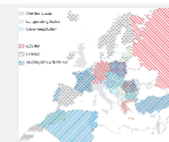


Progress report on ECMWF's Scalability Programme

Peter Bauer and the Scalability Team (RDxFDxCD, ECMWF)

Governance:

ECMWF, Member states, Regional consortia



ECMWF Scalability
Programme 1.0

Projects:

Observation processing:

- Lean workflow in critical path
- Object-based data processing
- Screening

Data assimilation:

- Flexible algorithms (C++)
- IFS integration
- Coupling with ocean and sea-ice

Numerical methods:

- Numerical methods
- h/v/t-discretization, multiple grids
- Prognostic variables

Model output processing:

- Broker-worker workflow
- Near-memory processing

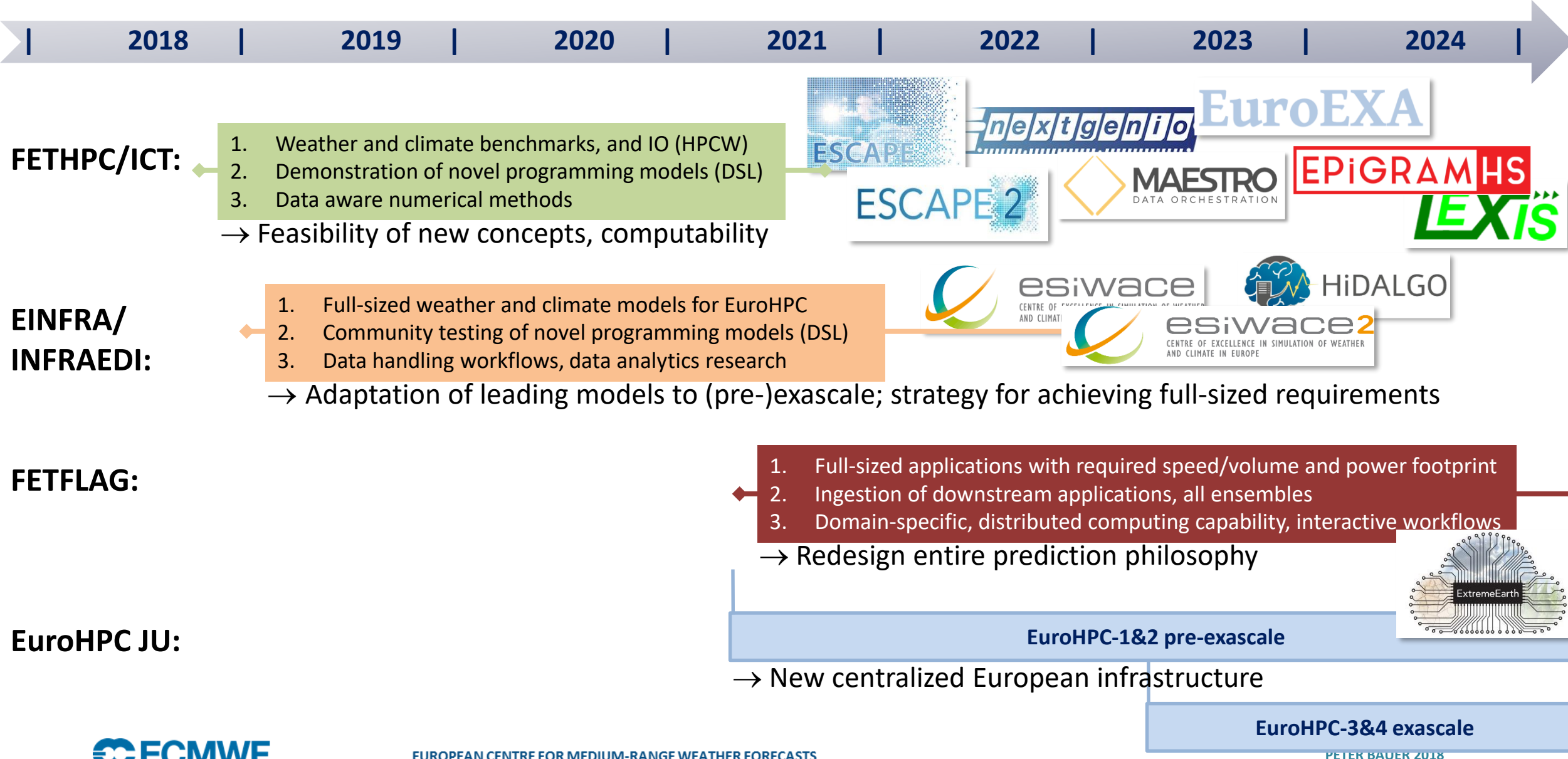
In the shorter term, implement low-hanging-fruit efficiency gains in present system to:

- Counterbalance cost of imminent science upgrades
- Trial portability/efficiency of *present* methodologies to *existing* hardware options
- Support planning (procurements w/ realistic budget requests, benchmarks, etc.)

In the longer term, test prepare and assess not-so-low-hanging fruit-efficiency gains in future system to:

- Counterbalance cost of more forward-looking science upgrade options
- Trial portability/efficiency of *future* methodologies to *future* hardware options
- Support planning (procurements w/ realistic budget requests, benchmarks, etc.)

