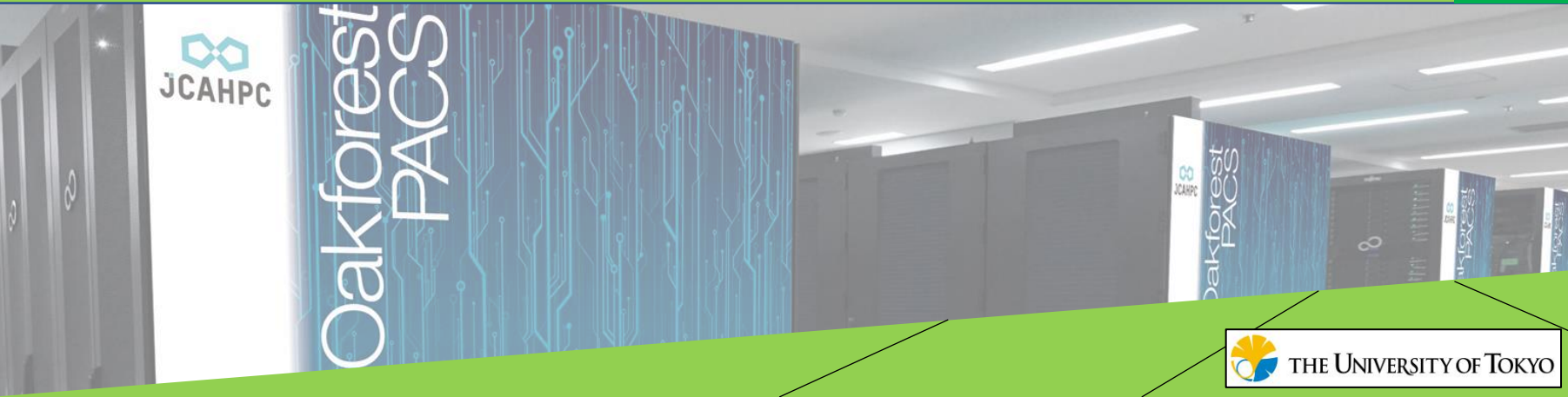
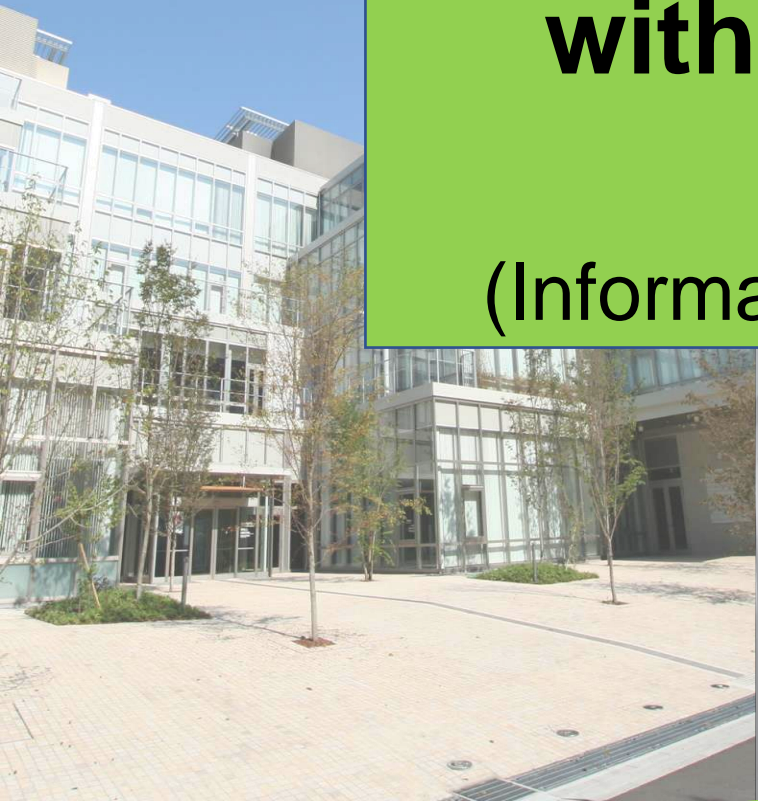


*H*ACApK library with low-rank structured matrices

Akihiro Ida

(Information Technology Center, University of Tokyo)



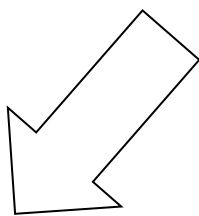
Overview of low-rank structured matrices

- Approximation technique for dense matrices from Integral operator.

