

Development of a Scalable Parallel Eigensolver for Large-scale Simulations and Data Analysis

Tetsuya Sakurai

Director, Center for Artificial Intelligence Research (C-AIR)

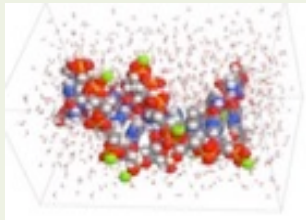
University of Tsukuba / JST CREST

Eigenvalue Problems in Simulation and Data Analysis

Electronic State Analysis

Modeling

Eigenvalue Problem



Basis Function Expansion

$$\psi_i(\mathbf{r}) = \sum_j \phi_j(\mathbf{r})$$

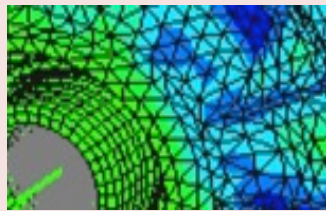


$$H\mathbf{x} = \lambda S\mathbf{x}$$

Interaction of Electrons

Nano Materials

Structural Analysis



Vibration Analysis

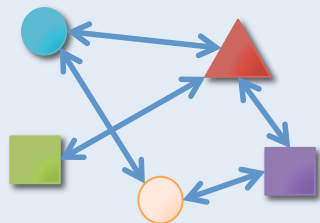


$$K\mathbf{x} = \lambda M\mathbf{x}$$

Interaction of Finite Elements

Car body design, etc.

Data Analysis



Data



Similarity of data elements

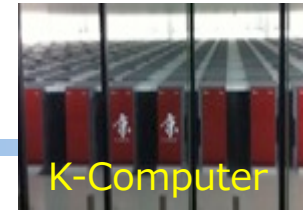
$$d_i = \sum_j A_{i,j} \quad L = D - A$$



$$L\mathbf{x} = \lambda D\mathbf{x}$$

Interaction of Data Elements

Applications



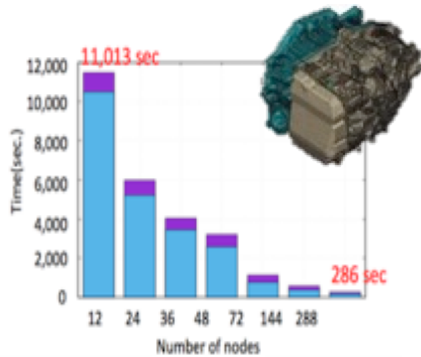
K-Computer



Cluster Systems

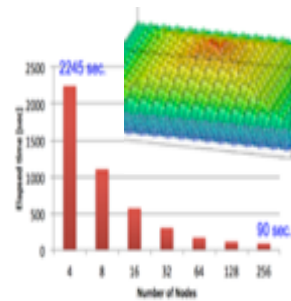
Simulation

Structural Analysis

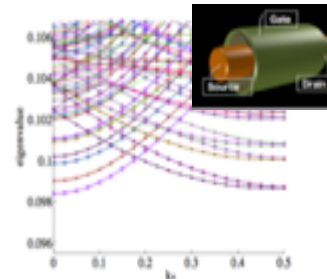


Vibration analysis

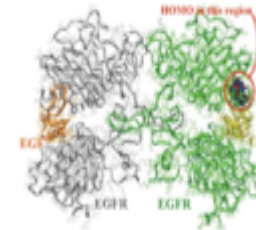
First-principles calculation



Quantum dots

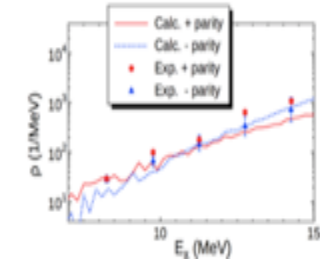


Next generation silicon device



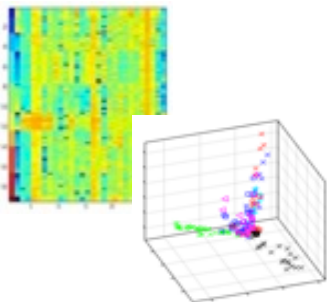
Molecular simulation

Elementary physics

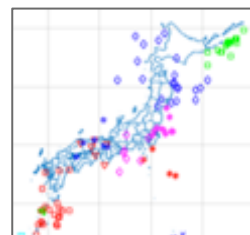


Lattice-QCD & Nuclei Physics

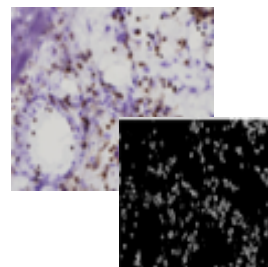
Data analysis, Deep learning



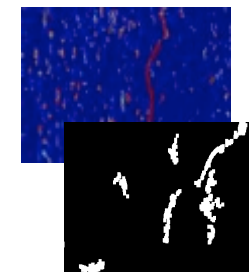
Gene data analysis



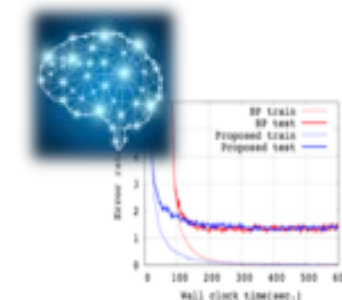
Earthquake risk evaluation



H&E image analysis



Anomaly detection of infrastructure



Neural network computation

A Scalable Solver for Nonlinear Eigenvalue Problems

ESSEX II – Equipping Sparse Solvers for Exascale

Gerhard Wellein

Bruno Lang

Achim Basermann

Holger Fehske

Georg Hager

Tetsuya Sakurai

Kengo Nakajima

Computer Science, University Erlangen

Applied Computer Science, University Wuppertal

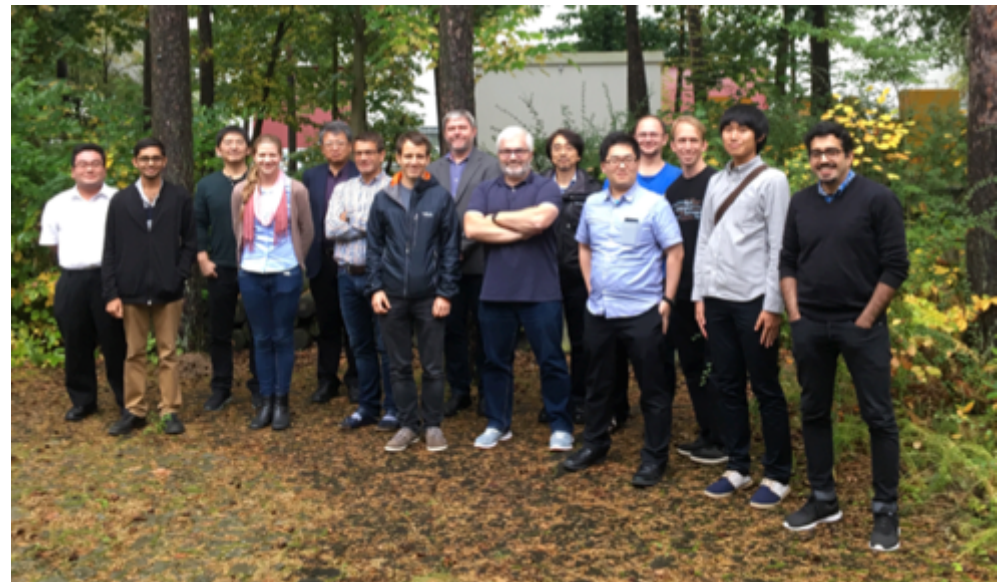
Simulation & SW Technology, German Aerospace

