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Adaptive Data-Center

for AI Machine Learning Acceleration

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Outline

Introduction

- ✓ Data center (DC) applications
- ✓ Market and development trends

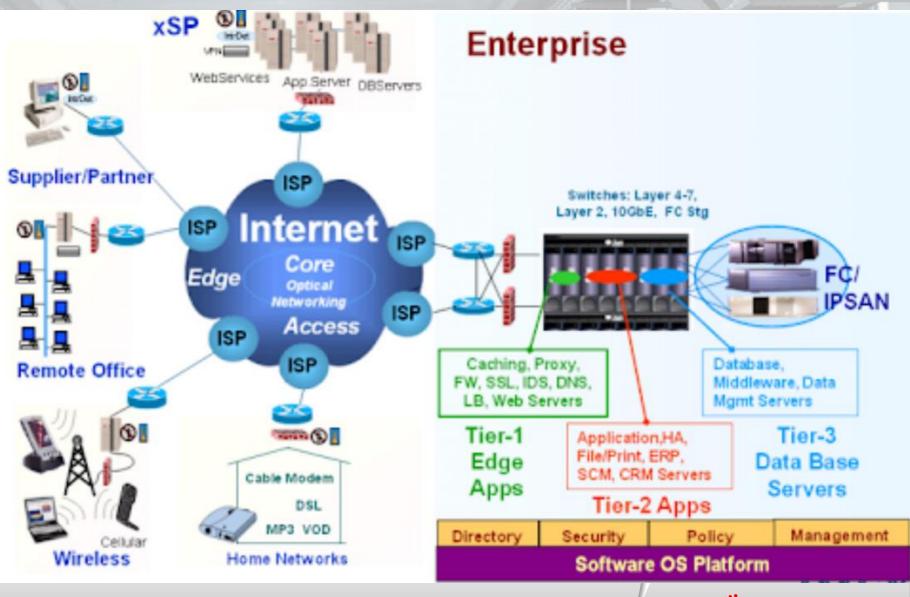
Data Center

- ✓ Basic DC configuration and challenges
- ✓ Configurable DC: Software-Defined Data Center (SDDC)
- ✓ Adaptability using ML
- ✓ Hardware/software codesign
- \checkmark Extended EDA for optimization
- OCP deployment
- Summary



Introduction – DC Applications

- Data center provides
 - ✓ Computation
 - ✓ Storage
 - ✓ Network
 - ✓ Security
- Clouds services
 - ✓ SaaS
 - ✓ PaaS
 - ✓ Edge /Orchestra /AI
- Improve infrastructure usage and IT OPEX
 - ✓ Expected 50% improvement

