# The Convergence of AI and HPC: A New Hybrid Architecture for High Performance Data Analytics

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# Agenda

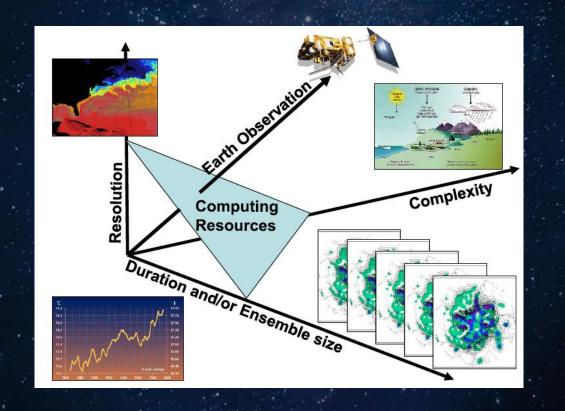
- 1. Factors driving the convergence of HPC, HPDA and Al Architectures
- 2. Combining the strengths of HPC and Al
- 3. BullSequana XH2000: A new hybrid architecture for HPC and Al
- 4. Codex Al Suite: A new framework for developing cognitive applications
- 5. BullExtreme Factory: HPC within a hybrid cloud services model
- 6. Expanding Extreme Factory for Deep Learning
- 7. Conclusions



# Convergence of HPC, HPDA and Al architectures

# Insatiable need for increasing compute power:

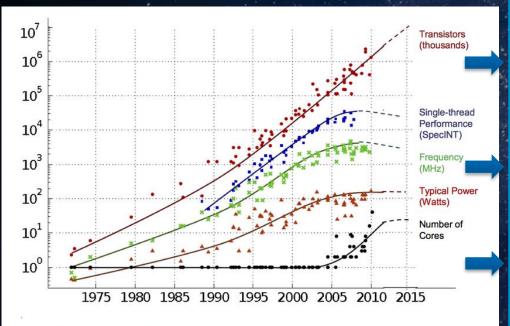
E.g. Ensemble Prediction Systems (EPS) in weather forecasting





# Increasing parallelism inevitable

Putting additional demands on the programmer

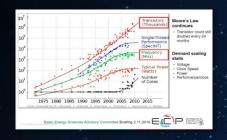


Original data collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond and C. Batten Dotted line extrapolations by C. Moore

- Number of transistors continues to grow roughly in proportion to Moore's law
- Dennard Scaling (power density roughly constant as transistors get smaller) has broken down
- Single thread performance and processor frequencies have now plateaued
- Computational power now coming from increasing number of cores per processor
- Increasing parallelism inevitable



# Factors Driving Innovation in HPC and DL



- End of Dennard Scaling places a cap on single threaded performance
- Increasing application performance will require fine grain parallel codes with significant computational intensity

- Al and Data Science emerging as important new components of scientific discovery
- Dramatic improvements in accuracy, completeness and response time yield increased insight from huge volumes of data





- Cloud based usage models, in-situ execution and visualization emerging as new workflows critical to the science process
- Tight coupling of interactive simulation, visualization, data analysis/Al





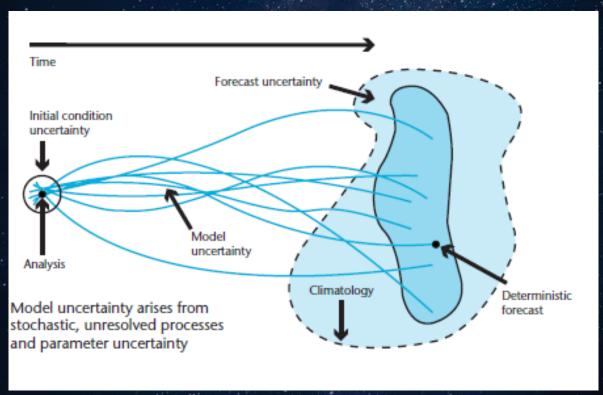
# Incorporating DL as part of the HPC Workflow