Weather & climate computing and data roadmap in H2020





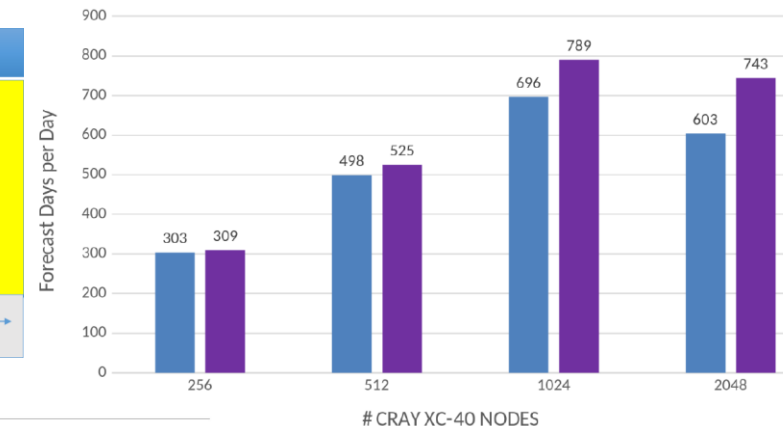
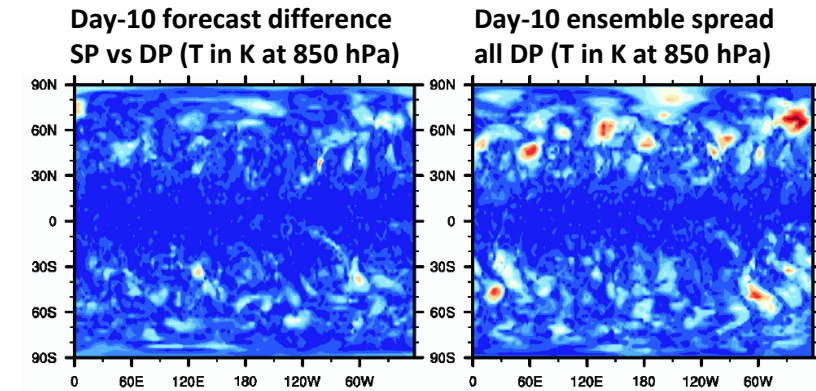
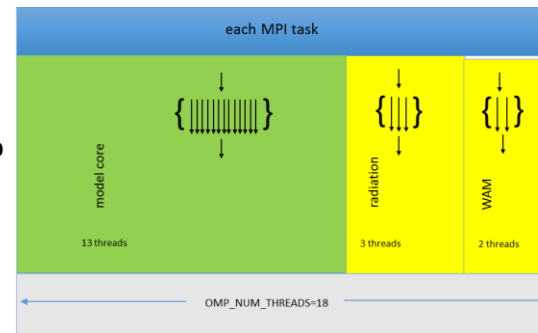
Low(ish)-hanging fruit: Computing

Precision:

- running IFS with single precision arithmetics can save 40% of runtime, IFS-ST offers options like precision by wavenumber, only for LT, in semi-implicit solver;
- storing ensemble model output at reduced precision can save 67% of data volume;

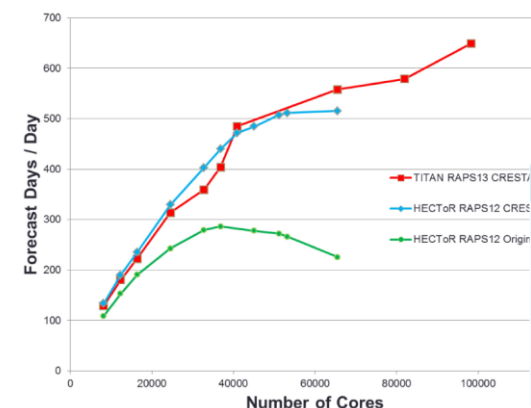
Concurrency:

- allocating threads/task (/across tasks) to model components like radiation or waves can save 20% (gain increases with resolution);
- implementation is cumbersome;



Overlapping communication & computing:

- through programming models (Fortran co-array vs GPI2 vs MPI), gave substantial gains on Titan w/Gemini,
- on XC-30/40 w/ Aries there is no overall performance benefit over default MPI implementation;



