A decorative graphic consisting of a thin yellow circle on the left side and a horizontal yellow bar that spans across the middle of the slide. The text is centered within this bar.

# Accelerating lattice QCD simulations using multiple GPUs

Ken-Ichi Ishikawa  
(Dept. of Physical Science,  
Hiroshima Univ.)

# [ Contents ]

- 1. Introduction to QCD
- 2. Lattice QCD
- 3. LQCD with GPU/Accelerator
- 4. Accelerating  $D[U]x=b$  solver using single GPU
- 5. Many flavor lattice QCD with multiple GPUs on single node
- 6. Towards parallel GPU computation
- 7. Summary

# [ 1. Introduction to QCD ]

## ■ QCD (Quantum Chromodynamics)

### ○ Describes properties of

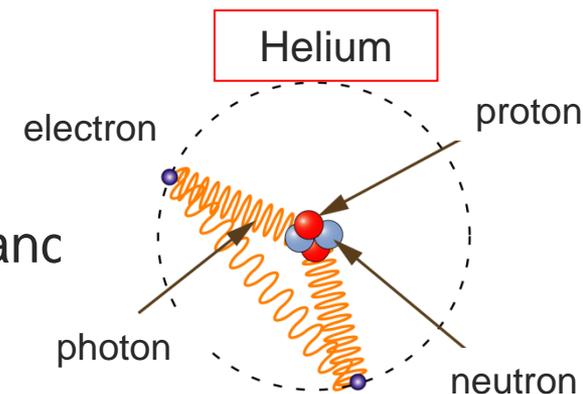
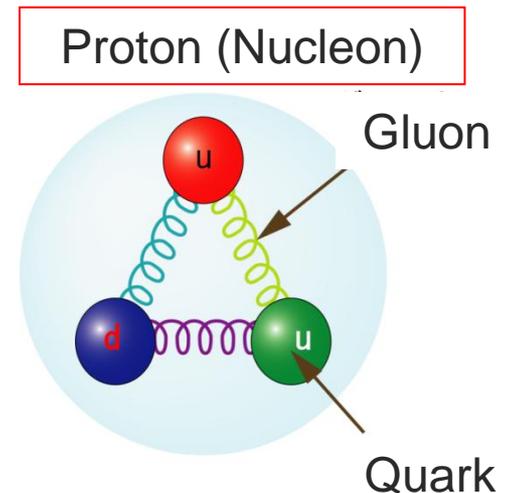
- Nucleons (Proton, Neutron)
- (Mesons, Baryons = Hadrons)
- Strong interaction

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

- Dynamics by exchanging gluons.

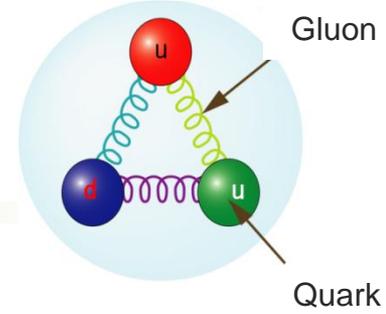
### ○ e.g. QED (Quantum Electrodynamics)

- Molecules/Atoms from nucleons/electrons and photons



# 1. Introduction to QCD (contd')

Proton (Nucleon)

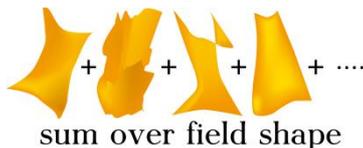


## ■ QCD (Quantum Chromodynamics)

### ○ 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).

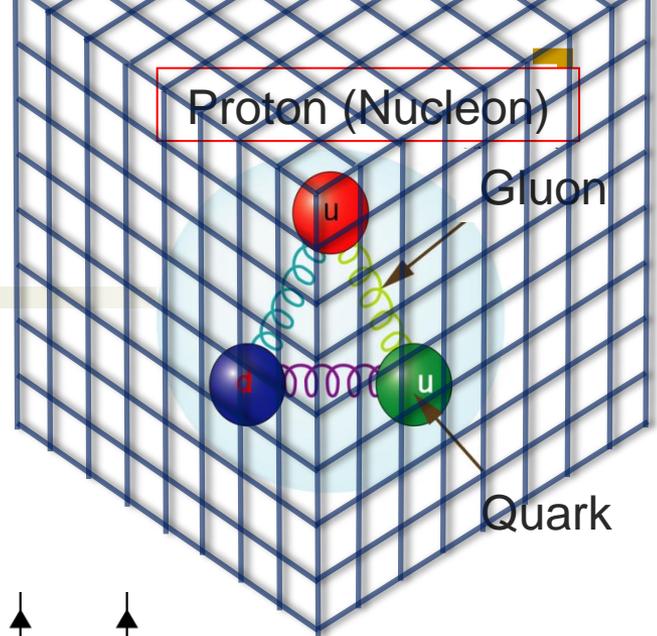
$$Z = \int D A D \bar{\Psi} D \Psi \exp \left[ \int d^4 x \left( -S_{gluon}(A) - S_{quark}(A, \bar{\Psi}, \Psi) \right) \right]$$



# [ 2. Lattice QCD

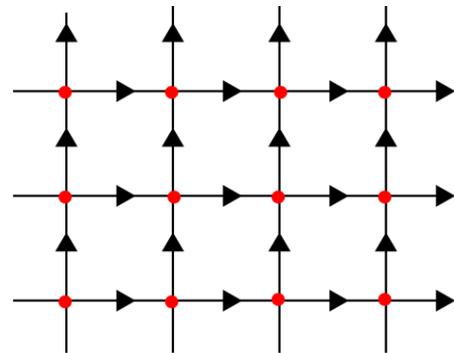
## ■ Lattice QCD [K.G.Wilson (1974)]

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



Quark field  
 $\psi(x)$

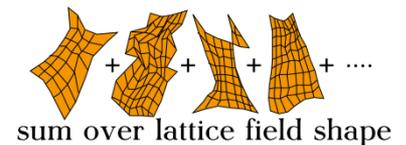
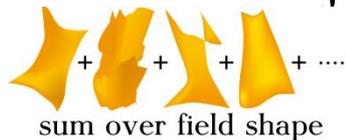
Gluon field  
 $A_\mu(x)$



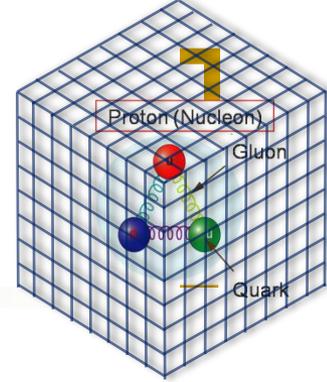
• Quark field  $\psi(n)$     $\rightarrow$    Gluon field  $U_\mu(n)$

- 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

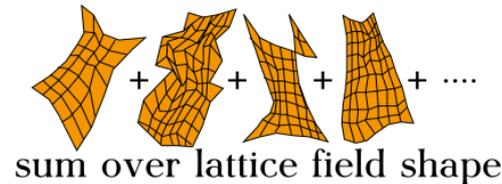
$S(U, \bar{\psi}, \psi)$ : Discretized version of QCD action

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

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

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

- Integration on many var's.



