

# Next generation applications

EXDCI WP3 – Applications Grand Challenges



S. Requena - PRACE / GENCI – EXDCI WP3 leader  
and all the WP3 contributors

September 7, 2017

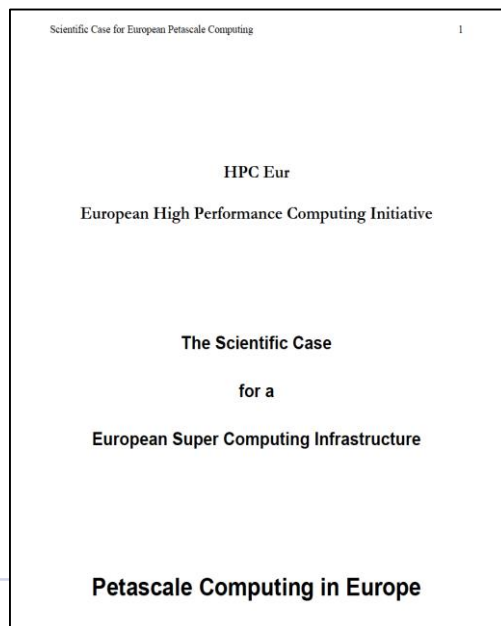


# Outline

- Introduction on EXDCI WP3
- Few words about the overall context
- Overview of some grand challenge applications
- First recommendations

# WP3 : Applications Roadmap Toward Exascale

- Objectives
  - Provide updated roadmaps of needs and expectations of scientific applications
  - Provide inputs to the update the **PRACE Scientific Case** in order to support PRACE in the deployment of its (Pre)Exascale pan European HPC research infrastructure



# WP3 structure in a nutshell (1/2)

## ● Organisation

### ● 4 working groups of 45 experts

#### ✓ T3.1 : Industrial and engineering applications

(EDF : Yvan FOURNIER, University of Aachen : Heinz PITSCH)

#### ✓ T3.2 : Weather, Climatology and Solid Earth Sciences

(Univ. Salento/CMCC : Giovanni ALOISIO, JCA Consultance: Jean-Claude ANDRE)

#### ✓ T3.3 : Fundamental Sciences

(CEA : A. Sacha BRUN, JSC: Stefan KRIEG)

#### ✓ T3.4 : Life Science & Health

(BioExcel CoE/ KTH : Rossen APOSTOLOV, CompBioMED / UCL : Peter COVENEY)

### ● T3.5: Coordination, consolidation of application roadmaps and inputs to the update of the PRACE Scientific Case

## ● Deliverables

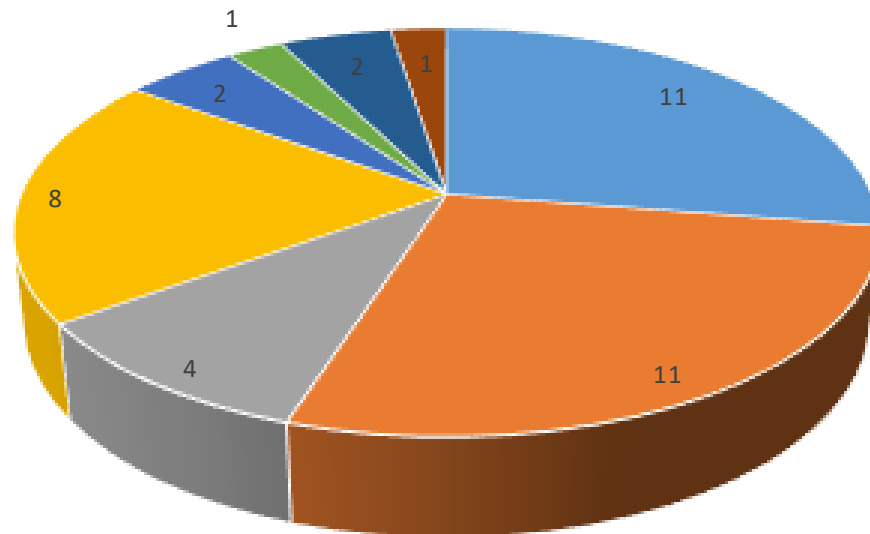
### ● D3.1: First set of reports and recommendations toward applications (**M15**)

### ● D3.2: 3<sup>rd</sup> version of the PRACE SSC Scientific Case – a full bottom-up new version of PRACE SSC Scientific Case, following the last one published in 2012 (**M27 = end 2017**)



## WP3 structure in a nutshell (2/2)

- Contribution from 45 experts of 8 countries
- Academia & industry
- Strong collaborations with the CoE :