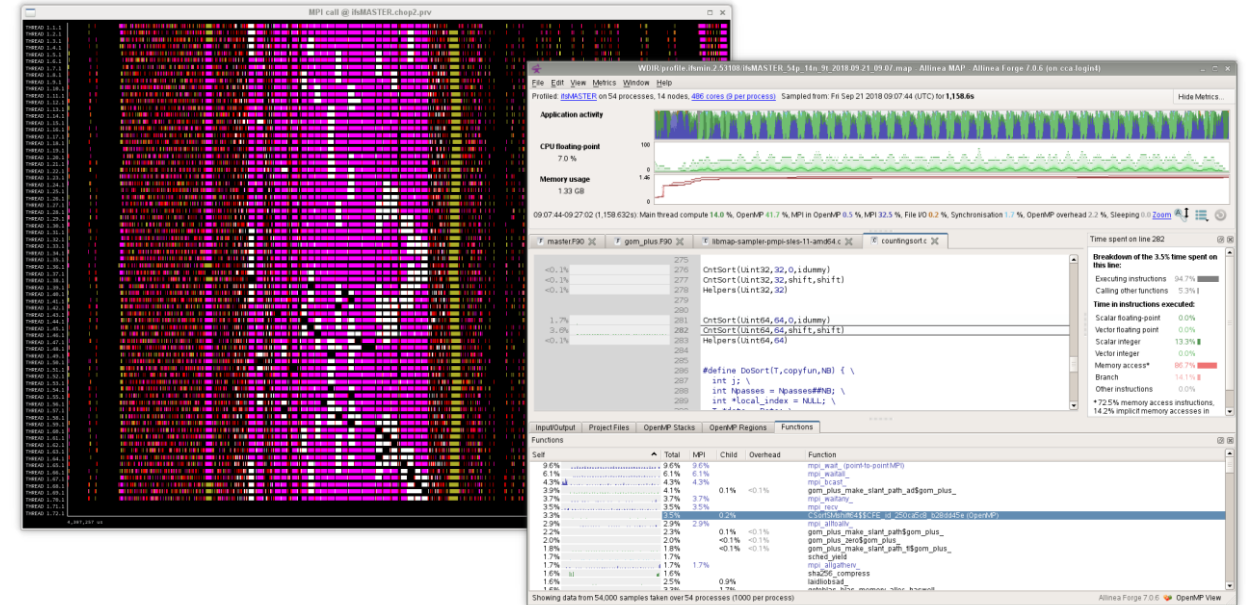
Tasks x threads	Nodes	Experiments	Forecast days/day
2160Tx12t	360	control	1104.9
2160Tx12t	360	control	1116.1
2160Tx12t	360	coarray2	815.6
2160Tx12t	360	coarray2	846.0
2160Tx12t	360	gpi2	788.9

Low(ish)-hanging fruit: Diagnostics & Architectures

Performance tools:

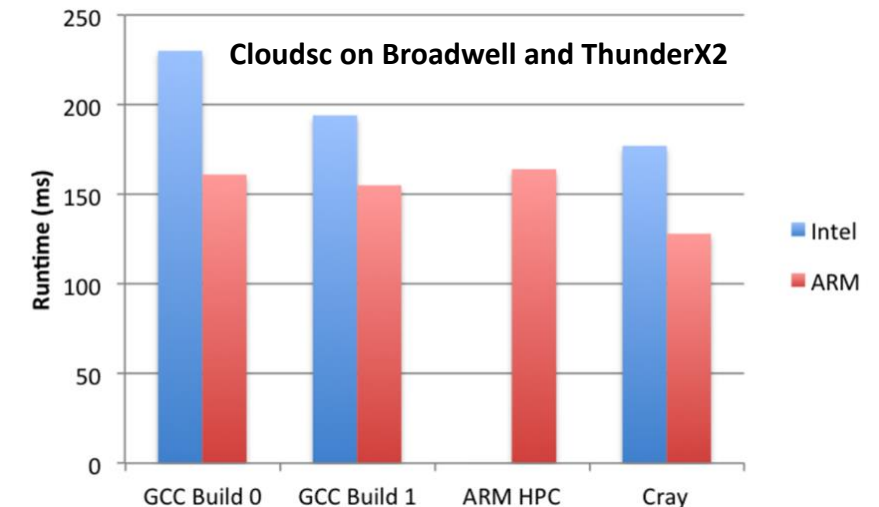
- Integrate easy-to-use performance tools with IFS, available to all
- ARM Forge MAP, BSC Extrae & Paraver (see POP CoE)

[From Patrick Gillies – check also Mario Acosta’s talk on Wednesday!]



Porting code to other processor types:

- OpenIFS and ESCAPE dwarfs ported to early access nodes - collaboration with U Bristol using Isambard Cray platform with Cavium ThunderX2 CPUs
- Long and short-wave MCICA solvers ported to GPU V100 with OpenACC (achieves 85% of peak memory bandwidth on V100) – collaboration with NVIDIA by hackathon for ECMWF staff