- Observable:  $O$

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

Computation of hadron masses, etc...

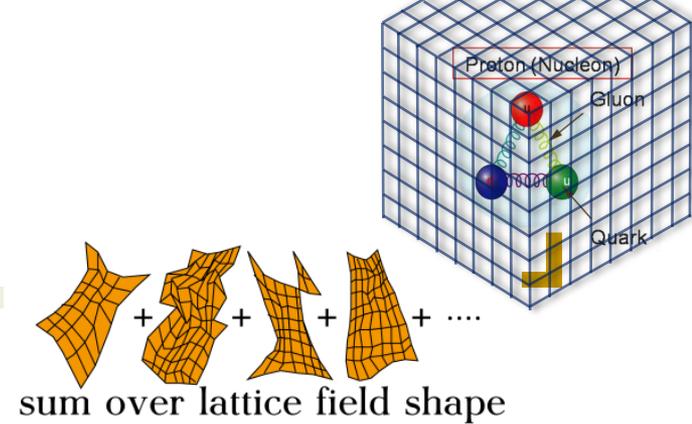
- Similar to Statistical Physics

Monte Carlo Integration  
with Supercomputer!!

## 2. Lattice QCD (cont'd)

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

- Monte Carlo Importance Sampling



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

- Markov chain Monte Carlo to generate sequence of  $U$

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

- Statistical Average

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

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

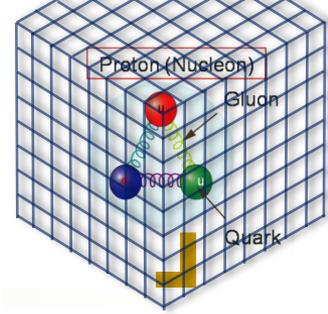
## Hybrid Monte Carlo (HMC) algorithm

[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  $D[U]$ .

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

- 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*.

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

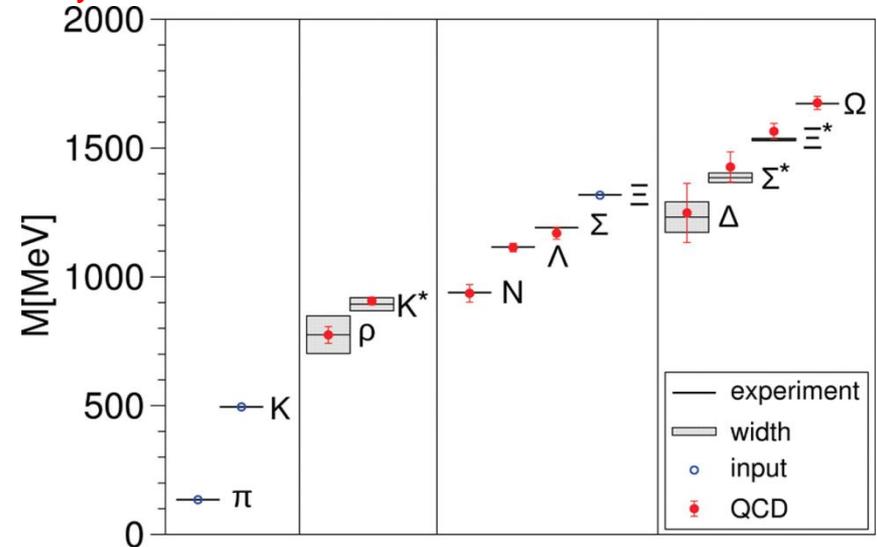
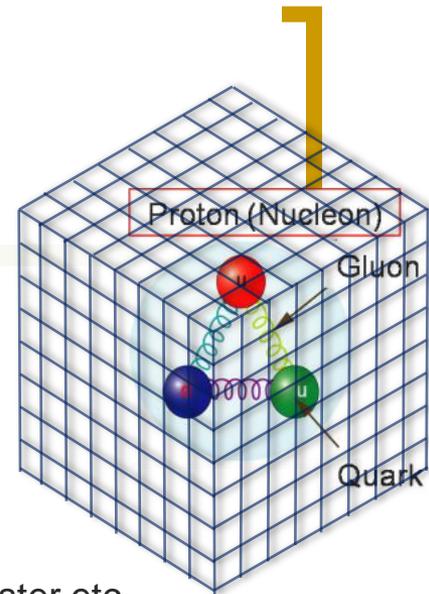
- To get better statistics, we need  $O(100)$  configurations of  $U$ . This needs  $O(1000)$  the HMC cycles. In each HMC cycle, MD needs  $O(100)$  time steps. At each time step we need two inversion of  $D[U]$ . In total we have to invert  $D[U]$  by  $2 \times 100 \times 1000 \times 100 = O(10^7)$  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.**



**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,  
**Q**PACE project (**Q**CD **P**arallel  
computing on the **C**ELL/B.E.)

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



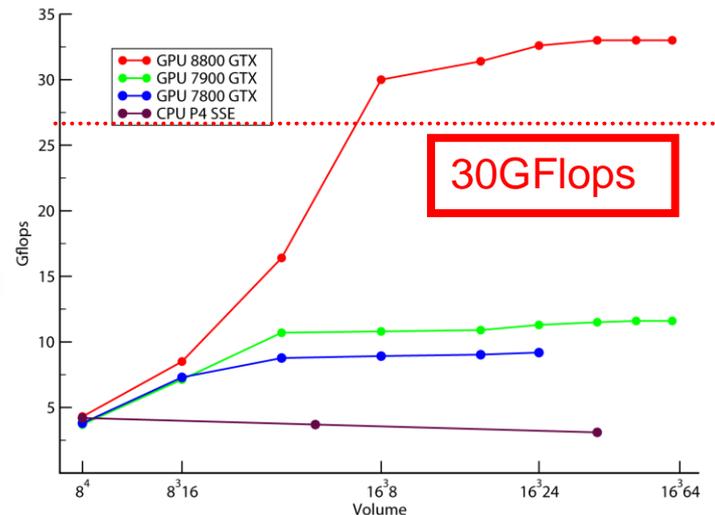
### 3. LQCD with GPU/Accelerator (cont'd)

## 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. > 30 GFlops(SP)
- Lattice Wilson kernel > 30 GFlops
- Difficult to program using Graphic API

