

# Quantum Spin S=1/2 or Qubit

$$|\phi\rangle = c_{\uparrow}|1\rangle + c_{\downarrow}|0\rangle$$

$$\hat{S}_j^{\alpha}|\phi\rangle = c'_{\uparrow}|1\rangle + c'_{\downarrow}|0\rangle$$

$$\begin{pmatrix} c'_{\uparrow} \\ c'_{\downarrow} \end{pmatrix} = \frac{1}{2}\hat{\sigma}^{\alpha} \begin{pmatrix} c_{\uparrow} \\ c_{\downarrow} \end{pmatrix}$$

$$\hat{\sigma}^x = \begin{pmatrix} 0 & +1 \\ +1 & 0 \end{pmatrix}$$

$$\hat{\sigma}^y = \begin{pmatrix} 0 & -i \\ +i & 0 \end{pmatrix}$$

$$\hat{\sigma}^z = \begin{pmatrix} +1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$\hat{S}_j^{\alpha} \doteq \frac{1}{2}\hat{\sigma}^{\alpha}$$

# Quantum Spins: Two Site TFIM

Decimal representation of orthonormalized basis

		0 th site		1 st site
$ 0\rangle_d$	=	$ \downarrow\rangle$	$\otimes$	$ \downarrow\rangle$
$ 1\rangle_d$	=	$ \uparrow\rangle$	$\otimes$	$ \downarrow\rangle$
$ 2\rangle_d$	=	$ \downarrow\rangle$	$\otimes$	$ \uparrow\rangle$
$ 3\rangle_d$	=	$ \uparrow\rangle$	$\otimes$	$ \uparrow\rangle$

$$L = 2 \quad \hat{H} = J \sum_{i=0}^{L-1} \hat{S}_i^z \hat{S}_{i+1}^z - \Gamma \sum_{i=0}^{L-1} \hat{S}_i^x$$

$$\hat{H} \doteq \begin{pmatrix} +J/2 & -\Gamma/2 & -\Gamma/2 & 0 \\ -\Gamma/2 & -J/2 & 0 & -\Gamma/2 \\ -\Gamma/2 & 0 & -J/2 & -\Gamma/2 \\ 0 & -\Gamma/2 & -\Gamma/2 & +J/2 \end{pmatrix} d \langle i | \hat{H} | j \rangle_d$$

# Larger TFIM

$$\hat{H} = J \sum_{i=0}^{L-1} \hat{S}_i^z \hat{S}_{i+1}^z - \Gamma \sum_{i=0}^{L-1} \hat{S}_i^x$$

