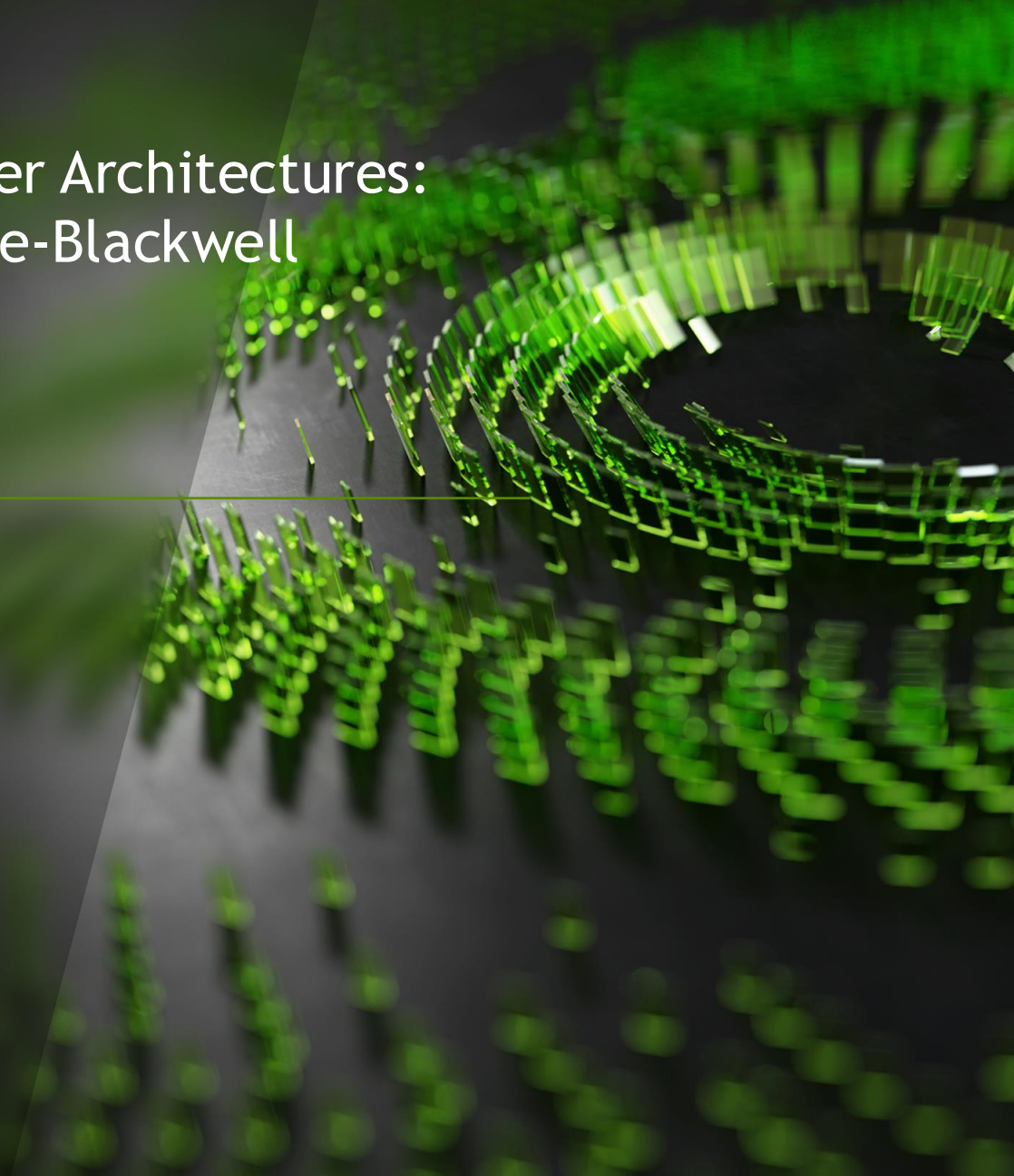


Rapidly Evolving HPC, SuperComputer Architectures: Nvidia Grace-Hopper, Grace-Blackwell

Giri Chukkapalli





EMERGING GRAND CHALLENGE APPLICATIONS

OCT 14, 2022

EMERGING GRAND CHALLENGE APPLICATIONS

- Full systems modeling
 - Earth systems: Complex Climate Models
 - Biological systems: Human Microbiome Modeling
 - Complex Engineering Systems: Aircraft, Power Plant etc.
 - Environmental modeling like distribution of Micro-Plastics
 - Novel Drug designs, Material designs (Inverse problems)

EMERGING GRAND CHALLENGE APPLICATIONS

- They span across several orders of magnitude space and time scales
- Span across large number of physical, chemical, quantum phenomena
- Example could be influence of Aerosols to Ocean Circulation

EMERGING GRAND CHALLENGE APPLICATIONS

- At the same time silicon process technologies running out of steam

EMERGING GRAND CHALLENGE APPLICATIONS

- Classic brute force ModSim algorithms may not be sufficient or possible
- Classic Fortran + MPI is not sufficient
- Homogeneous CPU Nodes or GPU Nodes are not sufficient
- Homogeneous ModSim clusters may not be sufficient

EMERGING GRAND CHALLENGE APPLICATIONS

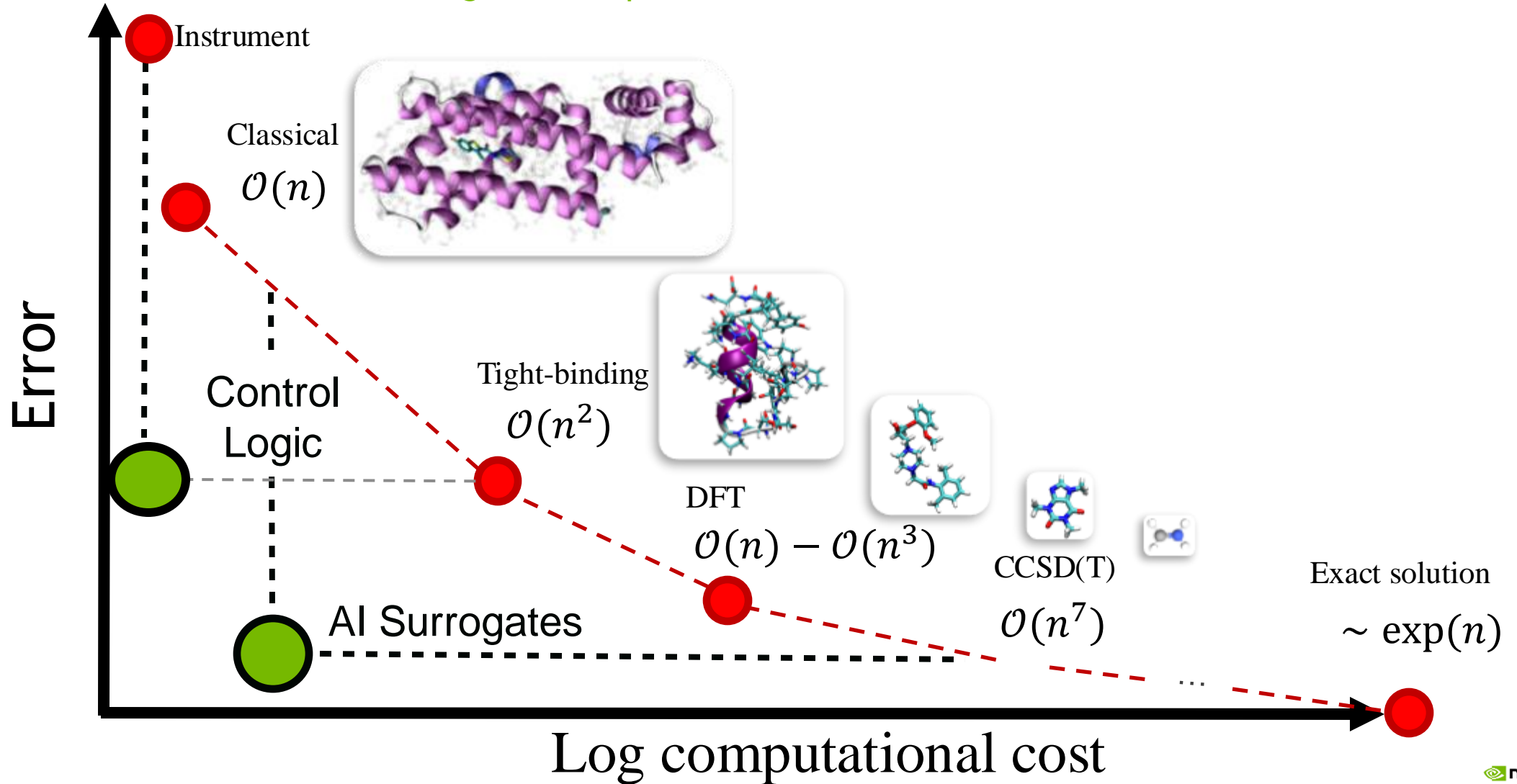
- Data and Algorithms (ML/DL) to the rescue
- ModSim + ML/DL + Data

ML/DL WILL CONTINUE MOORE'S OBSERVATION IN HPC

- ▶ ML/DL is rapidly becoming the 4th pillar of Scientific Discovery
- ▶ Approximations in Mod-Sim, Inverse problems, Insights from experimental/empirical data etc.
- ▶ Protein Structure Prediction became a solved problem
- ▶ Sub-Seasonal Forecast Competition is won by MSFT ML/DL Team
- ▶ Physics informed NNs in CFD

AI INTRODUCES NEW USE CASES FOR SCIENCE AND ENGINEERING

AI Bridges the Gap Between Simulation and Real-Time



Chemical and Material Discovery is about to have its ChatGPT moment

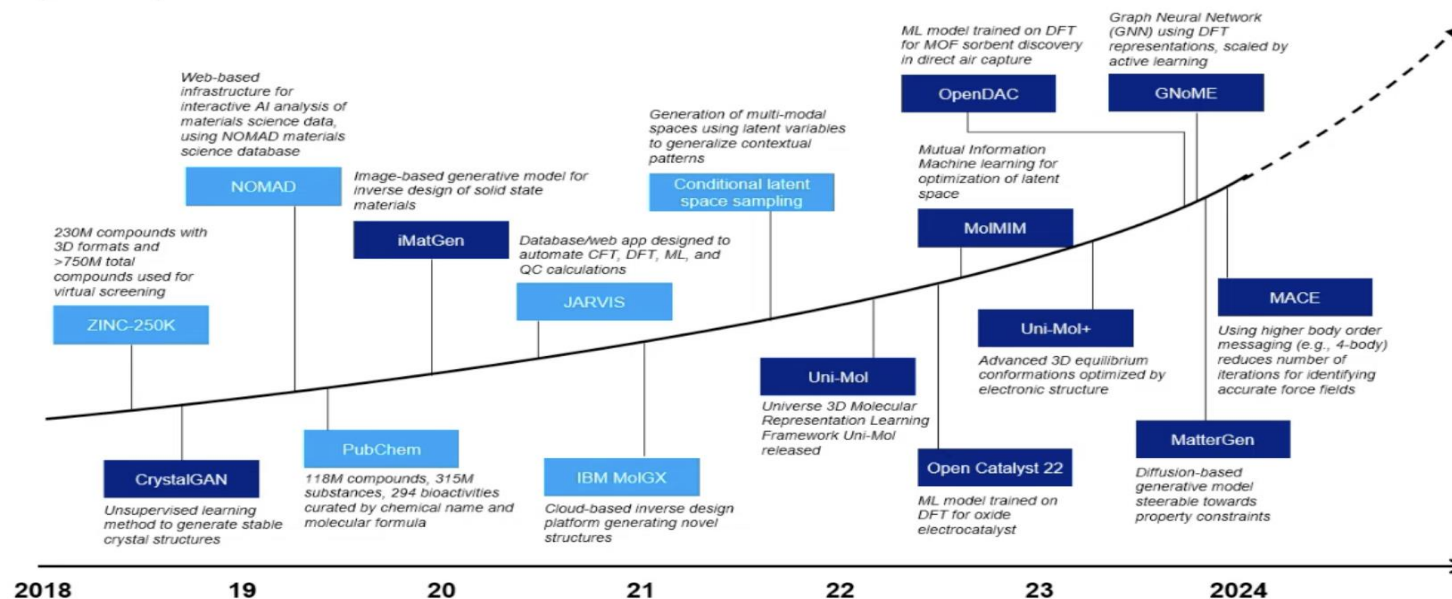
The pace is changing from years to months on new datasets and models.

