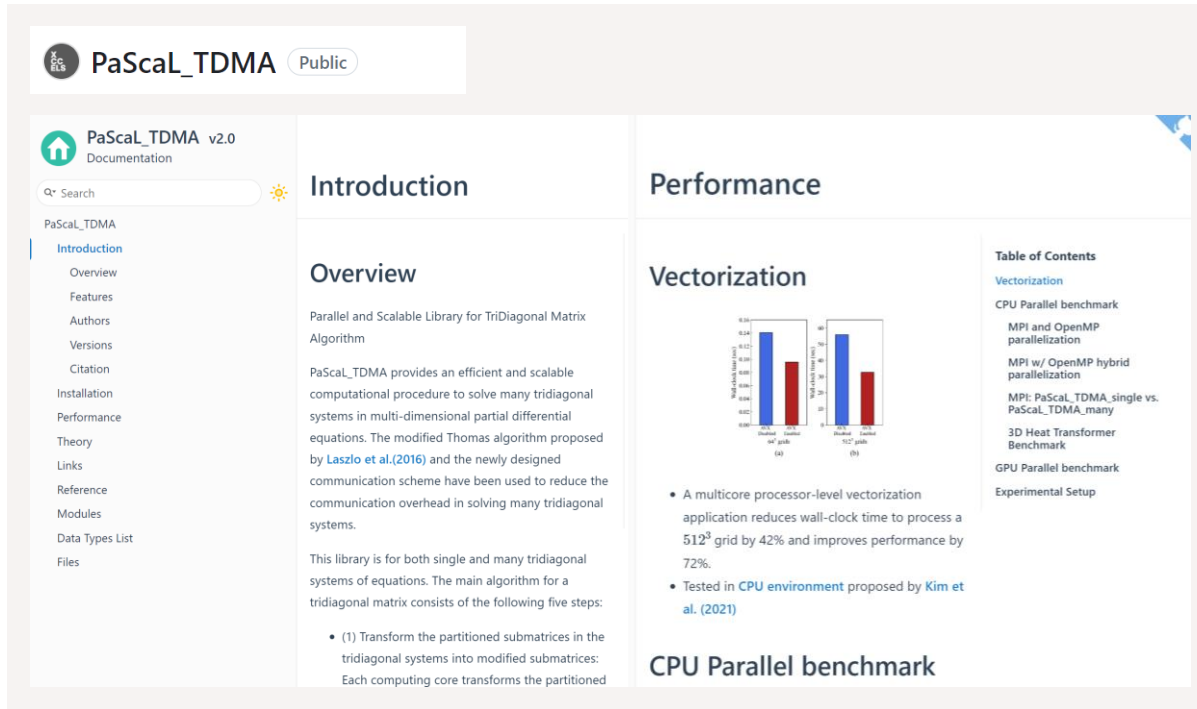
## 엑사스케일 선형 해석자 수치 라이브러리 개발 / HPC Library for Exascale Linear Solvers

1단계 - 2차 연도(2024)														
추진내용	추진 일정												책임기관 (소속기관)	
	1	2	3	4	5	6	7	8	9	10	11	12		
<b>2</b> 수치 라이브러리 통합 저장소 구축														강지훈 (KISTI)
<b>3</b> 국가슈퍼컴퓨터 5호기 대상 최적 병렬화 및 성능 평가														

## 2 Git Website

[https://xcclls.github.io/PaScaL\\_TDMA/](https://xcclls.github.io/PaScaL_TDMA/)

### Documentation for PaScaL\_TDMA library



Second year targets of  
엑사스케일 선형 해석자 수치 라이브러리 개발

### Why

2 수치 라이브러리 통합 저장소 구축

강지훈  
(KISTI)

### Role

- Develop GitHub repository & PaScaL\_TDMA Library documentation site

### Result


- Created xcclls/PaScaL\_TDMA repository
- Wrote up 5+ documentation web pages

### 3 Benchmark

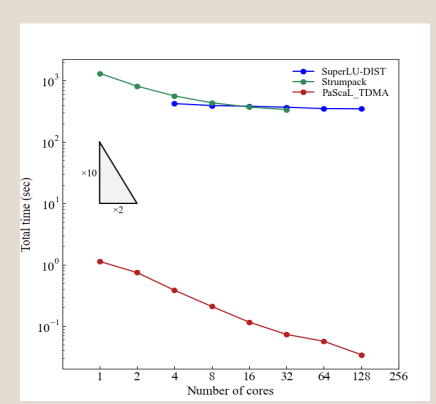
Performance benchmark of large sparse matrix using three libraries

## Parallel performance benchmark of large sparse matrix

using SuperLU-DIST, Strumpack and PaScaL\_TDMA



\*1안찬영, 1권오경 #1강지훈



	Total time (sec.)
<b>SuperLU-DIST</b>	348.092226
<b>Strumpack</b>	337.743508
<b>PaScaL_TDMA</b>	<b>0.034142</b>

Second year targets of  
엑사스케일 선형 해석자 수치 라이브러리 개발

<b>3</b>	국가슈퍼컴퓨터 5호기 대상 최적 병렬화 및 성능 평가	강지훈 (KISTI)
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**Why**

**Role**

**Result**

- Parallel benchmark of large sparse matrix using three libraries
- PaScaL\_TDMA is best suited to compute large tridiagonal matrices.
- (ongoing) Conference poster for targeting CDE in 2024

### 3 Benchmark

## Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA

\*1안찬영, 1권오경 #1강지훈

- Problem** We need to find suitable sparse matrix solvers that enhance efficiency.
- Challenge** Sparse matrix solver encounter challenges in computational efficiency, becoming a computational bottleneck.
- Solution** We provide a comparative performance benchmark of three parallel solvers.

