

Caffe-MPI: A parallel Framework on the GPU Clusters

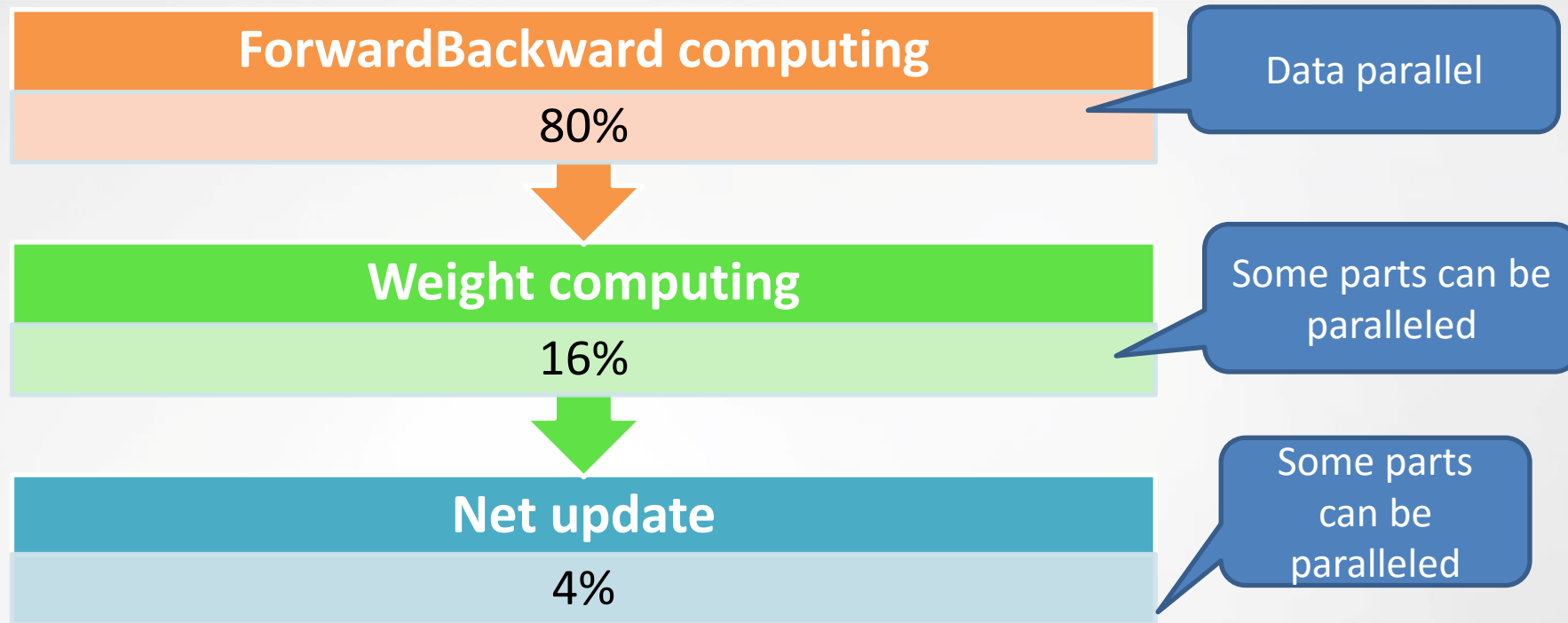
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Caffe-MPI

- What is Caffe-MPI?
 - Developed by Inspur
 - Open-source: <https://github.com/Caffe-MPI/Caffe-MPI.github.io>
 - Programmed by **MVAPICH**
 - Based on the Berkeley Vision and Learning Center (BVLC) single node version
 - A GPU Cluster version
 - Support 16+ GPUs to Train



Analysis of Caffe



- Caffe needs long training time for big data sets on a single node.