HPC	Deep Learning
Long history of modelling and simulation of physical phenomena. Track record of enabling grand challenge scientific discovery and proven return on investment in multiple science domains	New methods to improve predictive accuracy, insight into new phenomena and response time with previously unmanageable data sets
<ul> <li>Develop training data sets using first principal models</li> </ul>	Train inference models to improve accuracy and comprehend more of the physical parameter space
<ul> <li>Apply Bayesian regression methods to expedite/ensure training accuracy</li> </ul>	Implement inference models with real time interactivity
<ul> <li>Incorporate Al models in semi-empirical style applications to improve throughput</li> </ul>	<ul> <li>Analyze data sets that are simply intractable with classic statistical models</li> </ul>
Validate new findings from AI	<ul> <li>Control and manage complex scientific experiments or apparatus</li> </ul>



# Use of DL for uncertainty modelling

Example: Reducing the Parameter sweep in Ensemble Prediction Systems







# Organising HPC + DL Convergence

Future of HPC

### **Transformation**

HPC + DL couple simulation with live data in real time detection/control system

Experimental/simulated data is used to train a NN that is used to for detection/control of an experiment or clinical delivery system in real time.

The NN is improved continuously as new simulated / live data is acquired

### Augmentation

HPC + DL combined to improve simulation time to science > orders of magnitude

Experimental/simulated data is used to train a NN that is used to replace all or significant runtime portions of a conventional simulation.

The NN is improved continuously as new simulated / live data is acquired

### Modulation

HPC + DL combined to reduce the number of runs needed for a parameter sweep

Experimental/simulated data used to train a NN which steers simulation/experiment between runs

The steering NN can be trained continuously as new simulated / live data is acquired

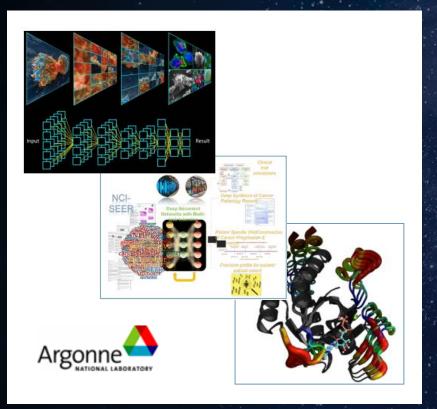
Potential for Breakthroughs in Scientific Insight





# Exascale Deep Learning Enabled Precision Medicine for Cancer

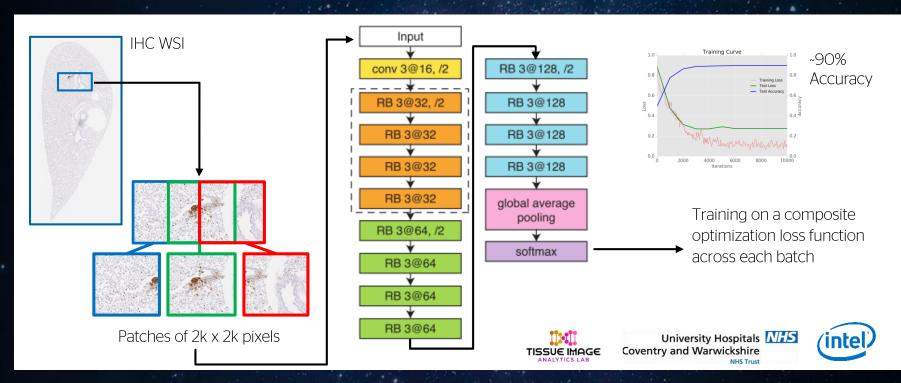
CANDLE accelerates solutions towards three top cancer challenges



- ➤ Focus on building a scalable deep neural network code called the CANcer Distributed Learning Environment (CANDLE)
- CANDLE addresses three top challenges of the National Cancer Institute:
  - Extraction of information from millions of cancer patient records to determine optimal cancer treatment strategies
  - Understanding the molecular basis of key protein interactions
  - 3. Developing predictive models of drug response, and automating the analysis



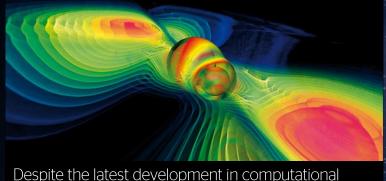
# Semi-supervised Deep Learning to detect tissue abnormalities





# Gravitational Waves – Next steps

Use of Deep Neural Networks to accelerate waveform detection



power, there is still a large gap in linking relativistic theoretical models to observations.

