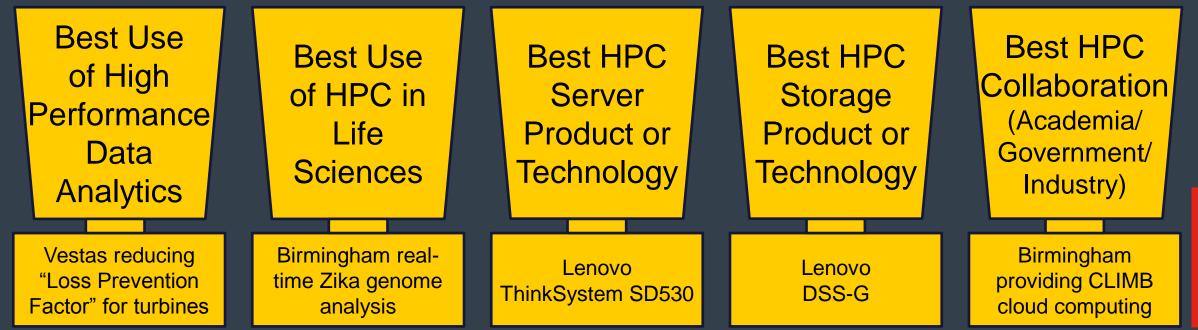
Tecnologie emergenti per nuovi ambili a plicativi: HPC/A.I./D.L.

Lenovo

Marco Briscolini, HPC&Al Sales Leader Data Center Group, Lenovo Global Technology Italy E-mail: mbriscolini@Lenovo.com

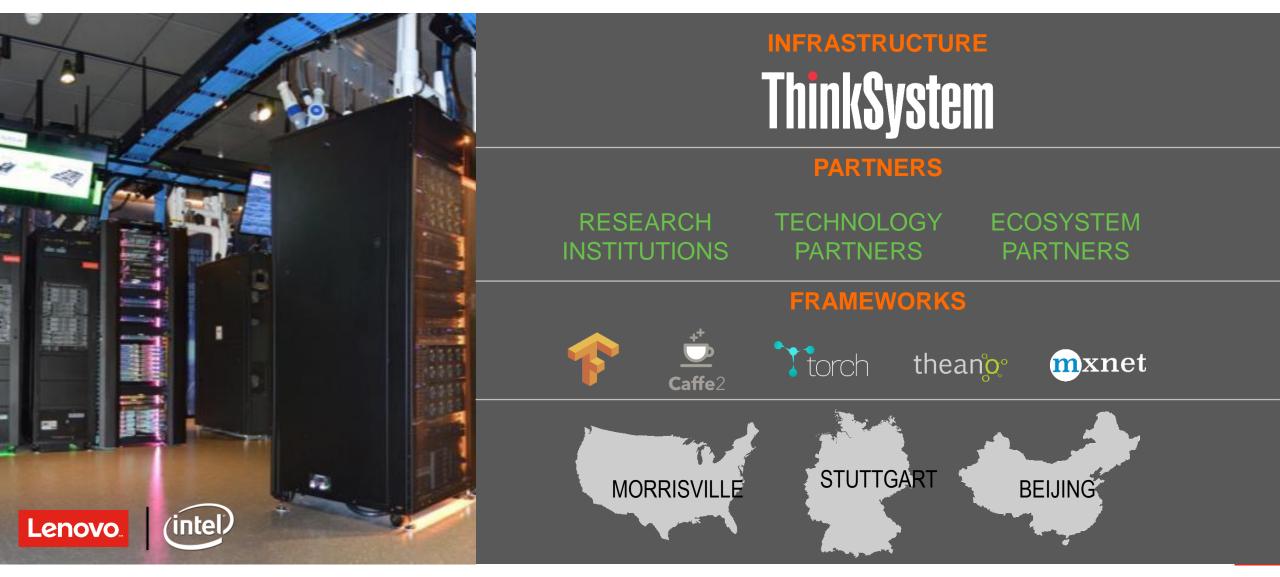
Lenovo... Best in Class





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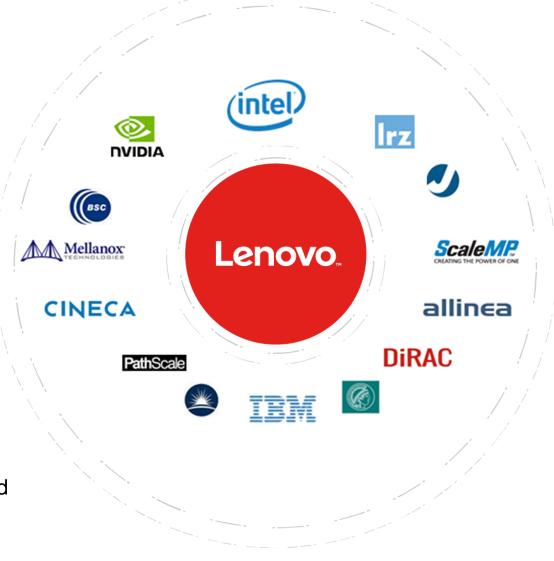
DISCOVER AI benefits with the Lenovo AI Innovation Centers



Innovation through Collaboration

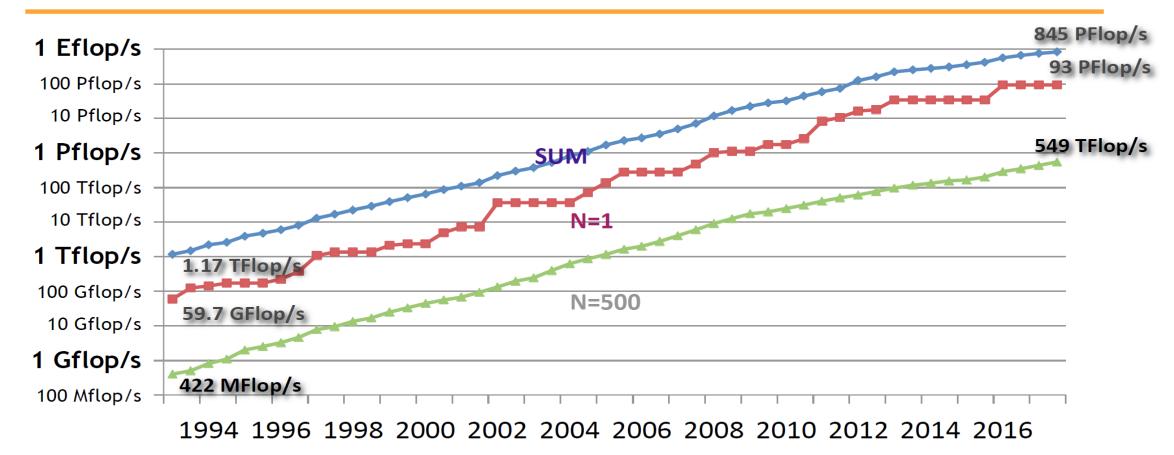
Delivering the power to help solve humanity's greatest challenges

