



# BDEC2: the digital continuum

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# What is BDEC?

- An international “think-tank” started by **HPC people** (circa 2009) to address **exascale** convergence of compute and (big) data.
- We met regularly (2-3 times per year) in intensive **3-day workshops**.
- We prepare **position papers** and roadmaps that are submitted to major national funding agencies and that guide investment strategies.
- Major **reports** published:
  - IESP Roadmap, IHPCA, 25(1), 2011.
  - Pathways to Convergence, IHPCA, 32(4), 2018.
- Follow us: [www.exascale.org/bdec](http://www.exascale.org/bdec)



# 10<sup>18</sup>

# BDEC2

## BIG DATA *AND* EXTREME-SCALE COMPUTING<sup>2</sup>



THE INTERNATIONAL JOURNAL of  
**HIGH  
PERFORMANCE  
COMPUTING  
APPLICATIONS**

## The International Exascale Software Project roadmap

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### Abstract

Over the last 20 years, the open-source community has provided more and more software on which the world's high-performance computing systems depend for performance and productivity. The community has invested millions of dollars and years of effort to build key components. However, although the investments in these separate software elements have been tremendously valuable, a great deal of productivity has also been lost because of the lack of planning, coordination, and key integration of technologies necessary to make them work together smoothly within individual petascale systems and between different systems. It seems clear that this development model will not provide the software needed to support the unprecedented parallel exascale computation on millions of cores, or the flexibility required to exploit new hardware mechanisms such as transactional memory, speculative execution, and graphics processing units. This report describes the community's plan to prepare for the challenges of exascale computing, ultimately combining their efforts into the national Exascale Software Project.



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BDEC2

BIG DATA *AND*  
EXTREME-SCALE  
COMPUTING<sup>2</sup>

**BIG DATA AND EXTREME-SCALE  
COMPUTING: PATHWAYS TO  
CONVERGENCE. Toward a Shaping  
Strategy for a Future Software and Data  
Ecosystem for Scientific Inquiry\***

M. Asch, T. Moore, M. Asch, R. Badia, M. Beck, P. Beckman, T. Bidot, F. Bodin, F. Cappello, A. Choudhary, B. de Supinski, E. Deelman, J. Dongarra, A. Dubey, G. Fox, H. Fu, S. Girona, W. Gropp, M. Heroux, Y. Ishikawa, K. Keahey, D. Keyes, W. Kramer, J.-F. Lavignon, Y. Lu, S. Matsuoka, B. Mohr, D. Reed, S. Requena, J. Saltz, T. Schulthess, R. Stevens, M. Swamy, A. Szalay, W. Tang, G. Varoquaux, J.-P. Vilotte, R. Wisniewski, Z. Xu and I. Zacharov

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# In the beginning...

