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3rd IFERC workshop on the usage of GPU based system
for fusion applications



GPU Acceleration of 5D Global Gyrokinetic Code GKNET by OpenACC Directives

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Acknowledgement

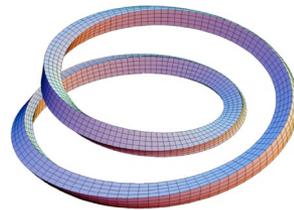
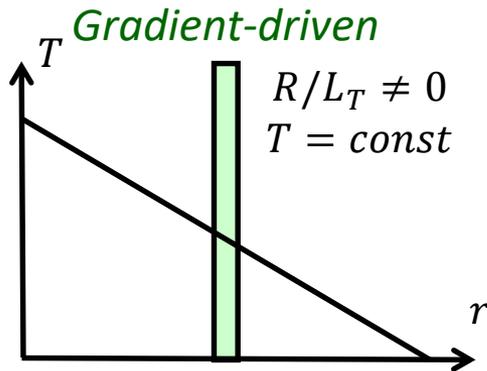
Masatoshi YAGI, Naoaki MIYATO, Haruki SETO (QST), Akira NARUSE (NVIDIA Japan)

Global/Local Gyrokinetics

Local δf approach

$$\cancel{\partial_t f_{eq} - [H, f_{eq}]} = \cancel{C(f_{eq})} + \cancel{S}$$

$$\partial_t \delta f - [H, \delta f] - [\delta H, f_{eq}] - [\delta H, \delta f] = C(\delta f)$$

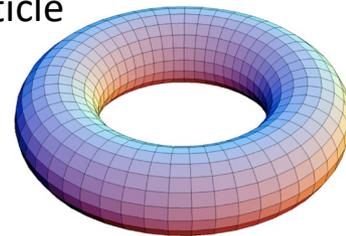
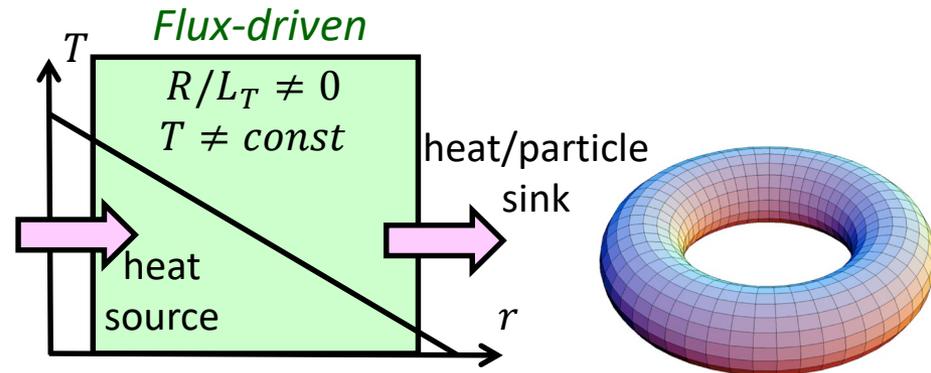


- ☺ Very powerful tool to estimate turbulent transport process
- ☺ Computationally efficient
-> multi(ion/electron)-scale simulation

Global full- f approach

$$\partial_t f_{eq} - [H, f_{eq}] = C(f_{eq}) + S$$

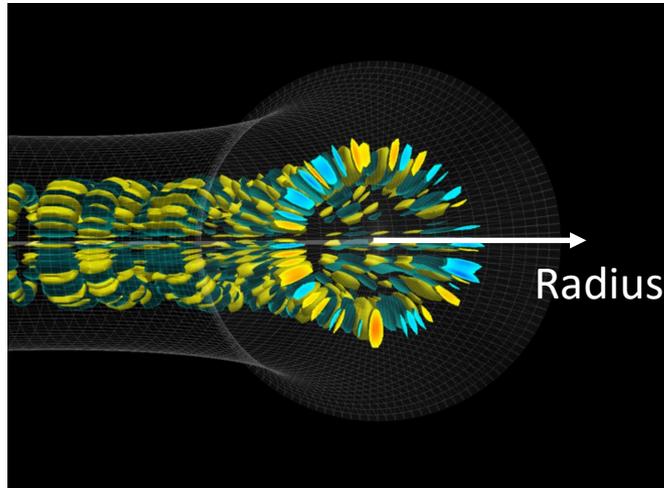
$$\partial_t \delta f - [H, \delta f] - [\delta H, f_{eq}] - [\delta H, \delta f] = C(\delta f)$$



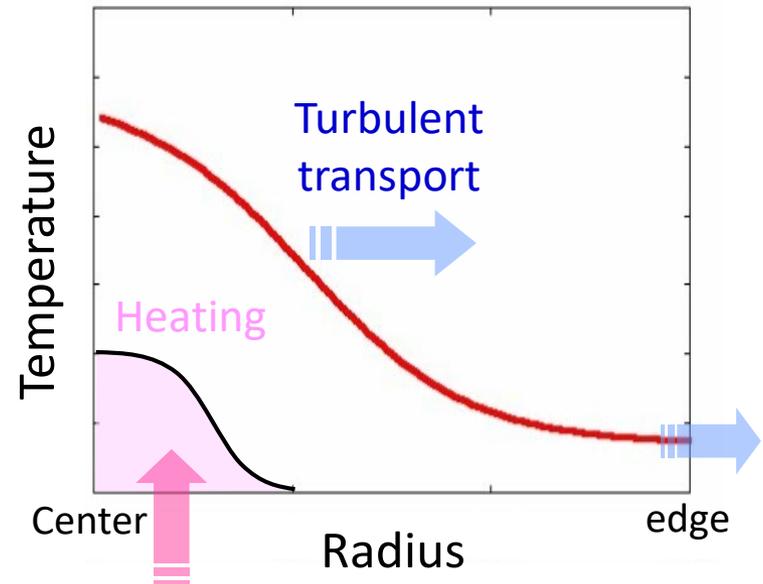
- ☺ Global effects can be treated precisely
- ☺ Mean E_r is self-consistently determined
-> Internal Transport Barrier (ITB)
- ☺ Both neoclassical & turbulent transport process can be traced

Typical Flux-driven Simulation

Animation of turbulence structure



Animation of plasma temperature



- ✓ In flux-driven simulations, the external heat source is introduced, which sustains the background temperature profile. As the result, we can treat non-decaying turbulence over the confinement time.
- ✓ By means of such flux-driven simulations, we try to **understand the relationship between background profile evolution and turbulence dynamics.**

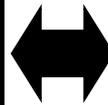
Global Gyrokinetic Code *GKNET*

5D phase-space Boltzmann equation
for distribution functions

$$\frac{\partial f_s}{\partial t} + \sum_{i=1}^5 \frac{dx_i}{dt} \frac{\partial f_s}{\partial x_i} = C_{coll}$$

Time derivatives
of distribution
function
for species s

Convection
In 5D space



3D real-space Poisson equation
for electrostatic potential

$$-\nabla^2 \phi = \sum_s e_s \int f_s d\mathbf{v}$$

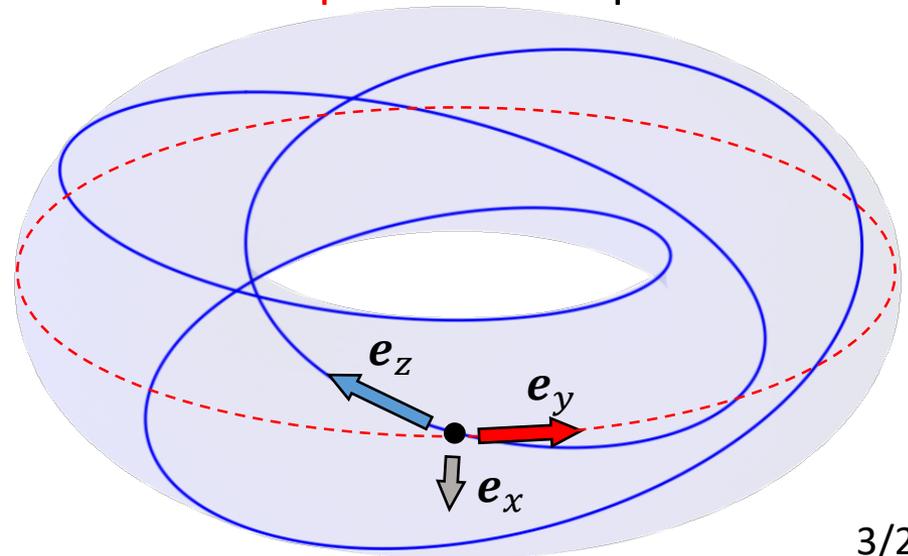
Electrostatic potential
(related to convection
in the Boltzmann
equation)

Total charge
density

- ✓ For performing flux-driven simulations in 5D phase-space, **GKNET (GyroKinetic Numerical Experiment for Tokamak)** has been developed and MPI-parallelized.
- ✓ To reduce the number of the simulation grid and the resultant calculation time, we recently introduced field aligned coordinate.