Institute for Physics, University Greifswald

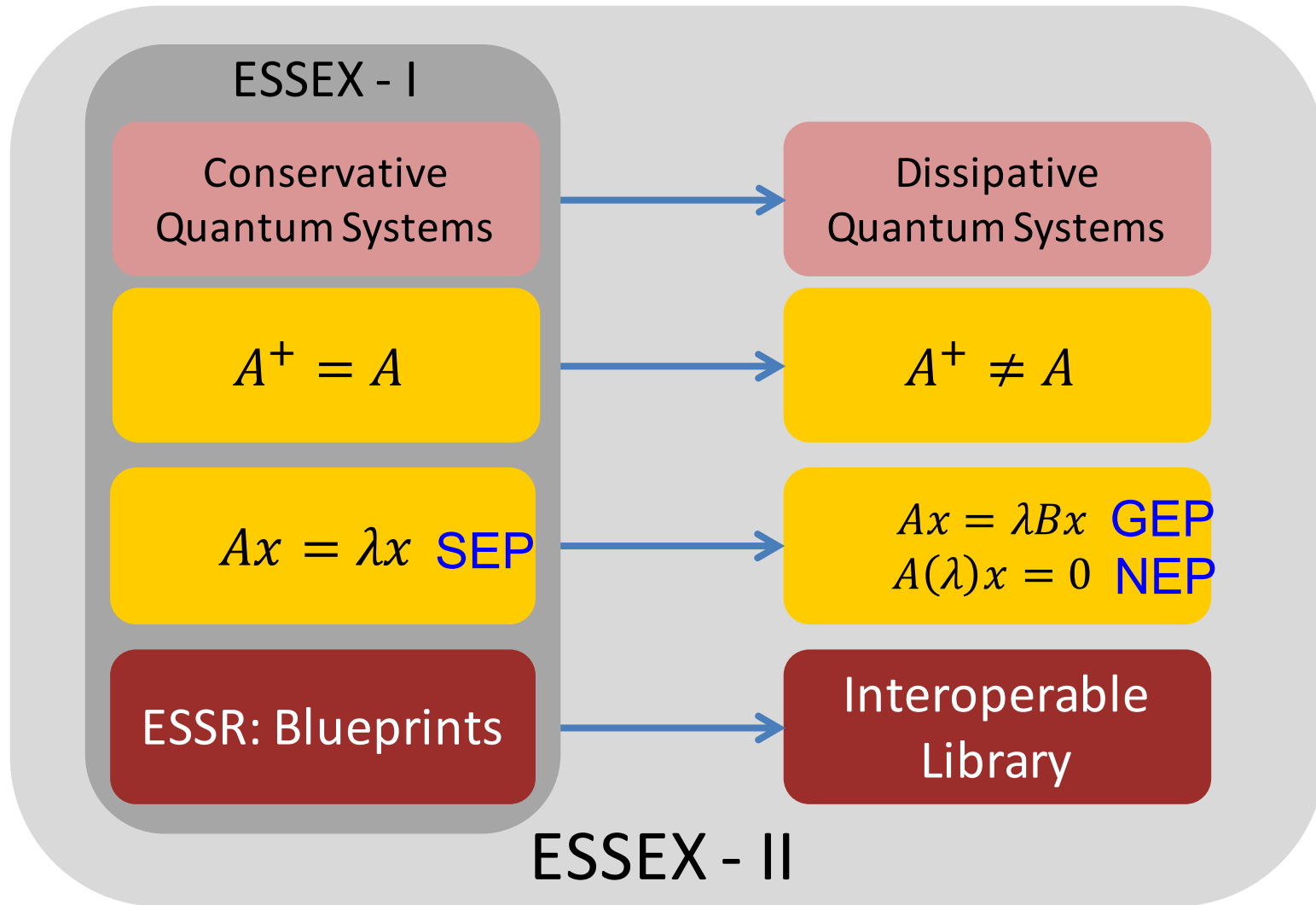
Erlangen Regional Computing Center

Applied Mathematics, University of Tsukuba

Computer Science, University of Tokyo



ESSEX-II



Nonlinear eigenvalue problems

Eigenvalue problem:

$$P(\lambda)\mathbf{x} = \mathbf{0} \quad P(\lambda) \in \mathbb{C} \rightarrow \mathbb{C}^{n \times n} \quad \mathbf{x} \in \mathbb{C}^n \setminus \{\mathbf{0}\}$$

Quadratic eigenvalue problem:

$$P(\lambda) = \lambda^2 A_2 + \lambda A_1 + A_0$$

Polynomial eigenvalue problem:

$$P(\lambda) = \lambda^k A_k + \lambda^{k-1} A_{k-1} + \cdots + A_0$$

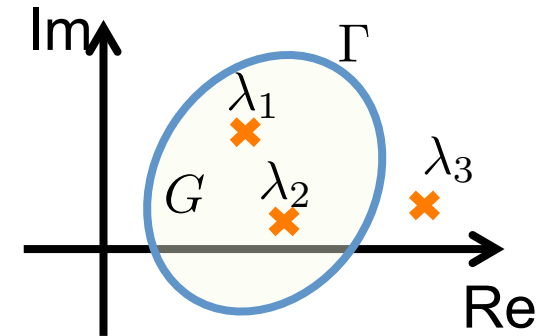
General nonlinear:

$$P(\lambda) = \sqrt{\lambda}A + e^{-\lambda}B$$

Scalable Parallel Eigensolver

Quadrature-type Eigensolver:

Sakurai-Sugiura method (SSM) computes all the eigenvalues inside a given Jordan curve.



- For generalized eigenvalue problems (GEPs)
 - [SS-Hankel \(Sakurai and Sugiura, 2003\)](#)
 - SS-RR (Sakurai and Tadano, 2007)
 - SS-Arnoldi (Imakura, Du and Sakurai, 2013)

- For nonlinear eigenvalue problems (NEPs)
 - [SS-Hankel \(Asakura, Sakurai, et al. 2009\)](#)
 - SS-RR (Yokota and Sakurai, 2013)

NEP in Complex Band Structure Calculation

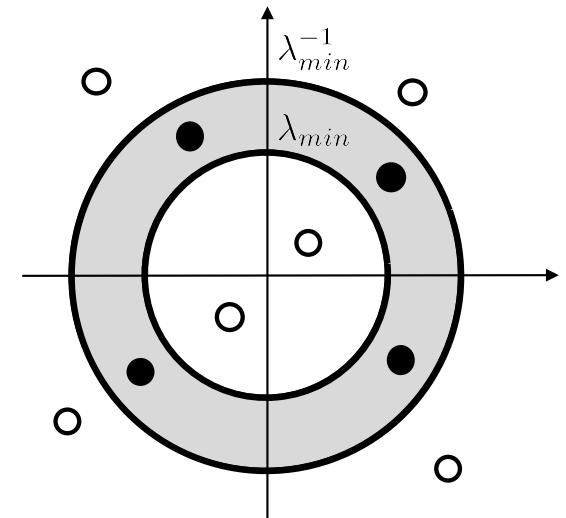
A nonlinear eigenvalue problem (NEP)

$$P(\lambda)\mathbf{x} = \mathbf{0} \quad (\lambda \in \mathbb{C}, P(\lambda) \in \mathbb{C}^{n \times n}, \mathbf{x} \in \mathbb{C}^n \setminus \{\mathbf{0}\})$$

appears in complex band structure (CBS) calculation
where

$$P(\lambda) = -\lambda^{-1}H_{i-1,i} + (E - H_{i,i}) - \lambda H_{i,i+1}$$

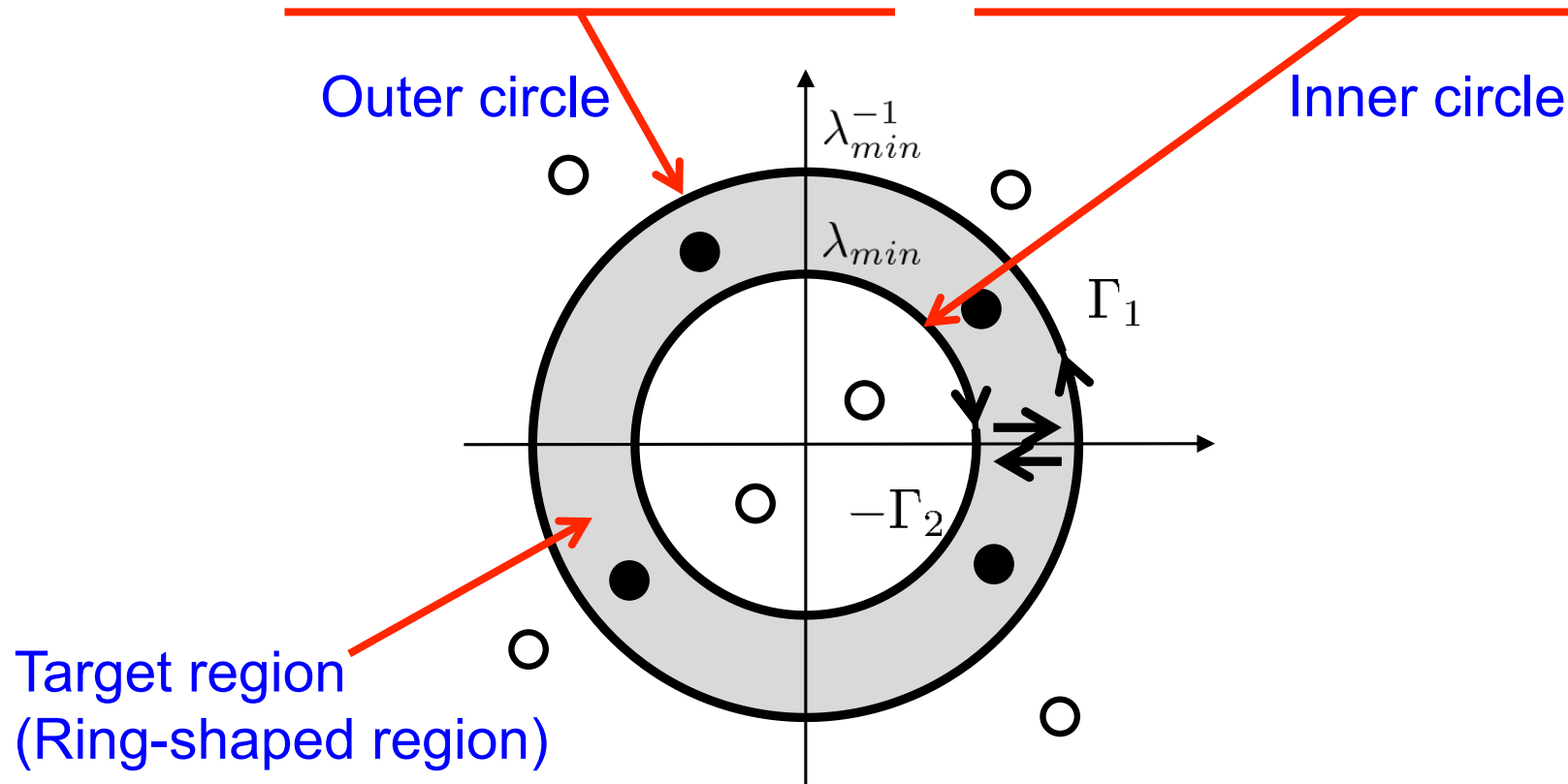
Eigenvalues in a ring region are required