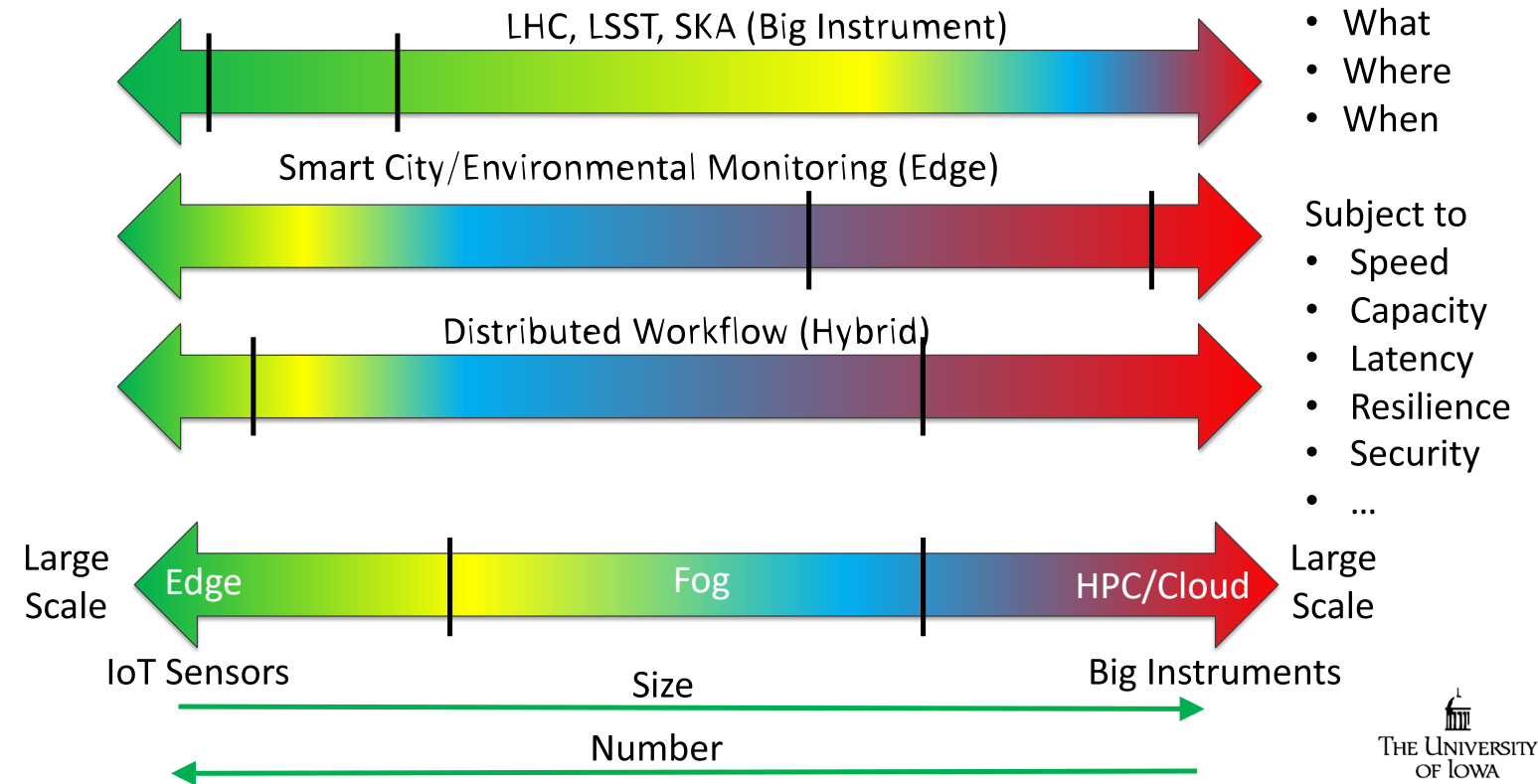
- Peta- to Exascale transition
- Big Data and IoT (the end of HPC?)
- Convergence of HPC and Big Data, but what about logistics?
- TransContinuum e-infrastructure:
  - Edge-to-edge, from IoT and Big Instruments through to the Centre (cloud, HPC)
  - Data everywhere
  - Compute near the data
  - Workflows
  - **AI everywhere**... the new (unavoidable) enabler!




# So, what is the Digital Continuum?



## Building fluid cyberinfrastructure



# Challenge: programming the Digital Continuum?



Size	Nano	Micro	Milli	Server	Fog	Campus	Facility
Example	IoT	Smart Device	Sage Node	Linux Box	Co-located Blades	1000-node cluster	Datacenter
Memory	0.5K	256K	8GB	32GB	256G	32TB	16PB
Network	BLE	WiFi/LTE	WiFi/LTE	1 GigE	10GigE	40GigE	N*100GigE
Cost	\$5	\$30	\$600	\$3K	\$50K	\$2M	\$1000M

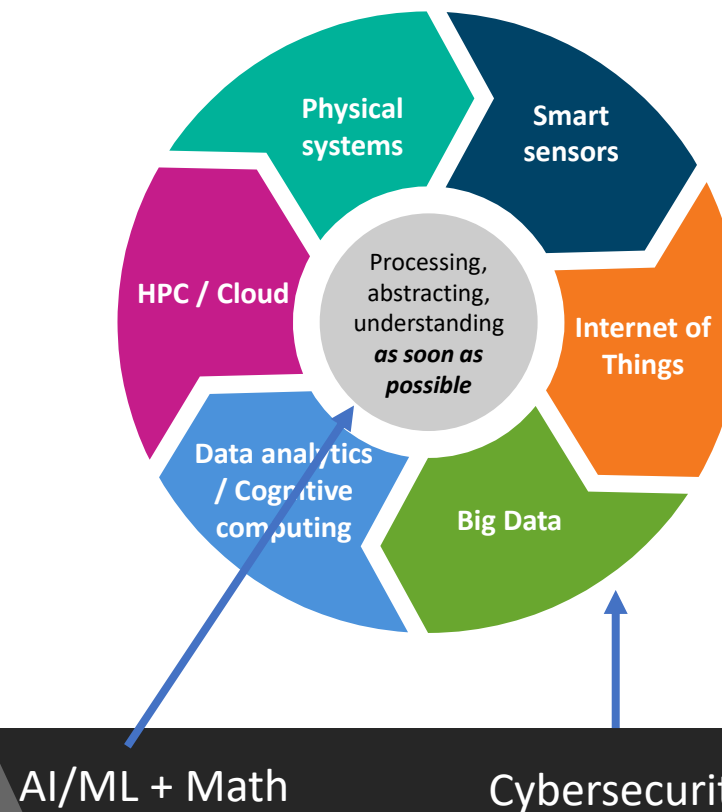
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# So, what is the Digital Continuum?



