

Can Climate Science Win the Hardware Lottery?

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HPC Circa 2007

*Exascale report from 2007 Town Halls
Entirely focused on modeling and
simulation*

Simulation $\neq$ Scientific Computing $\neq$ HPC

Modeling and Simulation at the Exascale for Energy and the Environment

Co-Chairs:

Horst Simon

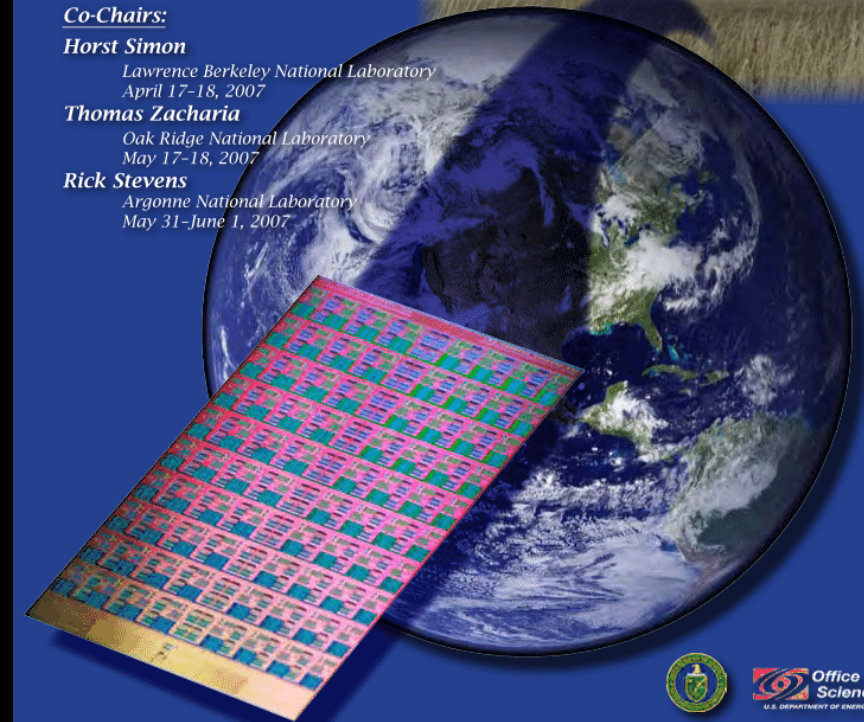
Lawrence Berkeley National Laboratory
April 17-18, 2007

Thomas Zacharia

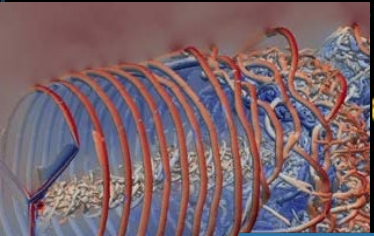
Oak Ridge National Laboratory
May 17-18, 2007

Rick Stevens

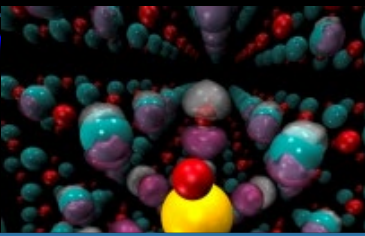
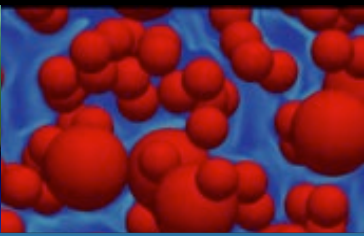
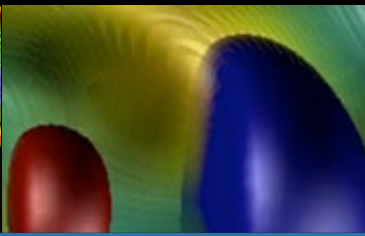
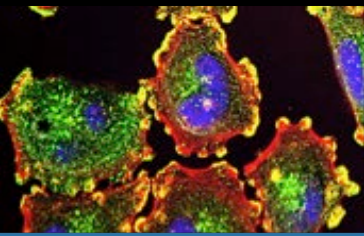
Argonne National Laboratory
May 31-June 1, 2007



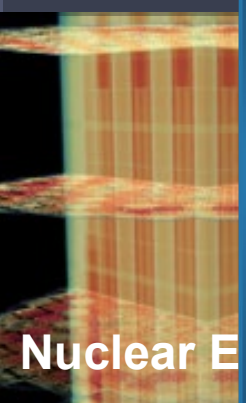
Continue to Rethink Applications



Wind Energy



CS



Nuclear Energy

- 24 projects with about 10 people per team
- Rely heavily on hardware features and software teams
- Several new to HPC, all with new capabilities
- We should have another 2 dozen in 10 years!!



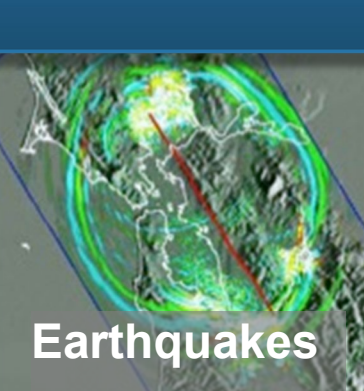
Power Grid



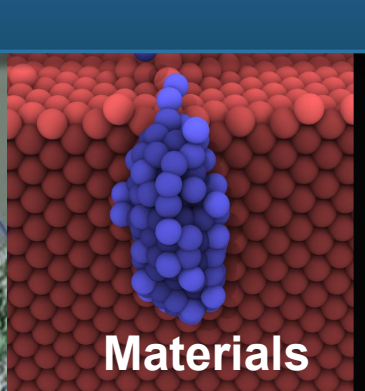
Manufacturing



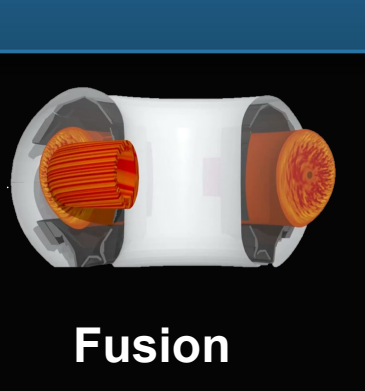
Climate



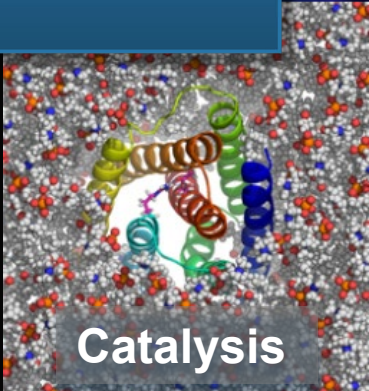
Earthquakes



Materials



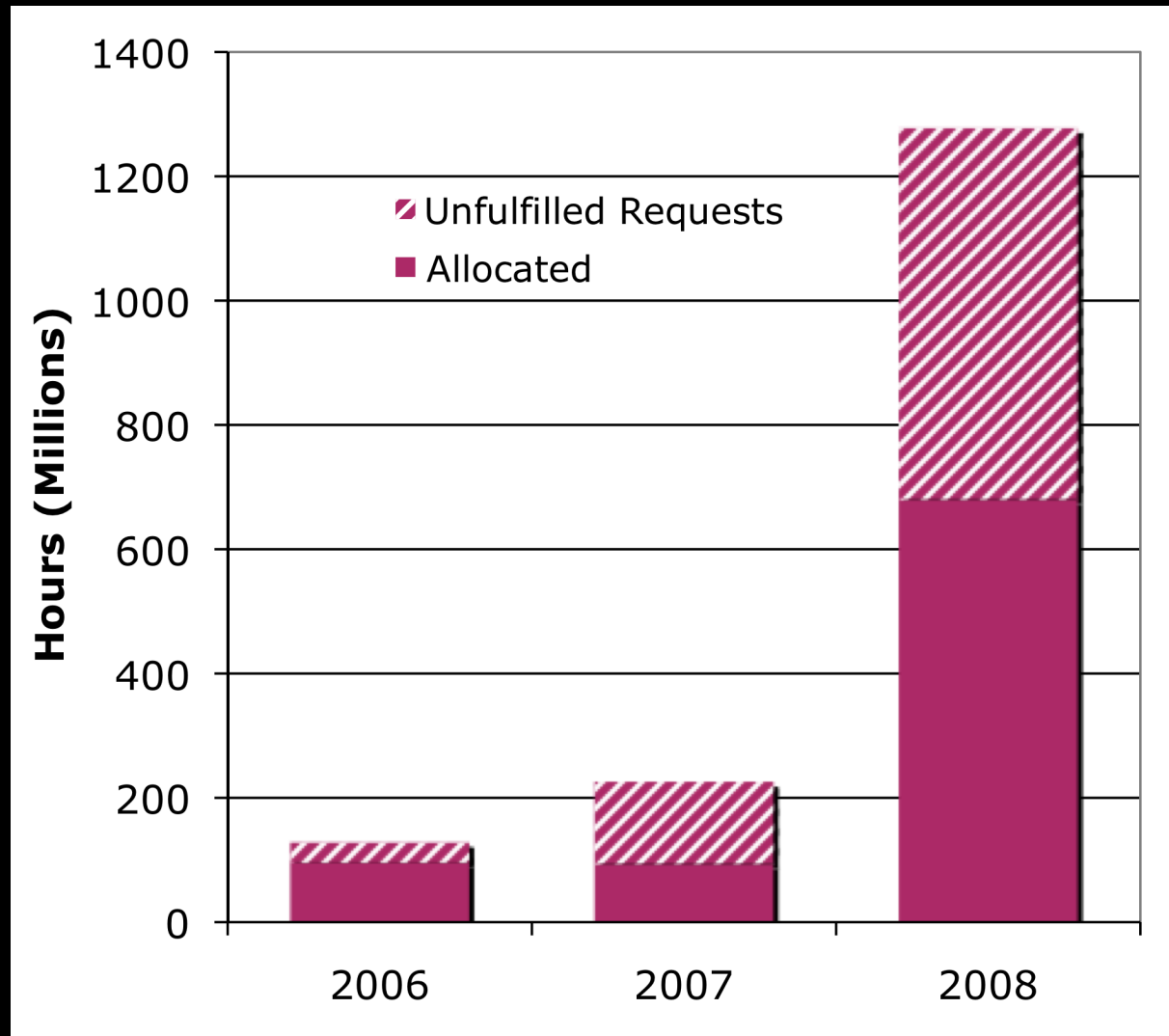
Fusion



Catalysis

DOE Demand for Computing is Growing (circa 2008)

Compute Hours Requested vs Allocated



- *Each year DOE users requests 2x more hours than allocated*
- *This 2x is artificially constrained by perceived availability*
- *Unfulfilled allocation = hundreds of millions of hours in 2008*
- *When allocation limits are removed, scaling and science increase*

New demands for HPC in Science



Simulation

From atoms
to the
universe



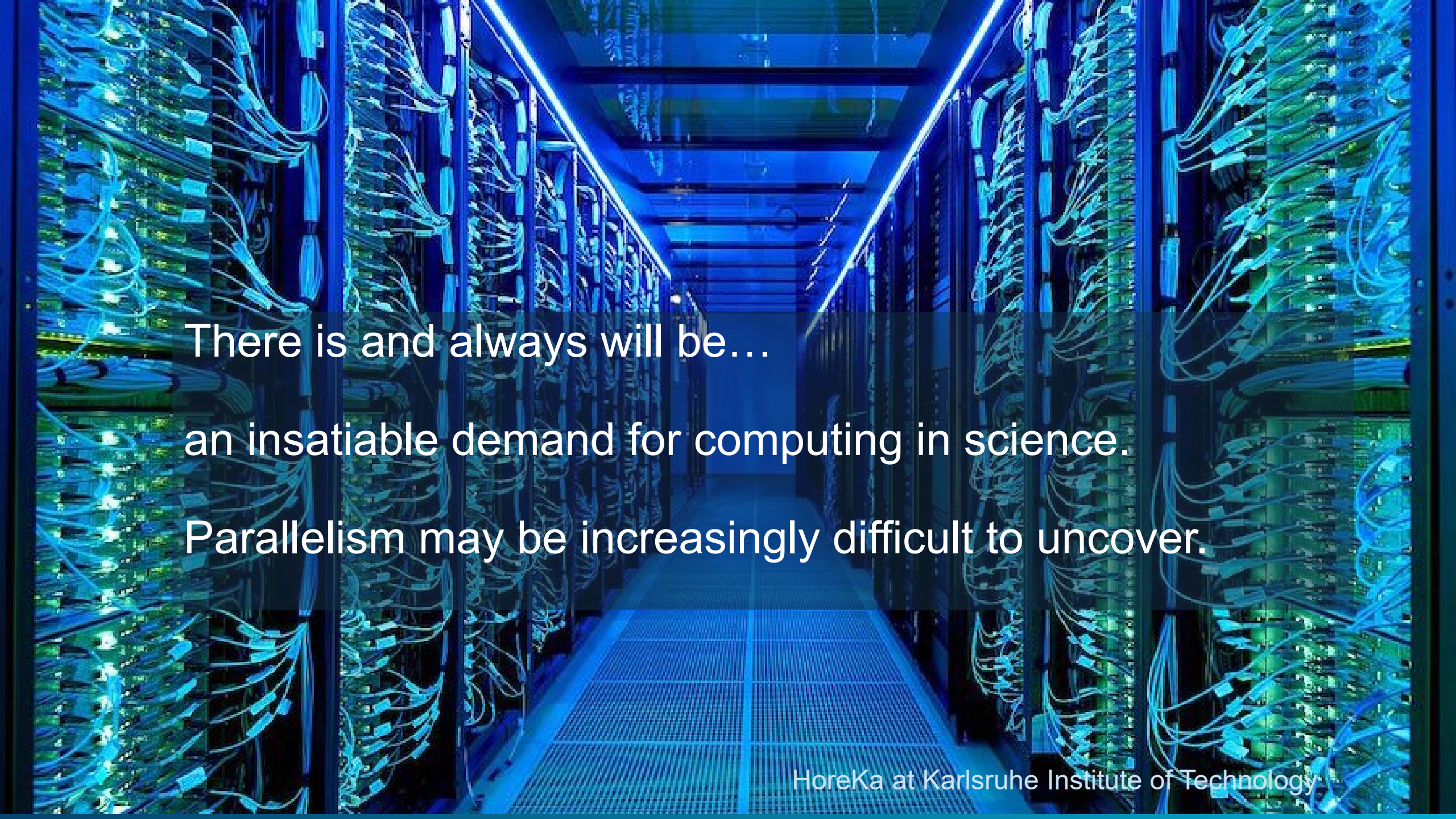
Data

Images, text,
to genomes



Learning

Interpret,
generate, and
automate



There is and always will be...
an insatiable demand for computing in science.
Parallelism may be increasingly difficult to uncover.