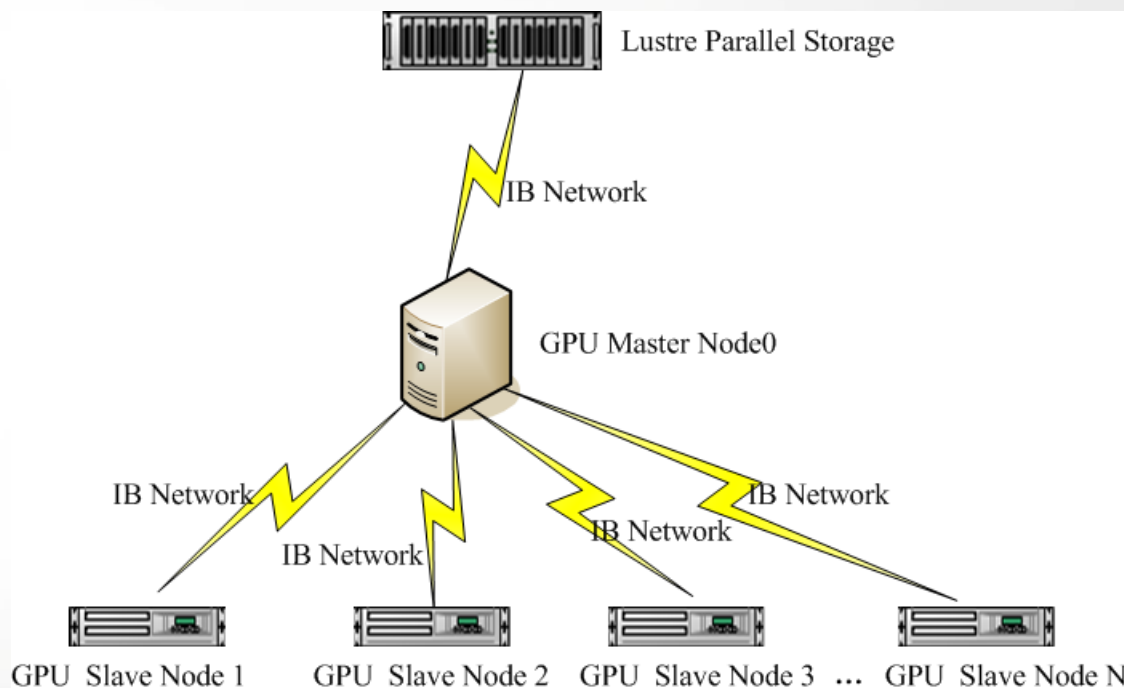


Caffe-MPI Architecture

- HPC Technology
 - Hardware arch: IB+GPU cluster+Lustre
 - Software arch: MPI+Pthread+CUDA
- Data parallel on GPU Cluster

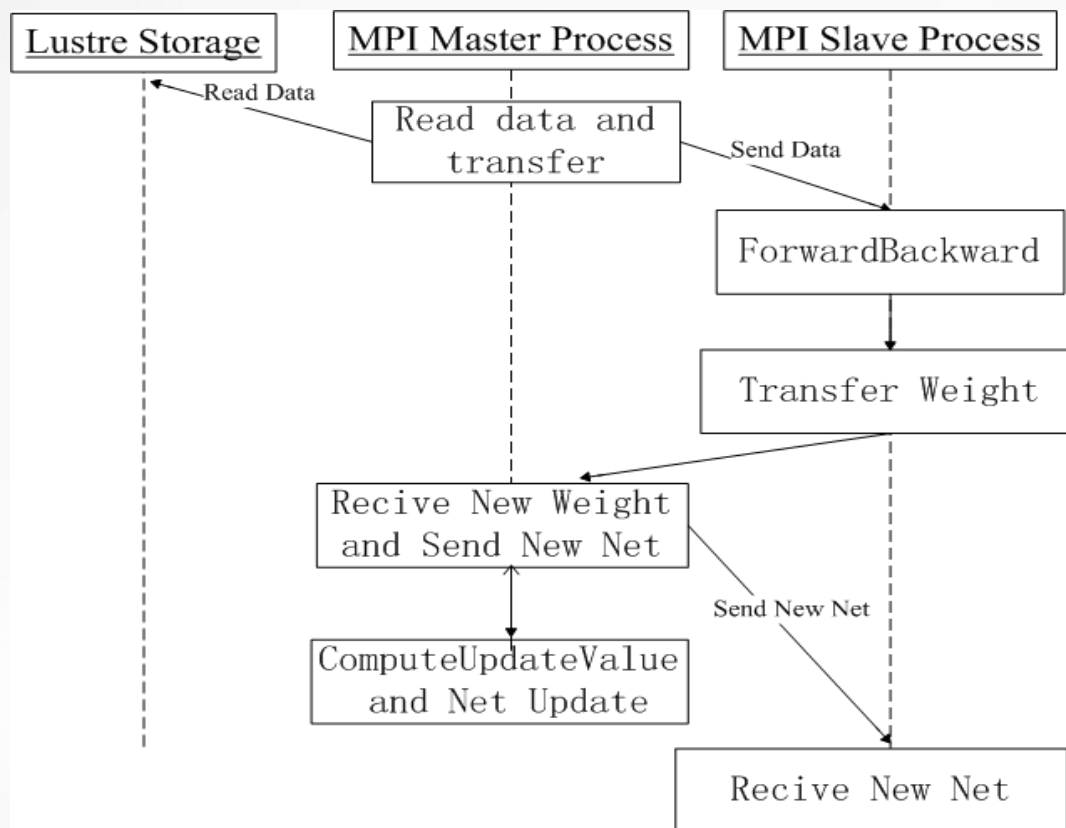
GPU Cluster Configuration

GPU master node	Multi GPUs
GPU Slave Node	Multi GPUs
Storage	Lustre
network	56Gb/s IB
Software	Linux/Cuda7.5/Mvapi ch2