Last 6 months

Not Exhaustive

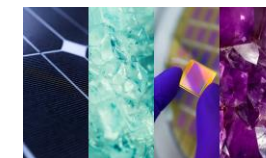
Age of accelerating AI/GenAI breakthroughs for chemicals and materials discovery

Yearly trend in publications



Source: Press search

 Datasets/Tools
 Models



GNoME/Deepmind
Nov' 2023



MatterGen/MSFT
Dec' 2023



MACE-MP-0
(Academia/SCC)
March 2024



MatterSim/MSFT
May 2024



CatTSunami
/Meta May 2024

Leaderboard for AI models for Material Discovery : <https://matbench-discovery.materialsproject.org/>

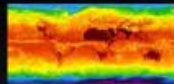
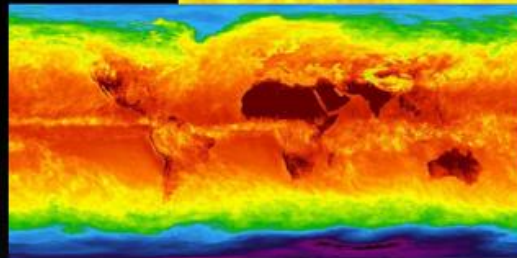
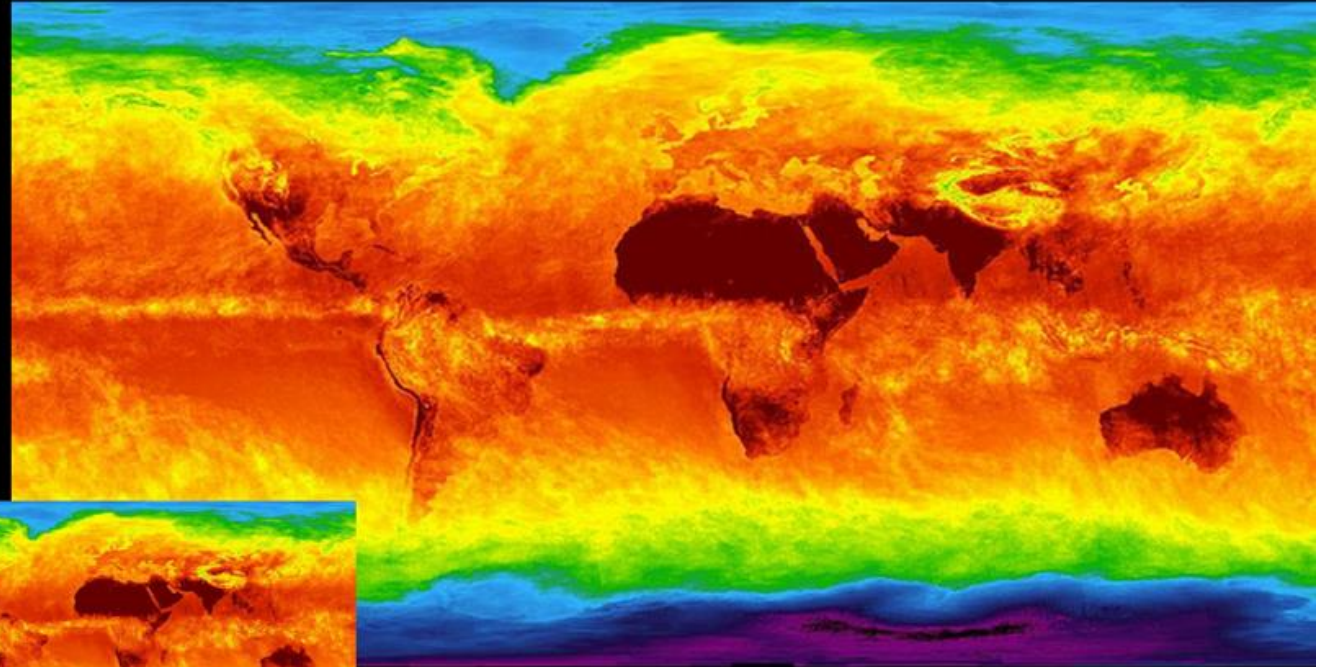


AI ALGORITHMS EVOLVING AT UNPRECEDENTED PACE

FourCastNet High Resolution for Data-Driven Weather Models

Comparison of resolutions for data-driven weather models since 2018 (Dueben & Bauer)

SOTA evolving rapidly
Recent Pre-print Kang Chen et al (2023) extend forecast to 10 days with 0.25° resolution using “cross modal Transformer”



Deuben & Bauer (2018), 6°, 60x30, 1.8K pixels, MLP

GNN, Keisler et al. (2022), 1°, 64,000 Pixels, Graph Neural Networks

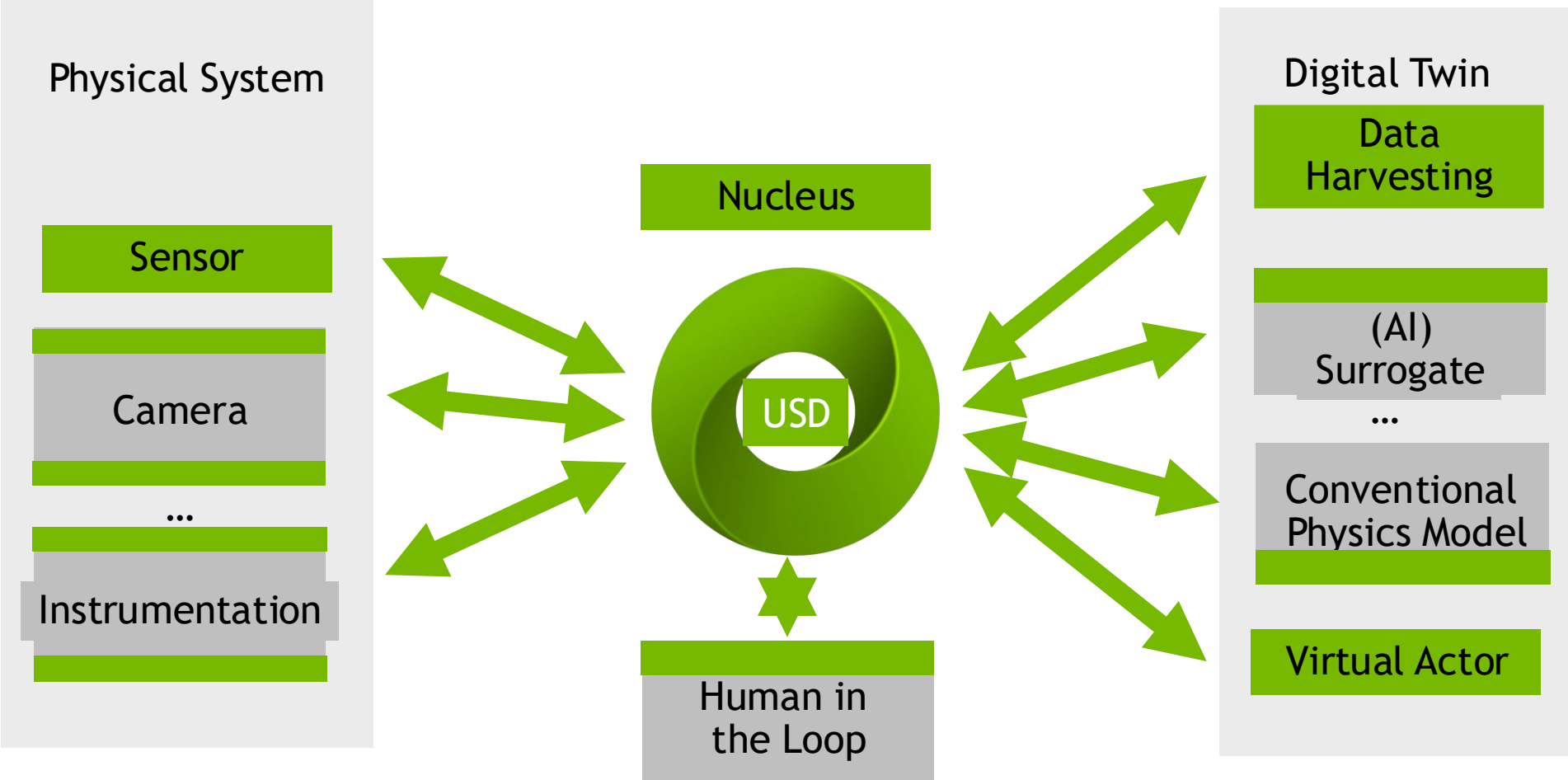
DLWP, Weyn et al. (2020). 2°, 16K pixels, Deep CNN on Cubesphere/(2021) ResNet

Weyn et al. (2019), 2.5° N.H only, 72x36, 2.6k pixels, ConvLSTM

WeatherBench, Rasp et al. (2020). 5.625°, 64x32, 2K pixels, CNN

FourCastNet, Pathak et al. (2022), 0.25°, ~1,000,000 Pixels, ViT+AFNO

Digital Twin For Science

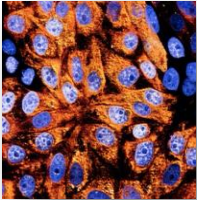


EMERGING GRAND CHALLENGE APPLICATIONS

- Supercomputers may need to integrate with sensor networks
- High throughput scientific instruments
- Most importantly, capable of solving HPC + ML in a tightly coupled loop