Hardware Lottery

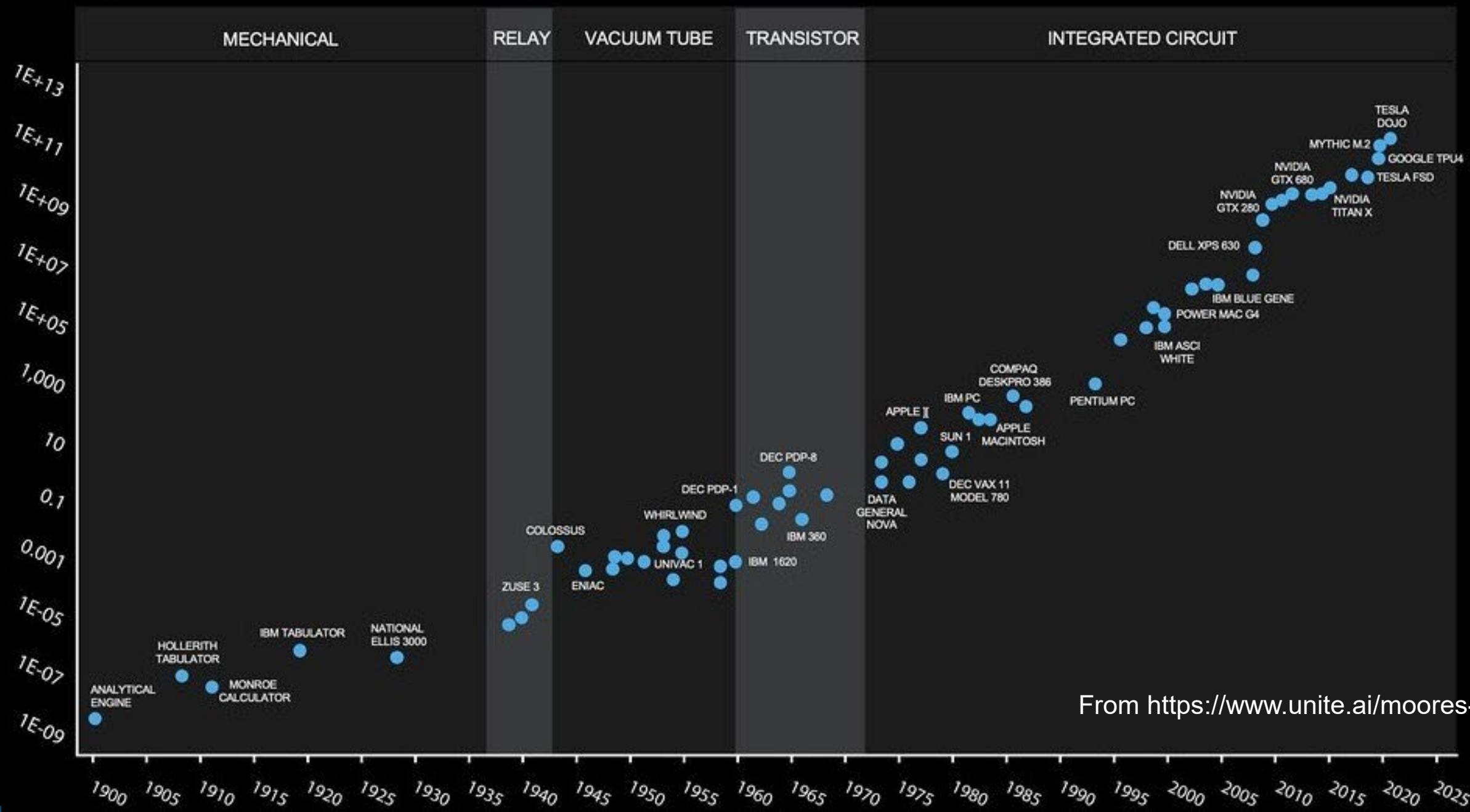
When a research idea wins because it is suited to the available software and hardware and

not because the idea is superior to alternative research directions

Sara Hooker

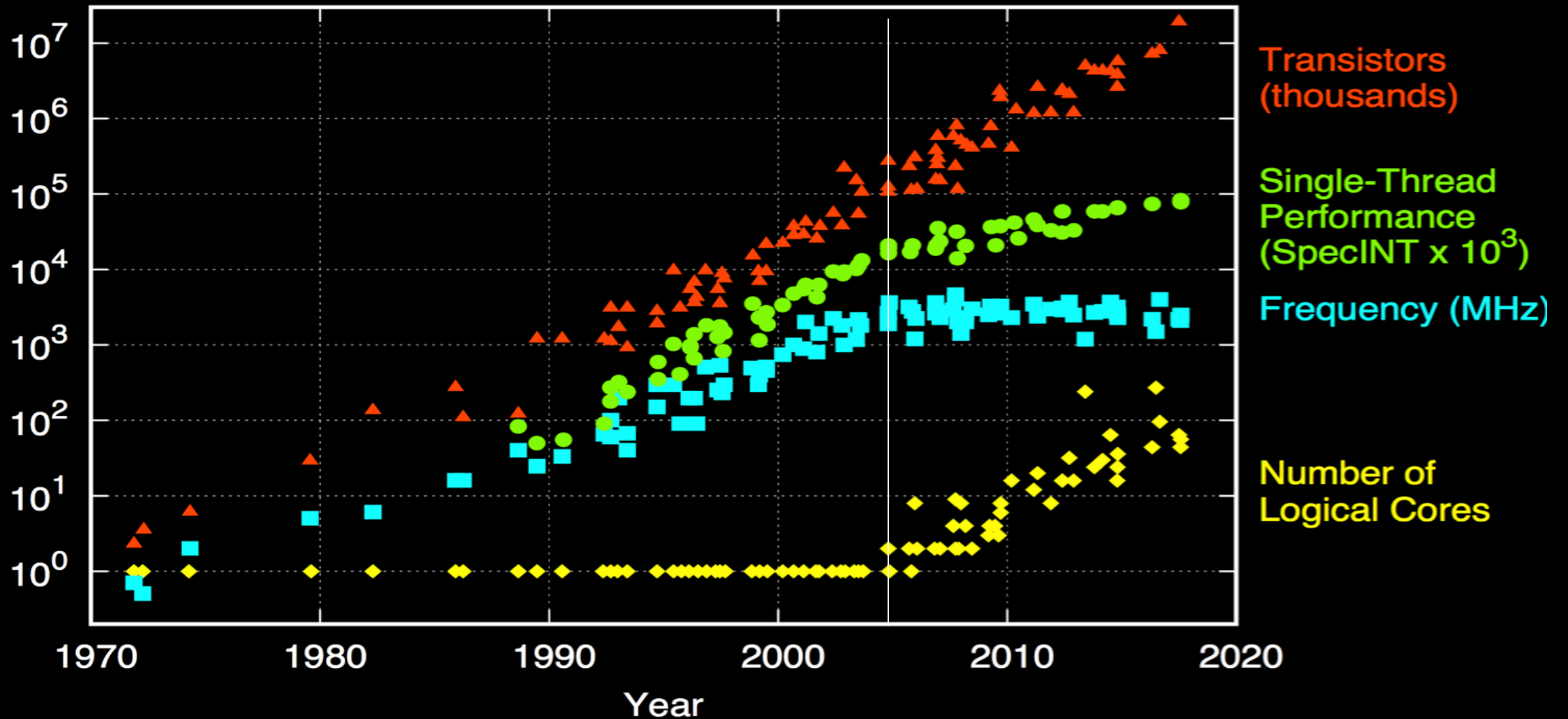


122 YEARS OF MOORE'S LAW



From <https://www.unite.ai/moores-law/>

Dennard Scaling is Long Dead; Moore's is Dieing



Exascale Architecture Plans (2008)

Petascale X 10x more energy X 100x more Performance per Joule = Exascale

Faster clocks
+ wider SIMD

Accelerators
(GPUs)

100x
more
cores

Exascale Era Architectures (US DOE Office of Science)

US DOE Office of Science Systems



Exascale
HPE AMD+AMD



Exascale
HPE Intel+Intel

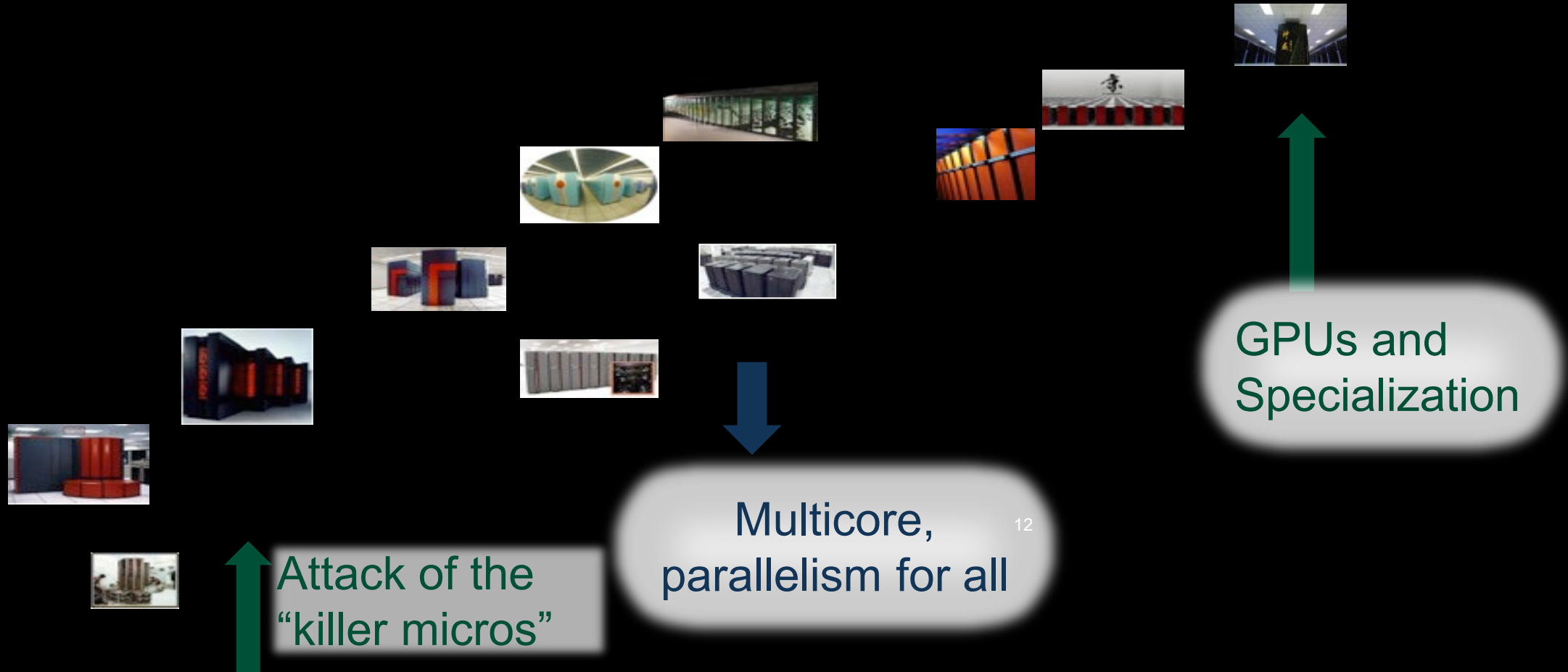


Pre-exascale
HPE AMD+NVIDIA

1 Architecture (3 GPUs), 1 Integrator!

Performance Growth

Top500 and Gordon Bell



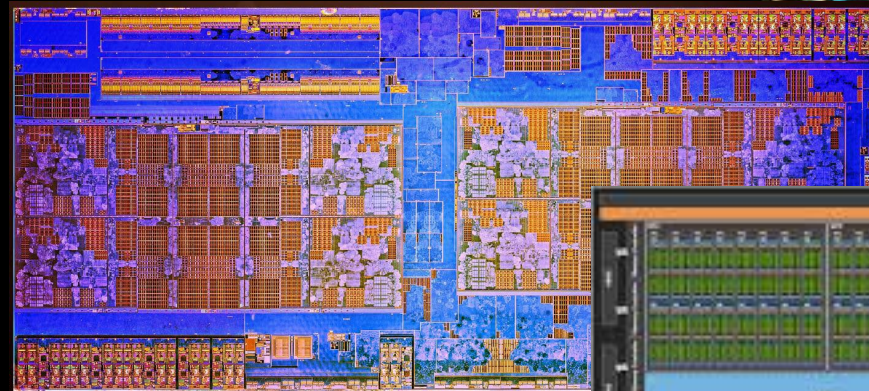
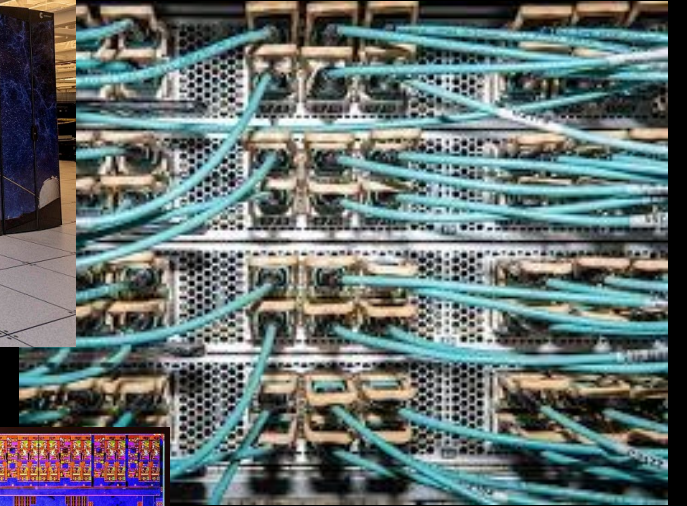
First-in-Class HPC Systems (Top500)