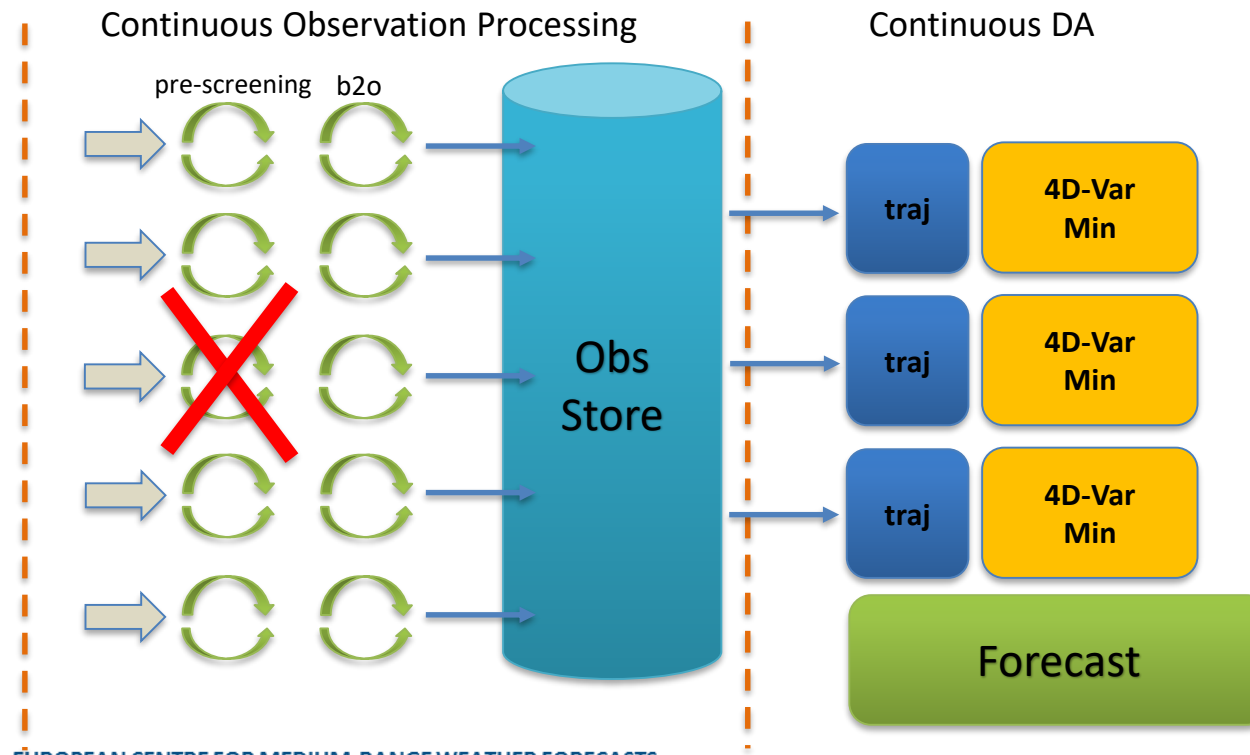


Not-so-low-hanging fruit: Pre-processing

Current processing chain is sequential; a failure at any point leads to delay in forecast production



COPE: Observations pre-screened in small batches as they arrive.
Decoupled system is more robust to failures.



Gains:

- resilience
- 15% cost in critical path

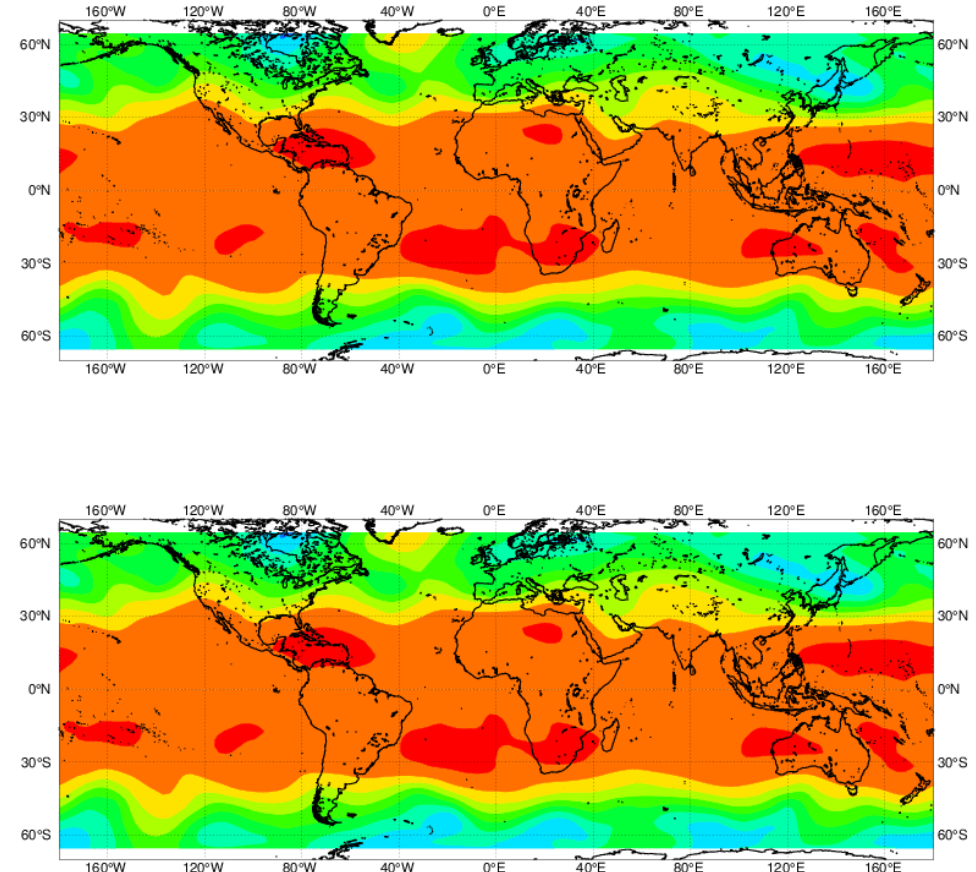
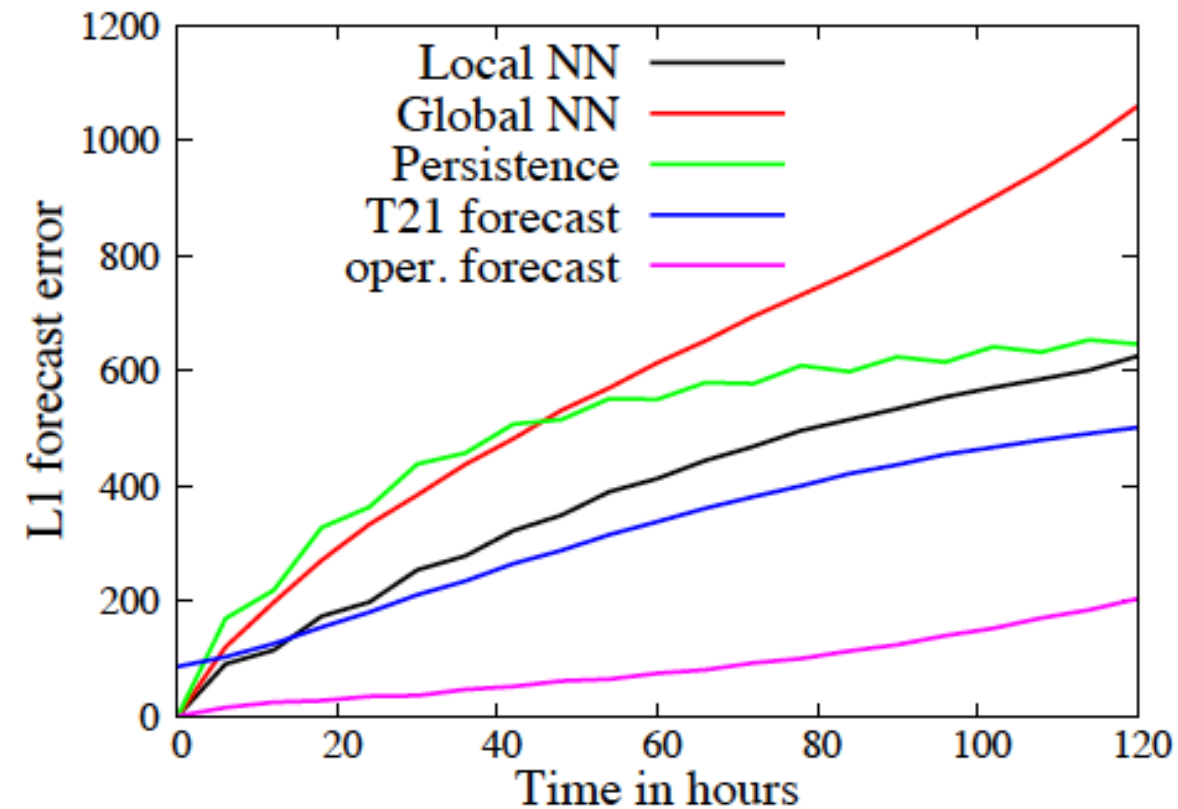
[From Peter Lean – check his talk on Friday!]

