

Parallel Processing

Thoai Nam

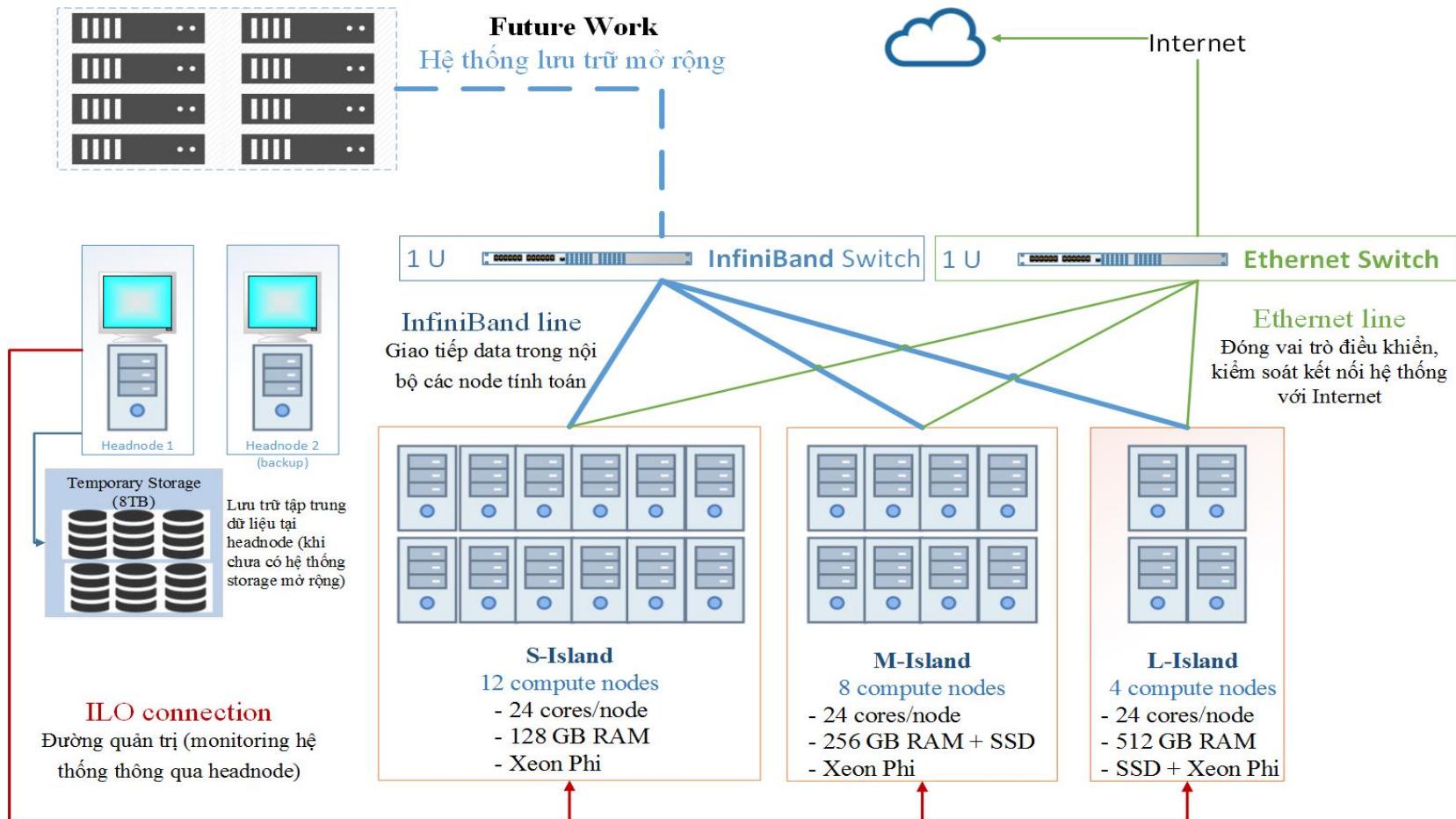
Faculty of Computer Science and Engineering

HCMC University of Technology



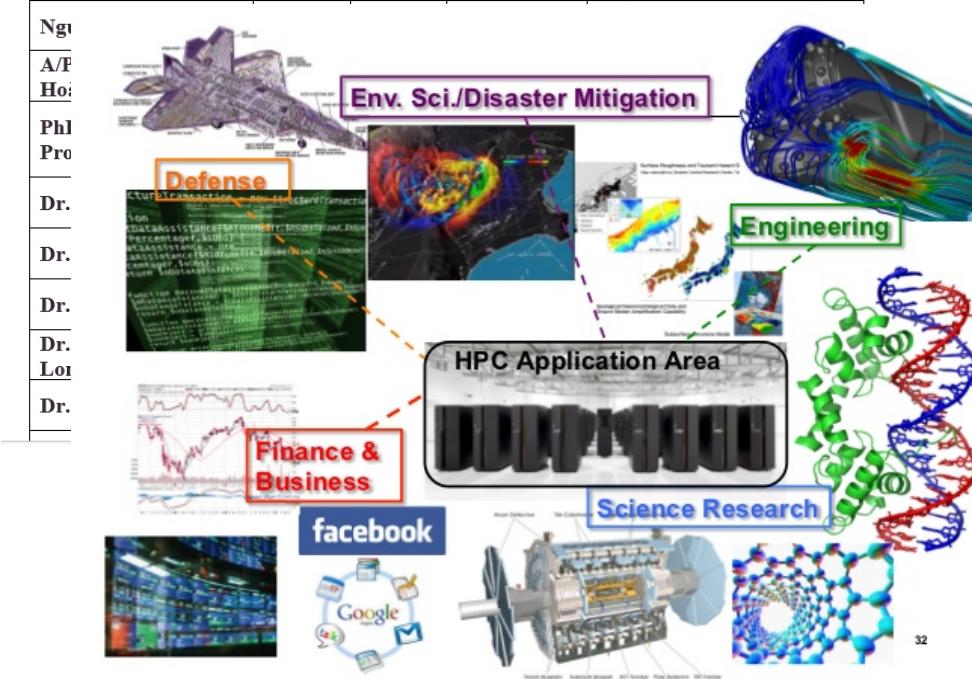
SuperNode-XP

50 TFlops machine

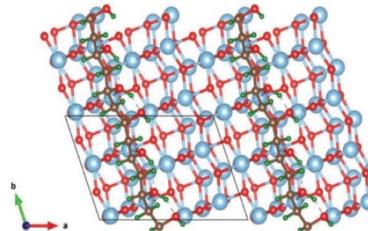


Applications on SuperNode-XP

User	Start Time	End Time	Apps	Organization/Unit
A/Prof. Nguyễn Thông	30/09/14		OpenTelemac	HCMUT – Faculty of Civil Engineering
Ms. Phạm Ngọc Thanh, Dr. Ing Jorg Franke	06/06/16		OpenFoam	Vietnam Germany University
Dr. Lê Thành Văn	20/08/16		Hadoop	HCMUT – Faculty of Computer Science
A/Prof. Trần Văn Hoài	25/10/16		BLAS	HCMUT – Faculty of Computer Science
Mr. Quân	01/11/16		DFFT	HCMUT – Faculty of Electrical and Electronic Engineering