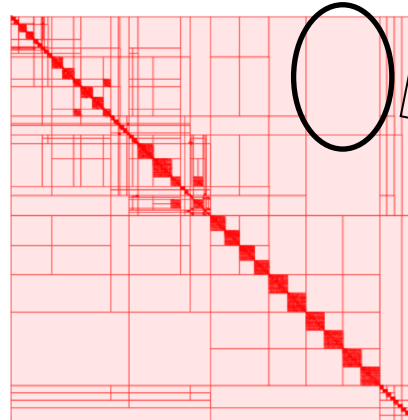
$$g[u](x) = \int_{\Omega} g(x, y) u(y) dy$$

\mathcal{H} -matrices

Discretization

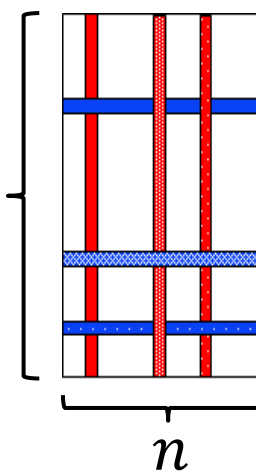


Permutation
Partition

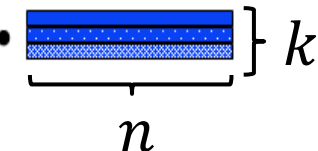
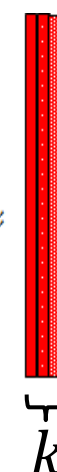


Low-rank
approximation

m



\approx



n

k

n

■ Full-Rank

■ Low-Rank

- $O(mn) \Rightarrow O(k(m+n))$
- Accuracy control : # of k

Full rank dense matrix

\mathcal{H} ACApK library for low-rank structured matrices

■ Library for simulations using BEM(boundary element method)

➤ Application

Integral operator

$$\mathcal{G}[u](x) = \int_{\Omega} g(x, y) u(y) dy$$

Degenerate kernel:

$$g(x, y) \cong \sum_{\nu=1}^r g_1^{\nu}(x) g_2^{\nu}(y)$$

, for far remote x, y

e.g. $g(x, y) \propto |x - y|^{-1}$

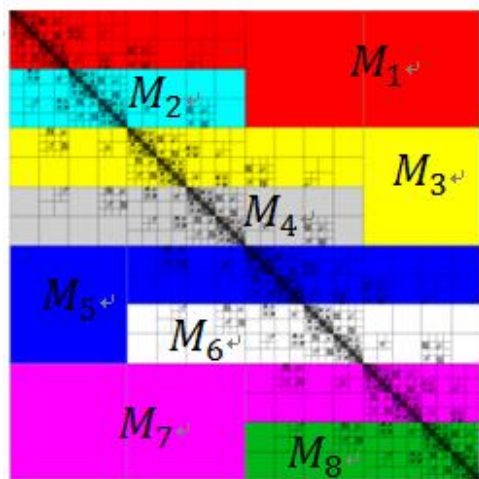
➤ Download site :

- <http://ppopenhpc.cc.u-tokyo.ac.jp>
- <https://github.com/Post-Peta-Crest/ppOpenHPC/tree/MATH/HACApK>

