

Accelerating lattice QCD simulations using multiple GPUs

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1. Introduction to QCD

■ QCD (Quantum Chromodynamics)

○ Describes properties of

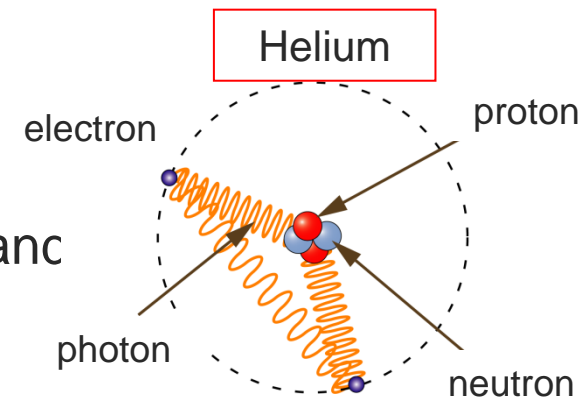
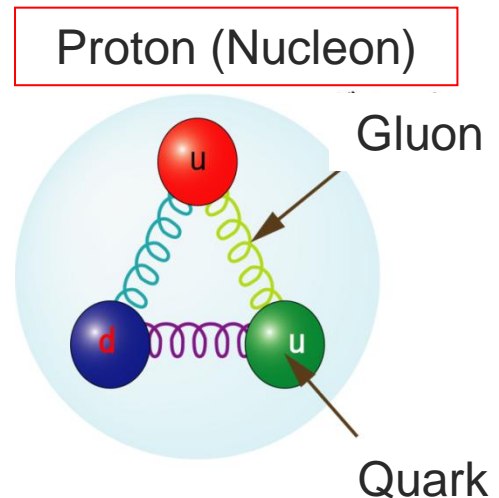
- Nucleons (Proton, Neutron)
- (Mesons, Baryons = Hadrons)
- Strong interaction

○ from more fundamental particles **Quarks and Gluons**

- Dynamics by exchanging gluons.

○ e.g. QED (Quantum Electrodynamics)

- Molecules/Atoms from nucleons/electrons and photons



1. Introduction to QCD (contd')

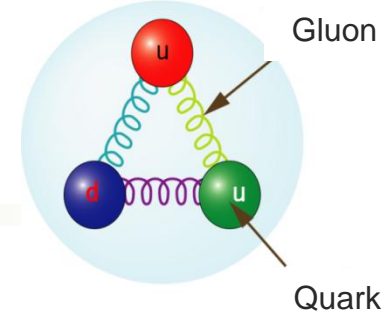
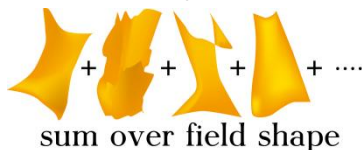
Proton (Nucleon)

■ QCD (Quantum Chromodynamics)

○ Quantum Field theory (QFT)

- Quark Field and Gluon Field with space-time index. They have color charge (based on SU(3)).
(Electric charge based on U(1))
 - Gluon action : similar to Maxwell's equation (photon action), Gluon carries its color charge.
(photon does not have electric charge.)
 - Quark action: similar to electron's equation (Dirac's equation/action)
Quark carries color charge. (electron has electric charge)
- Quantized by **Feynman's path-integral**. Partition function with action (similar to statistical physics).

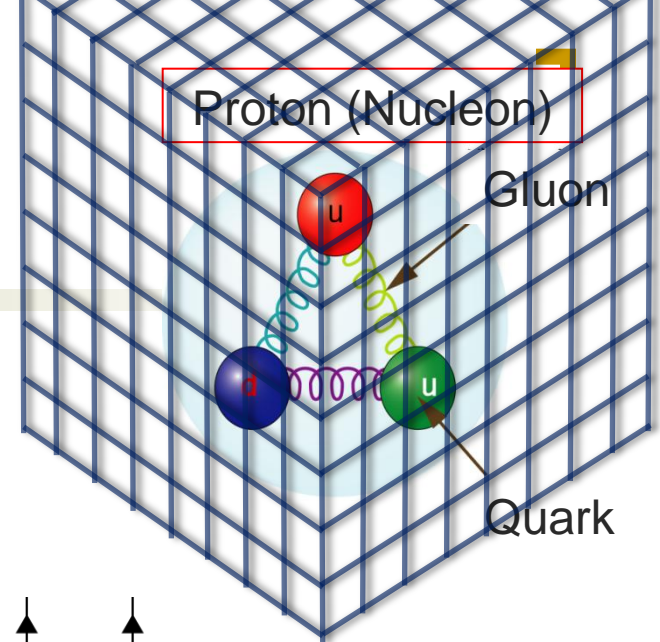
$$Z = \int D A D \bar{\Psi} D \Psi \exp \left[\int d^4 x \left(-S_{gluon}(A) - S_{quark}(A, \bar{\Psi}, \Psi) \right) \right]$$



[2. Lattice QCD

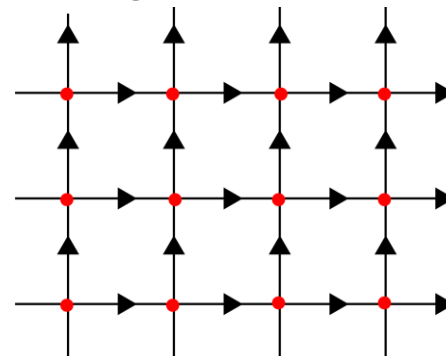
■ Lattice QCD [K.G.Wilson (1974)]

- 4D Space Time => 4D Lattice Box
- Fields on Discretized Space-Time



Quark field
 $\psi(x)$

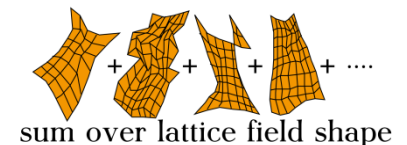
Gluon field
 $A_\mu(x)$



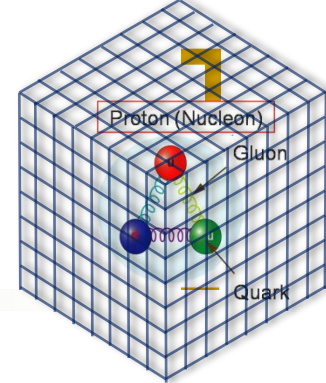
• Quark field $\psi(n)$ \longrightarrow Gluon field $U_\mu(n)$

- Integration on Field Shape => Integration on many var's.

$$\int D A D \bar{\Psi} D \Psi \quad \longrightarrow \quad \int \prod_{n,\mu} dU_\mu(n) \prod_n d\bar{\Psi}(n) d\Psi(n)$$



2. Lattice QCD (cont'd)



Lattice QCD [K.G.Wilson (1974)]

- LQCD Partition function

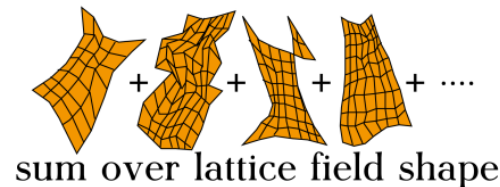
$S(U, \bar{\psi}, \psi)$: Discretized version of QCD action

$$Z = \int \prod dU d\bar{\psi} d\psi \exp[-S(U, \bar{\psi}, \psi)]$$

$$= \int \prod dU \exp[-S_{\text{eff}}(U)]$$

analytic integration on Grassmann vars $\bar{\psi}, \psi$.

- Integration on many var's.



- Observable: O

$$\langle O \rangle = \frac{1}{Z} \int \prod dU O(U) e^{-S_{\text{eff}}(U)}$$

Computation of hadron masses, etc...

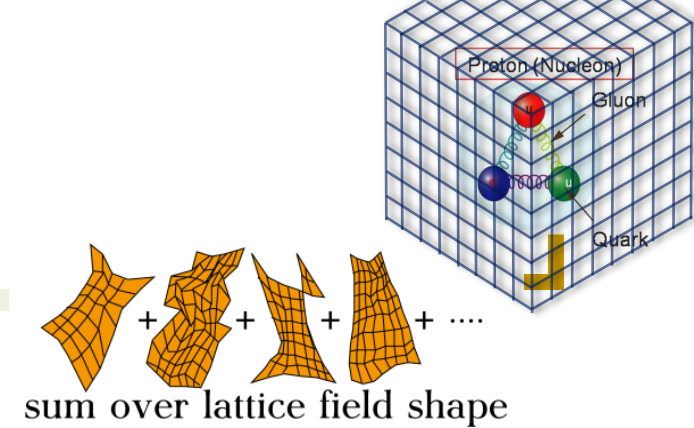
- Similar to Statistical Physics

Monte Carlo Integration
with Supercomputer!!

