

# Future Computing Technology Landscape: Challenges and Opportunities

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## Computing and HEP

- ▶ Computing is an essential enabling and empowering component of almost all aspects of HEP science
- ▶ Computing within HEP has a long history (~70 years) including notable contributions to High Performance Computing (HPC), High Throughput Computing (HTC), and large-scale Data Science
- ▶ Substantial resources are devoted to computation and data science as an essential aspect of the HEP scientific enterprise



Seattle Snowmass Summer Meeting,  
Future of Computing for HEP  
July 23, 2022

## HEP computing is complex and very diverse

### ► Activities cover the full range of the computational environment

- At all scales and team sizes — individual, local groups, large teams
- General-purpose computing, HPC, HTC, data management (including data motion; important role of networking)
- Different topical areas in HEP have overlaps as well as divergences; also can have quite different points of view
- Substantial inertia in evolving the current software base (size, diversity, complexity)
- Future technologies will impact each area in different ways — some will be heavily impacted, others not so much
- There is an enormous amount of information in the CompF papers, documentation, and talks — developing roadmaps for the future is a difficult task!
- Most CompF reports emphasize key aspects of technology and systems evolution

### ► Key issues for future of HEP computing

- Manage complexity and diversity
- Be ready to embrace specialization
- Emphasis on portability and reproducibility
- Exploit (different types) of loosely connected systems
- Argue for, and help develop, common interfaces/edge services
- To the extent possible, integrate approach with vendor and industry roadmaps

## Different flavors of computing

### ► High Throughput Computing ('Grid')

- Distributed systems with a relatively slow network (loosely-coupled jobs)
- Batch processing with a large number of relatively independent jobs

### ► High Performance Computing ('Supercomputing')

- Parallel systems with nodes designed for compute-intensive tasks and a fast network (tightly-coupled jobs)
- Batch processing with a small number of large individual jobs

### ► Interactive Analytics ('Cloud')

- Parallel systems with balanced I/O and networking
- Interactive processing with fast on-demand cluster configurations

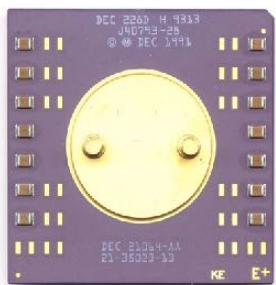
### ► Edge Computing

- Computing at the data source, at the network 'edge' (IoT devices, detectors, controls, —)
- Fast, dedicated local analysis and storage

## Computing: Recent past, present, and future

### Living in the past — 1980/2000

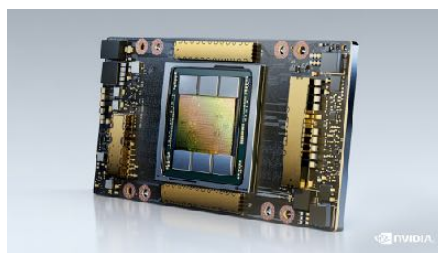
- ▶ Moore's law reigns
- ▶ Life is good!
- ▶ 2x rule of thumb for effort on performance
- ▶ CPU cost/performance ratios very favorable
- ▶ Scientific computing *rides on general computing advances*
- ▶ Global parallelism did not prove difficult to address



DEC Alpha  
CPU

### Facing the present — 2000/2025

- ▶ (Conventional) Moore's law ends
- ▶ Life starts becoming nontrivial
- ▶ CPU cost/performance stalls
- ▶ Local concurrency must be faced (rise of GPUs)
- ▶ Computing advances fragment
- ▶ Scientific computing *roadmap begins to look unclear*



NVIDIA A100  
GPU

### Confronting the future — 2025+

- ▶ Disruptive technologies? Likely niche applications
- ▶ Multiple tech roadmaps, but no major changes (too much hardware/software/community inertia)
- ▶ Life is going to be tougher
- ▶ Significant specialization
- ▶ Scientific computing *roadmap may require actual planning*



Anton 3 ASIC  
DE Shaw Research



# HEP as a computing technology driver/consumer

## ► Experiments: Requirements set primarily by throughput

- Data 'velocity' in burst and quasi-continuous modes controls computational requirements
- Needs vary across experiments (size, one-shot vs. multi-pass, technology history, community preferences, workflow complexity, other technical requirements)
- Experiments stress the total computing environment — computing as well as IO and data management, thus in-transit computing and processing-in-memory can play useful roles
- Most experimental workflows have limited arithmetic intensity (low flops/byte ratio) and limited data reuse
- HEP is not alone — similar issues are faced in other fields (e.g., light sources)

## ► Theory/Modeling/Simulation: HPC Requirements

- Most requirements (with some variations) similar to those of HPC applications in other fields
- Software development cycle and sustainability are key concerns



## General observations

### ► Advantages of general purpose computing (e.g., CPUs)

- Advances in hardware directly translate to improvements in application performance
- Higher-level software stack largely independent of (local) hardware implementation
- Responsibility for performance optimization lies largely outside the realm of (high-level) applications (reliance on compilers)
- Relatively fixed set of algorithms — focus on improved implementations; algorithmic development not directly connected to underlying hardware
- Although overall technology advances may be rapid, the effect on software development cycles (traditionally long) can be relatively small