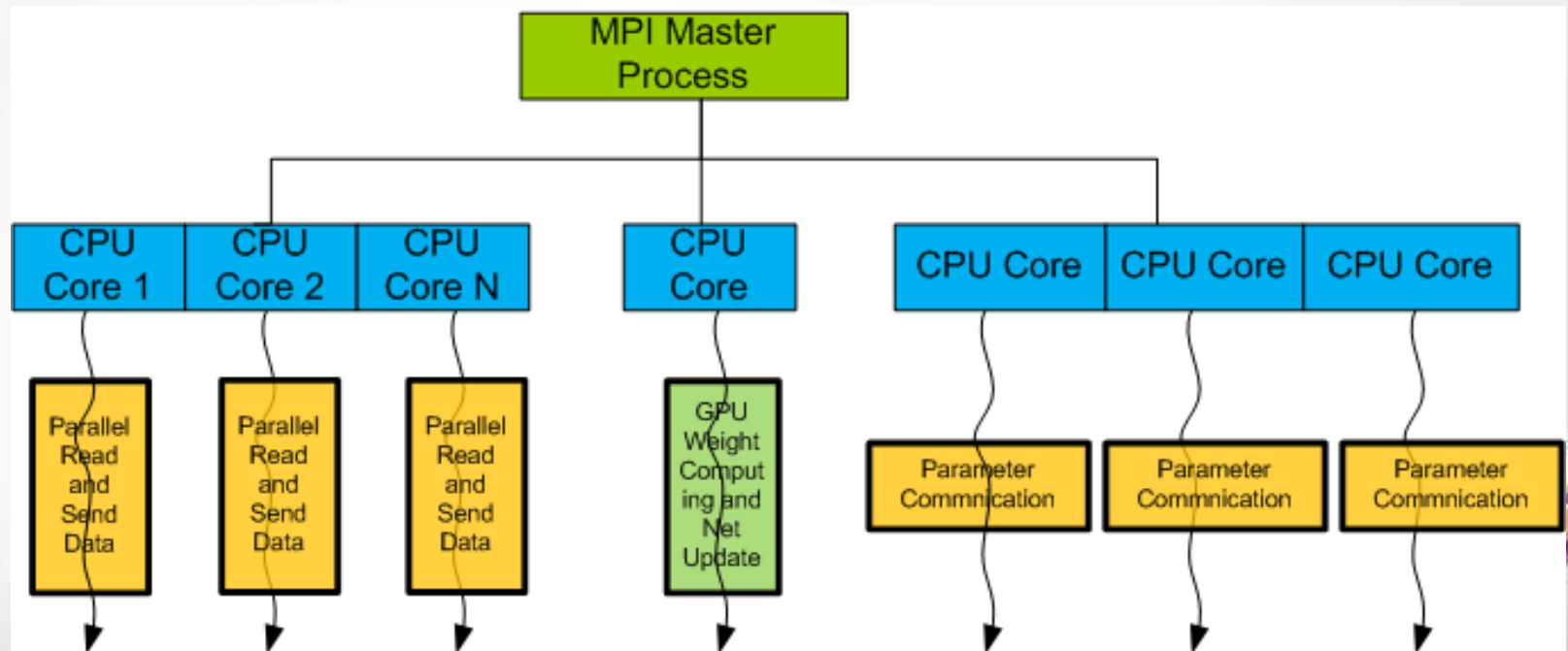
MPI Framework Design

- **MPI Master-Slave model**
 - Master Process: Multi Pthread Threads+CUDA Threads
 - Slave Process: CUDA Threads



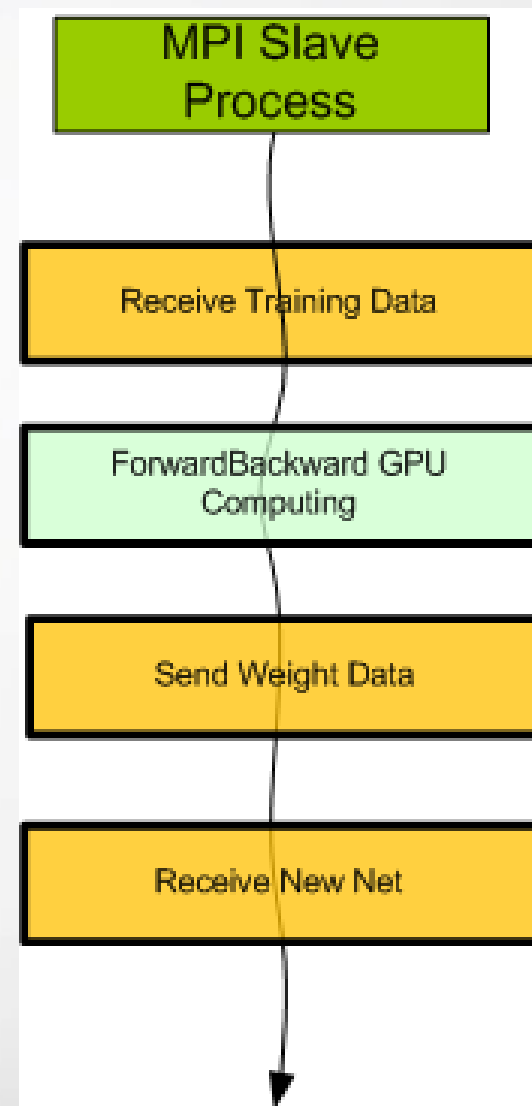
Design of Master Process

- Master Process (0 process)
 - Three functions
 - Parallel read data and send data
 - Weight Computing and The parameter update
 - The parameter communication