# 3 Benchmark

Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA



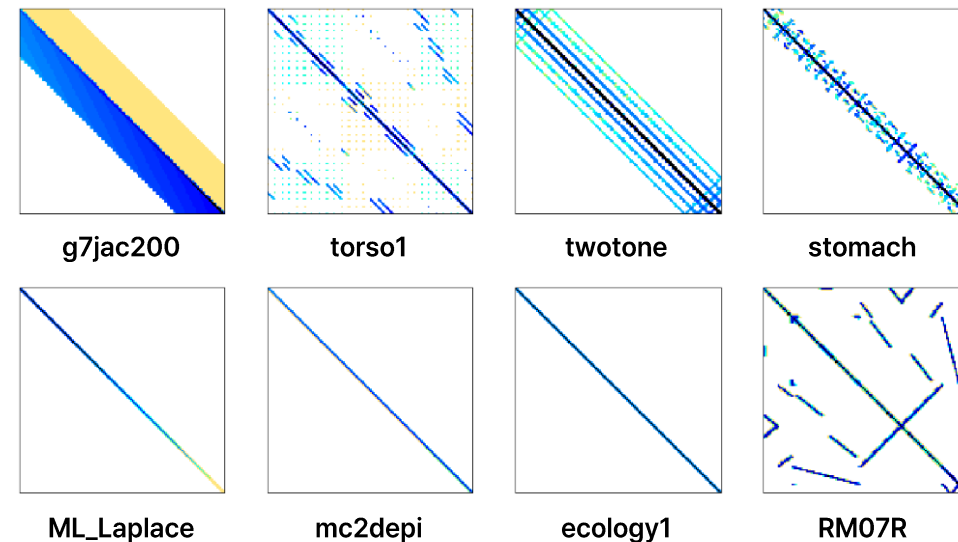
## Experimental setup

Benchmark total and factorization times with nine large sparse matrices

Comparative performance benchmark of three parallel solvers; SuperLU-DIST, Strumpack, and PaScaL\_TDMA

\* All the computations were executed on the Nurion manycore cluster at KISTI.

	r/c	nnz	ratio
<b>g7jac200</b>	59,310	717,620	2.38E-04
<b>torso1</b>	116,158	8,516,500	6.31E-04
<b>twotone</b>	120,750	1,206,265	8.40E-05
<b>stomach</b>	213,360	3,021,648	6.64E-05
<b>ML_Laplace</b>	377,002	27,582,698	1.95E-04
<b>RM07R</b>	381,689	37,464,962	2.57E-04
<b>mc2depi</b>	525,825	2,100,225	7.60E-06
<b>ecology1</b>	1,000,000	4,996,000	5.00E-06
<b>256<sup>3</sup> TDM</b>	16,777,216	50,331,646	1.79E-07





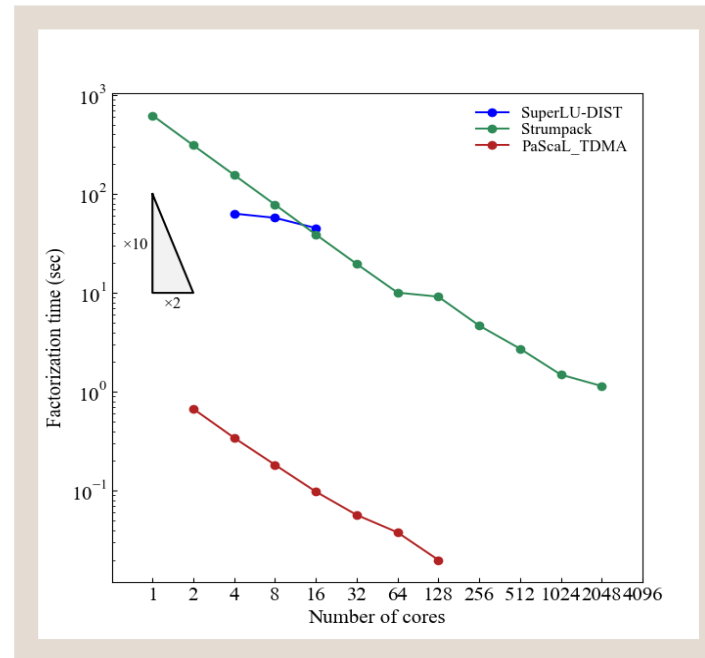
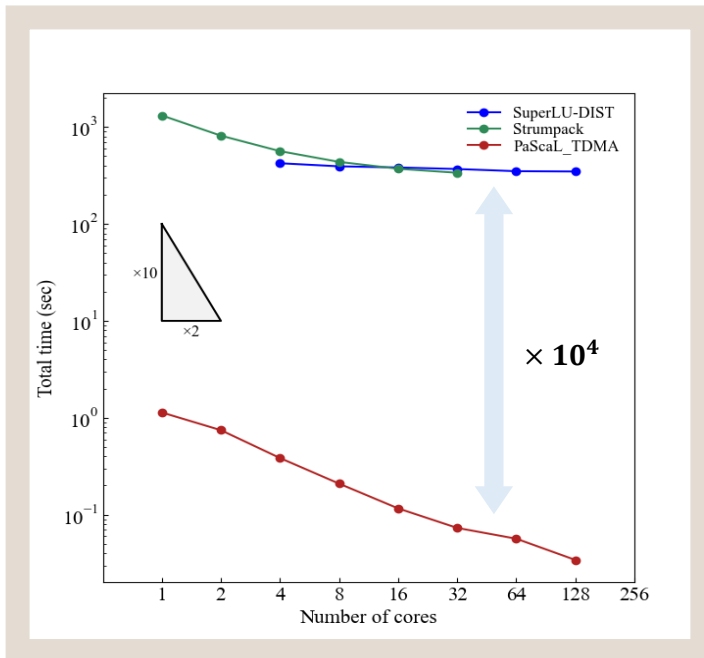
# 3 Benchmark

## Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA



### Overall performance

#### Benchmark of Three Parallel Solvers on a 256<sup>3</sup> Tri-diagonal Matrix



	Total time (sec)	speed up (2 <sup>5</sup> )
<b>SuperLU-DIST</b>	348.092226	1.2
<b>STRUMPACK</b>	337.743508	3.9
<b>PaScaL_TDMA</b>	<b>0.034142</b>	<b>15.5</b>

\* A 256<sup>3</sup> tri-diagonal matrix was tested in Nurion normal nodes, with a range of 1–4096 cores.