- Collaboration with industry partners and clients
- Barcelona Super Computing Center
 - Energy aware MPI runtime
 - HPC Virtual environment investigation
- CINECA
 - Open System Management
 - GPFS Filesystem optimization on Intel OPA
- LRZ
 - High dense DWC architecture
 - Energy Aware Runtime
- HPC Benchmarking and Innovation Centers
 - Industry leaders to bring the newest technology and skills together for greater outcomes
 - Collaboration with visionary client partners to bring focused knowledge and deep skills in specific areas of science



HPC technology trends

PERFORMANCE DEVELOPMENT

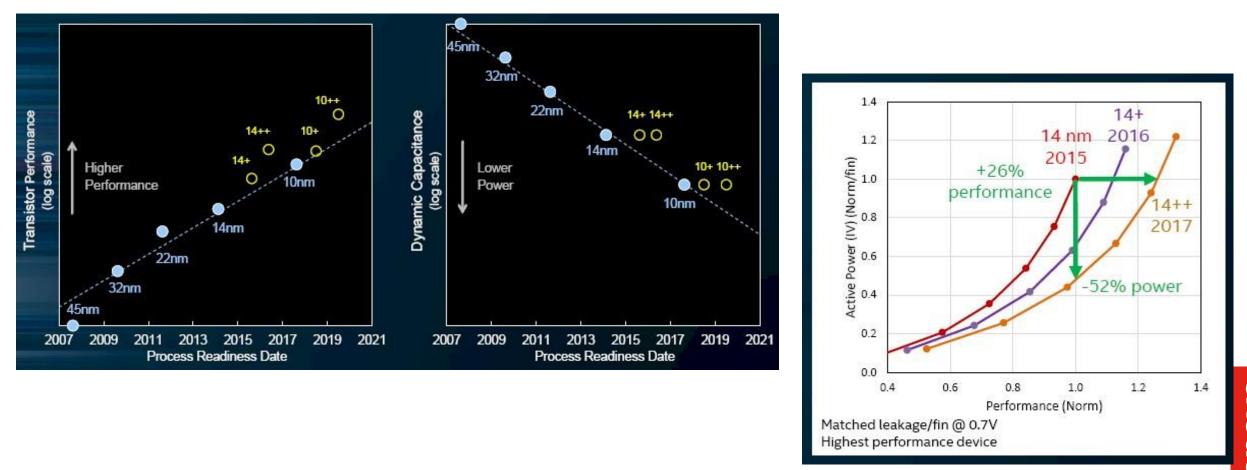


Lenovo

500

HPC technology trends

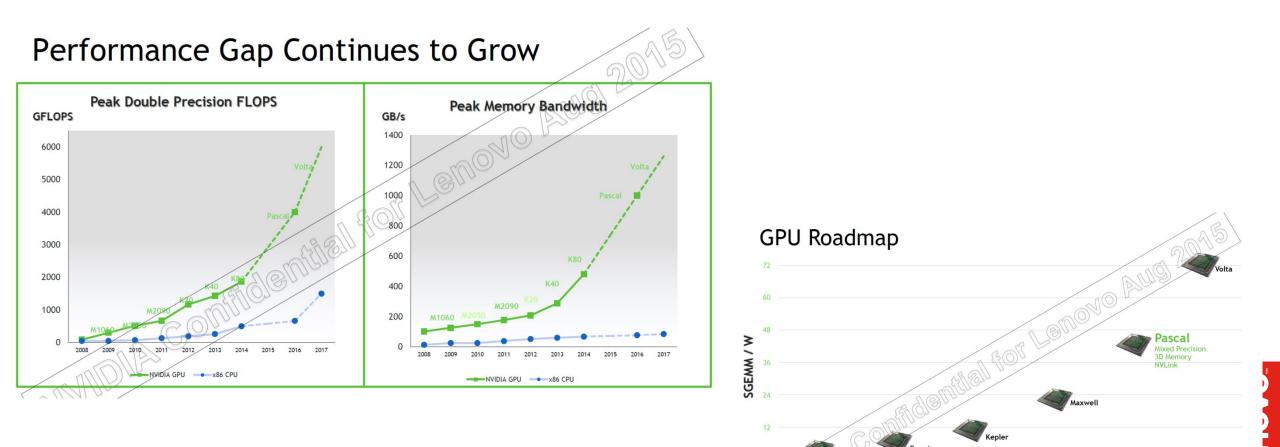
An example of technology evolution: the Intel x86 micropocessor



https://www.nextplatform.com/2017/04/27/mapping-intels-tick-tock-clock-onto-xeon-processors/