Max Plank Institute



### Background

The aLIGO (Advanced Laser Interferometer Gravitational Wave Observatory) experiment successfully discovered signals proving Einstein's theory of General Relativity and the existence of cosmic Gravitational Waves.

### Challenge

The initial a LIGO discoveries were successfully completed using classic HPC analysis using hundreds of CPU's where the bulk of the processing was done offline. Here the latency is far outside the range needed to activate resources, such as the Large Synaptic Space survey Telescope (LSST) which observe phenomena in the electromagnetic spectrum in time to "see" what aLIGO can "hear".

### Solution

A DNN was developed and trained using a data set derived from the CACTUS simulation using the Einstein Toolkit. The DNN was shown to produce better accuracy with latencies 4500x better than the original CPU based waveform detection.





# BullSequana XH2000 HPC and DL Hybrid Architecture

# BullSequana XH2000

A natural evolution for BullSequana X1000



### A field proven platform

- Blades are backward and forward compatible
- More than 60% of components in common
- Exascale ready
- Platform in production with Governmental, educational and industrial customers











Enhanced flexibility allows to reduce TCO, from one rack up to Exascale projects



# BullSequana XH2000 Designed for Hybrid Computing







Traditional scalar, memory bound or GPU based simulations



Al augmented simulations & Deep Learning training & inference



Big Data Analysis

### Using best in class CPU/GPU/FPGA & Interconnects







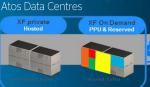




### Seamlessly integrated with Codex AI Suite and Google Cloud Platform













# BullSequana XH2000

Optimized TCO Supercomputer



### **Best in class Power Usage Effectiveness (PUE)**

- 4<sup>th</sup> generation ATOS DLC using warm water cooling up to 40°C (104°F) Inlet, no single fan.
- Capable of cooling highest TDP/Tcase processors on the market, build for the future

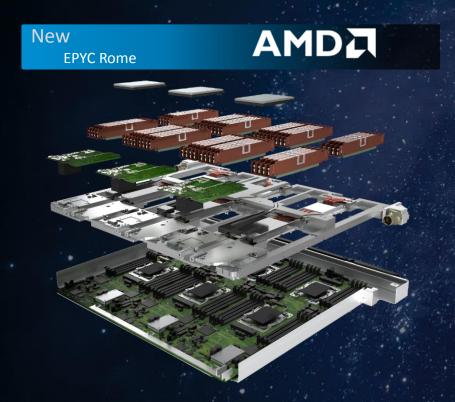
### Very high density and usage of available resources

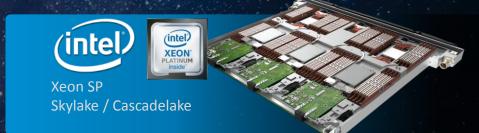
- All-in-one modular design (Compute, Networking, Power, Cooling)
- Up to 96 nodes per rack/in less than 2m³ (68ft³)



# BullSequana XH2000 Blades

The best of all worlds, available in one supercomputer







Volta P100/V100 NVLink2



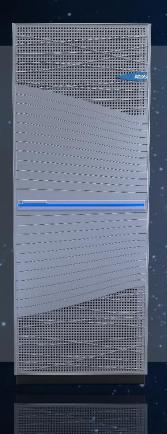


ThunderX2
ThunderX3 when available



# BullSequana XH2000 Networking

Best in class Interconnect flexibility



### New BXI-2 switch for BullSequana XH2000

- DLC cooled Atos cold plate technology
- 48 x 100Gb/s ports
- troubleshooting tool: Traffic generator