-Non-commutative

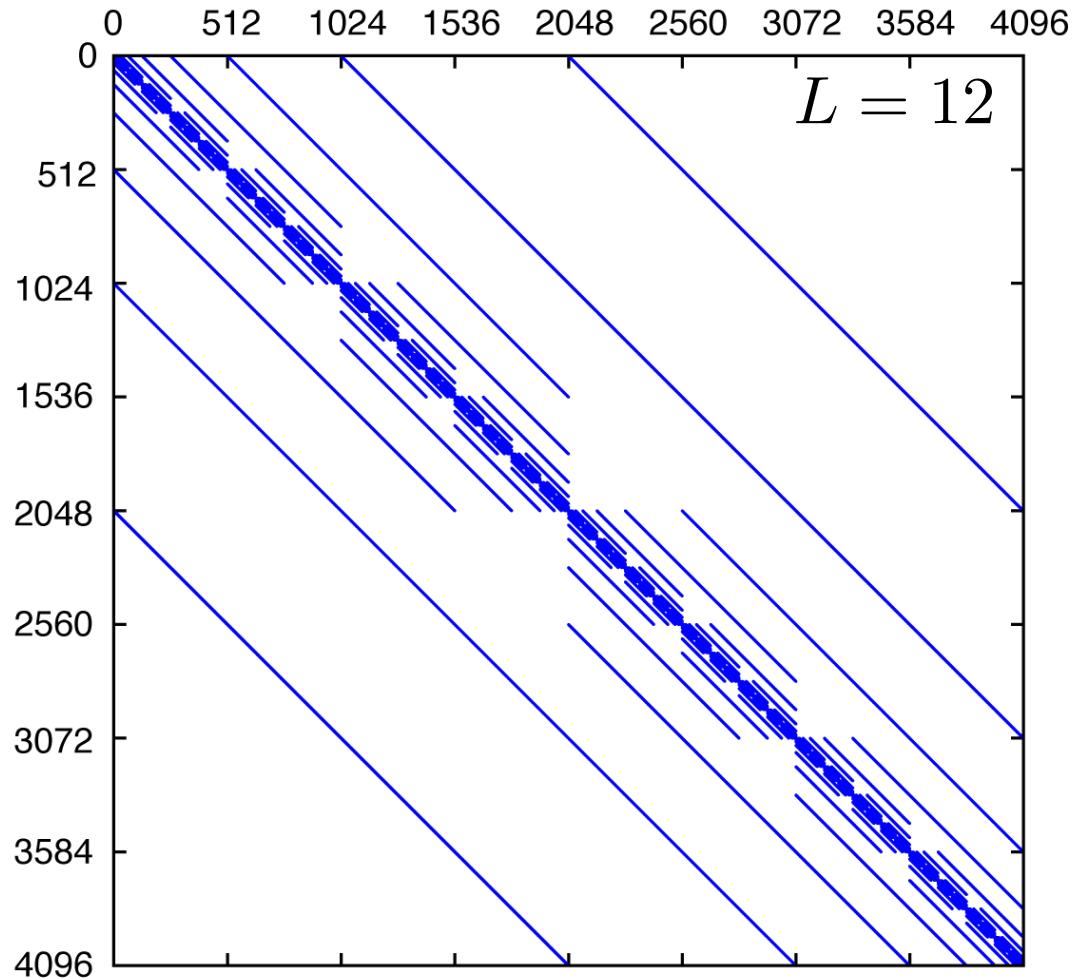
$$\left[ \sum_{i=0}^{L-1} \hat{S}_i^z \hat{S}_{i+1}^z, \sum_{i=0}^{L-1} \hat{S}_i^x \right] \neq 0$$

→ Quantum fluctuations  
or Zero point motion

-Sparse  
# of elements  $\propto O(2^L)$

-Solvable

-Hierarchical matrix?



# Eigenvalue Problems

Diagonalizing Hermitian matrices

Standard approach:

Call LAPACK (Linear Algebra PACKage)

subroutine zheev:

Householder reflection + QR algorithm

→  $O(N^3)$  numerical cost

# Difficulties in Many-Body Problems

- Longer simulation
- Summation over exponentially large  
# of configurations
  - ←Monte Carlo (2nd to 9th lecture), ...
- $O(N^3)$  numerical cost and  $O(N^2)$  memory
  - ←Krylov subspace method (11th to 12th), ...

# Computer

# Supercomputers in UTokyo

## Oakbridge-CX

Intel Xeon Platinum 8280  
1,368 nodes

## Wisteria/BDEC-01

Odyssey: A64FX 7,680 nodes  
Aquarius: Intel Xeon + [NVIDIA A100](#)  
45 nodes

A single CascadeLake: 4.8384 TFLOPS  
A single A64FX: 3.3792 TFLOPS  
A single A100: [19.5 TFLOPS](#)

## Ohtaka (CPU server)

ISSP, UTokyo@Kashiwa

Theoretical Peak: 6,881 TFlop/s  
AMD EPYC 7720 2.0GHz, 64 cores x 2  
Memory: 258GiB  
1,680 nodes