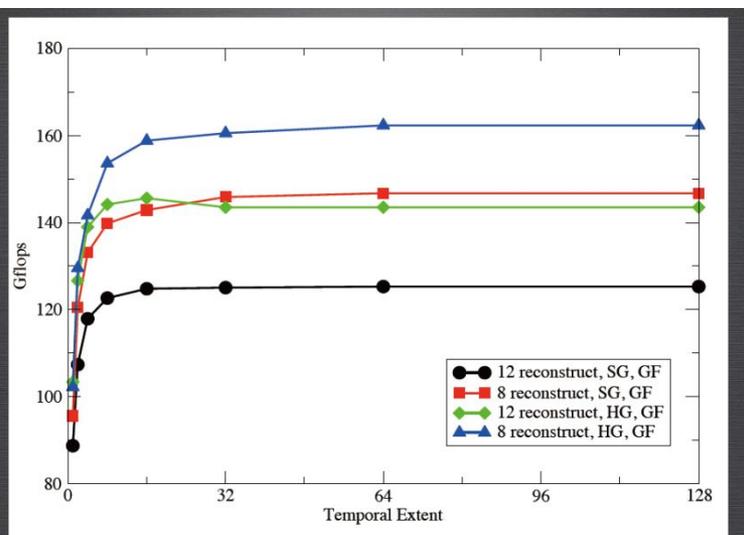
(OpenGL)



30GFlops

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 Unists”, 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 $D[U]$

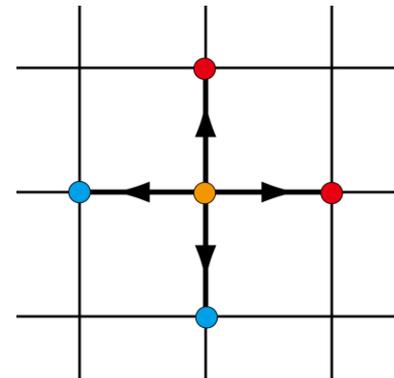
- There are several discretization scheme for Dirac equation  $D$ . ( $D$ : 4D 1<sup>st</sup> order diff. eq.)
- Wilson-Dirac discretization is the most generic one and used widely. And this has an important kernel called Hopping matrix, which is common to other discretization.
- Improvement on the Hopping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

## Hopping kernel (Hopping matrix)

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

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

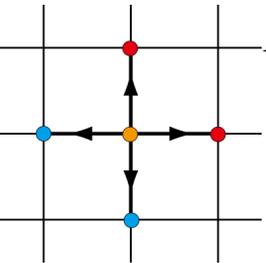
- 4D, first-difference operator.
- sites connected by gluon  $U$ , by which quark-color are mixed.
- quark-spin are mixed by Dirac's gamma matrix  $\gamma_{\mu}$
- Memory bandwidth intensive operation.**
  - Byte/Flop ~ 3 B (D.P.), ~1.5 B (S.P.)**



- 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 :=> 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)  $q = Dv$  [doble prec.]

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

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

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

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

GPU task  
CUDA code

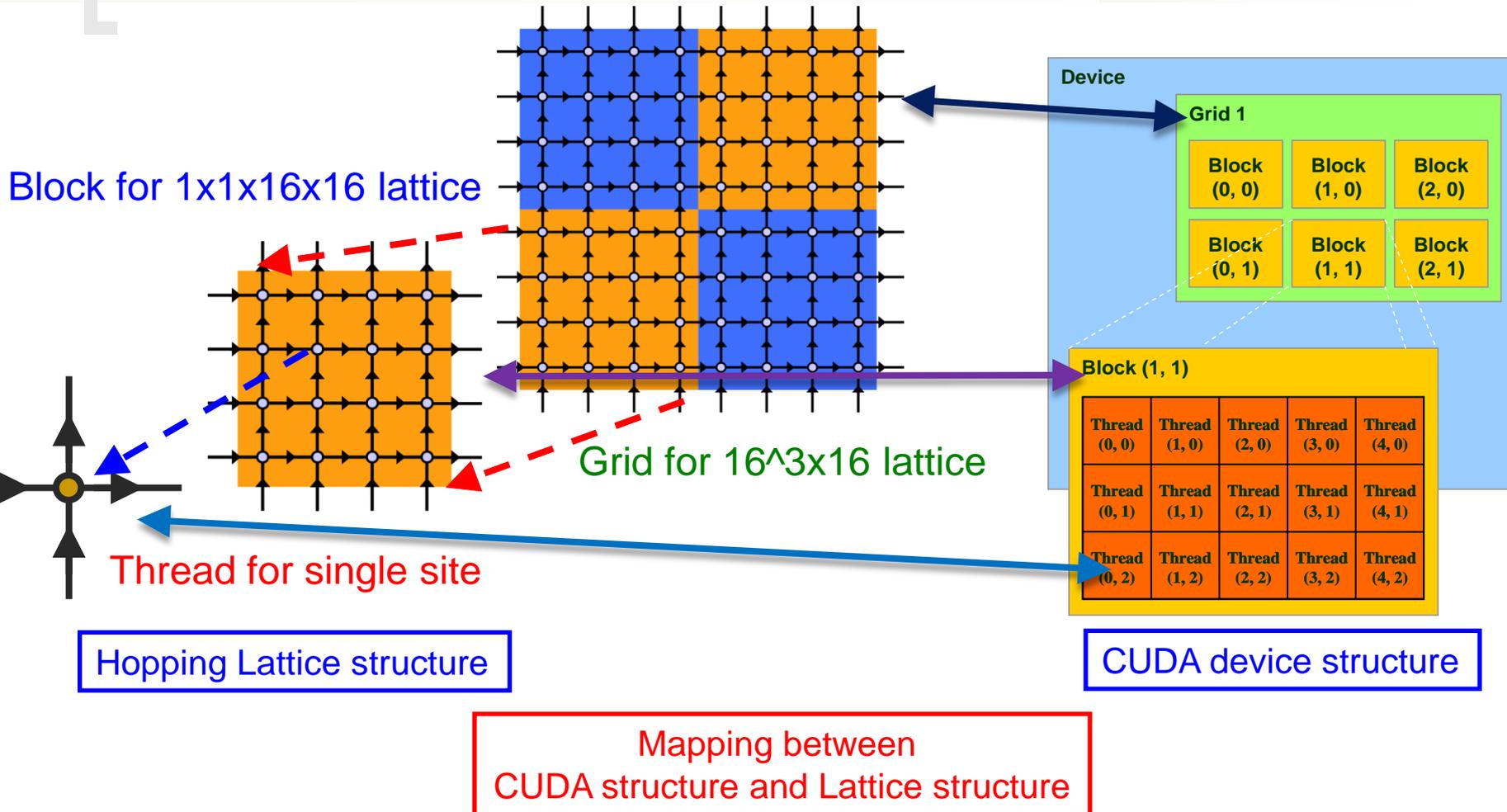
CPU task  
HOST code

- 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.

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

## GPU acceleration LQCD Kernel $D[U]$

- Hopping kernel CUDA code



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

## 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  $3 \times 3$  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.  $3 \times 2$  complex elements are sufficient to reconstruct  $3 \times 3$   $U$ .  $(12 \text{ real} \times 4) = 48$  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!