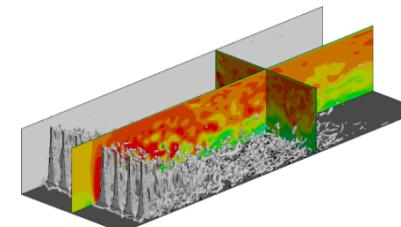
INOMAR center



VASP, Quantum Espresso Simulations

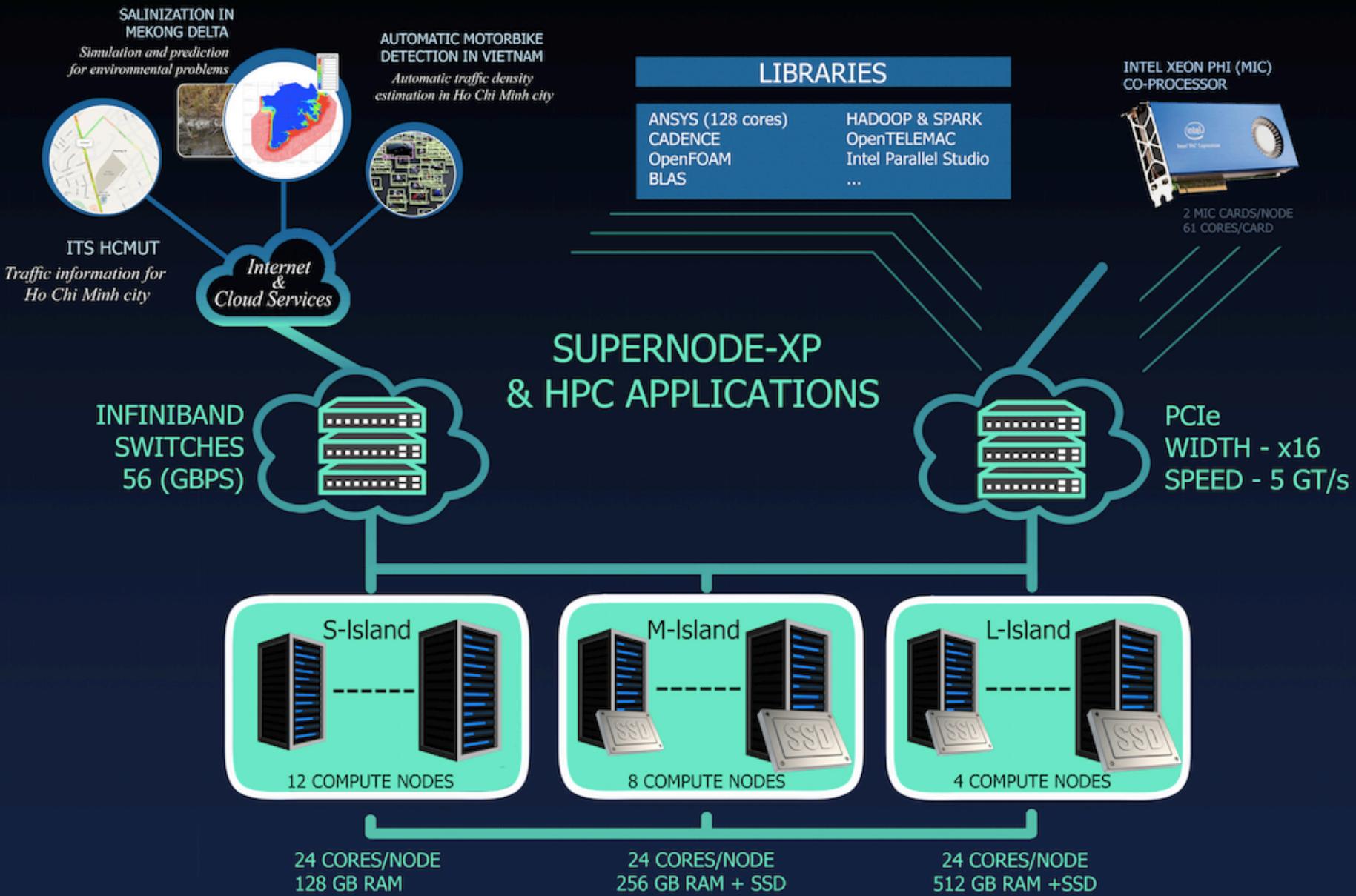
A/Prof. Dzung Hoang

Vietnam Germany University



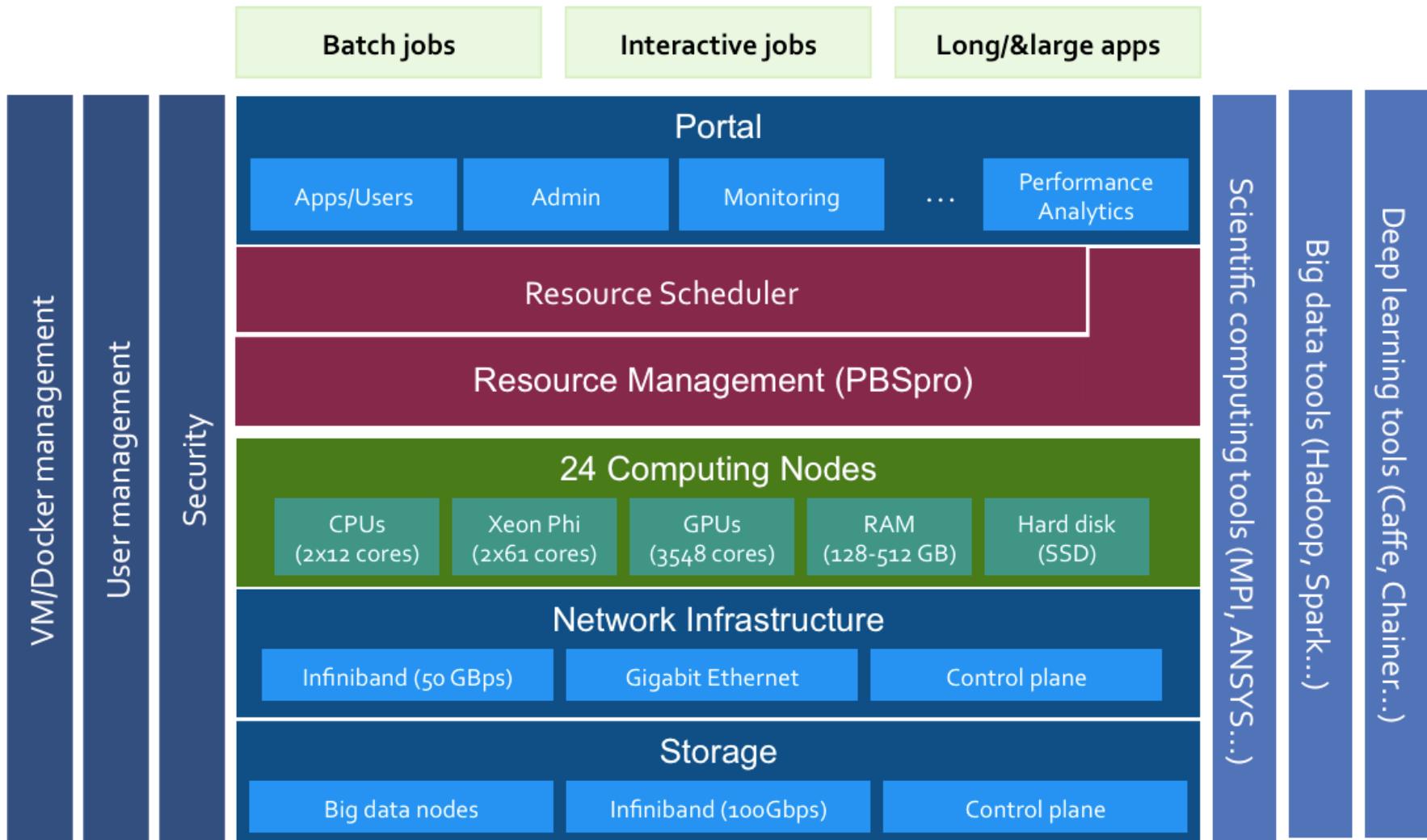
OpenFOAM sims

Prof. Joerg Franke

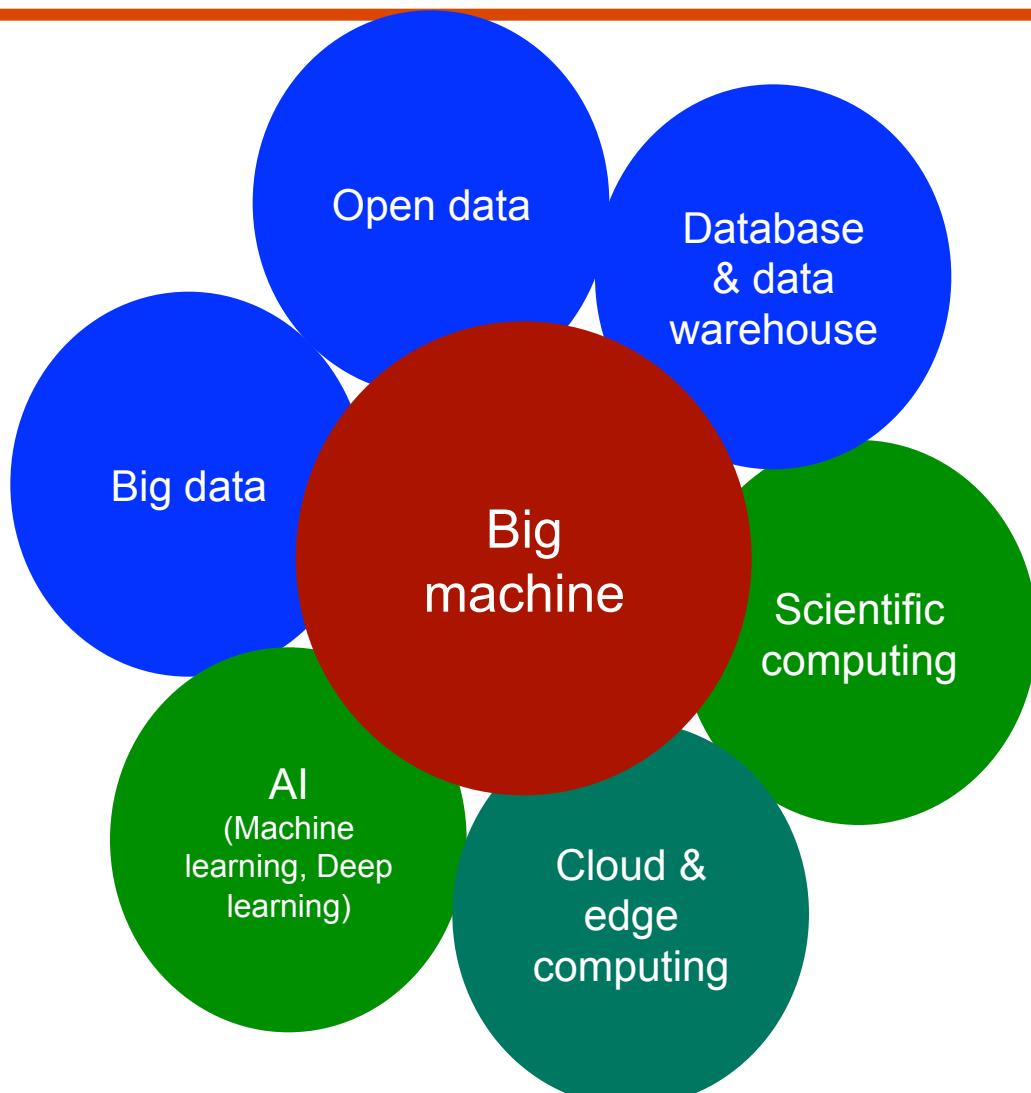




SuperNode-XP Architecture



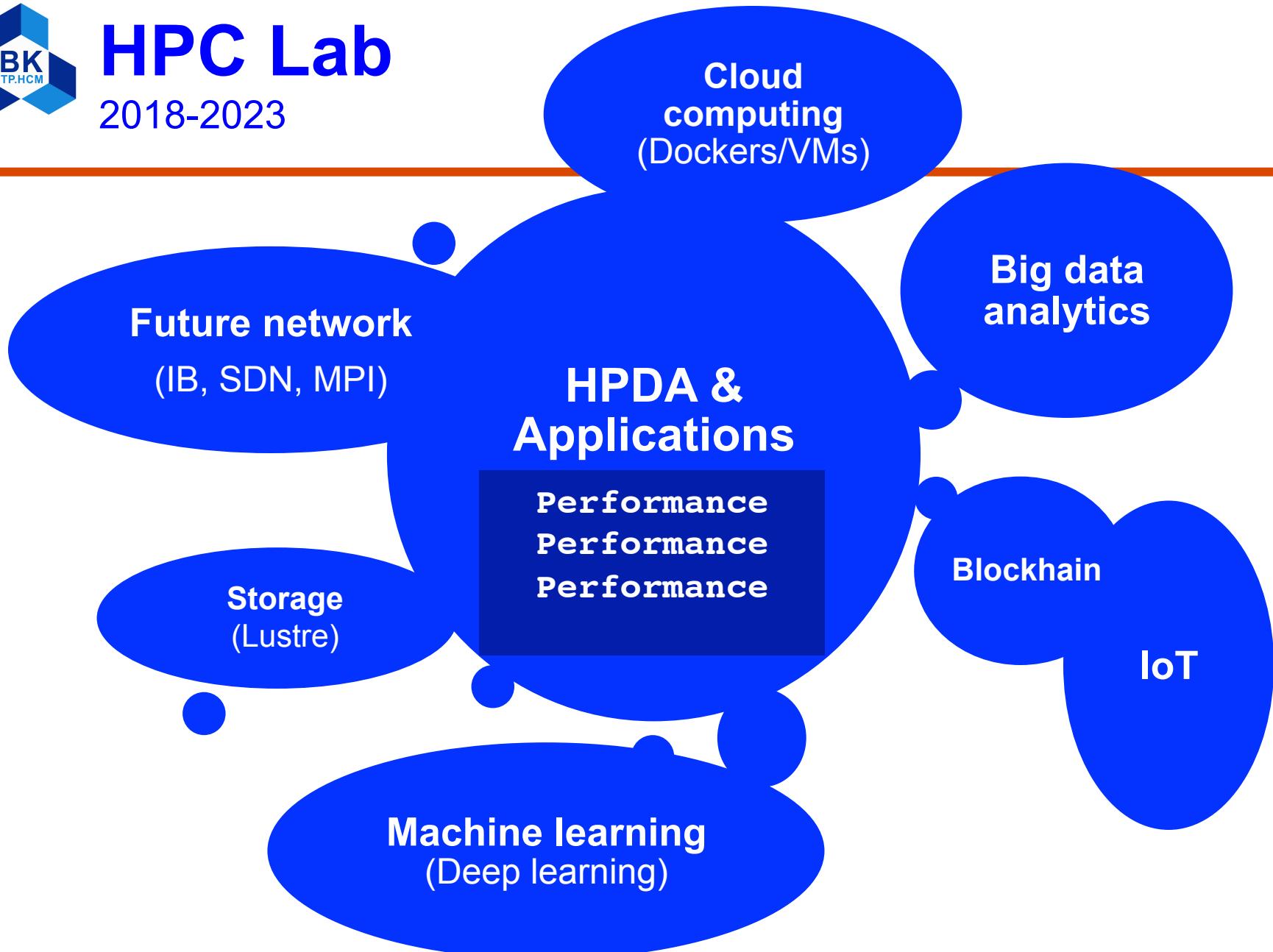
Smart cities: Data center





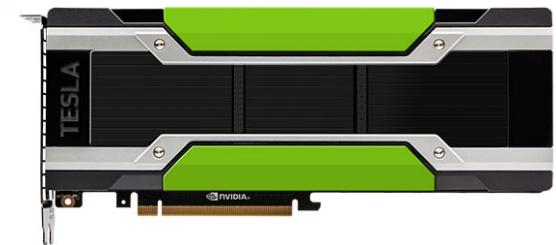