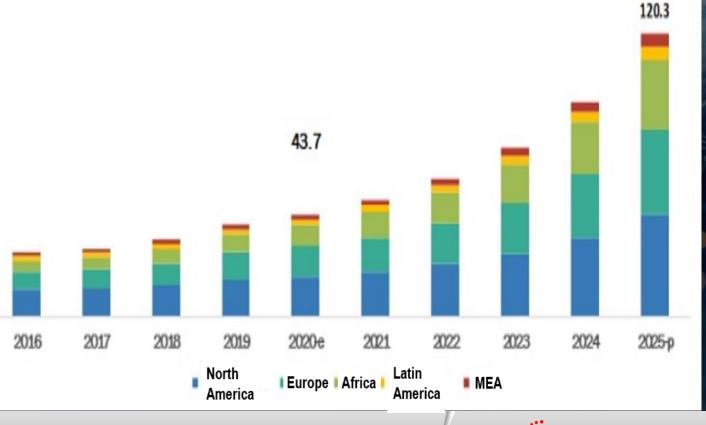




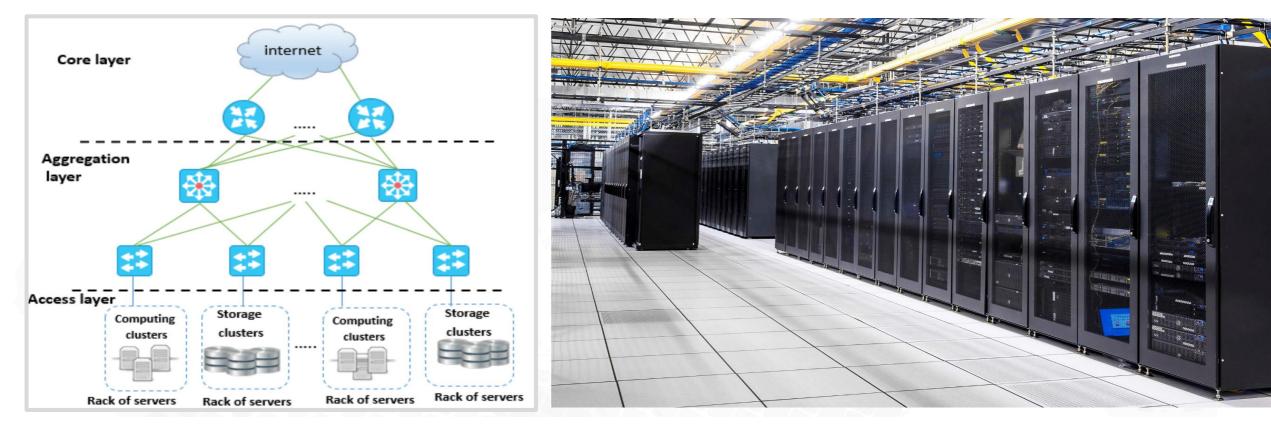
Introduction – Market

- Software-Defined Data Center (SDDC) market size is projected to grow
 - ✓ \$**43.7** billion in 2020
 - ✓ \$**120.3** billion by 2025
- At a Compound Annual Growth Rate (CAGR) of 22.4%
- 30 % DC failing to prepare for AI will no longer be operationally or economically viable by 2020





Introduction – DC Architectures



- Core layer: high-speed packet switching backplane going in and out of the data center
- Aggregation layer
 - ✓ Service module integration
 - ✓ Layer 2 domain definitions, spanning tree processing
 - ✓ Server-to-server multi-tier traffic flows through the aggregation layer
- Access layer: servers

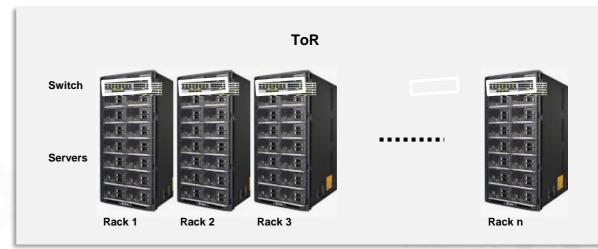
Introduction – DC Architecture Examples

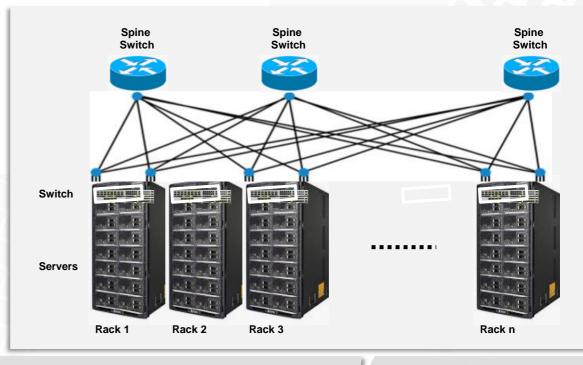
Switch-to-server connections

 Top of rack (ToR): one switch for each rack. Servers within the rack are connected to the switch via copper cable. All switches for the racks are connected to ToR switches, spine switches.

Features

- ✓ Copper stays "In Rack", lower cabling costs
- Modular, flexible "per rack" architecture, and higher speeds
- High capital and maintenance costs. The distributed architecture of ToR requires more physical switches
- ✓ Under utilized switches cost unnecessary power





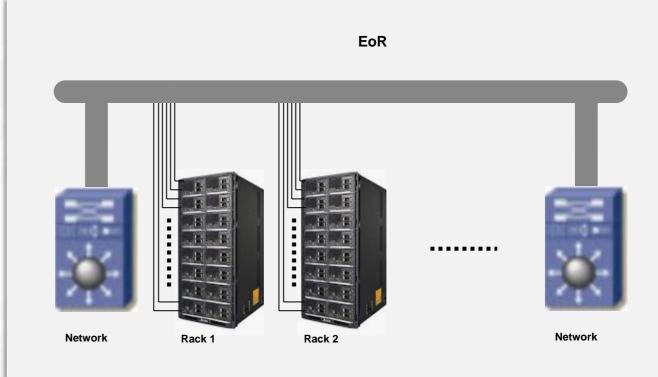


Introduction – DC Architecture Examples

 End-of-Row / Middle of Row (EoR / MoR): a dedicated networking rack at either end of a row of servers for the purpose of providing network connectivity to the servers. Within that row, both end can have a networking rack for reliability redundancy purpose

Features

- Allows collapsing of the access and aggregation layers into a single tier of highdensity chassis switches. It reduces the number of switches
- Provides improved performance by reducing the level compared to multi-tier approach
- Each server rack would have a twisted pair copper cabling routed through overhead cable trays to the switch rack