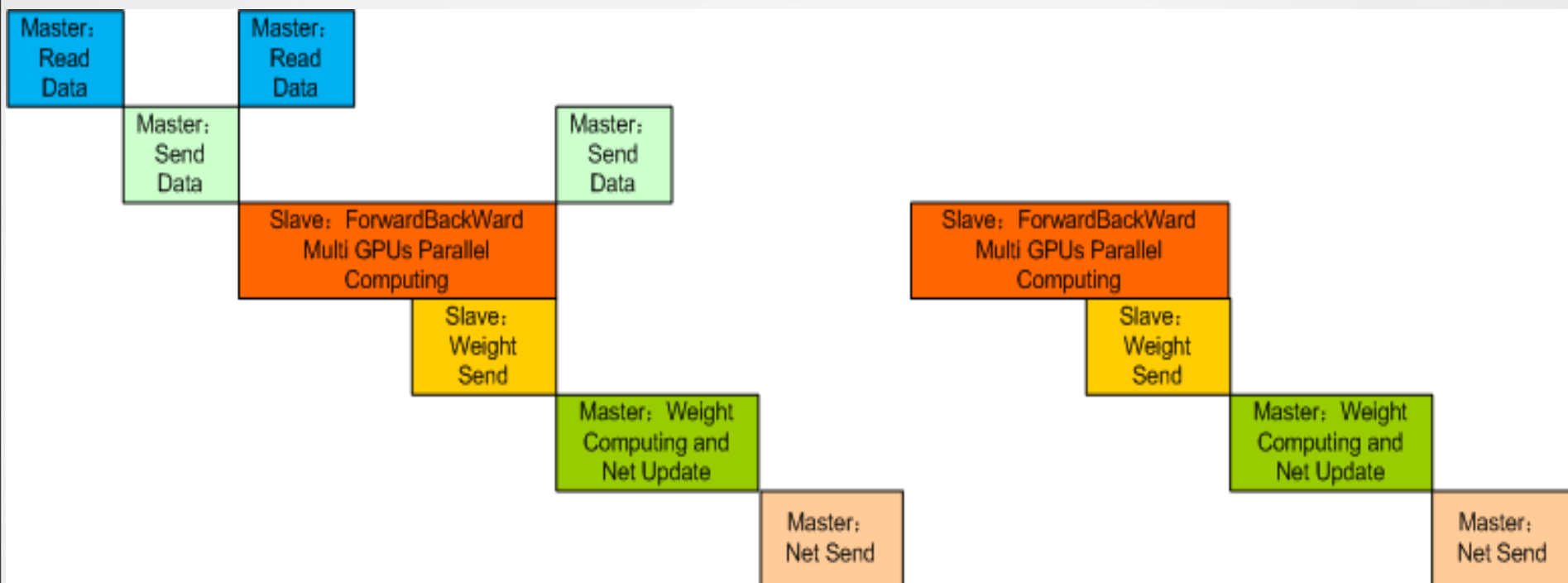


Design of Slave Process

- Slave process
 - CPU
 - To receive training data from the master process
 - To send weight data(GPU-to-GPU)
 - To receive new net data(GPU-to-GPU)
 - GPU
 - ForwardBackward computing
- Slave Node
 - The number of Slave process = the number of GPU