## HPC in the loop



Enabling Intelligent data processing at the edge:

- Fog computing
- Edge computing
- Stream analytics

Transforming data into information as soon as possible

Collaboration between edge devices and the HPC/cloud ensuring:

- Data security and Privacy
  - Lower bandwidth
  - Better use of HPC/Cloud
- creating a continuous flow



# BDEC2 Demonstrators

- **Definition:** a *proof-of-concept* platform designed to demonstrate some **common** capabilities that some of our BDEC2 applications and application communities need.
- **Objective:** produce a working version of an international, federated, **continuum-spanning demonstrator** that can be cooperatively operated and managed and that engages stakeholders at all levels.
- **Why?** Existing cyberinfrastructure was not designed to adequately deal with edge to cloud/HPC workflows, especially not extremely data intensive ones.
- **How?** A series of international *WORKshops*, bringing together computer scientists, application scientists, big data, IoT, AI and other stakeholders who are focused on achieving this goal.
- Follow us: [www.exascale.org/bdec](http://www.exascale.org/bdec)



# What are the Challenges?

- There is an **end-to-end** problem (spanning the continuum) - from AI@Edge to HPC in the Cloud.
- There is a **software stack** problem (HPC troglodytes).
- There is a **resource allocation** problem (on demand, shared infrastructure).
- There is a data movement and **logistics** problem (both directions).
- Robustness, security, sustainability and reliability of large, interlinked, **composed** infrastructures.
- **AI is everywhere**, and new infrastructures must support monitoring and control; infrastructure learns (not just the app).



# What is a good Demonstrator?

- Could **evolve** to support multiple application domains.
- Reveals **programming model** from edge to cloud.
- Shows **global workflow** (data, resources, users, etc.)
- Architecture is **reusable**, across multiple scales.
- Could evolve to run across several different **composed infrastructures**.



# What are the potential use-cases?



- **Big Instruments:**
  - Radio telescopes – LOFAR, SKA.
  - High energy physics – LHC .
  - Satellite data – Copernicus, SWOT, HIMAWARI, ...
  - Climate, earth sciences, oceanography.
- **IoT-like:**
  - Personalized medicine.
  - Autonomous vehicles.
  - Predictive maintenance.
  - Precision agriculture.
- **Digital Twins...**

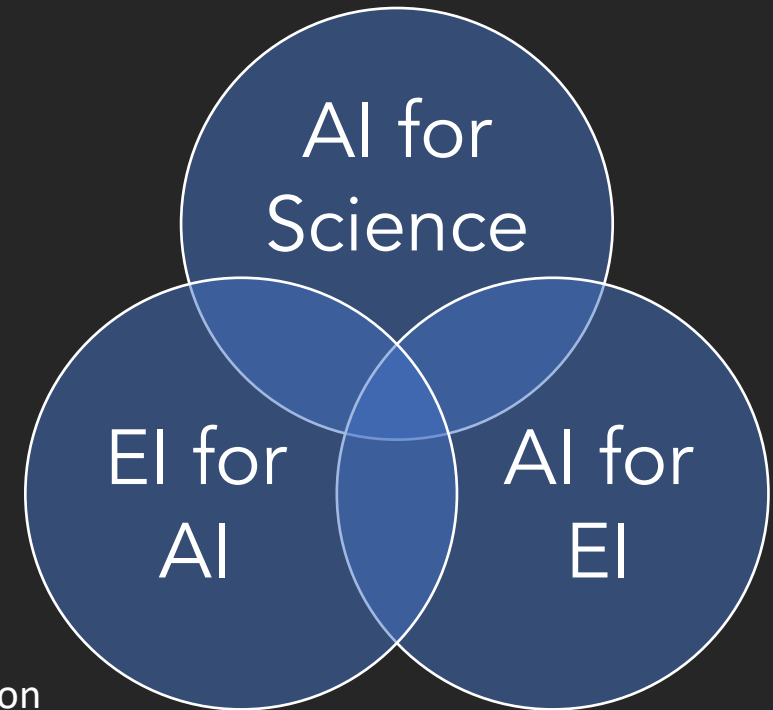
# What are the next steps?