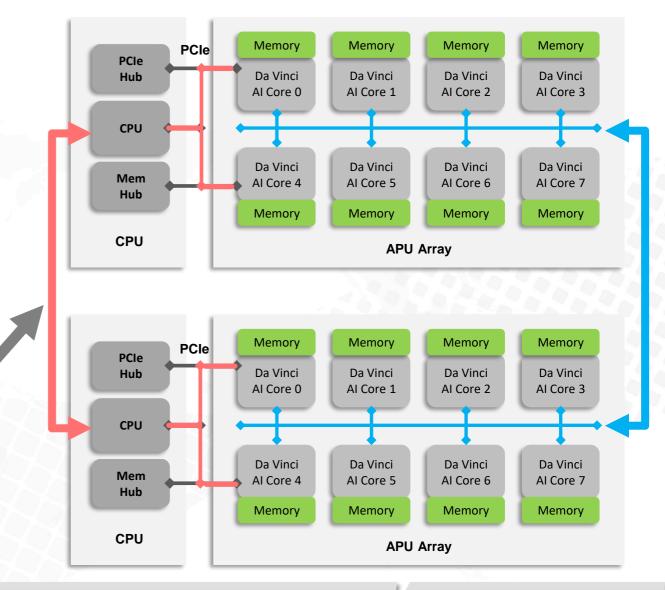




Introduction – AI Server/Core Configuration Example

Within the server, the main components are APU/GPU, memory and CPU. There are many possible configurations

- Communicate via CPU
 - ✓ Easy to synchronize
 - ✓ Low efficiency
- Communicate bypass CPU
 - High communication efficiency, especially for pipelined computations
 - \checkmark Need synchronized computation output and input





Challenges

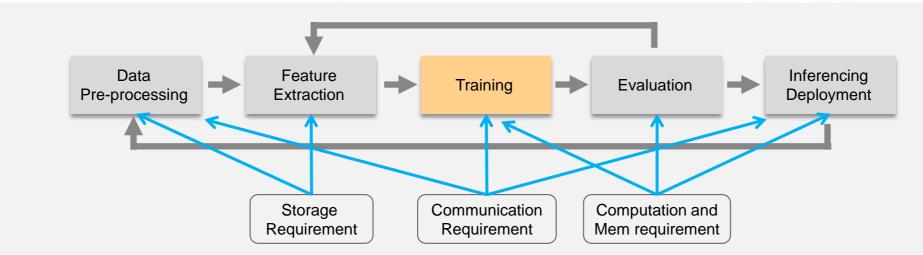
- Exponential growth of large and complex data due to digitalization. In addition, enterprises increasingly
 using various software, such as ERP, CRM, and SCM create a huge amount of data related to
 customers, operations, suppliers, and other stakeholders. Securely storing crucial business information
 with flexible DC is the key to success
- Lack of universally accepted virtualization standards and different vendor hardware and cloud solutions cause
 - ✓ Integration complexity which requires skill and knowledge
 - \checkmark Interoperability and efficiency is not optimized
- DC/AI core and storage configuration are not flexible enough to support heavy computation tasks, such as
 - ✓ Computer vision for image classification, object detection, and video understanding
 - ✓ Ranking and recommendation, such as news feed and search
 - ✓ Language processing for translation, speech recognition, content understanding, etc.





Challenges





- Different operation tasks require different memory and storage configurations. Machine learning is an intensive matrix multiplication task
- Huge operation efforts for recognition of systems for
 - ✓ High capacity, high bandwidth memory
 - ✓ Unstructured accesses benefit from caches
 - ✓ Larger on-chip memory for flexibility of compiler

Strategy – Overview

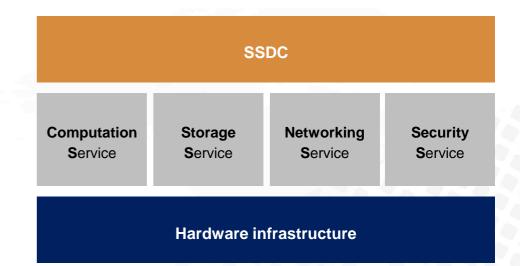
- Adaptive DC to support the secured and scalable data storage, software services and computation efficiency
 - Flexible DC resource configuration based on applications for optimized workload management with greater agility, speed, and security
- Automated and efficient model training and optimization without hassles associated with integration and deployment maintenance support
 - Machine learning requires huge data set and heavy computation for model training and inferencing deployment
 - ✓ Data storage, recovery and cybersecurity along with managing large volumes needs complicated and timeconsuming process. Need to better support various cloud strategies, scalability across heterogenous clouds
- Hardware/software co-design and EDA to scale the software with programmable building hardware
 - \checkmark Concurrency and control feature, especially for many cores
 - ✓ Computation feature that supports scalar and SIMD (Single Instruction, Multiple Data)
 - ✓ Data reuse with software-controlled SRAM
 - ✓ Latency hiding such as hardware for prefetcher, etc.