in WG3.2



in WG3.3



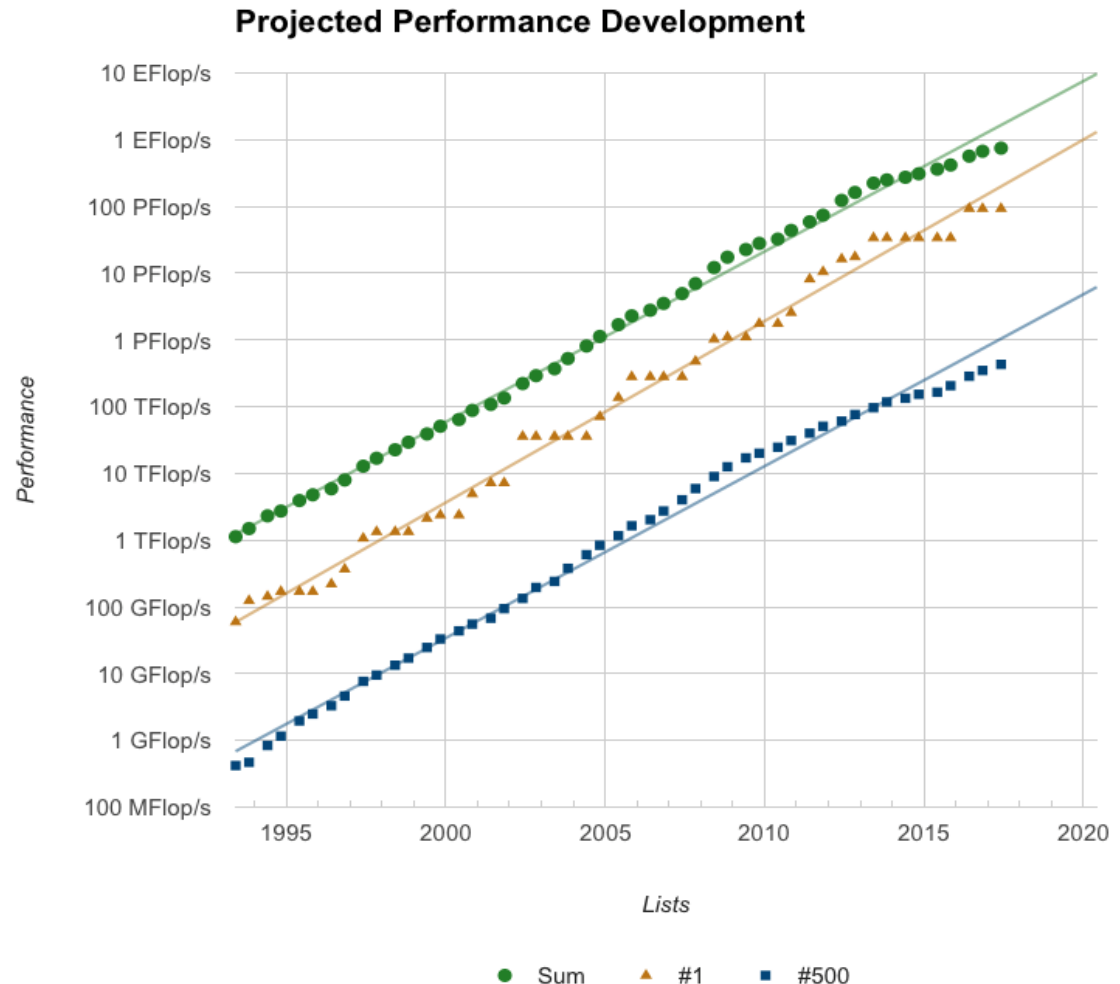
**COMPBIOMED**

leading WG3.4



# The context : the road to Exascale

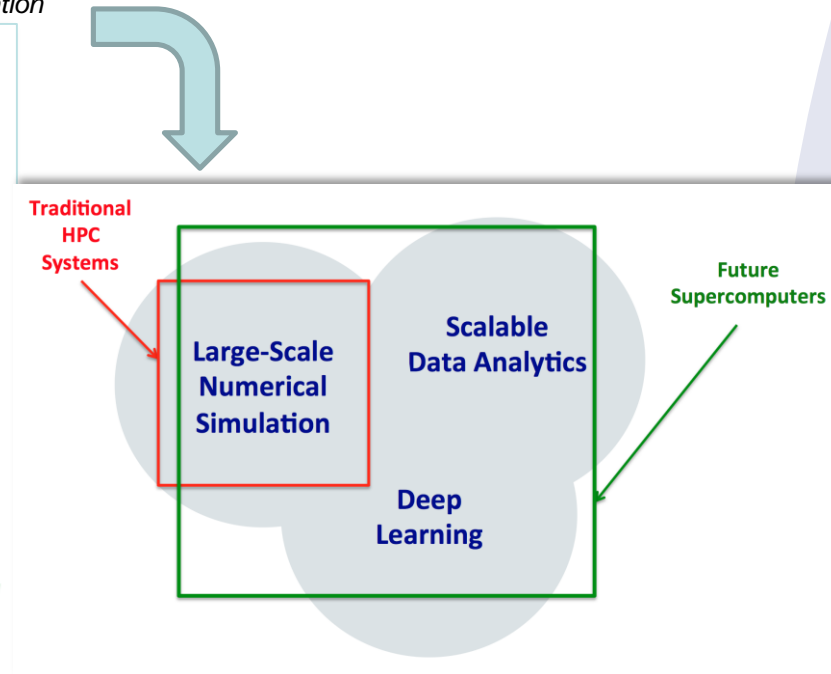
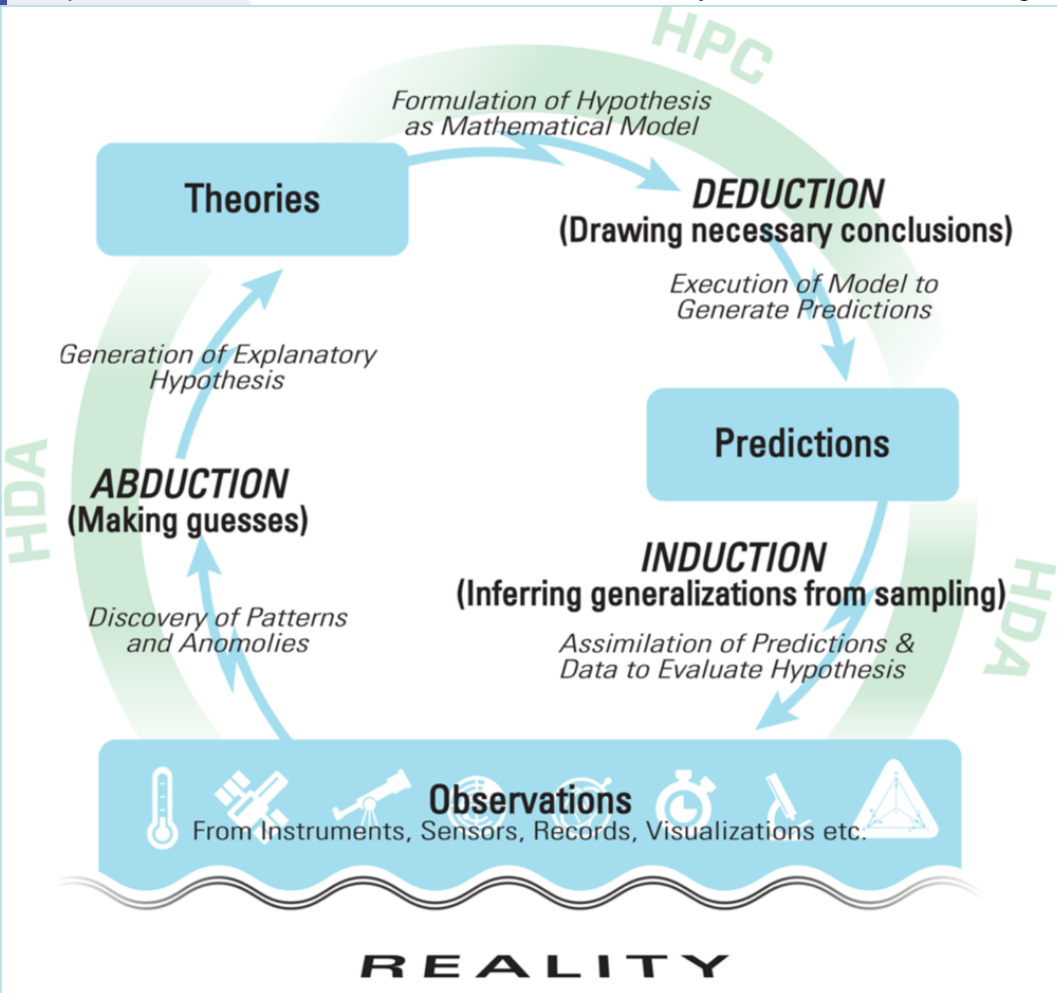
- Expected in
  - 2020 for China
  - 2021/22 for US, EU and Japan
- No more focus on peak performance
- Systems 50 to 100x faster than current ones on **real apps**



Source [top500.org](http://top500.org)