### ► Disadvantages

- Stagnation in application performance as the underlying technology stalls
- Performance engineering over a finite set of algorithms can only produce limited gains
- Different approaches to computing naturally arise to fight performance stalls
- Possible danger of being left in the “slow lane” as hardware/software technology evolve in different directions; software development cycles need to be sped up to keep pace

## Trend towards heterogeneity

### ► Transistor limitation issues

- Dennard scaling — reduce transistor sizes by 30% every generation but keep electric fields constant (increased transistor count, increased performance, reduced supply voltage keeps power use constant)
- Transistor density increase led to complex architectures capable of multiple optimizations (out-of order execution, speculation, pipelining, cache hierarchy —), adding more general capability
- Power consumption limits (leak currents, frequency and supply-voltage limits) + finite power budget drives trend to multiple (simpler) cores and customization (algorithmic or restricted parallelism)

Borkar & Chien, CACM 54, 67 (2021)

### ► Software/application ramifications

- Higher-level software stack can no longer ignore lower-level hardware realities, no more simply “riding the wave” of Moore’s law (Dennard scaling ended in 2004/2005)
- Algorithm choices controlled/restricted by low-level architecture — worst-case scenarios can involve poor trade-offs for many scientific applications (small winner pool, large loser pool — less diversity in scientific applications, bad for HEP computing given its intrinsic breadth)
- Management of hardware specialization is the major challenge going forward — involves all aspects of problem specification, solution strategies, algorithmic implementations, overall software environment, includes management of heterogeneity at local and system-scale levels

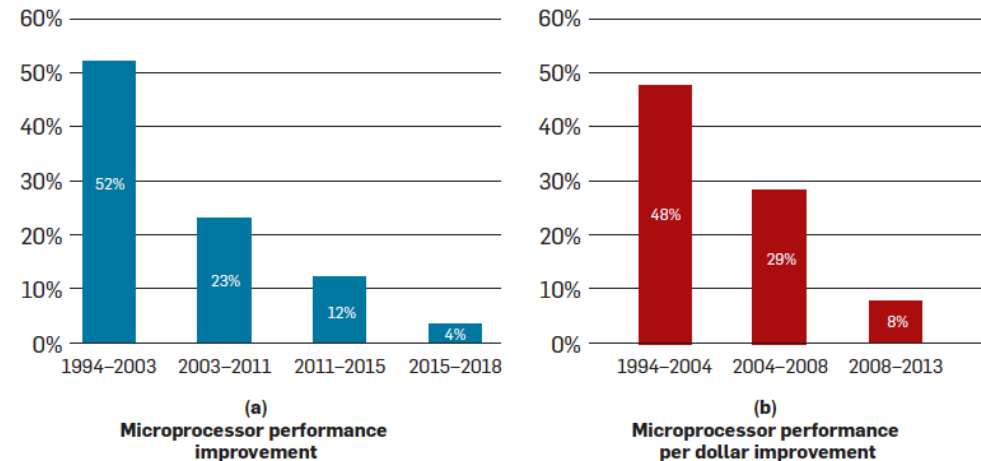
## Trend towards specialization

### ► 'Winners'

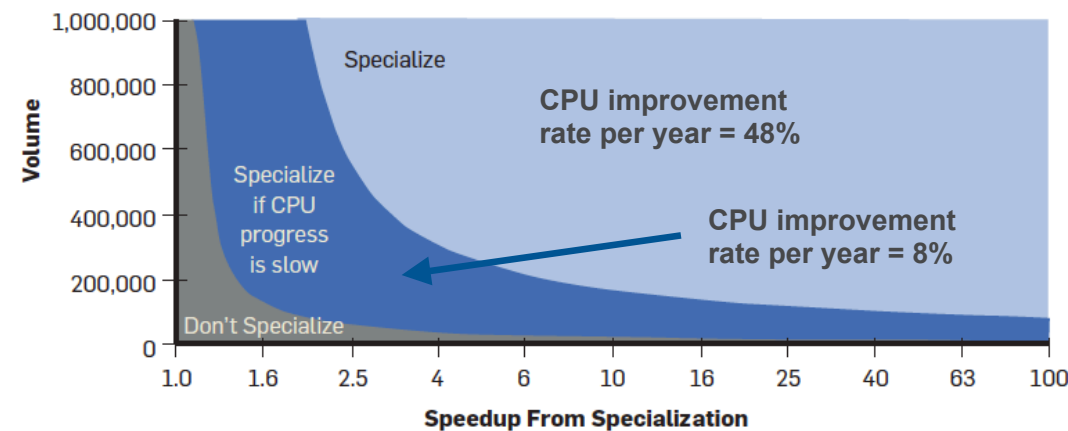
- Applications that can exploit significant parallelism
- Computational tasks that can be arranged in stable configurations with a regular cadence
- Limited memory accesses for a fixed computational effort
- Allow use of fewer degrees of precision (e.g., AI/ML applications)
- Low-power applications