2. Lattice QCD (cont'd)

Lattice QCD [K.G.Wilson (1974)]

- Monte Carlo Importance Sampling



$$Z = \int \prod dU e^{-S_{\text{eff}}(U)} \quad \langle O \rangle = \frac{1}{Z} \int \prod dU O(U) e^{-S_{\text{eff}}(U)}$$

- Markov chain Monte Carlo to generate sequence of U

$$U^{(1)} \rightarrow U^{(2)} \rightarrow U^{(3)} \rightarrow \dots \rightarrow U^{(j)} \rightarrow \dots \quad \text{Prob}[U] \propto e^{-S_{\text{eff}}(U)}$$

- Statistical Average

$$\langle O \rangle = \frac{1}{N_{\text{sample}}} \sum_{j=1}^{N_{\text{sample}}} O(U^{(j)}) \quad (N_{\text{sample}} \rightarrow \infty)$$

- To generate the Markov chain of U (gluon field)

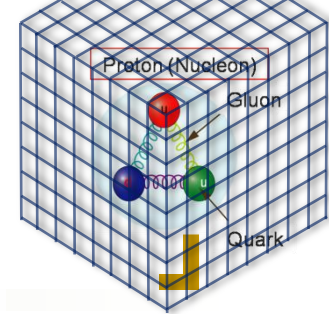
Hybrid Monte Carlo (HMC) algorithm

[Duane, Kennedy, Pendleton, Roweth(1987)]

is usually employed.

2. Lattice QCD (cont'd)

Lattice QCD



HMC algorithm [Duane, Kennedy, Pendleton, Roweth(1987)]

- The most time consuming part of the HMC algorithm is the inversion of quark matrix $D[U]$.

$$e^{-S_{eff}(U)} \Rightarrow D\phi^\dagger D\phi \exp\left[-S_{gluon}(U) - \left(D[U]\right)^{-1}\phi\right]$$

- Molecular Dynamics (MD) evolution is used to generate a sequence of U (Markov chain) in the HMC. This requires huge number of solution of the *linear equations*.

$$D[U]x = y \Rightarrow y = \left(D[U]\right)^{-1}x \quad D: \text{size} > 10^6 \times 10^6$$

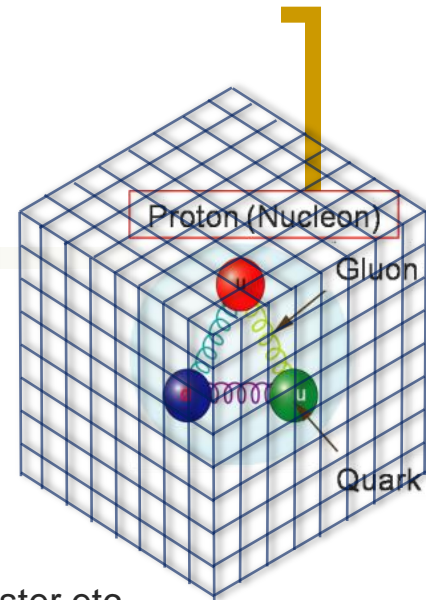
- To get better statistics, we need $O(100)$ configurations of U . This needs $O(1000)$ the HMC cycles. In each HMC cycle, MD needs $O(100)$ time steps. At each time step we need two inversion of $D[U]$. In total we have to invert $D[U]$ by $2 \times 100 \times 1000 \times 100 = O(10^7)$ times!.
- Speeding up the large scale linear equations solver is the key of LQCD simulations.

2. Lattice QCD (cont'd)

Lattice QCD

Hadron masses from LQCD (~2010).

- Proton size = 1~2fm (experimental)
- Typical Lattice size $L = 16 \sim 32$
- Typical Lattice spacing $a = 0.05 \sim 0.1$ fm
- Typical Lattice extent $La = 2 \sim 5$ fm
- Machines:
 - Supercomputers, BG/L/P, SX, SR, VPP.. , Custom made PC cluster etc.
 - Uses O(10) TFlops machines for several years.



QCDOC@BNL, ColumbiaU, UKQCD

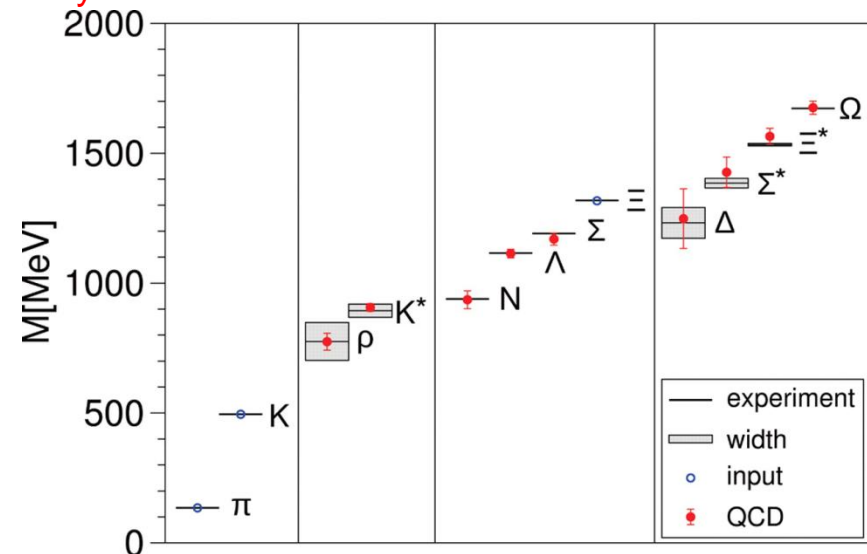
Jugene(BG/P)@JSC

BG/L@KEK

TachyonII@KISTI

PACS-CS
@CCS, Tsukuba

etc.....



**Durr et al. (Budapest-Marseille-Wuppertal collab.
"Ab Initio Determination of Light Hadron Masses",
Science 322, 1224(2008).**

[3. LQCD with GPU/Accelerator]

- For all computation in HMC, Parallel Supercomputers are usually employed, where 4D space-time is domain-decomposed and the task is distributed to each node.
- (O(10)Tflops x 2-3 years) machine can now well reproduce single hadron in the computer.
- But such machines are still too expensive... for a novice researcher or to do a trial computer experiment
- Improvements on the algorithm and computer arch.s are highly desired.
- GPGPU and Accelerator are the candidate to get more efficient and economical machines for LQCD.
- Speeding up the quark matrix linear solver $D[U]x=b$.

■ LQCD experience on GPGPU/Accelerator

○ CELL B.E. (PlayStation3)

Spary, Hill, Trew hep-lat/0804.3654;

S. Motoki & A. Nakamura Lat2007;

F. Belletti et al. LAT2007

A. Nobile et al, DESY group,
QPACE project (**Q**CD **P**arallel
computing on the **CELL**/B.E.)

For details of Lattice QFT on CELL .B.E.
see Next talk by Motoki.



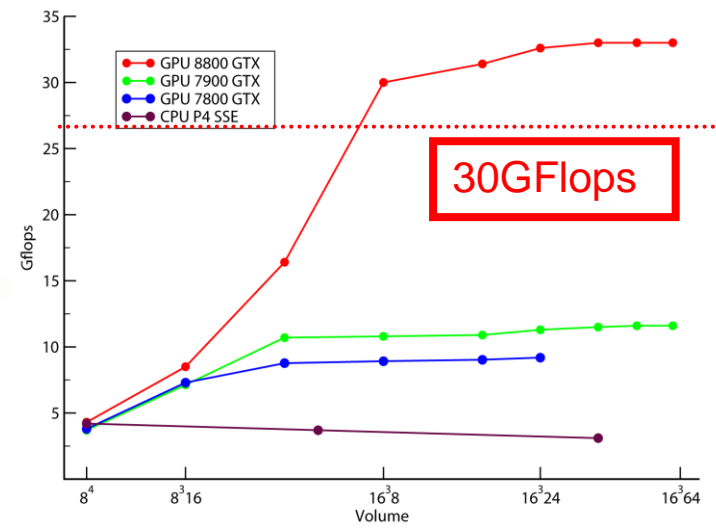
3. LQCD with GPU/Accelerator (cont'd)

GPGPU

“Lattice QCD as a video game”,
G.I.Egri, Z.Fodor, S.D.Katz, D.Nogradi,
K.K.Szabo, (2006)hep-lat/0611022.

- NVIDIA G80 arch. > 300 GFlops(SP)
- Lattice Wilson kernel > 30 GFlops
- Difficult to program using Graphic API

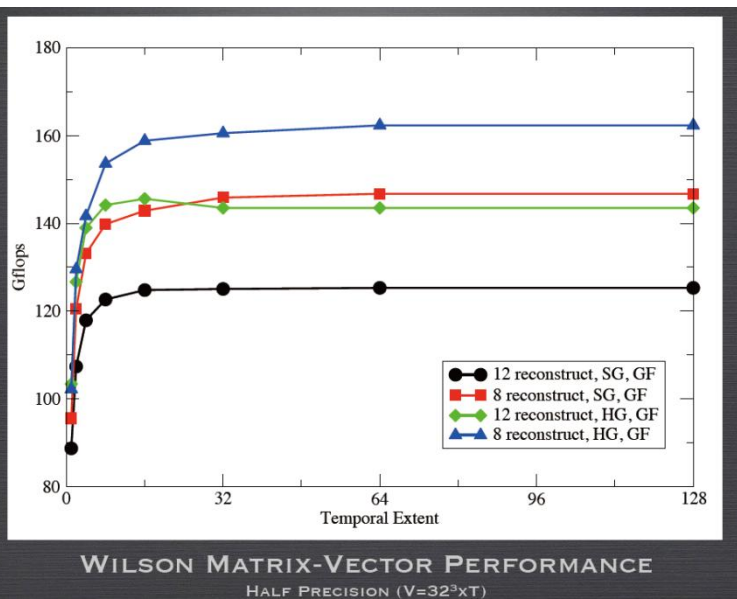
(OpenGL)



Quark solver speed (2006)

- CUDA version (2008~) with new GPU G200 arch.