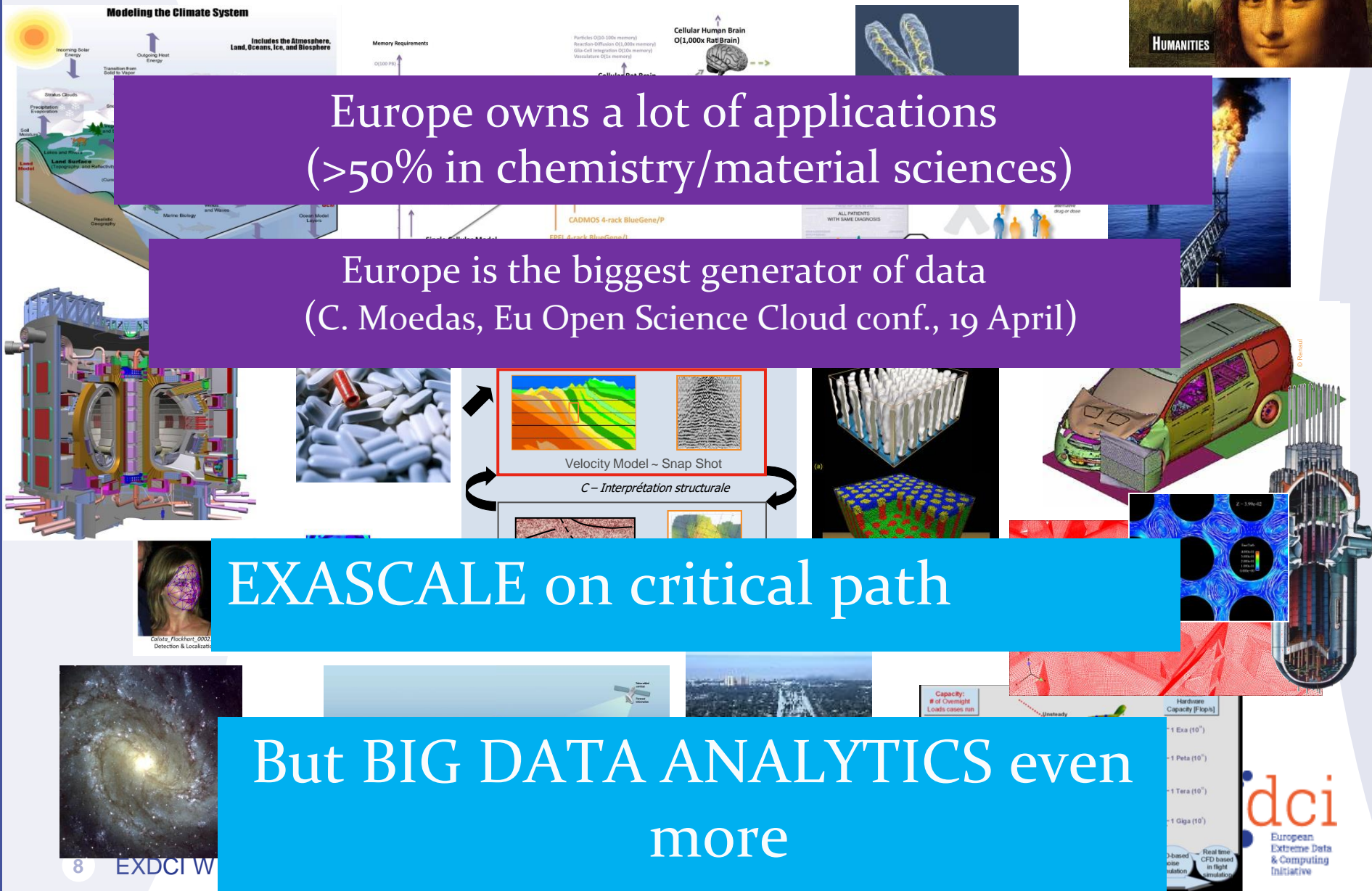
# The context : Convergence between HPC, HPDA and AI

Adapted from illustration in *Abduction and Induction: Essays on their relation and integration*



- Challenges : cohabitation of SW stacks, containers, security, smart resource managers, end to end workflows, edge computing (at the source), ...

# The context : applications are THE meaning of HPC and Big Data



Modeling the Climate System

Includes the Atmosphere, Land, Oceans, Ice, and Biosphere

Memory Requirements

Particles (10<sup>12</sup>-10<sup>14</sup> memory)  
Reaction-Diffusion (10<sup>12</sup>-10<sup>14</sup> memory)  
Cell-Cell Interactions (10<sup>12</sup>-10<sup>14</sup> memory)  
Vascularization (10<sup>12</sup>-10<sup>14</sup> memory)

Cellular Human Brain  
O(1,000x Rat Brain)