HPC technology trends

Hybrid computing: CPU+GPU makes the difference as performance and consumption



2018

2014

2016

Lenovo HPC in any shape or form



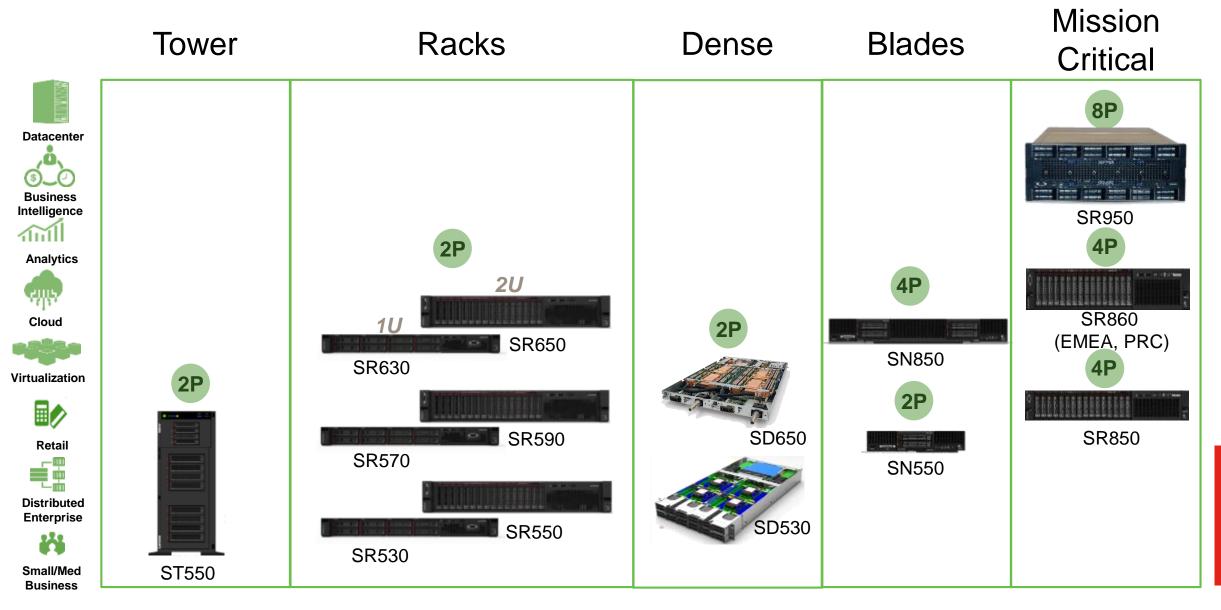
Supercomputing Divisional Departmental Workgroup

| Hardware | Software |
|----------|-----------|
| Service | Solutions |

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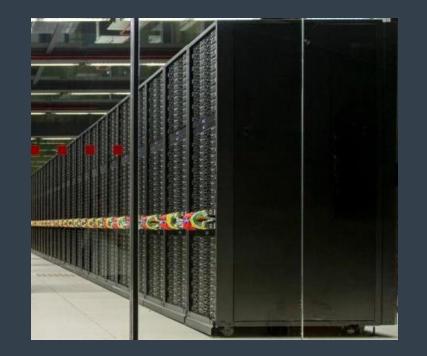
The Lenovo ThinkSystem Server Portfolio



Lenovo ThinkSystem HPC Dense



Our biggest product launch ever Celebrating 25 years in x86 servers Top BIN processors and networks New GPU capability shipping soon Highest performance/lowest cost



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ThinkSystem SD530

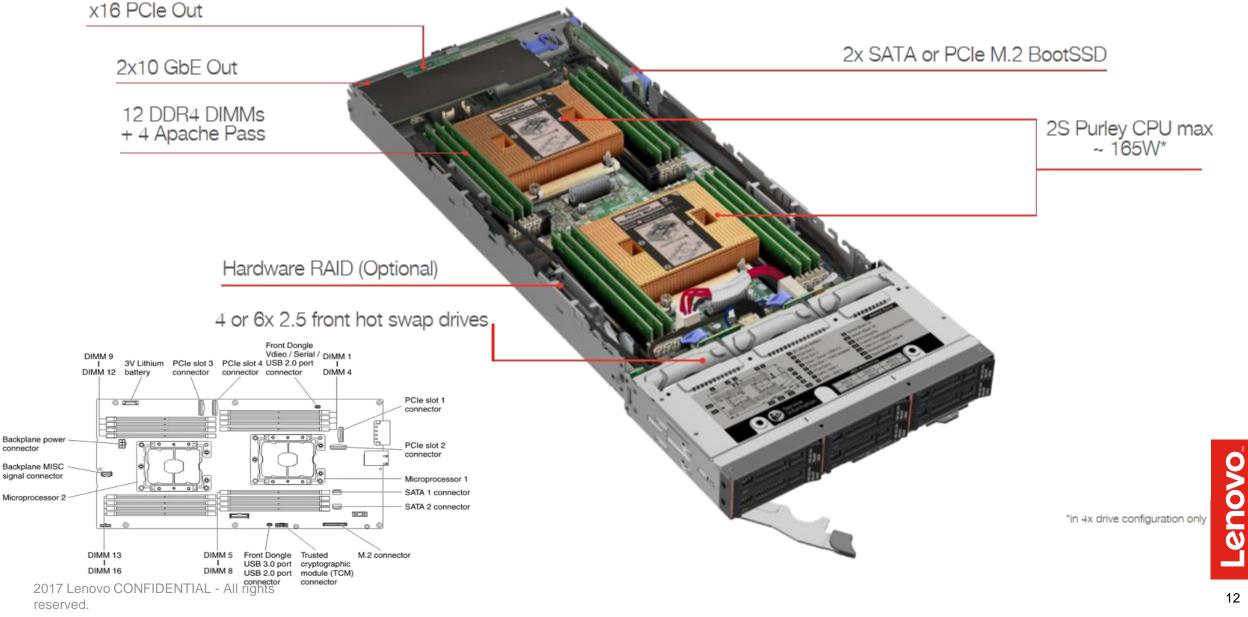
| Feature | SD530 |
|----------------------------|--|
| Processors | 2 Intel "Purley" Generation processors per nodeUp to 165W and 28 Cores |
| Form factor | 1U Half wide tray node / 2U4N Chassis |
| Memory Slots Max Memory | 12x DDR4 (R/LR) 2667MHz DIMM (up to 64GB) 4x ApachePass future ready |
| Storage | 6x SAS/SATA SF HS HDD/SSD or 2x NVMe, 2x M.2 SATA SSD |
| NIC | 2x 10G SFP+ or 10GBase-T |
| PCle | 2x x8 or 1x x16 PCIe (HS) in Rear Shuttle 2x x16 internal for expansions |
| Power | HS/1+1 redundant 1100W/1600W/2000W Platinum |
| USB ports | 1x USB 2.0 |
| Cooling | 5x hot swap redundant fans Support ASHRAE A2/3/4 (config limitation A3/4) |
| System MGMT / TPM | XCC Stark Mgmt Module, 1 / 2 shared GbE ports TPM, Pluggable TCM, Opt. KVM break-out |
| Expansions | GPU Expansions for 2 GPUs per node |
| Dimensions | 846mm depth, front access w/ rear I/O |