SSDC (Software-Defined Data Center)

- SDDC is a software-based data storage facility where all the resources are combined to provide the best service
 - ✓ Core CPU/APU/GPU for computation
 - ✓ Storage for data
 - ✓ Networking for communication
 - ✓ Security



SSDC can be planned at hierarchical level

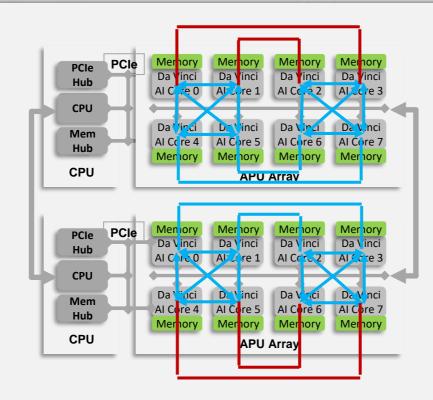
- Data center level which includes servers, storage and networking
- Server level which includes CPU/APU, memory and data flow interconnection

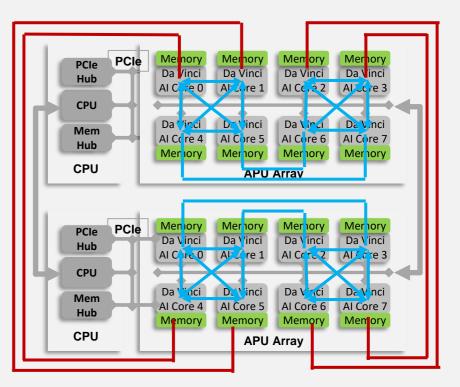
SSDC – Flexible Data Flow and Communication Examples

Flexible PCIe interconnect topology

- ✓ AI core to CPU
- ✓ AI core to AI core direct
- ✓ AI core to AI core via CPU
- ✓ Group of AI core to Group of AI core direct
- ✓ Group AI core to Group of AI core via CPUs

Examples





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SSDC – Standards

The SDDC solution should follow these standards

- Cloud Infrastructure Management Interface (CIMI) by Distributed Management Task Force (DMTF)
- Open Virtualization Format (OVF) specifications
- Organization for the Advancement of Structured Information Standards (OASIS)
- Cloud Application Management for Platforms (CAMP)
- OASIS Topology and Orchestration Specification for Cloud Applications (TOSCA) interfaces
- Storage Networking Industry Association (SNIA) Cloud Data Management Interface (CDMI)

Adaptive Model Training

Automated SSDC

- \checkmark Adaptive configuration based on applications
- ✓ SIMD (Single Instruction, Multiple Data) type of computations
- $\checkmark\,$ Data reuse and stages pipelined computation

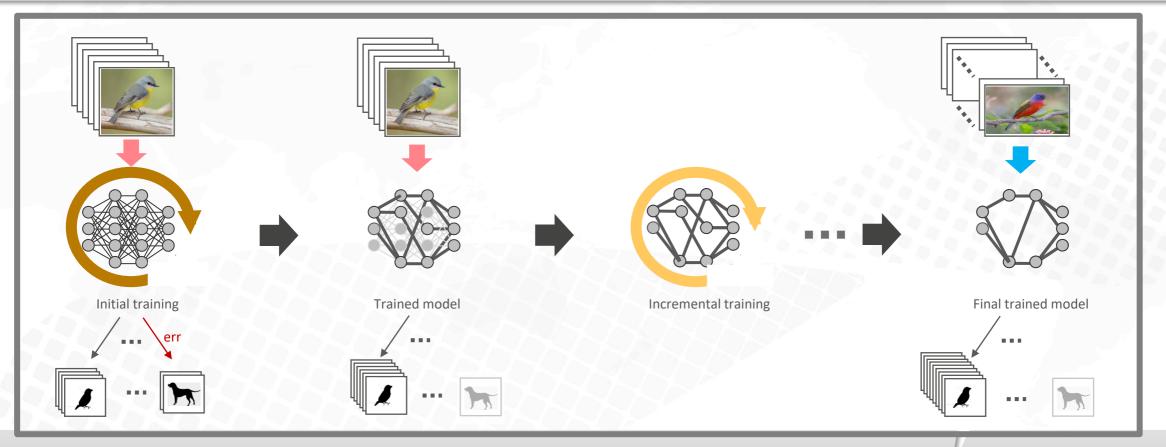
Domain specific AI machine learning

- \checkmark Complicated network orchestration and heterogeneous environments
- \checkmark Optimized training with node pruning
 - Network device distribution
 - Adaptive server and CPU nodes
- ✓ Automated application classification and network/DC configuration