[Okuda, PFR-2023]

$$\begin{cases} x = \rho \\ y = -\zeta + \int_0^\theta v(\rho, \theta') d\theta' \\ z = \theta \end{cases}$$



Global Gyrokinetic Code *GKNET* -Vlasov Solver-

5D phase-space Gyrokinetic Boltzmann equation

$$\frac{\partial f}{\partial t} + \frac{d\mathbf{R}}{dt} \cdot \frac{\partial f}{\partial \mathbf{R}} + \frac{dv_{\parallel}}{dt} \frac{\partial f}{\partial v_{\parallel}} = C_{coll} \quad \mathbf{R} = (x, y, z)$$

$$\frac{d\mathbf{R}}{dt} \equiv \{\mathbf{R}, H\} = v_{\parallel} \mathbf{b}(\mathbf{R}) + \frac{c}{eB_{\parallel}^*(\mathbf{R}, v_{\parallel})} \mathbf{b}(\mathbf{R}) \times [e\nabla\langle\phi(\mathbf{R})\rangle_{\alpha} + m_i v_{\parallel}^2 \mathbf{b}(\mathbf{R}) \cdot \nabla \mathbf{b}(\mathbf{R}) + \mu \nabla B(\mathbf{R})]$$

$$\frac{dv_{\parallel}}{dt} \equiv \{v_{\parallel}, H\} = -\frac{\mathbf{B}_{\parallel}^*(\mathbf{R}, v_{\parallel})}{m_i B_{\parallel}^*(\mathbf{R}, v_{\parallel})} \cdot [e\nabla\langle\phi(\mathbf{R})\rangle_{\alpha} + \mu \nabla B(\mathbf{R})]$$

Vlasov solver

- ✓ **Spatial discretization: 4th-order Morinishi scheme** [Idomura, JCP-2007]
- ✓ Time integration: 4th-order explicit Runge-Kutta scheme

Ex. 2D case of Morinishi scheme

$$\frac{\partial f}{\partial t} = \underbrace{-v_x \frac{\partial f}{\partial x}}_{\text{blue}} - \underbrace{v_y \frac{\partial f}{\partial x}}_{\text{green}} = -\frac{1}{2} \left[\underbrace{v_x \frac{\partial f}{\partial x} + \frac{\partial(v_x f)}{\partial x}}_{\text{blue}} \right] - \frac{1}{2} \left[\underbrace{v_y \frac{\partial f}{\partial y} + \frac{\partial(v_y f)}{\partial y}}_{\text{green}} \right] \quad (\because \frac{\partial v_x}{\partial x} + \frac{\partial v_y}{\partial y} = 0)$$



Discretize each derivative by using 4th-order central FDM

$$\left(\frac{\partial f}{\partial t}\right)_{i,j}^n = \underbrace{-\frac{1}{2} \left\{ \left[v_{x,i,j}^n \left(\frac{4}{3} \frac{f_{i+1,j}^n - f_{i-1,j}^n}{2\Delta x} - \frac{1}{3} \frac{f_{i+2,j}^n - f_{i-2,j}^n}{4\Delta x} \right) \right] + \left[\frac{4}{3} \frac{v_{x,i+1,j}^n f_{i+1,j}^n - v_{x,i-1,j}^n f_{i-1,j}^n}{2\Delta x} - \frac{1}{3} \frac{v_{x,i+2,j}^n f_{i+2,j}^n - v_{x,i-2,j}^n f_{i-2,j}^n}{4\Delta x} \right] \right\}}_{\text{blue}}}_{\text{green}} - \frac{1}{2} \left\{ \left[v_{x,i,j}^n \left(\frac{4}{3} \frac{f_{i,j+1}^n - f_{i,j-1}^n}{2\Delta y} - \frac{1}{3} \frac{f_{i,j+2}^n - f_{i,j-2}^n}{4\Delta y} \right) \right] + \left[\frac{4}{3} \frac{v_{x,i,j+1}^n f_{i,j+1}^n - v_{x,i,j-1}^n f_{i,j-1}^n}{2\Delta y} - \frac{1}{3} \frac{v_{x,i,j+2}^n f_{i,j+2}^n - v_{x,i,j-2}^n f_{i,j-2}^n}{4\Delta y} \right] \right\}$$

Global Gyrokinetic Code *GKNET* -Poisson Solver-

3D real-space Gyrokinetic quasi-neutrality condition

$$\nabla \cdot \left(\frac{m_i n(x)}{B(x, z)^2} \nabla_{\perp} \phi \right) - \frac{e^2 n}{T_e} [\phi - \langle \phi \rangle_f] = -2\pi e \iint \langle \delta f_i \rangle \frac{B_{\parallel}^*}{m_i} dv_{\parallel} d\mu$$

 $(L_0 + L_1)\phi(x, y, z) = s(x, y, z)$

$$L_0 = c_1(x, z) \frac{\partial^2}{\partial x^2} + c_2(x, z) \frac{\partial^2}{\partial y^2} + c_3(x, z) \frac{\partial}{\partial x} \frac{\partial}{\partial y} + c_4(x, z) \frac{\partial}{\partial x} + c_5(x, z) \frac{\partial}{\partial y} + c_6(x)$$

$$L_1 = l_1(x, z) \frac{\partial}{\partial x} \frac{\partial}{\partial z} + l_2(x, z) \frac{\partial}{\partial y} \frac{\partial}{\partial z} + l_3(x, z) \frac{\partial^2}{\partial z^2} + l_4(x, z) \frac{\partial}{\partial z}$$

Poisson solver

Step-1 : FFT along the y direction \leftarrow because all the coefficients are independent to y

Step-2 : Set the initial guess $\hat{\phi}_n^{(0)}(x, z)$, and then solve $\hat{L}_0 \hat{\phi}_n^{(1)}(x, z) + \hat{L}_{1,D} \hat{\phi}_n^{(1)}(x, z) = \hat{s}_n(x, z) - \hat{L}_{1,ND} \hat{\phi}_n^{(0)}(x, z)$ by using the 1D matrix solver

Step-3 : By repeating Step-2 (=Jacobi method), get the converged solution $\hat{\phi}_n$

\leftarrow because $\frac{\partial \phi}{\partial z}$ is higher order, a few iterations are enough for the convergence

Targets of This Talk

- ✓ We are aiming at the GPU acceleration of GKNET by using OpenACC directives on MARCONI 100 (CINECA, Italy).

A) Multi-GPU

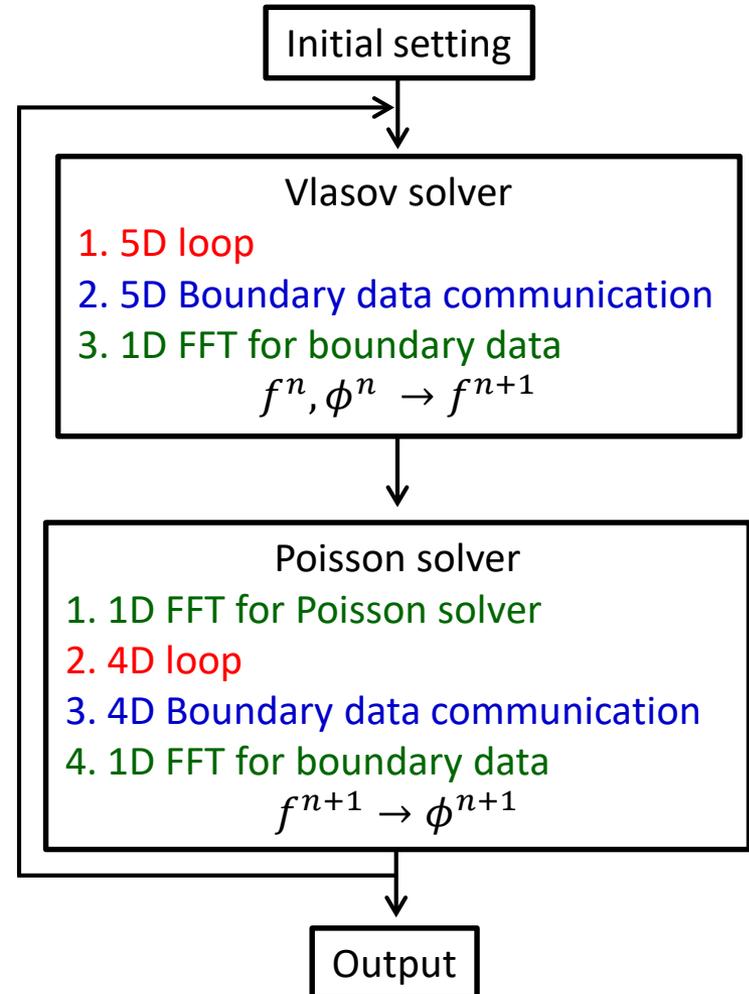
- ✓ By utilizing multi-GPU on each node, we accelerated 5D/4D loops in the Vlasov/Poisson solvers.