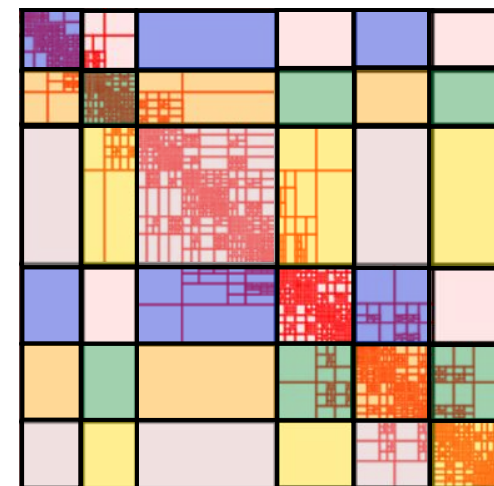
➤ (Lattice) \mathcal{H} -matrices : $\mathcal{O}(N^2) \Rightarrow \mathcal{O}(N \log N)$

➤ Parallel computing

For CPU cluster : MPI+OpenMP (partially GPU)



Assignment for \mathcal{H}



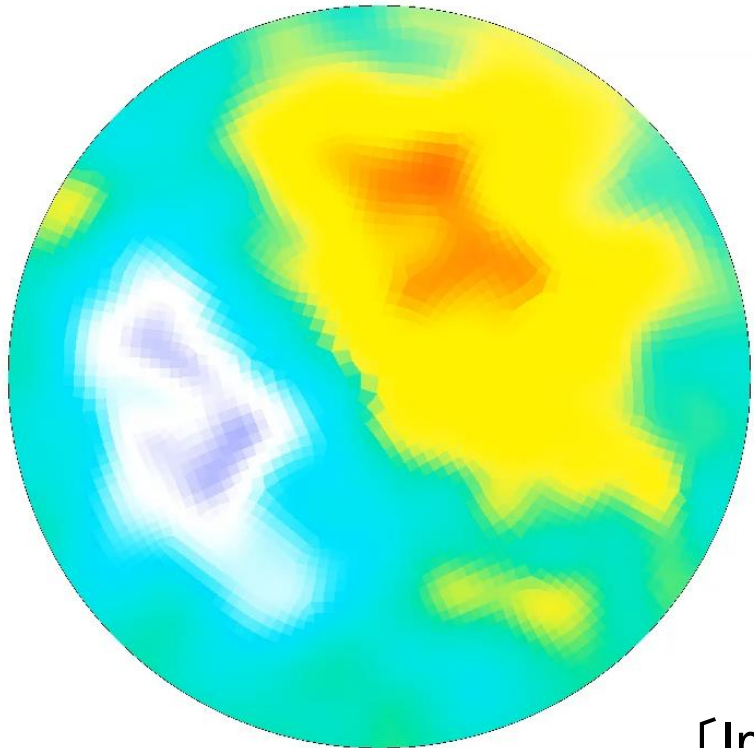
Assignment for lattice \mathcal{H}

Example analyses using $\mathcal{H}ACApK$