HPC Lab

2018-2023



HPDA: Big data analytics (Lrz-Germany, Intel+HPE)

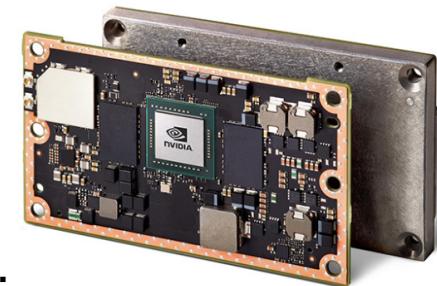
- SuperNode-XP
 - Xeon + Xeon Phi, GPUs
 - IB, Lustre storage, SSD
- Technology
 - Hadoop, Spark, PBSpro
 - Edge computing, Docker/VMs
 - Intel@ Parallel Studio XE
 - Libraries: ANSYS, OpenTelemac, Gromacs...
- Applications
 - (1) Data mining for large-scale data sets: Association rules, K-means, SVM...
 - (2) Security analytics
 - (3) Real problems in HCNC & VN on SuperNode-XP



Machine Learning + IoT

(Dr. Lê Thành Sách + Nvidia)

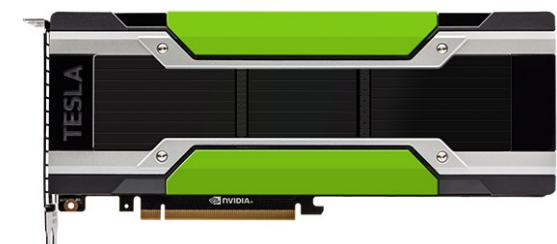
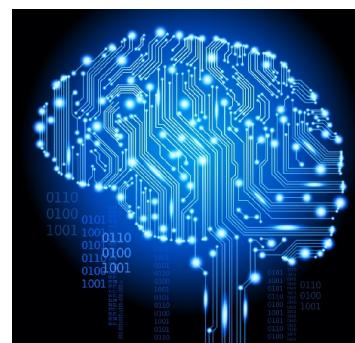
- SuperNode-XP + GPUs
 - P100, P4, GTX 1080Ti, Jetson TX2
- ML: Deep Learning
 - TensorFlow, Torch, Theano, Caffe, Chainer...
- Applications
 - (1) Image/data analytics
 - (2) Computing box (Edge computing)
 - (3) Real problems in HCMC & VN