B) Direct GPU-GPU data transfer

- ✓ Instead of CPU-GPU data transfer, direct GPU-GPU data transfer is introduced for 5D/4D boundary data exchange in the Vlasov/Poisson solvers.

C) Batched CuFFT

- ✓ For boundary data in the Vlasov/Poisson solvers & for solving the quasi-neutrality condition in the Poisson solver, the batched cuFFT is installed.



Utilization of Multi-GPU - 1

Typical 5D loop in Vlasov solver

```
def = acc_get_num_devices (acc_device_nvidia)
gpuid = mod(rank, def)
call acc_set_device_num(gpuid,
acc_device_default)

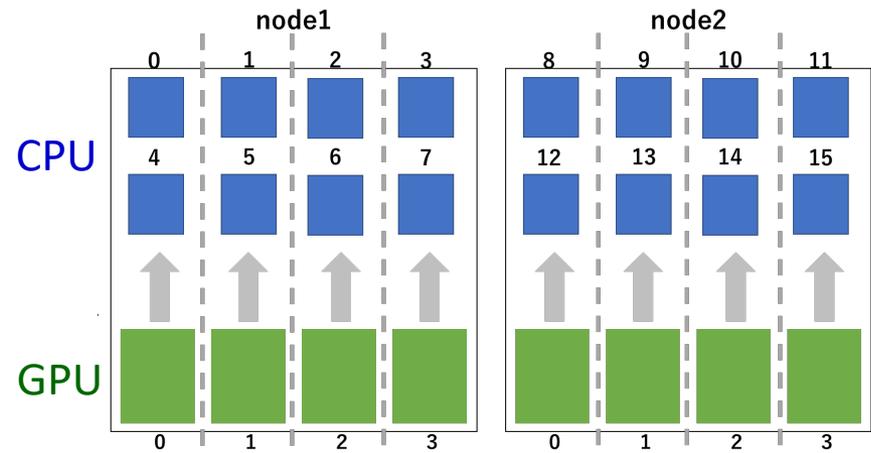
!$acc data copy(...) &
!$acc& copyin(...) &
!$acc& create(...)

!$acc wait
!$acc kernels
!$acc loop collapse(4) gang vector
DO x_i = 3, N_x_p+2
  DO y_i = 3, N_y_p+2
    DO z_i = 3, N_z_p+2
      DO v_i = 3, N_v+2
        DO u_i = 3, N_u+3

          Heavy calculation

        END DO
      END DO
    END DO
  END DO
END DO
!$acc end kernels
```

Image of GPU allocation to CPUs

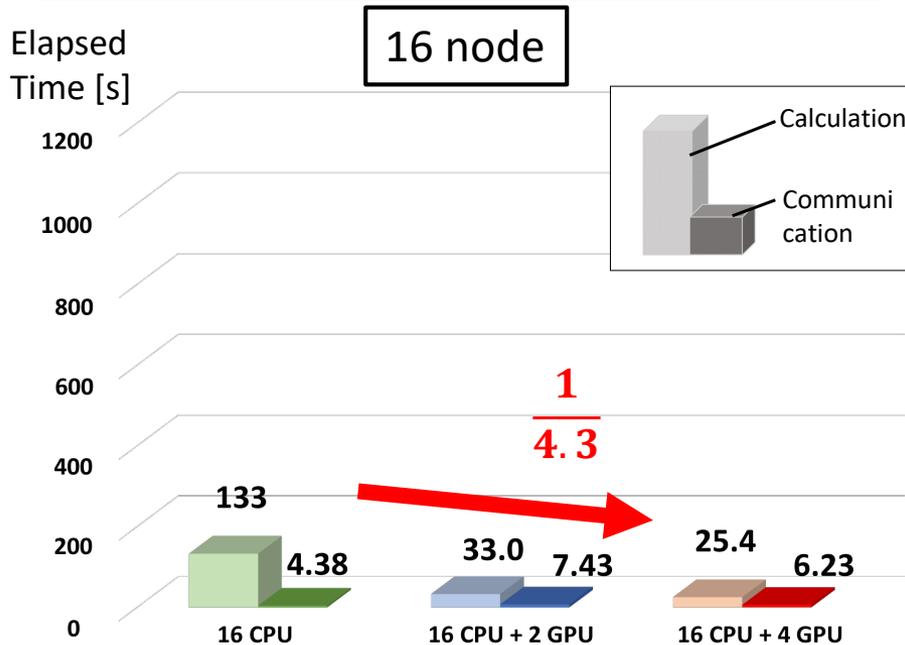


- ✓ Each CPU is explicitly linked to the GPU in same node.
- ✓ The OpenACC data directives (copy, copyin, etc.) are utilized for CPU-GPU data transfer.
- ✓ The 4D/5D loops are collapsed to one loop and then distributed to each GPU.