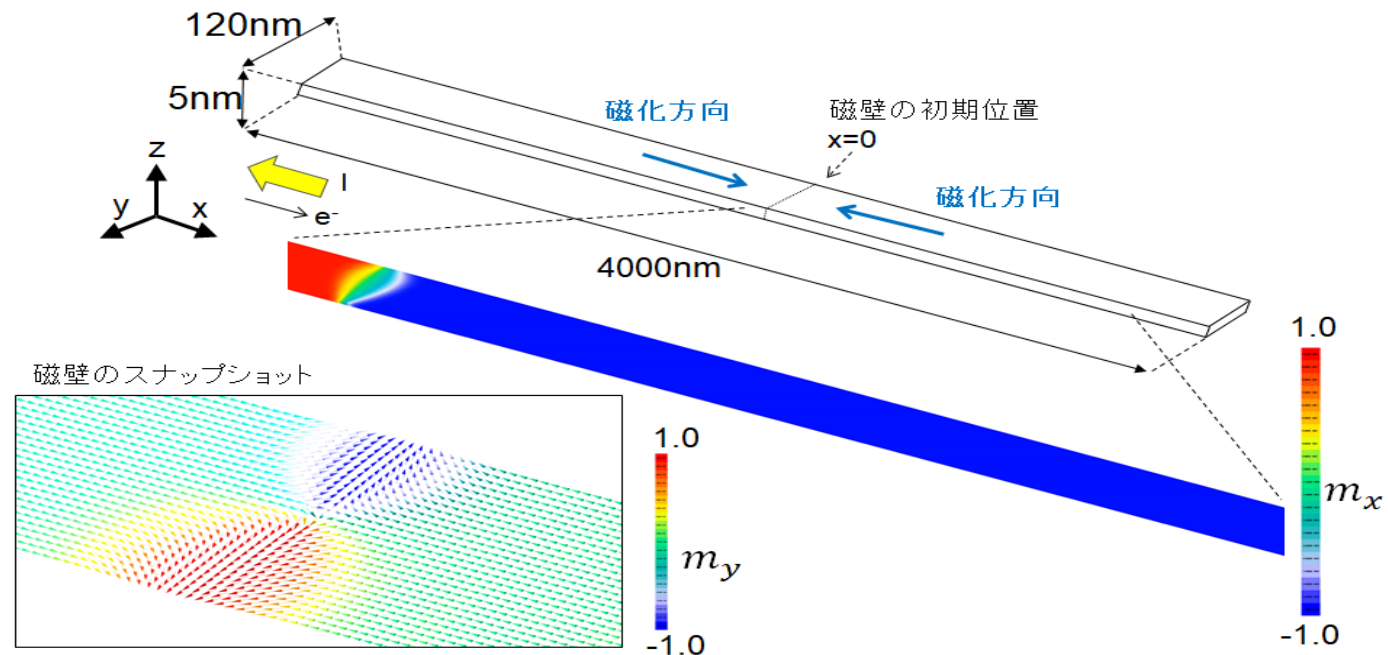
■ Micromagnetic Simulation

$$g(x, y) = \frac{1}{4\pi} \int_{\partial V} \varphi_1(\mathbf{y}) \nabla |\mathbf{x} - \mathbf{y}|^{-1} \cdot d\mathcal{S}$$

➤ Spin-torque oscillator



➤ Current-induced domain wall Motion

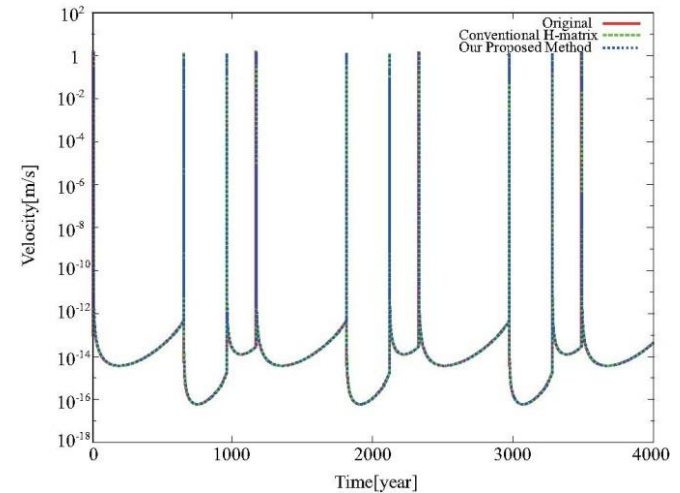
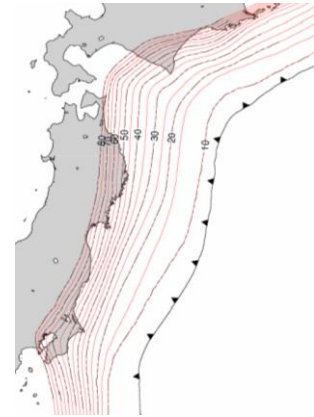


[Image courtesy : T. Ataka (Fujitsu Ltd.)]