	First Teraflop	First Petaflop	First Exaflop
	ASCI Red	Roadrunner	Frontier
Year	1997	2008	2022
Best Tech (nm)	500	10x → 65	10x → 6
Power (MW)	0.9	2x → 2.4	10x → 21.1
Efficiency (GF/W)	0.001	400x → 0.4	100x → 52
Memory (PB)	0.001	40x → 0.04	200x → 9
FPU's (K)	9	100x → 464	1000x → 534,000
Floorspace (m ²)	150	4x → 557	1x → 678

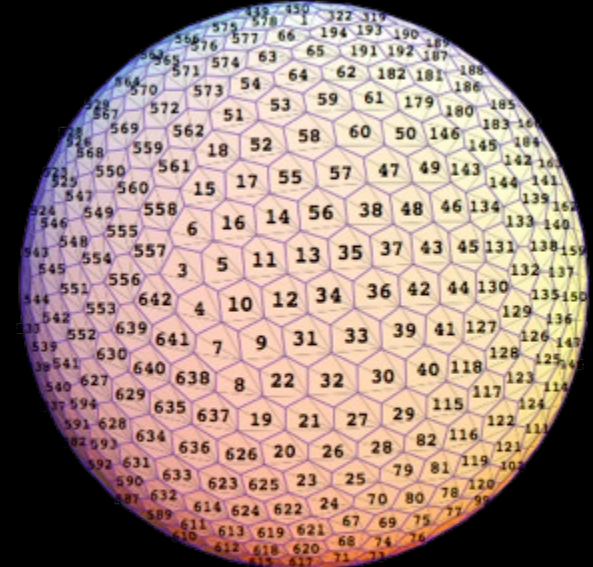
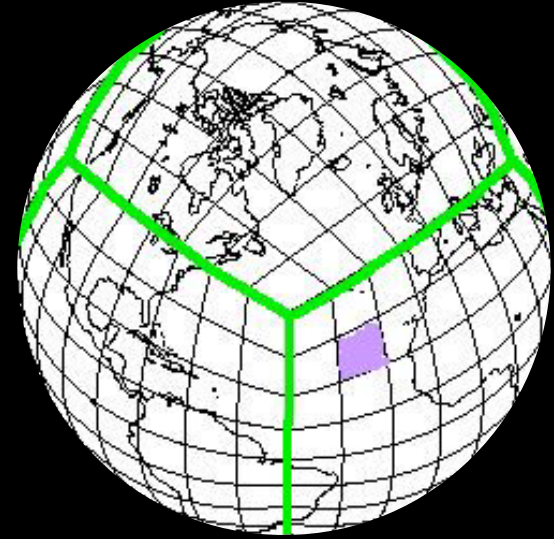
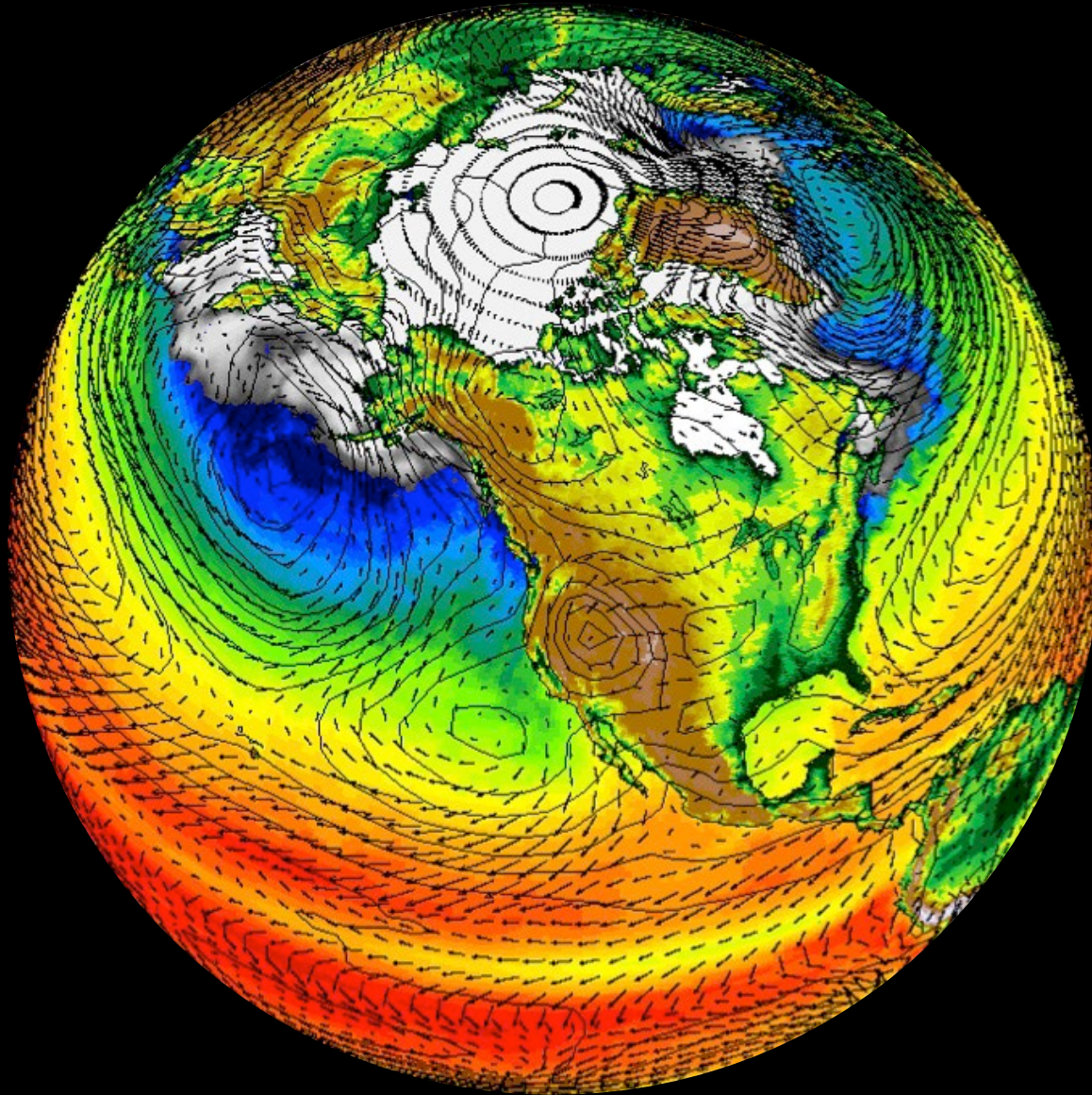
Kogge and Dally: Frontier vs the Exascale Report + Wikipedia for ASCI Red

Parallelism: Everything Everywhere All At Once

- Between Nodes
- Communication (overlap)
- Avoid synchronization
- Multicore
- GPUs/Accelerators
- SIMD
- Pipelineing



Players in the Climate Modeling Hardware Lottery

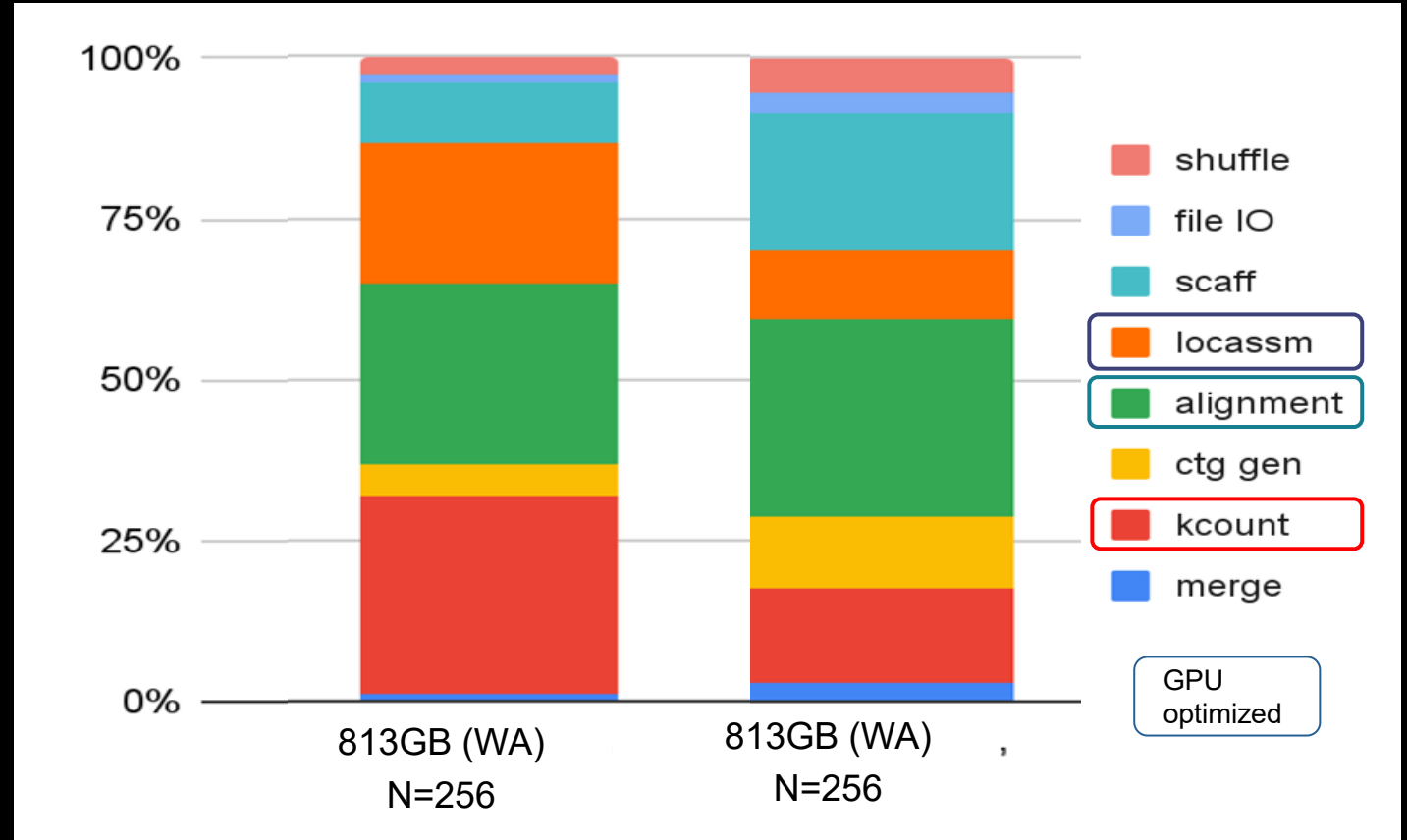


MetaGenome Assembly: Lottery Player?

From shared memory to HPC + GPUs

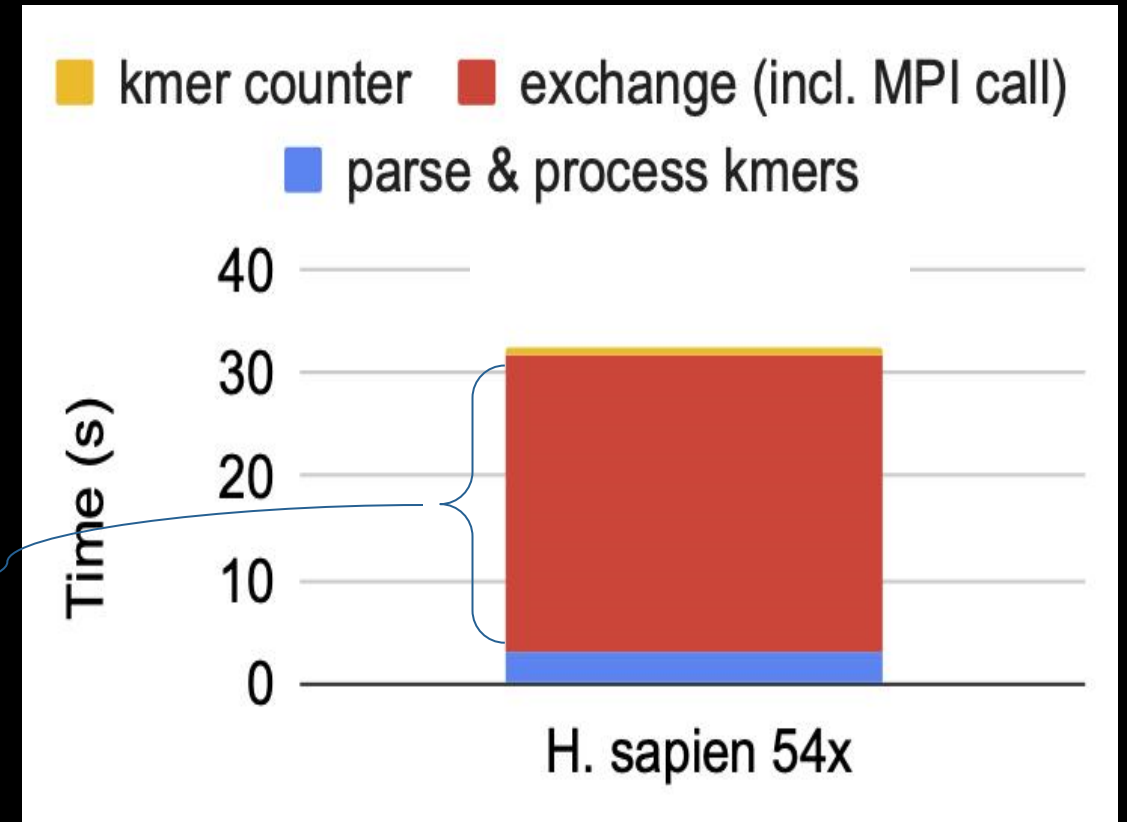
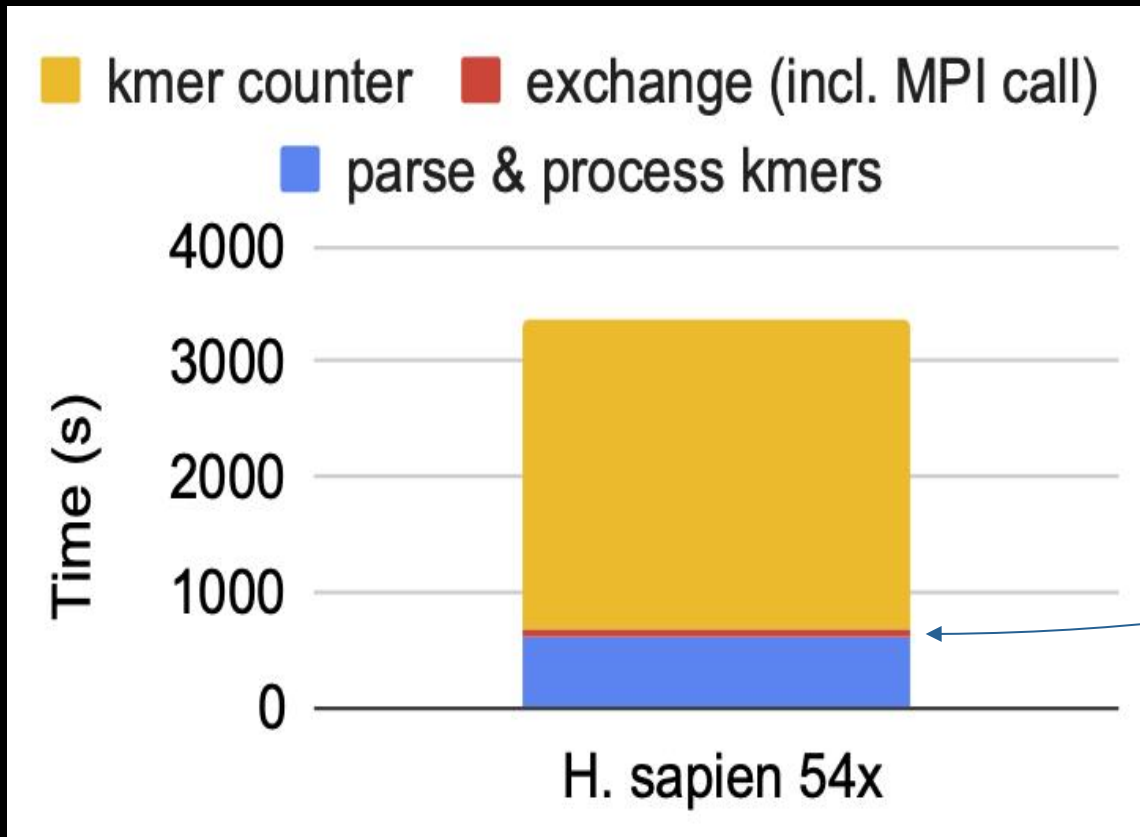
Repeat