ALL PATIENTS  
WITH SAME DIAGNOSIS

Velocity Model ~ Snap Shot

C-Interprétation structurale

EXASCALE on critical path

But BIG DATA ANALYTICS even more

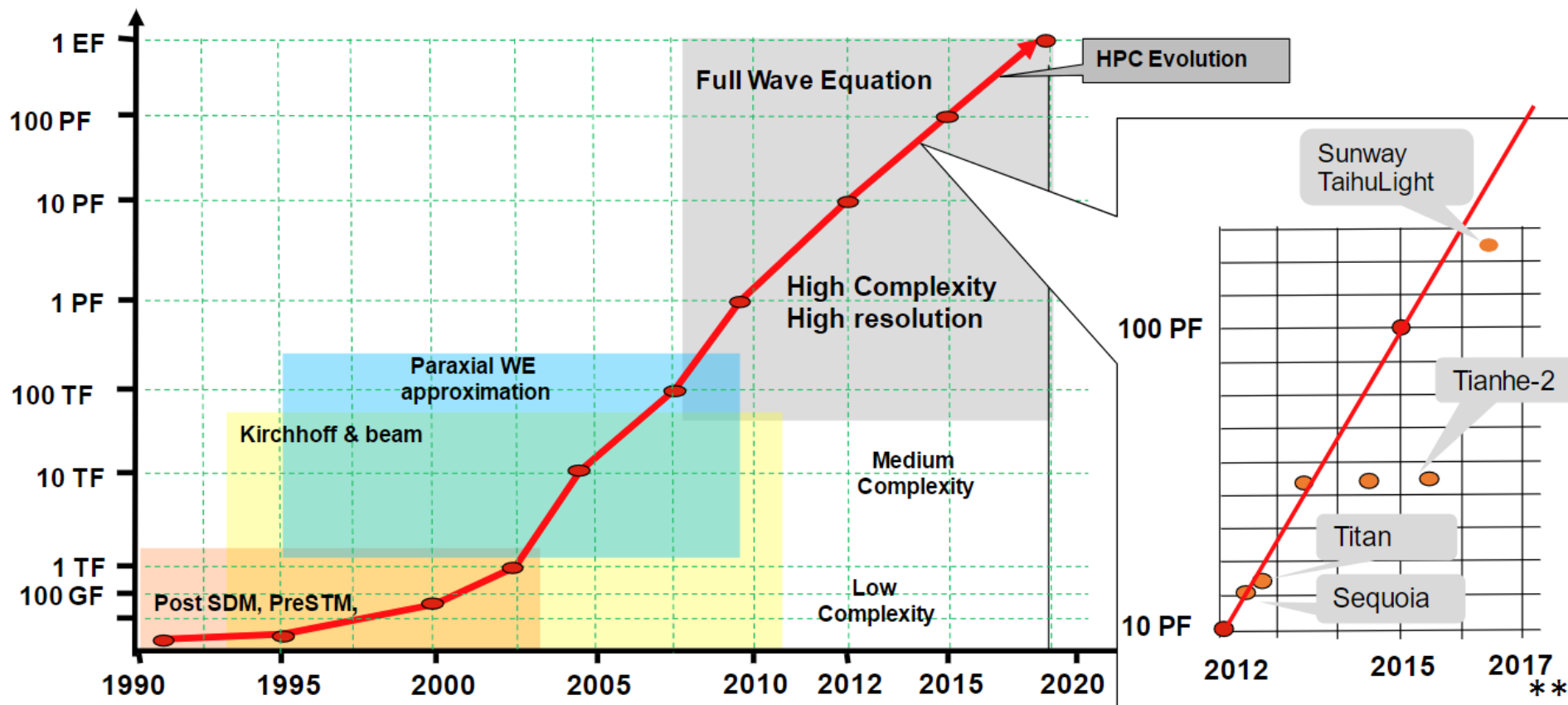
8 EXDCIW

dcii  
European  
Extreme Data  
& Computing  
Initiative



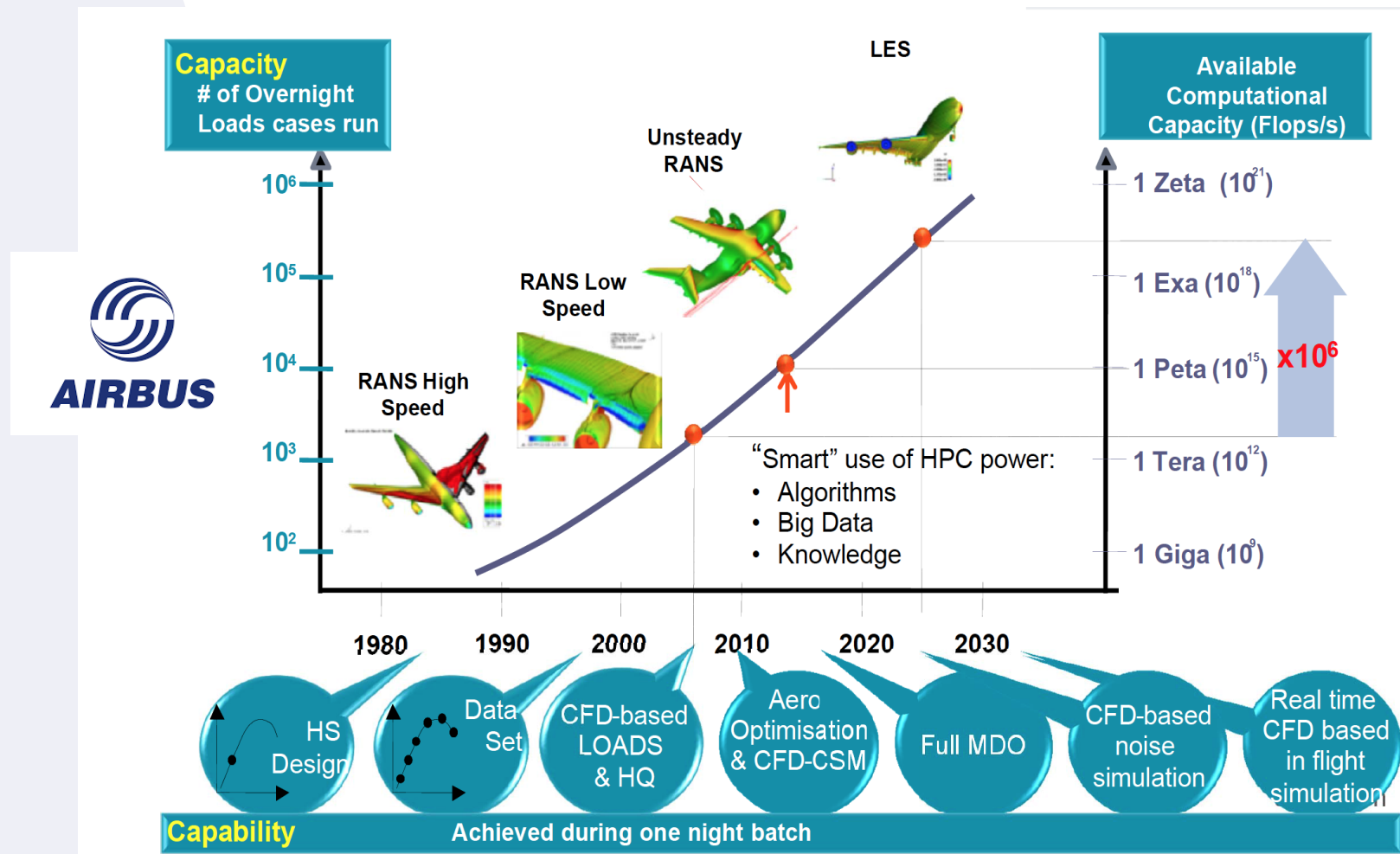
# Some Exascale applications challenges

- In the Oil & Gas : toward FWI in seismic exploration in 2020



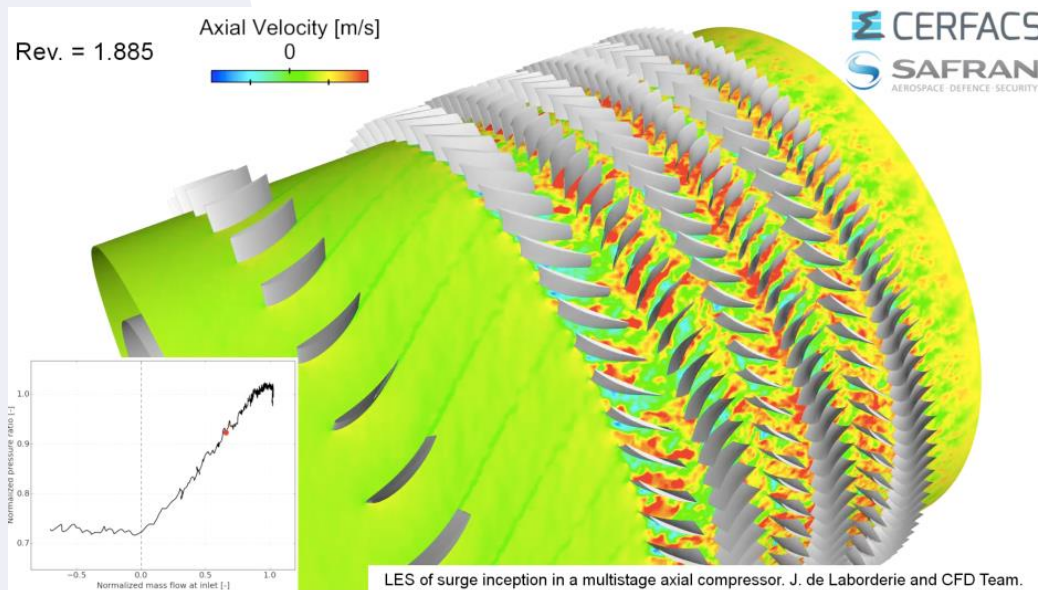
# Some Exascale applications challenges

- In aeronautics : toward a full multiphysics virtual bird in 2030 !



# Some Exascale applications challenges

- Toward a full LES simulation of a 3.5-stage gas turbine compressor
- Project « Cœur Numerique 2020 » led by Safran

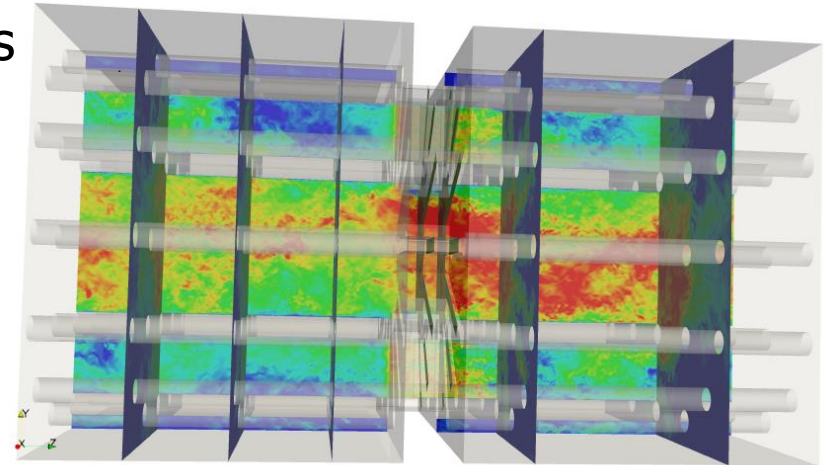


- importance of smart and scalable code couplers
- importance of MPMD on exascale (performance/intelligent load balancing)