Workloads of the Modern Supercomputer

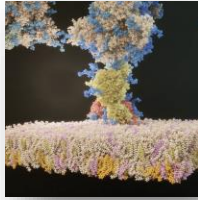
EDGE



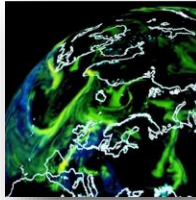
SIM + AI



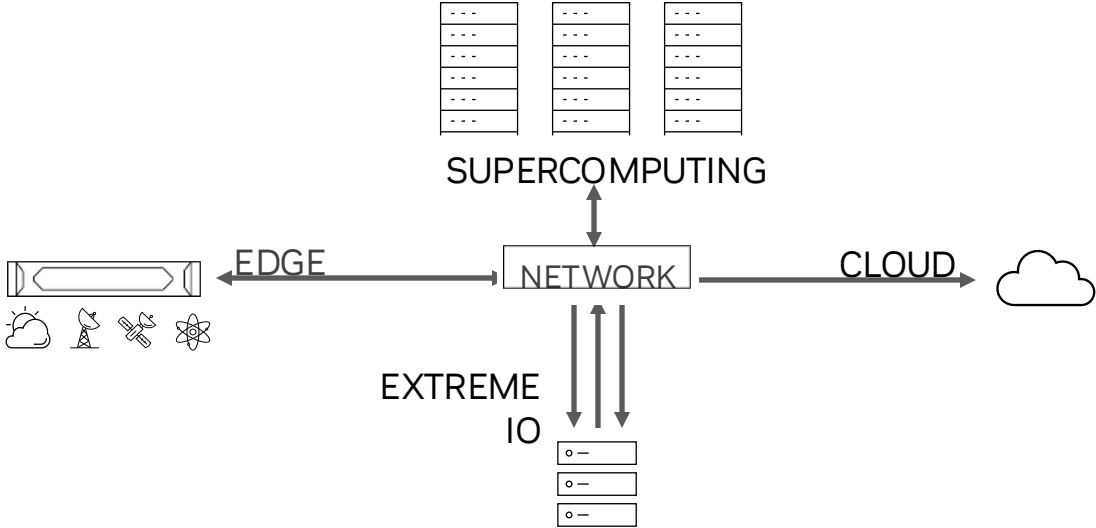
SIMULATION



DIGITAL TWIN



QUANTUM COMPUTING



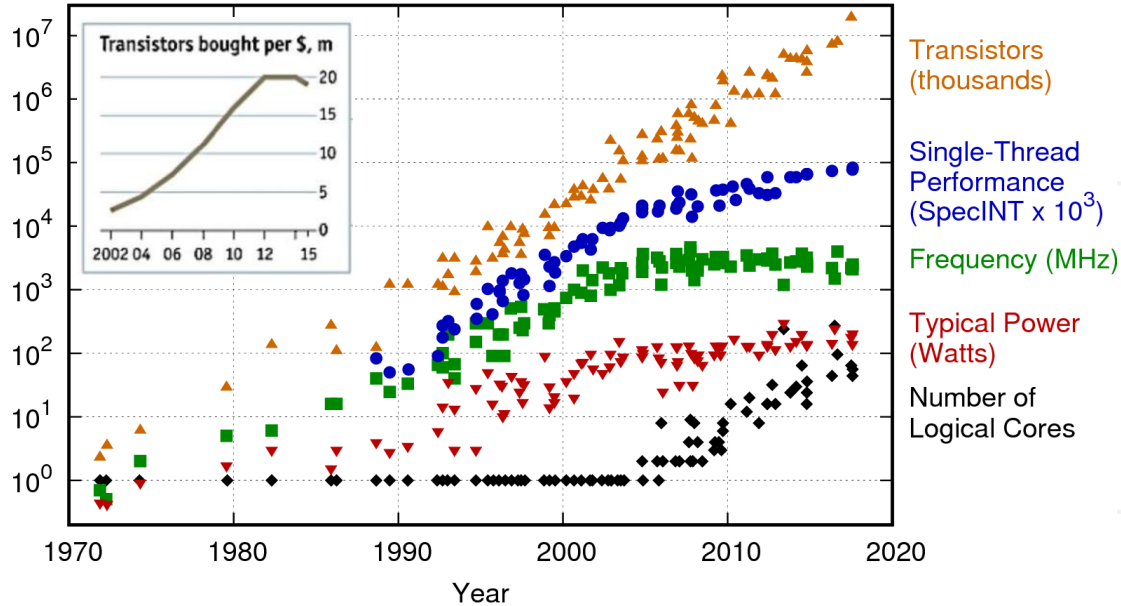


POST-MOORE SILICON SCALING

ISO PLAN + UNIFIED SDK + LIBRARIES

MOORE'S OBSERVATION

42 Years of Microprocessor Trend Data



Transistors (thousands)

Single-Thread Performance (SpecINT x 10³)

Frequency (MHz)

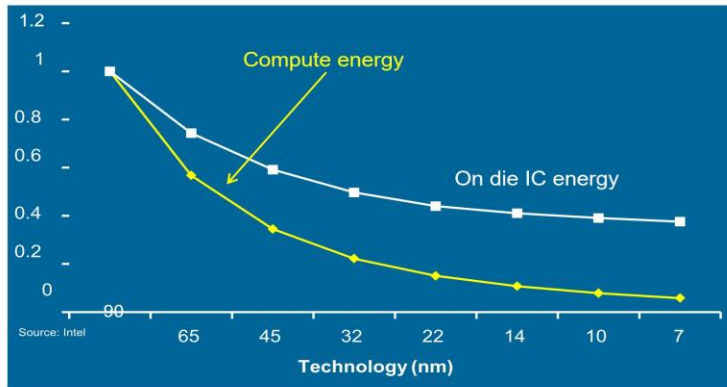
Typical Power (Watts)

Number of Logical Cores

Original:

- ▶ Transistor density doubles
- ▶ Power stays same and Cost stays same
- ▶ Last two are gone
 - ▶ Due to loss of Dennard scaling and Litho and other process complexities
- ▶ Different features scale differently with process
 - ▶ Logic, SRAM, Wires, and IOs
 - ▶ Distorting over time

Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten
New plot and data collected for 2010-2017 by K. Runn

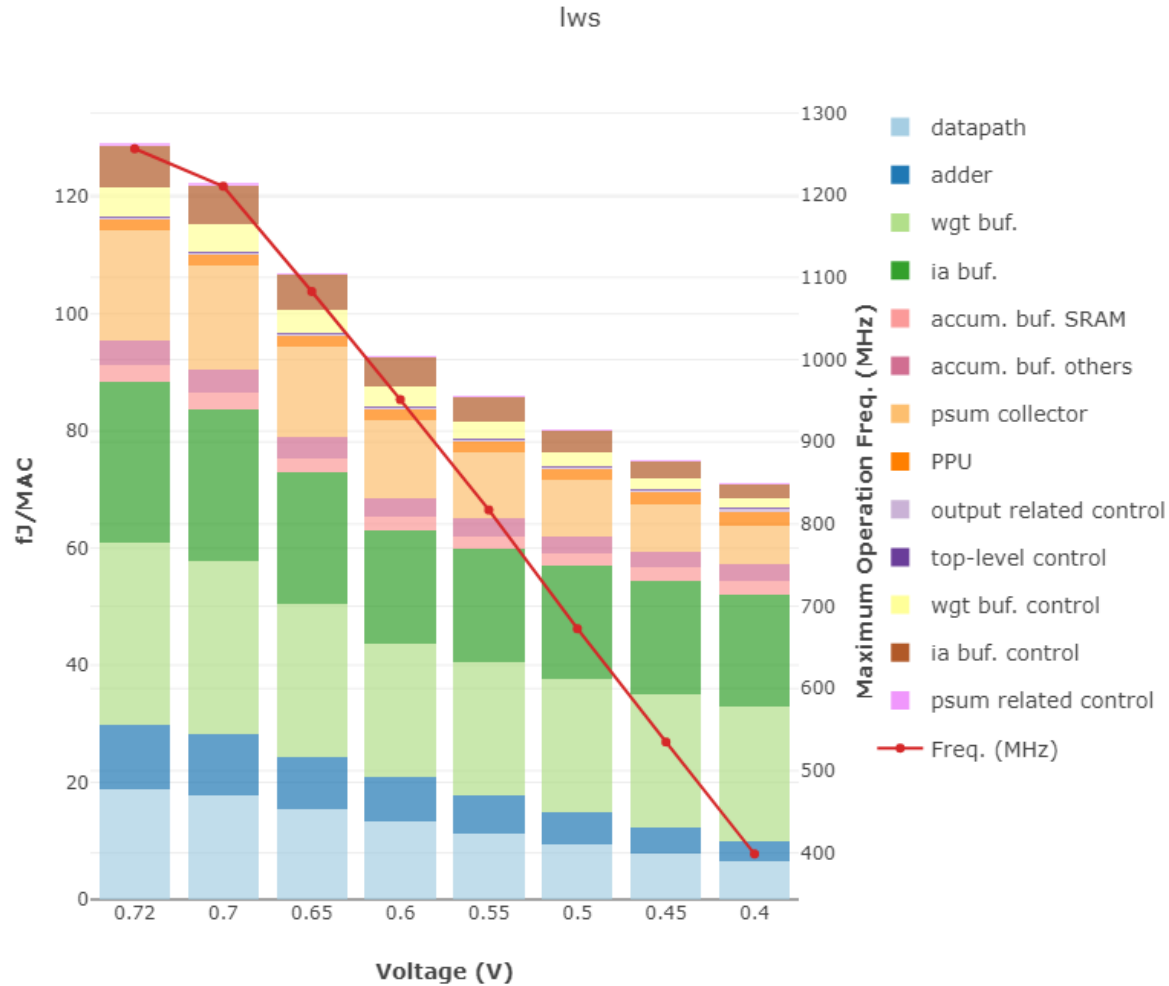


ENERGY DOMINATED BY MEMORY AND DATA

70 fJ/MAC

35 fJ/OP

29 TOPS/W



MORE THAN MOORE

- ▶ Nvidia maintains “More than Moore” by optimizing
 - ▶ End-to-End mapping of Applications to Supercomputing System
 - ▶ Algorithmic, SW, architectural, Packaging, Process Technology
 - ▶ Try to maintain 2X to 6X Gen-to-Gen perf improvement
 - ▶ Requires close collaboration with HPC/ML/DL Community
 - ▶ Transformer Acceleration in Hopper is the best recent example
- ▶ FLOPs not executed, Bytes that are not moved are the best FLOPs and Bytes



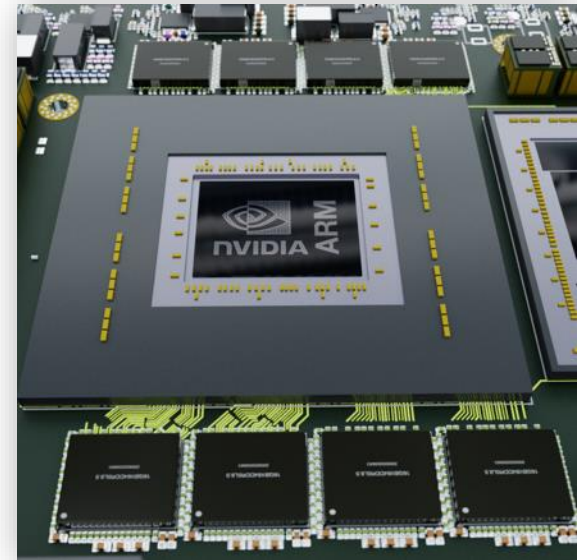
GRACE ARM CPU

ISO CAN + UNIFIED SDK + LIBRARIES

NV ARM OVERVIEW

Server Class ARM CPU

- ▶ 64bit Server Class Core and SoC
 - ▶ Arm V9.0 ISA Compliant aarch64 core
 - ▶ Full SVE-2 Vector Extensions support, inclusive of NEON instructions
 - ▶ Supports 48-bit Virtual and 48-bit Physical address space
- ▶ Balanced architecture between Single Core Perf, Core count, Memory and IO subsystems
- ▶ Supports Arm Server ready (SBSA), Boot compliant (SBBR) and manageability (SBMG) open standards
 - ▶ Linux, system management, HPC stacks will run out of the box



NVIDIA ARM CORE

Optimized for Single Thread Performance

Scalar Side

- ▶ Wide Super Scalar OoO single-threaded high performance Core Pipeline
 - ▶ Supporting Multi-stage branch prediction and advanced prefetching algorithms
 - ▶ 1LD + 4LD/ST pipes, 6 ALU pipes
- ▶ 64B Cacheline
- ▶ 64KB L1ICache, 64KB L1DCache
- ▶ 1MB Private L2Cache