- PaScaL\_TDMA computes the 256<sup>3</sup> tri-diagonal matrix in the shortest total time of 0.034142 seconds.
- The solver also shows the best improvement in computation speed due to parallelization.

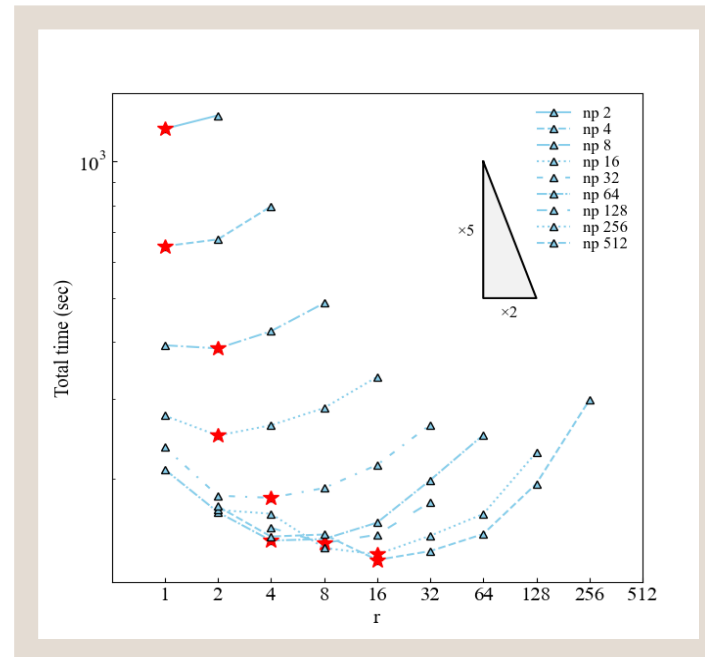
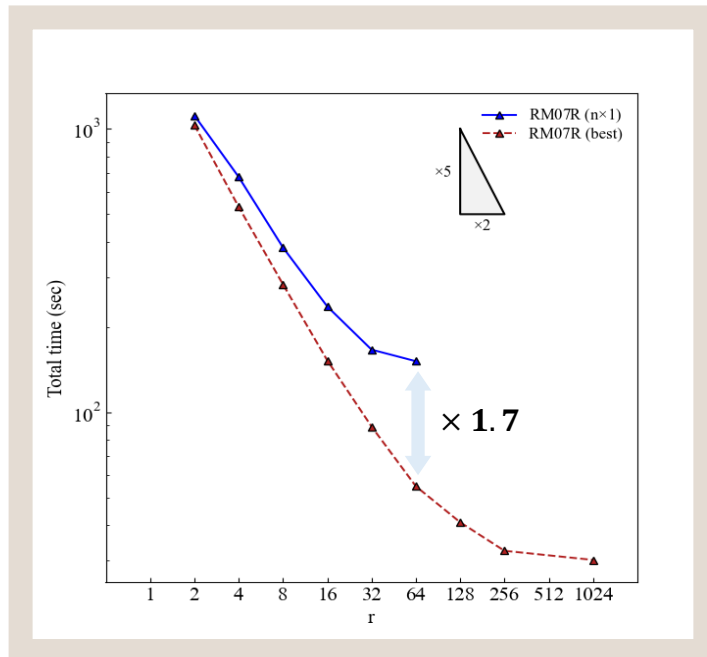
# 3 Benchmark

## Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA



### Baseline 1: SuperLU-DIST (1/3)

#### Optimization of -r -c Parameters for RM07R



np	r	c
2	1	2
4	1	4
8	2	4
16	2	8
32	4	8
64	4	16
128	8	16
256	16	16
512	16	32

- Two parameters, -r and -c, affect the performance of total and factorization times.
- As the number of cores increases, the optimized value of parameter -c tends to rise from 1 to 2, 4 ... 16.