Computing Eigenvalues in a Ring-shaped Region

Using SSM, we can compute the target eigenvalues by setting two circles (clockwise and anti-clockwise)

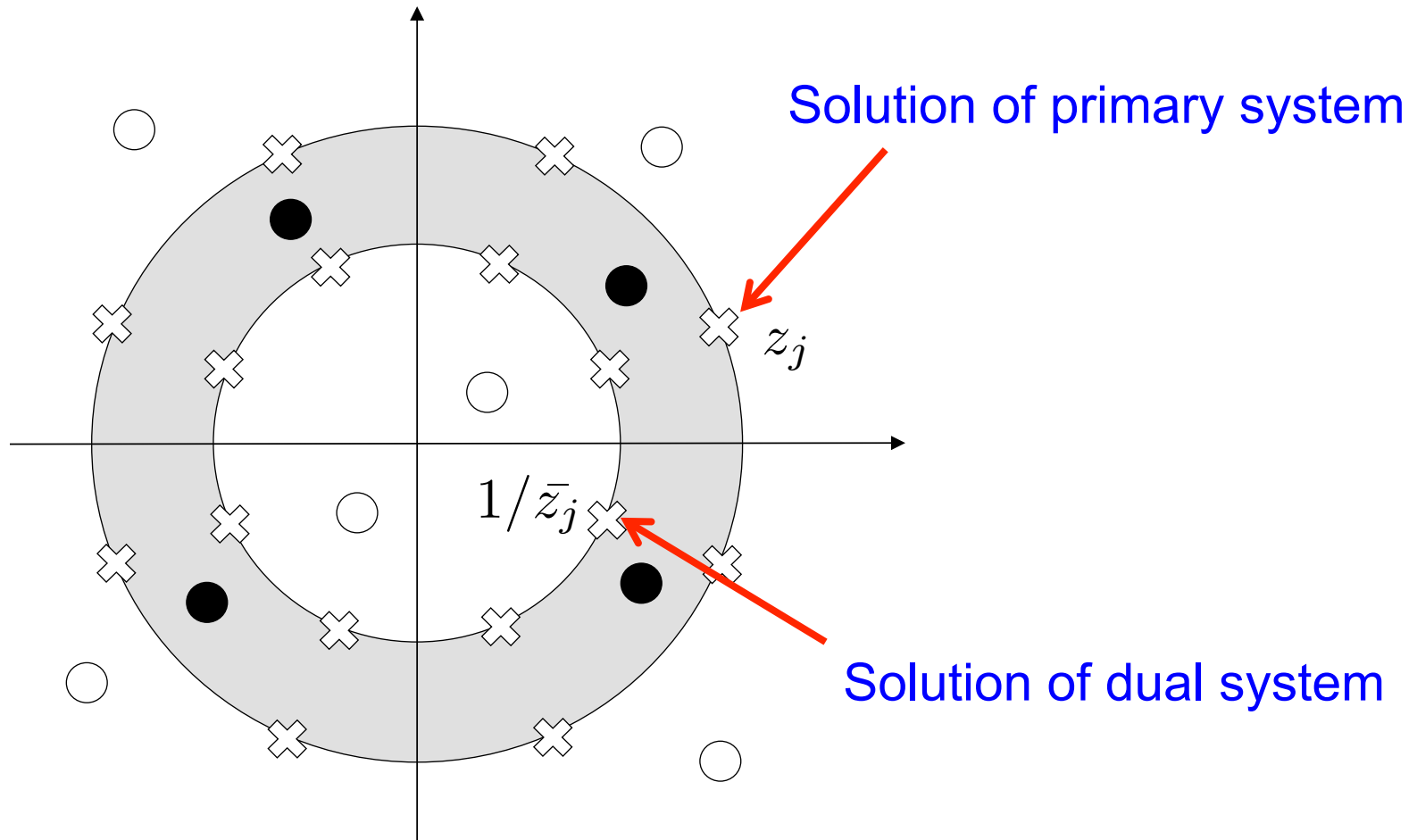
$$S_k := \frac{1}{2\pi i} \oint_{\Gamma_1} z^k P(z)^{-1} V dz - \frac{1}{2\pi i} \oint_{\Gamma_2} z^k P(z)^{-1} V dz$$



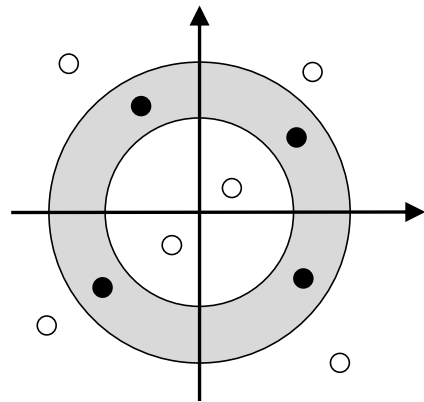
Utilization of BiCG

- BiCG method solves two linear systems:
Primary system and Dual system

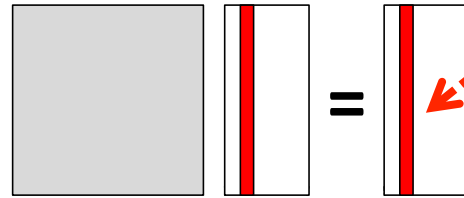
We can obtain solutions at z_j and $1/\bar{z}_j$ from one BiCG.



Three level parallelism of SSM in CBS calculation

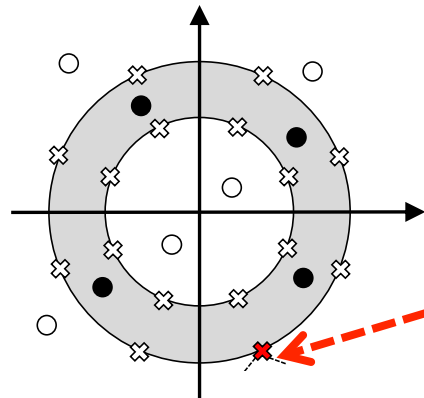


Linear systems
at each quadrature point

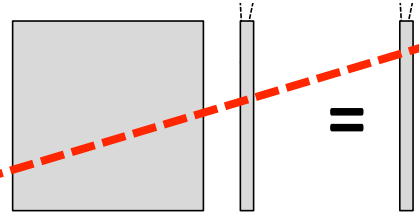


Top level:

Linear system for each
right-hand side vector

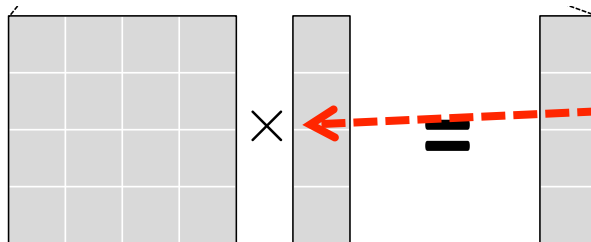


Linear systems
at each quadrature point



Middle level:

Linear system for each
quadrature point



Bottom level:

Sparse matrix-vector
product

Performance Evaluation

Performance evaluation

■ BN-CNT with 1,024 atoms

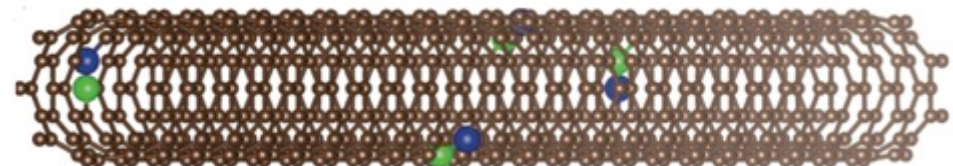
- # of grid points = $72 \times 72 \times 640 = 3,317,760$ (matrix size)
- # of eigenvalues = 22, $\lambda_{min} = 0.5$

■ Parameters for SSM and BiCG

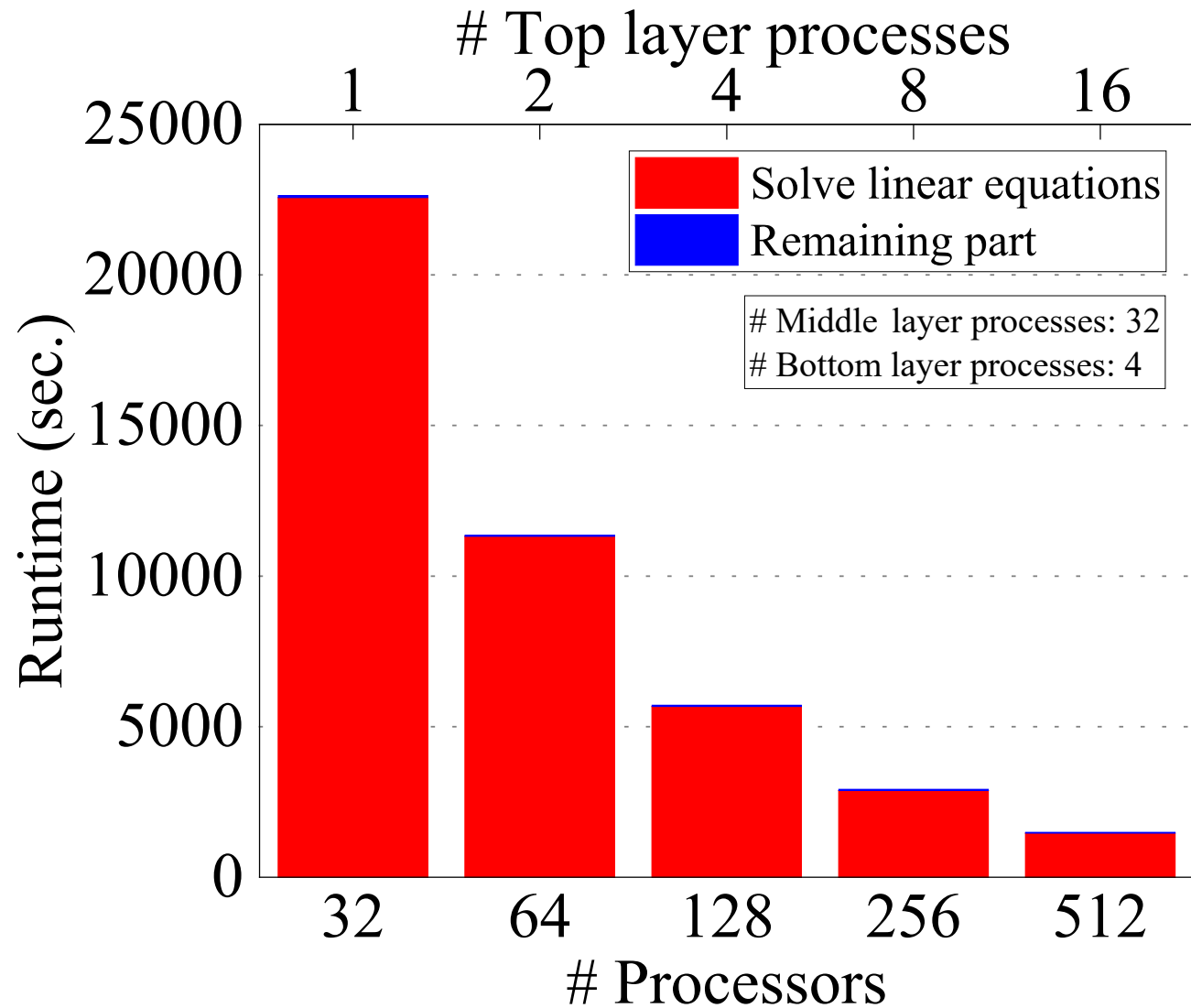
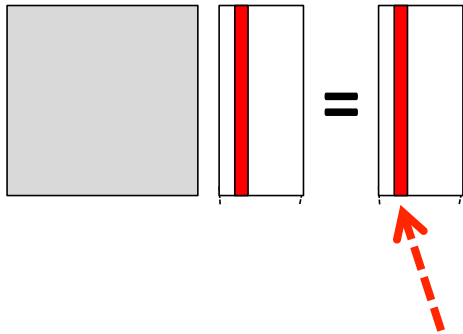
- $N=32$, $M=8$, $L=16$
- Stopping criterion for BiCG : Relative residual $< 1e-10$

■ Computing environment: [Oakforest-PACS](#)

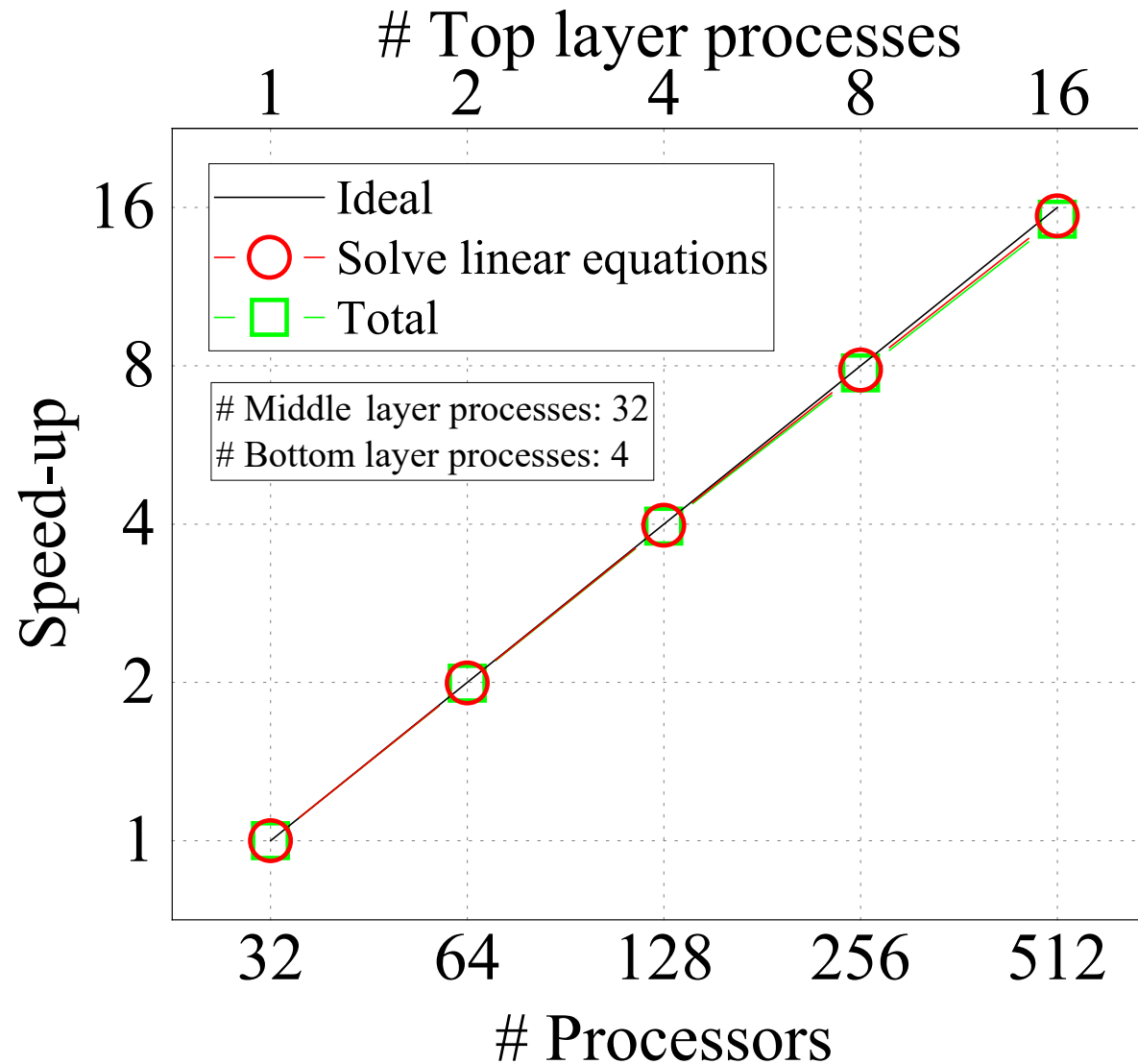
- Parallelization setting
 - 17 threads/MPI
 - 4 MPI/node