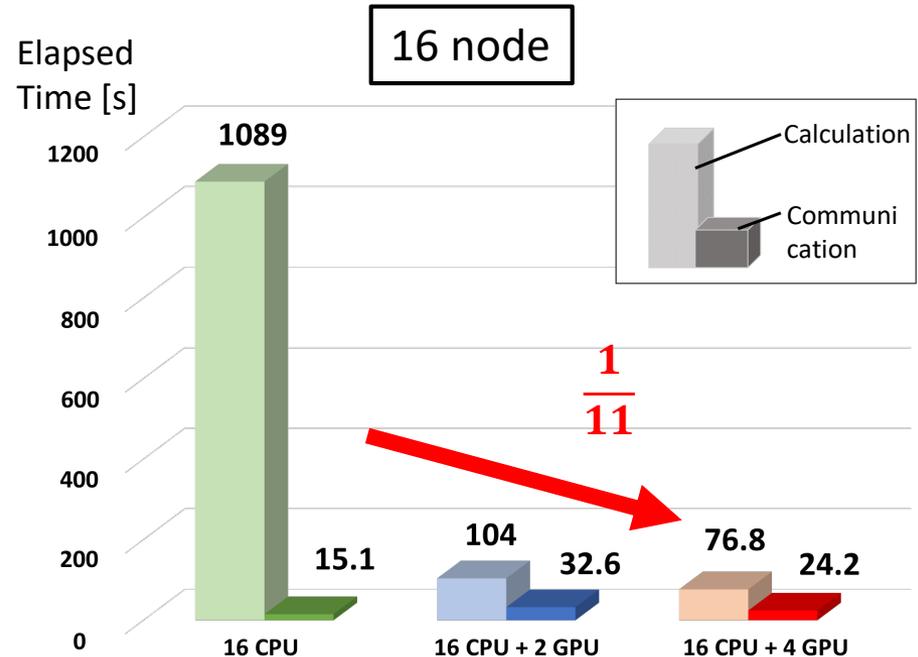
Utilization of Multi-GPU - 2

Benchmark test of Vlasov solver

Mesh number: $64 \times 64 \times 144 \times 32 \times 8$



Mesh number: $128 \times 128 \times 144 \times 64 \times 8$

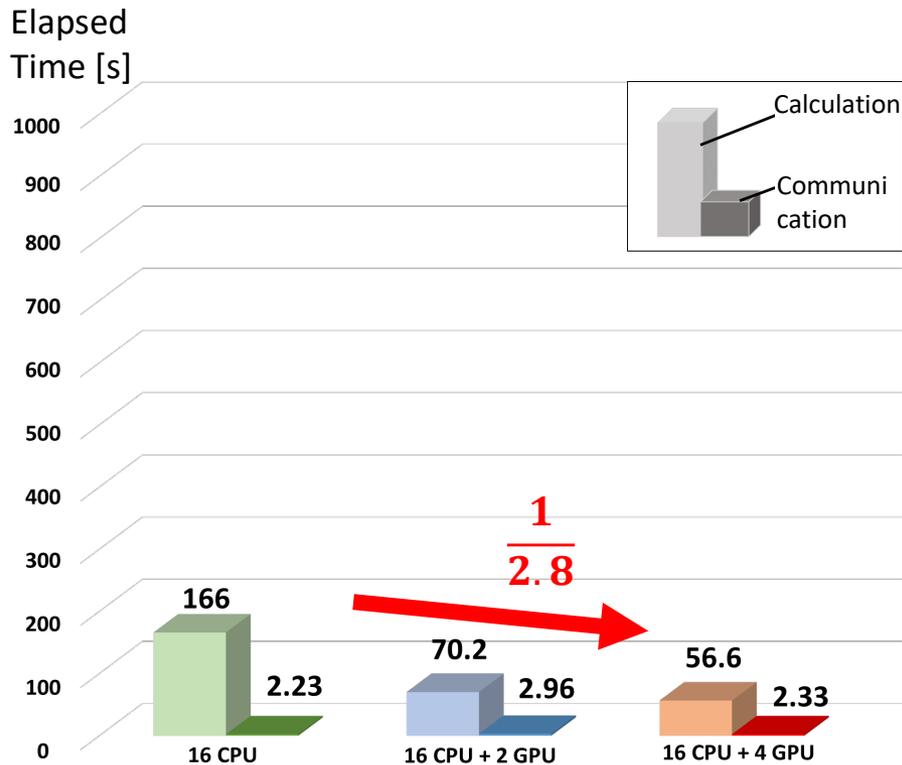


- ✓ By increasing the number of GPUs on each nodes, the calculation time is reduced, which tendency is enhanced in the larger problem size case.
- ✓ But the speed-up rate is still low because the FFT part is not GPU parallelized yet.

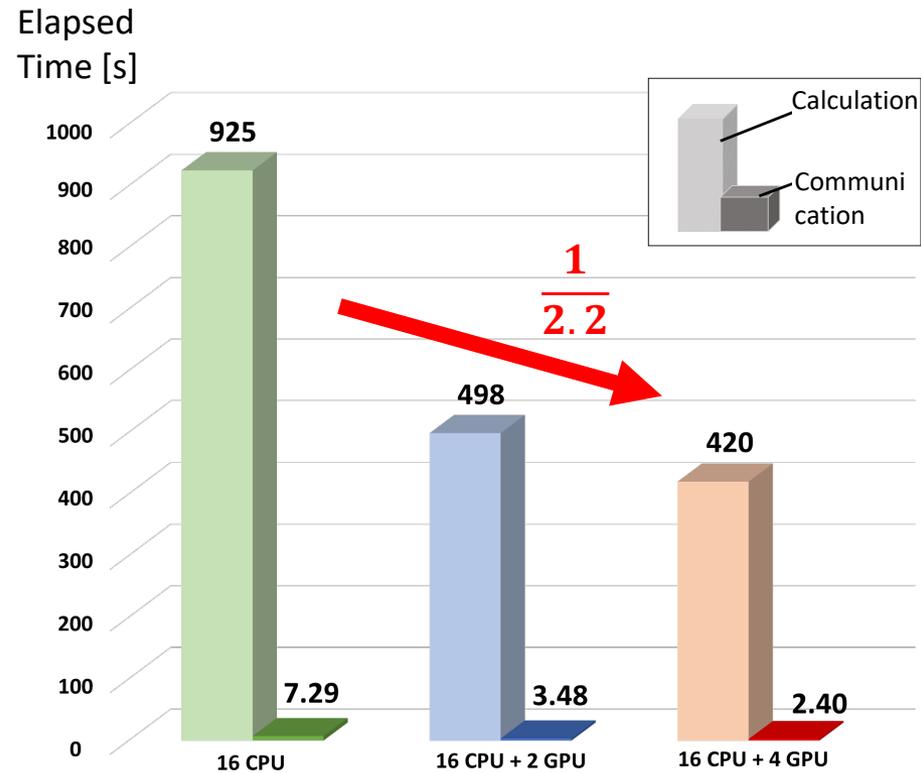
Utilization of Multi-GPU - 3

Benchmark test of Poisson solver

Mesh number: $64 \times 64 \times 144 \times 32 \times 8$



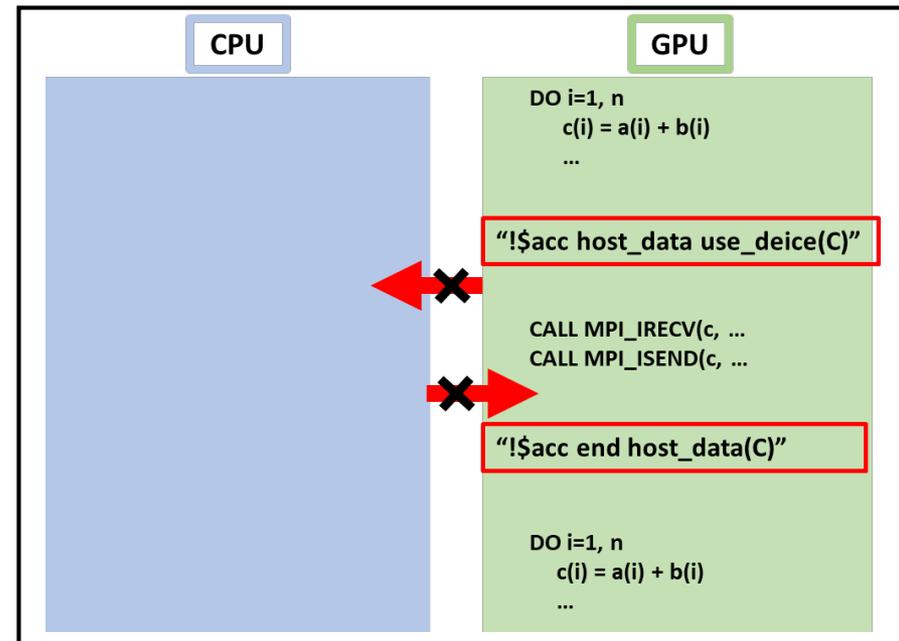
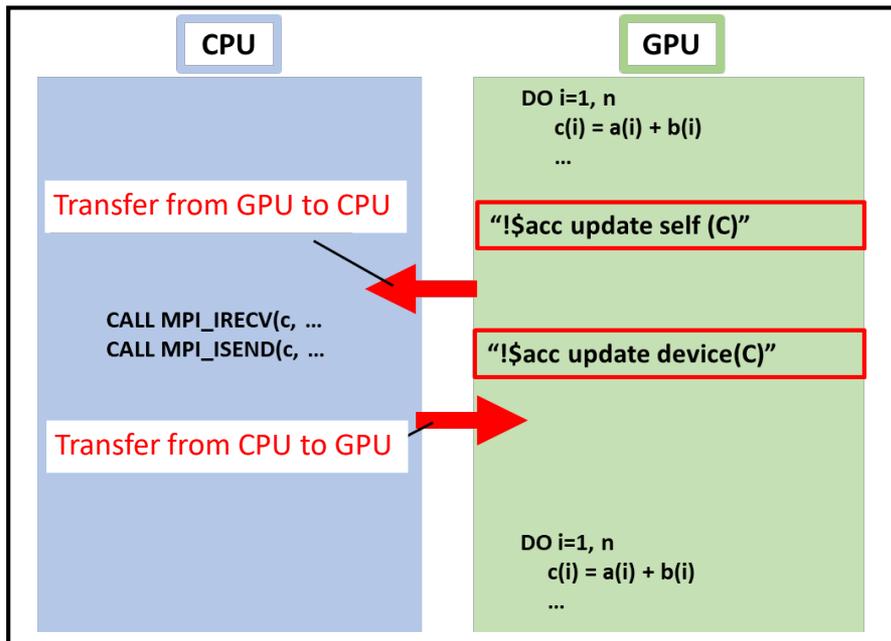
Mesh number: $128 \times 128 \times 144 \times 64 \times 8$



✓ Same tendency can be observed but the FFT part is not GPU parallelized yet.