-

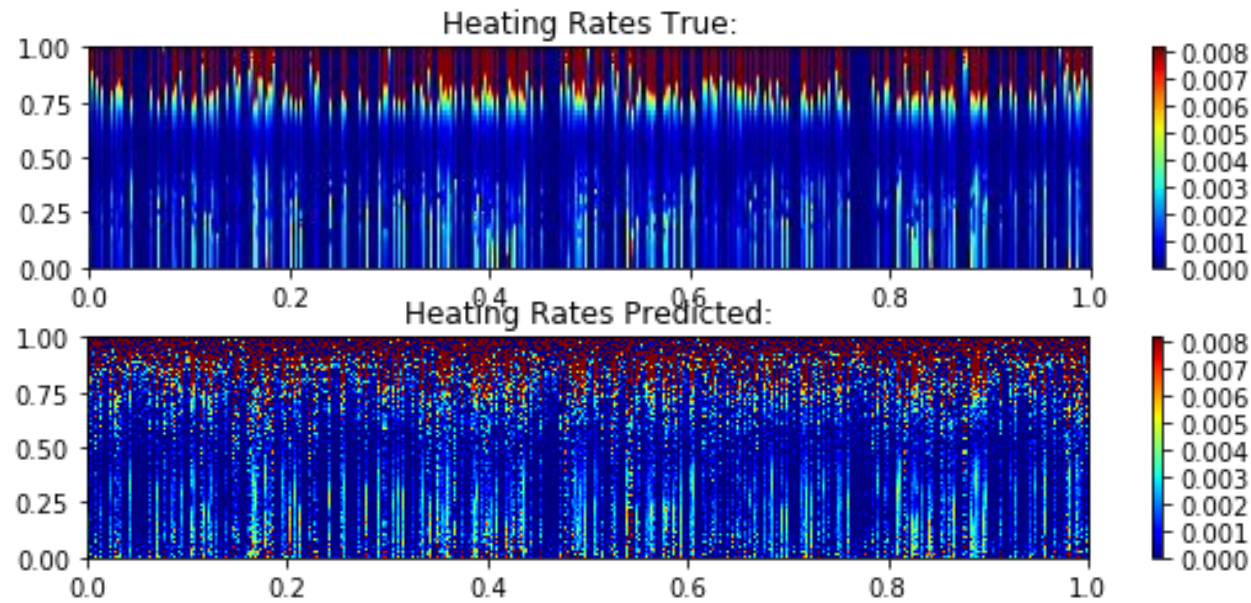
[illegible]

Not-so-low-hanging fruit: AI methods for forecasting

1. Take ERA-5 z500/6° LAT/LON reanalyses/forecasts forecasts = operational forecasts, T21 forecasts, persistence
2. Train NN with truth
3. Run NN forecasts for z500 with all 9x9 grid points predicting tendency = local NN
4. Run NN forecasts for z500 with all grid points predicting tendency = global NN