### New IB HDR switch for BullSequana XH2000

- DLC cooled Atos cold plate technology
- 40 x 200Gb/s ports / 80 x 100Gb/s ports

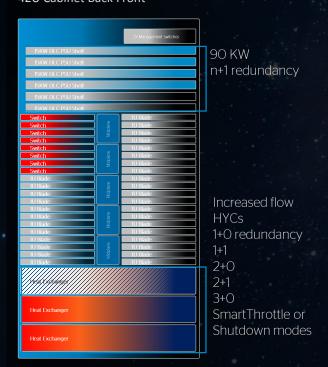
New High-Speed Ethernet network for BullSequana XH2000

**Atos** 

# BullSequana XH2000

Increased flexibility and agility

42U Cabinet Back Front



**42U Cabinet Front View** 



PDU + Power controller up to 6 x 15KW DLC shelves

2 x Leaf Eth switches

up to 10 IB/BXI/High Speed Ethernet switches

4 to 20 compute blades (front)

up to 12 compute blades (Back)

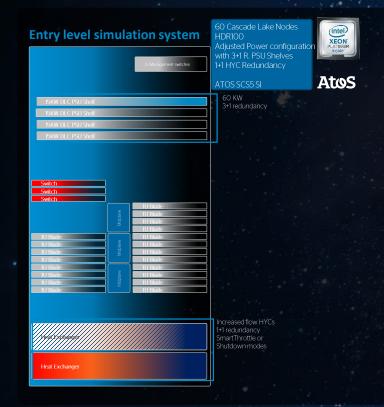
up to 3 Hydraulic chassis

**42U Cabinet Rear View** 



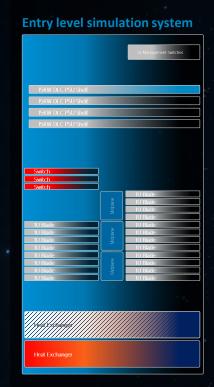


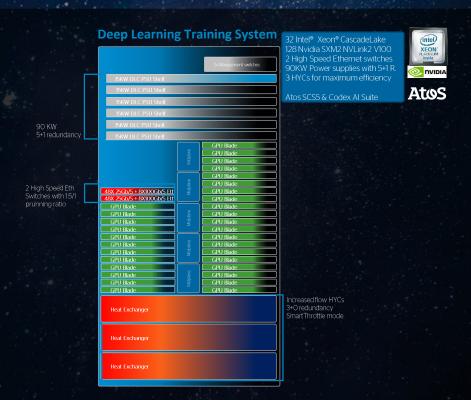
Unlimited solutions in real world





Unlimited solutions in real world

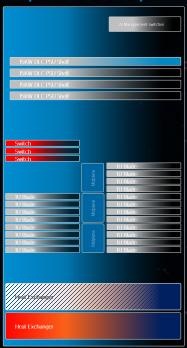






Unlimited solutions in real world

### **Entry level simulation system**



### **Deep Learning Training System**

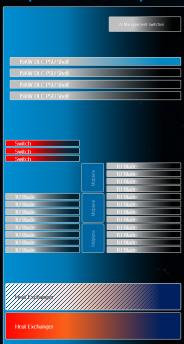


# Al augmented HPC System **AMD** NVIDIA 🌅 5 HDR 200 Gb/S switches 90KW Power supplies with 5+1 R. **Atos** HDR Switch GPU Blade 2+1 redundancy SmartThrottle mode

Atos

Unlimited solutions in real world

### **Entry level simulation system**



### **Deep Learning Training System**



### Al augmented HPC System

		2x Management switches
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### 192 AMD EPYC Rome CPU 4 HDR 100 Gb/S switches 90KW Power supplies with 5+1 R.