Direct GPU-GPU Data Transfer - 1

- ✓ We transferred the boundary data from GPU to CPU before the MPI data communications (Left case).
- ✓ To improve this part, direct GPU-GPU data transfer is utilized by using “acc_host_data_use_device” (Right case).

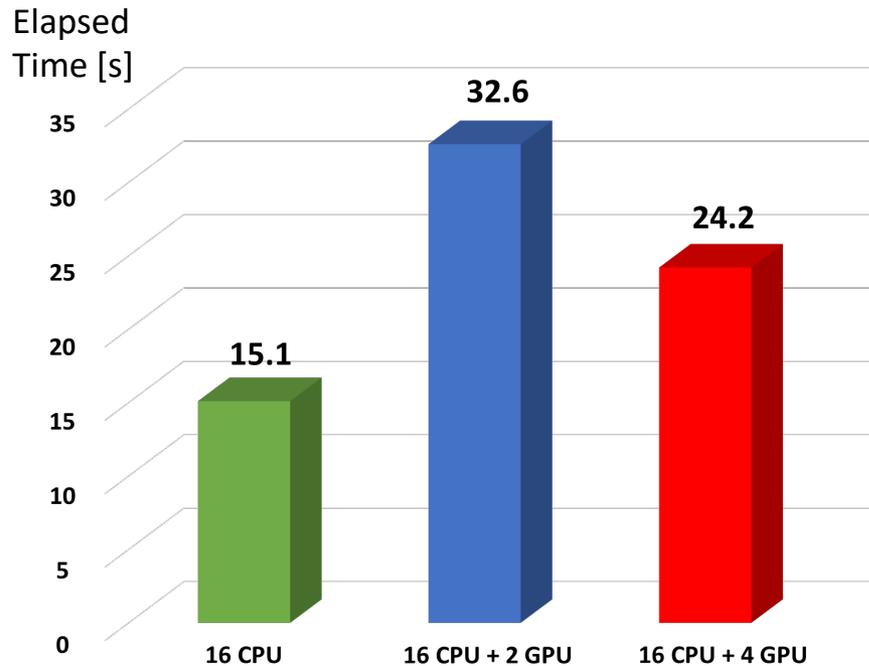


Direct GPU-GPU Data Transfer - 2

Benchmark test of Vlasov solver

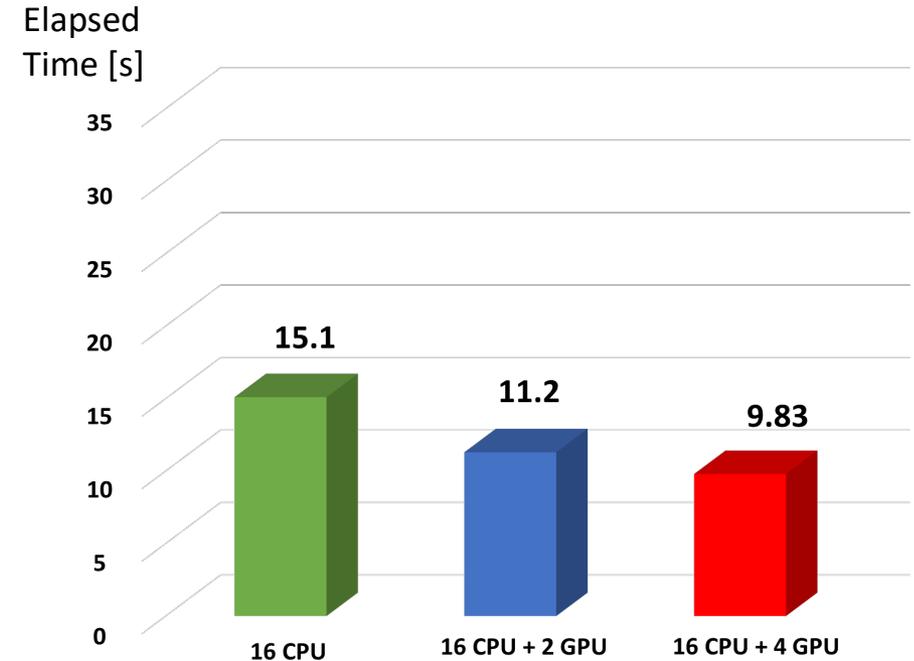
CPU-GPU communication

	16CPU	16CPU+1GPU	16CPU+4GPU
Calculation	1089	104	76.8
Communication	15.1	32.6	24.2



Direct GPU-GPU communication

	16CPU	16CPU+1GPU	16CPU+4GPU
Calculation	1089	104	76.8
Communication	15.1	11.3	9.83



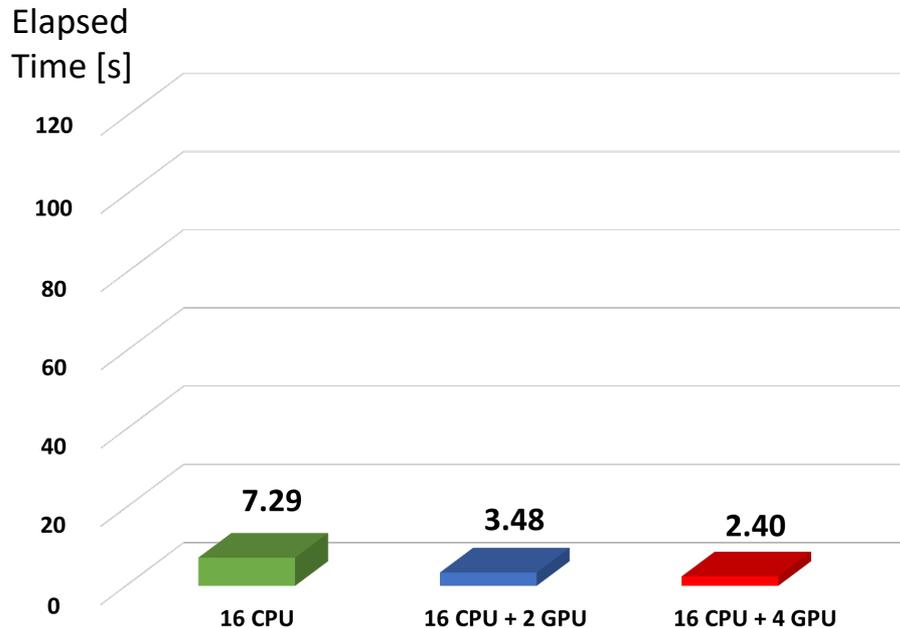
- ✓ Since the CPU-GPU data transfer is skipped by the direct GPU-GPU one, the communication time is reduced.

