



HYPERION RESEARCH

Research Findings: HPC-enabled AI



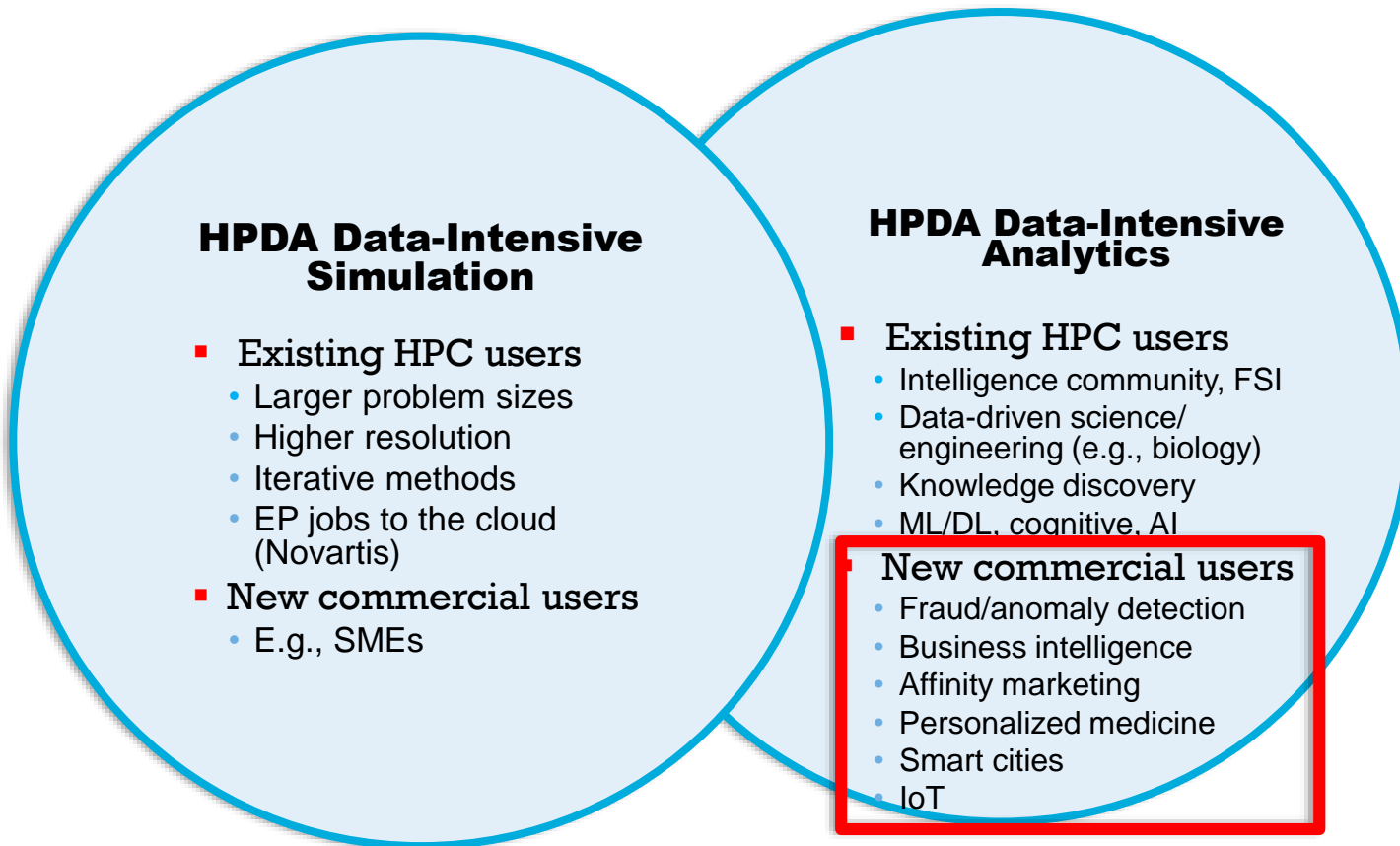
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Recent Worldwide Studies/Projects for U.S. Federal Agencies