- Now CUDA (a C/C++ simple extension) is available.
- Easy to learn, but requires hardware/memory model knowledge



- C. Rebbi, “Blastigng Through Lattice Calc. using CUDA”, Lat08.
- F. Di Renzo, “GPU computing for 2-d spin systems: CUDA vs OpenGL”, Lat08
- And many studies....

M. Clark et al. Work shop@CCS,
10-12 March, 2009
They got **140 Gflops** (SP) for
Wilson-D[U] computation using
Nvidia GTX280.

GPGPU

- AMD GPU, Firestream, OpenCL
- applications to LQCD
 - “Pseudo-random number generators for Monte Carlo simulations on Graphics Processing Units”, V.Demchik, hep-lat/1003.1898.
 - “Monte Carlo simulations on Graphics Processing Units”, V.Demchik and A.Strelchenko, hep-lat/0903.3053v2.
They employed CAL of ATI Stream.
- For more detailed GPGPU usage in LQCD application, see,
“QCD on GPUs: cost effective supercomputing”
by M.A. Clark , Lattice2009, PoS(LAT2009)003
[hep-lat/0912.2268].



GPU acceleration LQCD Kernel $D[U]$

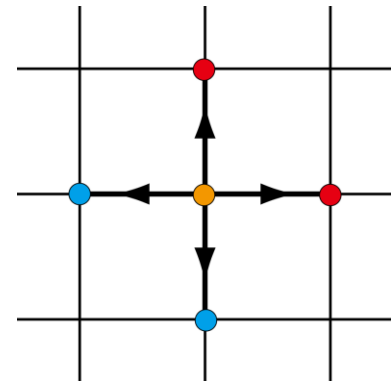
- There are several discretization scheme for Dirac equation D . (D : 4D 1st order diff. eq.)
- Wilson-Dirac discretization is the most generic one and used widely. And this has an important kernel called Hopping matrix, which is common to other discretization.
- Improvement on the Hopping kernel is very important.

Hopping kernel (Hopping matrix)

$$M(n, m) = \sum_{\mu=1}^4 \left[(1 - \gamma_{\mu}) \otimes U_{\mu}(n) \otimes \delta_{n+\hat{\mu}, m} + (1 + \gamma_{\mu}) \otimes U_{\mu}^{\dagger}(m) \otimes \delta_{n-\hat{\mu}, m} \right]$$

$$D_W[U](n, m) = 1 \otimes 1 \otimes \delta_{n, m} - \kappa M(n, m)$$

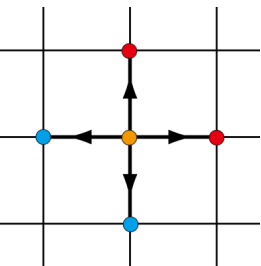
- 4D, first-difference operator.
- sites connected by gluon U , by which quark-color are mixed.
- quark-spin are mixed by Dirac's gamma matrix γ_{μ}
- Memory bandwidth intensive operation.
 - Byte/Flop ~ 3 B (D.P.), ~ 1.5 B (S.P.)



- GPU acceleration LQCD Kernel $D[U]$
 - My personal experience on the CUDA programming for the Hopping matrix (single GPU)

Hopping kernel (Hopping matrix)

$$M(n, m) = \sum_{\mu=1}^4 \left[(1 - \gamma_{\mu}) \otimes U_{\mu}(n) \otimes \delta_{n+\hat{\mu}, m} + (1 + \gamma_{\mu}) \otimes U_{\mu}^{\dagger}(m) \otimes \delta_{n-\hat{\mu}, m} \right]$$



$$D_W[U](n, m) = 1 \otimes 1 \otimes \delta_{n, m} - \kappa M(n, m)$$

- We want to solve $D_W[U]x = y$ for x with given y and U .
- Lin. eq. with large sparse matrix \Rightarrow Iterative solver (CG, BiCGStab etc..)
- CUDA/GPU: Single precision is very fast (Tflops), but we need double precision solution x .
- We use the mixed precision iterative solver (generalization of iterative refinement/Richardson iteration). This guarantees double precision accuracy with almost S.P. arithmetics.

GPU acceleration LQCD Kernel $D[U]$

- **Mixed precision solver (strategy)** [early proposal by Buttari, Dongarra, Langou, Langou, Luszczek, Kurzak (2007)]

- To solve $Dx = b$.

(0) [given r and x satisfy $r = b - Dx$. (double prec.)]

(1) [Solve $Dv = r$ in single precision]

(2) $q = Dv$ [double prec.]

(3) $x = x + v$ [double prec.]

(4) $r = r - q$ [double prec.]

[new r and x still satisfy $r = b - Dx$.]

(5) [Check $|r|$ and goto (1)]

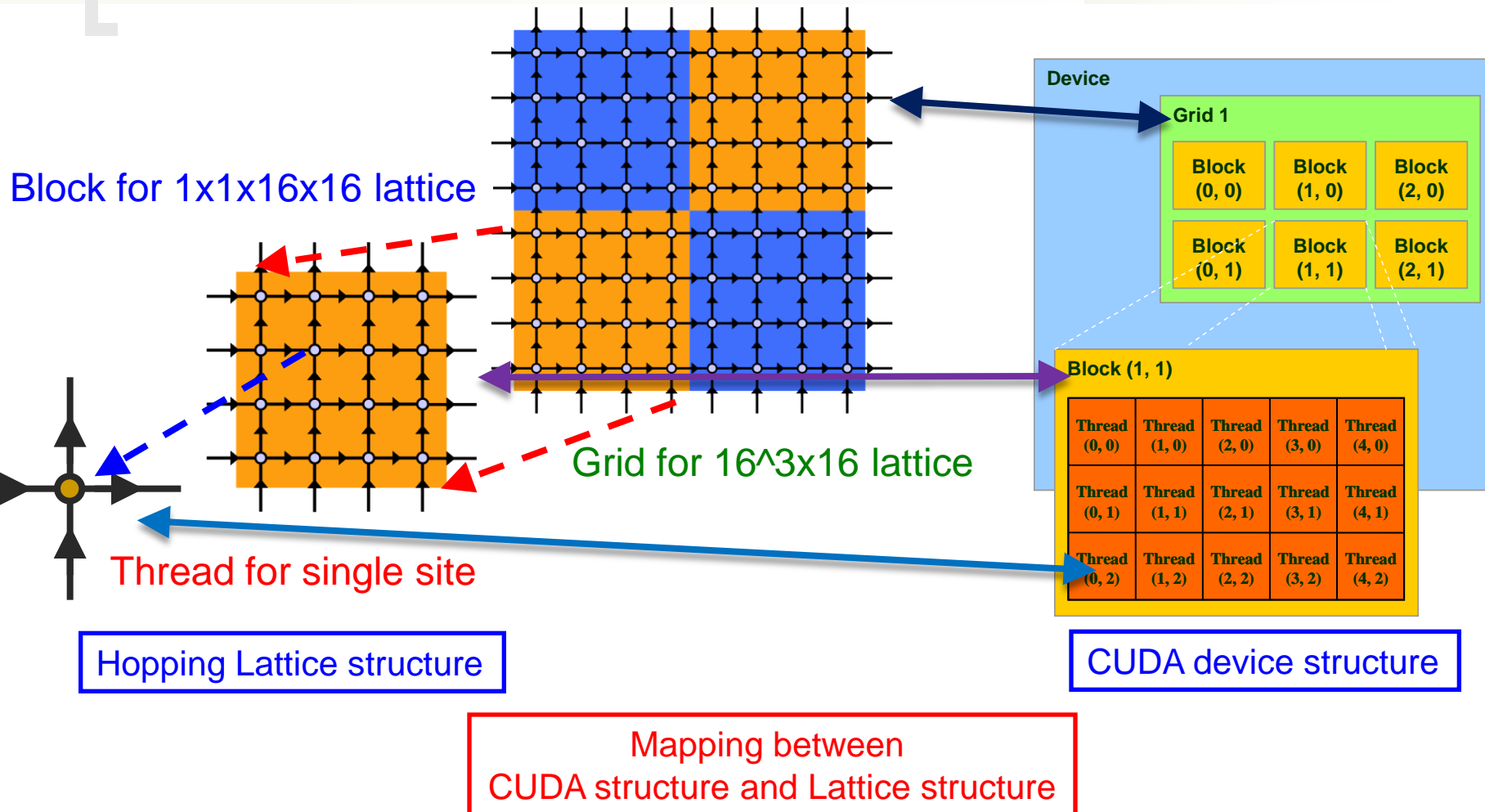
GPU task
CUDA code

CPU task
HOST code

- The iterative refinement technique with full single precision solver (10^{-7}) can solve full double precision (10^{-14}) solution within 3-5 refinement iterations.
- Most computing time is spent in the S.P. solver.
- GPU is employed for Full single precision solver.