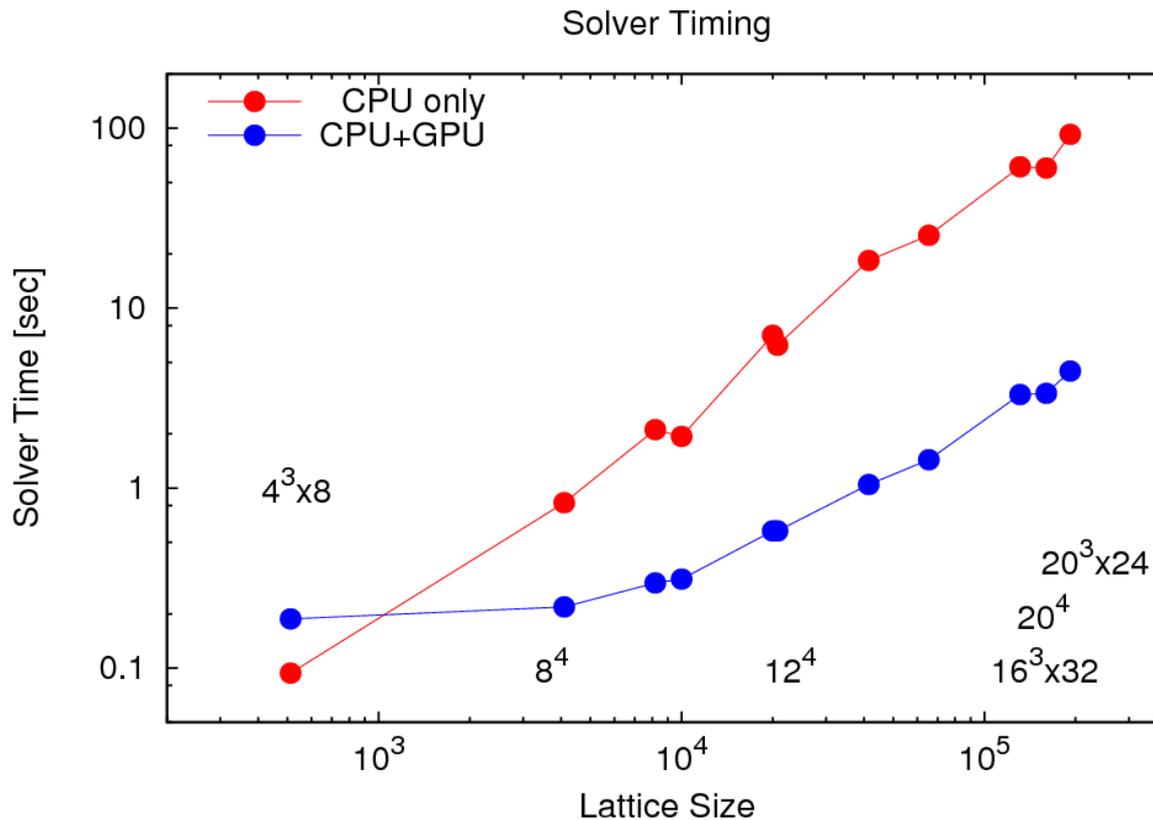
# 4. Accelerating $D[U]x=b$ solver using single GPU.

- 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)



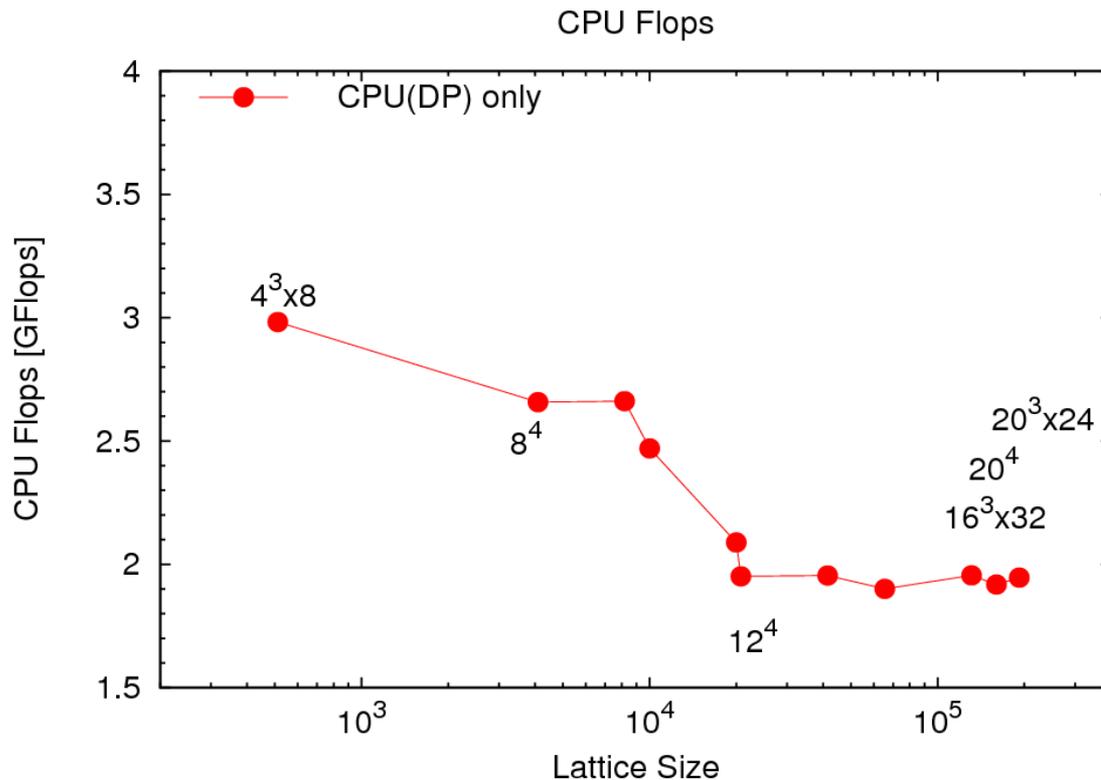
Small volume is not effective for GPU because parallelism is less than num of GPU cores(maximum threads).