Not-so-low-hanging fruit: AI methods for parameterizations



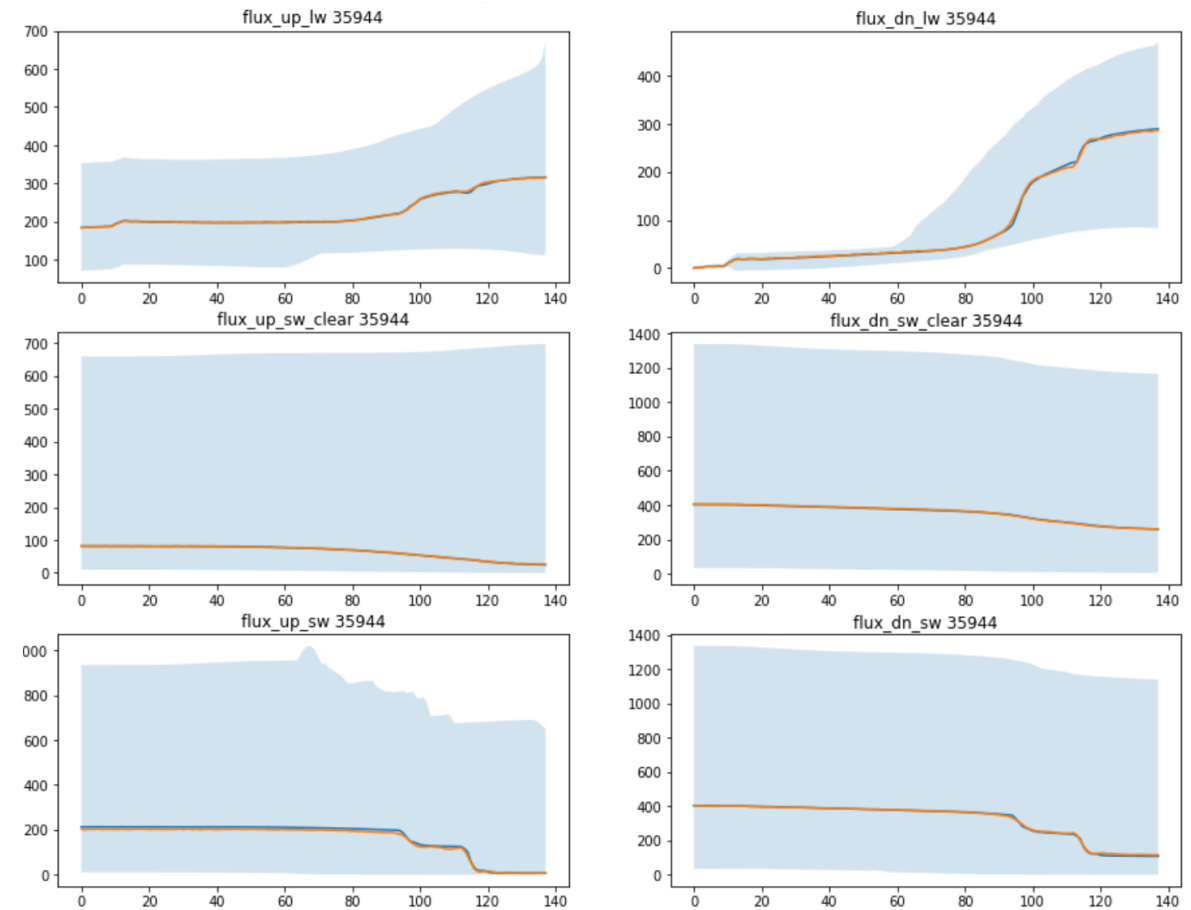
Data Set: 150,000 profiles total (25,000 locations with different solar zenith angles), divided into training=126,000, validation=24,000

Input to the network: 128 x 137 x 19 (128 batch size, 137 full levels, 19 variables SW clear sky)

Output of the network: 128 x 138 x 2 (up and down flux on each half level)