Features of the Computing & Communication



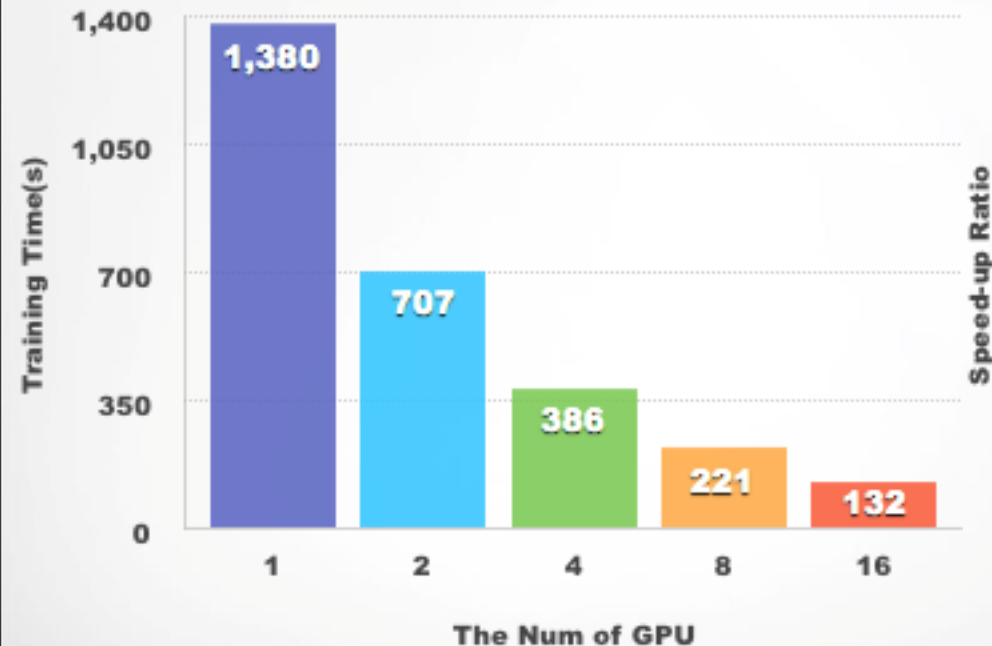
- GPU parallel computing
- Computing & Communication asynchronous parallel
- Communication Optimization
 - GPU RDMA: Weight Data and Net data between GPUs

Total Time = $\max(T_{\text{Read Data} + \text{Send Data}}, T_{\text{ForwardBackWard Computing} + \text{Weight Computing and Net Update} + \text{Net Send}})$

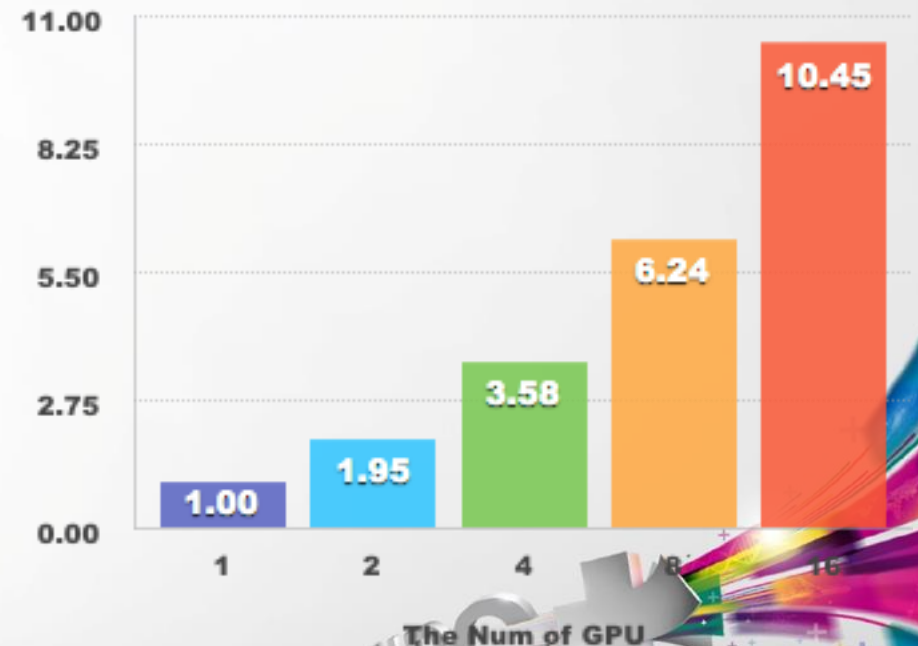
The Performance of Caffe-MPI

- Speed-up Ratio: $16\text{GPU}/1\text{GPU}=10.45\text{X}$
- Scalability efficiency: 65%

GoogleNet(Iteration=1000,BatchSize=64)



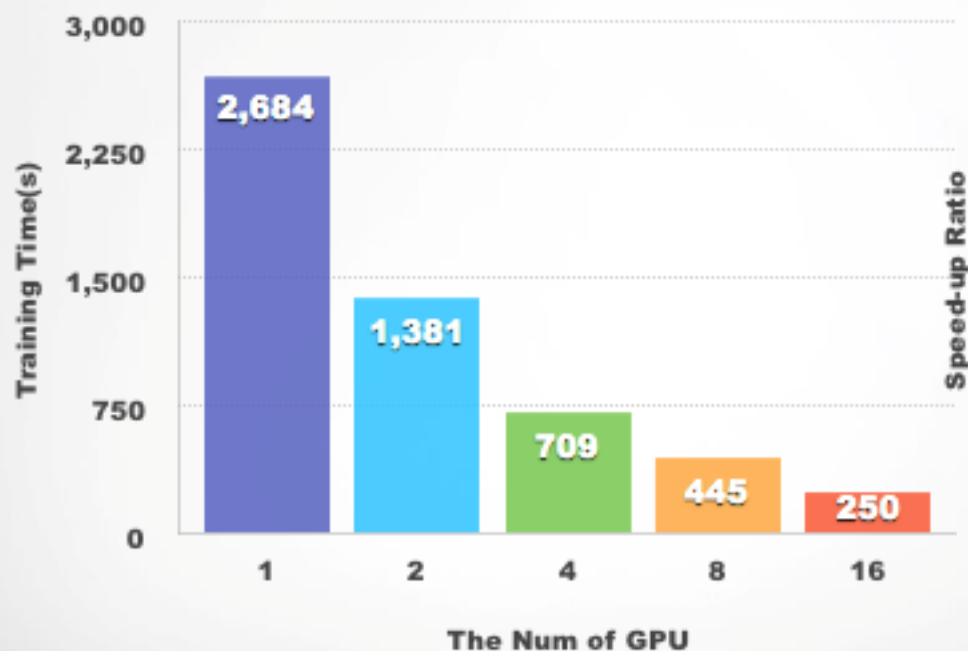
GoogleNet(Iteration=1000,BatchSize=64)