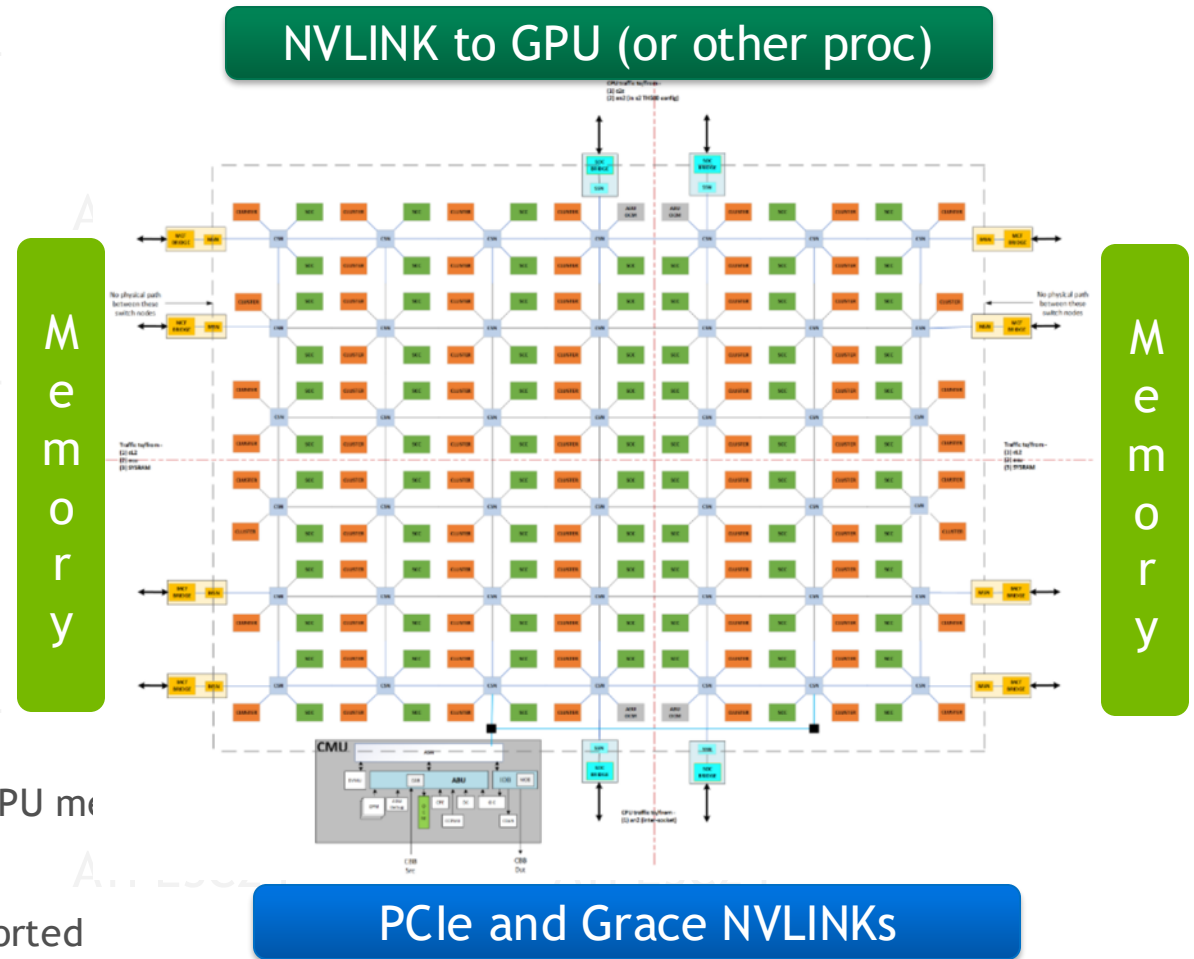
Vector Side

- ▶ Four 128-bit SVE2 Execution pipes capable of 16 DP FLOPS per Cycle
- ▶ Support 64bit, 32bit, 16bit and bfloat16 FP and int8
- ▶ Complex datatypes and math
- ▶ Enables easier vectorization through Predicate and mask instructions
- ▶ Lane widening and narrowing instructions
- ▶ Classic and non-temporal Gather Load and Scatter Store instructions
- ▶ Crypto extensions

PROCESSOR SOC

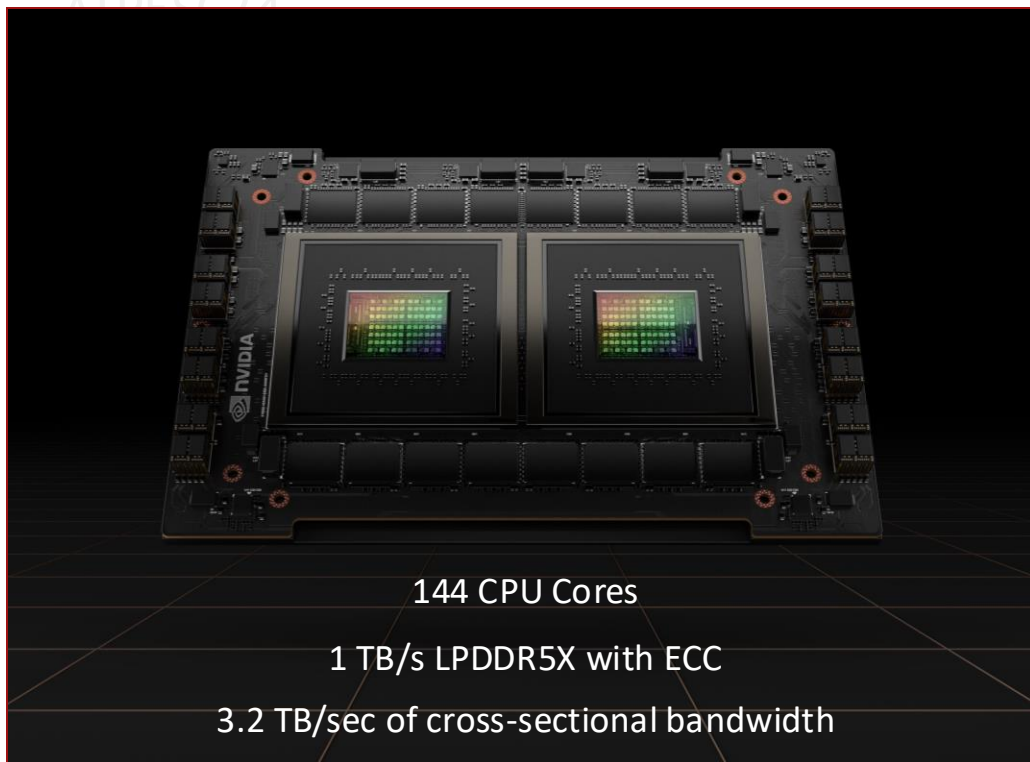
Augmented custom logic to support memory movement

- ▶ Monolithic SoC
 - ▶ Up to 117MB shared L3 cache
 - ▶ >3TB/s on-die mesh bisection BW
- ▶ Extensive set of Core and un-Core perf counters
- ▶ Thermal monitoring and power management
- ▶ DVFS support with multiple voltage domain
- ▶ Individual core power and clock gating support
- ▶ Tx and Rx paths optimized for 400Gbps fabric
- ▶ ARM V9 ISA virtualization and security support
- ▶ Custom SoC level logic support for GPUDirect, CPU-GPU movement and synchronization
- ▶ 120GB, 240GB, 480GB LPDDR5 CapaATPESC24es supported



NVIDIA Grace and Grace Hopper

High Performance for an Energy Constrained World



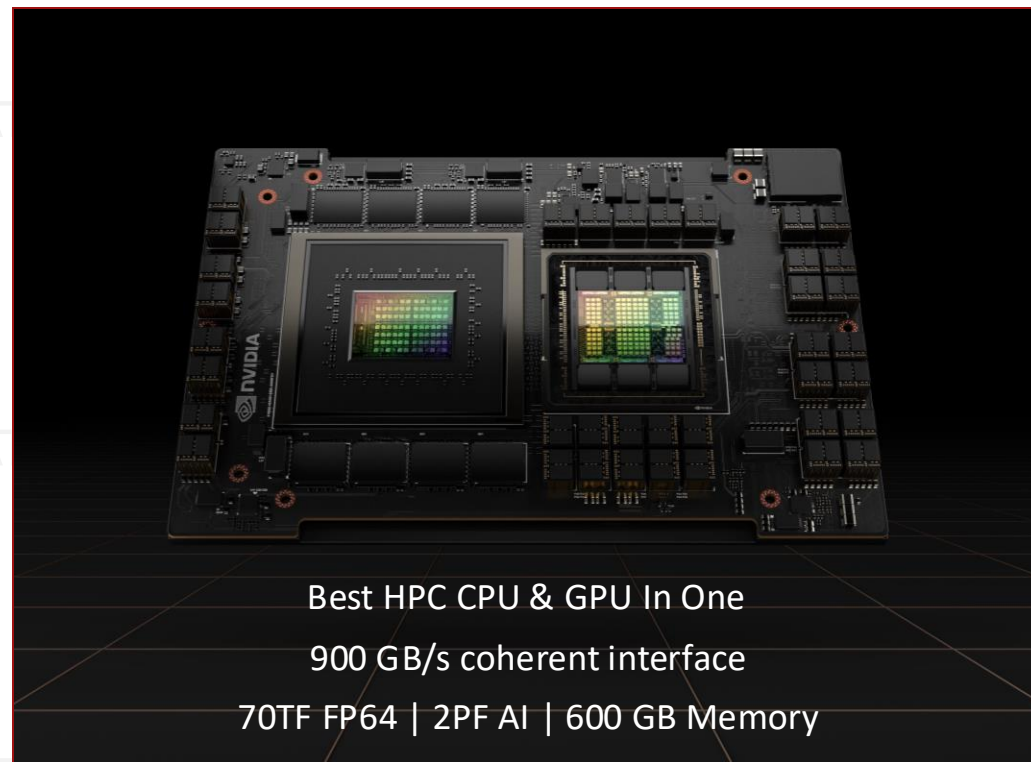
144 CPU Cores

1 TB/s LPDDR5X with ECC

3.2 TB/sec of cross-sectional bandwidth

Grace CPU Superchip

High-performance CPU for HPC and cloud computing



Best HPC CPU & GPU In One

900 GB/s coherent interface

70TF FP64 | 2PF AI | 600 GB Memory

Grace Hopper Superchip

CPU+GPU designed for giant-scale AI and HPC

NVIDIA ARM HPC SW ECOSYSTEM

HPC Applications

Simulation

Data Analytics

AI

HPC SDK

COMPILERS

NVC++

NVC

NVFORTTRAN

NVCC

CORE C++ LIBRARIES

CUB

Thrust

libc++

PROFILING AND DIAGNOSTICS

CUDA-gdb

Nsight Compute

Nsight Systems

CUPTI

COMMUNICATION LIBRARIES

MPI

HPC-X

OpenSHMEM

UCX

NCCL

SHARP

HCOLL

NVSHMEM

RAPIDS

cuDF

cuxfilter

cuSpatial

cuSignal

cuGRAPH

cuML

DL INFERENCE

Triton Inference Server

TensorRT

TRAINING (DLFW)

Pytorch

TensorFlow

DL LIBRARIES AND SOFTWARE

DALI

cuDNN

CUTLASS

MATH LIBRARIES

cuBLAS

cuTENSOR

cuSOLVER

cuSPARSE

cuFFT

cuRAND

MATH API

(OpenBLAS)

CUDA Runtime

Platform Software

Fabric Manager

GPUDirect

NVML

NVIDIA Linux GPU Driver for ARM

DCGM

(Slurm)

(Docker)

(Singularity)

(Third party)