Direct GPU-GPU Data Transfer - 3

Benchmark test of Poisson solver

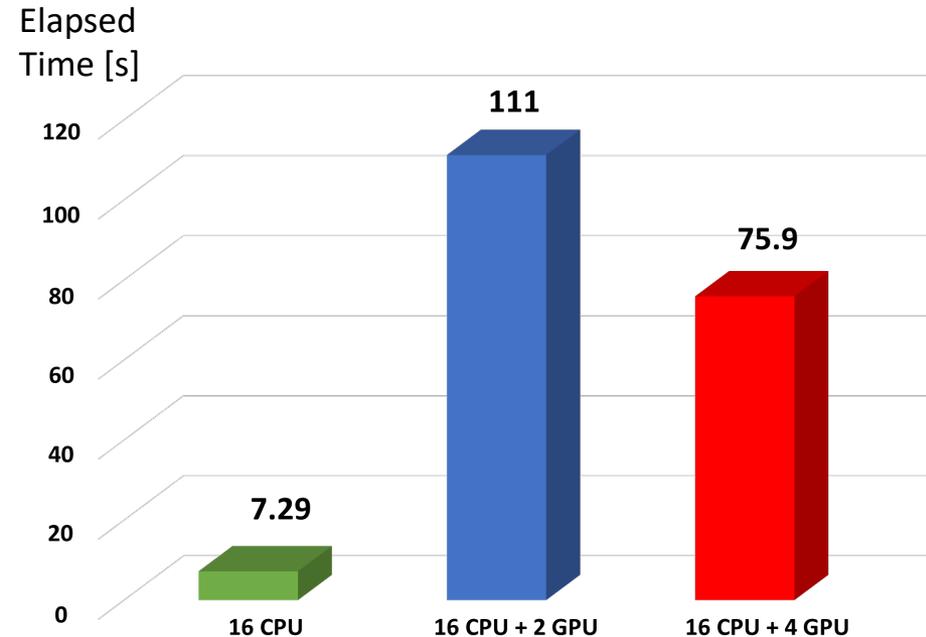
CPU-GPU communication

	16CPU	16CPU+2GPU	16CPU+4GPU
Calculation	925	498	420
Communication	7.29	3.48	2.40



Direct GPU-GPU communication

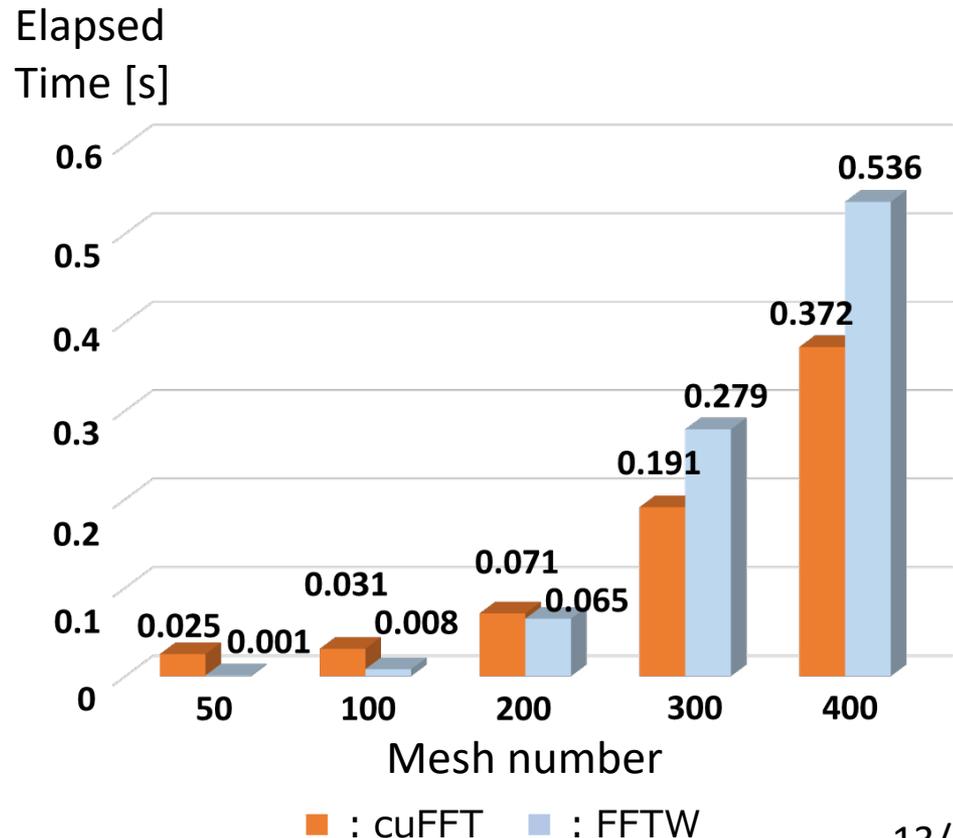
	16CPU	16CPU+2GPU	16CPU+4GPU
Calculation	925	498	420
Communication	7.29	111	75.9



- ✓ The direct GPU-GPU data transfer has not been successfully installed, but the data communication time of Poisson solver is originally small.

Installation of cuFFT - 1

- ✓ In the original GKNET, FFTW is used for FFT calculation on CPUs. To accelerate this part, cuFFT is installed.
- ✓ As is shown by the simple 3D FFT test below, the efficiency of cuFFT is confirmed only in the large-size problem.
- ✓ But, taking into account for the CPU-GPU data transfer for using FFTW, cuFFT is expected to be faster than FFTW even in the small-size problem.



Installation of cuFFT - 2

Call of cuFFT

...

```
ierr1 = cufftPlanMany(plan, 1, ..., CUFFT_Z2Z, Nbatch)
```

```
if (ierr1 /= CUFFT_SUCCESS) then  
  print *, 'cufftPlanMany: error', ierr1  
end if
```

...

```
!$acc host_data use_device(in, out)
```

```
ierr1 = cufftExecZ2Z(plan, in, out, CUFFT_FORWARD)
```

```
if (ierr1 /= CUFFT_SUCCESS) then  
  print *, 'cufftExecC2C: error', ierr1  
end if
```

```
!$acc end host_data
```

...

```
ierr1 = cufftDestroy(plan)
```

```
if (ierr1 /= CUFFT_SUCCESS) then  
  print *, 'cufftDestroy: error', ierr1  
end if
```

- ✓ Since we need 1D FFT for 3D data as follows, batched cuFFT is utilized.

$$f(x, k_y, z) = \sum_i f(x, y_i, z) \exp(ik_y y_i)$$

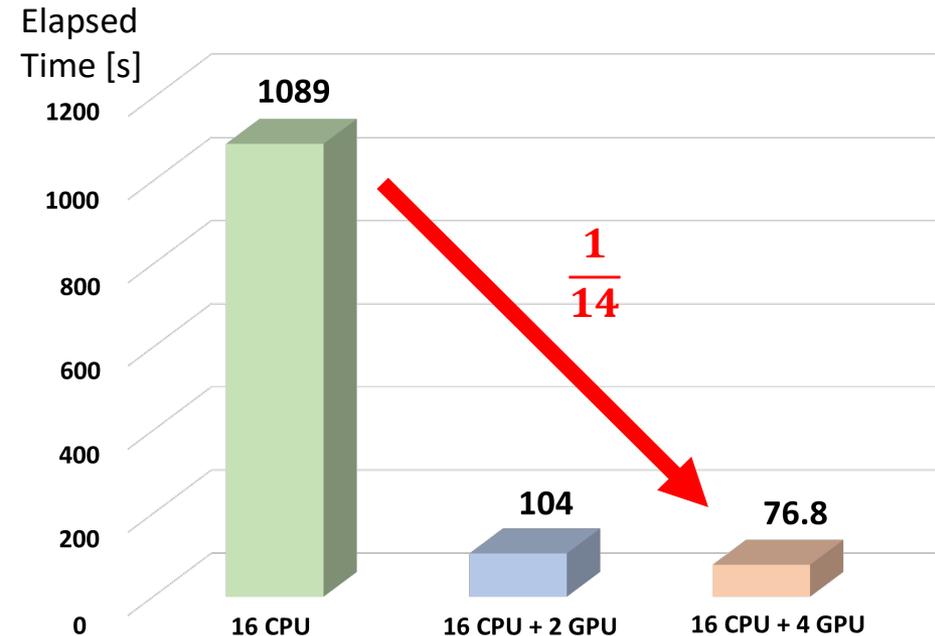
- ✓ In this case, Nbatch=N_x*N_z is set.
- ✓ In addition, direct GPU-GPU data transfer is also utilized.

Installation of cuFFT - 3

Benchmark test of Vlasov solver

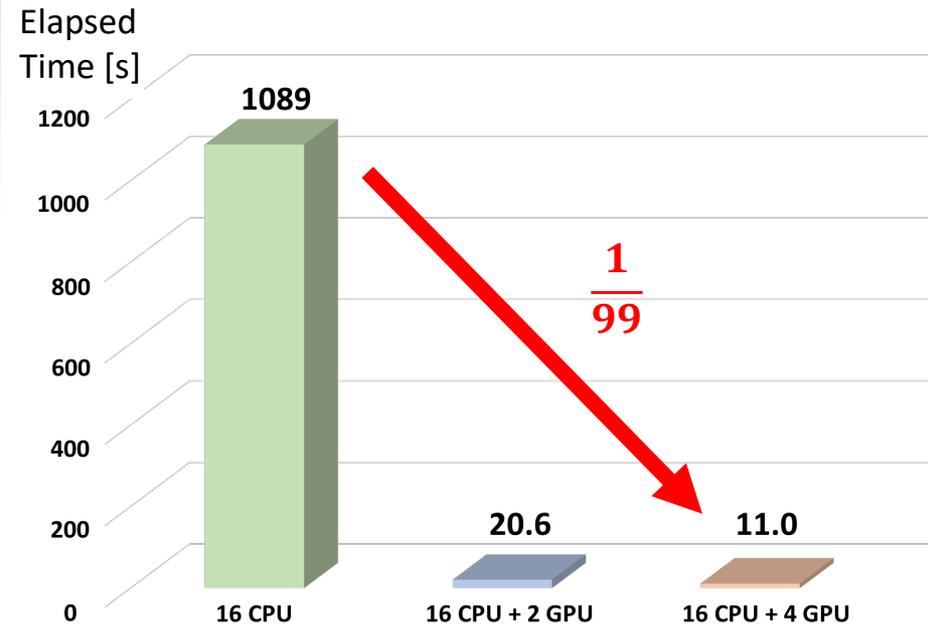
FFTW

	16CPU	16CPU+1GPU	16CPU+4GPU
Calculation	1089	104	76.8
Communication	15.1	11.3	9.83



cuFFT

	16CPU	16CPU+1GPU	16CPU+4GPU
Calculation	1089	20.6	11.0
Communication	15.1	11.3	9.83