GPU acceleration LQCD Kernel $D[U]$

- Hopping kernel CUDA code



GPU acceleration LQCD Kernel $D[U]$

Hopping kernel CUDA code

Memory structure (data ordering)

- CUDA requires appropriate data ordering to get efficient memory bandwidth (like vector processor).
- We have to organize quark/gluon field component (spin and color) appropriately to match with CUDA memory alignment.
- Gluon field on single site in a direction, U , has 3×3 complex elements. $(\text{real} \times 4) \times 18 = 72$ bytes. (not match 16 bytes alignment) We make use of $SU(3)$ property of U to reduce data size. 3×2 complex elements are sufficient to reconstruct 3×3 U . $(12 \text{ real} \times 4) = 48$ bytes matches 16 bytes alignment.
- We use $y(3,4,T,Z,Y,X)$ (in Fortran array form) for quark field.
- Complex-Color-Spin $(3,4)$ indexes are encoded to float4 array.
- A part of TZ plane is assigned to CUDA thread parallelization. Remaining TZ,Y,X indexes are assigned to CUDA block parallelization.
- Texture fetching is used to read memory (to cache data y and U)

My understanding on the CUDA programming for LQCD application is:

- The CUDA thread parallelization is similar to vectorization on vector machines.
- The block parallelization is similar to usual parallelization as MPI.
- Calling the CUDA kernel is similar to MPI job submitting.
- This strategy is very familiar to lattice guys....

Various tuning tips using CUDA for LQCD!

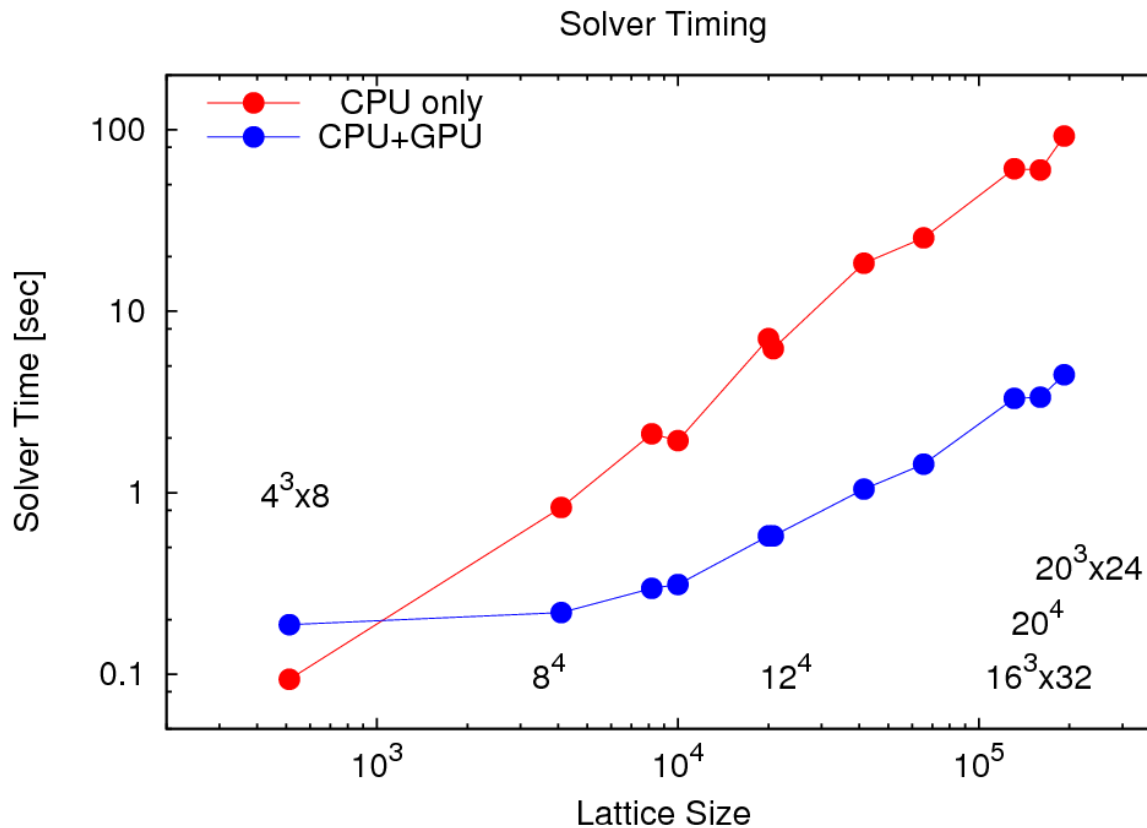
4. Accelerating $D[U]x=b$ solver using single GPU.

- Some results [K.-I.I. and Y.Osaki, study in 2008]
 - CPU: Core2Duo@3GHz, GPU: GeForce GTX 280, CentOS.
 - O(a)-improved Wilson-Dirac Fermion
 - Red/black site prec'd, Nested-BiCGStab (mixed prec. solver)
 - Random gluon field U .
 - Programming language
 - HOST: Fortran90, BiCGStab, BiCGStab calls single precision BiCGStab(GPU) as a preconditioner.
 - GPU: CUDA and C/C++. Single precision BiCGStab.
 - Residual history, performance, Lattice Volume (space-time) dependence etc.

4. Accelerating $D[U]x=b$ solver using single GPU.

Volume dependence of performance

○ CPU(DP) only vs CPU(DP)+GPU(SP)



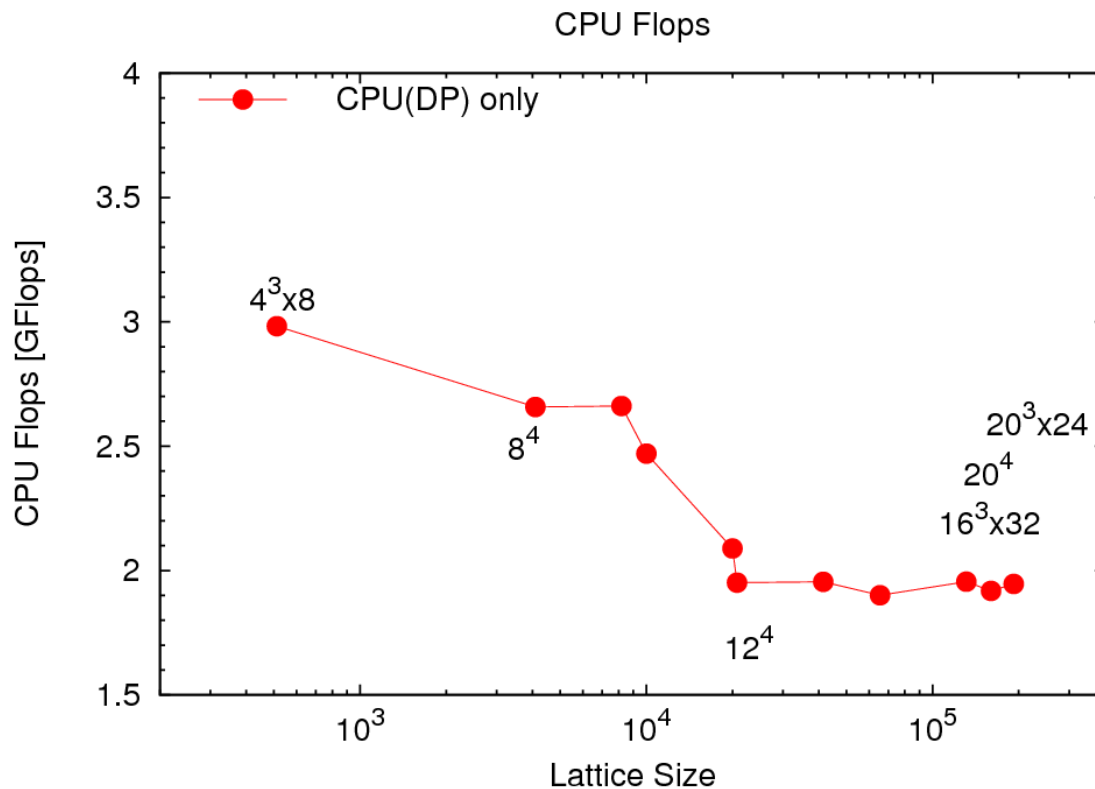
Small volume is not effective for GPU because parallelism is less than num of GPU cores(maximum threads).

For larger lattice we can achieve ~10x~20x speed up.

4. Accelerating $D[U]x=b$ solver using single GPU.

Volume dependence of performance

○ CPU performance (GFlops)



CPU off-cache performance is about 2GFlops (D.P.).

Hopping kernel requires about 3bytes/flop.

Effective memory bandwidth is about 6GByte/s.

[Stream benchmark, Triad = 7.7GB/s (meas'd)]

Lattice QCD is bandwidth intensive application.