Jetson TX2



GTX 1080 Ti



P100, P4



Future Network

(Dr. Phan Trường Khoa (UCL – UK))

- SDN (Software-Defined Networking)
 - P4: a language for programming the data plane of network devices
 - OpenFlow, OpenDayLight
- MPI
 - MPI one-sided communication
- Applications:
 - Routing analysis
 - Security
 - Optimization for big data applications
- Ref:
 - P4: <http://p4.org/>
 - IEEE SDN: <http://sdn.ieee.org>
 - OpenFlow: <http://archive.openflow.org>
 - OpenDayLight: <https://www.opendaylight.org>

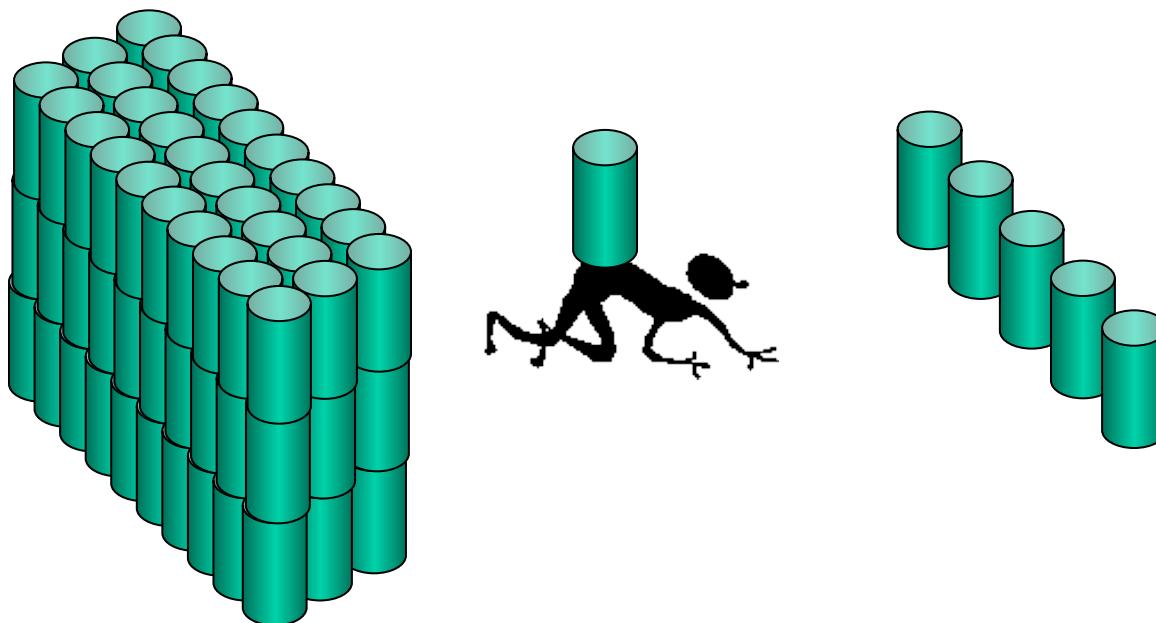


How to do

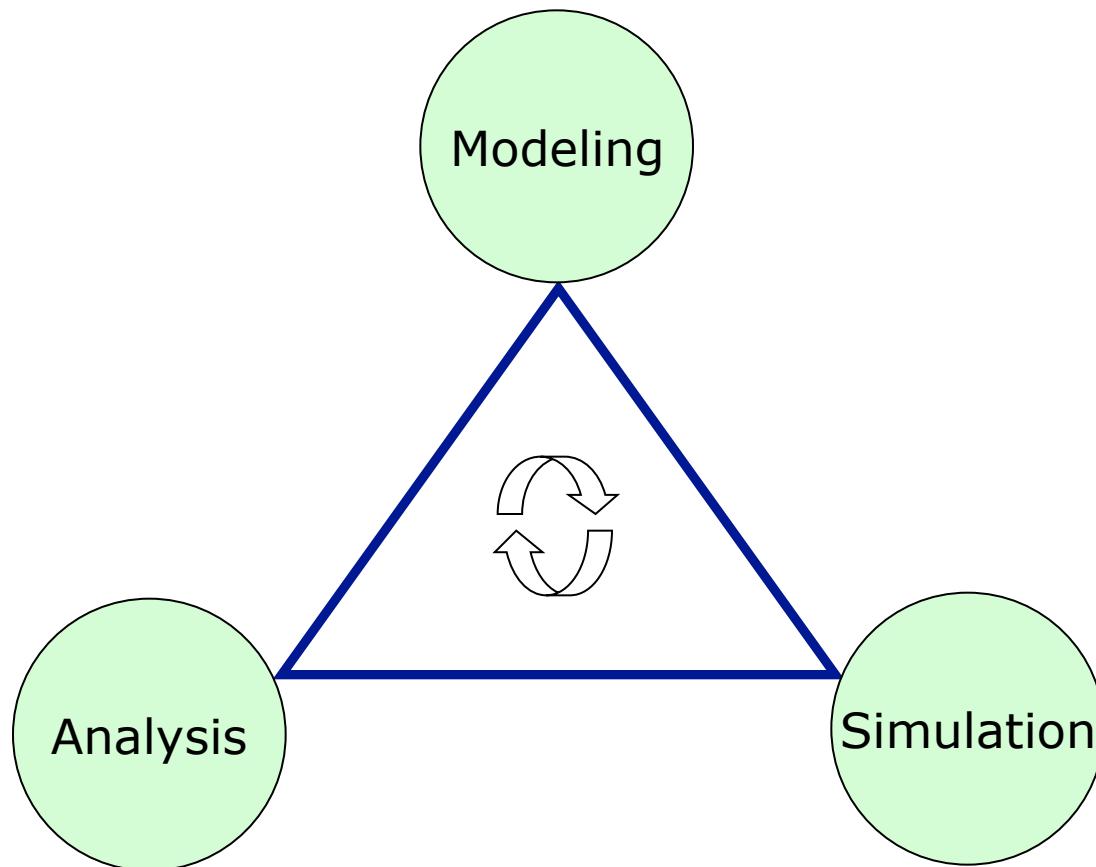
Parallel processing

Sequential Processing

- 1 CPU
- Simple
- Big problems???