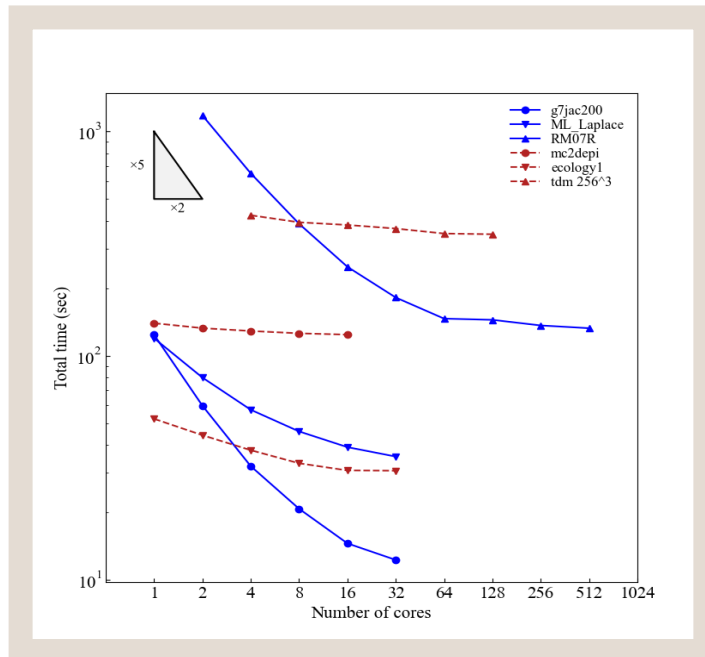
# 3 Benchmark

## Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA

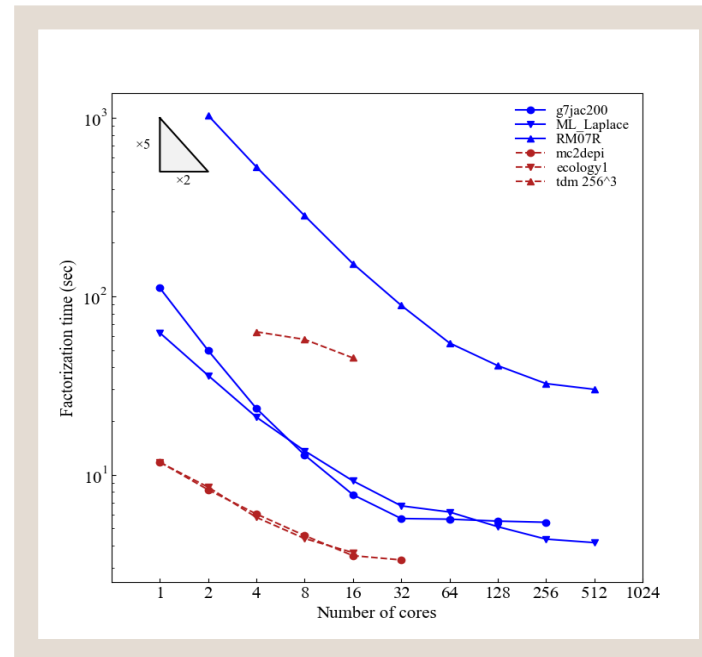


### ■ Baseline 1: SuperLU-DIST (2/3)

#### Benchmark of SuperLU-DIST on Six Large Sparse Matrices



(a) Total Time



(b) Factorization Time

	r/c	ratio	speed up (2^5)
<b>g7jac200</b>	59,310	<b>2.38E-04</b>	<b>10.1</b>
<b>RM07R</b>	381,689	<b>2.57E-04</b>	<b>8.1</b>
<b>ML_Laplace</b>	377,002	<b>1.95E-04</b>	<b>3.4</b>
<b>mc2depi</b>	525,825	<b>7.60E-06</b>	<b>1.1</b>
<b>ecology1</b>	1,000,000	<b>5.00E-06</b>	<b>1.7</b>
<b>256<sup>3</sup> TDM</b>	16,777,216	<b>1.79E-07</b>	<b>1.2</b>

\* These six matrices have been tested in Nurion normal nodes with a range of 1-1024 cores.

$$\text{nnz\_ratio} = \frac{\text{The number of nonzero}}{\text{The number of all elements}}$$

- The efficiency of parallelization tends to decrease as the nonzero ratio decreases in SuperLU-DIST.