- Two demonstrators to be developed:
  - Global Data Logistics Network.
  - Learning everywhere.
- Multi-lateral funding:
  - Set up an international funding scheme for the above 2 demonstrators.
  - Solicit funding agencies.
  - Encourage private sector involvement (GAFAM, ABC, etc.).
- Strategic Research Agenda for EU (coordinated calls):
  - ETP4HPC
  - BDVA
  - IOTI
  - 5G, ECSO, Robotics

# AI in the Continuum

- 3 categories:
  - AI for Science – applications
  - AI for EI – piloting e-infrastructure
  - EI for AI – making AI efficient
- AI for Science:
  - **Steering** of simulations
  - **Embedding** ML in simulation methods
  - Customized computational **kernels**
  - **Tuning** applications parameters
  - **Generative** models to compare with simulation
  - Student (AI) Teacher (Sim) models -> **learned functions**
  - **Guided search** through parameter spaces
  - **Hybrid** architectures HPC + Neuromorphic





## In Ten Years...

- **Learned Models Begin to Replace Data**
  - queryable, portable, pluggable, chainable, secure
- **Experimental Discovery Processes Dramatically Refactored**
  - models replace experiments, experiments improve models
- **Many Questions Pursued Semi-Autonomously at Scale**
  - searching for materials, molecules and pathways, new physics
- **Simulation and AI Approaches Merge**
  - deep integration of ML, numerical simulation and UQ
- **Theory Becomes Data for Next Generation AI**
  - AI begins to contribute to advancing theory
- **AI Becomes Common Part of Scientific Laboratory Activities**
  - Infuses scientific, engineering and operations

## ARTIFICIAL INTELLIGENCE

# DOE readies multibillion- dollar AI push

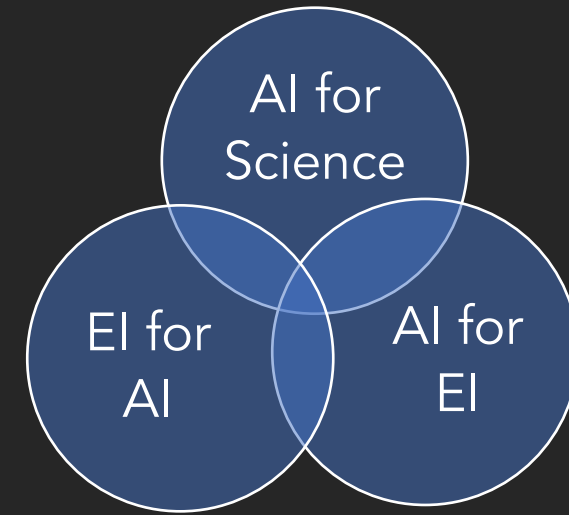
U.S. supercomputing leader  
is the latest big backer  
in a globally crowded field

By **Robert F. Service**, in Washington, D.C.

**T**he U.S. Department of Energy (DOE) is planning a major initiative to use artificial intelligence (AI) to speed up scientific discoveries. At a meeting here last week, DOE officials said they will likely ask Congress for between \$3 billion and \$4 billion over 10 years, roughly the amount the agency is spending to build next-generation “scale” supercomputers.



# AI in the Continuum



- AI for EI:

- Manage AI **expectations** (users in the loop)
- Improve system **operation**: cost, reliability, security
- Improve app and workflow **performance**
- **Closed-loop systems**
- **Test-beds**

- EI for AI

- **Programming**: tools to be used across the continuum, performance in the continuum context
- **Distributed service composition**: data placement, accomplish complex AI workflows in distributed, unreliable environment
- **Data**: retain provenance, enforce access obligations, purpose-driven storage
- **Communications** and Protocols: enable communication in this “more Internet than the Internet” environment
- **Authentication and Authorization** : establish a chain of trust





# BDEC Final Community Report

- Consensus

- Data volumes are increasing exponentially
- Data movement is the bottleneck
- AI will be transformational
- The Internet is ossified
- There is no common platform for data

- Recommendations: (made to EC)

- Design a **data ecosystem** based on international standards, for the security and transfer of data from devices to computing machines.
- Conceive and develop transcontinuum **workflow-enabling** software stacks.
- Investigate how **machine learning** can play a role at all stages of the digital continuum, all the way from the edge to the centre.
- Adapt HPC centres to this new, **data-centric science** and the machine learning algorithms on which it relies.
- Build, together, a comprehensive shared, **sustainable e-science e-infrastructure** to address the major societal challenges in the UN's Sustainable Development Goals in their Agenda 2030. Not forgetting fundamental science challenges.



# Thank YOU

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- References:

- [www.exascale.org/bdec](http://www.exascale.org/bdec)

- Asch et al. Pathways to Convergence. Int J. HPC Appl. 32(4), 2018.