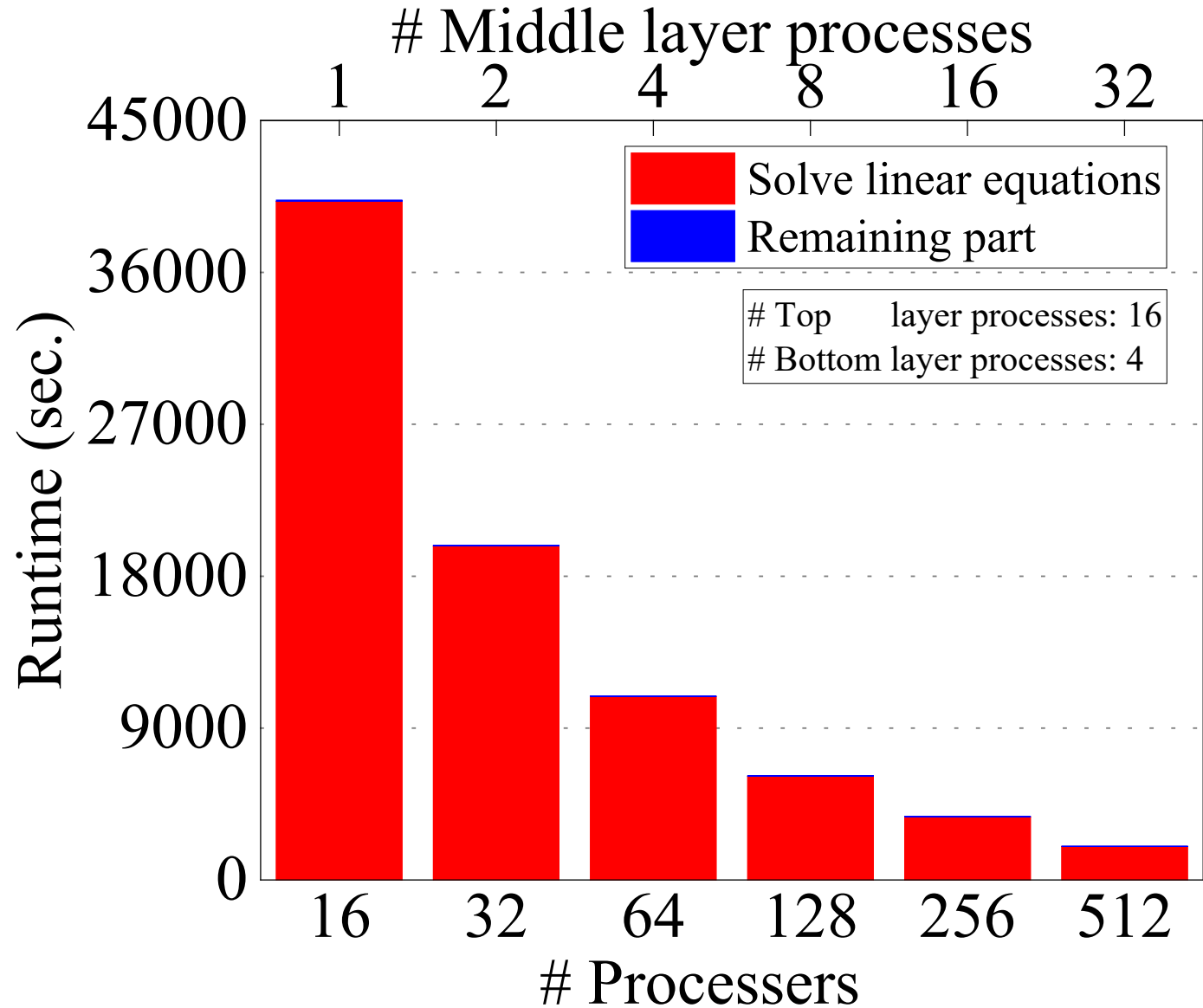
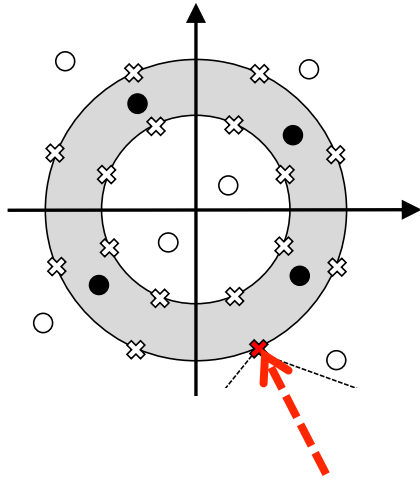
Top layer



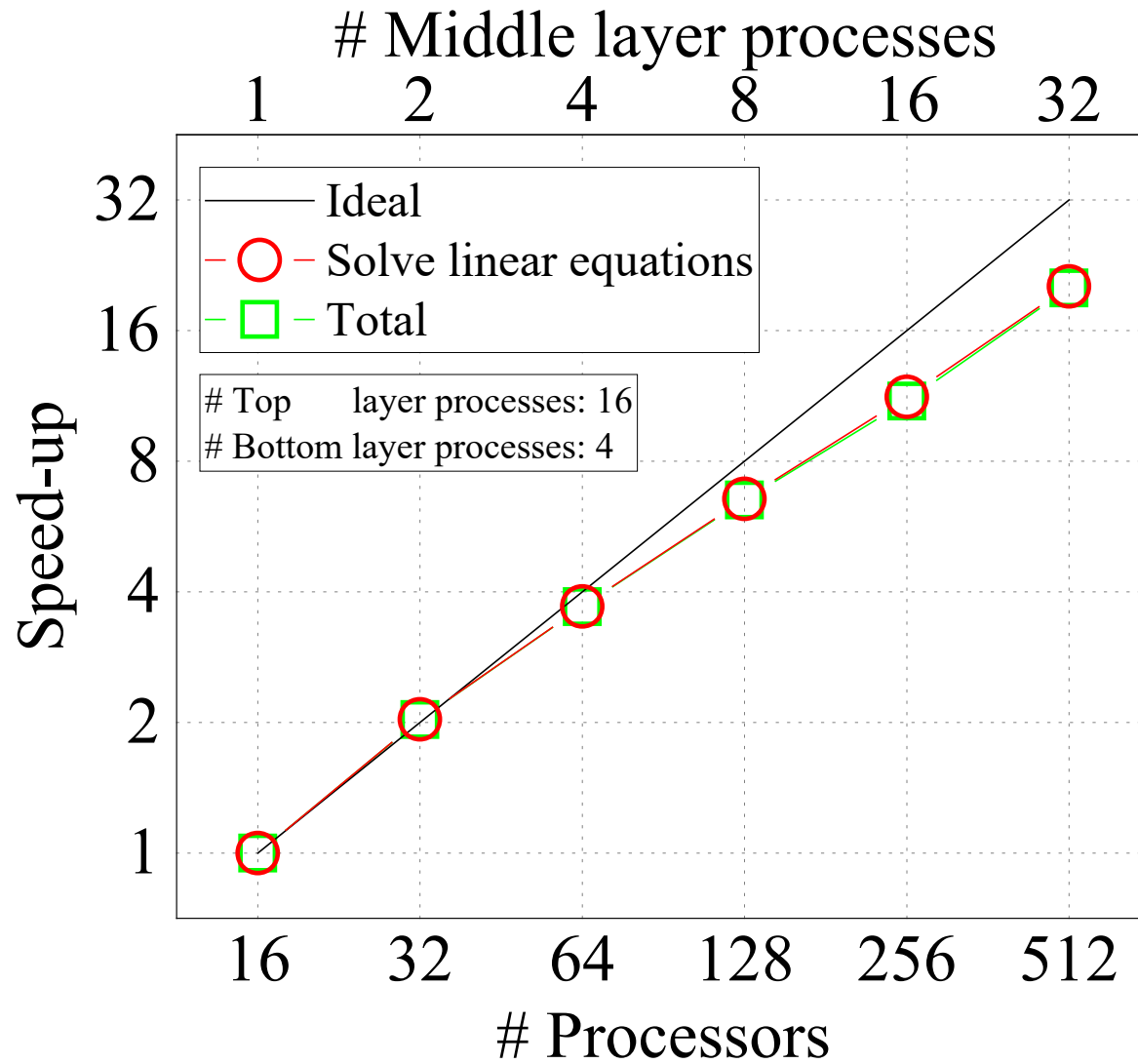
Top layer (speedup ratio)