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))

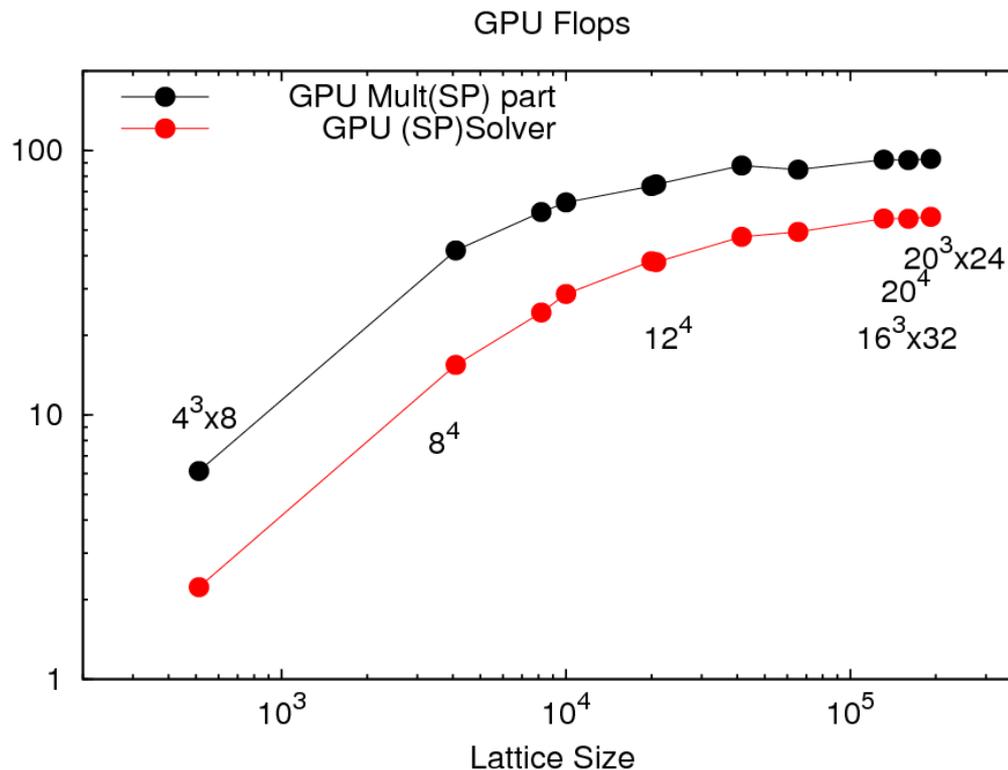
GPU performance reaches about 100 GFlops (S.P.).

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)

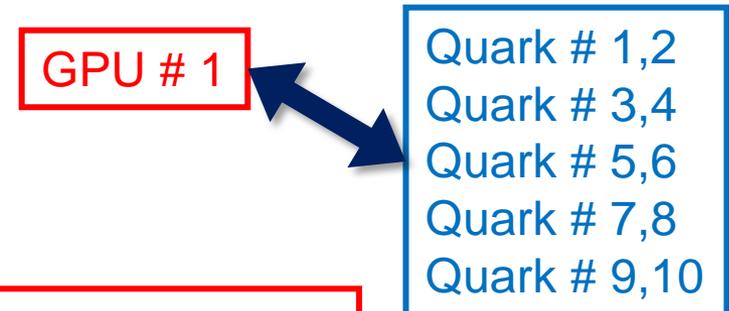
- 2GPU case



- 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.**

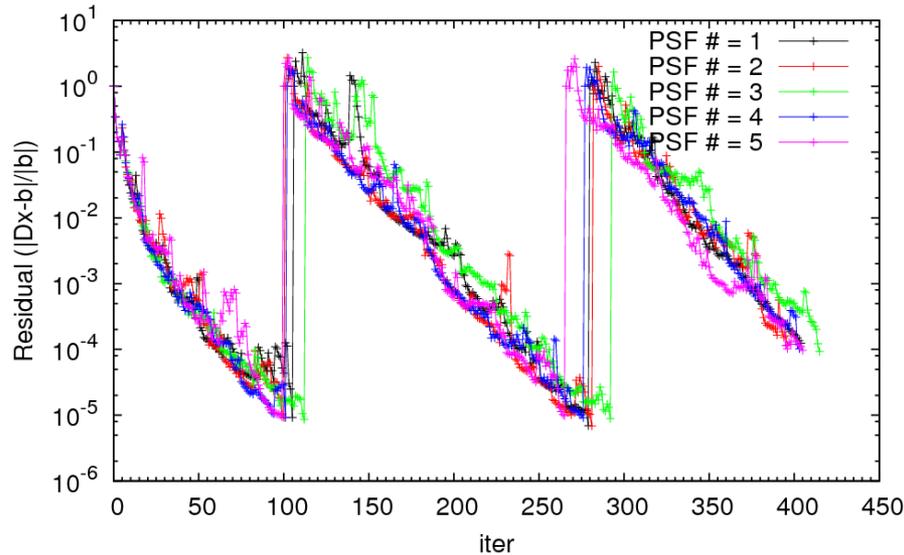
Quark pair is always assigned to single external scalar field.

# 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)

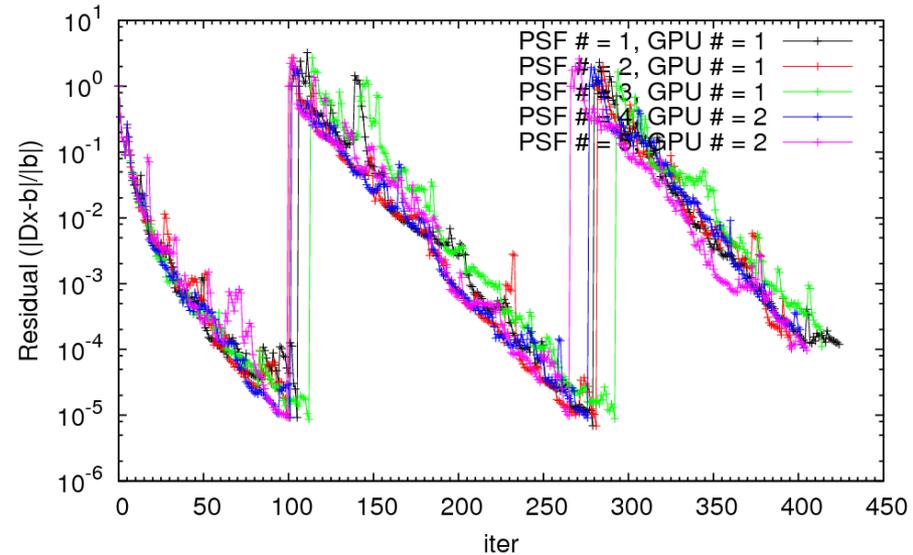
Global Blocked DP-BiCGStab + Normal 1-GPU SP-BiCGStab  
Solver Residual History