Atos SCS5
Atos Smart Data Management Suite AtoS

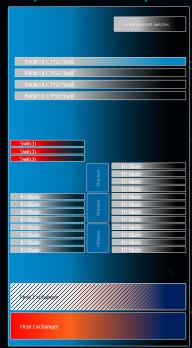
### **Memory Bound Apps Sim. System**

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Unlimited solutions in real world

### **Entry level simulation system**



### **Deep Learning training System**



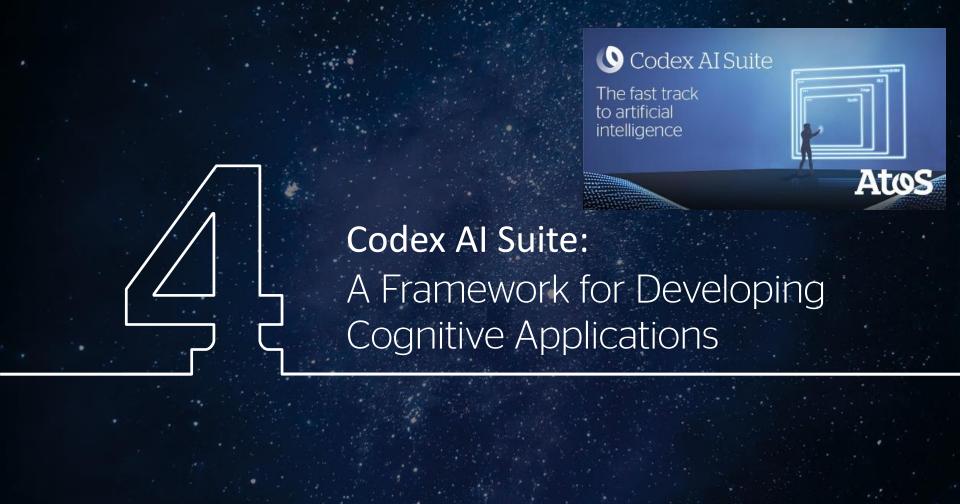
### Al augmented HPC system

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### **Memory bound simulation**

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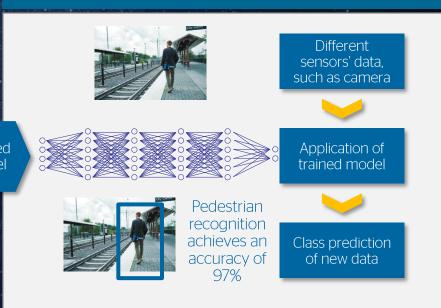
### Codex Al Suite tackles the large-scale Deep Learning Problem

New requirements, but different from training to running











# Main Requirements for a Deep Learning Self Service model

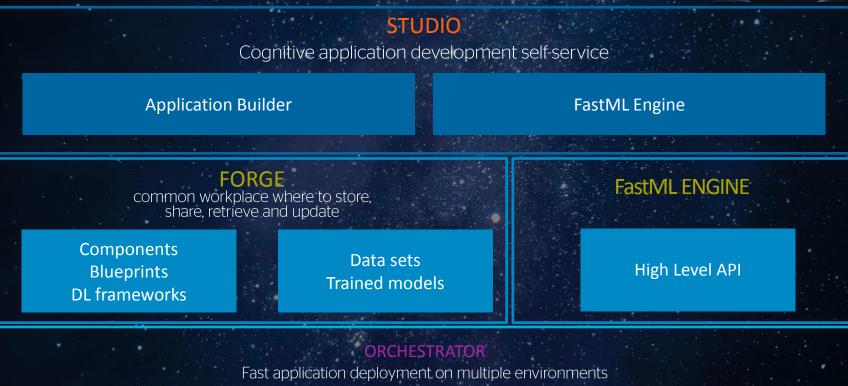
Training Management Models Management Dataset Management

Inference Management Deep Learning Framework Management Optimised Resource Management