The Node



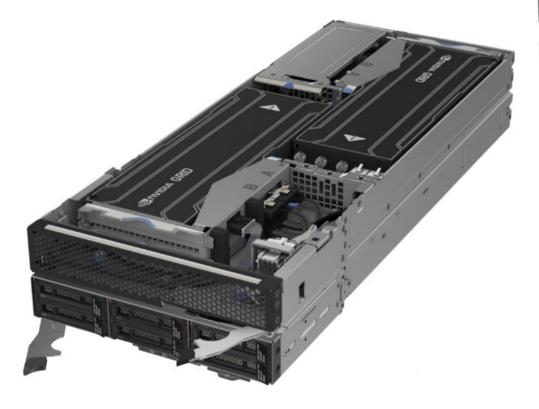
GPU Tray

Node limitation

- Only support 2x2 Backplane (4 HDD config)
- Must be identical GPU cards in the same GPU tray

Chassis limitation

- Must have 2 x GPU tray in a chassis for air flow balance
- Only support 2000W PSU with GPU tray





Supported Accelerators

- NVIDIA Tesla M10 32GB (250 W)
- NVIDIA Tesla M60 16GB (300W)
- NVIDIA Tesla P40 24GB (250 W)
- NVIDIA Tesla P100 16GB (250 W)
- NVIDIA Tesla V100 16GB (300 W)

... more planned for the future



Lenovo **Direct "Hot" Watercooling**



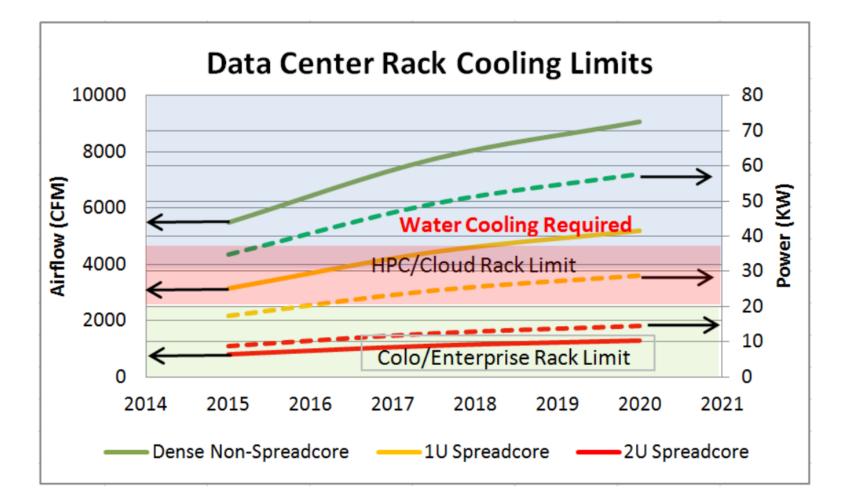
Up to 50°C Hot Water Cooling Up to 90% Heat Removal Efficiency World Record Energy Reuse Efficiency Many patents on market leading design





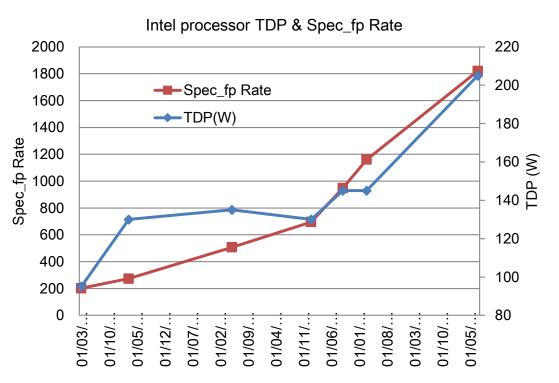
Data Center Level Cooling and Power Limits

- Node power density trends cannot be cooled at data center level
 - Partial rack population or rack level power capping may be required for Dense and 1U



Intel Xeon Server processor history

| Release date | Code | Processor | core/chip | TDP(W) | Spec FP | Spec_fp Rate |
|--------------|-------------|--------------------------|-----------|--------|---------|--------------|
| 2006/6/26 | Woodcrest | Intel Xeon 5160 | 2 | 80 | 17.7 | 45.5 |
| 2007/11/12 | Harpertown | Intel Xeon x5460 | 4 | 120 | 25.4 | 79.6 |
| 2009/3/30 | Nehalem | Intel Xeon x5570 | 4 | 95 | 43.8 | 202 |
| 2010/3/16 | Westmere-EP | Intel Xeon x5690 | 6 | 130 | 63.7 | 273 |
| 2012/5/1 | SandyBridge | Intel Xeon E5-2690 | 8 | 135 | 94.8 | 507 |
| 2014/1/9 | IvyBridge | Intel Xeon E5-2697v2 | 12 | 130 | 104 | 696 |
| 2014/9/9 | Haswell | Intel Xeon E5-2699v3 | 18 | 145 | 116 | 949 |
| 2015/3/9 | Bradwell | Intel Xeon E5-2699v4 | 22 | 145 | 128 | 1160 |
| 2017/7/11 | Skylake | Intel Xeon Platinum 8180 | 28 | 205 | 155 | 1820 |