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



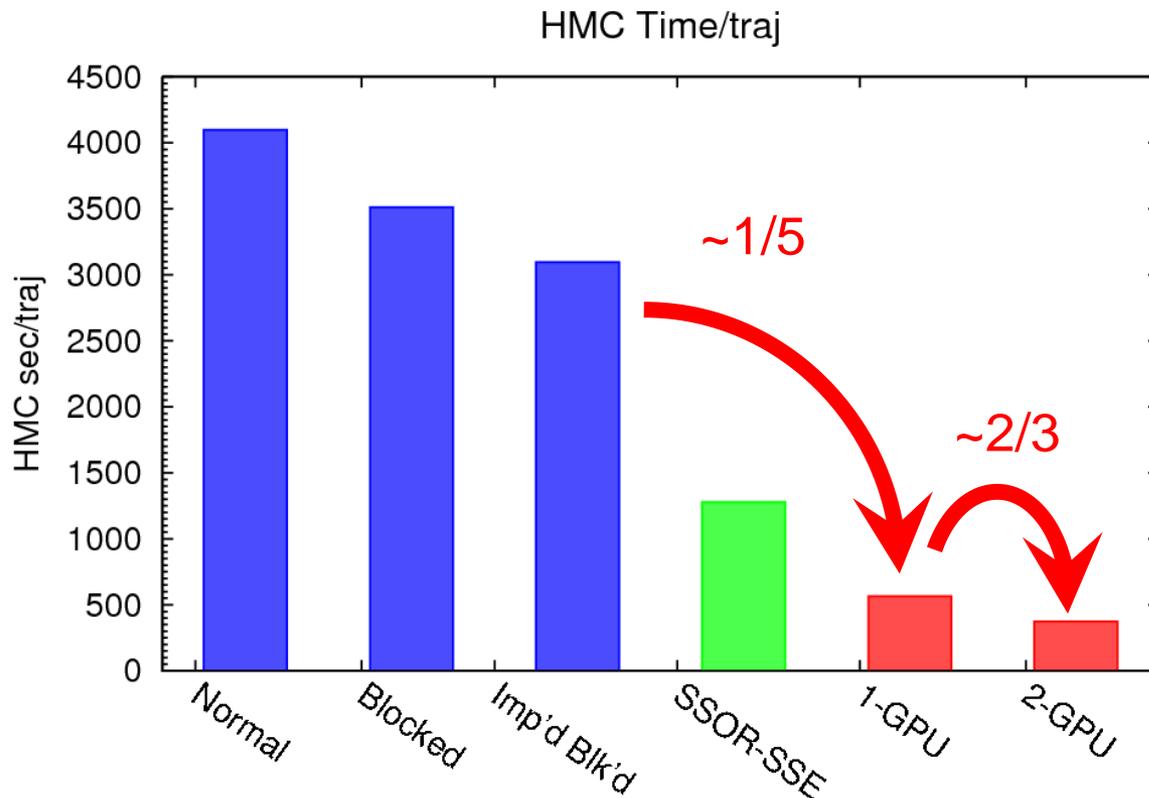
GPU solver is called 3 times for  
GPU#1 and 2 times for GPU#2.  
Full Timing : 2.69 sec

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.

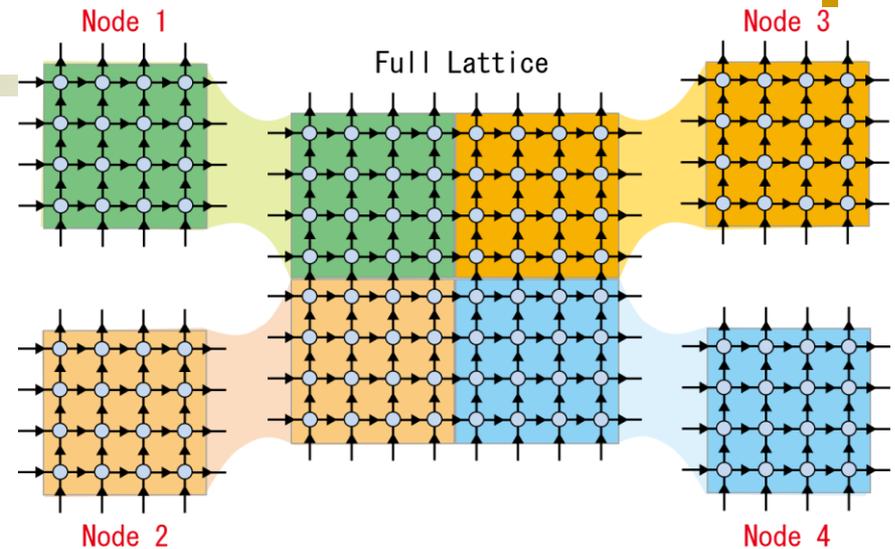
# [ 6. Towards parallel GPU computation ]

- 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).

## 6. Towards parallel GPU computation

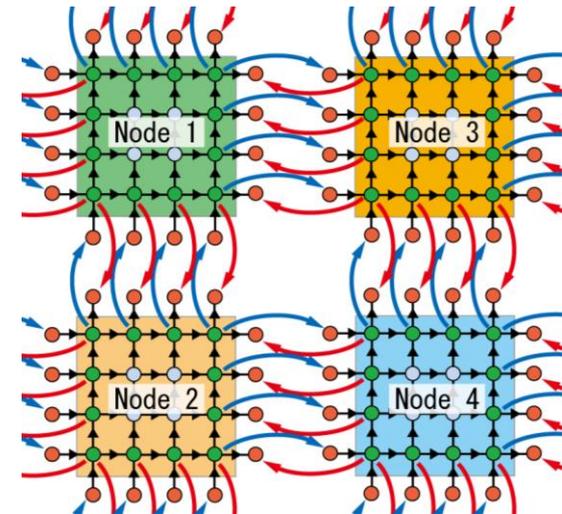
# LQCD parallelization

- 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.



## 6. Towards parallel GPU computation

### ■ Parallel GPU implementation

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

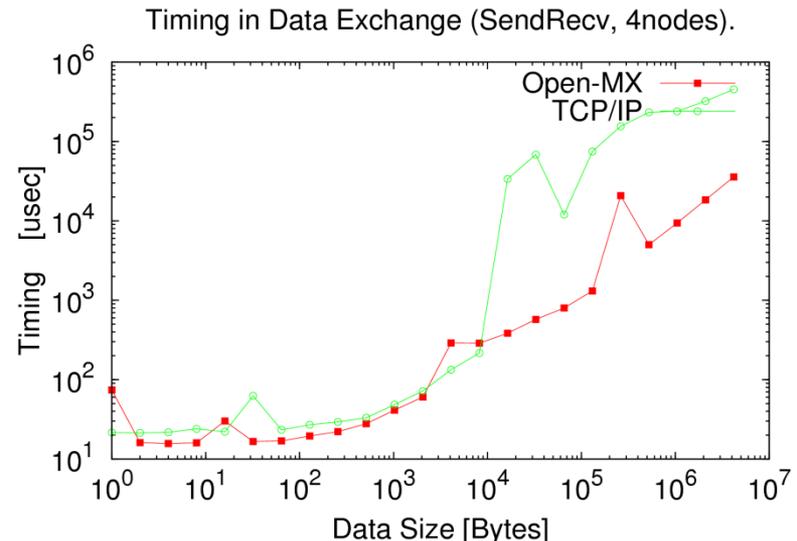
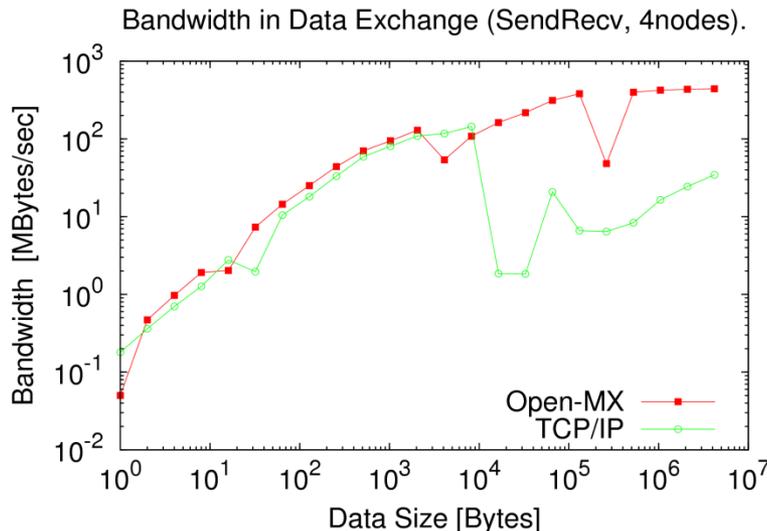
- Machine: 4PC's with
- CPU : Intel Core i7 920@2.67GHz (4cores)
- GPU : GeForce GTX 285 × 2 (2GPUs in 1 node)
- Memory 6GBytes
- LAN Adapter : Intel Gigabit ET Quad Port Server. Poor man's network. (COTS, cheap)
- CentOS 5.4, CUDA 2.3, OpenMPI
- TCP/IP is too slow. Instead We employed Open-MX (Myrinet 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.**