NVIDIA supported

Work in Progress



HOPPER GPU

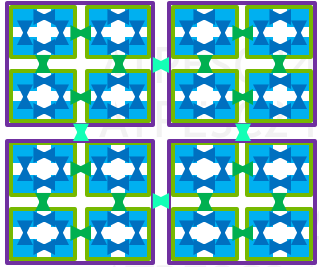
ISO CAN + UNIFIED SDK + LIBRARIES

NEW HOPPER SM DOES MORE THAN IMPROVE RAW SPEEDS AND FEEDS

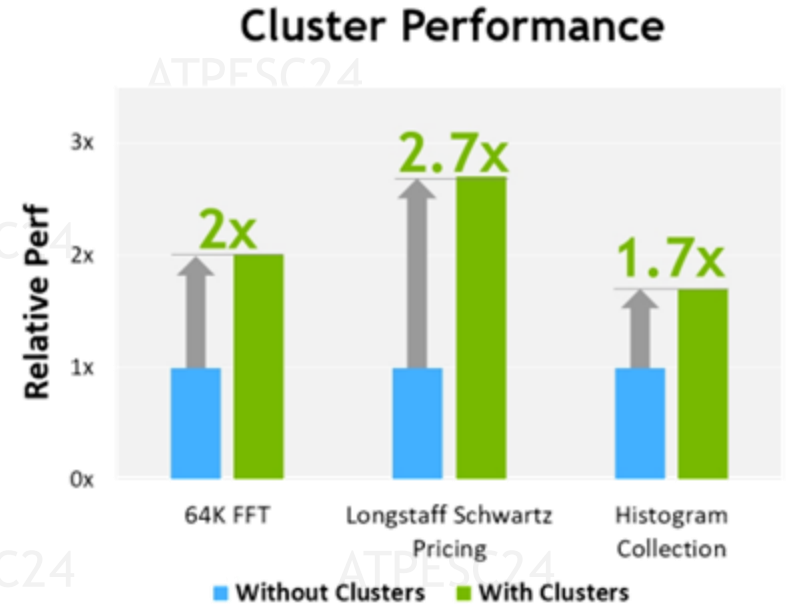
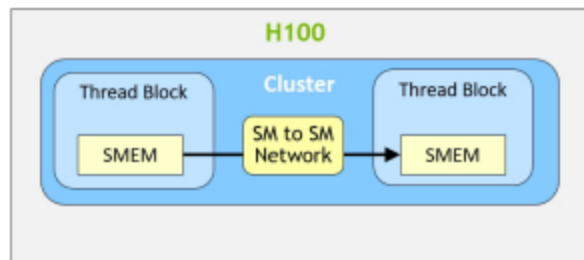
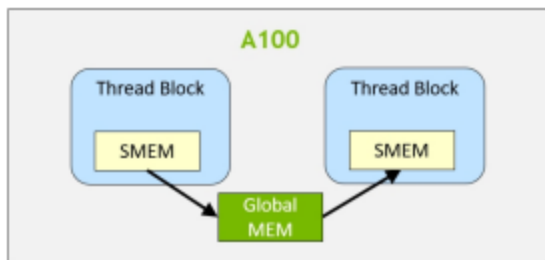


- New **Thread Block Clusters**
 - Turn locality into efficiency
- Support **Distributed Shared Memory** between SMs
- New **Asynchronous Transaction Barriers**
 - Increased support for asynchronous programming
- New **Tensor Memory Accelerator**
- Fully asynchronous data movement
- New **DPX** instruction set
 - Special Purpose Acceleration
- New **Transformer Engine** for AI Model Acceleration

THREAD BLOCK CLUSTERS



- New feature introduces programming locality within clusters of SMs
- About 7X higher throughput vs. using global memory
- Shared memory blocks of SMs within a GPU Processing Cluster (GPC) can communicate directly (w/o going to HBM)
- Leveraged with CUDA cooperative groups API

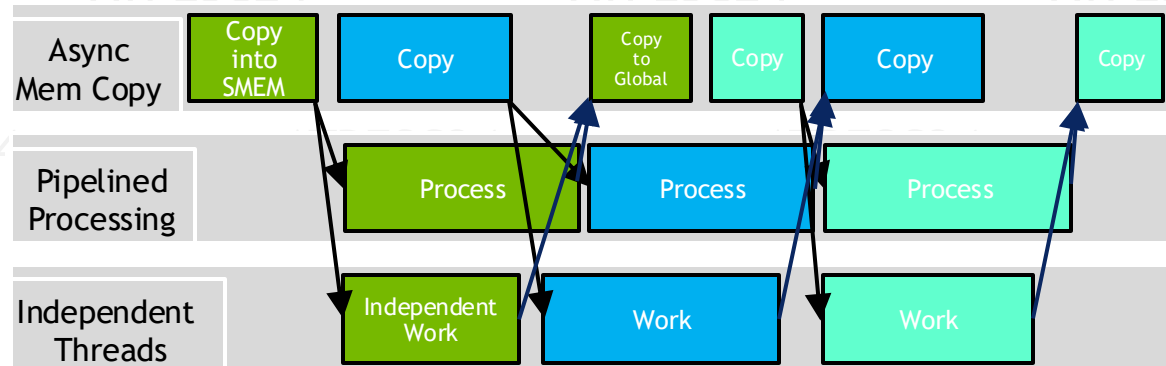


For details, see “NVIDIA H100 Tensor Core GPU Architecture” white paper available for download

ASYNCHRONOUS ENHANCEMENTS

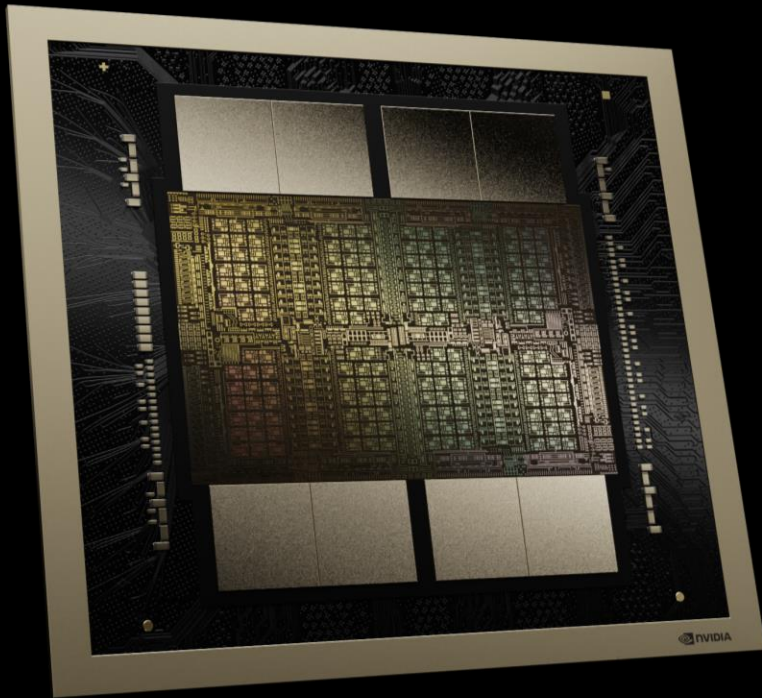
Hopper enables end-to-end fully asynchronous pipelines

- Async Transaction Barriers - Atomic data movement with synchronization
- More efficient Waiting on Barriers
- Async Mem_copy via Tensor Memory Accelerator (TMA)



NVIDIA Blackwell

An Engine for High-Fidelity Simulation and Trillion Parameter AI



Blackwell GPU

FP64 FMA	40 TFLOPS
FP64 TC	40 TFLOPS
FP32 FMA	80 TFLOPS
HBM Bandwidth	Up to 8 TB/s
HBM Capacity	Up to 192 TB

Simulation capabilities 1.3x – 2x higher than H100



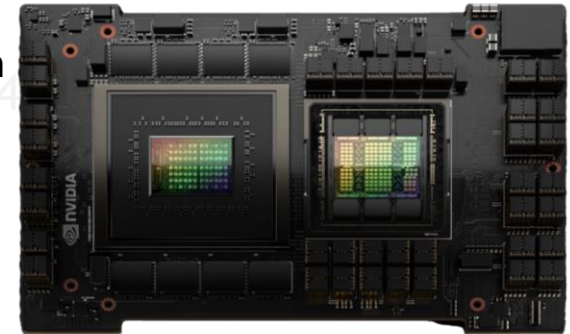
GRACE-HOPPER ARCHITECTURE & RATIONALE

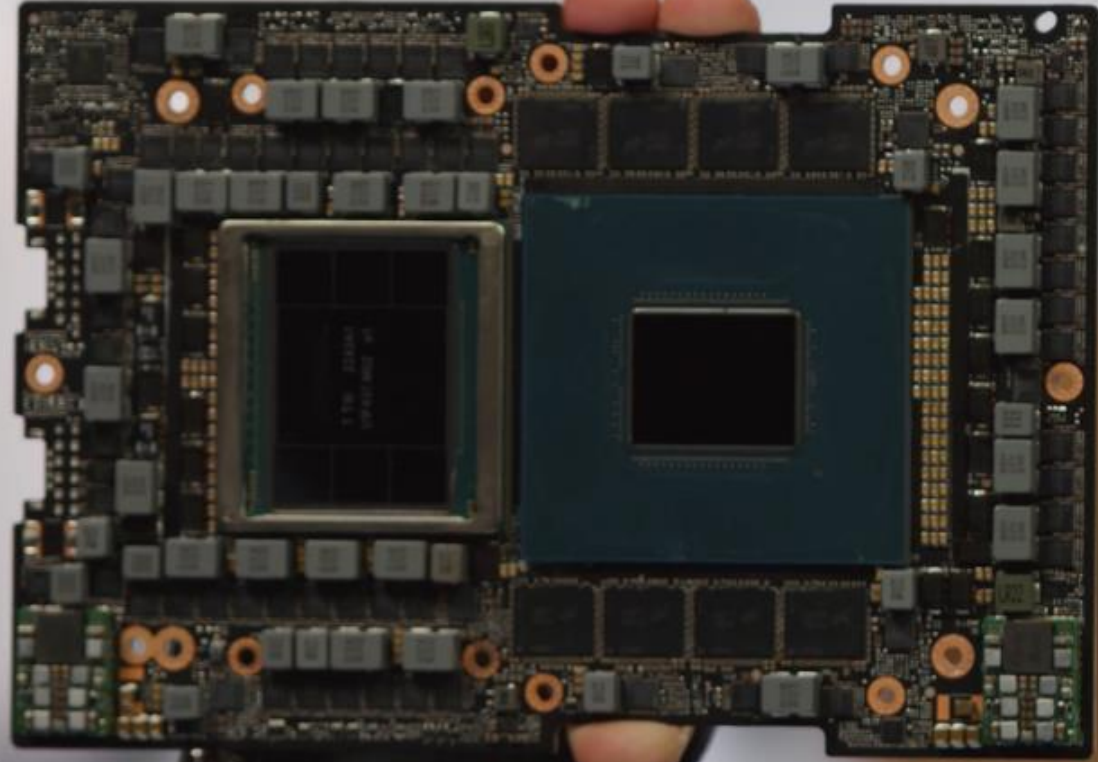
ISO CAN + UNIFIED SDK + LIBRARIES

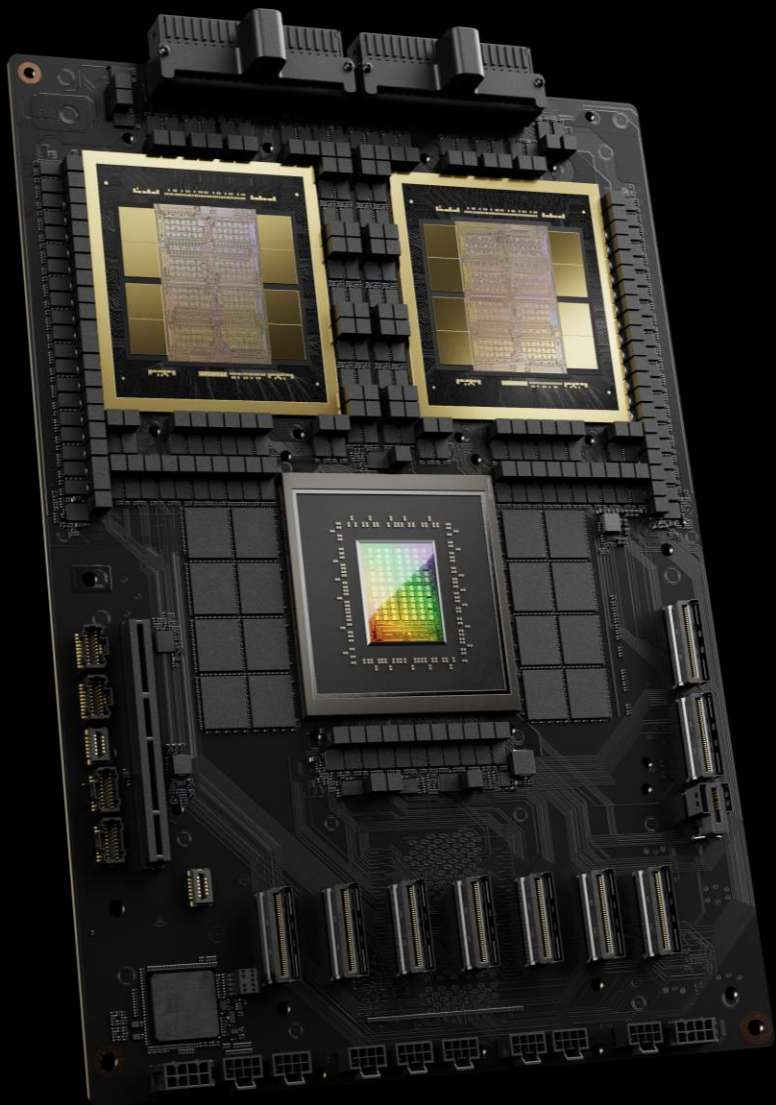
GRACE-HOPPER

A revolutionary Architecture

- Nvidia GPUs
 - Latency hiding Throughput Machines => Async Computational Graph solvers
 - Can effectively map *Dataflow-Complex* portions of the Algorithm
 - Custom (compute and memory) IP Blocks for energy efficiency
- Nvidia CPUs
 - SuperScalar, OOO Core based tightly coupled SoC with balanced Bytes/s/FLOPS
 - Can effectively map *ControlFlow-Complex* portions of the Algorithm
 - Strong Vector and Tensor performance in future
- Unified Compute Substrate