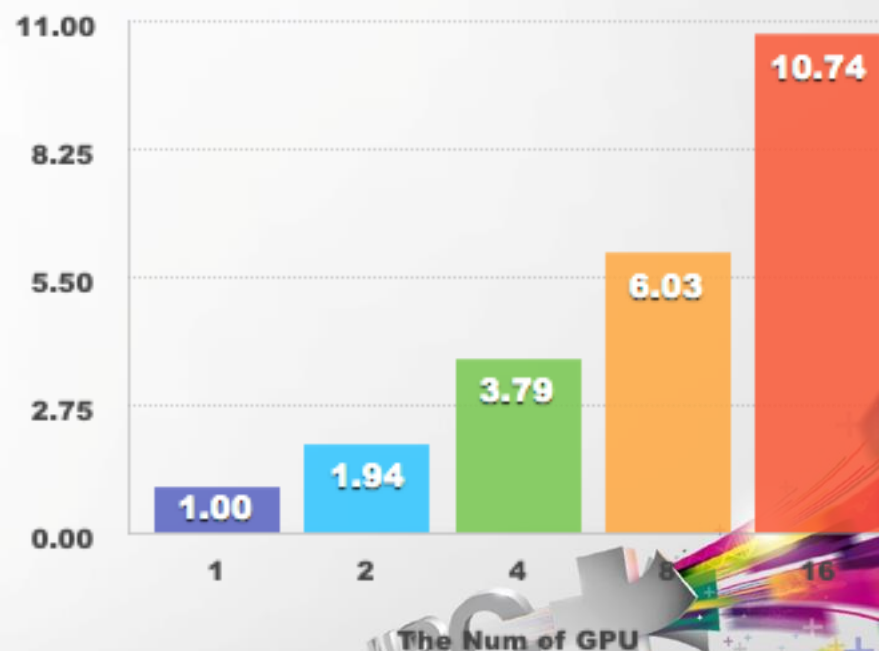
Tuning 1 : Change BatchSize

- Speed-up Ratio: $16\text{GPU}/1\text{GPU}=10.74\text{X}$
- Scalability efficiency: 67%

GoogleNet(Iteration=1000,BatchSize=128)