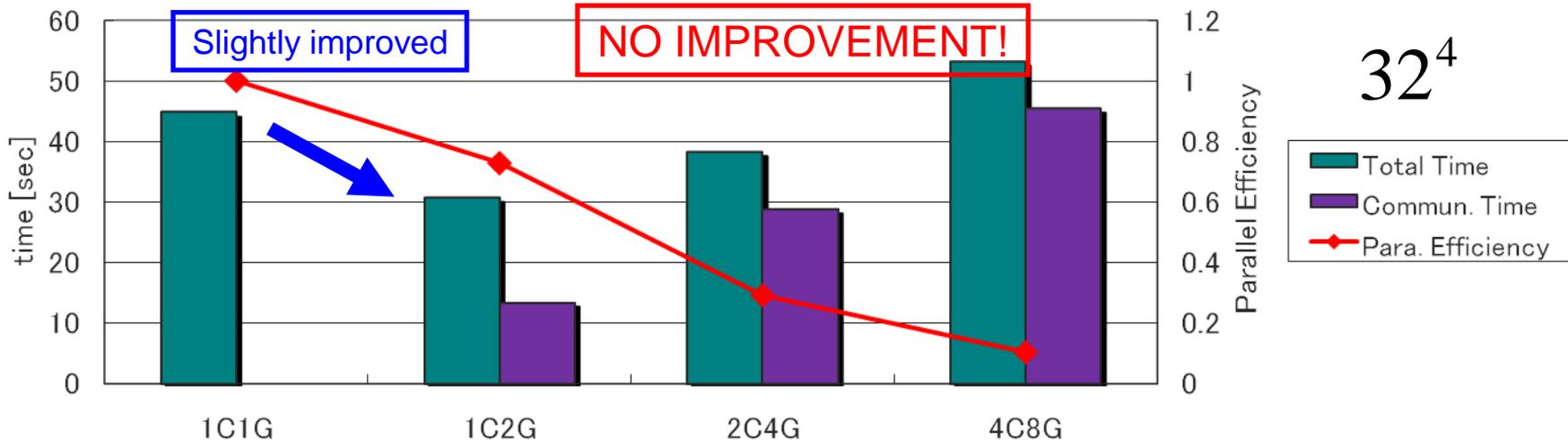
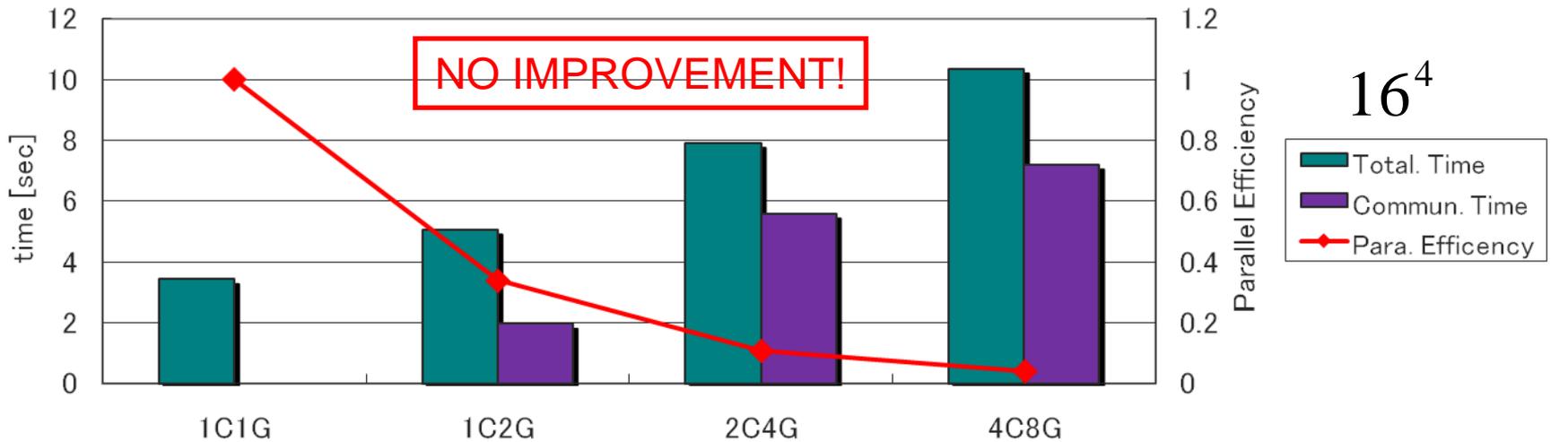