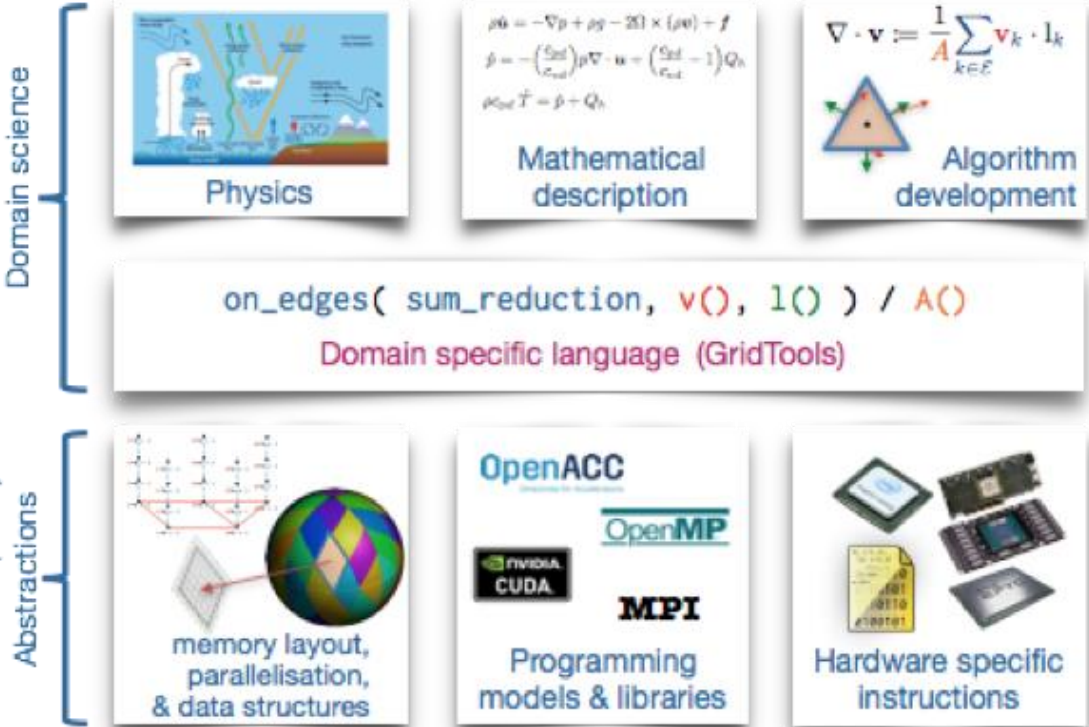
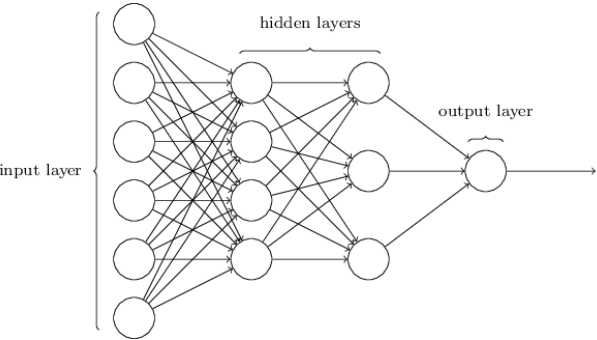
Network: four 1D convolutional layers followed by two fully connected layers; 194k trainable parameters

Shortwave and longwave flux profiles
(reference, NN, shading = natural variability)



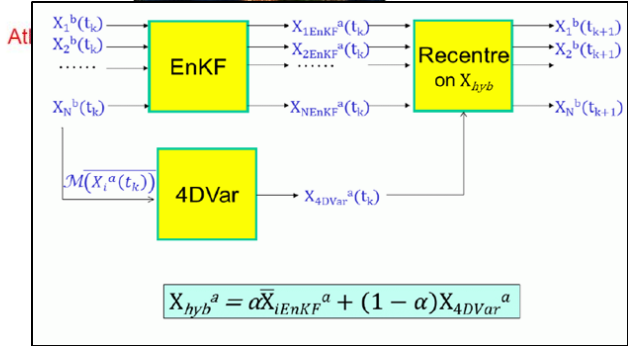
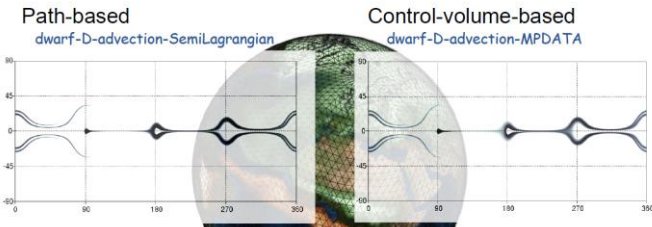
Far-hanging fruits: Algorithms – programming - hardware