東京大学 物性研究所  
THE INSTITUTE FOR SOLID STATE PHYSICS  
THE UNIVERSITY OF TOKYO

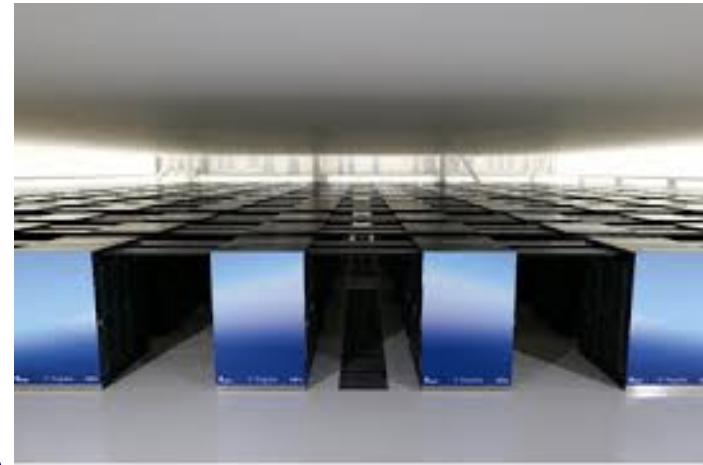
# “Fugaku” Supercomputer

Fugaku 富岳 @RIKEN, Kobe

-158,976 nodes

A64FX

48 cores/node  
32 GiB/node



- 1st in Top500 & 25th in Green500  
on November 2021



# How Computer Handles Numbers

# Numbers in Computers

## Integer

Integer(4): 31bit+1bit

$$-2^{31} \leq j \leq 2^{31} - 1 \quad (j \in \mathbb{Z})$$

## Real number

IEEE Std 754-2008 binary64  
(double-precision floating-point number)

Double precision/real(8):

Sign (1 bit) + Exponent (11 bit) + Significand (52 bit)

$$(-1)^s \times 2^e \times m$$

$$-1022 \leq e \leq 1023 \quad (e \in \mathbb{Z}) \quad m = \frac{\sum_{\ell=0}^{p-1} d_\ell \cdot 2^\ell}{2^p} \quad p = 53$$

# Performance of computer

How many times does the computer multiply/add per second?

How much data does the computer memorize?

How much data does the computer read/write per second?

# FLOP/s

## Floating-point Operations Per Second

An example: Intel Xeon Phi Nights Landing

Intel Xeon Phi 7250 (1.4GHz, 68 cores, 112GB)

- $1.4 \times 10^9$  instructions per second

(instruction to perform double precision add or multiply)

-Intel AVX-512 instruction\*

double precision floating point number (8byte=64bit)