- 1) Histogram k-mers
- 2) Walk k-mer graph
- 3) Align reads to contigs
- 4) Do 1-3 locally
- 5) Graph walk (scaffold)



Muaaz Awan, J. Deslippe, S. Hofmeyr, R. Egan, A. Buluç, L. Olier, K. Yelick, SC21 Best Paper Finalist

K-mer Counting: Finding Data Parallelism



- K-mer counter on Summit. (Note scales -- red k-mer exchange time is roughly equal.)
- Reduce CPU/GPU communication by parsing as well as processing on GPU

Post-Exascale Architecture Plans 2024 (Strawperson-v0)

Exascale X 2x more energy X 500x more Performance per Joule ??

GPUs

Influenced by “us” to make AI better (e.g., sparsity)?

Specialized
for AI

Specialized for
Simulation

Designed by “us”...?

Declining Benefits of Moore's Law Scaling

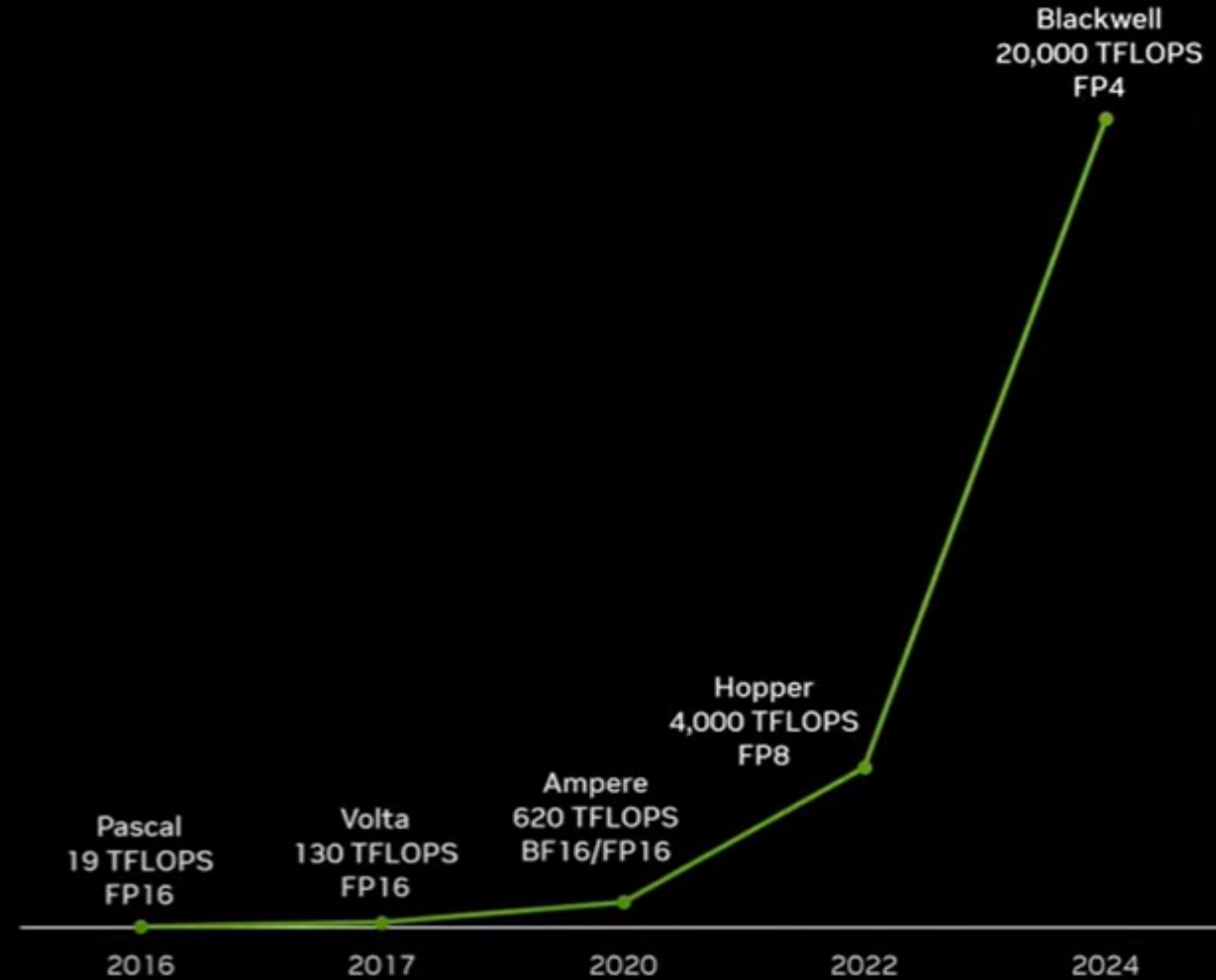
	2008	2012	2015	2022	10X
Gate Length (nm)	65	32	16	6	
Energy ⁻¹	1	1.8	2.8	4.5	4.5X

2nm fabs coming? But how much will it help?

Kogge and Dally: Frontier vs the Exascale Report: Why so long? and Are We Really There Yet?

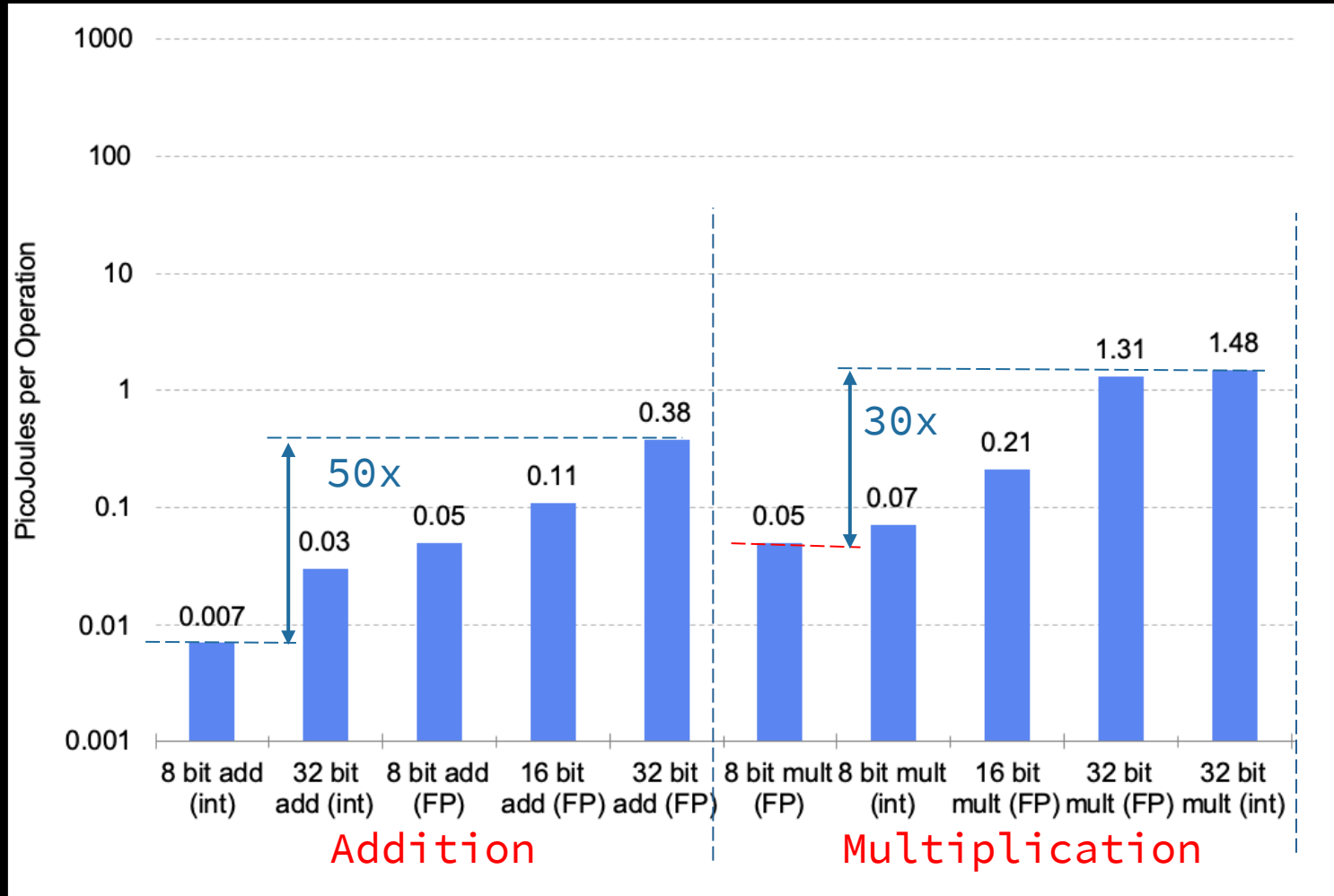
Another Exponential?