Adaptive Autonomous – Example

- Deep learning model training and optimization
 - ✓ Automated learning
 - ✓ Model pruning for optimization
 - ✓ Potentially reduced precision. Incremental training necessary





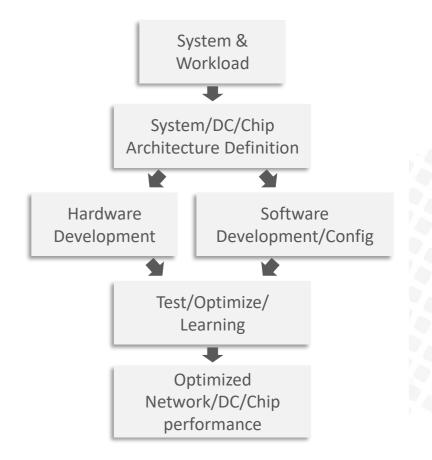
Hardware/Software Co-Design EDA

To improve the efficiency of software execution

- ✓ Hardware design should support the optimal software execution
 - Type of computation tasks intent learning
 - Type of data and their size structured data
 - Priority of latency, throughput, bandwidth computation flow optimization
- Server, CPU/APU, storage should be configured to best execute the tasks – intelligent configuration model
 - Twine models for simulation and learning
 - o Dynamically adjust model based on continuous incremental learning

Hardware/software co-design EDA tasks

- Determine the hardware feature requirement and configuration
- Build configuration model library and learning algorithm
- Develop key KPIs for optimization measurement, such as various of workload scheduling measurement, throughput stages, etc.



Hardware/Software Co-Development Approach



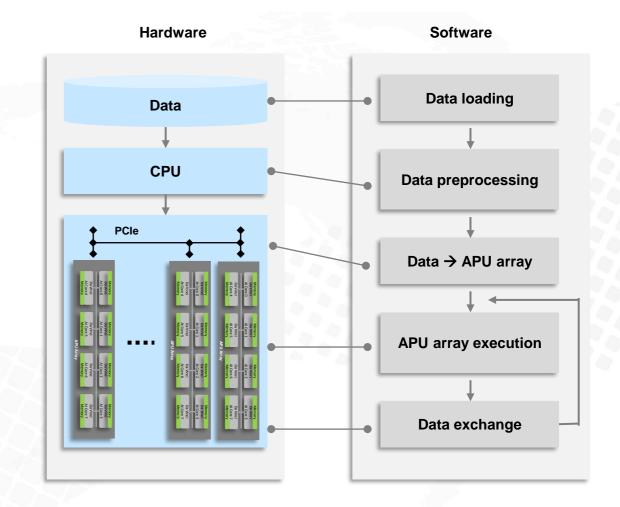
Hardware/Software Co-Design EDA – Example

Co-design task and challenge

- Computation
 - Both maximal sharing and performance
 - Parallel, streamlined, Pipelined, asynchronous, etc.
- ✓ Storage
 - Storage depth (size) and width (speed)
 - Pool storage vs synchronization/pipelining storage
- ✓ Communication network
 - Between storage data
 - o Between storage data and CPU
 - o Between APU and storage data
 - o Between APU and APU

Co-design library components

- ✓ Scalar for computation and storage
- Vector macro for computation and memory
- Repetition and branching timing and control
 Etc.





Hardware/Software Co-Design EDA – Analysis

Plan all types of tasks for deep learning models, such as

- Computation dominated
 - Top MLP
- Communication dominated
 - Feature extraction and analysis
- ✓ Memory bandwidth dominated
 - o Bottom MLP
 - EMB lookup
- ✓ Memory capacity dominated
 - Dense features
 - Sparse features

Benchmark model library

- ✓ Training and recommendation
- ✓ Incremental collection and justification



Summary

Adaptive SDDC requires the supports from

- ✓ Hardware and software co-design
- ✓ AI machine learning to automate the analysis of the intended tasks
- ✓ Libraries for configuration models, machine learning models, execution framework and scalar, macro instruction set, etc.
- ✓ Automated template recommendation including configuration and algorithm for service tasks
- ✓ Twin model simulation and dynamical model adjustment

Community sharing

- ✓ Increase OCP availability
- ✓ Advanced training model availability from various applications





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