# Codex Al Suite at a glance





Atos

# The Codex Al Suite leverages our partners to accelerate Al solutions

Codex AI Studio is the common workbench to develop use cases

Deep Learning engine Google ML services







Forge



Orchestrator





Partners (components, algorithms, applications, models)





Enterprise, Edge, HPC, On-premise, Cloud

# 3 key Codex Al Suite benefits



01

Increases data scientist productivity

02

Provides a complete cognitive eco-system ready to work

03

Delivers optimal performance at low cost

- Day-to-day mundane tasks are performed by Codex Al Suite
- Data scientists leverage the use case experience
- Use case policy management

- Customers benefit from an integrated & cost-effective solution
- Applications take immediate benefit of latest technologies with no additional investment

- High-performance resources are allocated only when need be
- Hybrid cloud resources are allocated whenever possible to share resources.



# HPCaaS architecture

With the XCS web GUI



\*DC = Dashboard Component



HPC

Data base

Directory

service

# HPCaaS and DLaaS architecture

With the XCS web GUI



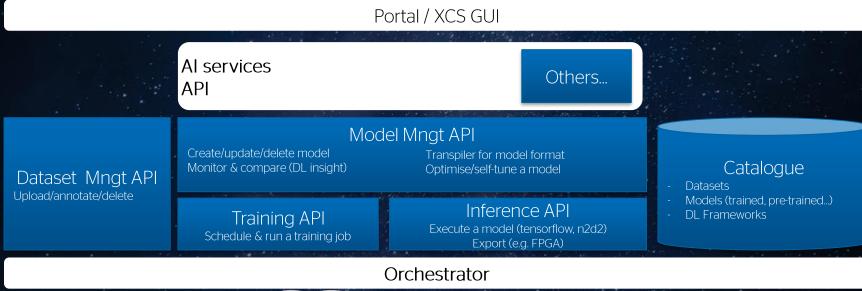
HPC HPC Batch schedulers HPC **REST API** DCs web server HPC Data base HTTPS XCS GUI web server Authorization Directory Dashboards OAuth service • Web Design DLSS Data base DLSS DLSS HTTPS DLSS DCs **REST API** SSH web server DL frameworks DL datasets

\*DC = Dashboard Component

Bringing the two threads together

Bull Extreme Fcatory

Sodex Al Suite











# Summary

- 1. To continue drive scientific endeavour through simulation and modelling, increasing parallelism and the use of Al are inevitable
- 2. The BullSequana XH2000: a new hybrid architecture for HPC and Al which combines leading technologies in a highly dense water cooled system, delivering market leading TCO
- 3. Codex Al Suite: a new open framework for developing cognitive applications which couples HPC and Al for on-premise or cloud deployments
- **4. BullExtreme Factory:** A 3<sup>rd</sup> generation portal for HPC and Al application deployment within a hybrid cloud services model
- 5. Visit our booth (24) to find out more



# Thanks to our partner Intel



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For more information please contact:

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## **Codex AI Suite differentiators**



End-to-end, fast development of complex enterprise use cases

Enable to easily combine ML, DL, analytics and data management to deliver complex and accurate use cases



Forge



Applications are infrastructure-agnostic

Enable applications to run on HPC, enterprise and Edge servers, on clouds and on-premises



Orchestrator



**HPC** business can deliver NG applications

From precision medicine, to imaging diagnosis, driver assist, autonomous maintenance, ..



ML Engine







# Extreme Factory UK – working with STFC Hartree Centre

Public HPCaaS offering to UK academics and Industry





# JADE – working with Oxford University and The Hartree Centre

Public DLaaS offering to UK academics and Industry



### Key scientific areas include:

- National Deep Learning Service
- GPU enabled Computing
- DL enabled HPC application development
- Prototyping and Algorithm Development

### **Atos Solution:**

- BullSequana X1000 supercomputer
  - ~4 Pflop/s Intel® Xeon® Scalable processors
     & Intel® Xeon Phi™, GPUs
- Extreme Factory (XCS and XRV)
  - Professional services to tailor the interface
- Business Development resources to co-sell