Example analyses using $\mathcal{H}ACApK$

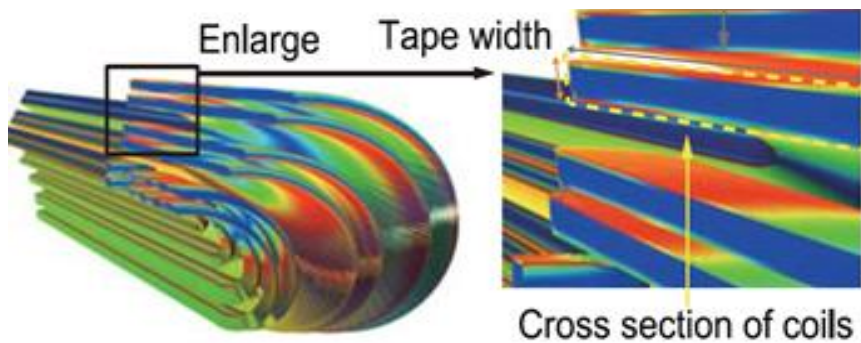
Earthquake cycle Integral operator with $g(x, y) \propto |x - y|^{-3}$

➤ eq. motion $\tau_i = \sum_{j=1}^N K_{ij} (u_j - V_{pl} t) - \frac{\xi G}{2V_s} V_i$
 ➤ friction law
 $\tau_i = \mu \sigma_n^{eff} = \tau_* + A_i \ln(V_i / V_*) + B_i \ln(V_* \theta_i / L_i)$
 $\frac{d\theta_i}{dt} = \exp(-V_i / V_c) - V_i \theta_i / L_i \ln(V_i \theta_i / L_i)$



Superconductor Analyses

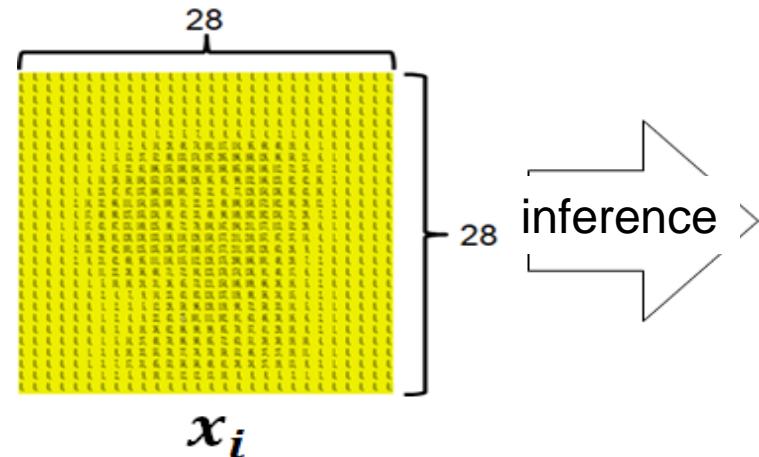
$$g(\mathbf{r}) = \frac{\mu_0 t_s}{4\pi} \int_{S'} \frac{(\nabla \times \mathbf{n}' T') \times \mathbf{r} \cdot \mathbf{n}}{r^3} dS'$$



[Image courtesy : Amemiya Lab. (Kyoto Univ.)]