New Approach



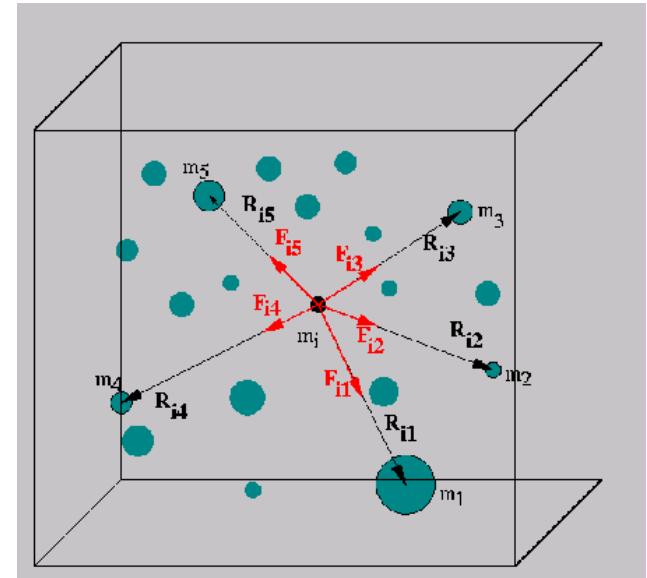


Grand Challenge Problems

- A grand challenge problem is one that cannot be solved in a reasonable amount of time with today's computers
- Ex:
 - Modeling large DNA structures
 - Global weather forecasting
 - Modeling motion of astronomical bodies

□ The N² algorithm:

- N bodies
- N-1 forces to calculate for each bodies
- N² calculations in total
- After the new positions of the bodies are determined, the calculations must be repeated





- **10^7** stars and so **10^{14}** calculations have to be repeated
- Each calculation could be done in $1\mu\text{s}$ (10^{-6}s)
- It would take $\sim 3 \text{ years}$ for one iteration ($\sim 26800 \text{ hours}$)
- But it only takes **10 hours** for one iteration with **2680** processors



Solutions

- Power processor
 - 50 Hz -> 100 Hz -> 1 GHz -> 4 Ghz -> ... -> Upper bound?
- Smart worker
 - Better algorithms
- Parallel processing



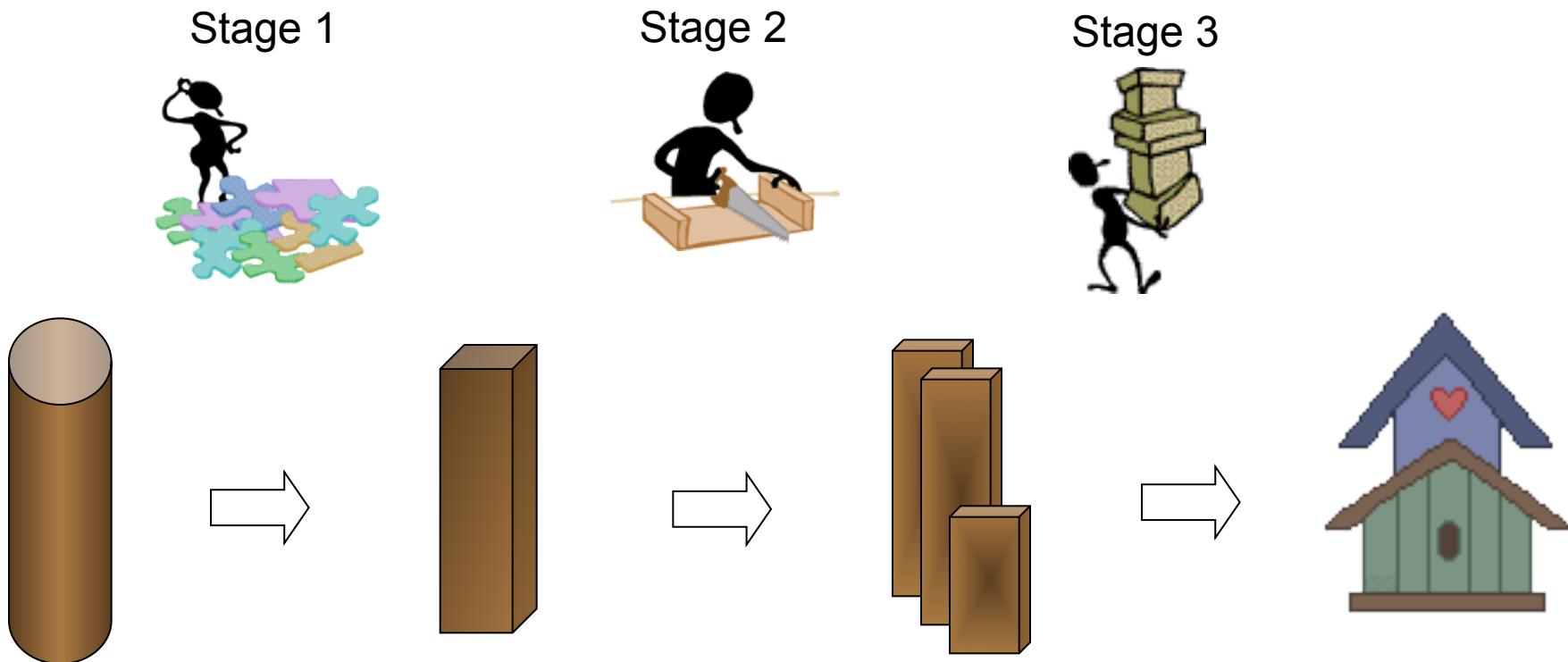
Parallel Processing Terminology

- Parallel processing
- Parallel computer
 - Multi-processor computer capable of parallel processing
- Throughput:
 - The throughput of a device is the number of results it produces per unit time.
- Speedup

$S = \text{Time}(\text{the most efficient sequential algorithm}) / \text{Time}(\text{parallel algorithm})$
- Parallelism:
 - Pipeline
 - Data parallelism
 - Control parallelism