1000X AI Compute in 8 Years



Jensen Huang's Nvidia GTC Keynote

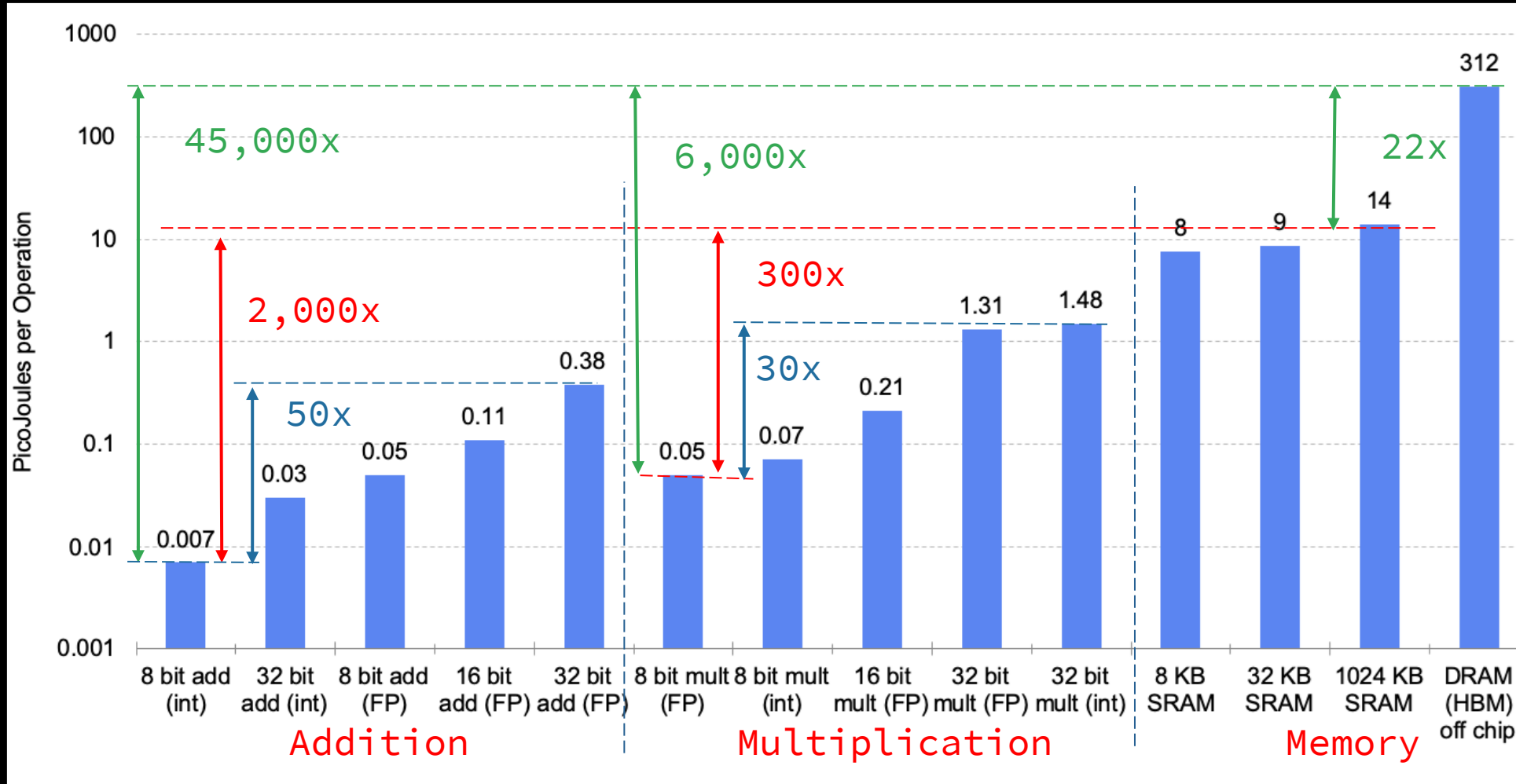
Narrow Data Types Save Energy



Smaller data
uses less energy
per op

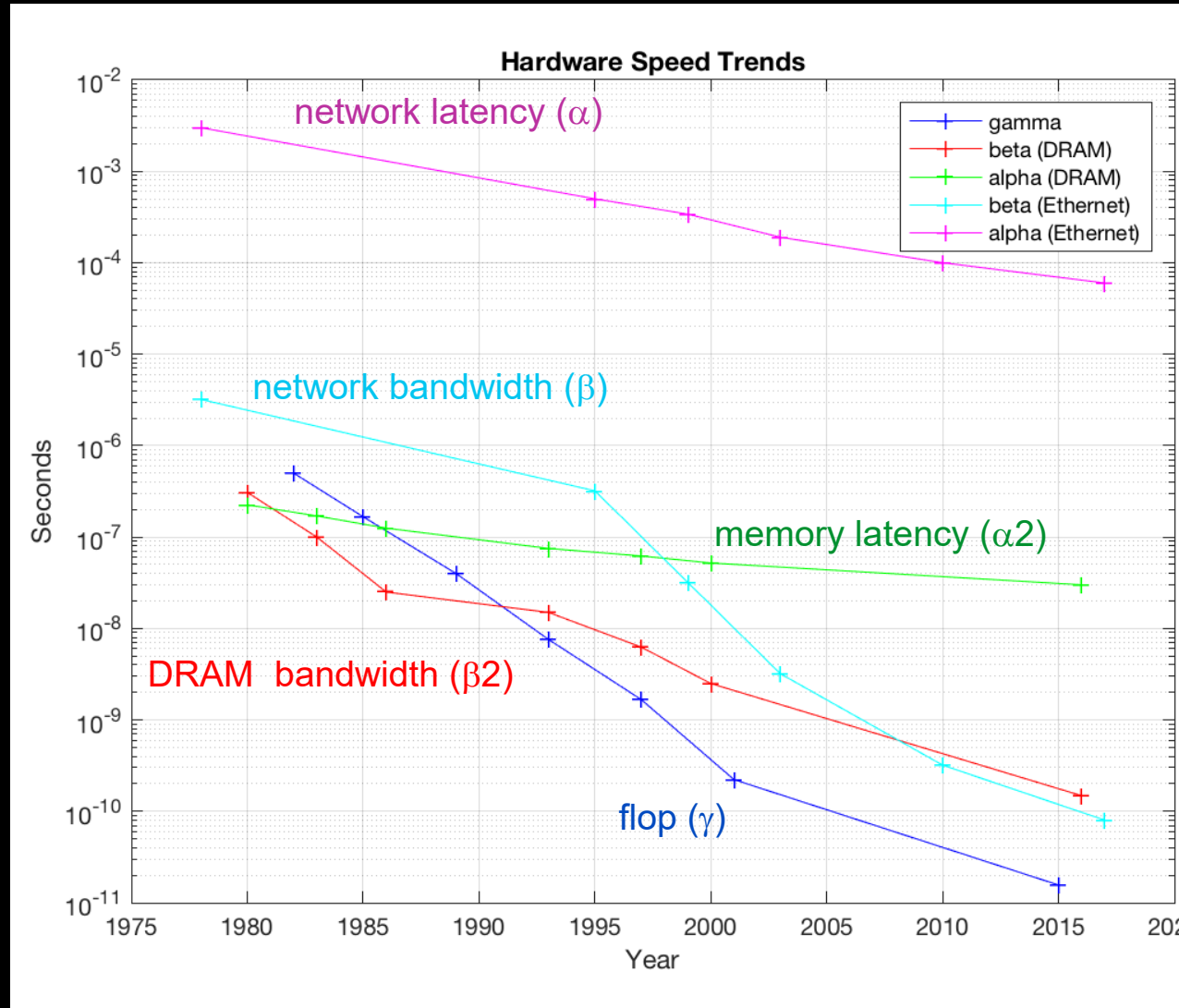
Graph from Dave Patterson

Narrow Data Types Save Energy



Smaller data also uses less on-chip memory capacity and off-chip memory bandwidth

Data Movement is the Most Expensive

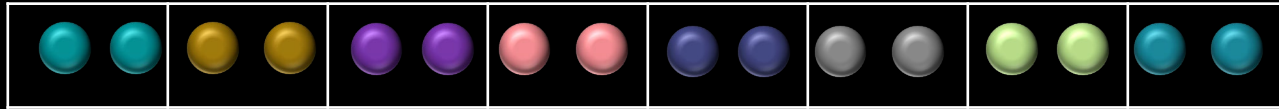


Data from Hennessy / Patterson, Graph from Demmel

Avoid Communication and Maximize Parallelism

Compute on all pairs of particles, DNA, tokens, or...

Obvious solution

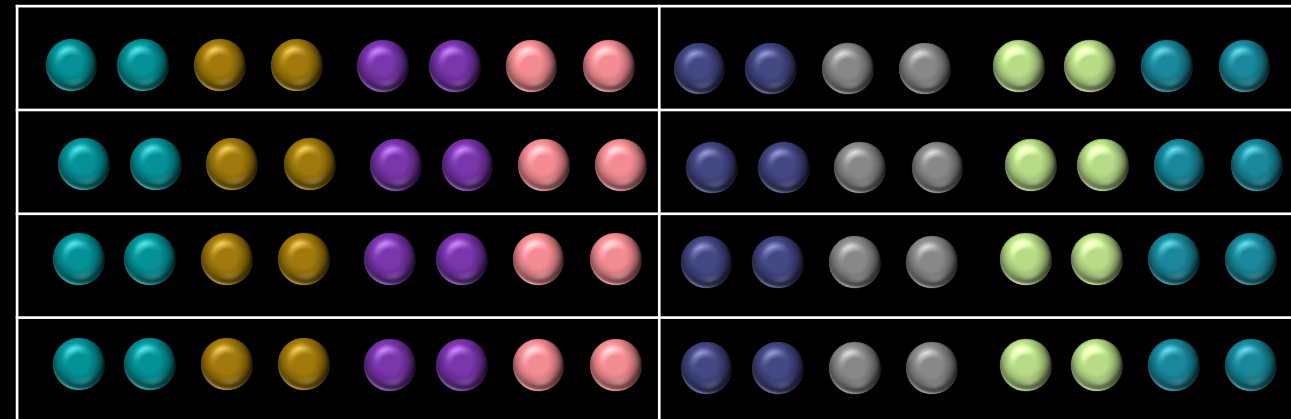


16 particles on 8 processors
Pass all particles around (p steps)

Decreases

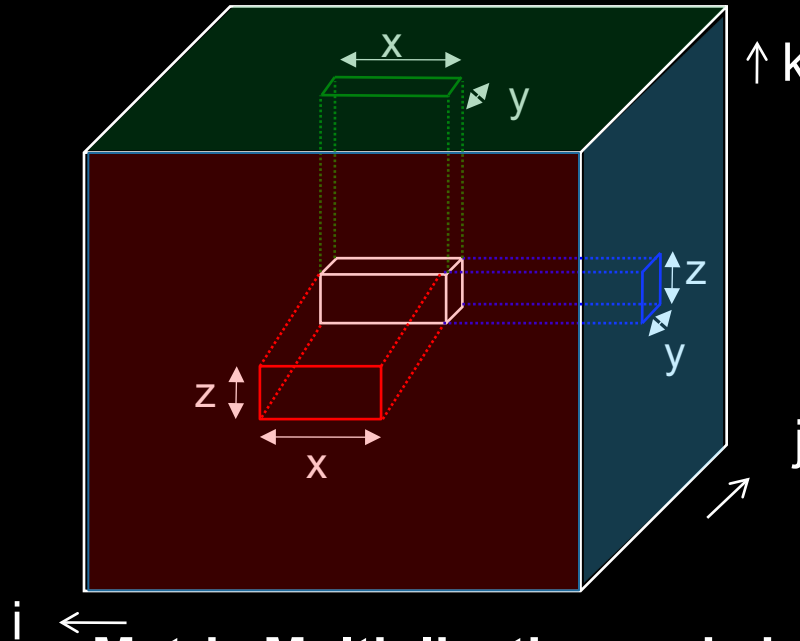
- #messages by factor c^2
- #volume sent by factor c

Better solution



$c = 4$ copies of particles
8 particles on each

Communication-Optimal Matrix Multiply



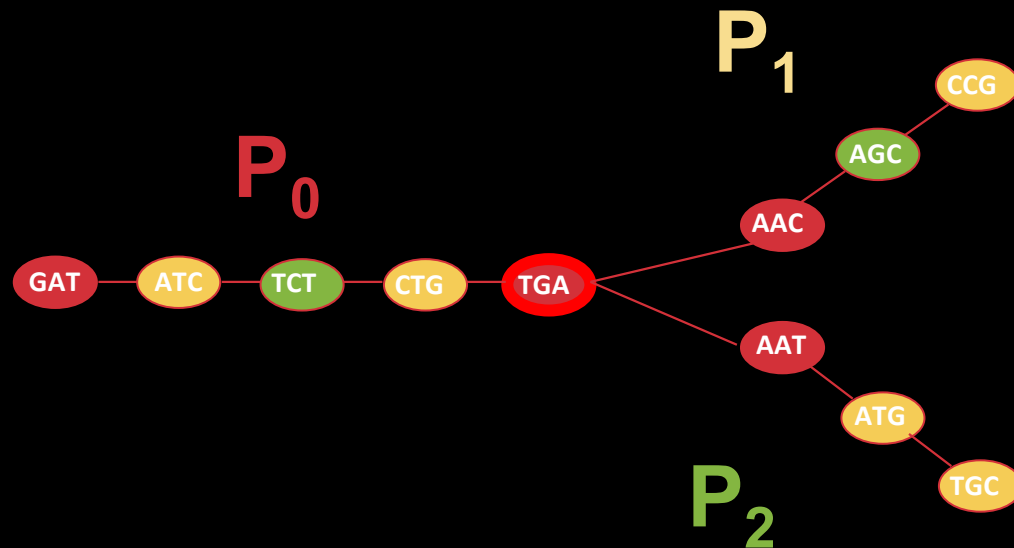
- 2D algorithm: never chop k dim
- 3D: Assume + is associative; chop k, which is $\rightarrow$ replication of C matrix

$i \leftarrow$ **Matrix Multiplication code has a 3D iteration space**
Each point in the space is a constant computation (*/+)

```
for i
  for j
    for k
      C[i,j] ... A[i,k] ... B[k,j] ...
```

Avoiding Communication in Graph Walk (DFS)

Caching for temporal locality:
depends on graph and layout



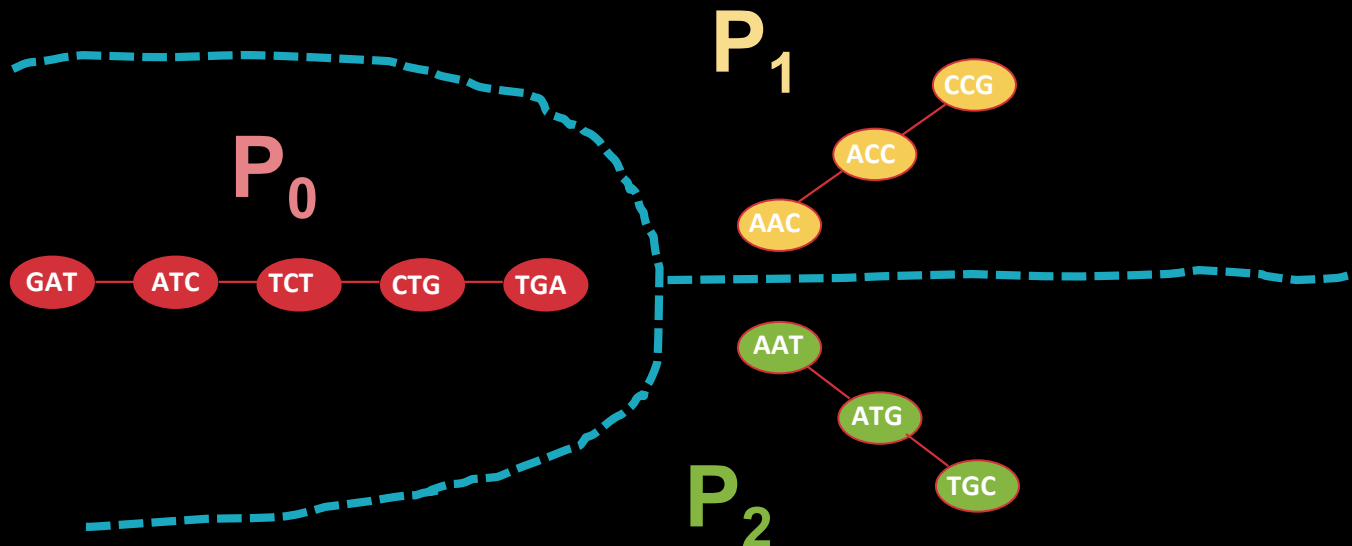
Up to 2.8x faster -- 76%
reduction of off-node
communication!

Avoiding Communication in Graph Walk (DFS)

It's all about the data

Caching for temporal locality:
depends on graph and layout

Layout for spatial locality: locality-
award hash function, i.e., an “oracle”



Up to 2.8x faster -- 76%
reduction of off-node
communication!