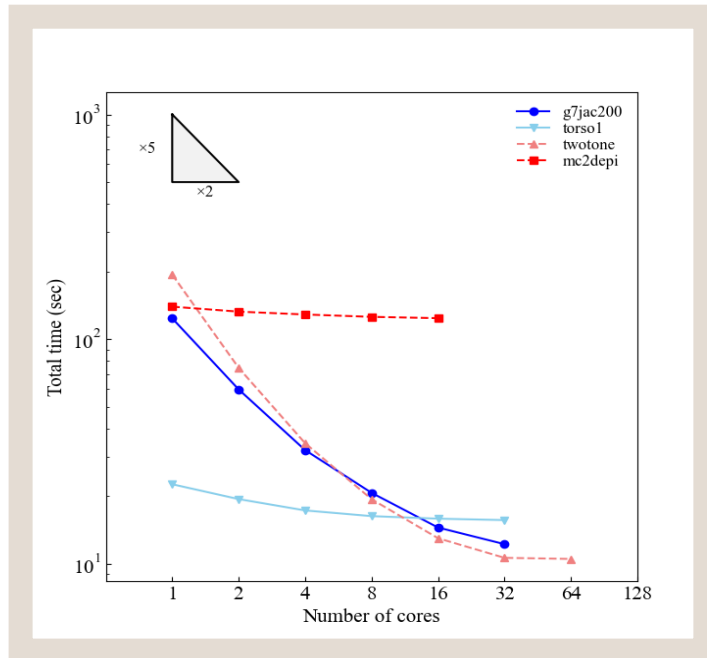
# 3 Benchmark

Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA

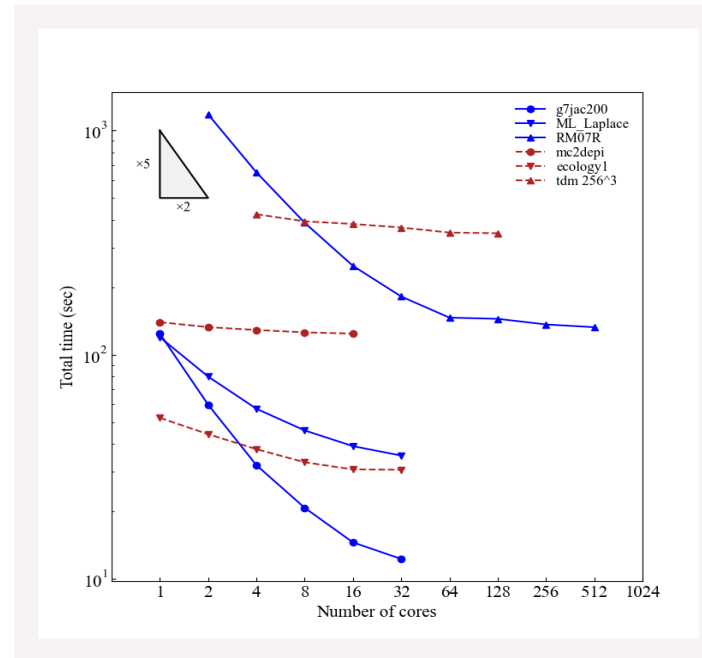


## ■ Baseline 1: SuperLU-DIST (3/3)

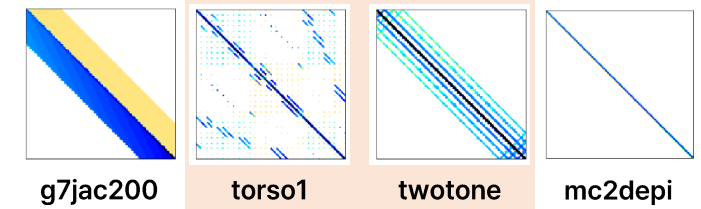
### Benchmark of SuperLU-DIST on Two Large Sparse Matrices



(a) Total Time in two outliers



(b) Total Time in six matrices



	r/c	ratio	speed up (2^5)
<b>torso1</b>	116,158	<b>6.31E-04</b>	<b>1.44</b>
<b>g7jac200</b>	59,310	<b>2.38E-04</b>	<b>10.1</b>
<b>RM07R</b>	381,689	<b>2.57E-04</b>	<b>8.1</b>
<b>ML_Laplace</b>	377,002	<b>1.95E-04</b>	<b>3.4</b>
<b>twotone</b>	120,750	<b>8.40E-05</b>	<b>18.2</b>
<b>stomach</b>	213,360	<b>6.64E-05</b>	<b>3.0</b>
<b>mc2depi</b>	525,825	<b>7.60E-06</b>	<b>1.1</b>
<b>ecology1</b>	1,000,000	<b>5.00E-06</b>	<b>1.7</b>
<b>256<sup>3</sup> TDM</b>	16,777,216	<b>1.79E-07</b>	<b>1.2</b>

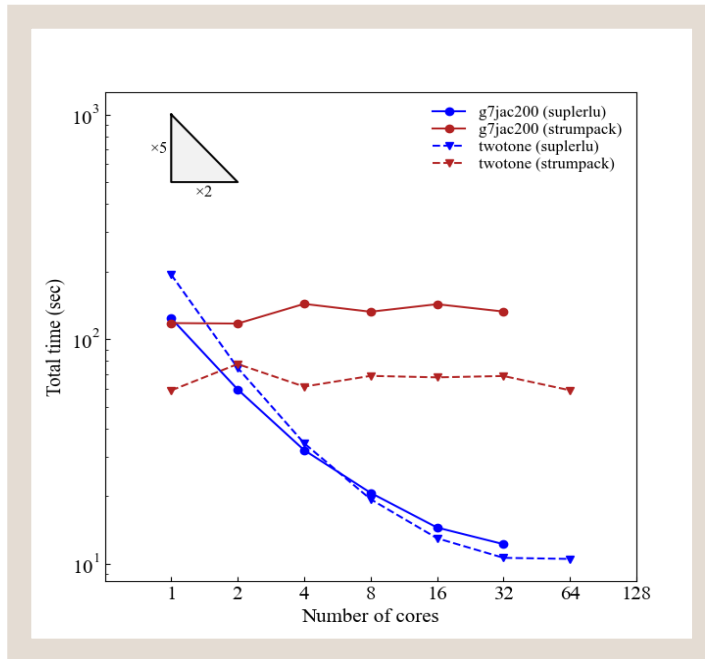
- The parallelization of two outlier matrices tends to increase as the nonzero ratio decreases in SuperLU-DIST.

# 3 Benchmark

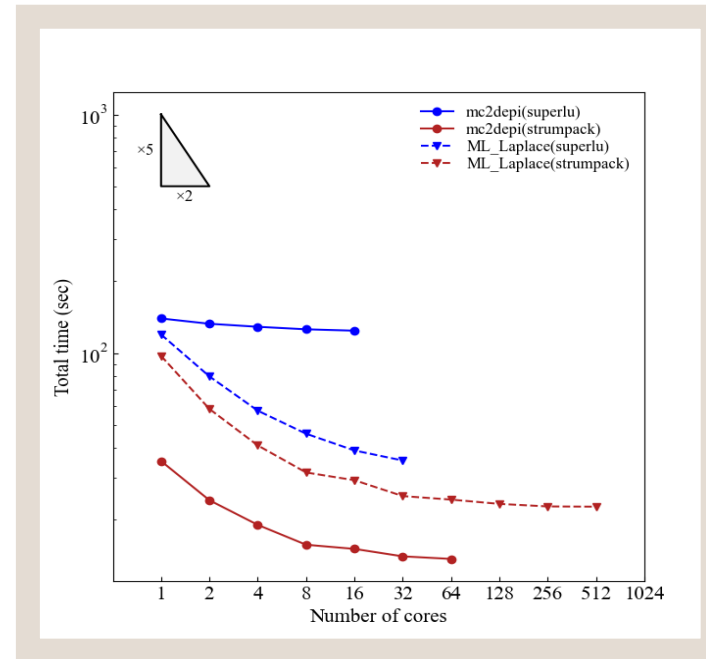
Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA



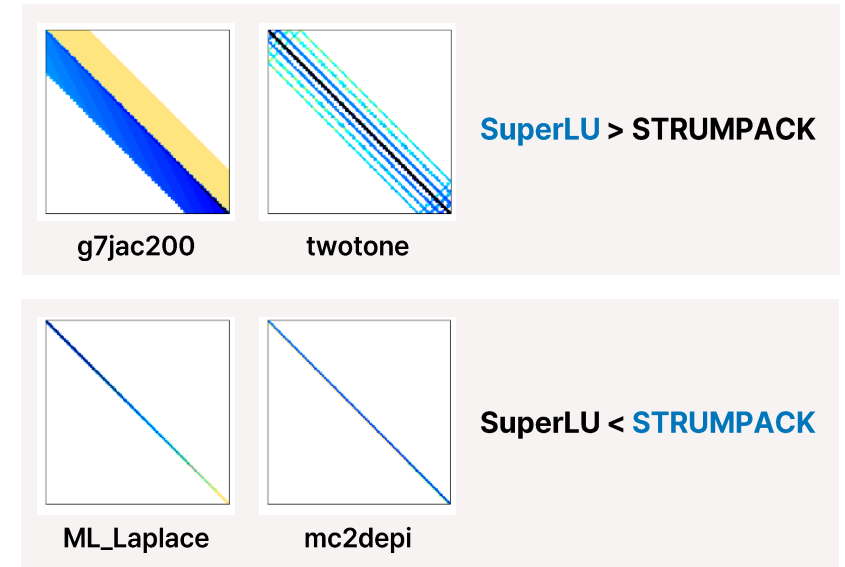
## Baseline 1 & 2: SuperLU-DIST vs. STRUMPACK



(a) SuperLU > STRUMPACK



(b) SuperLU < STRUMPACK



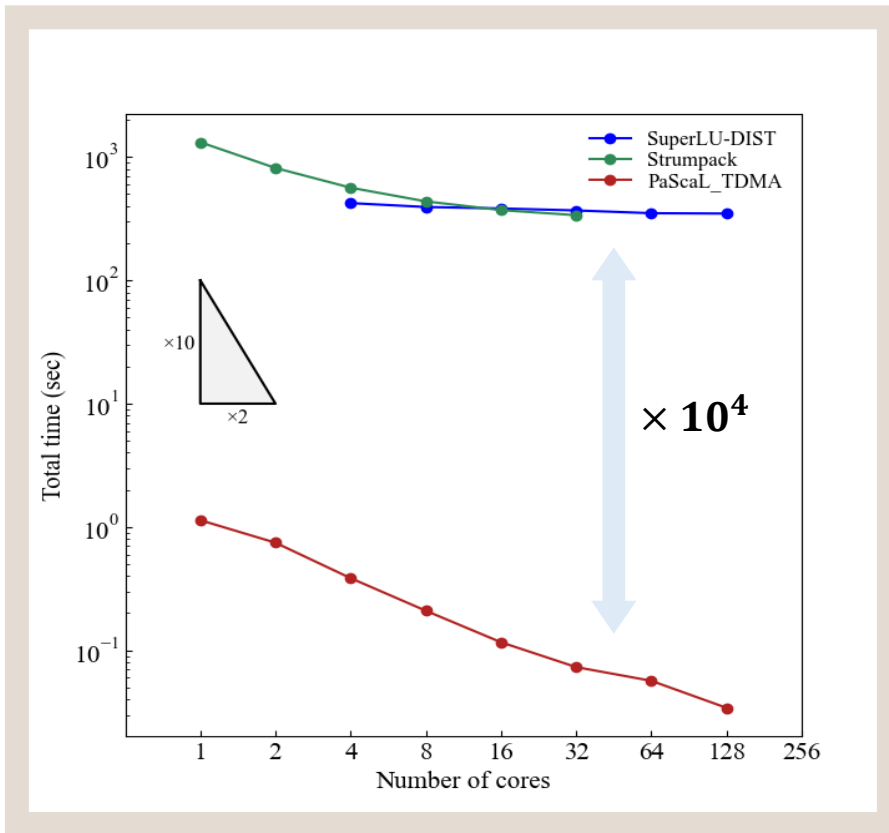
- As the diagonal components become denser, SuperLU may tend to outperform STRUMPACK.

# 3 Benchmark

Parallel Performance Benchmark of Large Sparse Matrix using SuperLU-DIST, Strumpack, and PaScaL\_TDMA



## Conclusion & Discussion



- PaScaL\_TDMA performs tri-diagonal matrix computations 10,000 times faster than baseline solvers.
- PaScaL\_TDMA remains stable with increased cores and maintains high computation speed. (SuperLU-DIST: over 16 cores, STRUMPACK: 64 cores)
- The results of these performance benchmarks highlight the necessity for optimized solvers based on matrix structure.

## 3 Benchmark

Parallel Performance Benchmark of Large Sparse Matrix  
using SuperLU-DIST, Strumpack, and PaScaL\_TDMA



### ■ Future Work

- **It is essential to conduct a comprehensive performance comparison of the three solvers, including a time analysis with MPI functionality.**
- **Matrix structure impacts parallelization performance; analyzing this can provide valuable insights for developing more suitable libraries.**
  - Diagonal matrix form reduces parallelization impact in SuperLU-DIST.
  - Increased density of diagonal components decrease parallelization impact in STRUMPACK.

# Conclusion

1 Lecture Materials: developed three lecture materials on MPI



\* MPI 병렬컴퓨팅교육 운영 2회 (강지훈, 권오경) :  
상반기 1회 완료

2 Git Website: documentation for PaScaL\_TDMA library

3 Benchmark: performance benchmark of large sparse matrix using three libraries

엑사스케일 선형 해석자 수치 라이브러리 개발 / HPC Library for Exascale Linear Solvers

1단계 - 2차 연도(2024)														
추진내용	추진 일정												책임기관 (소속기관)	
	1	2	3	4	5	6	7	8	9	10	11	12		
수치 라이브러리 통합 저장소 구축														강지훈 (KISTI)
국가슈퍼컴퓨터 5호기 대상 최적 병렬화 및 성능 평가														

평가항목	가중치(%)	연차	연차별 목표(조건/환경)
(정량) 유사 수치 라이브러리 대비 성능	20	(1단계) 1, 2차 연도	삼중대각행렬 해석자에 대해 5호기(NURION KNL)에서 1024노드 (65,536코어, 3PF)까지 경쟁 라이브러리(SuperLU, Strumpack, ScaLAPACK)와의 성능 비교 및 70%의 성능 달성