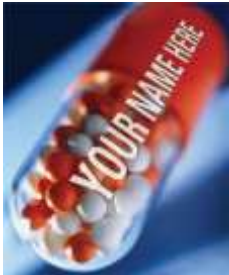
- The Evolution of AI Hardware and Software Ecosystems
- The Evolution of Field Competencies in Machine/Deep Learning and Resultant Industries
- AI Primer for Senior Decision-Makers
- AI Hardware Technology, Vendor Status and Trends



Converging HPC-Enterprise Market



Most Economically Important HPDA-AI Use Cases



Precision Medicine

Automated Driving Systems

Fraud and anomaly detection

Affinity Marketing

Business Intelligence

Cyber Security

Edge/IoT/Smart Cities



High Growth Areas: HPDA-AI (May 2019)

- HPDA is growing faster than overall HPC market.
- AI subset is growing faster than all HPDA.



Table 1

Forecast: Worldwide HPC-Based AI Revenues vs Total HPDA Revenues (\$ Millions)

	2018	2019	2020	2021	2022	2023	CAGR 18-23
WW HPC Server Revenues	13,706	14,495.000	15,780	17,376	18,983	19,947	7.8%
Total WW HPDA Server Revenues	3,153	3,598	3,932	4,737	5,467	6,450	15.4%
Total HPC-Based AI (ML, DL, and Other)	747	938	1,094	1,399	1,810	2,725	29.5%

Source: Hyperion Research 2019

Table 2

Forecast: Worldwide ML, DL & Other AI HPC-Based Revenues (\$ Millions)

	2018	2019	2020	2021	2022	2023	CAGR 18-23
ML in HPC	532	675	875	1130	1479	1940	29.5%
DL in HPC	177	216	301	392	510	665	30.3%
Other AI in HPC	38	47	66	80	95	120	25.9%
Total	747	938	1,094	1,399	1,810	2,725	29.5%

Source: Hyperion Research 2019

Where Do You Run HPC workloads? (Choose ALL that apply)

On-premise HPC data center	67.2%
On-premise enterprise data center (business operations)	36.2%
More than one external cloud (e.g., Amazon, Google, Microsoft)	29.3%
On-premise private or hybrid cloud	19.0%
One external cloud (e.g., Amazon, Google, Microsoft)	19.0%
Not sure/don't know	1.7%
Other	3.5%

Machine Learning Goes Back At Least to the 1950s

Thomas <
Bayes
Statistician
1702-1761



Decade ↕	Summary ↕
<1950s	Statistical methods are discovered and refined.
1950s	Pioneering machine learning research is conducted using simple algorithms.
1960s	Bayesian methods are introduced for probabilistic inference in machine learning. ^[1]
1970s	' AI Winter ' caused by pessimism about machine learning effectiveness.
1980s	Rediscovery of backpropagation causes a resurgence in machine learning research.
1990s	Work on machine learning shifts from a knowledge-driven approach to a data-driven approach. Scientists begin creating programs for computers to analyze large amounts of data and draw conclusions – or "learn" – from the results. ^[2] Support vector machines (SVMs) and ^[3] recurrent neural networks (RNNs) become popular. The fields of ^[4] computational complexity via neural networks and super-Turing computation started.
2000s	Support Vector Clustering ^[5] and other Kernel methods ^[6] and unsupervised machine learning methods become widespread. ^[7]
2010s	Deep learning becomes feasible, which leads to machine learning becoming integral to many widely used software services and applications.

Source: Wikipedia

1990s: HPC-Enabled Machine Learning

- By the early 1990s, George Washington University Hospital (Washington, DC) was routinely using a Cray supercomputer to help detect breast cancer after training it to identify early indicators, called microcalcifications, on X-ray films with better-than-human ability.