4. Accelerating $D[U]x=b$ solver using single GPU.

Volume dependence of performance

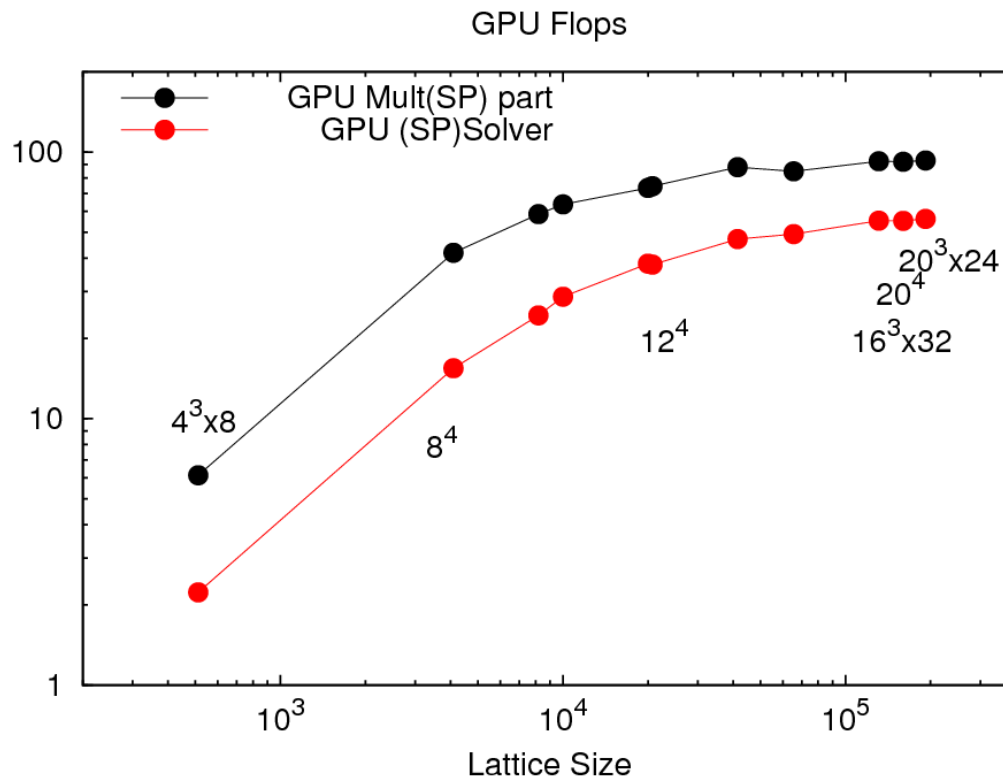
GPU performance (GFlops(SP))

GPU performance reaches about 100 GFlops (S.P.).

Hopping kernel requires about 1.4bytes/flop.

Effective memory bandwidth is about 140GByte/s!!
[Bandwidth test : 115GB/s]

Texture fetching and high memory bandwidth is very important to achieve this performance.



To get good efficiency large volume should be assigned to single GPU.

4. Accelerating $D[U]x=b$ solver using single GPU.

■ Single GPU performance in LQCD.

- We found 10x-20x speed up using single GPU.
- 16^4 lattice is enough for experimental simulations or other lattice simulations.
- But this is not our final goal because 16^4 is still small volume. (32^4 or larger is wanted).
- To enlarge lattice size, multiple GPU or parallel GPU usage is required.

■ Application class:

(1) Multiple GPUs in a single node.

for other kind of lattice simulations

(2) Multiple GPU, Multiple nodes

for O(10)Tflops machine for 32^4 lattice

5. Many flavor lattice QCD with multiple GPUs on single node

- Schrodinger Functional (SF) simulation with $N_f=10$ QCD.
 - Class (1) Multiple GPUs in a single node application.
 - $N_f=10$ QCD as a Technicolor model beyond Standard Model of elementary particles.
 - 16^4 lattice is enough. Parameter searching type simulation.
 - Simulation cost \propto to number of quarks (10 quarks $\Leftrightarrow N_f=10$) (Proton/neutron : $N_f=2$ or 3 QCD)
 - The Technicolor model is not yet established experimentally. Numerical simulation helps the validation from theoretical side. But simulation cost is too high!! , and most computer time is given to normal $N_f=2$ QCD simulations.....
 - GPU acceleration can help this situation.

5. Many flavor lattice QCD with multiple GPUs on single node

Schrodinger Functional (SF) simulation with $N_f=10$ QCD.

- We would like to simulate $N_f=10$ QCD more economically with GPU but
 - Simulation cost \propto to number of quarks (10 quarks $\Leftrightarrow N_f=10$) (Proton/neutron : $N_f=2$ or 3 QCD). 5x heavy for $N_f=10$ than $N_f=2$.
 - 10 quarks are independent each others in the HMC algorithm. Solver for each quark can be done independently.
 - Task can be distributed to each GPU (GPU = one quark) in single node.

- We tested $N_f=10$ SF QCD using 2GPUs in a node.

5. Many flavor lattice QCD with multiple GPUs on single node

Schrodinger Functional (SF) simulation with $N_f=10$ QCD.

N.Yamada, M.Hayakawa, K.-I.I., Y. Osaki, S.Takeda, S.Uno.
arXiv:1003.3288, arXiv:0910.4218(Lat09).

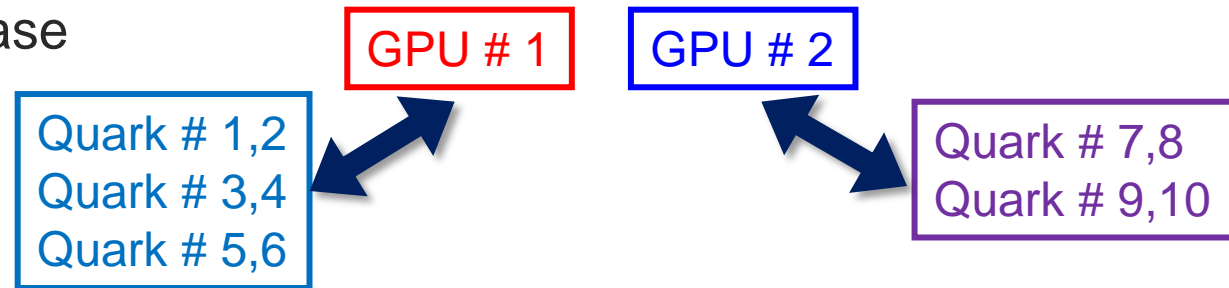
- Test on
 - CPU: Core i7 920 (2.67GHz, 4 core)
 - DP: 43 GFlops, SP: 85 GFlops (peak)
 - 2xGPUs: 2 x (Nvidia GeForce GTX 285)
 - 240 core @ 1.48 GHz, SP: 710 GFlops (peak)
 - OS, compiler
 - CentOS 5.2 (Linux), Intel Fortran / C++, CUDA 2.3.
 - Cost, BTO one PC box + 2GPU cards + Intel compiler.

5. Many flavor lattice QCD with multiple GPUs on single node

Schrodinger Functional (SF) simulation with $N_f=10$ QCD.

Quark assignment (Using blocked BiCGStab solver algorithm)

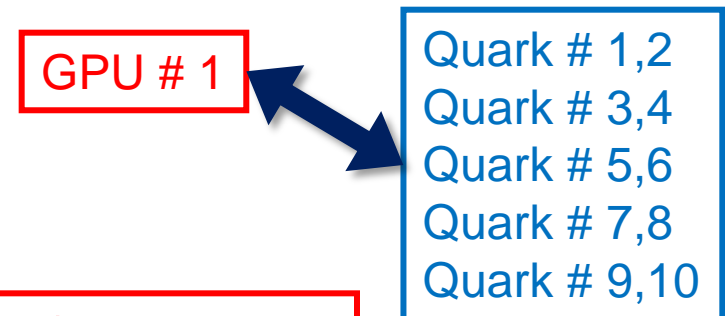
2GPU case



- GPU#1 = 3 solver calls, GPU#2 = 2 solver calls, in a single MD step.

1GPU case

- GPU#1 = 5 solver calls.



Ideally 40 % ($3/5 = 0.6$) Timing improvement is expected.

Quark pair is always assigned to single external scalar field.