# Some Exascale applications challenges

- EDF grand challenge in fuel assemblies

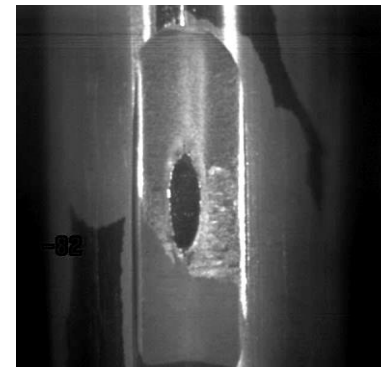
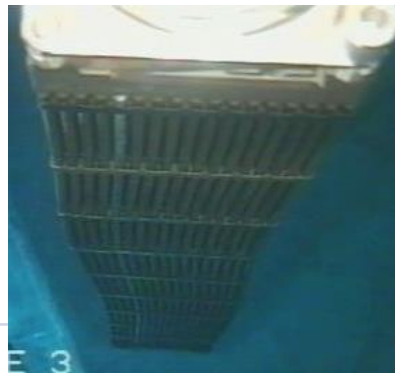
- 157 - 241 fuel assemblies in a PWR
- Each fuel assembly has 17\*17 fuel rods or guide tubes and 8 to 10 grids
  - complex, non symmetrical geometry
  - Different vendors and models
- Many constraints / stakes
  - If head loss/lift too high, stronger springs needed to keep FA down, leading to possible bowing and **deformation**
  - Good heat exchange: need mixing grids to generate turbulence
  - Low vibration: loss of cladding integrity may result from **vibration induced fretting**
- Complex geometry
  - Need for better automated **meshing tools** capable of generating very large and high quality meshes



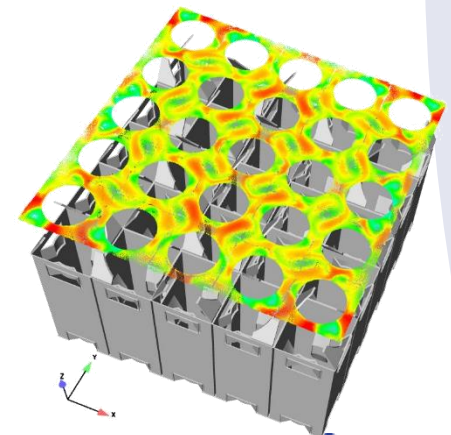
Billion-cell LES to obtain pressure loads on fuel rods;  
This is only a small subset of the full geometry



deformation



fretting-induced wear of  
cladding



EXDCI April 2017



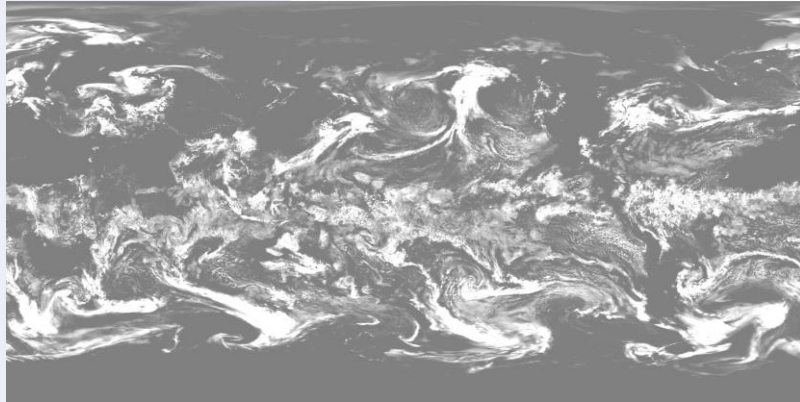
# Some Exascale applications challenges



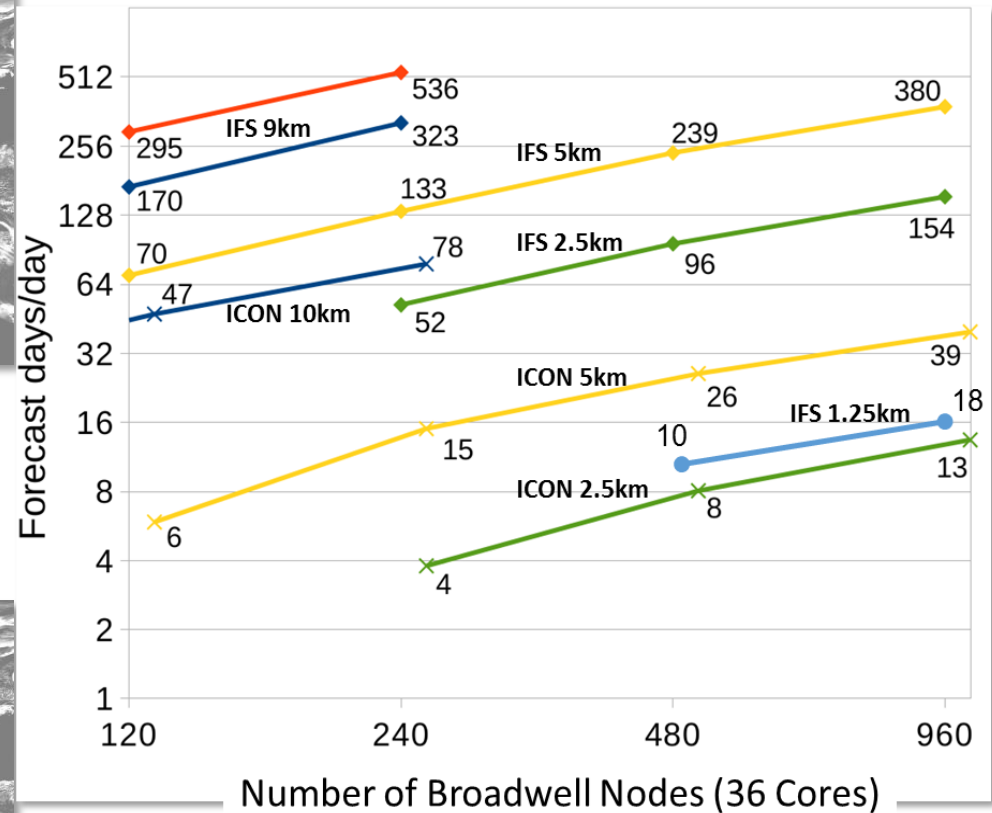
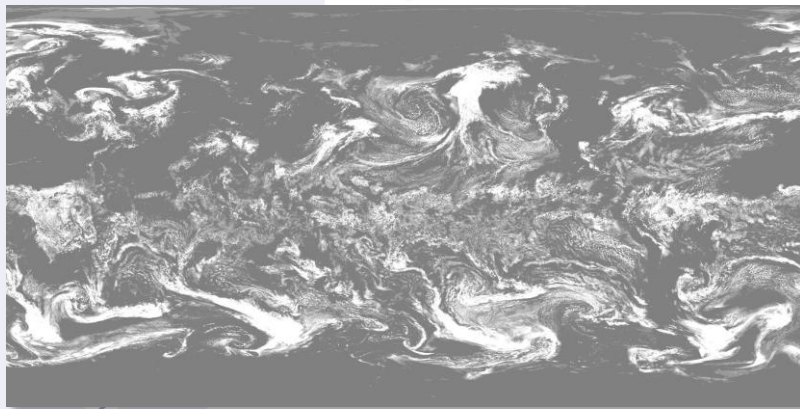
10 km

## Target for addressing key science challenges in weather & climate prediction:

Global 1-km Earth system simulations @ ~1 year / day rate

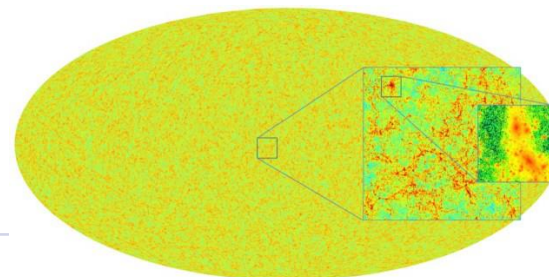
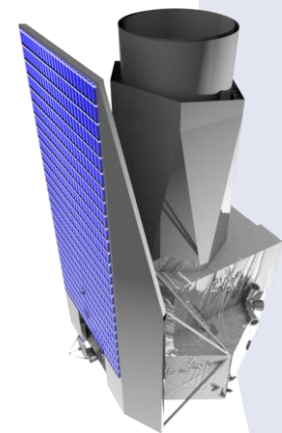
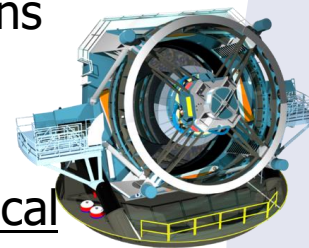