1993 Finalist
Computerworld
Smithsonian
Awards



Coupled Environments

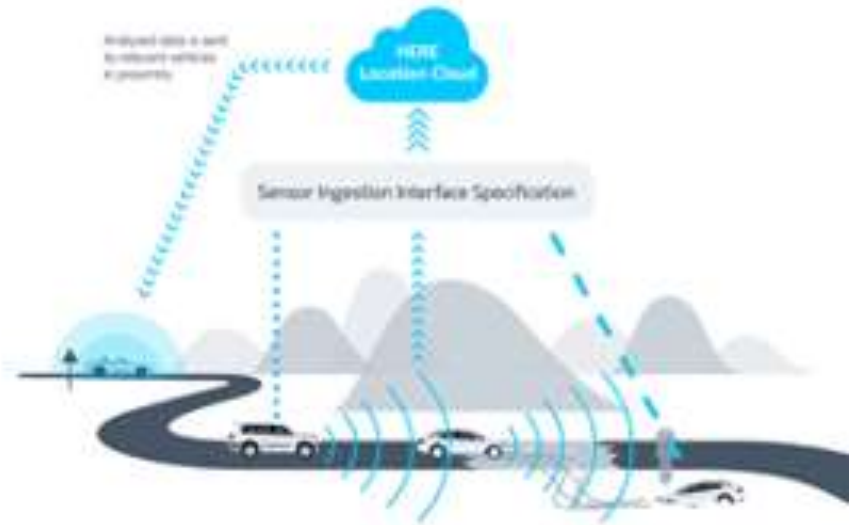
■ Automated Driving Systems

- Embedded processor for local control (car-car, car-environment)
- Private cloud for citywide and beyond (“air traffic control”)

■ Healthcare/Precision Medicine

- Healthcare systems are already private cloud-based.
- Future: couple in-office HPC decision-support engine to private cloud.

■ 5G Will Reduce Local-Cloud Latency Issue



AI Is Still Near the Start

Low IQ (Weak Inferencing)

High IQ (Strong Inferencing)

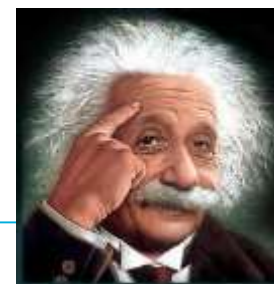
Today: Bounded Problems

- Many observations but few choices to make
- “One trick dogs”: 10 AI solutions in a box to solve 10 problems
- Already very useful:
 - Image & voice recognition
 - Advanced driver assistance
 - Reading an MRI



Future: Unbounded, Too

- Many observations, many choices to make
- Versatile decision-makers capable of serious experiential learning
- Examples:
 - Discerning human motivation
 - Fully automated driving
 - Diagnosing/”curing” a cancer



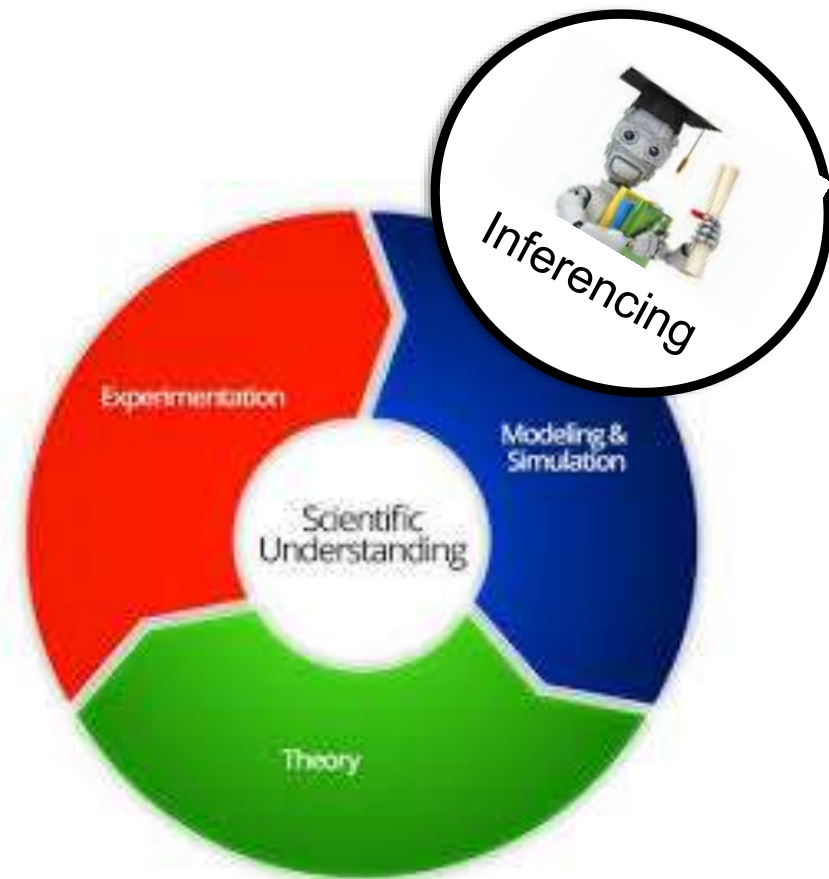
Hyperion Research Worldwide Survey of Leading AI Experts