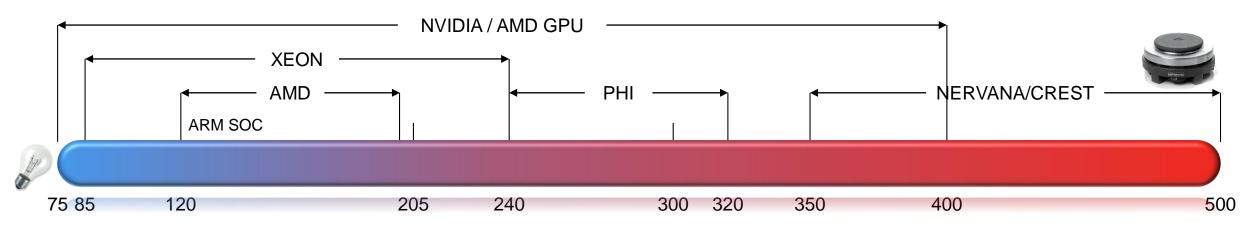


Processor performance trend

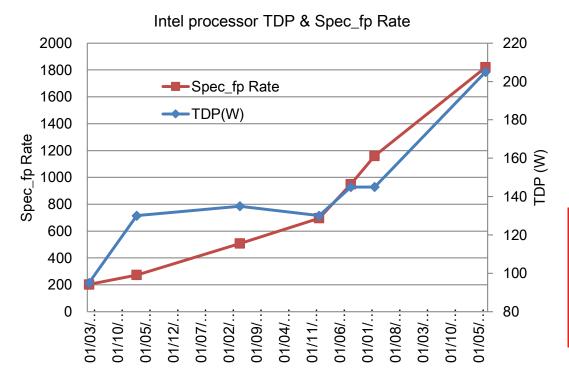
- Spec_fp rate with 2 processors/node has increased 40 times the past 11 years (2006 2017).
- The number of cores on the chip increase 14 times.
- After being flat, since 2014 TDP increases linearly with Spec_fp rate.
- Current maximum TDP is 205W. Knighs Mill Xeon phi processor will be 305 W

To sustain increased performance servers will have to be less dense or use new cooling technology

Industry Thermal Challenges



- Maintaining Moore's Law with increased competition is resulting in higher component power
- Increased memory count, NVMe adoption, and I/O requirements are driving packaging and feature tradeoffs (superset of features doesn't fit in 1U)
- Shared cooling fan power savings no longer exist for dense 2S nodes architectures due to nonspreadcore CPU layout high airflow requirements
- Industry moving away from dense nodes (e.g. Open 19 and OCS Olympus)



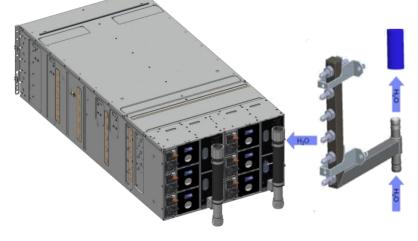
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Heat Extraction/Performance Expectations

| Compute Tray Configuration (dual compute nodes) | Single Rack Steady State Power (w/Linpack, Turbo ON/OFF) | 90% Heat Removal | | 85% Heat Removal | |
|---|---|------------------|--------------|------------------|--------------|
| Compute Tray Configuration (dual compute hodes) | | Heat-to-water | Heat-to-air* | Heat-to-water | Heat-to-air* |
| 2 server / tray with TDP=165W CPU, 16x16GB DDR4 Memory, 2x NIC | 33.5kW | 29.3kW | 4.2kW | 27.6kW | 5.9kW |
| 2 server / tray with TDP=145W CPU, 16x 16GB DDR4 Memory, 2x NIC | 30.7kW | 26.7kW | 4.0kW | 25.3kW | 5.4kW |
| 2 server / tray with TDP=135W CPU, 16x 16GB DDR4 Memory, 2x NIC | 29.4kW | 25.6kW | 3.8kW | 24.1kW | 5.3kW |
| 2 server / tray with TDP=120W CPU, 16x 16GB DDR4 Memory, 2x NIC | 27.3kW | 23.7kW | 3.6kW | 22.4kW | 4.9kW |

Assumptions: 36x compute trays (dual node), 72 servers + 6x switches / Rack

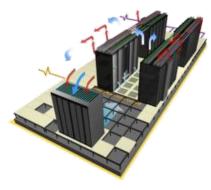
 Performance is dependent on many facility variables such as ambient temp, water temp, flow rate, etc... Based on testing, it is reasonable to assume 85-90% heat-to-water given the availability of typical environmental input parameters.





Choice of Cooling

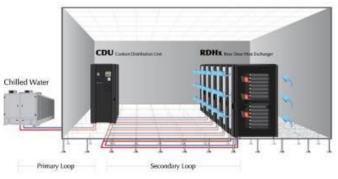
Air Cooled



- Standard air flow with internal fans
- Fits in any datacenter
- Maximum flexibility
- Broadest choice of configurable options supported
- Supports Native Expansion nodes (Storage NeX, PCI NeX)

PUE ~1.5 ERE ~ 1.5

Air Cooled with Rear Door Heat Exchangers



- Air cool, supplemented with RDHX door on rack
- Uses chilled water with economizer (18C water)
- Enables extremely tight rack placement

PUE ~1.2

ERE ~ 1.2

Direct Water Cooled



- Direct water cooling with no internal fans
- Higher performance per watt
- Free cooling (45C water)
- Energy re-use
- Densest footprint
- Ideal for geos with high electricity costs and new data centers
- Supports highest wattage processors

PUE <= 1.1

ERE ~ 0.3 with hot water

Choose for broadest choice of customizable options

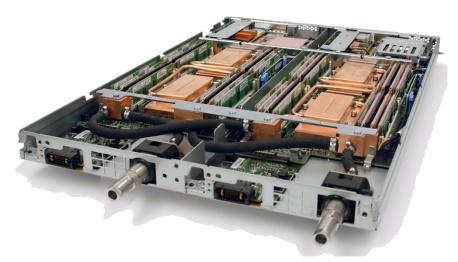
Choose for balance between configuration flexibility and energy efficiency

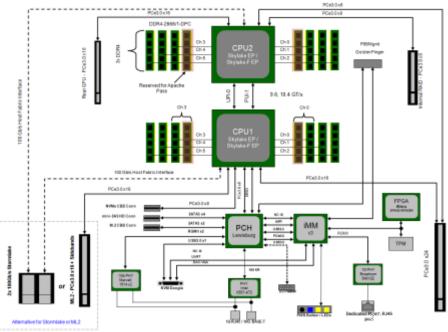
Choose for highest performance and energy efficiency

ThinkSystem SD650 (OceanCat)