Grace Blackwell SuperChip

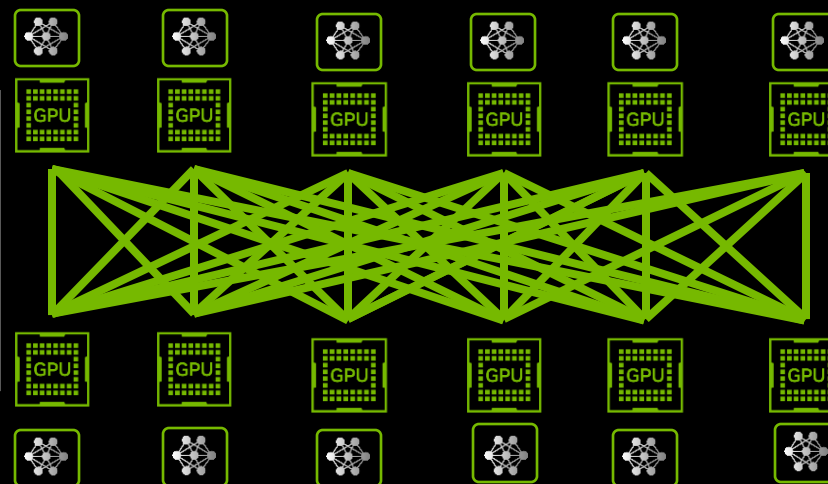
FP8	20 PFLOPS	2.5X Hopper
NEW FP6	20 PFLOPS	2.5X
NEW FP4	40 PFLOPS	5X
HBM Model Size	740B param	6X
HBM Bandwidth	34T param/sec	5X
NVLINK All-Reduce with SHARP	7.2 TB/s	4X

GB200 With NVL72 Enabling Trillion Parameter AI

NVL72: One Big GPU

GB200 NVL72

- 36 GRACE CPUs
- 72 BLACKWELL GPUs
- One NVLink Domain
- 130 TB/sec All-to-All Bandwidth
- 18x vs HDR





PROJECT CEIBA UPGRADED



GPU 20,000 GB200 NVL72

AI 400 EF

Hosted by AWS

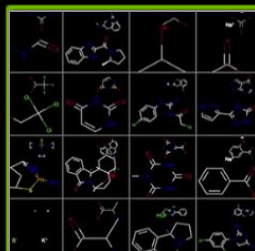
NVIDIA AI Accelerated Computing Platform



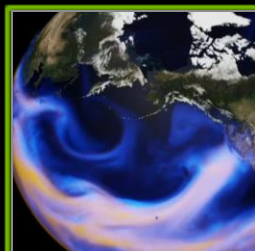
DATA
PROCESSING



CAD, CAE, SDA



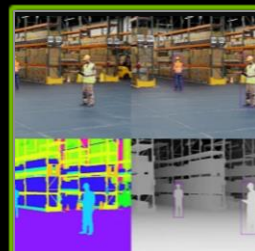
COMPUTER-AIDED
DRUG DESIGN



CLIMATE
SIMULATION



QUANTUM
SIMULATION



ROBOTICS
&
INDUSTRIAL
DIGITAL
TWINS

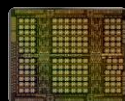


ENTERPRISE
AI

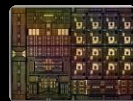
CUDA-X LIBRARIES



CPU



GPU

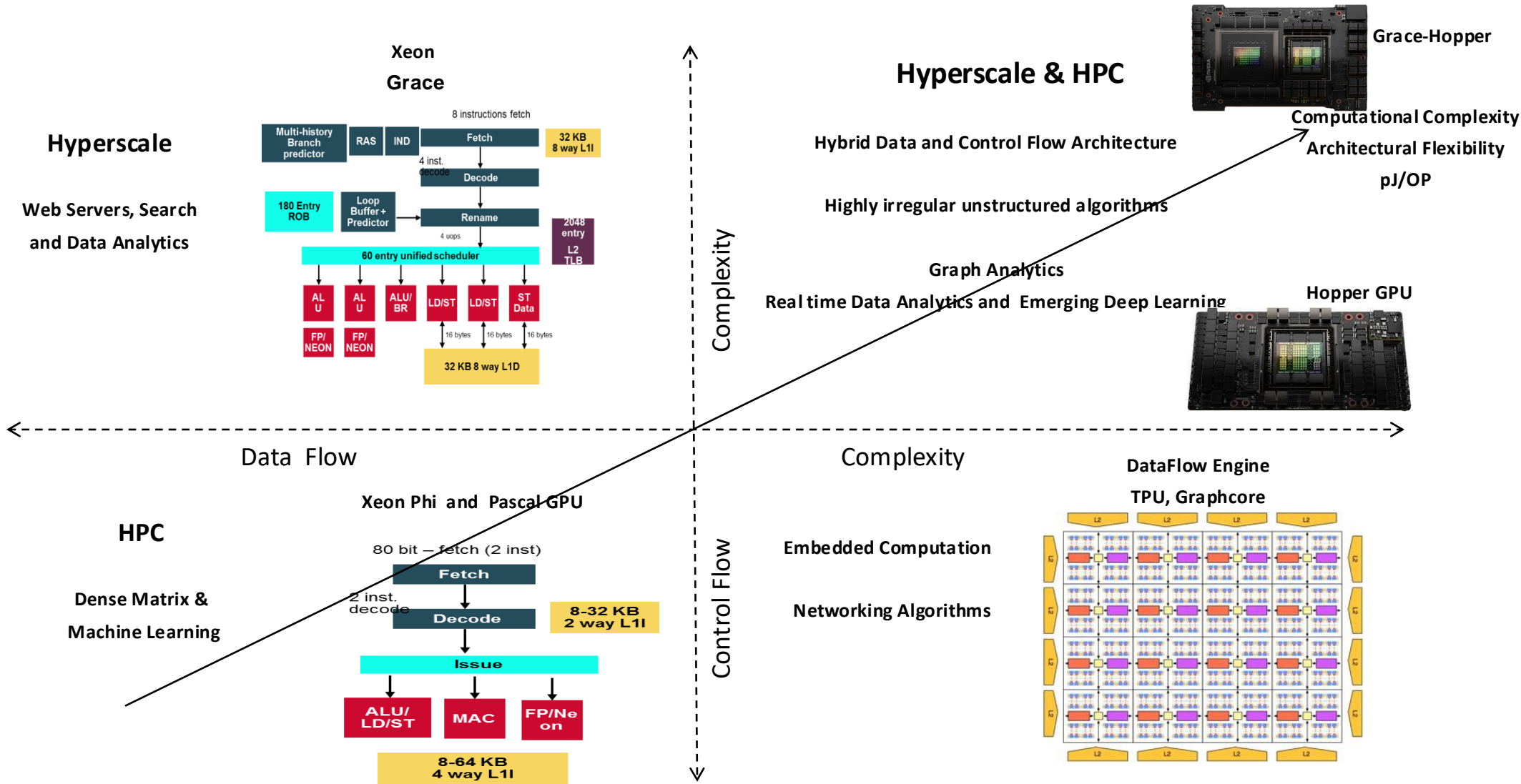


DPU

ACCELERATED COMPUTING



Application Complexity & Architectural Flexibility

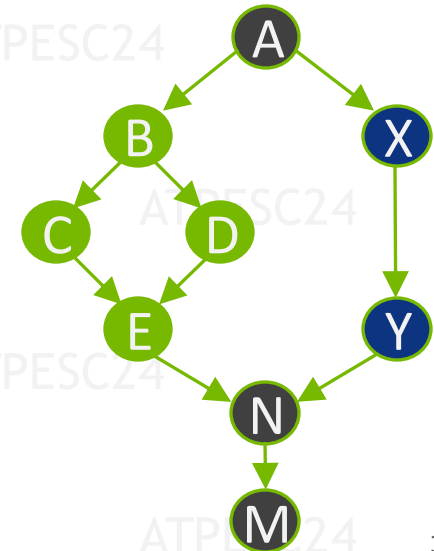
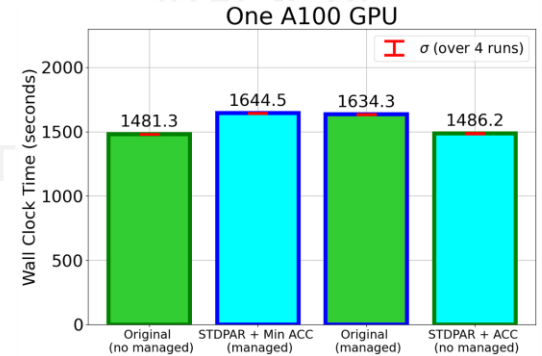


COMPUTATIONAL GRAPH + GRACE HOPPER

Same programming model for CPU and GPU, plus existing applications ready-to-run Day 1.

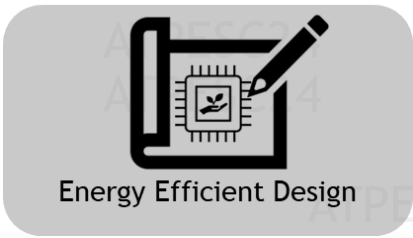
- True Unified Memory + High Bandwidth Link make data movement less of a bottleneck
- Existing CPU programming models continue to work on Grace CPU with high performance

Senders/Receivers enable defining hybrid execution graphs to take advantage of the strengths of each processor.