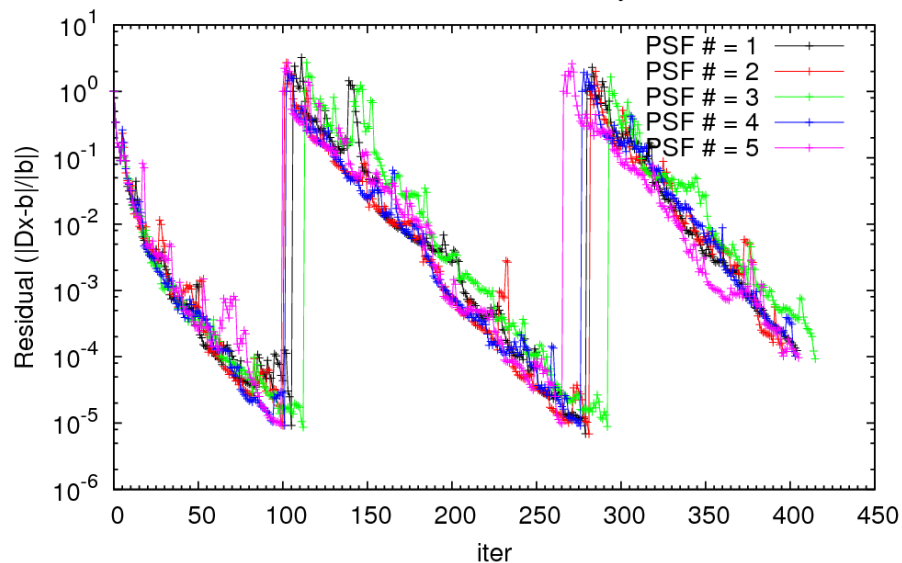
5. Many flavor lattice QCD with multiple GPUs on single node

16^4 , $\beta = 4.52$, $\kappa = 0.15805$, $N_F = 10$

$g_{SF}^2 \approx 10$, $aM_{PCAC} \approx 0.001$

Effect of Multiple GPUs (1GPU \Rightarrow 2GPU)

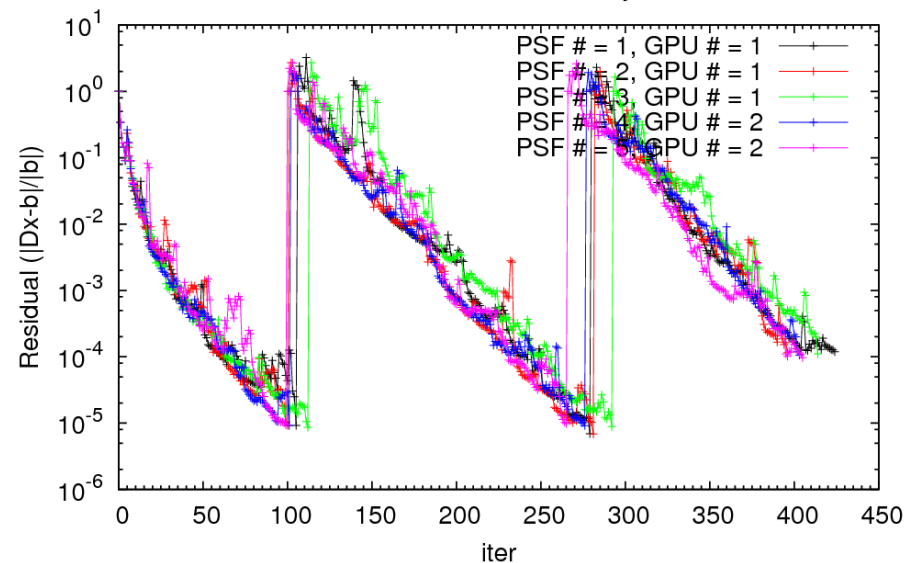
Global Blocked DP-BiCGStab + Normal 1-GPU SP-BiCGStab
Solver Residual History



GPU solver is called 5 times in
a CPU solver iteration.

Full timing : 4.06 sec

Global Blocked DP-BiCGStab + Normal 2-GPU SP-BiCGStab
Solver Residual History



GPU solver is called 3 times for
GPU#1 and 2 times for GPU#2.

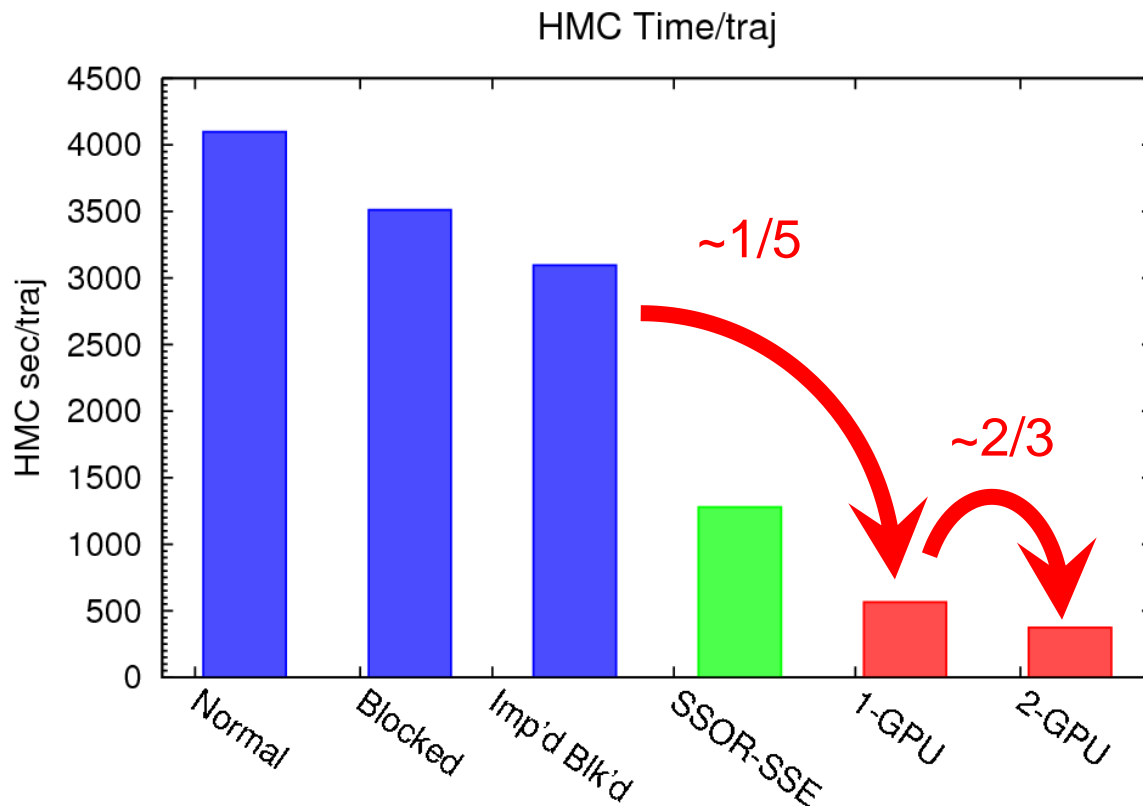
Full Timing : 2.69 sec

We observed 34% reduction in timing.
(Ideally it should be $3/5 = 0.6 = 40\%$ reduction)

5. Many flavor lattice QCD with multiple GPUs on single node

$$16^4, \beta = 4.52, \kappa = 0.15805, N_F = 10$$
$$g_{\text{SF}}^2 \approx 10, aM_{\text{PCAC}} \approx 0.001$$

- Timing comparison for single HMC step.



I omitted the details of CPU side improvements with multi-core and mixed precision preconditioning.

We accomplished a factor 2-3 improvement for CPU only. But this does not win GPU.

GPU is still faster than CPU code.

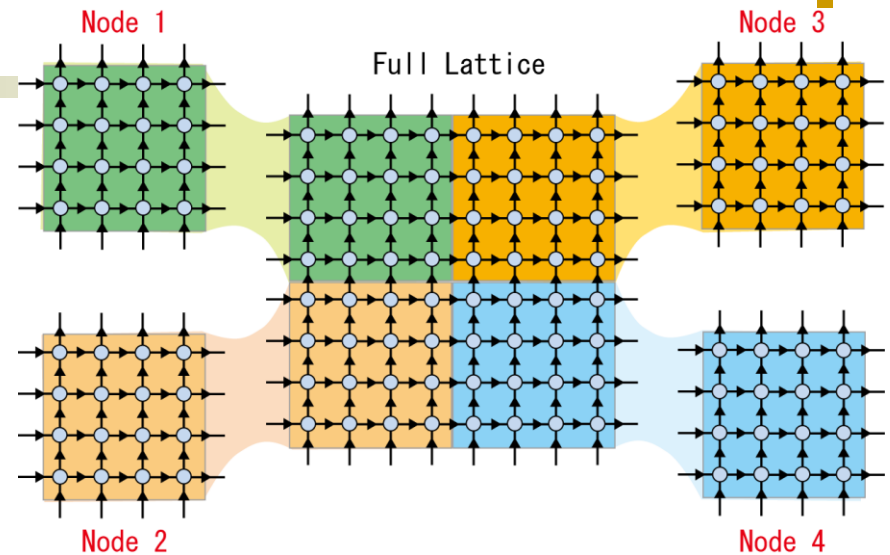
Using 2 GPUs we can accelerate (reduce 34%) in the timing.

[6. Towards parallel GPU computation]

- To catch up the most advanced LQCD simulation we need true parallel GPU computation. $O(10)$ Tflops machine for 32^4 lattice.
- Like in the parallel computation, the communication overhead is important.
- Unfortunately, GPU \leftrightarrow GPU direct communication device is not available in COTS GPU cards.
- Three steps, GPU \leftrightarrow CPU \leftrightarrow CPU \leftrightarrow GPU, communication is required in general.
- We investigated this overhead for LQCD application ($D[U]x=b$ solver).