Pipeline

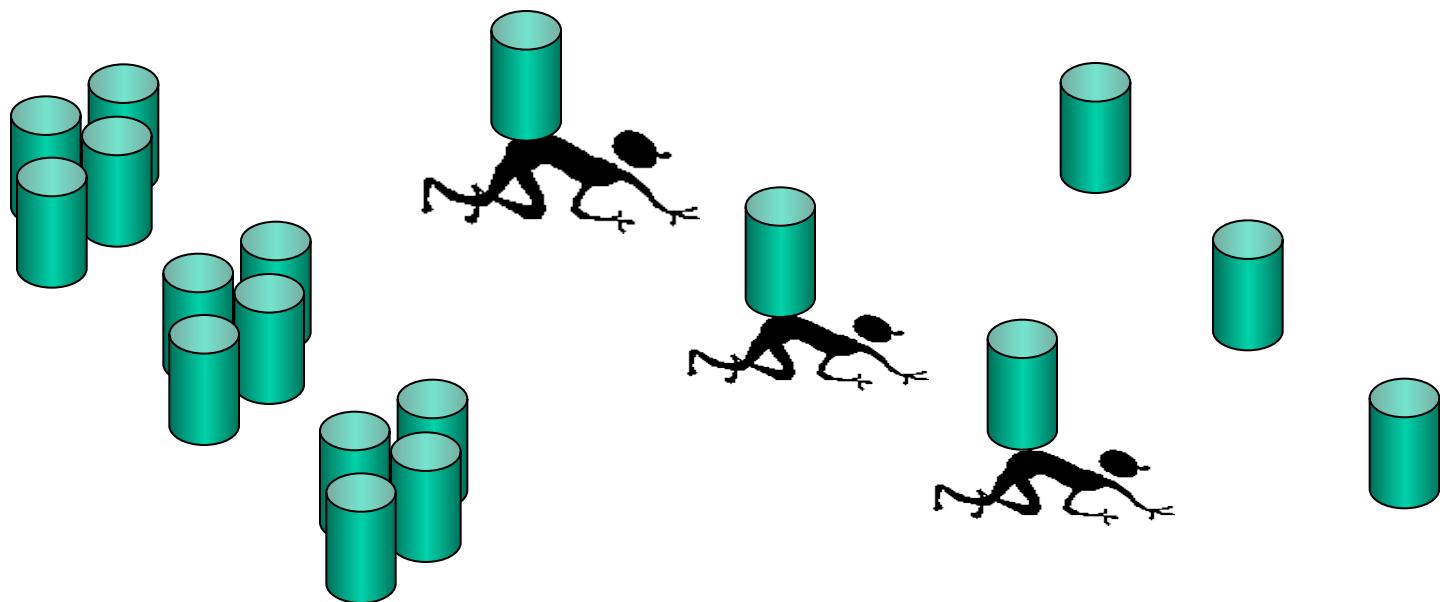
- A number of steps called **segments** or **stages**
- The output of one segment is the input of other segment



Data Parallelism

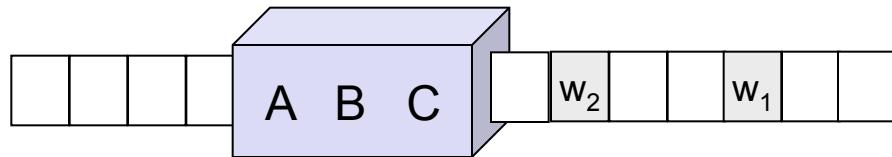
- Distributing the data across different parallel computing nodes

Applying the same operation simultaneously to elements of a data set

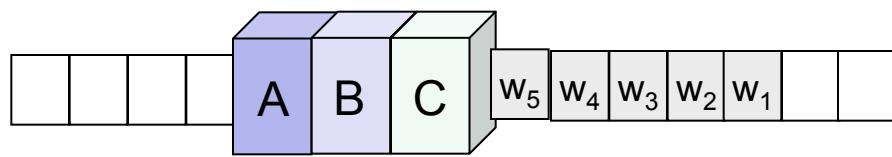


Pipeline & Data Parallelism

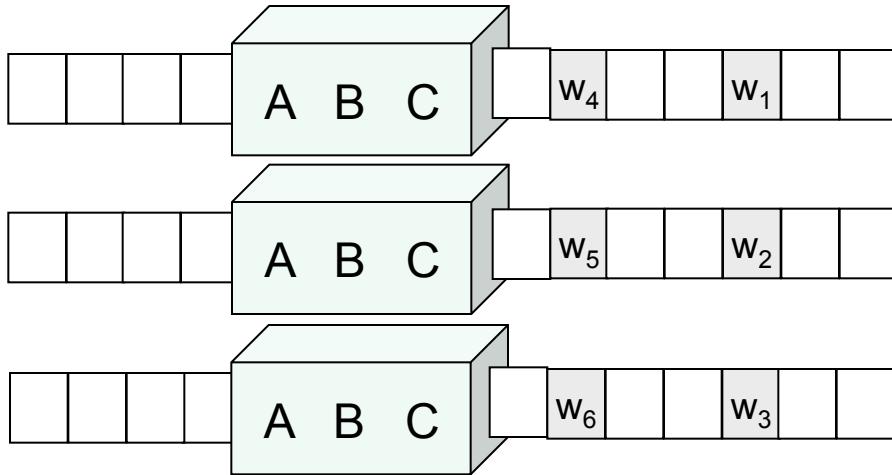
1. Sequential execution



2. Pipeline



3. Data Parallelism





Pipeline & Data Parallelism

□ Pipeline is a special case of control parallelism

□ $T(s)$: Sequential execution time

$T(p)$: Pipeline execution time (with 3 stages)

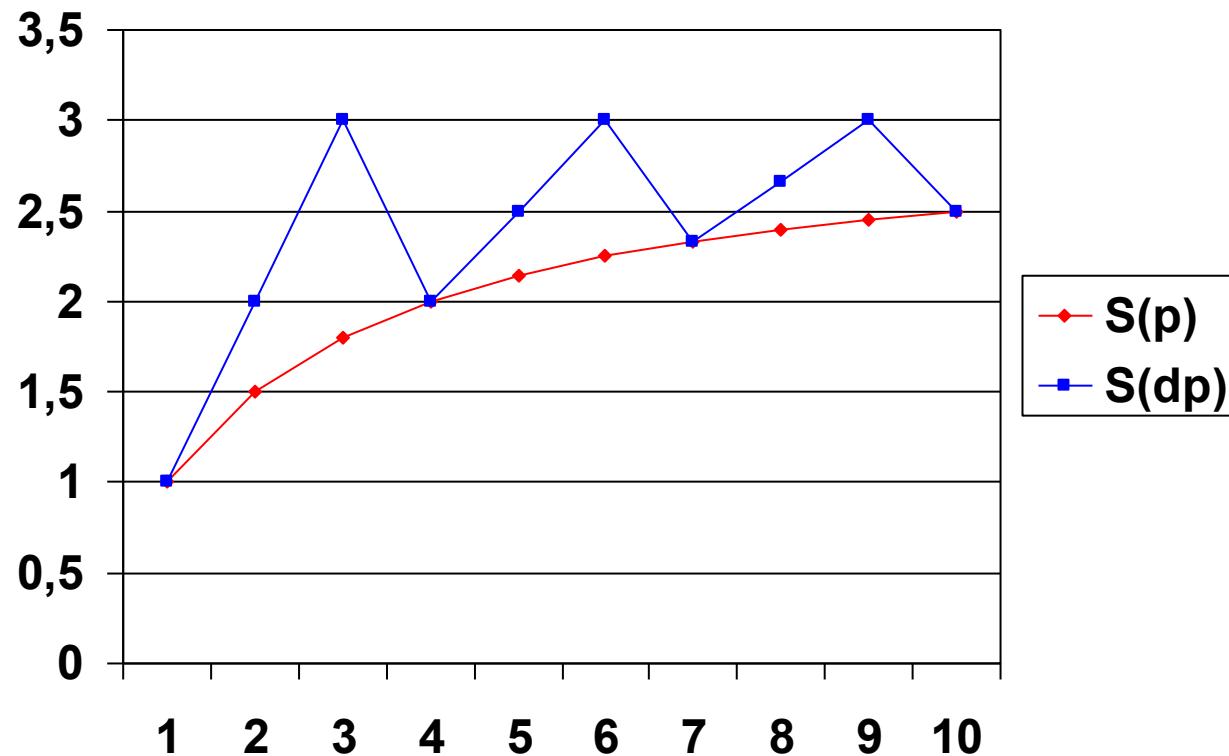
$T(dp)$: Data-parallelism execution time (with 3 processors)