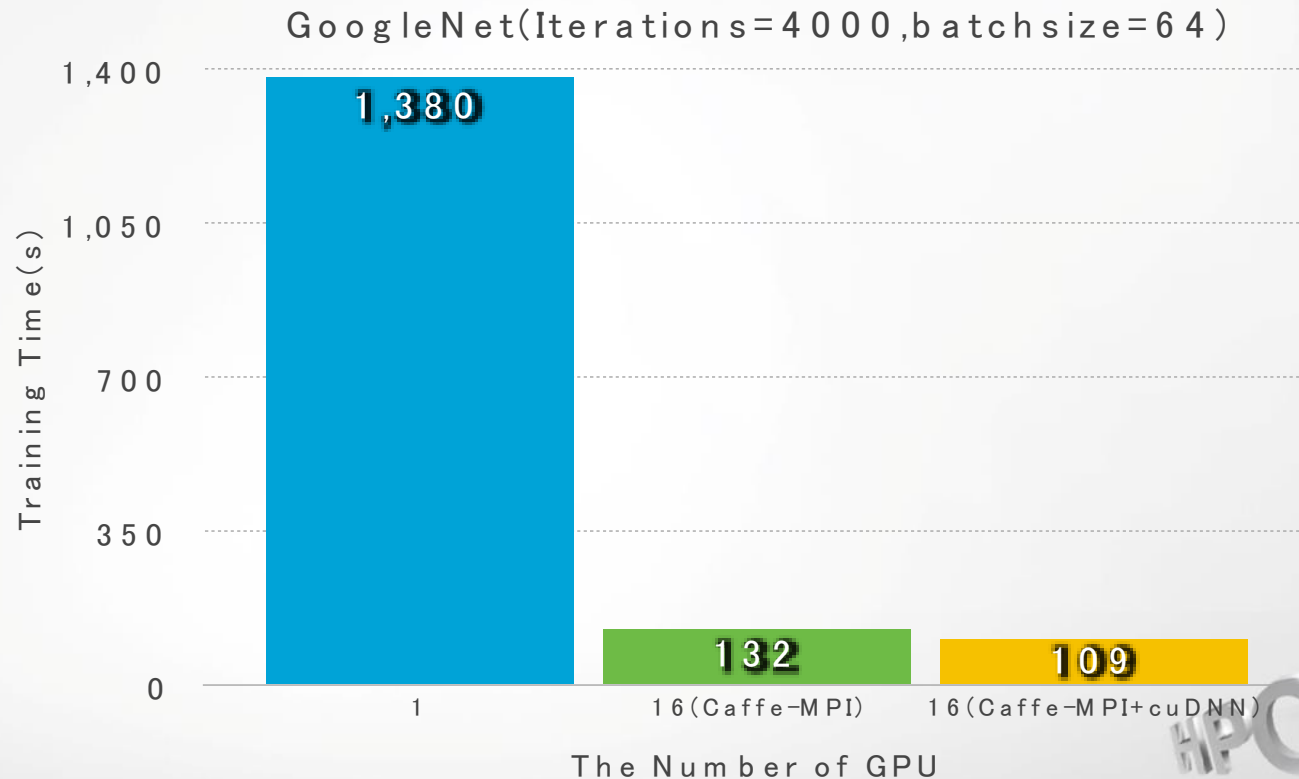


GoogleNet(Iteration=1000,BatchSize=128)



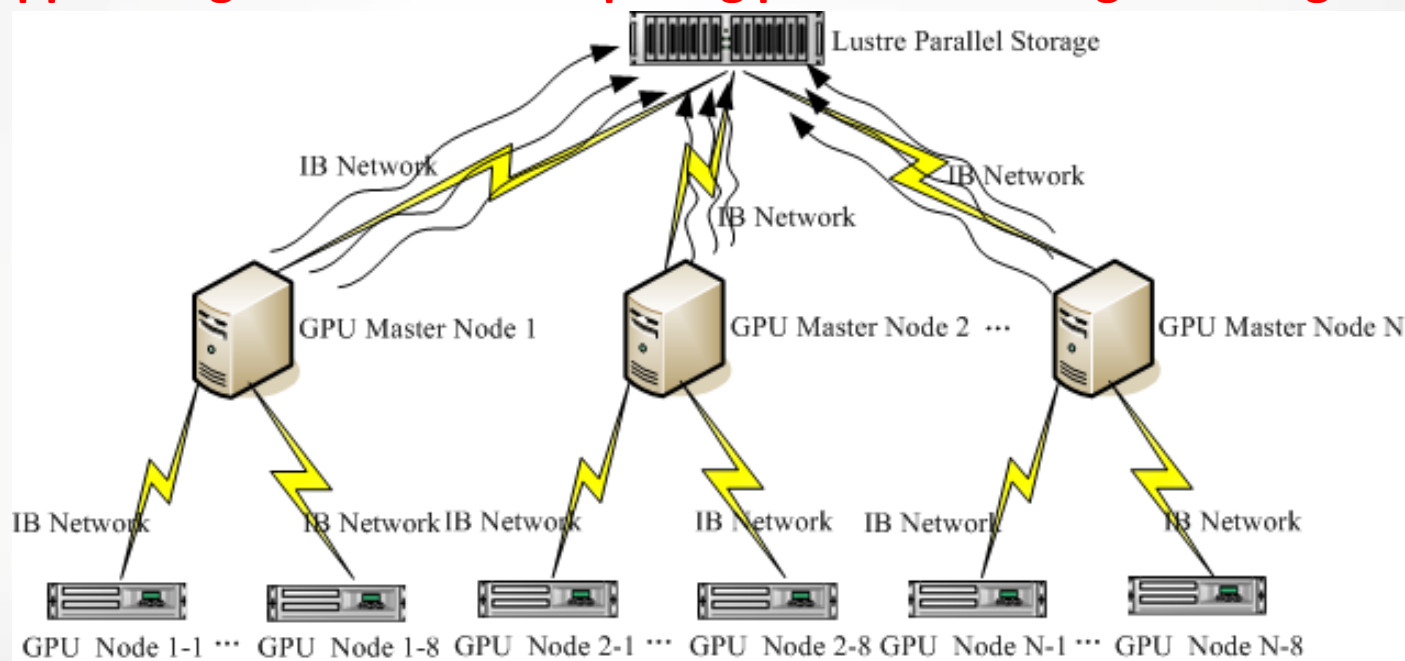
Tuning 2 : Caffe-MPI+cuDNN

- 21% Performance improvement by cuDNN
- Speed-up: 16GPU vs. 1GPU = 12.66x
- Scalability: 79%