1 km



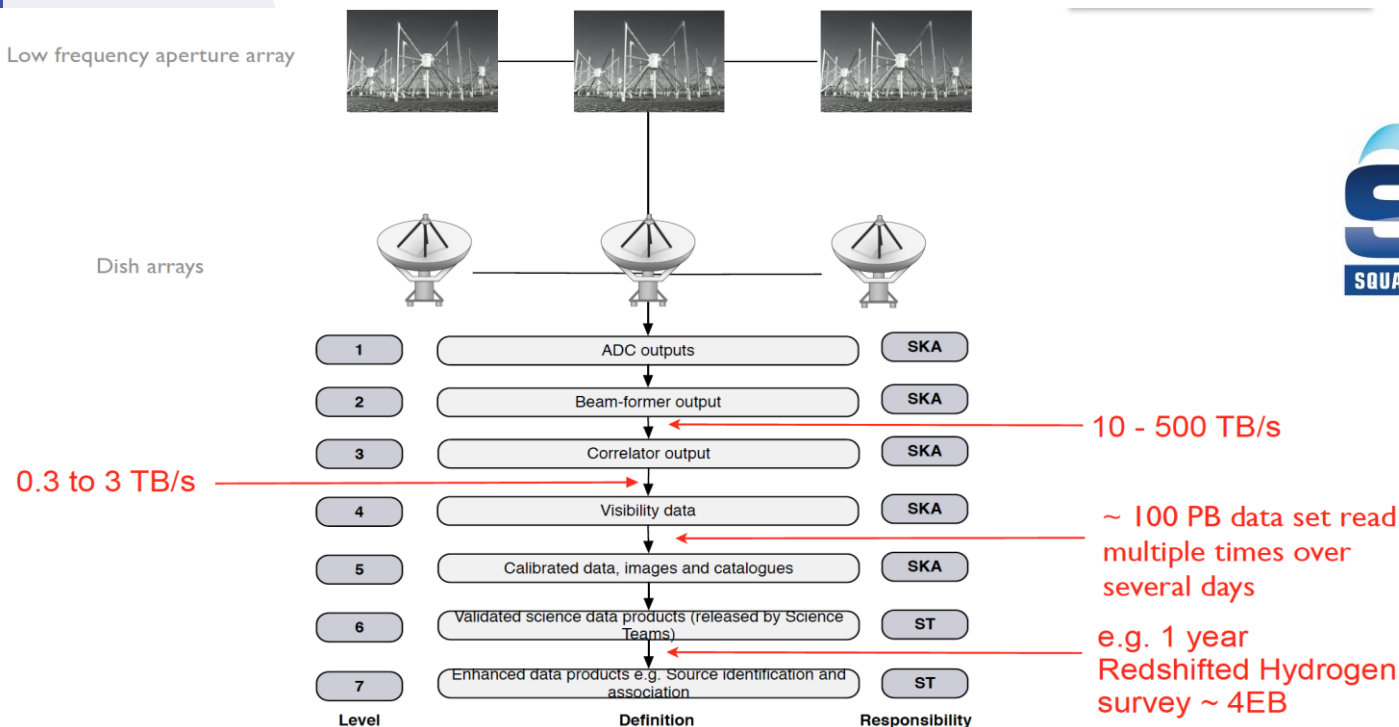
# Some Exascale applications challenges

- Massive cosmological simulations toward the next Era of Galaxy Surveys
  - Next gen satellites (Euclid, WFIRST) and telescopes (LSST) for 2020-2022
  - Ongoing massive simulations to feed / optimise future observations
  - Understanding evolution Universe wrt initial dark matter / energy
- May 2017 : Team from Univ. Zurich performed the biggest cosmological simulation with 2 billion particles using 4000+ GPU on PizDaint (CSCS)
  - PKDGRAV3 : N-Body simulation using Fast Multipole Methods (FMM)
  - Aggressive optimisations performed
    - reduced memory footprint, mix of single/double precision, ...
    - load balancing between CPU and GPU, on the fly post processing of the data
  - Excellent scaling up to 18 000 GPU
    - Benchmark performed using 8 trillion particles on Titan (ORNL/USA)



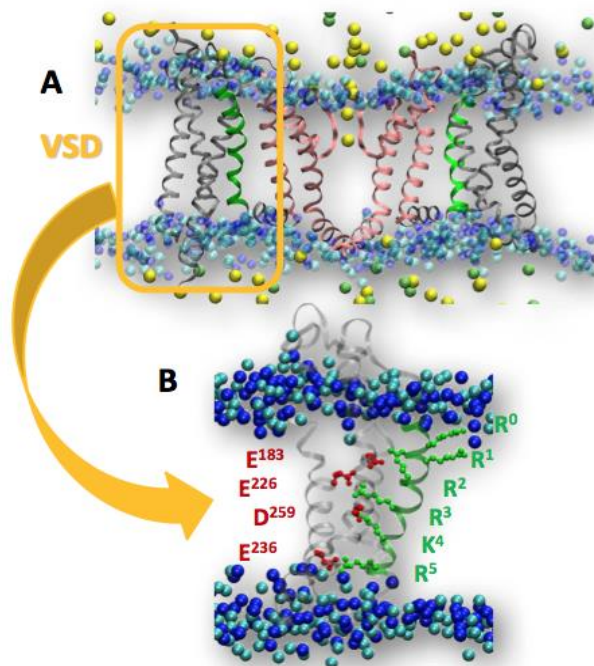
# Some Exascale applications challenges

- Postprocessing of massive amount of data issued by the SKA radiotelescope



# 2013 estimate by SKA South Africa	MeerKAT Pre-Cursor 2014-15	SKA Phase 1 2017-19	SKA Phase 2 Est. 2020-24
Data into CSP	2 Tbps	50 Tbps	up to 5 Pbps
Data into SDP	0.4 Tbps	20 Tbps	up to 500 Tbps
Into Storage	35 Gbps	300+ Gbps	up to 2 Tbps
Computing load	200 TFlops	30+ PFlops	3+ EFlops

## Molecular Dynamics on the exascale

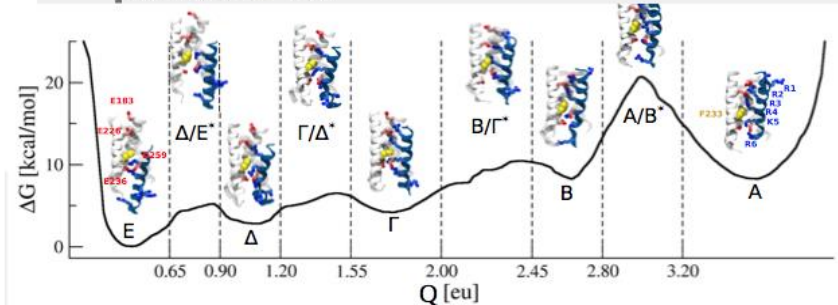


Example: ion channel in a nerve cell. Opens and closes during signalling. Affected by e.g. alcohol and drugs. 200 000 atoms