8 double-precision multiply-add\*\* operations

$\rightarrow 1.4 \times 10^9 \times 16 \times 68$  FLOP/s per processor

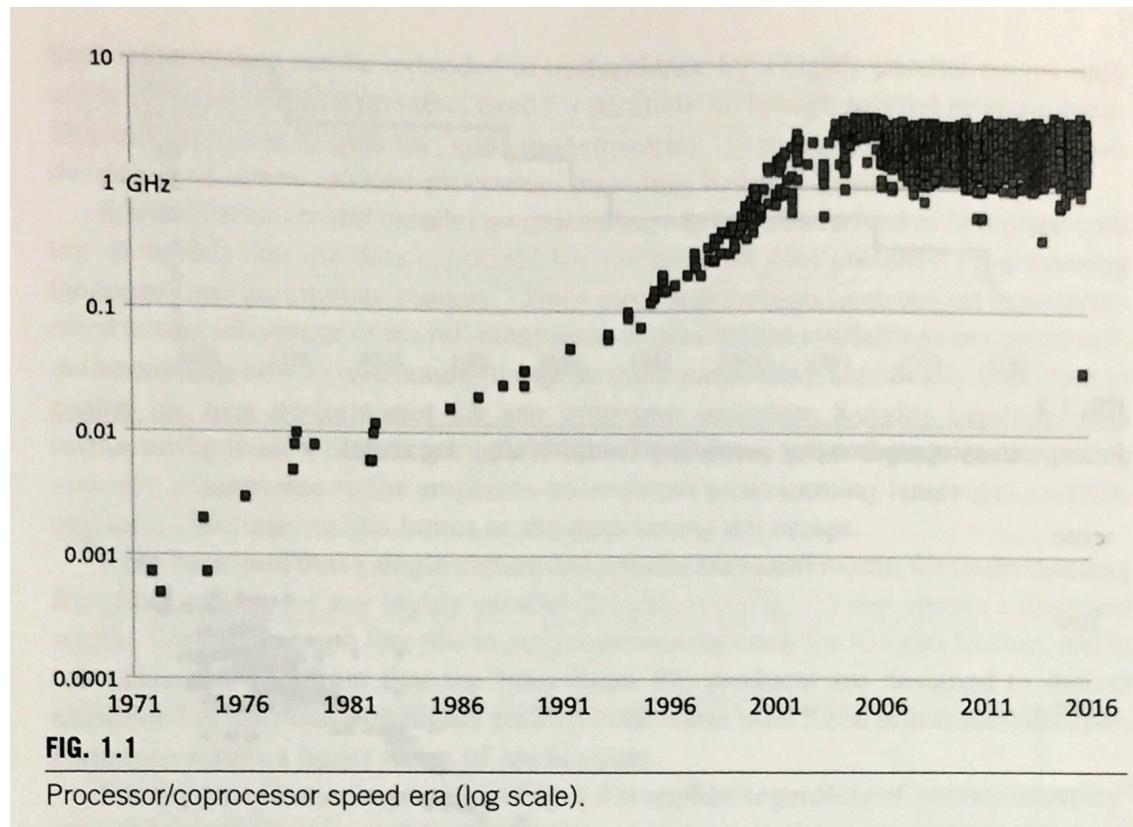
\*Other instruction set:

SVE (for example, A64FX of Fugaku)