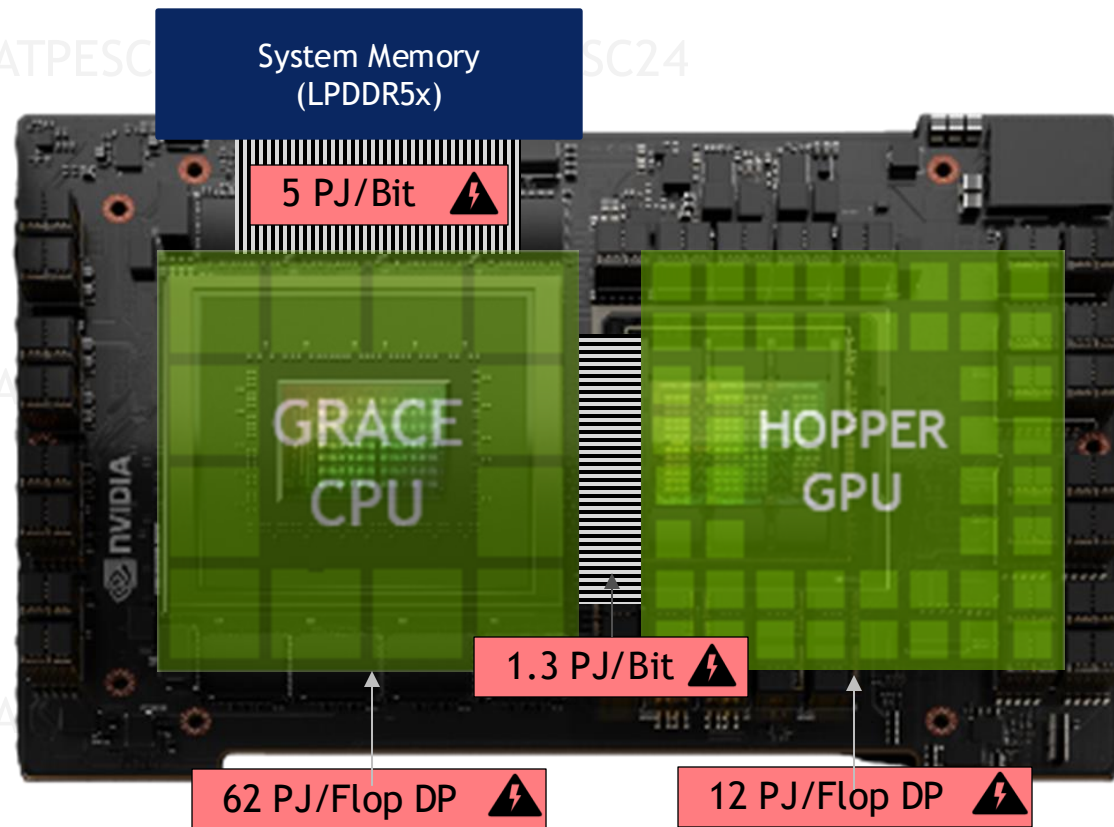
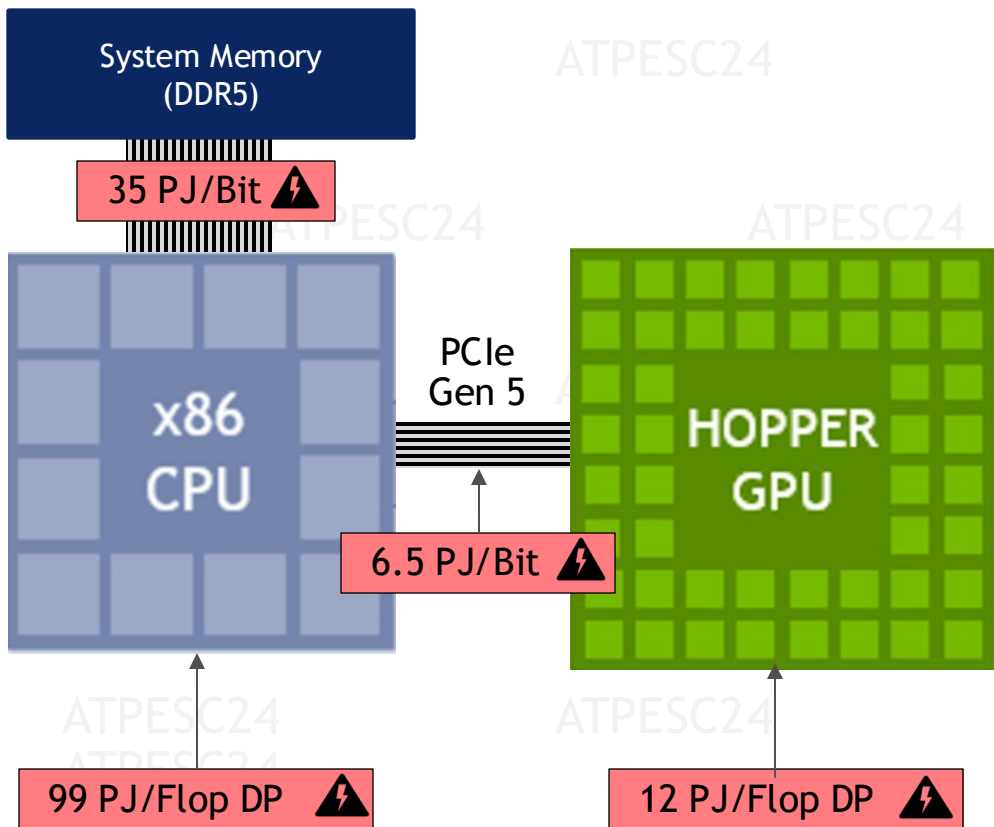
UNIFIED COMPUTE SUBSTRATE

- Grace + Hopper Enables broader set (all!) of codes to be accelerated
- Grace + Hopper enables architectural mapping of both control and data flow complex portions of the algorithm efficiently
 - Enables mapping of Multi-Scale, Multi-Physics Apps, post-Foundation AI and Complex workflows
- Standards-based parallelism enables productive portability
- Non-accelerated, fully-accelerated and mixed controlflow-dataflow complex applications can be run on Grace-Hopper
- Mapping these complex workflows to Distributed Heterogeneous Compute platforms require Omniverse like Digital Twin Frameworks



NVIDIA SUPERCHIPS SAVE ENERGY

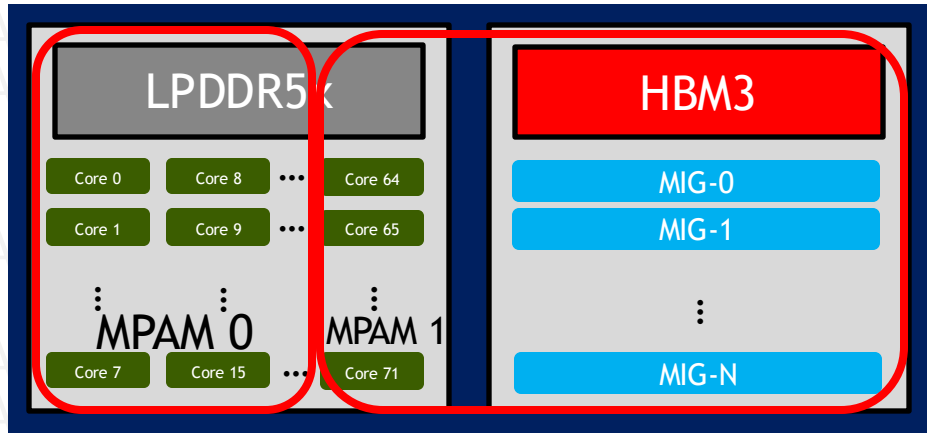
Low Power Data Motion and Computation



CO-SCHEDULING

Alternative to CPU-only partition

One Job exclusive to the node



Job A - 64 Grace CPU MPAM
Job B - 8 Grace Cores MPAM + Hopper GPU

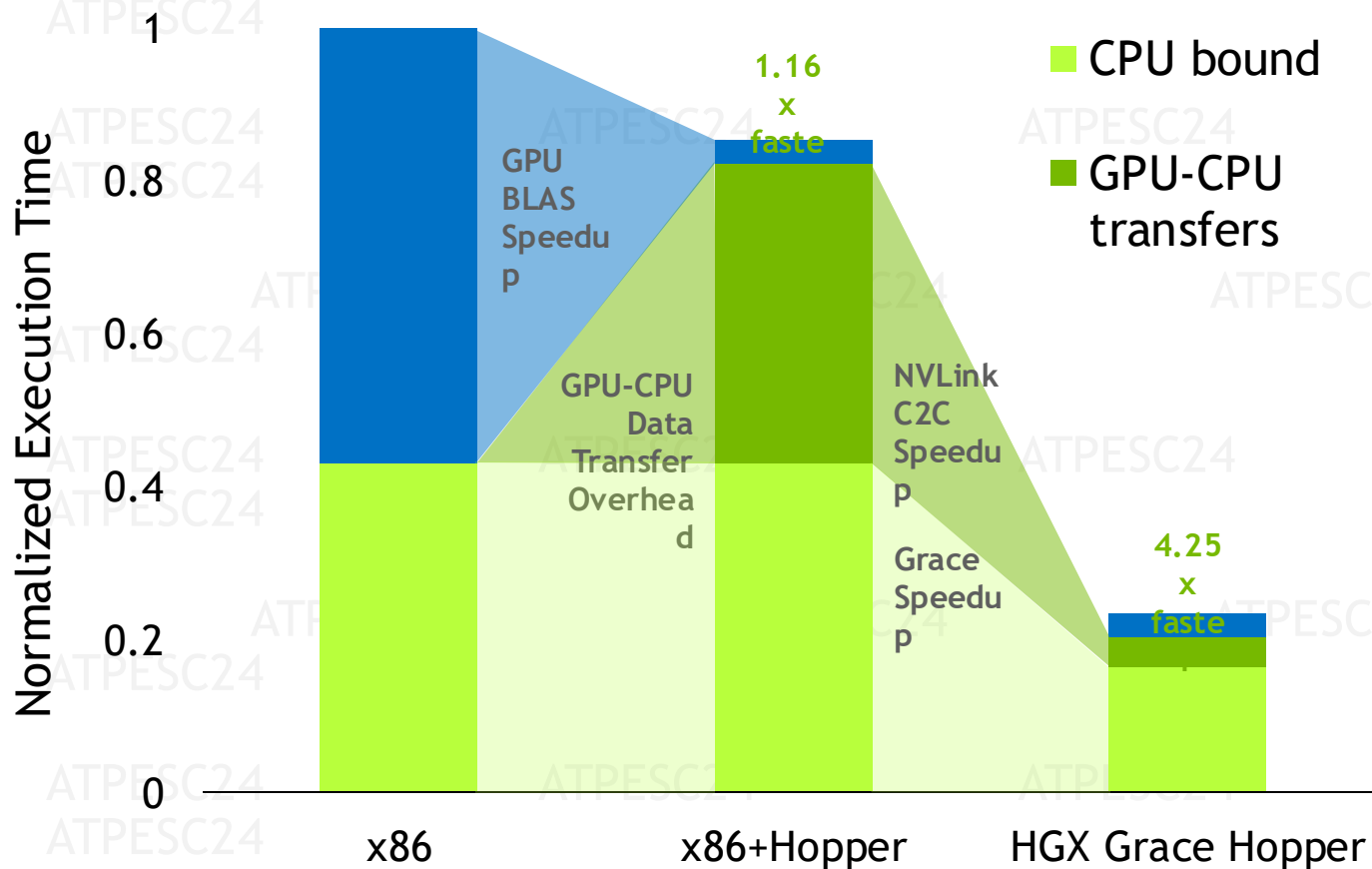


Job A - 8 Grace Cores MPAM + Hopper MIG
Job B - 8 Grace Cores MPAM + Hopper GPU
Etc.