Partners



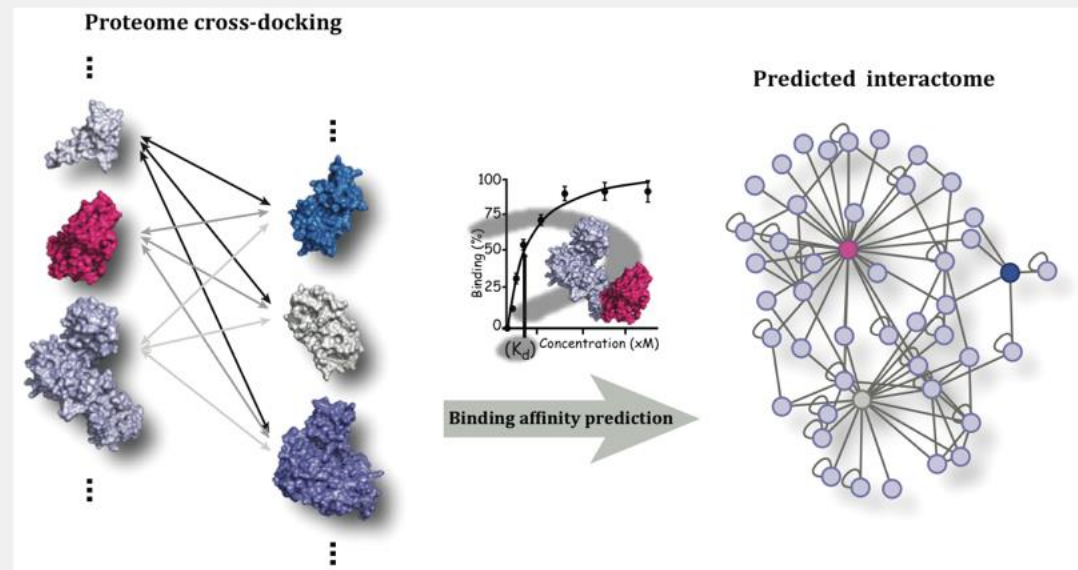
- Understanding proteins and drugs
- A 1 μs simulation: 10 exaflop
- Many structural transition: many simulations needed
- Study effect of several bound drugs
- Study effect of mutations
- All this multiplies to >> zettaflop
- Question: how far can we parallelize?



Horizon 2020  
European Union Funding  
for Research & Innovation



## Predicting interactomes by docking... a dream?



- ~20'000 human proteins
- Interactome prediction will require  $20'000^2$  docking runs
- Which will require > 10 billions CPU hours and generate about 100 exabytes of data
- Interest in simulating/understanding the impact of disease-related mutations that affect/alter the interaction network

### Partners

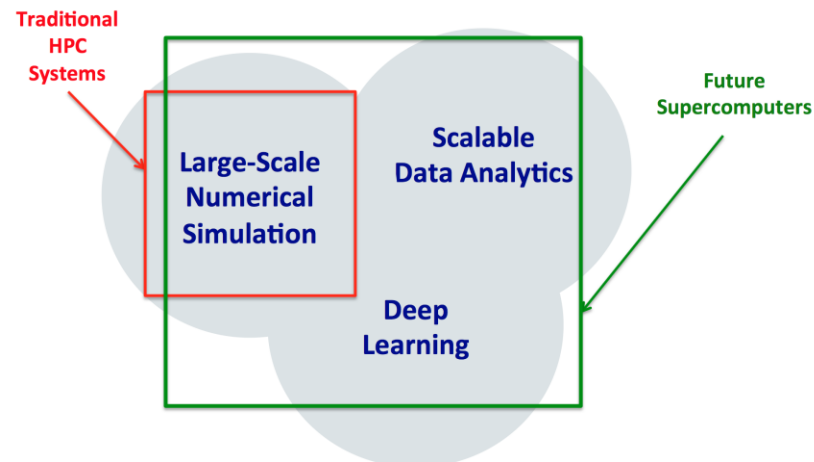


### Funding



# D3.1 – WP3 first global recommendations

- The context



- **R1 : Toward the convergence of in-situ data analysis and deep-learning methods for efficient post processing of pertinent structures among huge scientific datasets**
- **R2 : New HPC services toward improved link with large scale instruments and urgent computing**
  - Interactive supercomputing, co-scheduling, smart app based checkpoint/restart, malleability of applications, ...
- **R3 : Development at the EU level of new CoE**
  - Engineering and industrial applications
  - Opensource software industrialisation, promotion and long term support



# Toward the convergence of in-situ data analysis and deep-learning methods for efficient post processing of pertinent structures among huge scientific datasets

**S. Fiore<sup>1</sup>, G. Aloisio<sup>2</sup>, P. Ricoux<sup>3</sup>, S. Brun<sup>4</sup>, JM. Alimi<sup>5</sup>,  
M. Bode<sup>6</sup>, R. Apostolov<sup>7</sup>, S. Requena<sup>8</sup>**

*<sup>1</sup>Euro-Mediterranean Center on Climate Change Foundation, Italy and ENES*

*<sup>2</sup>University of Salento, Italy*

*<sup>3</sup>Total, France*

*<sup>4</sup>CEA, France*

*<sup>5</sup>LUTH, CNRS, OBSPM, France*

*<sup>6</sup>RWTH Aachen University, Germany*

*<sup>7</sup>KTH and BioExcel CoE, Sweden*

*<sup>8</sup>GENCI, France*

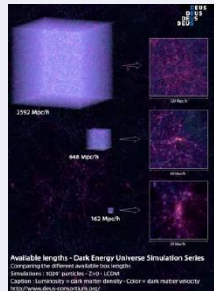
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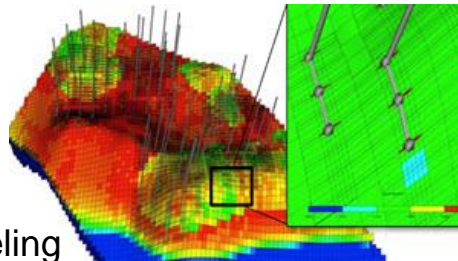
# The context (1/2) – Big Data is a today problem !

Explosion of the data from the computational side :

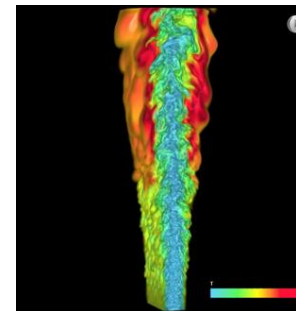
- High spatial and temporal resolutions
- Multiscale and multiphysics simulations
- Ensemble and UQ simulations



Cosmology  
DEUS project  
150 PB raw data



Reservoir modeling  
of gigamodels 350 TB/run



HiFi turbulent  
DNS combustion  
S3D : 1PB / 30mn



Ongoing convergence between HPC and HPDA platforms

- Data centric architectures
- Tiered storage
- Prg language support for data analysis, DSLs, ...

**BUT :**



**Simulation time : from days to months**



**Post processing time :  
from months to years**



# The context (2/2)



Explosion of the data from the computational side :

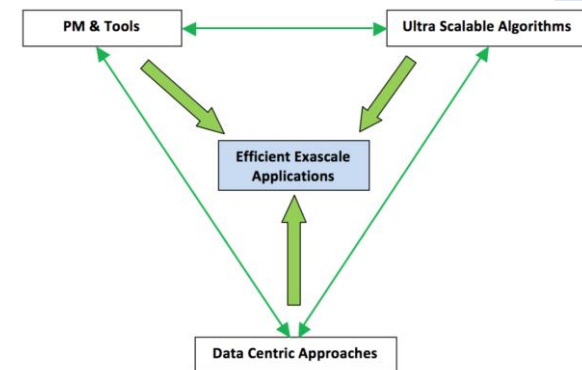
- High spatial and temporal resolutions



Necessity to reduce the time to discovery :  
amount of data impossible to process in a reasonable time for humans !  
New approaches are needed !