Jump on the AI Bandwagon

If anything will continue to improve in performance, it will be AI

Beat 'em

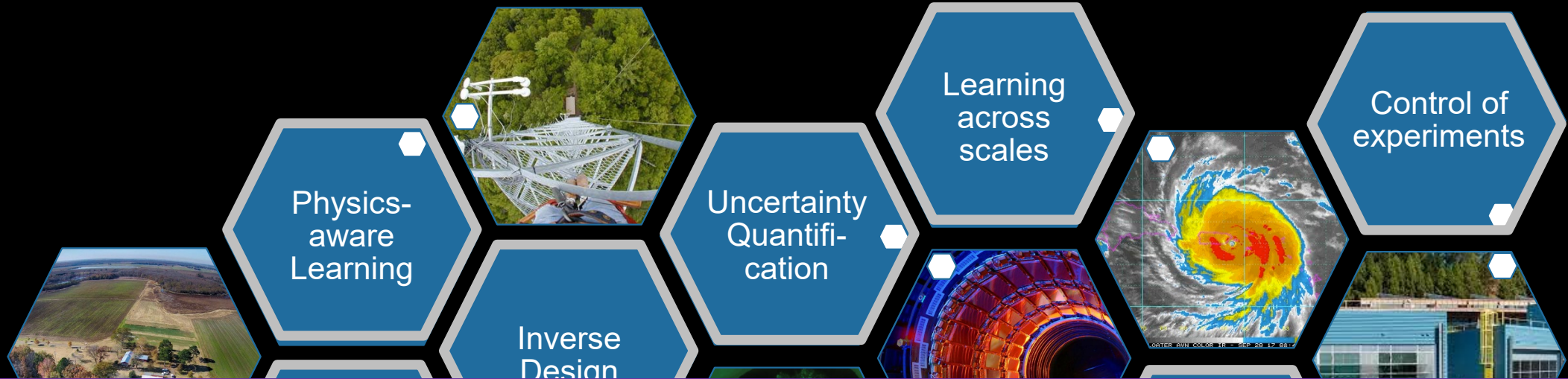
- Design science processors
- Or co-opt theirs

Join 'em

- Use AI methods for science



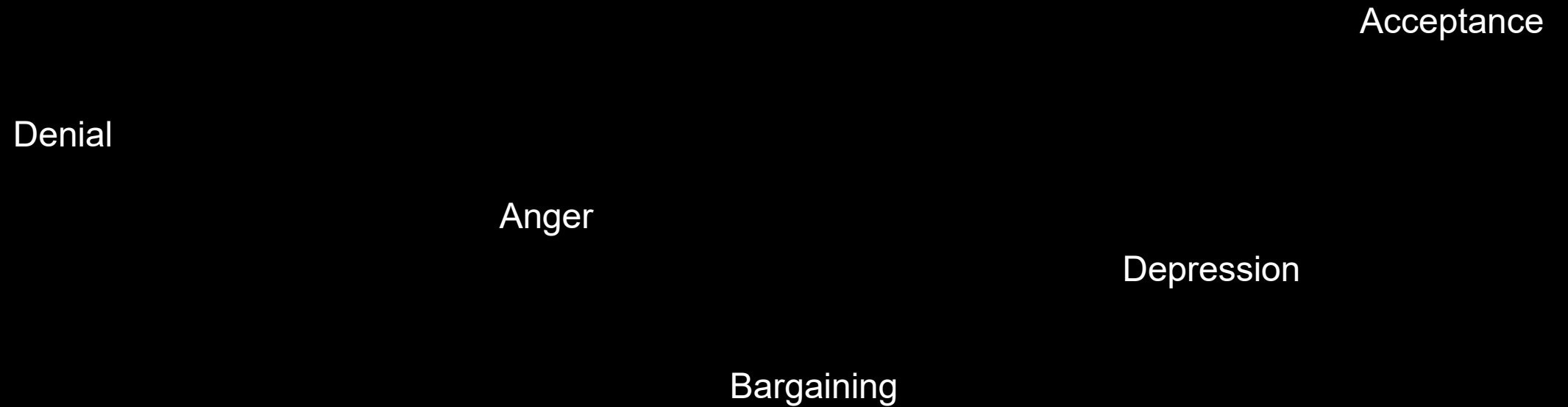
AI in Science



The Computational Science and Engineering community should have a leadership role in addressing safety, alignment, and explainability in AI for science and engineering

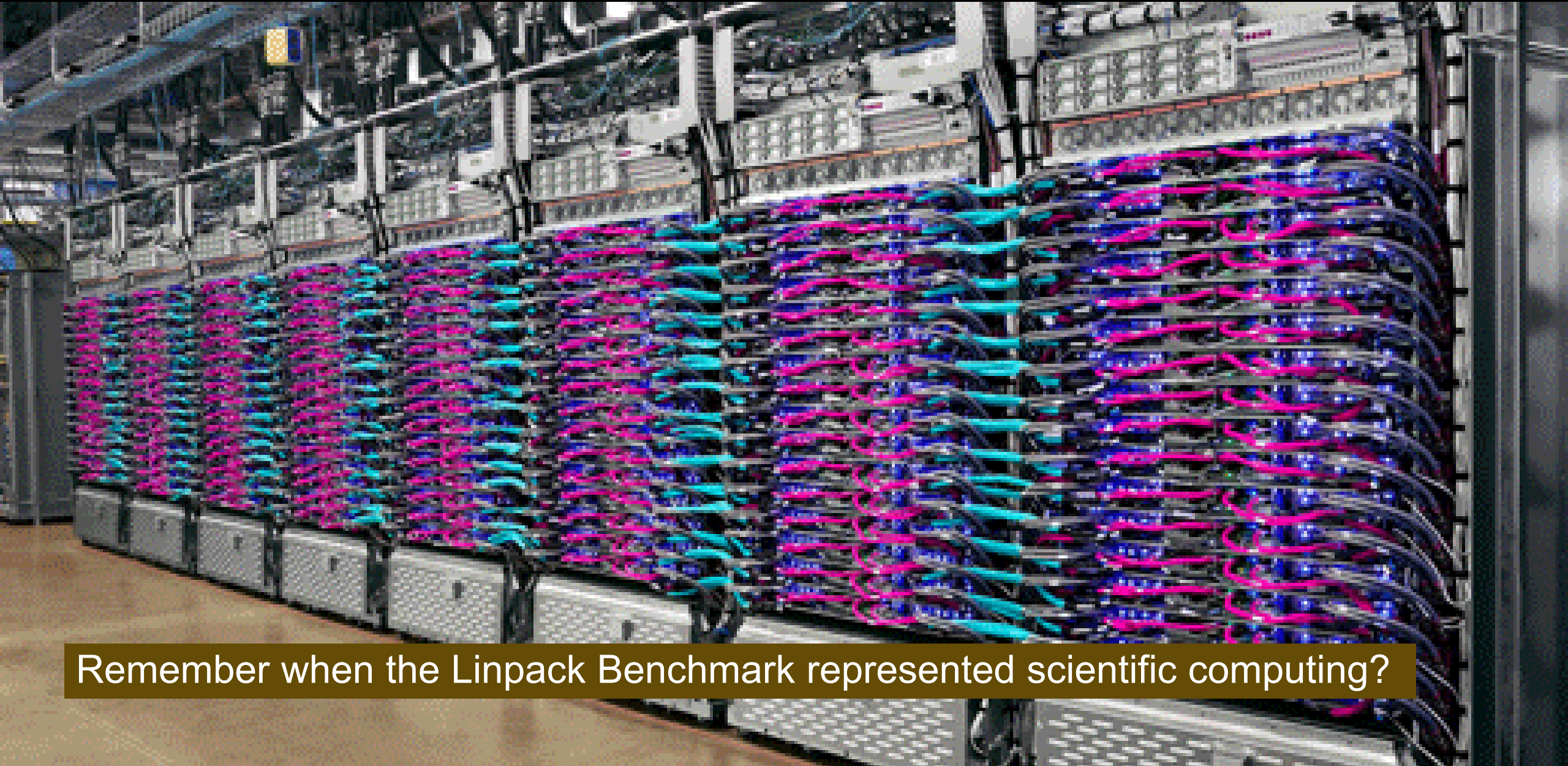


Five Stages of AI



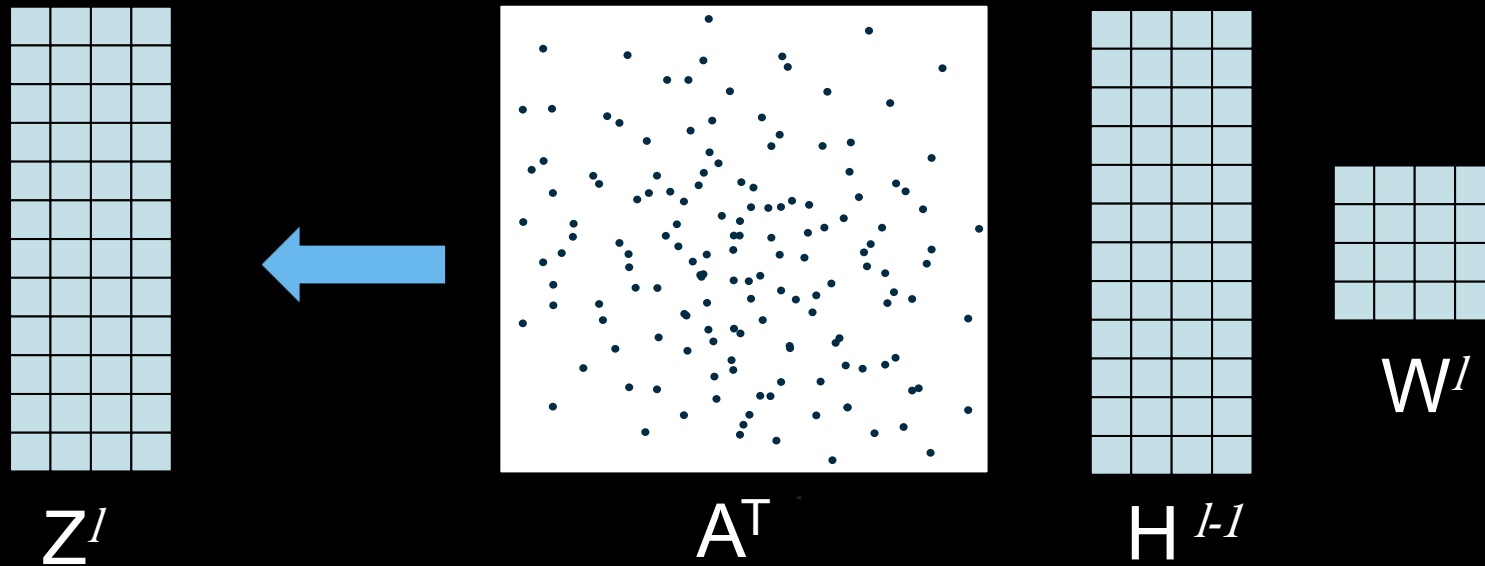
And this includes AI researchers!

Specialization: Are we looking under the lamppost?



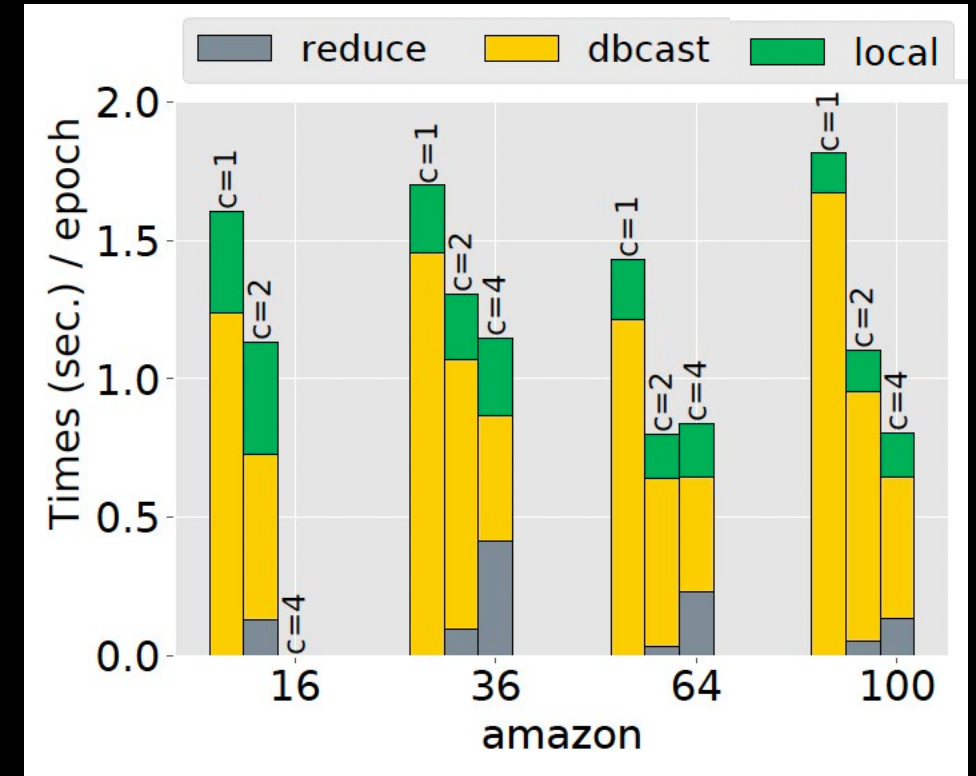
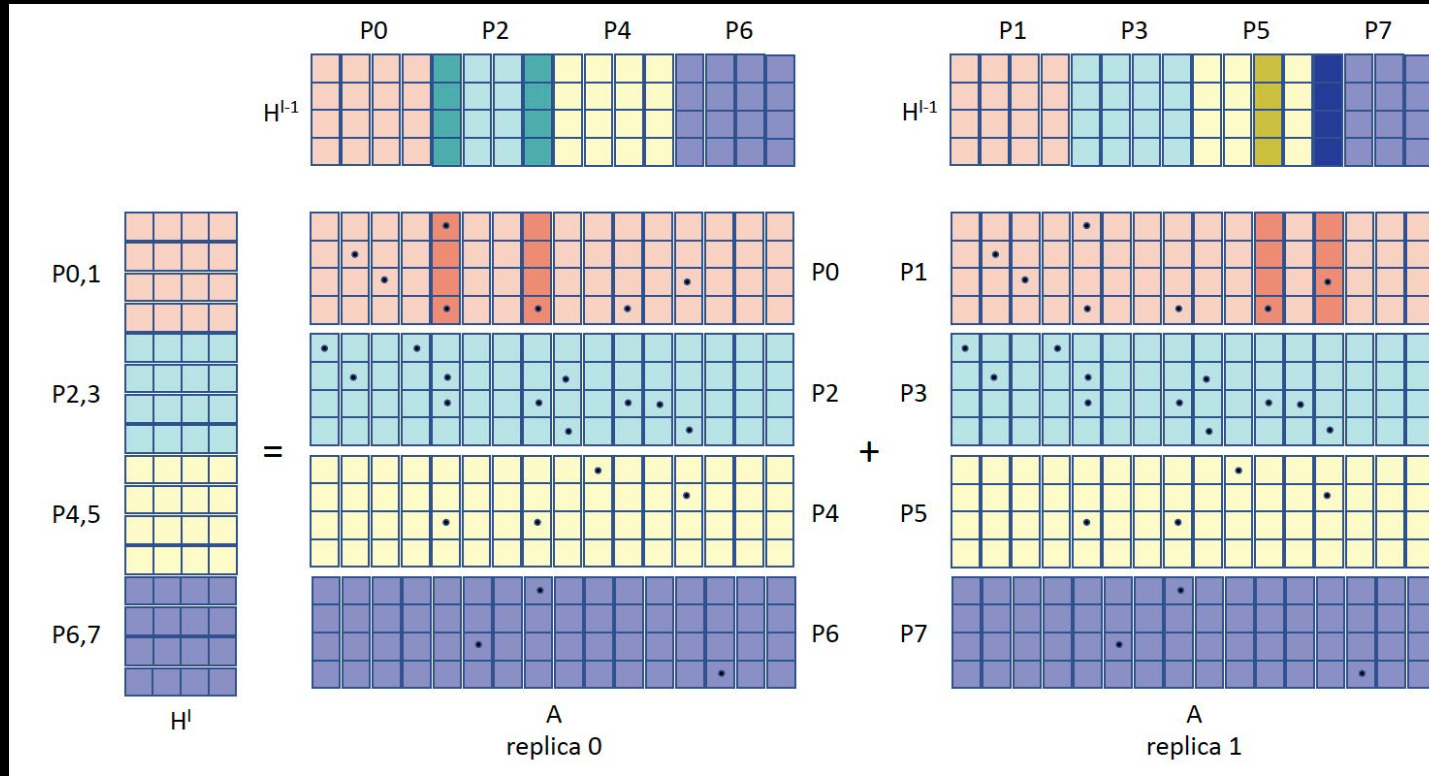
Remember when the Linpack Benchmark represented scientific computing?