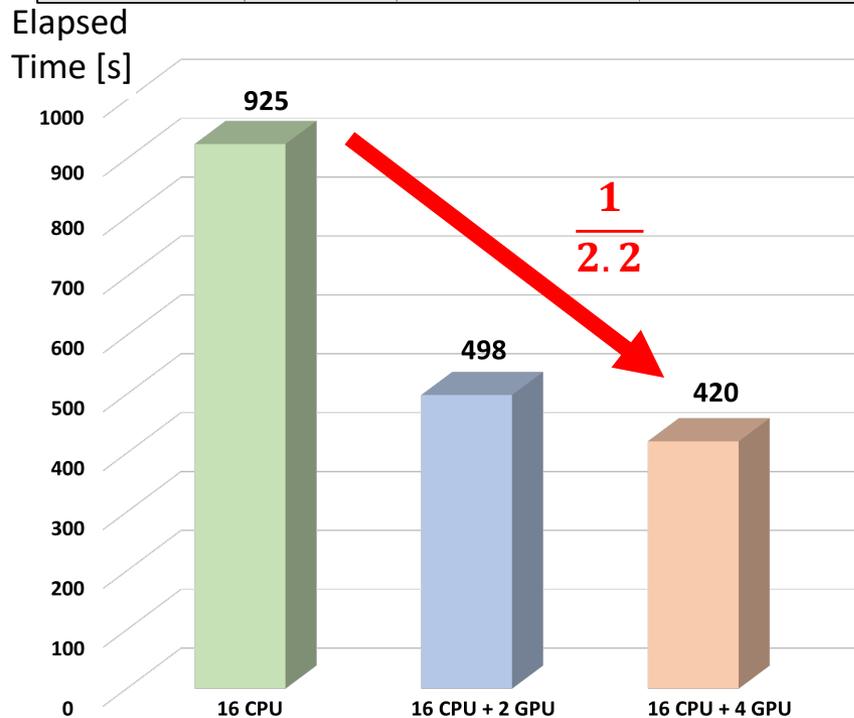
- ✓ The problem size in 3D real space is relatively low, but the calculation time is largely reduced by the installation of cuFFT because the CPU-GPU data transfer is skipped.

Installation of cuFFT - 4

Benchmark test of Poisson solver

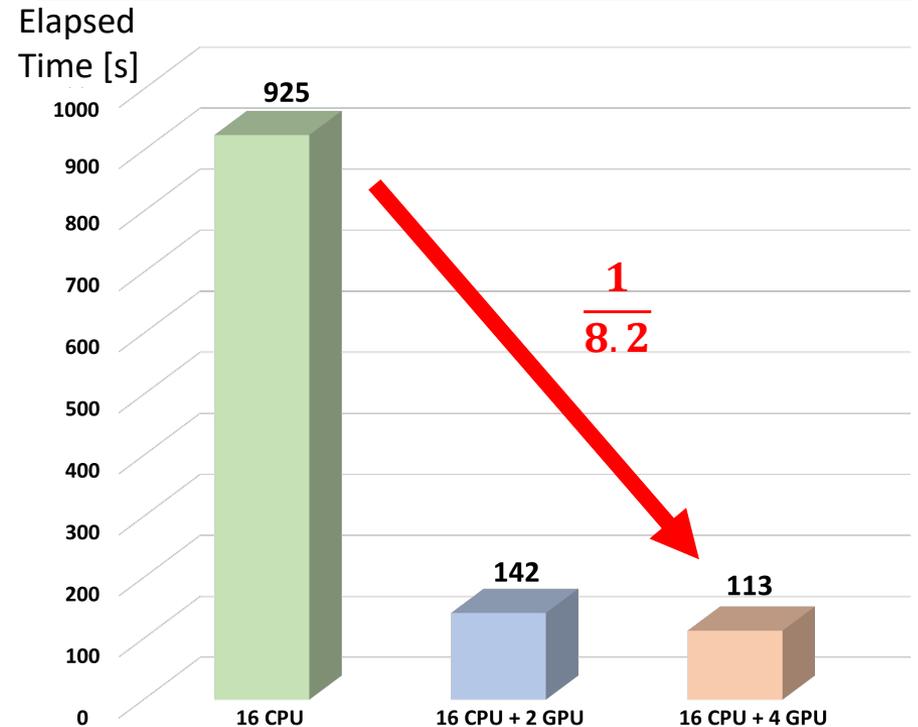
FFTW

	16CPU	16CPU+2GPU	16CPU+4GPU
Calculation	925	498	420
Communication	7.29	3.48	2.40



cuFFT

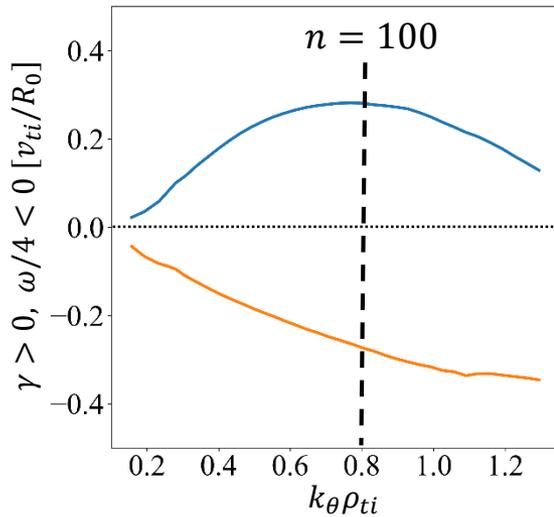
	16CPU	16CPU+2GPU	16CPU+4GPU
Calculation	925	142	113
Communication	7.29	3.48	2.40



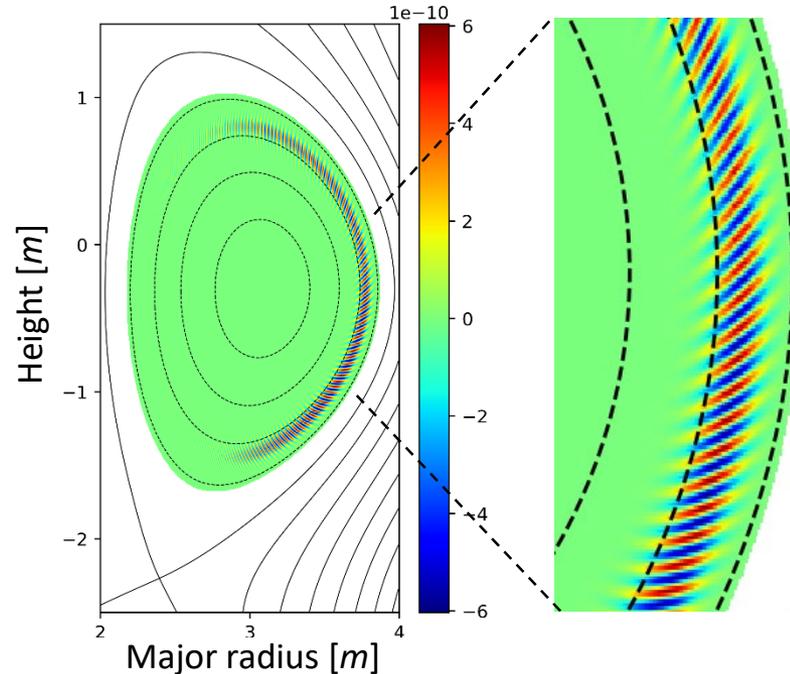
- ✓ Same tendency can be observed but the speed-up rate is still low. This is considered to originate from the 1D matrix solver for the Jacobi method.