\* 2023 연차보고서 / 1-2. 평가 주안점의 차년도 목표



2 PaScaL\_TDMA repo / website



3 Benchmark two baselines on Nurion knl, from 1 node to 32 nodes. (50%)





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**Thank you for listening**

Presenter: Chanyoung Ahn

# Appendix

## Diagonal Correlation Factor (Fail)

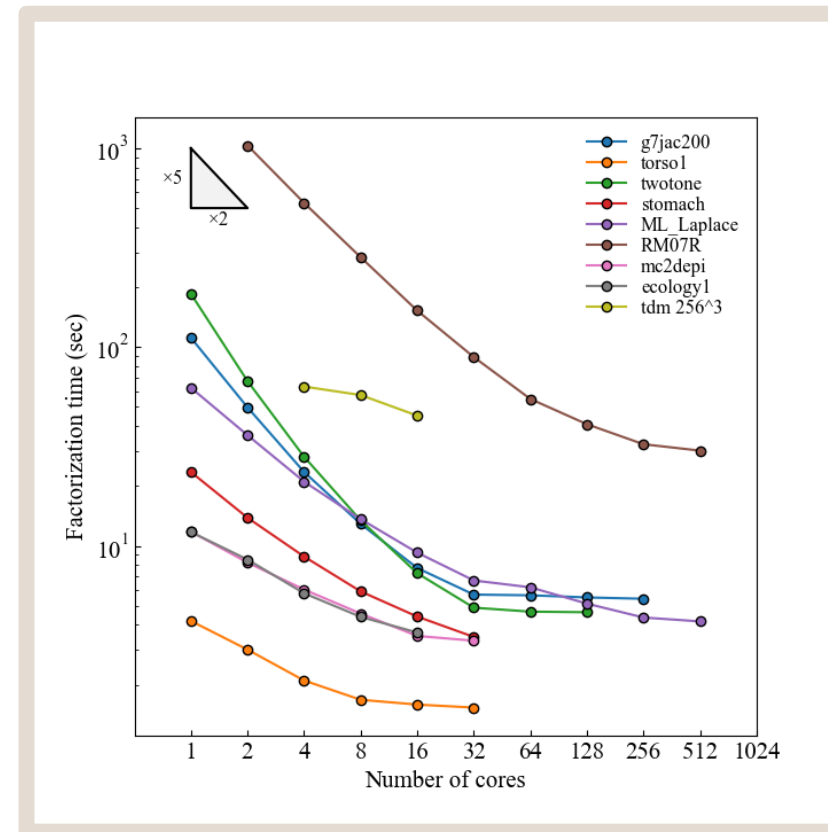
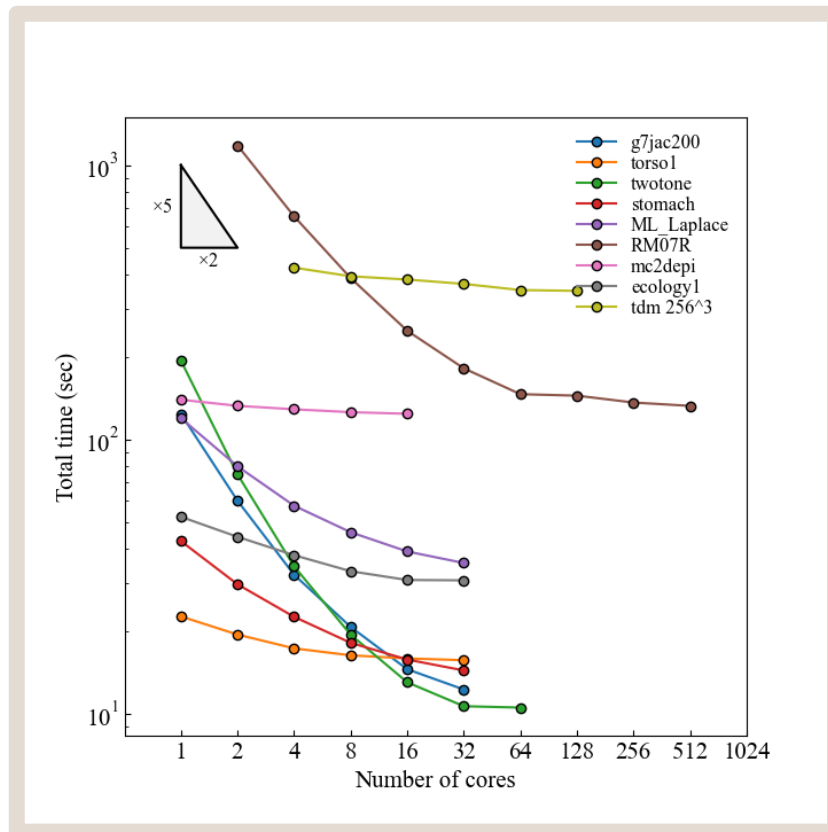
	r/c	nnz	nonzero ratio	Diagonal correlation factor	superlu speedup	strumpack speedup	PTDMA
g7jac200	59,310	837,936	2.38E-04	0.933829	10.1	0.9	
torso1	116,158	8,516,500	6.31E-04	0.118738	1.4	3.8	
twotone	120,750	1,224,224	8.40E-05	0.985866	18.2	0.9	
stomach	213,360	3,021,648	6.64E-05	0.999528	3.0	2.6	
ML_Laplace	377,002	27,689,972	1.95E-04	0.999949	3.4	3.9	
RM07R	381,689	37,464,962	2.57E-04	0.805946	8.1	error	
mc2depi	525,825	2,100,225	7.60E-06	0.999998	1.1	2.5	
ecology1	1,000,000	4,996,000	5.00E-06	0.999999	1.7	3.0	
<b>256<sup>3</sup> TDM</b>	<b>16,777,216</b>	<b>50331646</b>	<b>1.79E-07</b>	<b>0.999999</b>	<b>1.2</b>	<b>3.9</b>	<b>15.5</b>