Importance of HPC for Advancing AI

Importance	Number	Percent
Extremely important	47	87
Somewhat Important	7	13
Not very important	0	0
Not sure/don't know	0	0
Total	54	100

Intelligent Simulation

- AI adds fourth branch to the scientific method, ***inferencing***. Complements theory, experiments & established simulation methods.
- Inferencing is the ability to guess, based on incomplete information
- Simulation is becoming much more data-intensive—esp. iterative methods.
- When inferencing is applied to data-intensive simulation, the result is ***intelligent simulation***.



Building Consumer Trust in ADS

- RAND Corp. estimates 8.8 billion miles of physical testing would be needed to attain 95% consumer trust in self-driving vehicles. This would take 400 years.
- Adding HPC simulation can reduce time frame to 5-10 years.



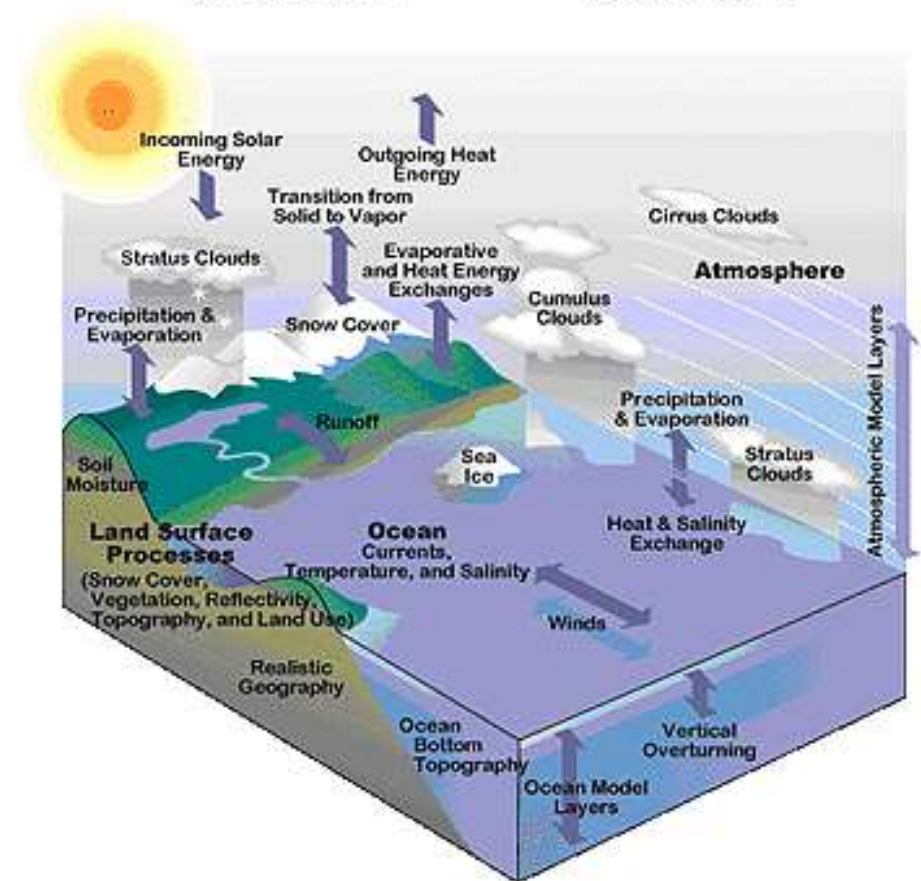
Climate Knowledge Discovery

- Climate research is inherently data-intensive.
- Ensemble models & adding new factors (e.g., carbon cycle) have made it much more so.
- Climate knowledge discovery algorithms add a data analytics approach.
- The first IEEE workshop on this topic (2008), was called "Data Mining for Climate Change and Impacts."