**Max bandwidth ~440 MB/s, lowest latency 16usec (Open-MX)**  
**~140 MB/s, 24usec (TCP/IP)**

## 6. Towards parallel GPU computation

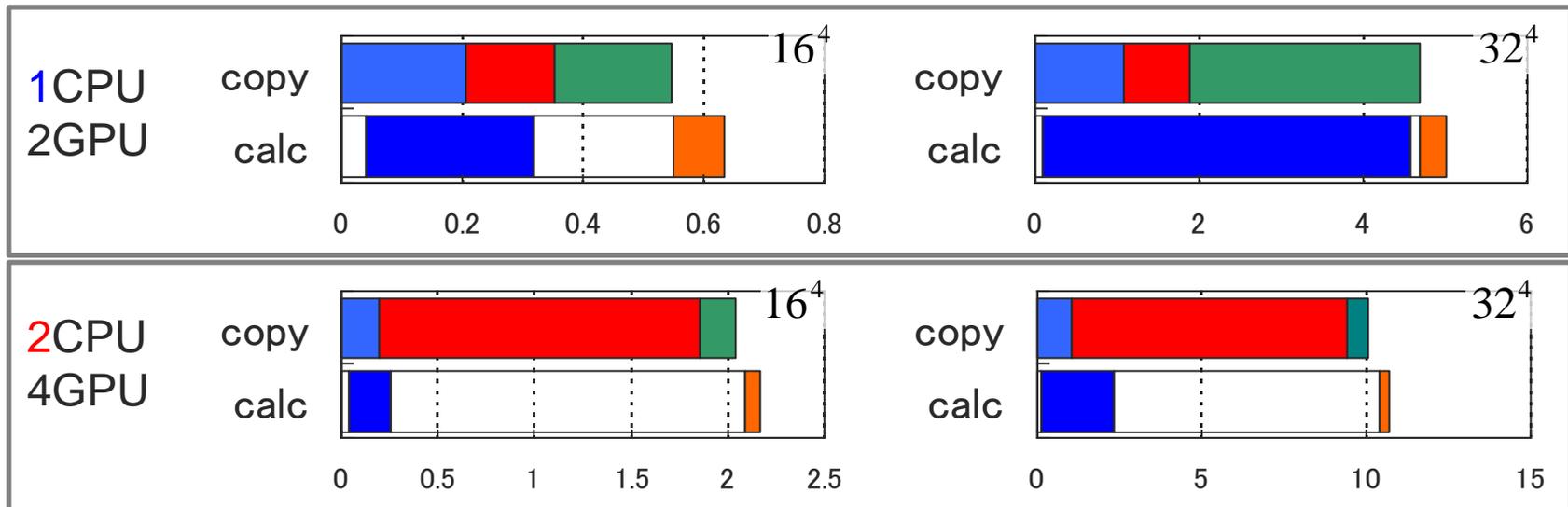
### Strong Scaling test, solving $D[U]x=b$

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

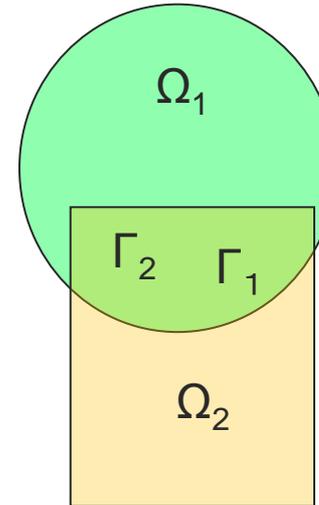
- **Another approach: Solver Preconditioning.**
- **Overlapped Additive Schwarz Preconditioner**
  - Schwarz iteration (common in fluid dynamics research?)

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

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

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

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

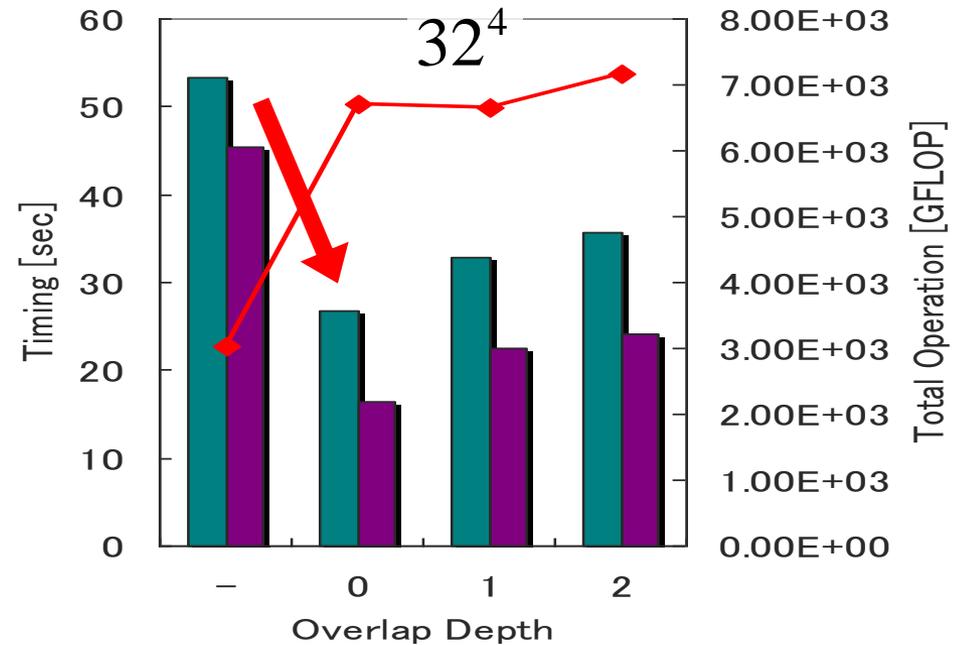
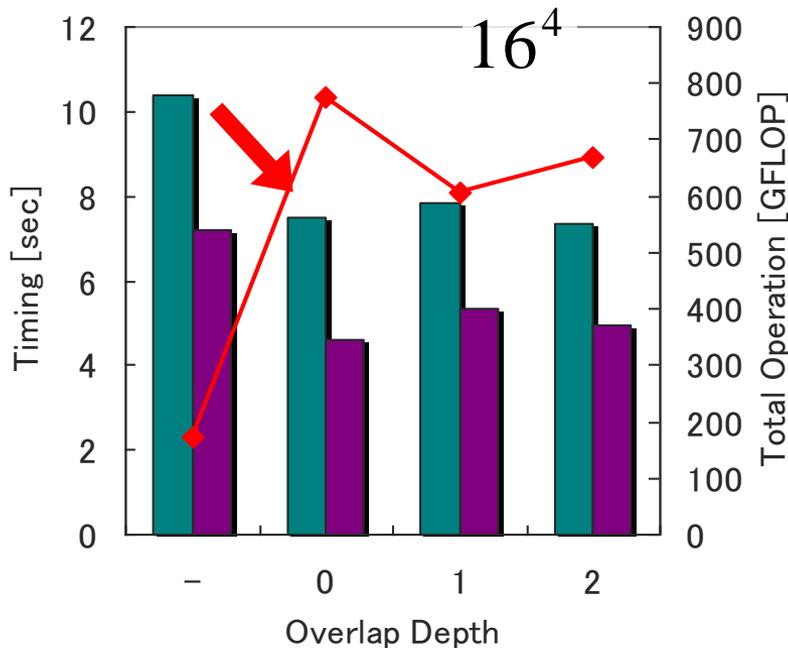


- Use this iteration as a preconditioner for iterative solver.
- Computation in each region  $\Omega$  is completely independent.
- No communication is required. GPU supports this domain solver.
- Due to the overlapping domain region, total Flop count increases.
- Balance: **Reduction of communication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