Pearson Correlation Factor

$$r_{XY} = \frac{\sum_i^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_i^n (X_i - \bar{X})^2} \sqrt{\sum_i^n (Y_i - \bar{Y})^2}}$$

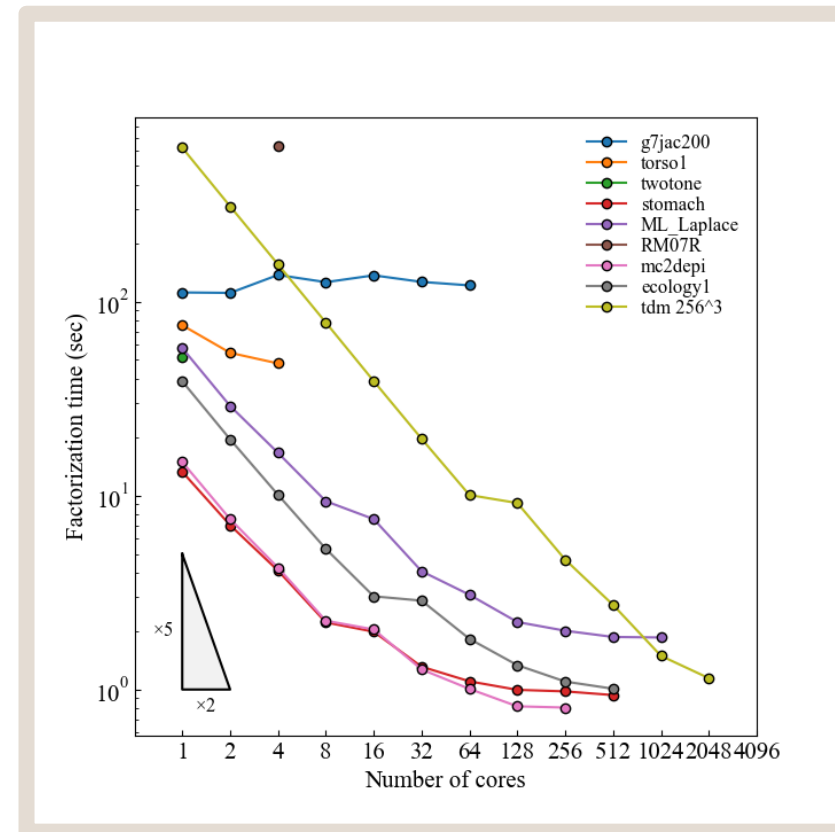
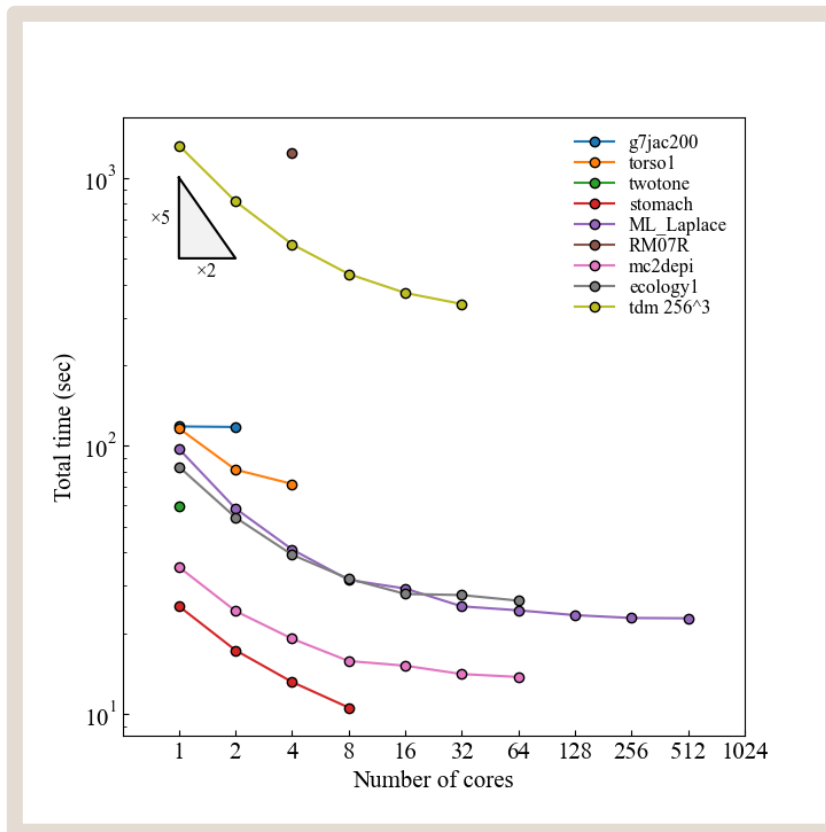
# Appendix

## All Matrices Benchmark in SuperLU-DIST



# Appendix

## All Matrices Benchmark in STRUMPACK



## Appendix

### Useful links

[https://xcels.github.io/PaScaL\\_TDMA/](https://xcels.github.io/PaScaL_TDMA/)

Public repository / PaScaL\_TDMA Library

[https://github.com/cold-young/2024\\_KISTI\\_Intern/](https://github.com/cold-young/2024_KISTI_Intern/)

Public repository / MPI with segmentation materials, documentation for Supercomputing .....

<https://github.com/cold-young/2024-CDE-KISTI/>

Private repository / All benchmark files: three benchmark file, raw data, and visualization utils

[Google Drive](#)

Google Drive / All benchmark data; Total and factorization times of three solvers