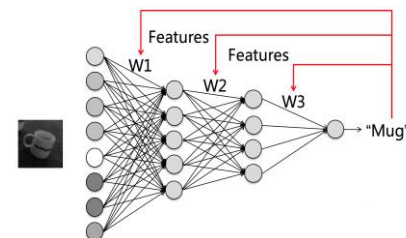
# EESI2 recommendations toward in-situ/in-transit post processing

- EESI2 FP7 project (sept 2012 – May 2015) - vision and roadmaps toward Exascale Eu applications and software tools
- 15 scientific recommendations proposed to EC  
→ « **In Situ Extreme Data Processing and better science through I/O avoidance in HPC systems** »
- Objective : Funding of appropriate R&D programs
  - Design and implement in-situ and in-transit post-processing frameworks  
→ maximize data utilisation while loaded and minimize raw data movements to remote storage units in order to save energy
  - Perform data reduction, ordering, filtering, compression, ...
  - And feature extraction through topological segmentation or trajectory based flow feature tracking, ...



# And now comes machine/deep learning

- Machine/Deep learning : the reborn of AI
  - use of HPC and neural networks for automatic (face, picture, speech, form, fraud, ...) detection, medical diagnostic, drug design,... in big data analytics
- Idea : to use ML/DL methods for improving in-situ, perform AI assisted computational steering, replace models with learned functions ...
  - smart feature extraction into massive amount of data
  - already some initial work done in Europe
  - and funding available in US : 17.7M\$ from NSF for 12 data science projects including ML/DL ones
- Strong expertise to pool, both in academia and industry, in Europe
  - France (Inria, CNRS, UPMC), Spain (BSC), Germany (Tum), UK (Alan Turing Inst.), ...
  - IBM Zurich, Sony CSL, Facebook FAIR, ...



# Proposed recommendation

- Extend current EESI2 recommendation toward in-situ/in-transit methods with machine/deep learning features
  - Assess concretely the potential of these techniques to 10 to 12 scientific pilots
  - Bridge together communities of scientific computing and ML/DL
- Organisation of a (urgent) call for proposals funded by EC or national agencies
  - Experts from domain science : climate, astrophysics, fusion, combustion, seismic, biology, chemistry, ...
  - Experts in applied maths and machine/deep learning methods
  - Experts from HPC centres
  - Target of 10 M€ for a first call, **could be scaled out to a joint BDVA**



**ETP4HPC joint project ?**





# EXDCI WP3 and SRA3 : 5yr - specific challenges per communities

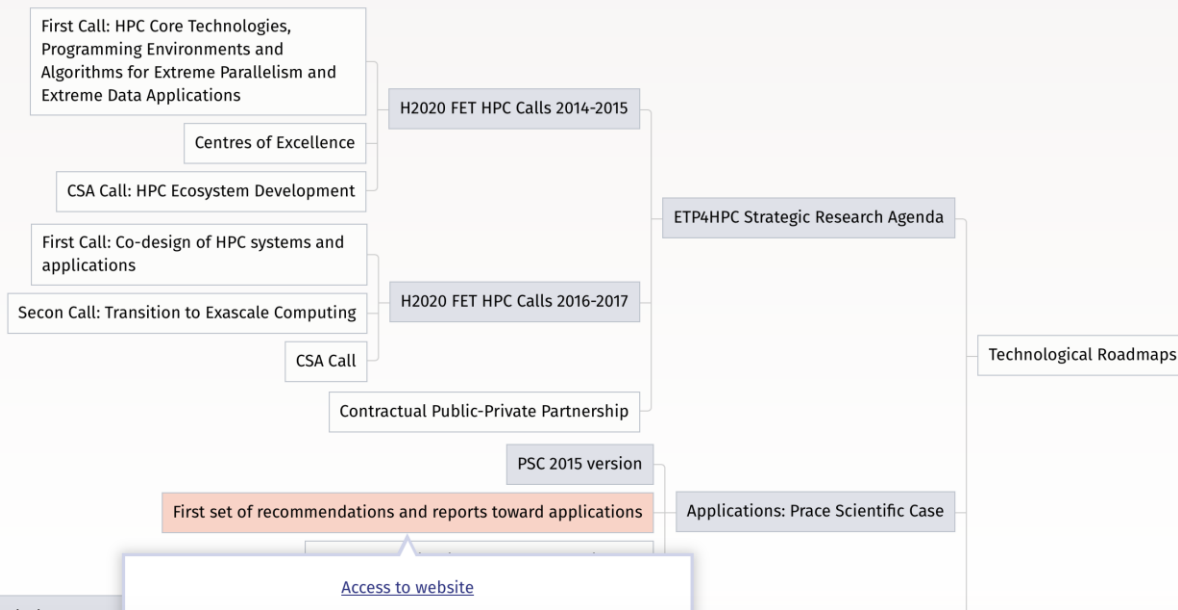
	WG3.1 Industrial & engineering apps	WG3.2 WCES	WG3.3 Fundamental Sciences	WG3.4 Life Sciences & Health
<b>HPC System Architecture and Components</b>	move to GPU or others accelerators virtualisation of resources	co design, GPU/MIC move hierarchical storage	GPU, MIC, FPGA, specialised HW ok with opt. libs, memory per core (astrophysics, fusion), mem & network BW	Large width vector units, low- latency networks, high- bandwidth and large memory; fast CPU<-> accelerator transfer rates, Heterogeneous acceleration, floating-point
<b>System Software and Management</b>	smart runtime systems, scalable couplers, data-aware schedulers	scalable couplers, smart scheduling	scalable couplers, mix of capacity/capability	urgent computing, link with instruments, co-location of compute & data, dynamic scheduling, support for workflows
<b>Programming Environment</b>	use of standards, sustainability	use of DSL, solution & performance portability	task programming / runtimes, use of DSL, standards	portability, fast code driven by python interfaces
<b>Energy and Resiliency</b>	scalable monitoring tools, data compression, application based FT	mixed precision	reduced precision, data compression	Distributed computing techniques to handle resiliency/fault tolerance
<b>Balance Compute, I/O and Storage Performance</b>	in-situ/in transit post processing	in-situ/in transit post processing active storage techniques	dynamic load balancing in mesh refinement and spectral element techniques	I/O driven, in memory database, Data-focused workflows, handling lots of small files in bioinformatics
<b>Big Data and HPC usage Models</b>	compression of data, remote viz, UQ/optimisation, ML/DL, mix of capacity/capability	UQ, compression of data, multi-site experiments to support multi-model analysis experiments, in- memory analytics, HPC- through-the cloud, ML/DL	remote data viz, ML/DL	convergence HPC-HTC, integration of HPDA tools (Hadoop, Spark, ...) inc ML/DL Cloud Computing security / privacy
<b>Mathematics and algorithms for extreme scale HPC systems</b>	ultra-scalable solvers, // in time, automatic/adaptive meshers, model order reduction , meshless and particle simulations, coupling stochastic and deterministic methods	parallelisation in time, ensemble simulations	scalable solvers, adaptive meshers, // in time, FMM and h-matrices	multiscale/physics workflows tools, ensemble simulation, model order reduction

## D3.2 – WP3 final recommendations

- Ongoing process planned for end of 2017
  - Use and development of “**standard tools**” (à la MPI+X) including domain specific libraries and languages (DSL) to abstract underlying complex HW
  - **Ultra-scalable solvers** inc. communication/synchronisation avoiding, parallelisation in time, high order discretisation, H-matrices, ...
  - Smart and scalable **meshers and code couplers** for supporting next gen multi-scale and multiphysics simulations
  - Importance of EU-wide **optimisation and UQ frameworks**
  - **Training** toward data analysis and AI
- Final deliverable **D3.2 : Update of the Scientific Case end 2017**
  - Specific and global recommendations provided
  - EXDCI WP3 to feed PRACE SSC effort

Download the D3.1 report  
[www.exdci.eu](http://www.exdci.eu)

EXDCI at a glance



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