Bottleneck in GNN Training



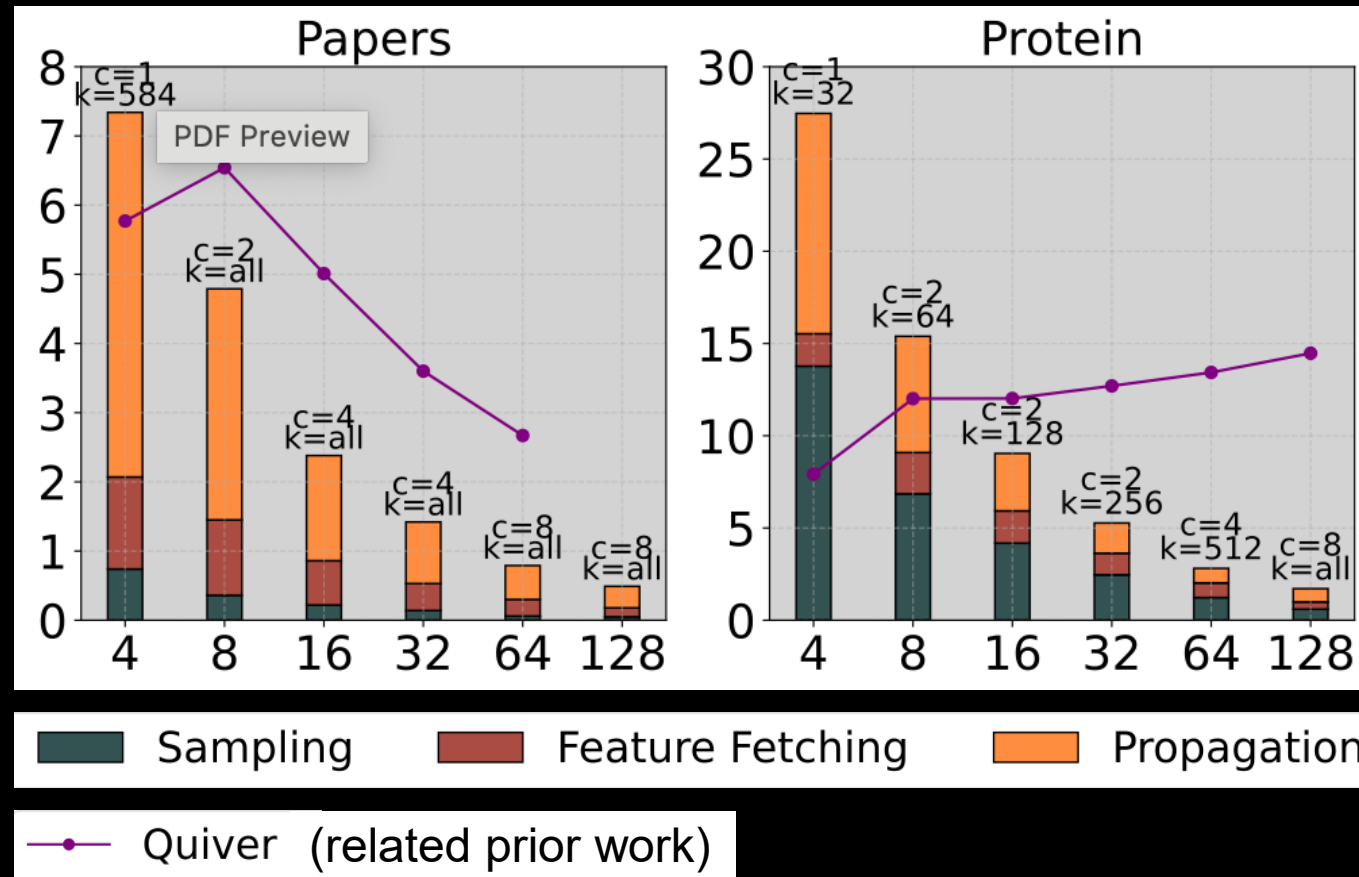
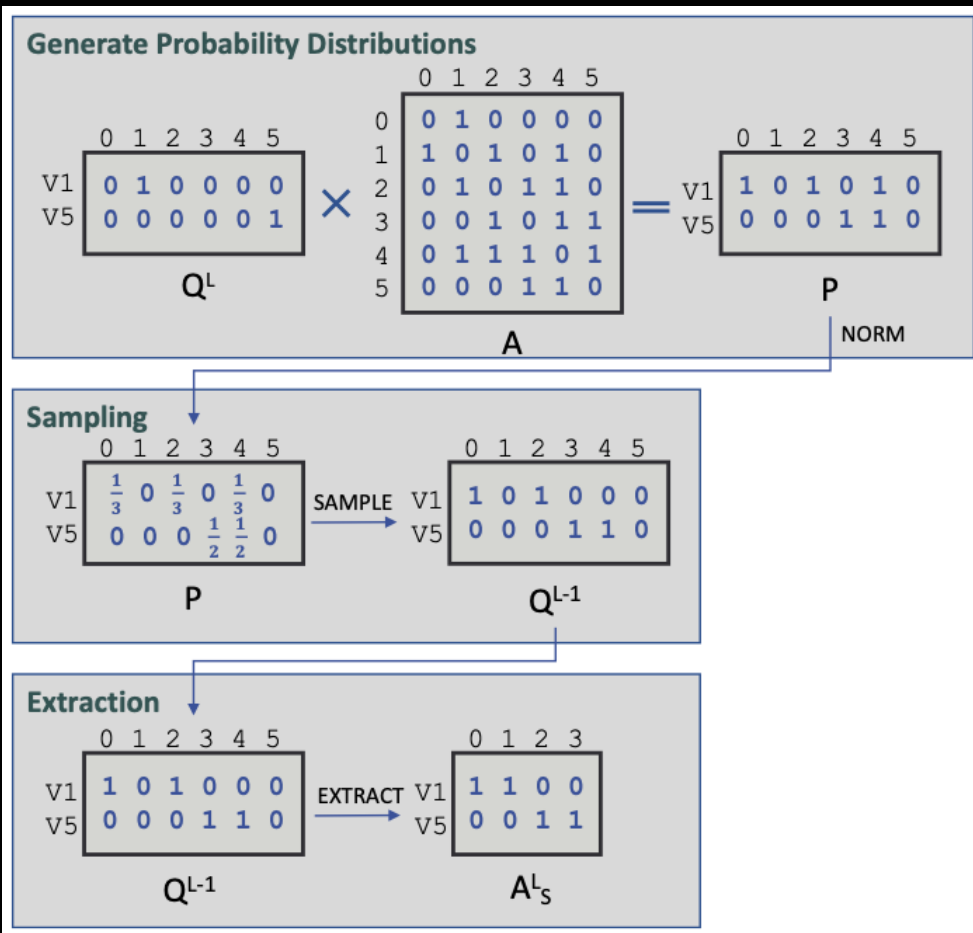
- $A^T H^{l-1}$ sparse-dense matmul (SpMM)
- $(A^T H^{l-1}) W^l$ dense-dense matmul (DGEMM)
- **SpMM is the bottleneck, not DGEMM!**

Avoiding Communication in GNNs



Tripathy, Yelick, Buluc, Reducing Communication in Graph Neural Network Training, SC'20
Mukhopadhyay, Tripathy, Selvitopi, Yelick, Buluc, Sparsity-aware communication ..., ICPP '24

Sampling as Communication-Avoiding Sparse Matrix Operations



Parallelism Kernels – Dice in the Hardware Lottery

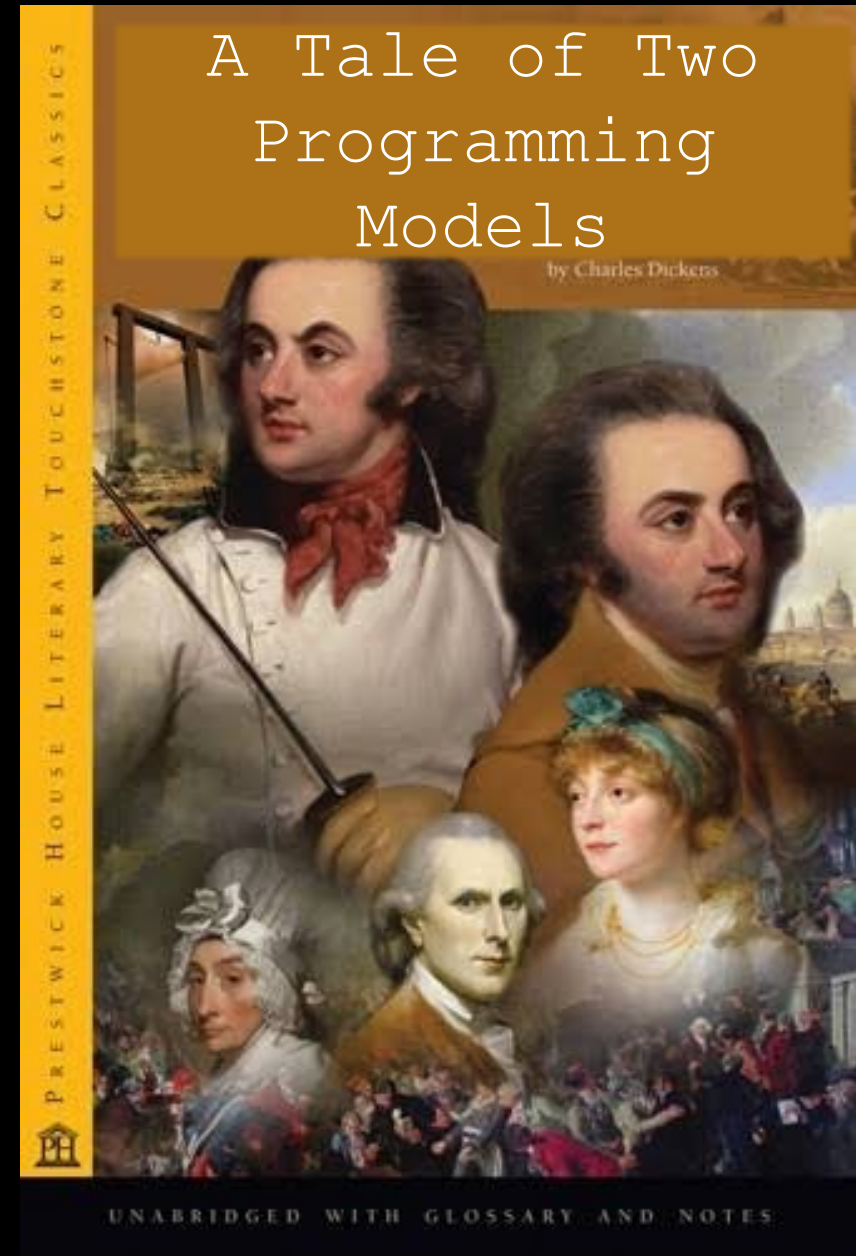
7 Dwarfs of Simulation	7 Giants of Big Data
Particle methods	Generalized N-Body
Unstructured meshes	Graph-theory
Dense Linear Algebra	Linear algebra
Sparse Linear Algebra	
Spectral methods	Hashing
Structured Meshes	Sorting
Monte Carlo methods	Alignment
	Basic Statistics