Tuning 3: Parallelizing Read and Send Data

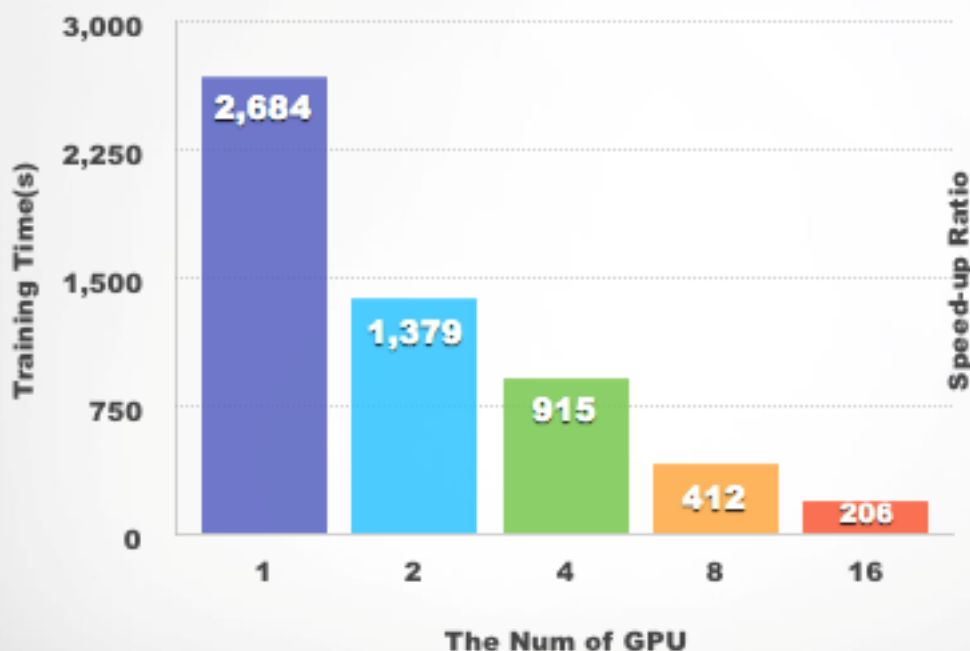
- Parallelizing read training data from Lustre Storage and send data to different GPUs
 - GPU Cluster is divided into sub groups
 - Each group has a master node
 - Each master node read and send data in parallel with Multi Processes and Multi Threads
- **Support large-scale GPU computing platform for large training data set**



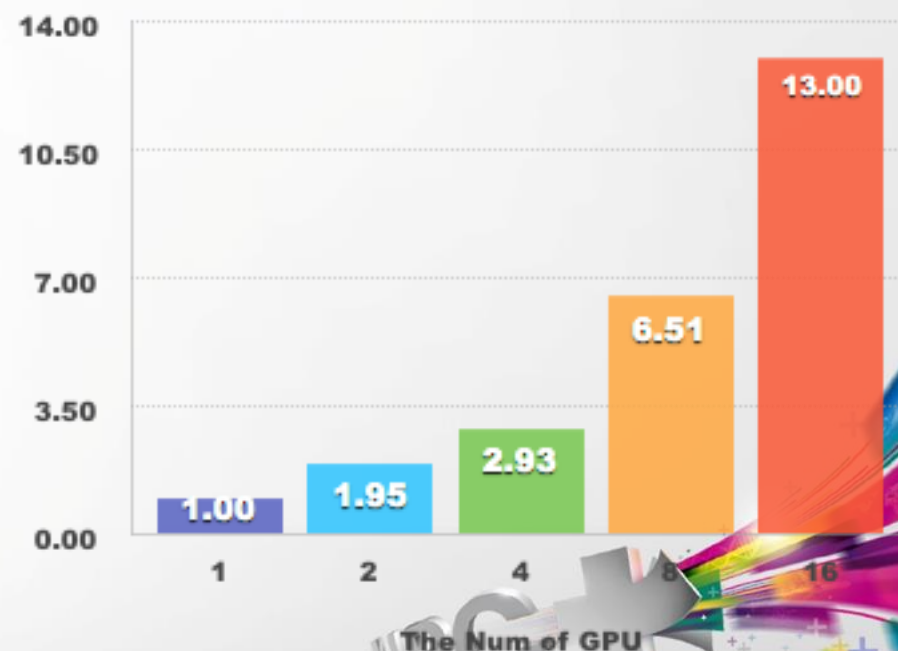
The Performance of Caffe-MPI

- Speed-up Ratio: $16\text{GPU}/1\text{GPU}=13\text{X}$
- Scalability efficiency: 81%

GoogleNet(Iteration=1000,BatchSize=128)



GoogleNet(Iteration=1000,BatchSize=128)



Caffe-MPI Plan

- Plan:
 - Support cuDNN 4.0
 - MPI tuning
 - Symmetric model



Conclusions

- Caffe-MPI
 - 13x performance improvements: 16 GPU vs. 1GPU
- Support 16+ GPU for large data sets
 - Improved master-slave model
- Open source



THANKS