An Approach for Predicting CO₂ Emissions using Data Mining Techniques

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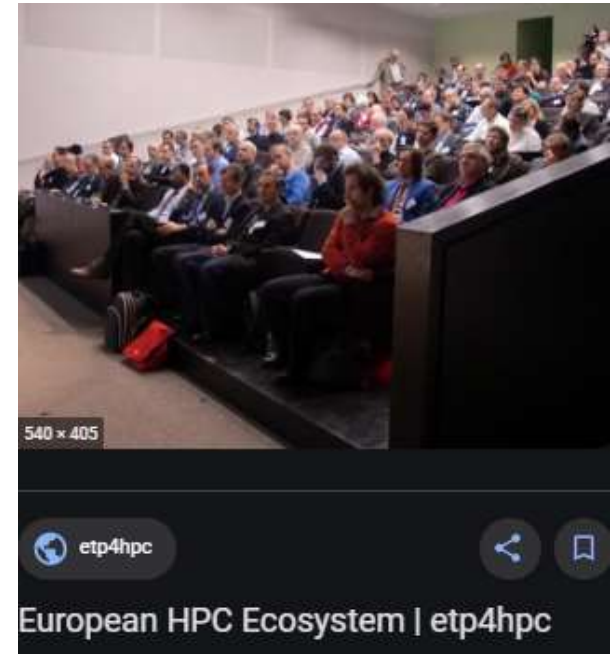
One Swim Lane in the Future



- Today: Mostly simulation and analytics on same compute-centric HPC system (budget reasons)
- In 2 years: More buyers will acquire separate system for analytics-AI.
 - But orthogonal findings of simulation & analytics runs will still need to be combined in the researcher's brain.
- Farther ahead: same system efficiently performs concurrent simulation & analytics runs – and integrates the orthogonal results.
 - Finally, economically efficient!

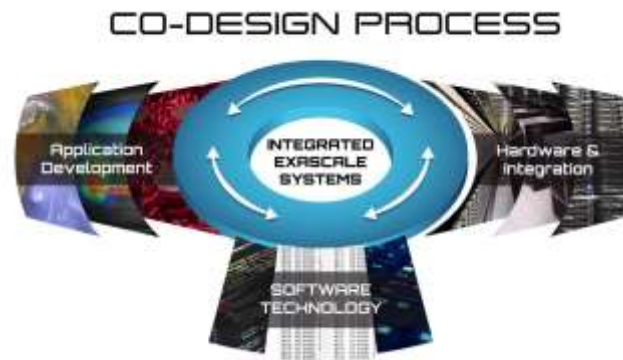
An Exploding Ecosystem

- The AI ecosystem has been growing quickly
- Targeted problems are more complex
- One result: new, AI-specific hardware
 - From companies large and small
 - China is very active
- The categories of processors and technologies continue to grow as well, and now include:
 - GPUs
 - TPUs
 - FPGAs
 - ASICs & eASICs
 - Neuromorphic Chips
 - IPU
 - Inference Chips
 - Training Chips
 - Dataflow processors
 - Vector processors
 - 3D stacking
 - Optical interconnects



Co-Design

- AI chips will be centered on co-design, with specific tasks in mind. Examples:
 - Low-power ASICs at the edge
 - Custom AI chips in hyperscale data centers or the cloud
- GPUs will remain important but not for all AI workloads.
- Software and model-designed hardware is the direction forward.



Power

- Power consumption is critical to chip design for AI workloads.
 - Low power chips can be placed closer to the edge.
 - Latency for near real time training and inference require the compute to be next to the stored data.
- Processing and memory also need to be closer together.
 - With faster interconnect/fabric speeds



Cloud Companies Joining the Processor Development Party

- Google uses tensor cores to accelerate machine learning workloads.
 - Only available on Google cloud for now
 - Google announced the third generation TPU last year.
- Amazon, at their re:Invent conference in November of 2018, announced their inference chip, Inferentia.
 - Designed to accelerate machine learning, especially inferencing.