\*\*Multiply-add:  $a \leftarrow a + (b \times c)$

# Increasing FLOP/s

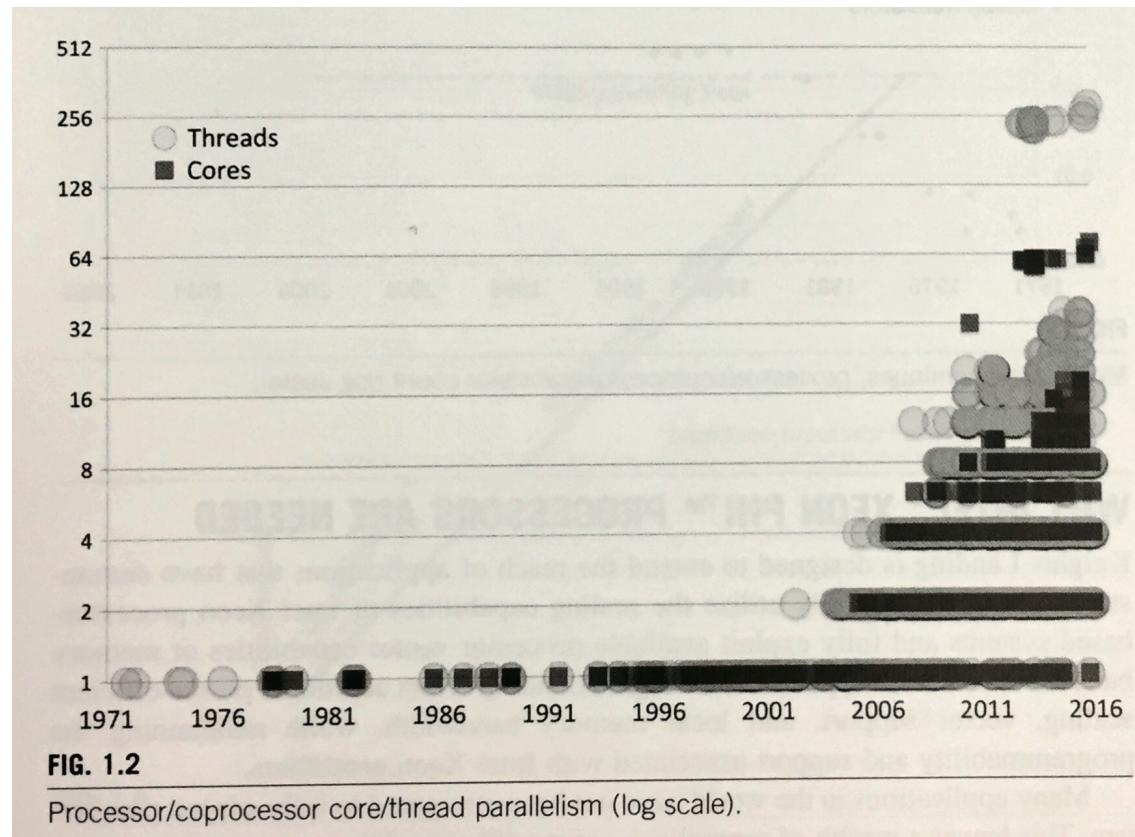
Clock rate saturated



J. Jeffers, J. Reinders, and A. Sodani,  
Intel Xeon Phi Processor High Performance Programming

# Increasing FLOP/s

## Number of cores



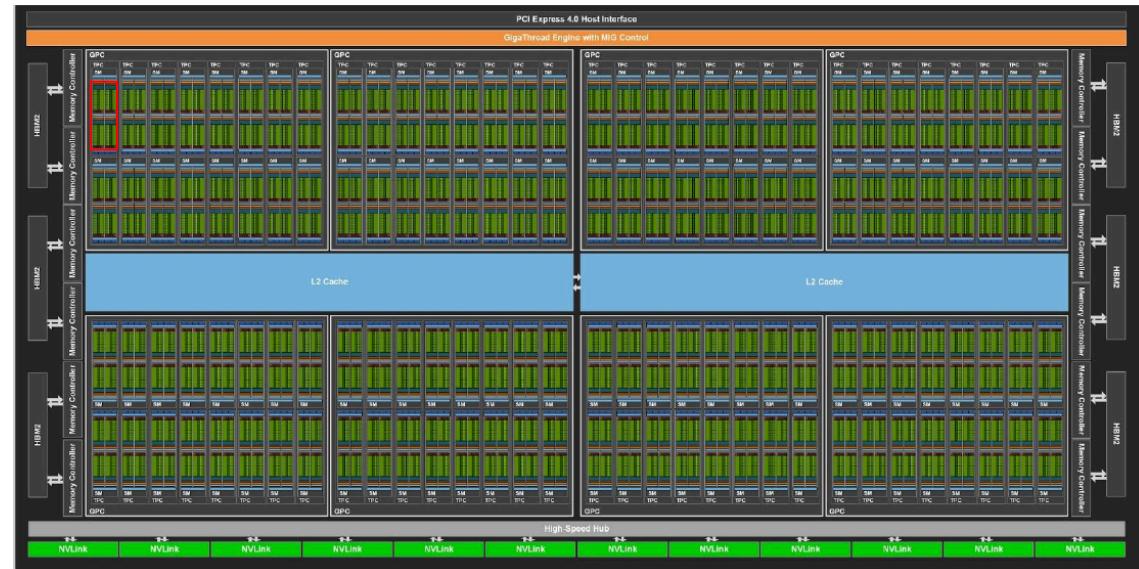
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# Increasing FLOP/s: GPGPU