Phil Colella

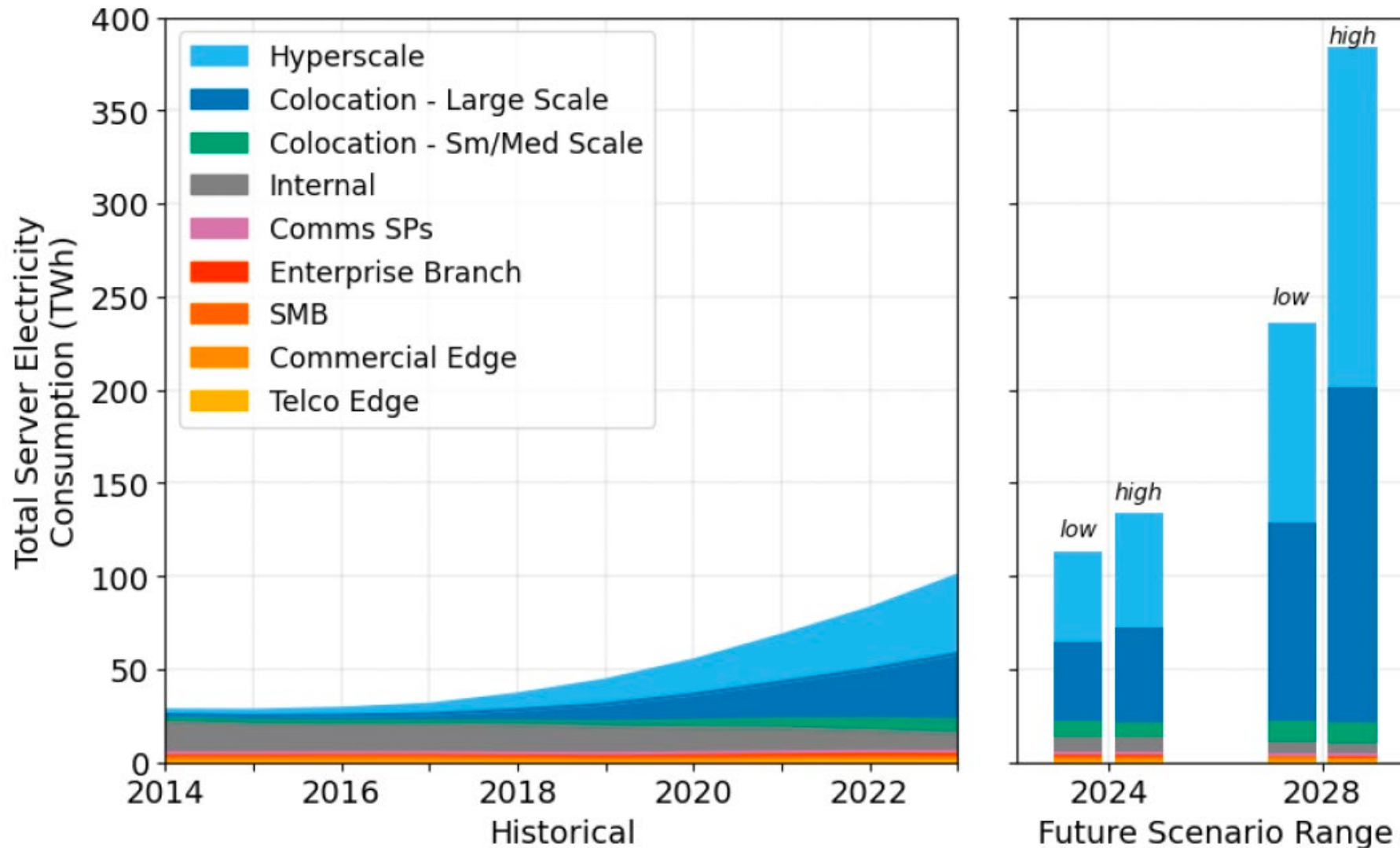
NRC Report + our paper

Partitioned Global Address Space

- 1-Sided Communication
 - Locality control
 - Never cache remote data
-
- Generalized Sparse Matrices
 - Graph BLAS
 - With Semiring operations



Will Computing and Data Center Power Dominate?

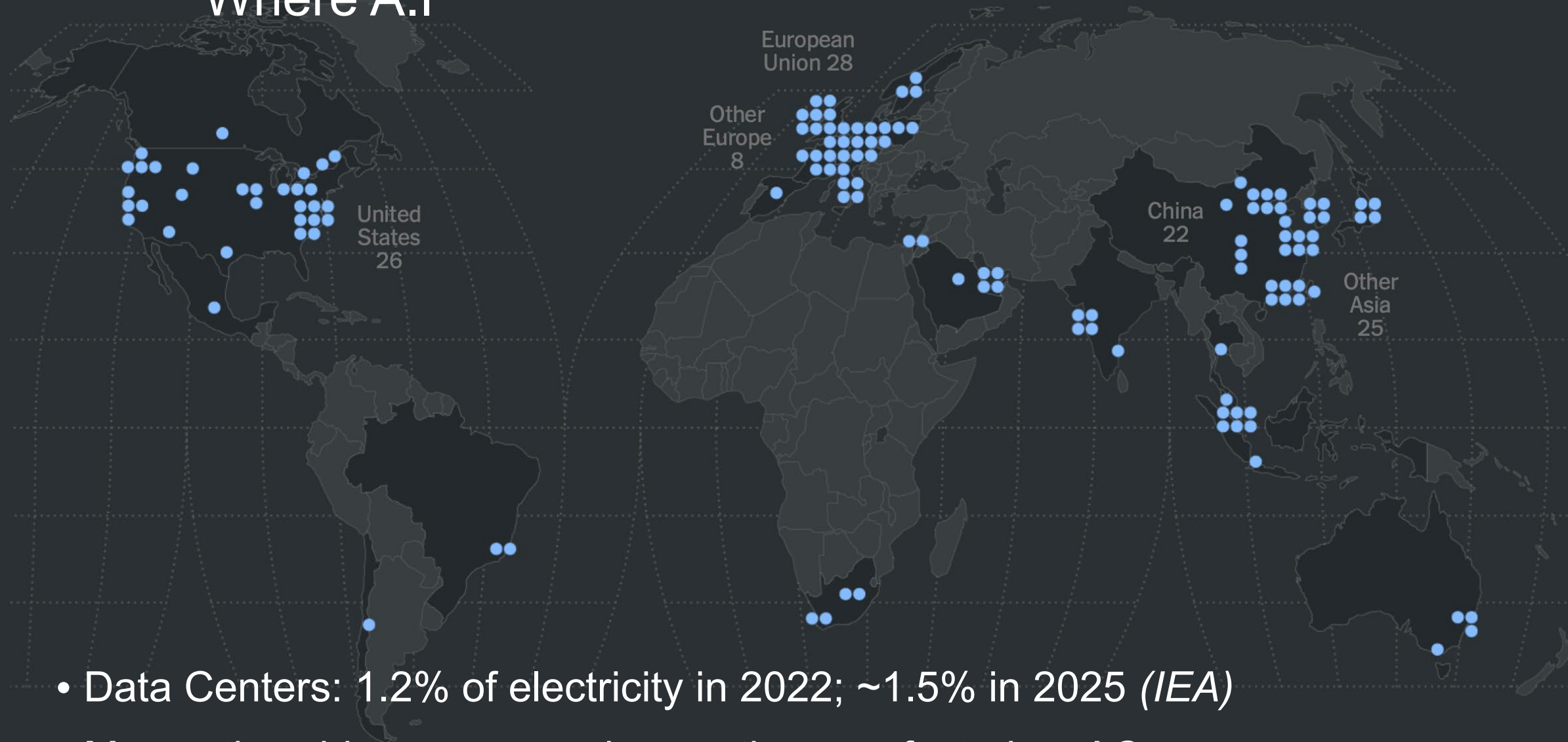


- AI driving growth
- LBNL Reports on US trends

WHERE A.I. DATA CENTERS ARE LOCATED

Only 32 nations, mostly in the Northern Hemisphere, have A.I.-specialized data centers.

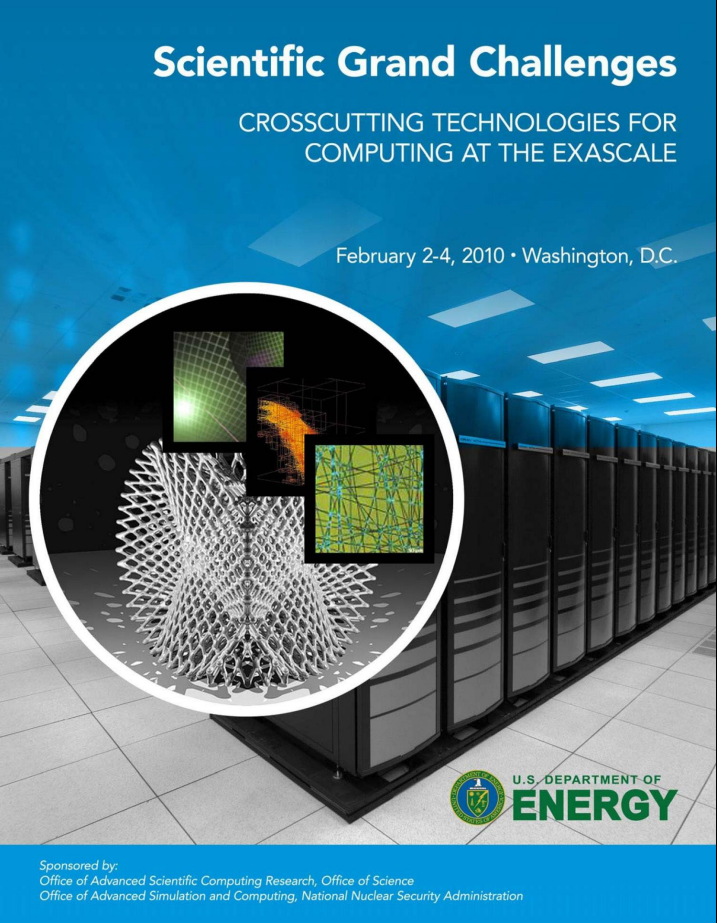
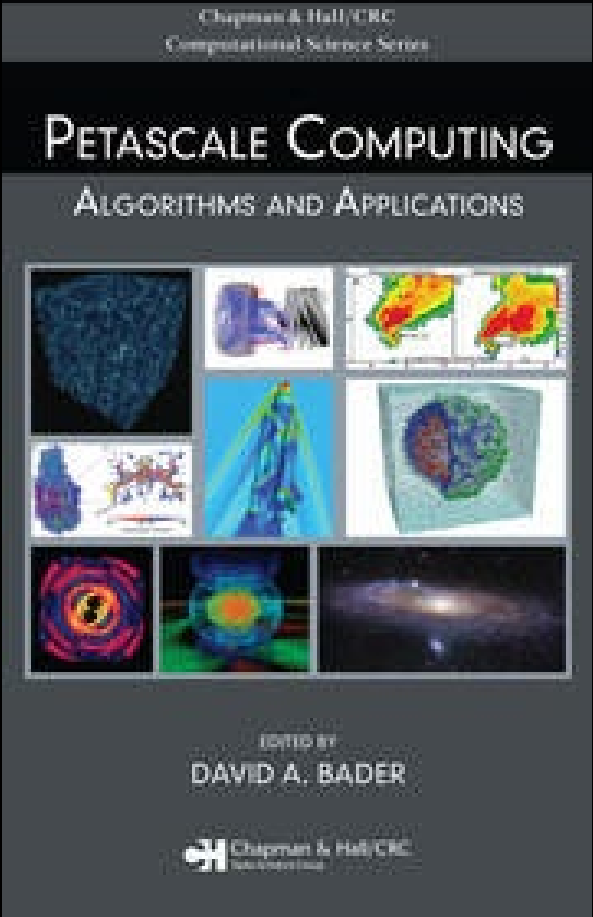
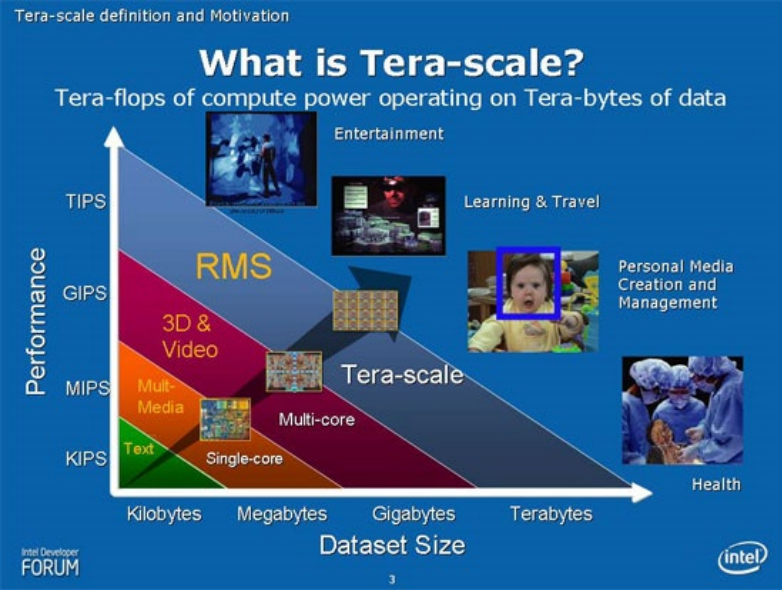
Where A.I



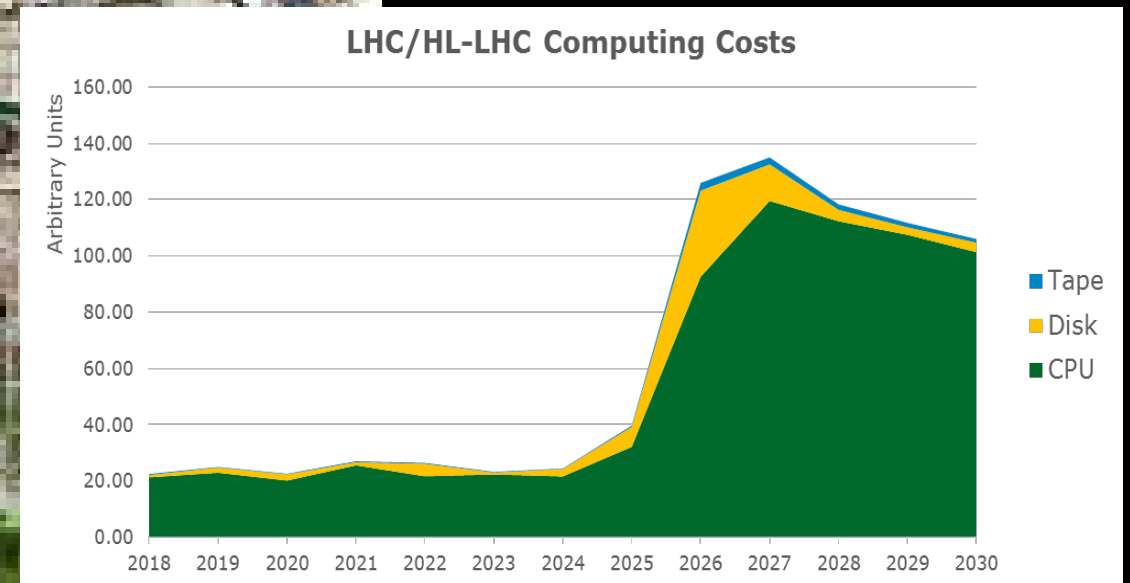
- Data Centers: 1.2% of electricity in 2022; ~1.5% in 2025 (*IEA*)
- Many other drivers: economic growth, manufacturing, AC

Source: Oxford University • Note: Count of data centers in China excludes facilities in Hong Kong and Taiwan.

It's hard to think exponentially



It's also hard to stop: Prediction of Atlas computing +\$1B





Thanks to Moore's Law!

It's been a great run!