Machine learning

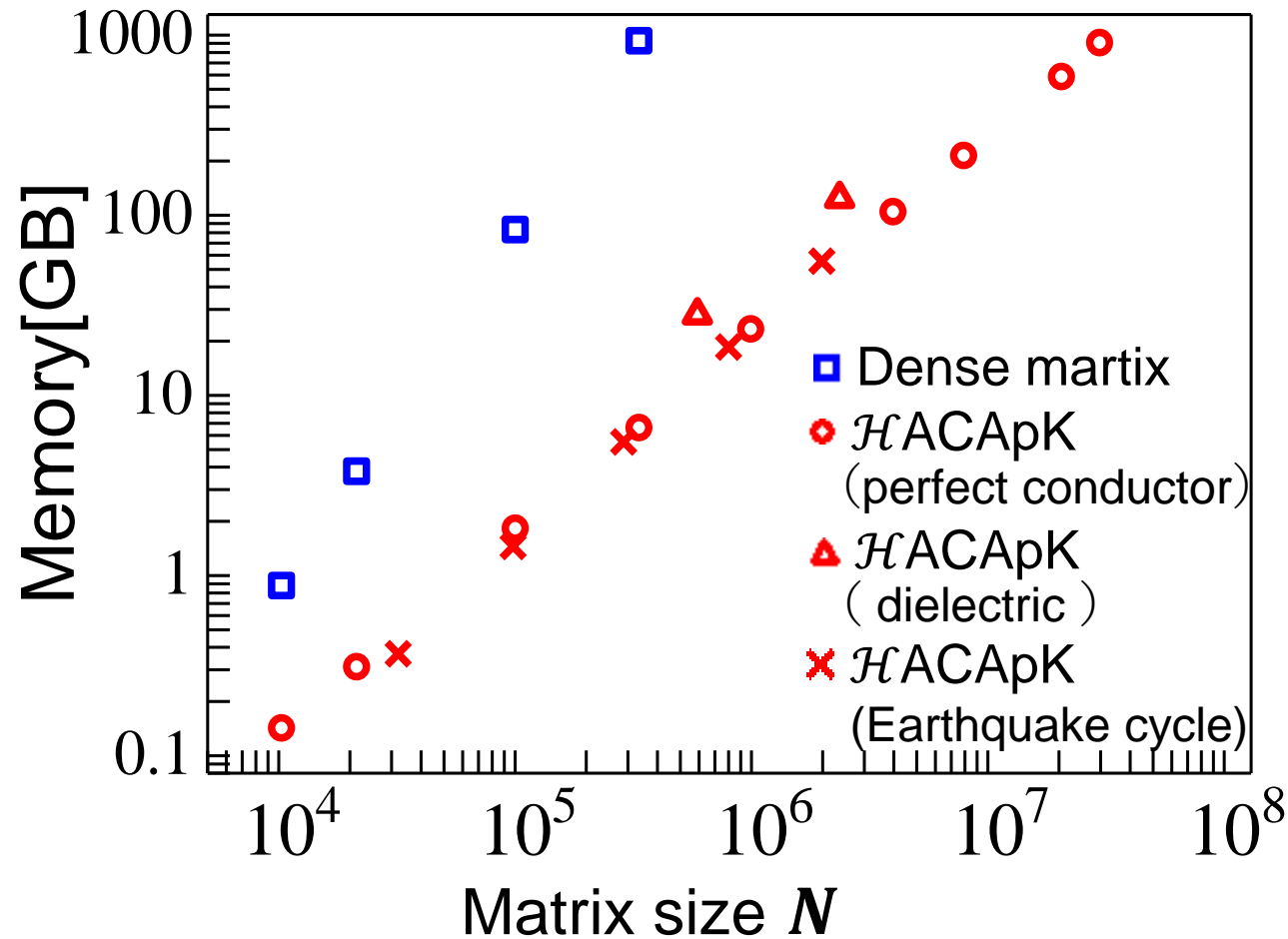
$$A_{ij} := \exp\left(-\frac{\|\mathbf{x}_i - \mathbf{x}_j\|_2^2}{2\sigma^2}\right)$$