GA Q1/2018

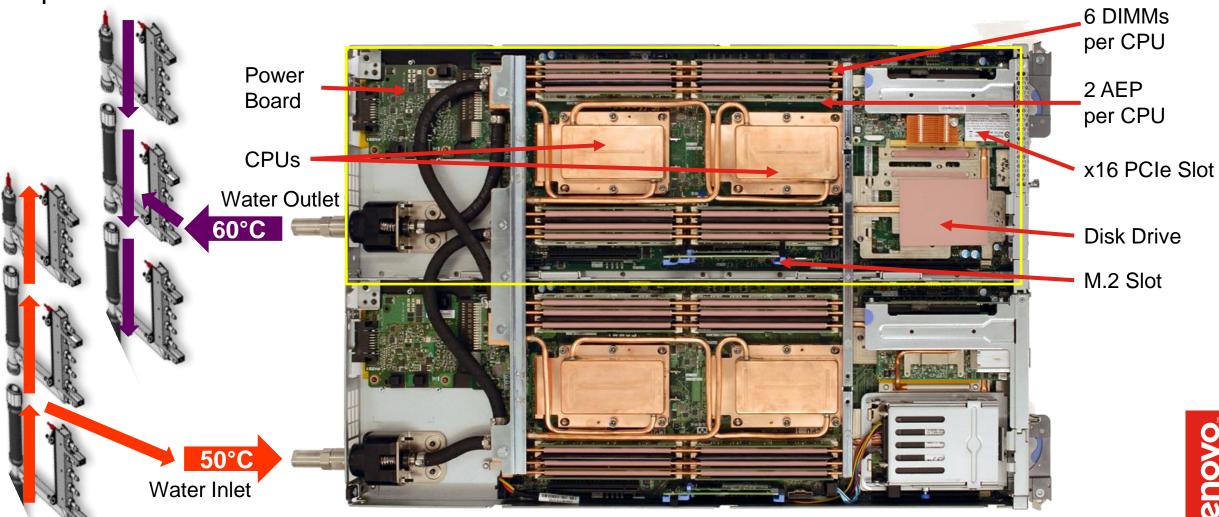
| Feature | SD650 |
|----------------------------|--|
| Processors | 2 Intel "Purley" Generation processors per node Socket-F for Intel Omnipath supported >120W all Skylake Shelves supported |
| Form factor | 1U Full wide tray double-node / 6U12N Chassis |
| Memory Slots Max Memory | 12x DDR4 (R/LR) 2667MHz DIMM 4x Intel Apache Pass DIMM |
| Storage | 2x SATA slim SSD, 1x NVMe, 2x M.2 SATA SSD |
| NIC | 1x 1 GBaseT, 1x 1 GbE XCC dedicated |
| PCle | 1x x16 PCIe EDR Infiniband, HDR with internal x16 1x x16 ML2 for 10Gbit Ethernet Internal: 1x x8 for RAID, 1x x16 for pot. expansions |
| Power | 1300W/1500W/2000W Platinum and 1300W Titanium |
| USB ports | Up to 1x front via dongle cable + 1x internal (2.0) |
| Cooling | No fans on chassis, PSU fans only Up to 50°C warm water circulated through cooling tubes for component level cooling |
| System MGMT / TPM | XCC, dedicated port or shared TPM, Pluggable TCM |
| Dimensions | 915mm depth, front access w/ front I/O |





ThinkSystem SD650 (OceanCat)

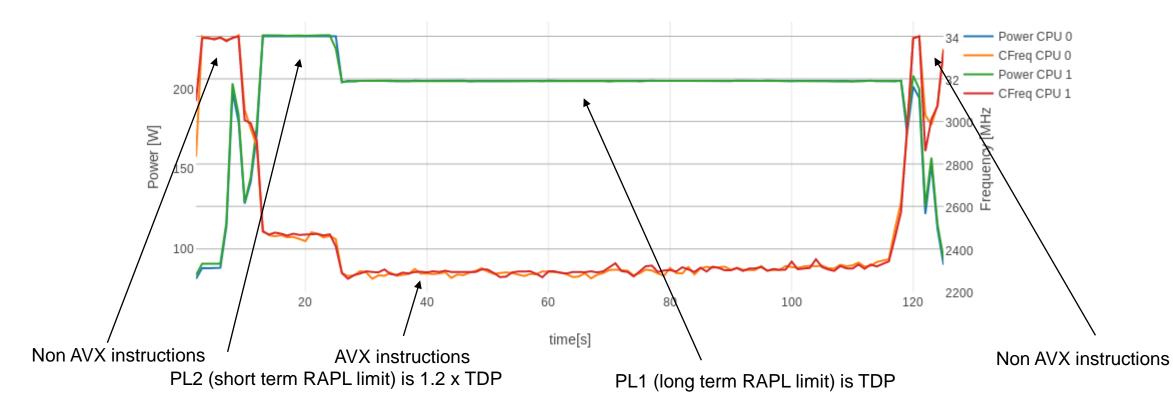
Top-Down View



two nodes sharing a tray and a waterloop

HPL DC Power & frequency on SD650 with 8168

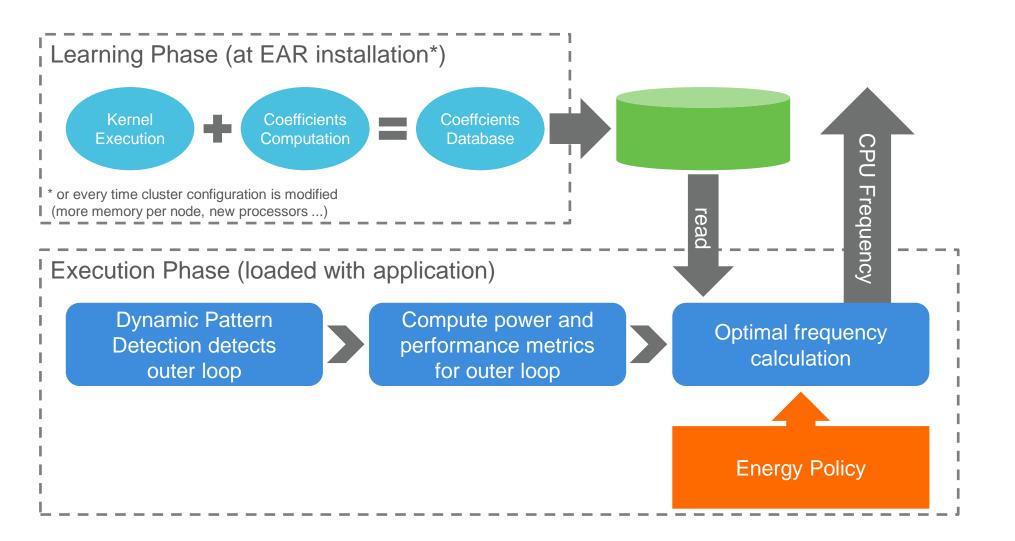
Compare Power and Frequency



SD650 with 2 sockets 8168 and 6 x 16GB DIMMs; room temp = 21°C, inlet water = 40°C, 1.5 lpm/tray