64 FP32 CUDA cores  
32 FP64 CUDA cores  
in a unit (SM)

NVIDIA Ampere architecture whitepaper

128 SM (108 SM in A100)

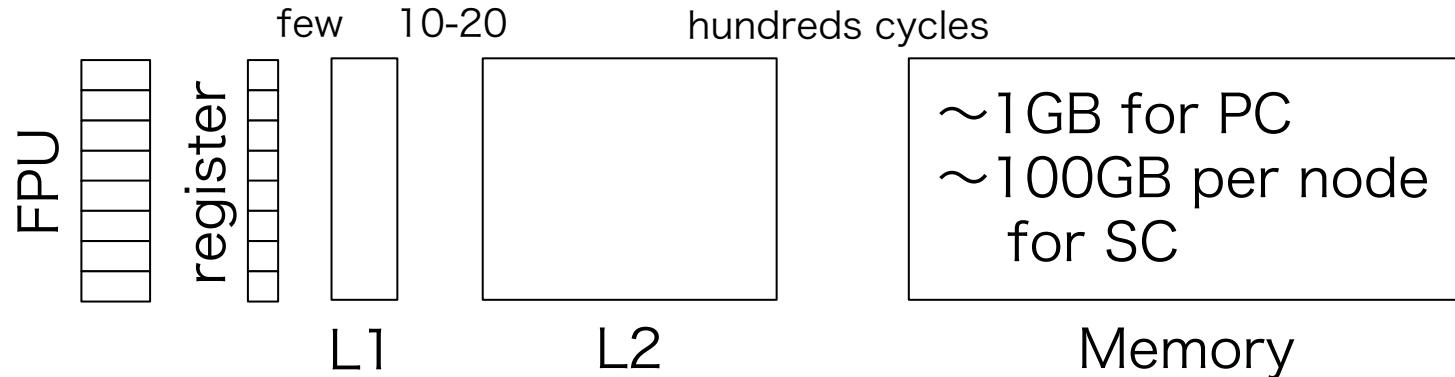


A single CascadeLake: 4.8384 TFLOPS  
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# Cache, Memory, and Disk: CPU



Hierarchy:  
Register  
L1 Cache  
L2 Cache  
Memory  
Disk  
Network

[Intel Xeon Phi 7250](#)

32KB per core  
512KB per core  
~1.6GB per core (~10GB/s per core)

Byte per flop (B/F)

# Computational and Memory Costs

## 1. Vector-vector product

$$\sum_{j=0}^{N_H-1} u_j^* v_j$$

Computational:  $\mathcal{O}(N_H)$

Memory:  $\mathcal{O}(N_H)$

## 2. Matrix-vector product

$$v_i = \sum_{j=0}^{N_H-1} A_{ij} u_j$$

Computational:  $\mathcal{O}(N_H^2)$

Memory:  $\mathcal{O}(N_H^2)$

## 3. Matrix-matrix product

$$C_{ij} = \sum_{k=0}^{N_H-1} A_{ik} B_{kj}$$

Computational:  $\mathcal{O}(N_H^3)$

Memory:  $\mathcal{O}(N_H^2)$

# Exercises (Not report problems)

1. Status of  $N$  qubits is represented as a complex vector in  $2^N$ -dimensional Hilbert space. If you can use whole system of [Ohtaka](#), how many qubits can you store in the memory?
2. Usually, B/F of modern supercomputers is less than 1. Which kind of calculations is suitable for them?

# Method of Evaluation

Based on 2 Reports:

- Exercise about algorithms
- Exercise of computer simulation

## Closely Related Lectures

- 計算科学概論 (for undergraduate)  
Mon. 3rd period
- (archive/seminar) 計算科学科学技術特論A,B,C

# Lecture Schedule

- #1 Many-body problems in physics and why they are hard to solve
- #2 Classical statistical model and numerical simulation
- #3 Classical Monte Carlo method
- #4 Applications of classical Monte Carlo method
- #5 Molecular dynamics and its application
- #6 Extended ensemble method for Monte Carlo methods
- #7 Quantum lattice models and numerical approaches
- #8 Quantum Monte Carlo methods
- #9 Applications of quantum Monte Carlo methods
- #10 Linear algebra of large and sparse matrices for quantum many-body problems
- #11 Krylov subspace methods and their applications to quantum many-body problems
- #12 Large sparse matrices and quantum statistical mechanics
- #13 Parallelization for many-body problems