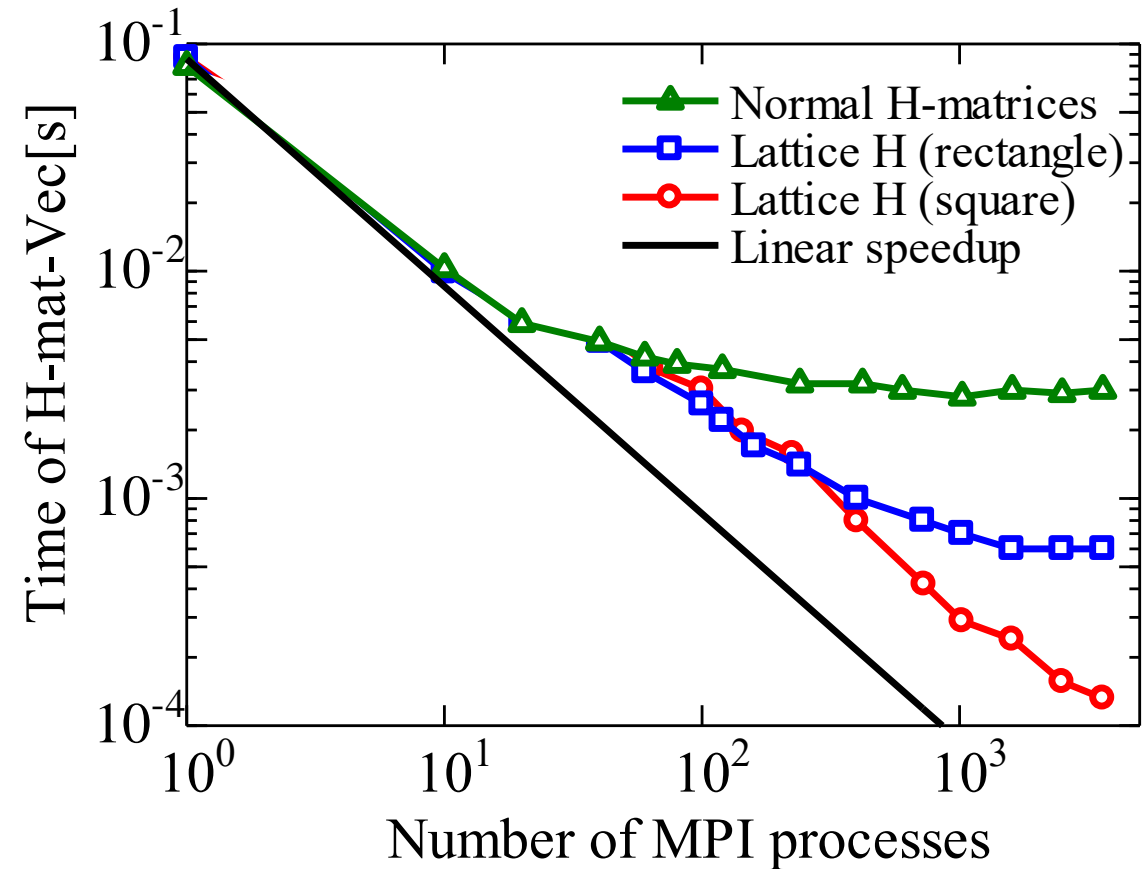
Performance of \mathcal{H} ACApK library

Memory usage

- Dense matrix: $\mathcal{O}(N^2)$ ➤ \mathcal{H} ACApK: $\mathcal{O}(N \log N)$



Parallel scalability



- Calculation condition
- Micromagnetic Simulation : $N = 20,910$
 - Computer: Fujitsu PRIMERGY CX2550

Conclusion

- **HACApK**: Library for low-rank structured matrices
- Integral operator to which low-rank structured matrices can be applied
 - ▶ $g[u] = f$, where $g[u](x) := \int_{\Omega} \kappa(x, y) u(y) dy$
 - Degenerate kernel: $\kappa(x, y) \cong \sum_{\nu=1}^r \kappa_1^{\nu}(x) \kappa_2^{\nu}(y)$, for far remote x, y
 - Naïve discretization yields **dense matrices** : $\mathcal{O}(N^2)$
- For large-scaled simulations
 - ▶ Approximation technique for dense matrices
 - **(Lattice) H-matrices** : $\mathcal{O}(N^2) \Rightarrow \mathcal{O}(N \log N)$
 - ▶ Parallel computing
 - Hybrid MPI+OpenMP for CPU clusters (partially ported to GPU)
- Download site : <https://github.com/Post-PetaCrest/ppOpenHPC/tree/MATH/HACApK>
 - Open source
 - MIT license

Thank you for watching