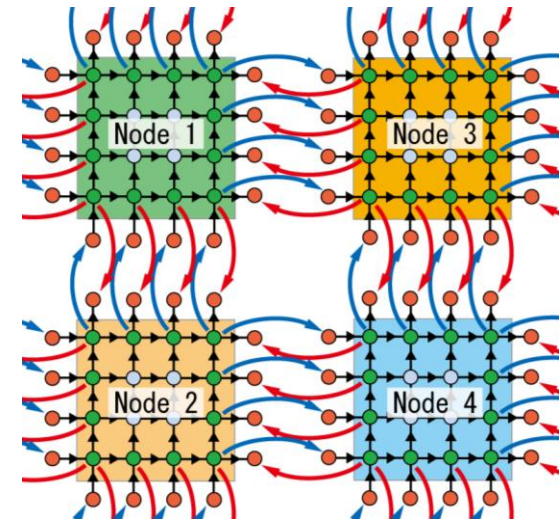
LQCD parallelization

- Data distribution:



- Data communication:

- Exchange edge site data
- Communication hiding behind internal site computation.
- Edge site computation after receiving the adjacent site data from next GPUs.



■ Parallel GPU implementation

○ Some results [K.-I.I. and Y.Osaki]

- Machine: 4PC's with
- CPU : Intel Core i7 920@2.67GHz (4cores)
- GPU : GeForce GTX 285 × 2 (2GPUs in 1 node)
- Memory 6GBytes
- LAN Adapter : Intel Gigabit ET Quad Port Server. Poor man's network. (COTS, cheap)
- CentOS 5.4, CUDA 2.3, OpenMPI
- TCP/IP is too slow. Instead We employed Open-MX (Myrinet Express over Generic Ethernet Hardware) library [<http://open-mx.gforge.inria.fr/>].

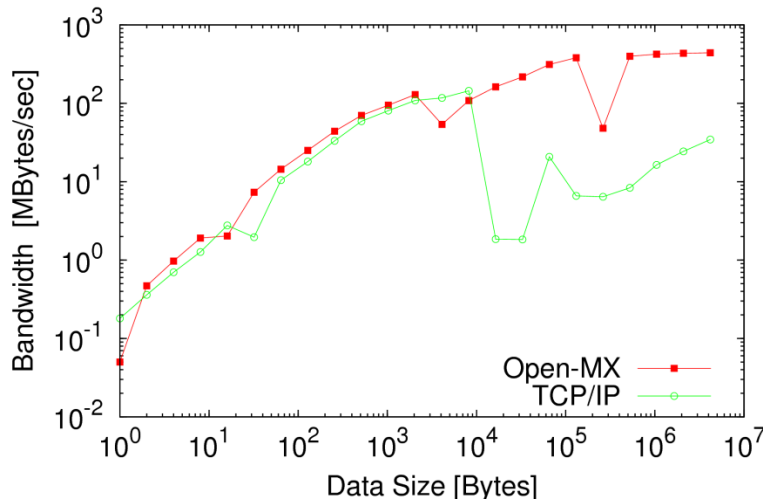
6. Towards parallel GPU computation

Parallel GPU implementation

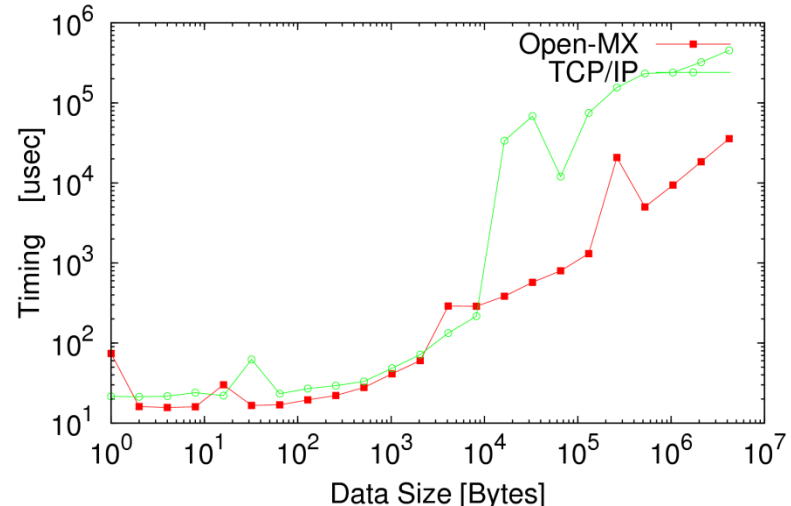
Some results [K.-I.I. and Y.Osaki]

- TCP/IP is too slow. Instead We employed Open-MX (Myrinet Express over Generic Ethernet Hardware) library [<http://open-mx.gforge.inria.fr/>].
- Network performance.

Bandwidth in Data Exchange (SendRecv, 4nodes).



Timing in Data Exchange (SendRecv, 4nodes).

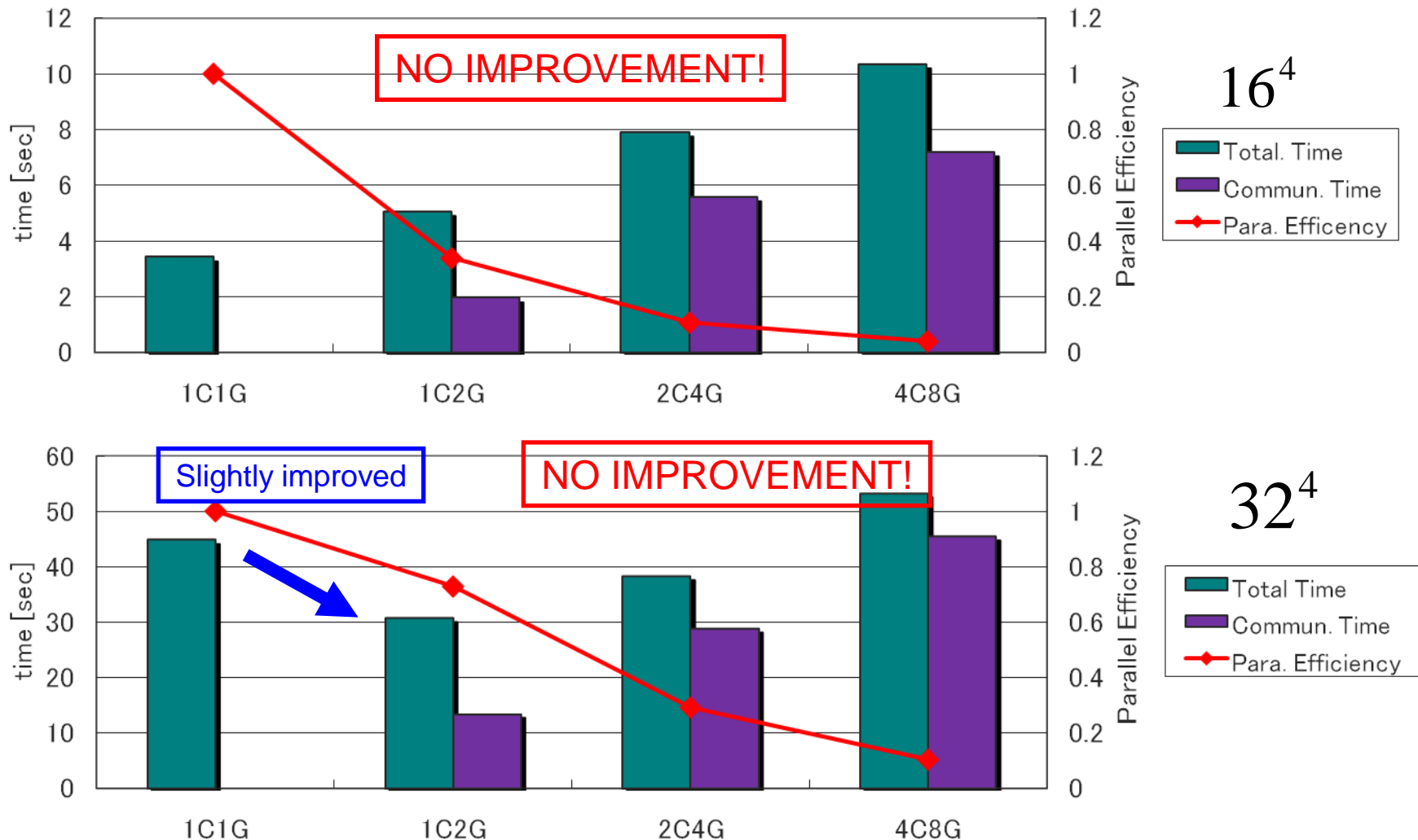


Max bandwidth ~440 MB/s, lowest latency 16μsec (Open-MX)
~140 MB/s, 24μsec (TCP/IP)