GRACE+HOPPER MAKES ACCELERATION MORE ACCESSIBLE

Delivers Superior Performance and Efficiency for HPC

ABINIT



White Paper - Grace Hopper Superchip Architecture

PARALLEL EXPRESSION

- Moving from Prescriptive to descriptive parallel expression
 - ISO-Language Parallelism
- Parallel Expression as a Computational Graph
- Decouple Scheduling from the expression of algorithm (aka Halide/XLA)
- OS/Runtime provide HW capabilities as hints to the scheduler
- ▶ Lowering through MLIR
- ▶ Scheduler/Runtime tools like XLA

CONCLUSIONS

Grand Challenge applications are becoming complex, Heterogeneous, data driven and ML/DL aided

Silicon Process Technology alone loosing steam

Architectural innovations, co-Design, ML/DL aiding the continuation of “Moore’s Observation”
Grace-Hopper architecture is a step in that direction

Accurate definition of performance is critical

Efficiently mapping Heterogeneous applications to heterogeneous HW is becoming complex
STD PAR, MLIR, XLA

Power and Energy are becoming the hard barriers to performance

GRACE, HOPPER WHITE PAPERS

- ▶ Hopper:
 - ▶ <https://resources.nvidia.com/en-us-tensor-core>
- ▶ Grace:
 - ▶ <https://resources.nvidia.com/en-us-grace-cpu/nvidia-grace-cpu-superchip#page=1>
- ▶ Grace-Hopper:
 - ▶ <https://resources.nvidia.com/en-us-grace-cpu/nvidia-grace-hopper>
- ▶ DGH-GH200 White paper:
 - ▶ <https://resources.nvidia.com/en-us-dgx-gh200/technical-white-paper>

GTC 2023 Sessions to Watch

For more information on these topics

GTC23 Spring

- [Asynchronous Acceleration in Standard C++ \[S51755\]](#)
- [A Deep Dive into the Latest HPC Software \[S51074\]](#)
- [Accelerating HPC applications with ISO C++ on Grace Hopper \[S51054\]](#)
- [How to Write a CUDA Program \[S51210\]](#)
- [cuNumeric and Legate: How to Create a Distributed GPU Accelerated Library \[S51789\]](#)
- [Connect with the Experts: C++ Standard Parallelism and C++ Core Compute Libraries \[CWES52064\]](#)

NVIDIA Developer Blogs

- [Developing Accelerated Code with Standard Language Parallelism](#)
- [Accelerating Standard C++ with GPUs](#)
- [Accelerating Fortran DO CONCURRENT](#)
- [Bringing Tensor Cores to Standard Fortran](#)
- [Accelerating Python on GPUs with NVC++ and Cython](#)

Legate and cuNumeric Resources

- <https://github.com/nv-legate>
- [Accelerating Python Applications with cuNumeric and Legate](#)

Open-source codes

- LULESH - <https://github.com/LLNL/LULESH>
- STLBM - <https://gitlab.com/unigehpfs/stlbn>
- MiniWeather - <https://github.com/mrnorman/miniWeather/>
- POT3D - <https://github.com/predsci/POT3D>
- StdExec - <https://github.com/nvidia/stdexec>

C++ algorithms and execution policy reference

- <https://en.cppreference.com/w/cpp/algorithm>

NVIDIA HPC Compilers Forum

- <https://forums.developer.nvidia.com/c/accelerated-computing/hpc-compilers>



BACKUP

ISO CAN + UNIFIED SDK + LIBRARIES

outline

- Future HPC and AI Applications
- Post-Moore HW Landscape
- Nvidia Grace-Hopper
- Future SW stack
- Conclusions
- White Papers & References

GRACE CPU SUPERCHIP

The Full Power of the Grace



Specifications	Grace SuperChip
Architecture	Armv9, SVE2 with 4x 128b pipeline/core
Cores / Speed	144 cores up to 3.2GHz
Memory	LPDDR5x soldered down, 1TB/s BW Up to 1TB per superchip
Cache	L1: 64KB i-cache + 64KB d-cache per core L2: 1MB per core L3: 234MB per superchip
Power	500W including LPDDR5x memory
Interfaces	Up to 8x PCIe Gen5 x16 HS interface
Specrate2017_int_base*	740
Process Node	TSMC 4N
Availability	Q4 2023

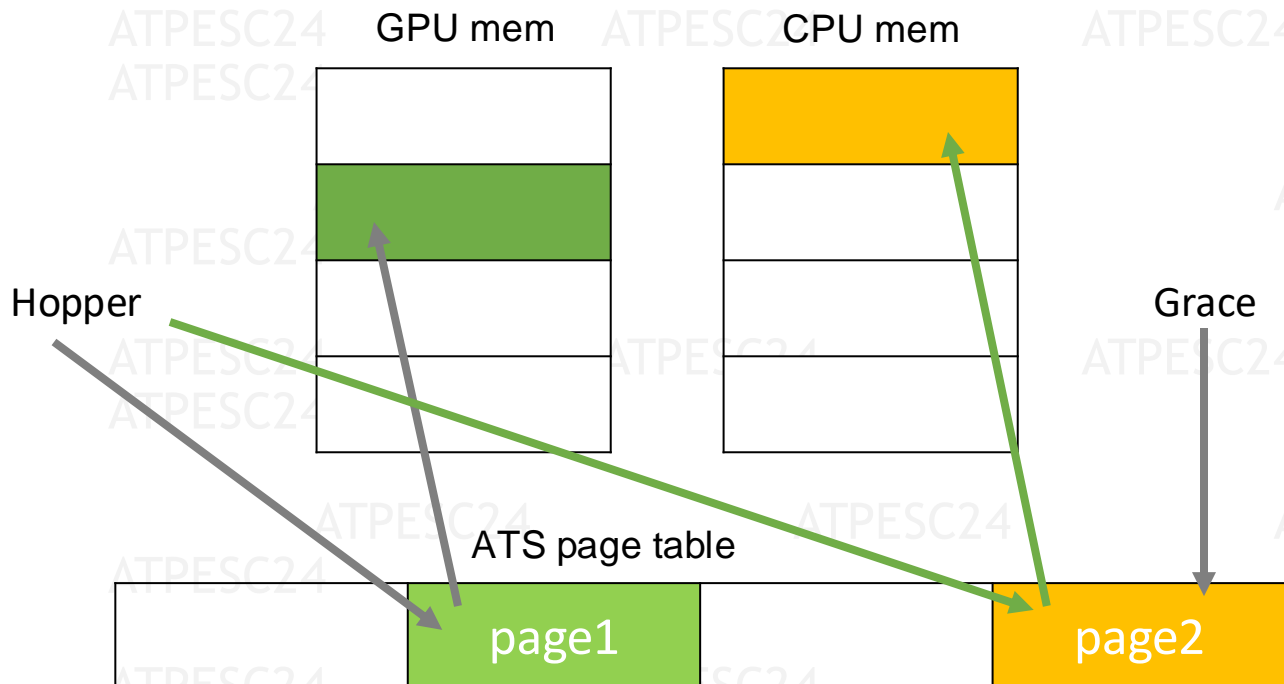
GH NODE ARCH COMPARISON

- ▶ Summit vs DGX-A100 vs GH : ratio of CPU perf to GPU perf in a Node
- ▶ GH is the first Heterogeneous Compute Substrate
 - ▶ Reticle size CPU + Reticle Size GPU tightly coupled in a unified, coherent address space
- ▶ CPU-only codes, fully GPU acceleratable codes, Mixed controlflow complex/dataflow complex codes
 - ▶ With MPAM on CPU and MIG on GPU all codes can run on GH, GH-Next efficiently
 - ▶ Optimization of Applications, Runtime, Scheduler/resource manager are necessary

	Node Arch		
	Summit P9+V100	DGX Rome+4*A100	Grace-Hopper G+H
CPU Cores	22	64	72
CPU FLOPS (TFLOPS)	1.08	2.30	3.46
GPU FLOPS (TFLOPS)	21.00	78.00	51.00
CPU Mem BW (GB/s)	170.00	190.00	500.00
GPU Mem BW (GB/s)	2700.00	8156.00	3686.40
CPU <=> GPU BW (GB/s)	150.00	256.00	900.00
GPU/CPU FLOPS	19.43	33.85	14.76
GPU/CPU Mem BW	15.88	42.93	7.37
C<=>G link BW to FLOPS ratio	7.14	3.28	17.65

GRACE HOPPER

Unified Programming Model



- Address Translation Service (ATS) enables all CPUs and GPUs in the node to share a single page table
- System-allocated memory is accessible by all CPU and GPU threads
- Runtime system backs system-allocated memory with physical memory on first touch, either on LPDDR5 X or HBM3, depending on whether a CPU or a GPU thread accesses it first .

Grace-Hopper Memory Model

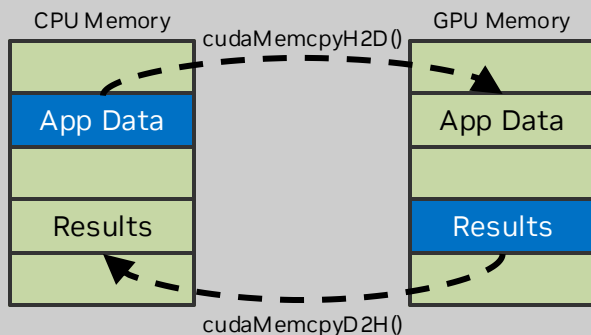
Full CUDA support with additional Grace memory extensions

Explicit Copy

Application explicitly moves data between CPU & GPU as needed

PCIe: ~60 GB/s PCIe transfers (H2D/D2H)

Grace: Faster transfers; up to 450 GB/s C2C transfers

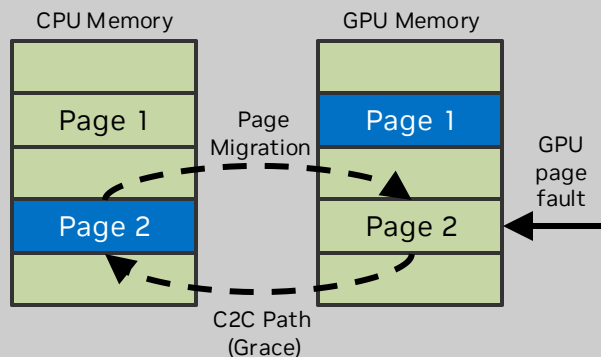


Managed Memory

CPU and GPU can access memory on-demand and data migrated locally for higher BW access

PCIe: Requires migration to GPU

Grace: Migrations not required and faster migrations when they happen

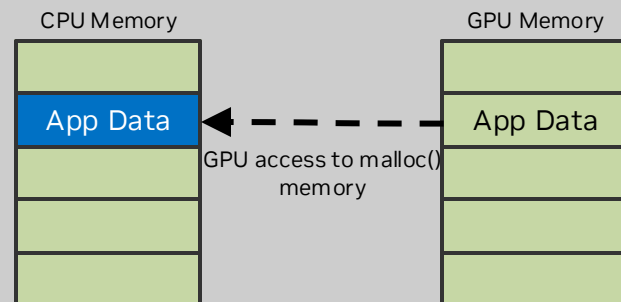


System Allocated


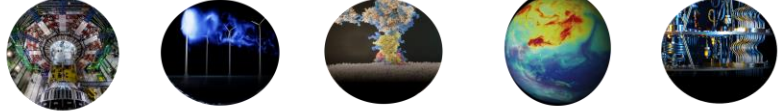
GPU can access memory allocated from `malloc()`, `mmap()`, etc.

PCIe: Access possible with explicit call to `cudaHostRegister()` at PCIe speeds

Grace: `cudaHostRegister()` not needed; access at NVLink C2C speeds



TRANSITION TO POST EXASCALE ERA

FEATURE	<p style="text-align: center;">EXPERIMENTS SIMULATION Viz</p>  <p style="text-align: center;">TERA THROUGH EXASCALE</p>	<p style="text-align: center;">EDGE HPC + AI SIMULATION DIGITAL TWIN QUANTUM COMPUTING</p>  <p style="text-align: center;">POST EXASCALE</p>
USAGE	BATCH & MOSTLY LOCAL TO A SITE	INTERACTIVE & DISTRIBUTED WITH MULTIPLE SITES
WORKLOAD	SINGLE SIMULATION/ENSEMBLE	WORKFLOW COMPRISED OF SIMULATION ENSEMBLES, AI TRAINING AND INFERENCE, LIVE DATA ANALYTICS
EXPERIMENTS	OFFLINE DATA ANALYSIS FOR EXPERIMENTS	MIX OF REAL-TIME ANALYSIS TIGHTLY COUPLED WITH OFFLINE
DIGITAL TWINS	IN-SITU VISUALIZATION OFFLINE	INTERACTIVE_VISUAL MODEL COUPLED WITH PHYSICAL ASSET
ROBOTICS	OFFLINE	INTEGRATED WITH VIRTUAL MODELS
PROGRAMMING MODELS	FORTRAN, C++, MPI, OPENMP	STANDARD PARALLELISM SUPPORT IN FORTRAN, C++, MPI, OPENMP, OPENACC, PYTHON, JULIA, PYTORCH, TENSORFLOW
SYSTEM CONFIGURATION	MONOLITHIC	MODULAR
DISTRIBUTED SITES	GRID	CLOUD

Generative AI - The Revolutionary New Tool for Scientific Discovery