Lenovo Energy Aware Runtime (LEAR)

Functional Overview



iarcelona upercomputing tenter entro Nacional de Supercomputació



Lenovo... ThinkSystem HPC & AI



Common AI/ML Terminology

| Inference | Using data training instead of programming to generate output |
|-----------------------|---|
| Intelligence Agent(s) | A device that interacts with its environment to achieve a goal |
| Ontology | A set of objects, concepts and data, about which a device is aware |
| Machine Learning | Utilizing algorithms to train data that improves results over time via experience |
| Perception | Using input from external devices (cameras, sensors) to make decisions |
| Deep Learning | Utilizing neural networks to train data in a way similar to the way human learning occurs |
| Cognitive computing | Systems that mimic human brain (self-learning systems) |

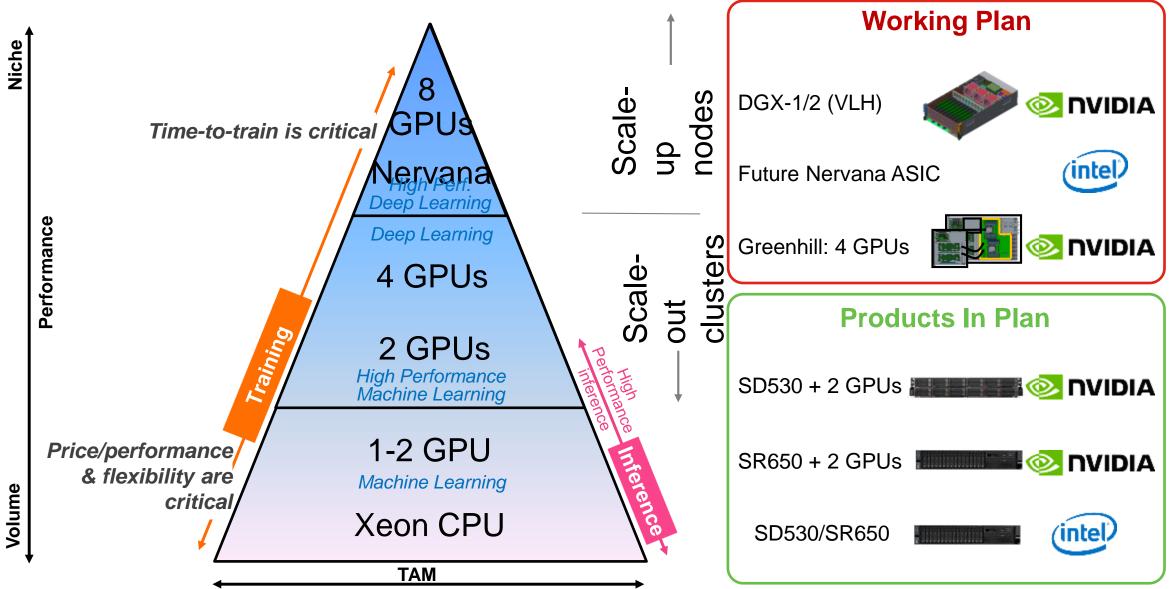
Key Terms for A.I. in your case

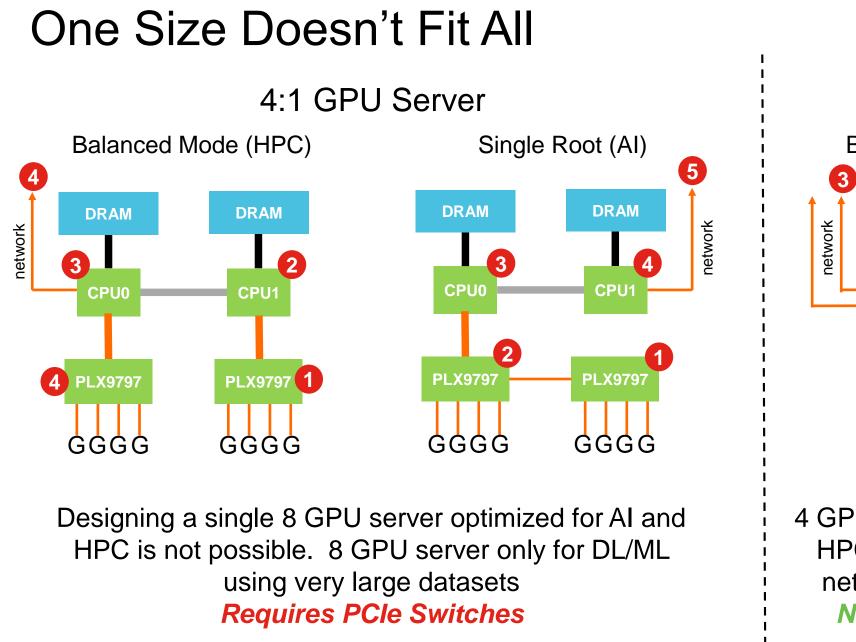
| Predictive Analytics | Image or voice recognition | Natural language processing | | |
|---|---------------------------------|--------------------------------|--|--|
| Autonomous machine interaction or Robotics | Unstructured data processing | ΙοΤ | | |
| Training systems | | | | |