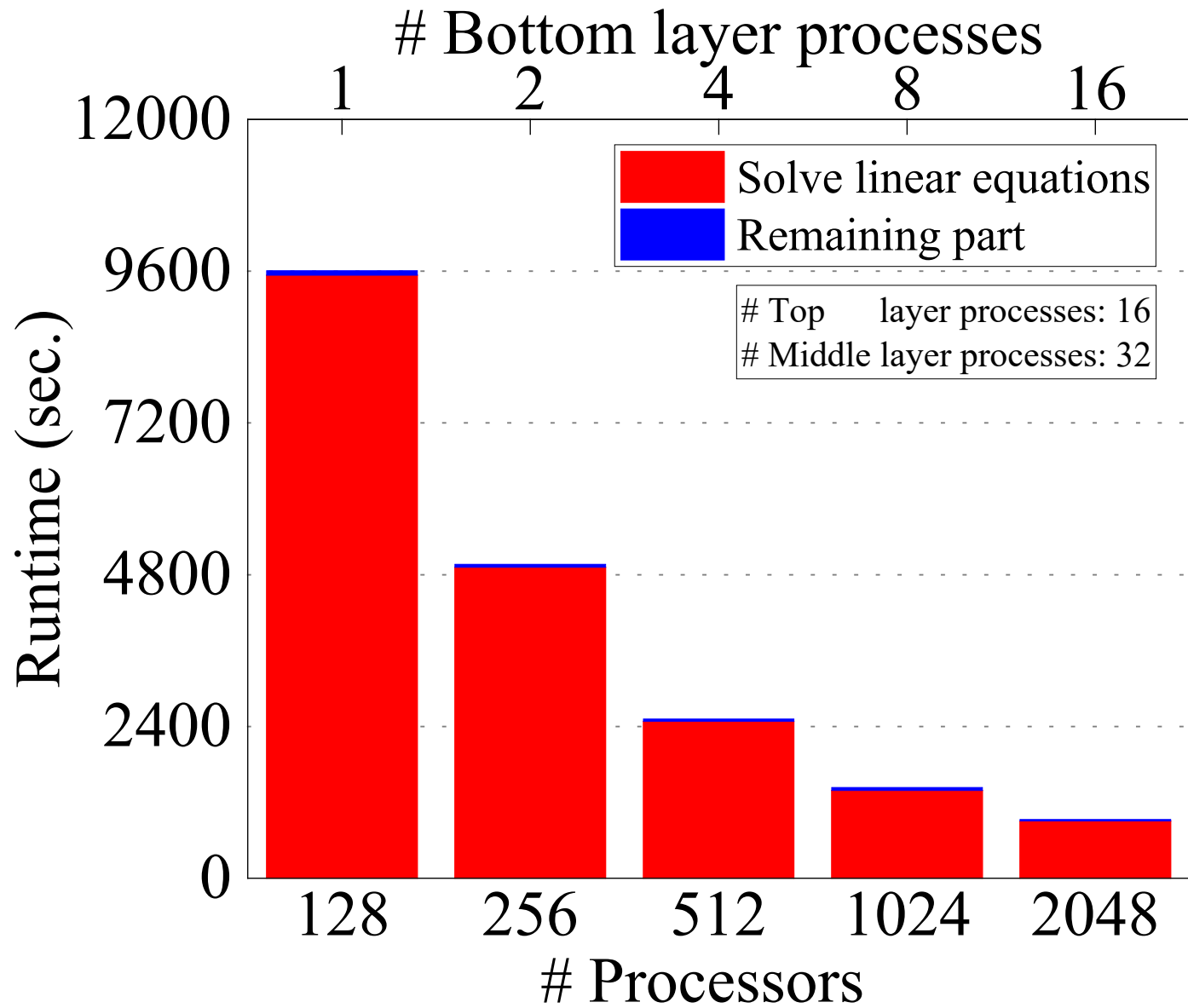
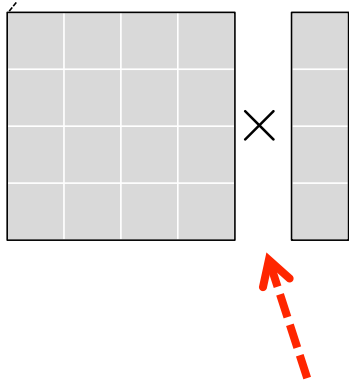
Middle layer



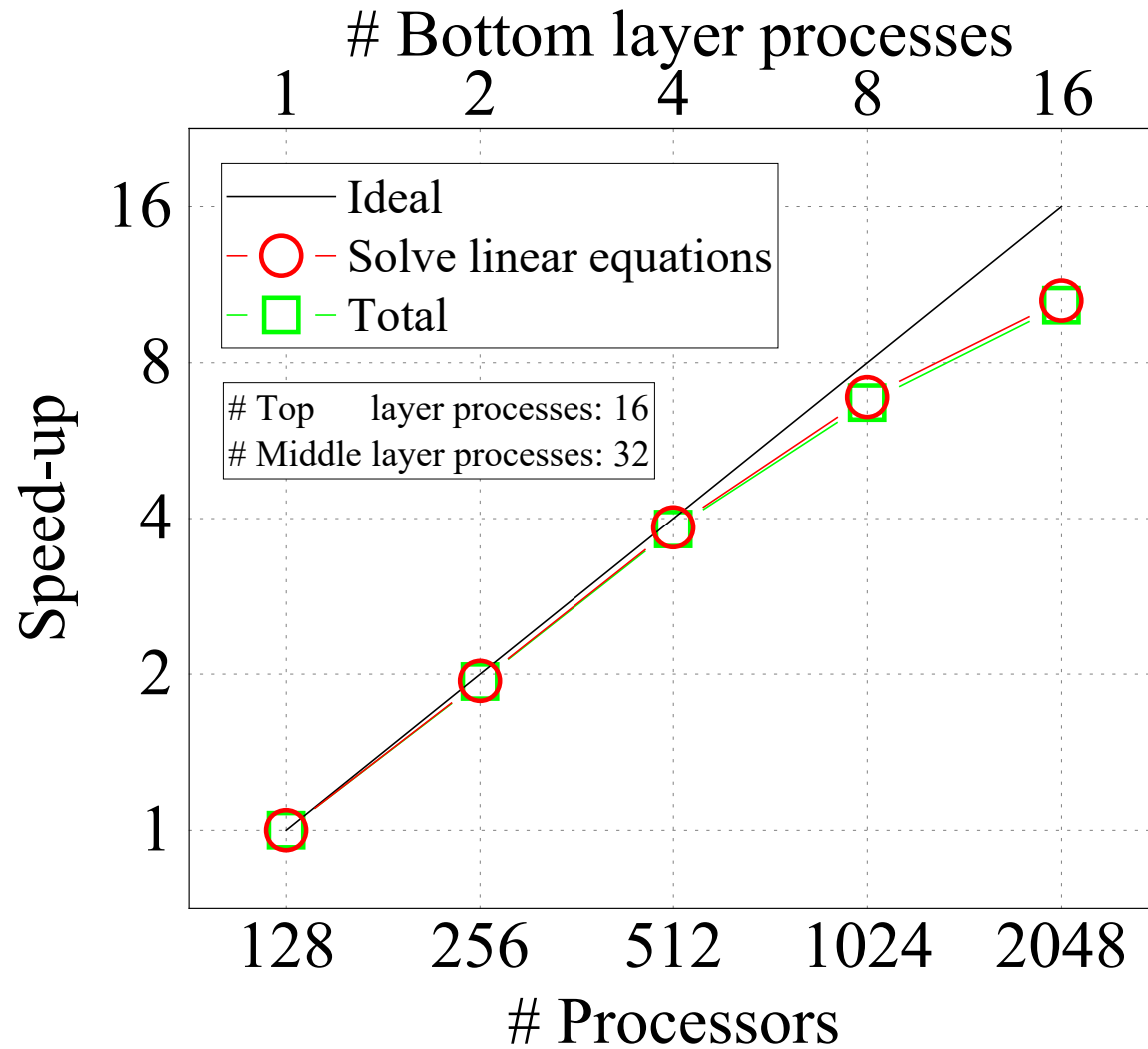
Middle layer (speedup ratio)