Nonlinear Simulation of JT-60SA Plasma - 1

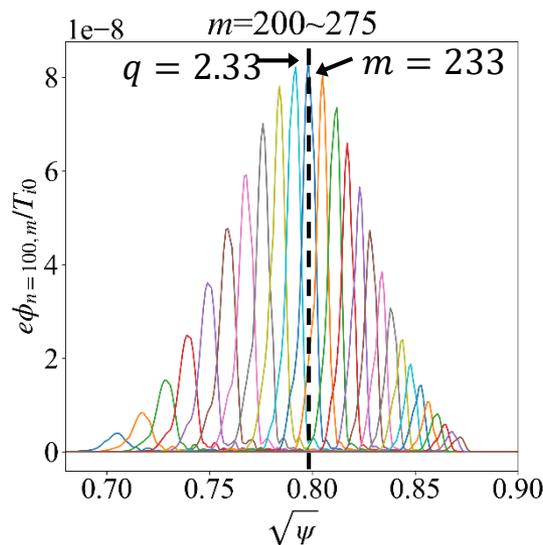
Dispersion relation



$\phi_{n=100}$ in a linear phase



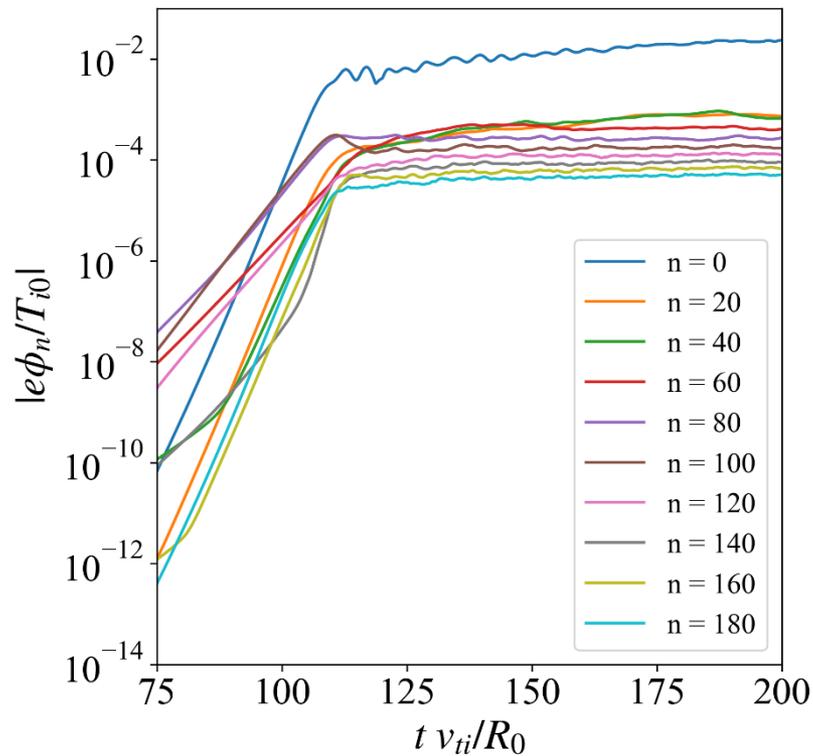
The poloidal harmonics of $\phi_{n=100}$



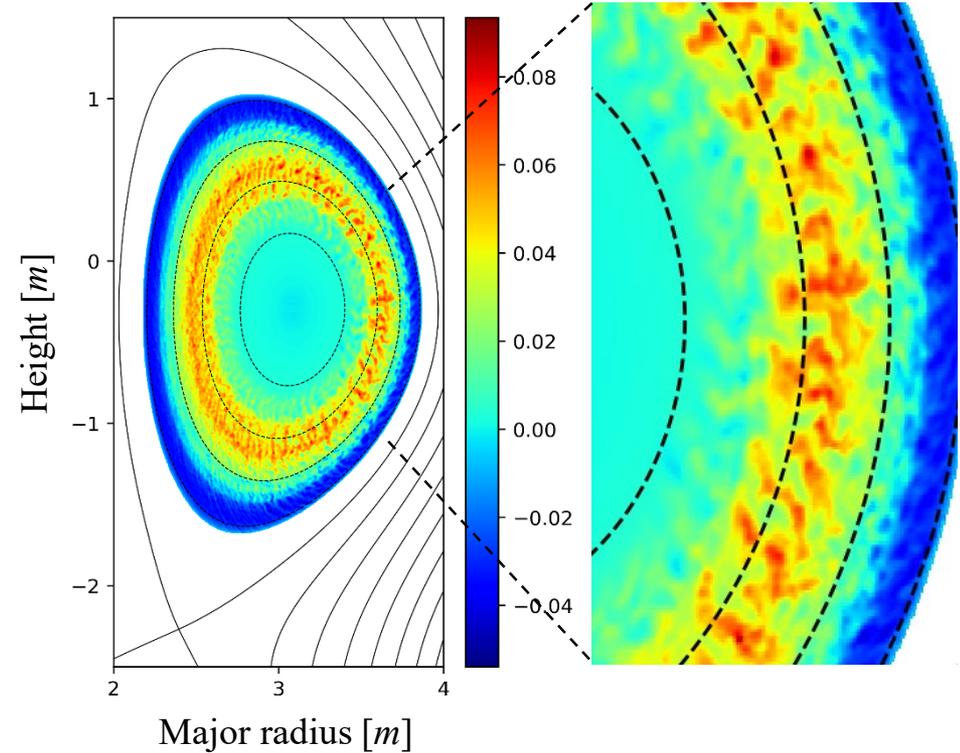
- ✓ The system size is $a_0/\rho_{ti0} = 294$ and the applied mesh number in real space is (720, 256, 32), respectively.
- ✓ Very high poloidal mode number instabilities, such as $(m, n) = (233, 100)$, have been resolved.

Nonlinear Simulation of JT-60SA Plasma - 2

Time development of the amplitude of the electrostatic potential



$\phi(\zeta = 0)$ after the nonlinear saturation



- ✓ The zonal flow has been generated at the position corresponding to the linear instabilities.

Nonlinear Simulation of JT-60SA Plasma - 3

- ✓ Consider the number of meshes required to resolve the mode $m_{\text{res}} = 160 \times 2.3 \approx 370$ resonating with $n = 160$, which is unstable in this case.
- ✓ Assuming that 8 times as many meshes as the mode number are required, we need $N_{\theta} = 370 \times 8 \approx 30$ in the flux-surface coordinate system (ρ, θ, ζ) .
- ✓ On the other hand, in the field aligned coordinate system (x, y, z) , $N_z = 32$ is enough as this case.
- ✓ By utilizing the field aligned coordinate system, the required meshes are reduced to $3000 \div 32 \approx 1/94$.

Summary - 1

Summary-1

- ✓ We have accelerated global gyrokinetic code by using OpenACC directives.
- ✓ Especially we have tried (A) the utilization of multi-GPU, (B) the direct GPU-GPU data transfer, (C) the installation of cuFFT.
- ✓ We have achieved **83 times speed-up for Vlasov solver and 9.1 times speed-up for Poisson solver.**
- ✓ As the result, **the total speed-up rate was 13.**

	Vlasov		NEUTRAL		Total
	Calc.	Comm.	Calc.	Comm.	
MPI	1089	15.1	925	7.29	2036
Multi-GPU	76.8	24.2	420	2.40	523
Multi-GPU+ Direct transfer	76.8	9.83	420	2.40	509
Multi-GPU+ Direct transfer + cuFFT	11.0	9.83	130	2.40	153

1
13

Summary - 2

Summary-2

- ✓ We have also implemented a field-aligned coordinate system to GKNET.
- ✓ Realistic tokamak geometries, including up-down asymmetric equilibria, have been also implemented.
- ✓ The result shows that the linear ITG instability with high poloidal modes and resultant zonal flow generation are properly traced. In this case, it is estimated that **the number of computational grids can be reduced to 1/94 compared to that of the flux surface coordinate system.**

Future plans

- ✓ The development of GKNET will be extended to address tokamak edge turbulence. Currently, an interface code is being developed in the SOL/divertor region.
- ✓ In addition, we will resolve some issues for GPU acceleration (Matrix solver, GPU-GPU direct transfer, etc.)