Diagnostics

Fraining systems or scoring or inference systems

DCG AI and Deep Learning HW portfolio





=Latency Hops from farthest GPU

2:1 GPU Server Balanced and "Single Root" DRAM DRAM 2 CPU1 CPU0 GG GG

4 GPU systems are ideal for both HPC and AI. Minimal hops to network and from GPU-GPU *No PCIe Switch required*

Lenovo Artificial Intelligence



Lenovo Intelligent Computing Ochestration

Integrated SW suite for HPC & AI

- Strong OpenHPC base
- Highly customizable
- Easy to use new design



<u>enovo</u>

LiCO - Lenovo Intelligent Computing Orchestration



A single software stack to efficiently manage both HPC & AI workloads

For HPC

Simplify monitoring and job submission for HPC

- Easy-to-use interface for users to submit and manage jobs
- Full access to native tools in the stack for more technical users
- Built on an OpenHPC software base, with Lenovo value-add capabilities and optimizations

| Customer Applications | | | | |
|---|---------------------------|---------|------------|--|
| Portal powered by Web GUI | | | | |
| HPC workflow templates AI pre-trained models, templates, SDK/API | | | | |
| HPC application libraries AI frameworks | | | | |
| Scheduler Shared storage | | | | |
| Server | Storage | Network | DataCenter | |
| XEON PHI Inside | (intel) XEON inside | ADIEVE. | | |
| Lenovo Scalable Infrastructure | | | | |

For Al

Easy access to train and optimize AI models

- Execute jobs, monitor training progress through a single GUI
- Out of the box scaling for both Intel and NVIDIA environments
- Easily try different frameworks, system types to determine best fit

LiCO for High Performance Computing

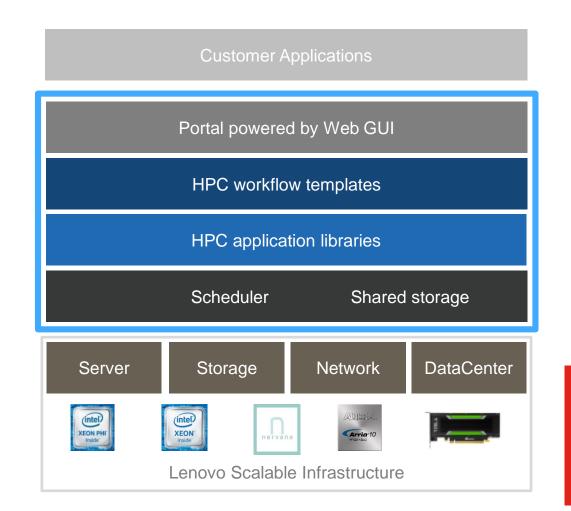
Validated stack of open tools to simplify cluster environments

• Single GUI

- consolidates functionality for both administrators and users
- Built on an OpenHPC foundation
 - Ganglia, Slurm/Munge, Torque/Maui, Singularity, MPI, xCAT, Confluent

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- Lenovo value-added capabilities
 - developed through client collaboration
 - Open web portal (Oxford & South Hampton)
 - Energy-Aware Runtime (BSC)

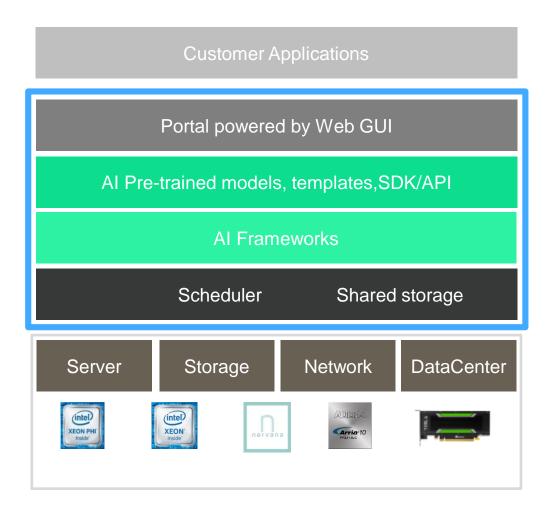




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LiCO for Artificial Intelligence

Enables easy experimentation to optimize training models





- Intuitive web portal
 - manage workflow and application development for Data Scientists
- Pre-validated stack of open AI tools and libraries, ready to deploy
 - Caffe, TensorFlow, CUDA, NVIDIA cuDNN, Intel MKL-DNN, etc.
- Optimized for scalability, regardless of underlying hardware processors

AI Use Cases that Lenovo is exploring today

A more efficient and effective

Radiology Practice

CodaLab

Lenovo.

For better quality control on

Manufacturing Lines

Mark III

Lenovo

For proactive actions improving

Water & Energy Conservation



Lenovo

To advance particle physics at

CERN's Large Hadron Collider



AI: Better quality control for Manufacturing

Breakthrough innovation in Manufacturing with Lenovo Al

Current State of Manufacturing:

- Better Quality Control directly related to:
 - More yield at higher speed
 - Lower production costs by faster adjustments to process

Where AI can help:

- Leverage cameras and sensors through-out the production lifecycle
- Better manage quality through product age of customizations

Al Inference and Training:

- Image / Pattern recognition and analysis
- ML training of products at different stages of production

See demo in action at SC17, Booth 1353



AI: To advance particle physics in the Large Hadron Collider

Lenovo Al progressing advanced Particle Physics

Current State and Challenges :

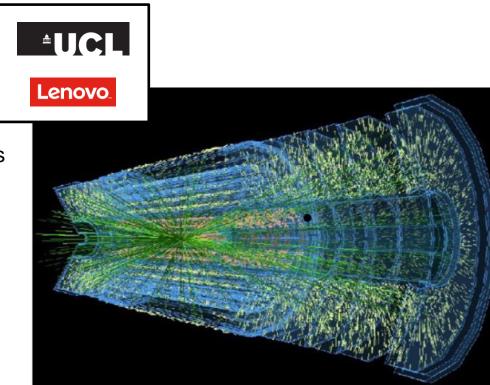
- Reconstructing particle trajectories from thousands of sensor measurements in the detector is an important data analytics task
- Traditional computational methods demand high amounts of resources and don't scale well as the LHC pushes to higher collision frequencies

Where AI can help:

• <u>Pattern recognition</u>: Reconstructing particle trajectories using imaging data from the collider much more efficiently than traditional methods

Al Inference and Training:

 Machine learning methods using binary image data from experiments combined with integral transforms for pattern recognition





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Different is better