### ► Specialization trends

- Slowdown in CPU performance gain makes specialization more attractive
- As the threshold for specialization is lowered, more applications can benefit from it (**good**)
- This can be a driver for fragmentation (**bad**)



Credit: Thompson & Spanuth CACM 64, 64 (2021)



Economic model for the rationale behind specialization



## Guessing the future

### ► Technology roadmap disruptions are hard to predict

- If past history is a guide, it is dangerous to predict a major change with certainty — either the technology does not arrive or the predicted timescale for it is significantly off
- Most cutting-edge technology roadmaps are stable on the timescale of 2 or 3 years (or even less!)
- Global user communities, entrenched software base, transition costs, control technology timelines, it is difficult for HEP to be an active player (other markets are much too dominant)
- Truly disruptive technology ideas (quantum, photonics) are not competitive enough yet to make a practical difference (aside from possibly niche applications)
- Most gains in the near-term (5-10 years) are likely to come from a combination of multiple factors
- From the HPC perspective, if we are currently at the exascale, then we may expect 20+ exaflops (2025+) and 100+ exaflops, (2030+), keeping the power budget fixed to current values

### ► Incremental improvements

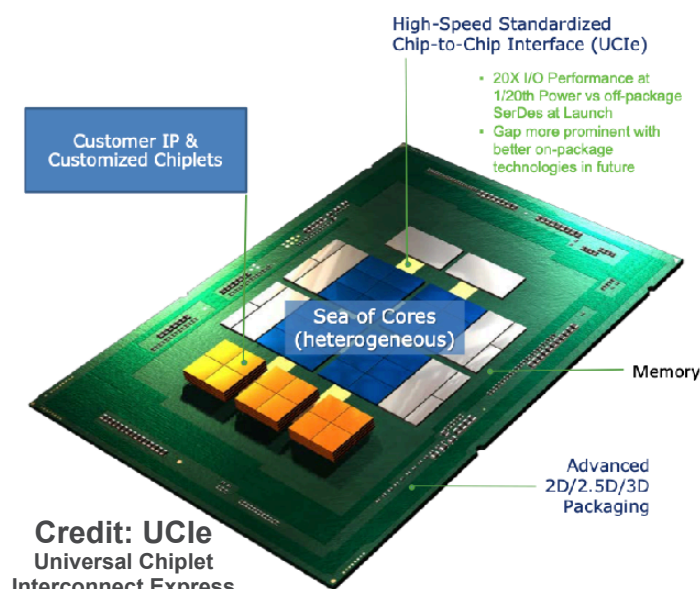
- Hardware architecture
- Data motion
- Packaging advances
- Thermal environment
- Computation in place and in flight
- AI/ML-driven large-scale data integration with HPC
- Diversity of networked resources (local systems, large-scale facilities, cloud)
- Still plenty of room for big gains: see, e.g., **Leiserson et al. Science 368, 1079 (2020)**

# Ramification for major HEP applications: Co-design vs. specialization

## ► Future technologies

- Diversity of approaches appears to characterize the 2025+ technology roadmap
- Low-power CPU, CPU/GPU, FPGA, AI/ML architectures (e.g., TPUs), all compete in this space
- Specialized architectures (ASICs) can be part of the solution — negatives are cost and potential obsolescence as hardware technology evolves
- Co-designed approaches that focus on algorithmic flexibility and ability to leverage vendor and open source methodologies for portability are likely preferred
- Hardware co-design possibilities may be better than in the past via chiplet integration (supported by major players — AMD, Google, Intel, Meta, TSMC, —)
- Software stack will require more low-level expertise than in the past, but physicists can still be targeted as the primary code writers via use of performant higher-level frameworks (e.g., as in present-day AI/ML)

### OPEN CHIPLET: PLATFORM ON A PACKAGE



Credit: UCle  
Universal Chiplet  
Interconnect Express

Heterogeneous Integration Fueled by an Open Chiplet Ecosystem  
(Mix-and-match chiplets from different process nodes / fabs / companies / assembly)

SoC vs. motherboard

## A Speculative Summary

- ▶ The trend towards heterogeneity and specialization is irreversible, as CPU performance gains will remain modest
- ▶ Over the next decade, no radically new computing technology is likely to get us back to the Moore's Law era of the recent past
- ▶ There will be winners and losers, the losers are those who:
  - Cannot get a worthwhile performance boost from specialization/co-design
  - Do not have large enough requirements (or enough funding) to exploit specialization
- ▶ For HEP computing to be on the winning side, we should
  - Actively consider new algorithmic approaches to solving our problems, if possible (e.g., AI/ML approaches, data restructuring)
  - Actively consider coordinating mechanisms to form a big enough market (this will require a combination of co-design and perhaps a more limited version of specialization) — national facilities and commercial or public clouds could also aid in providing this function

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