6. Towards parallel GPU computation

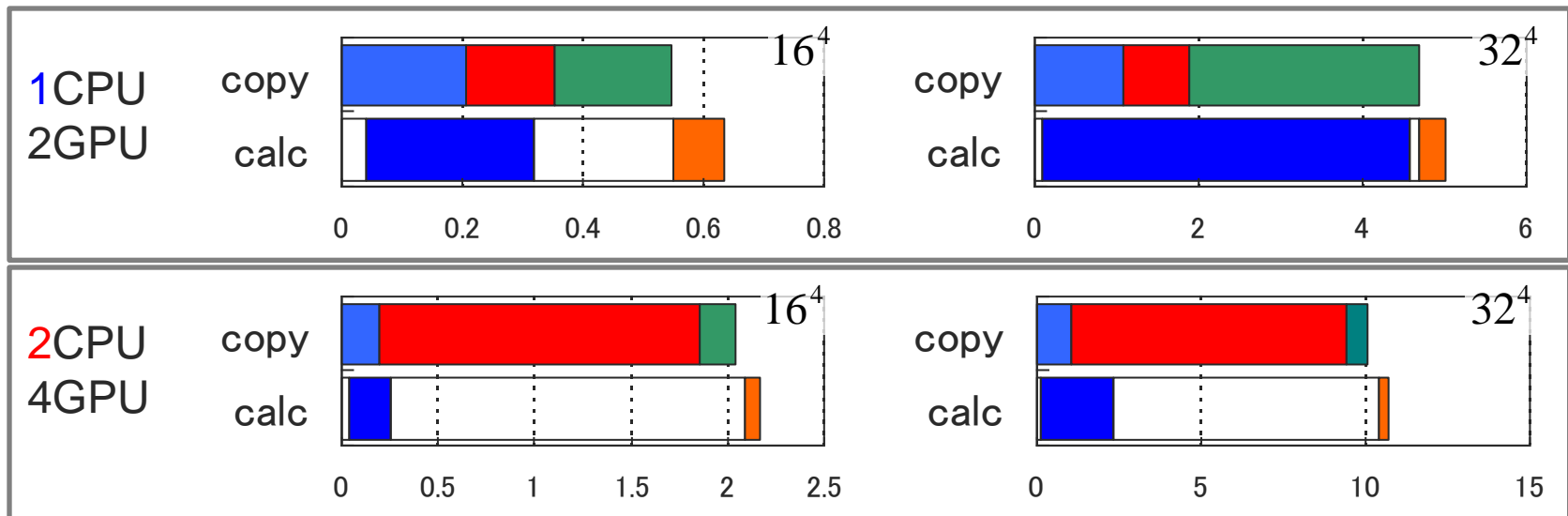
Strong Scaling test, solving $D[U]x=b$

- $\kappa=0.126$
- $csw=1.0$
- $accuracy=10^{-14}$
- MixedPrecBiCGStab solver



6. Towards parallel GPU computation

- Strong Scaling test
- For 16^4 lattice, there is no improvement.
 - Increasing num. of GPU and node, total time increase....
- For 32^4 lattice, there is almost no improvement.
 - Speed up by 30%: 1node1GPU => 1node 2GPU
 - No speed up: increasing nodes.
- Node-node communication (MPI) is still too slow.



■ MPI(CPU ↔ CPU) time

■ Strong Scaling test

- We have tested:
- Simple domain decomposition with communication hiding.
- Gigabit Ethernet (x4 port) and Open-MX enhanced network.
- The network performance does not balance with the GPU speed. Communication hiding fails.
- We need Expensive network hardware (Infiniband, 10GEthernet, True Myrinet.....)?
- Another approach: Solver Preconditioning. Additive Schwarz Preconditioner.

6. Towards parallel GPU computation

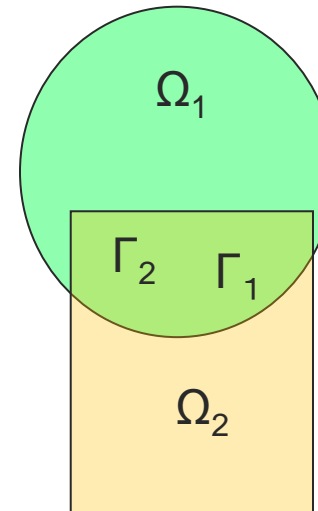
- **Another approach: Solver Preconditioning.**
- **Overlapped Additive Schwarz Preconditioner**
 - Schwarz iteration (common in fluid dynamics research?)

$$x_0, r = b - Ax_0,$$

$$\text{Solve : } A_{\Omega_i} x_{\Omega_i} = r_{\Omega_i}, \text{ in } \Omega_i$$

$$\text{Update : } x = x + f(x_{\Omega_i}),$$

$$\text{Compute : } r = b - Ax,$$

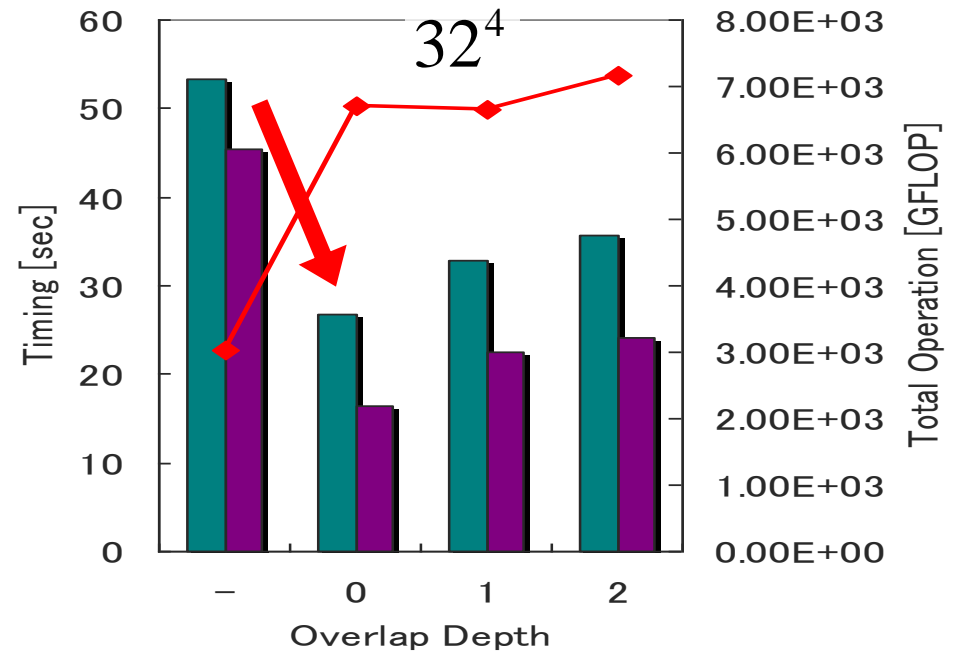
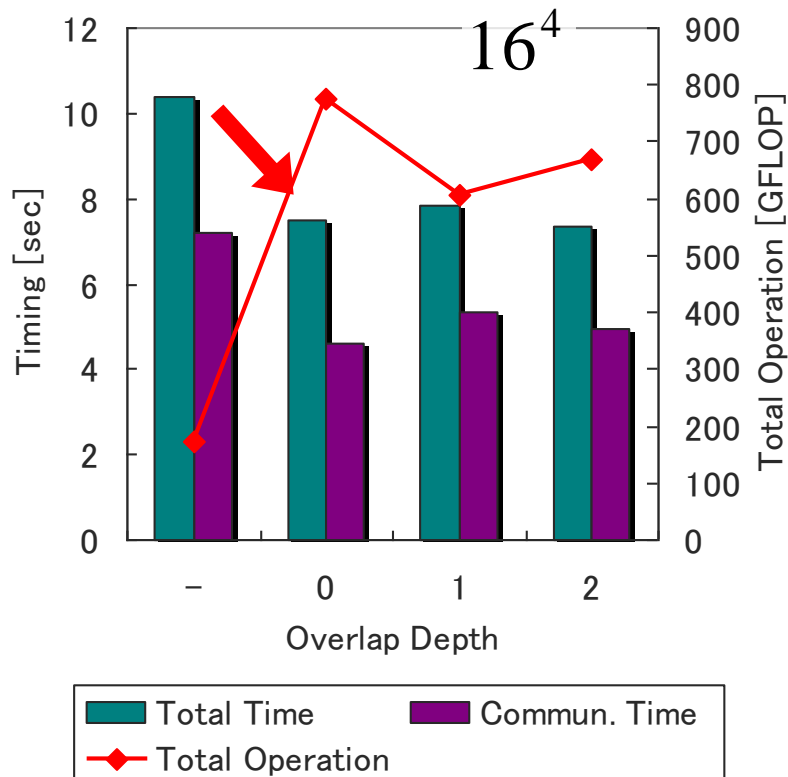


- Use this iteration as a preconditioner for iterative solver.
- Computation in each region Ω is completely independent.
- No communication is required. GPU supports this domain solver.
- Due to the overlapping domain region, total Flop count increases.
- Balance: **Reduction of communication overhead** and **Flop count increase**.

6. Towards parallel GPU computation

Overlapped Additive Schwarz Preconditioner

Preliminary results [K.-I.I. and Y.Osaki]



Additive Schwarz preconditioner without overlapping is effective for larger lattice. Comm. Overhead is well reduced.

[7. Summary]

- Lattice QCD demands huge resource of computing power.
- Studying possibility of GPU acceleration has begun in LQCD community.
- Single/Multiple GPU/s in a node is very effective and attractive as shown here.

But, for more advanced QCD simulation,

- We need more survey on Parallel GPU computation and algorithm.

감사합니다 Thank you!