Bottom layer

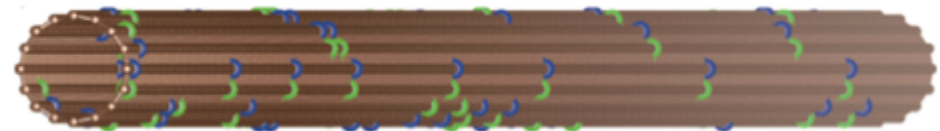


Bottom layer (speedup ratio)

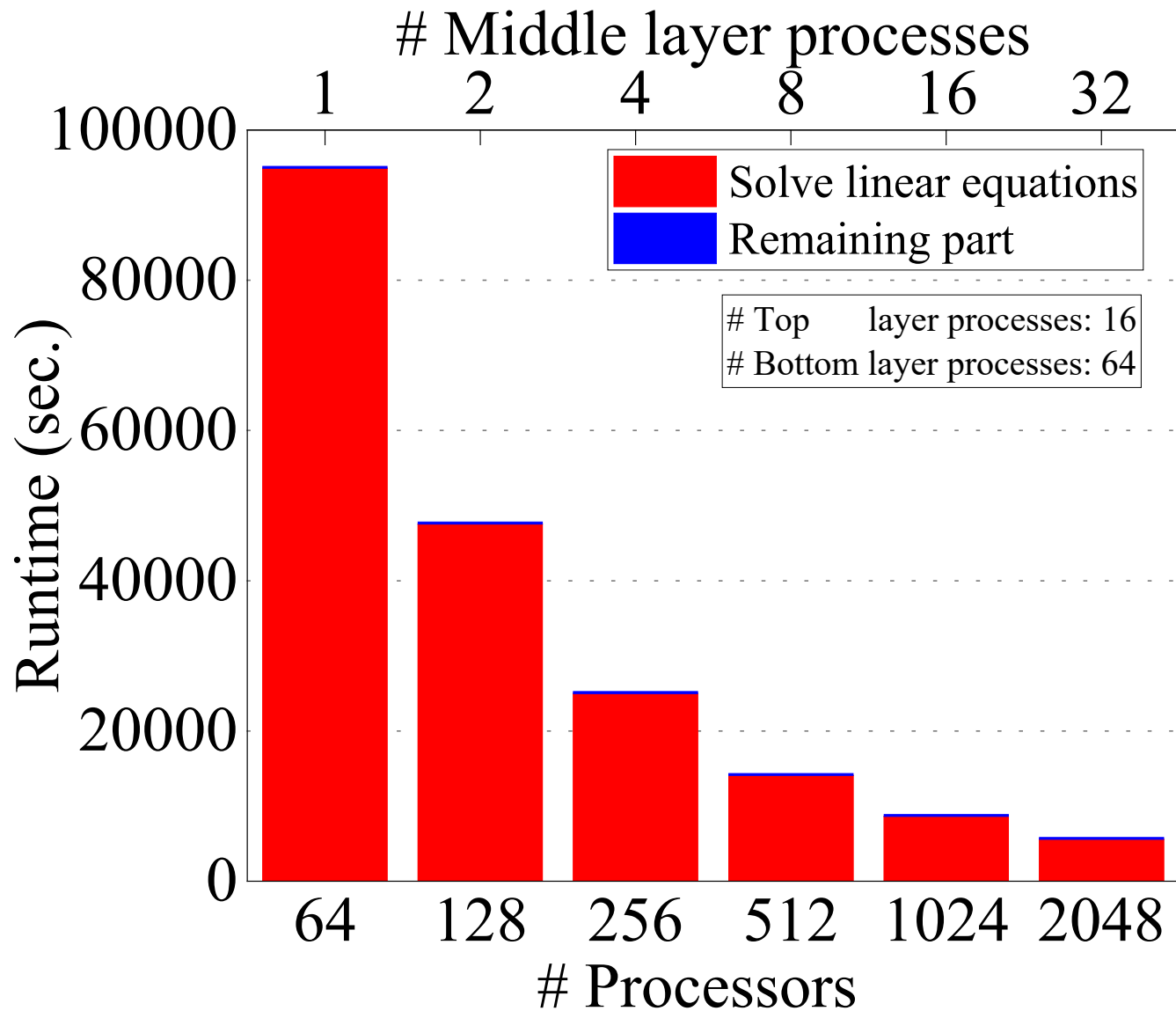


Performance evaluation

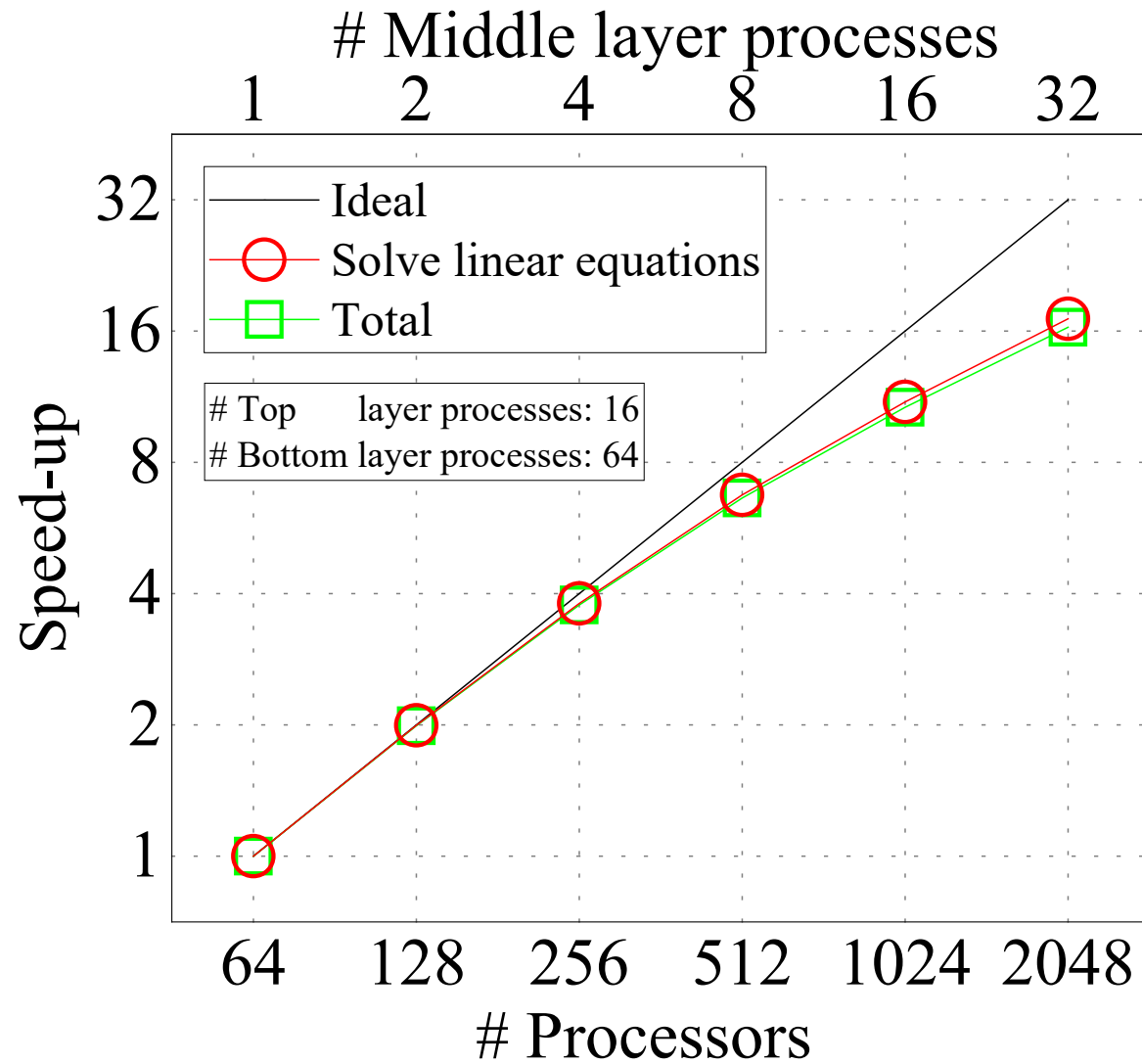
- **BN-CNT with 10,240 atoms**
 - # of grid points = $72 \times 72 \times 6400 = 33,177,600$ (matrix size)
 - # of eigenvalues = 34, $\lambda_{min} = 0.5$
- Parameters for SSM and BiCG
 - N=32, M=8, L=16
 - Stopping criterion for BiCG : Relative residual < $1e-10$
- Computing environment: **Oakforest-PACS**
 - **Parallelization setting**
 - 4 threads/MPI
 - 16 MPI/node