$S(p)$: Speedup of pipeline

$S(dp)$: Speedup of data parallelism

Widget	1	2	3	4	5	6	7	8	9	10
$T(s)$	3	6	9	12	15	18	21	24	27	30
$T(p)$	3	4	5	6	7	8	9	10	11	12
$T(dp)$	3	3	3	6	6	6	9	9	9	12
$S(p)$	1	$1+1/2$	$1+4/5$	2	$2+1/7$	$2+1/4$	$2+1/3$	$2+2/5$	$2+5/11$	$2+1/2$
$S(dp)$	1	2	3	2	$2+1/2$	3	$2+1/3$	$2+2/3$	3	$2+1/2$

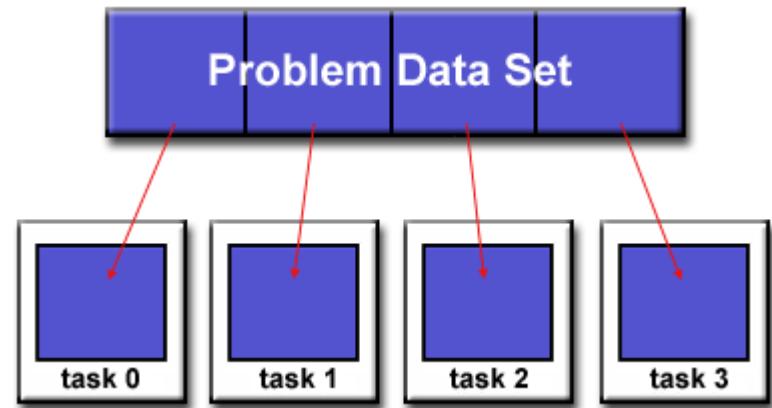
Pipeline & Data Parallelism



Control Parallelism

- Task/Function parallelism
- Distributing execution processes (threads) across different parallel computing nodes

Applying different operations to different data elements simultaneously





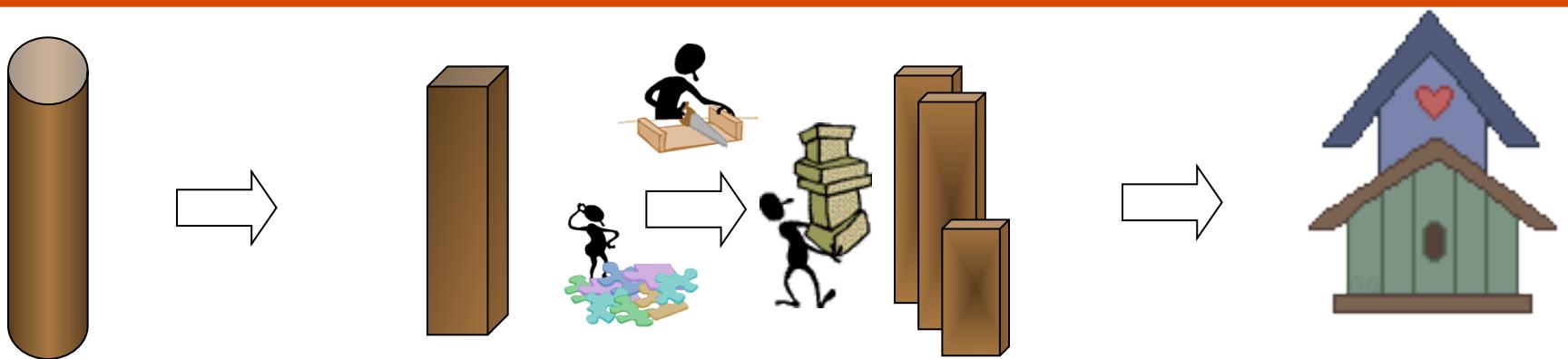
Example

{Milk, Sugar, Bread}

{Milk, Sugar, Bread, Tea}
{Bread, Milk, Coffee, Meat}
{Milk, Sugar}
{Milk, Bread, Sugar, Salt}
{Apple, Orange, Banana, Sugar, Milk}
...
{Milk, Bread, Sugar, Beer}

- Pipeline?
- Control parallelism?
- Data parallelism?

Throughput: Woodhouse problem



- ❑ 5 persons complete 1 woodhouse in 3 days
- ❑ 10 persons complete 1 woodhouse in 2 days
- ❑ How to build 2 houses with 10 persons?
 - (1) 10 persons building the 1st woodhouse and then the 2nd one later (sequentially)
 - (2) 10 persons building 2 woodhouses concurrently; it means that each group of 5 persons complete a woodhouse



Throughput

- The **throughput** of a device is the number of results it produces per unit time

- **High Performance Computing (HPC)**
 - Needing large amounts of computing power for short periods of time in order to completing the task as soon as possible

- **High Throughput Computing (HTC)**
 - How many jobs can be completed over a long period of time instead of how fast an individual job can complete



Scalability

- An algorithm is scalable if the level of parallelism increases at least linearly with the problem size.
- An architecture is scalable if it continues to yield the same performance per processor, albeit used in large problem size, as the number of processors increases.

- Data-parallelism algorithms are more scalable than control-parallelism algorithms