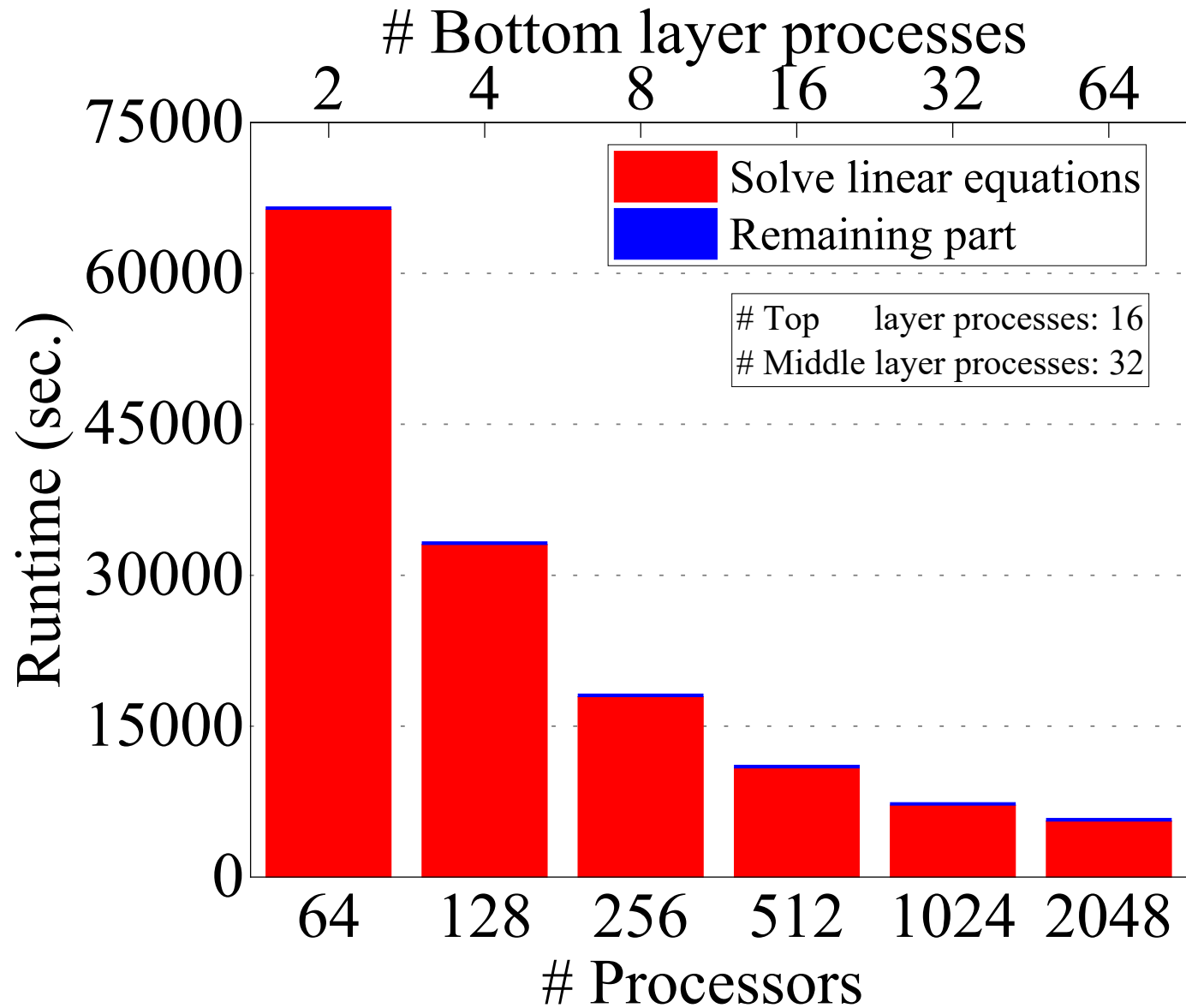
Middle layer



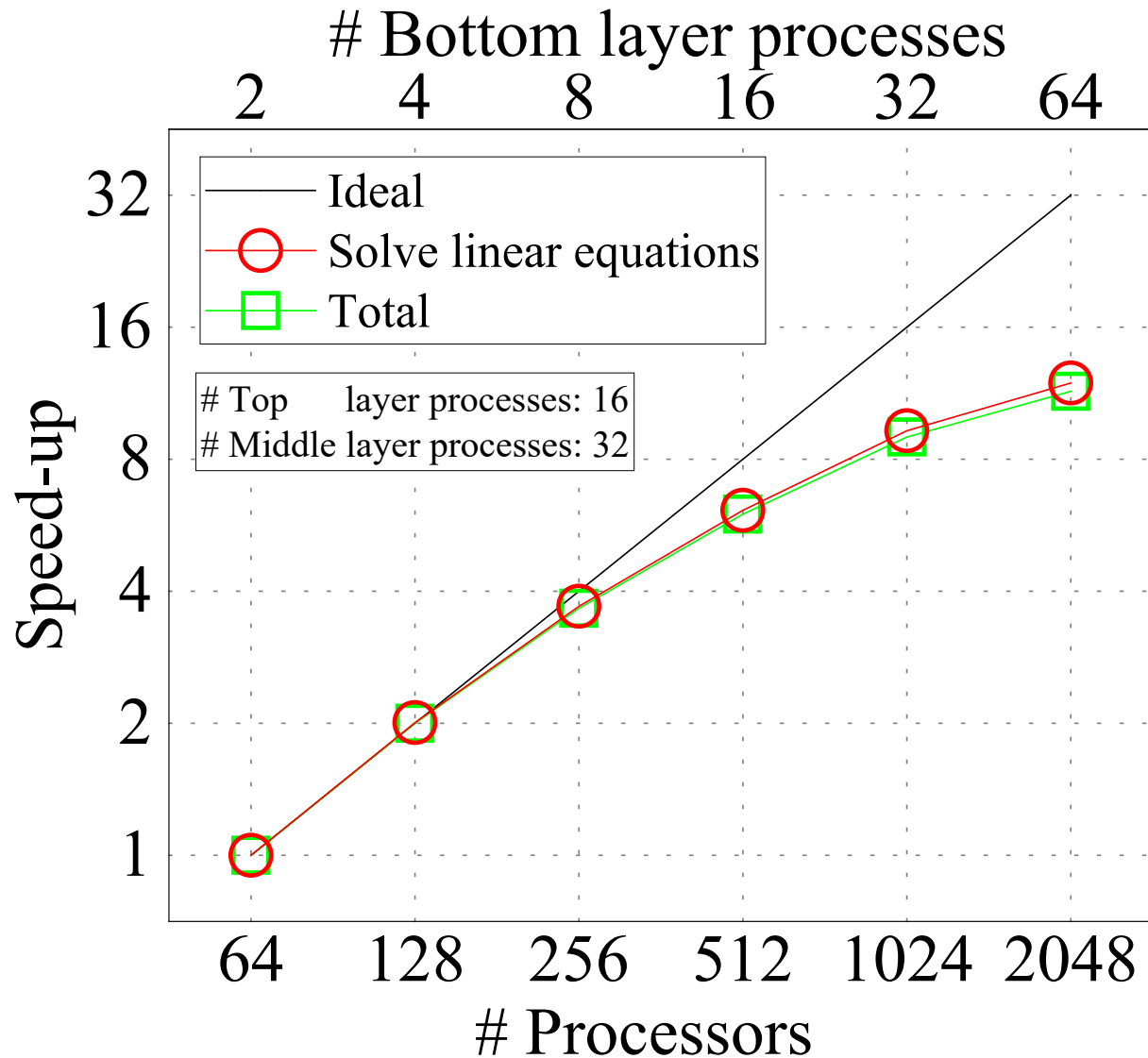
Middle layer (speedup ratio)



Bottom layer



Bottom layer (speedup ratio)



Conclusions

■ Simulation and Data Analysis

- Large-scale problems
- High performance eigenvalue solver
- Various applications

■ Developing a scalable parallel eigensolver

- Sakurai-Sugiura method (SSM)
 - Quadrature-type parallel eigensolver
 - Hierarchical parallel structure

■ Nonlinear Sakurai-Sugiura method for ring-shaped region

- Application for complex band structure calculation
 - Performance evaluation on Oakforest-PACS
 - BN-CNT with 10,240 atoms