Neural networks

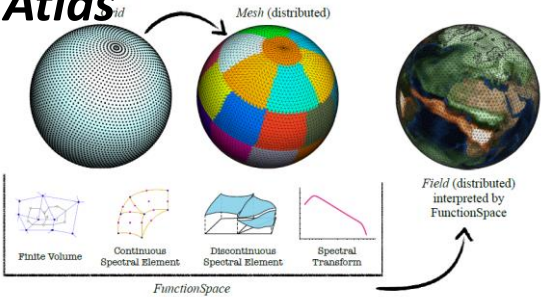


Mathematics&algorithms

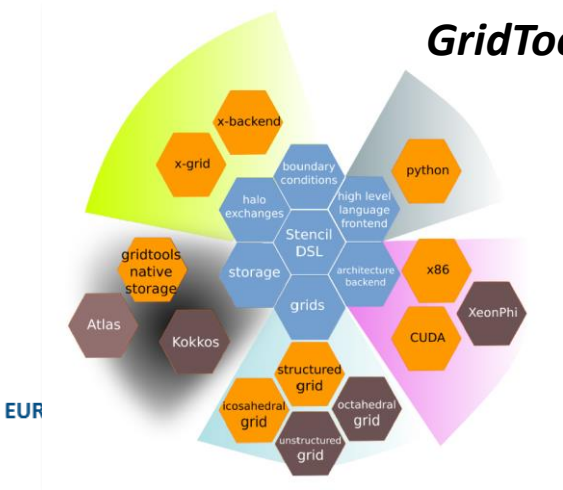
Rossby-Haurwitz test case after 7 days



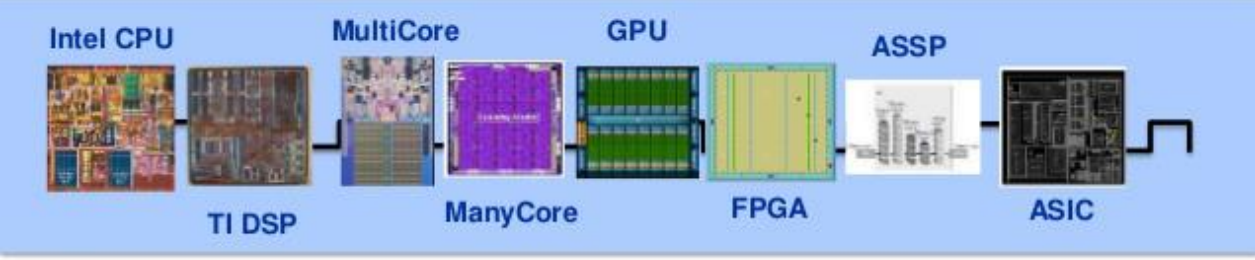
Atlas



GridTools



Processors

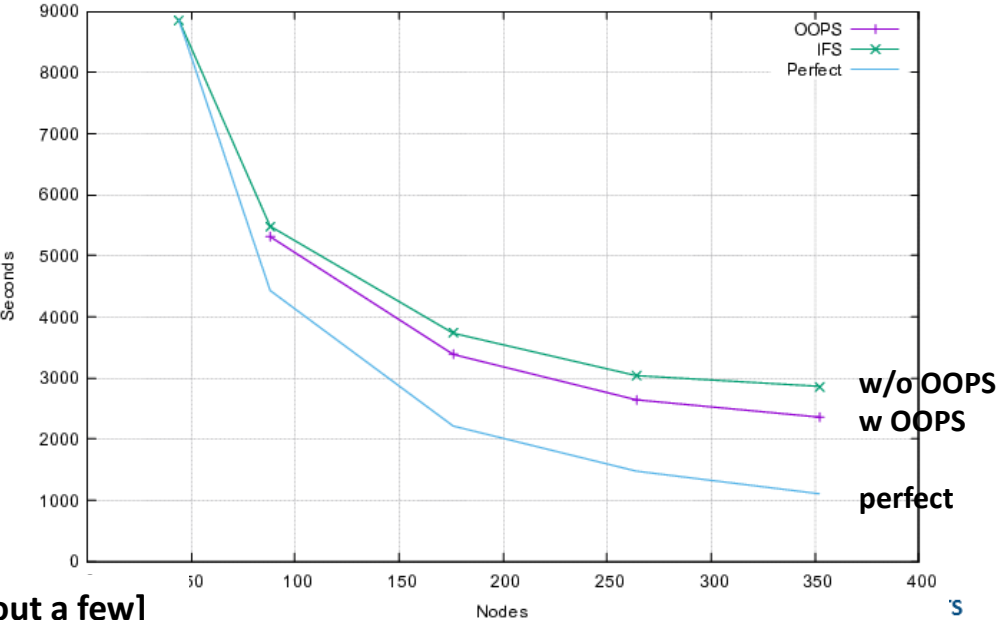
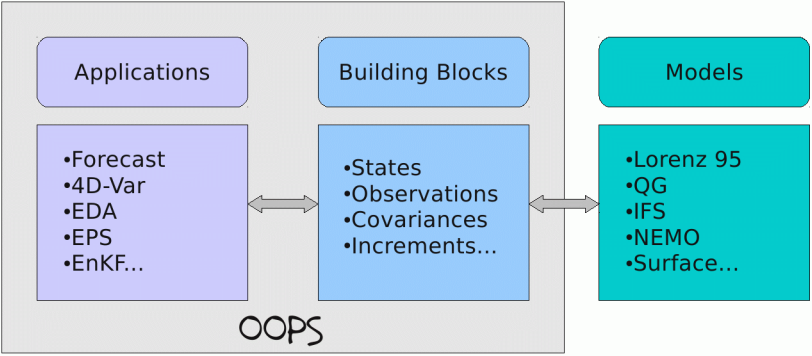


Far-hanging fruits: Algorithms – programming - hardware

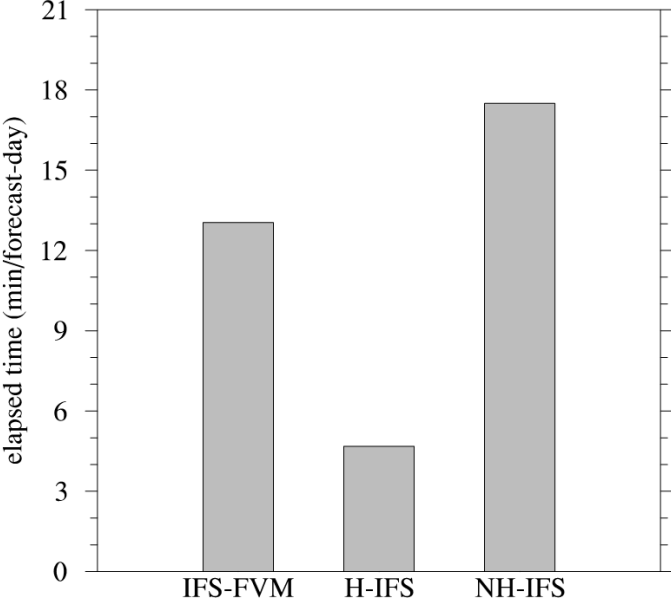
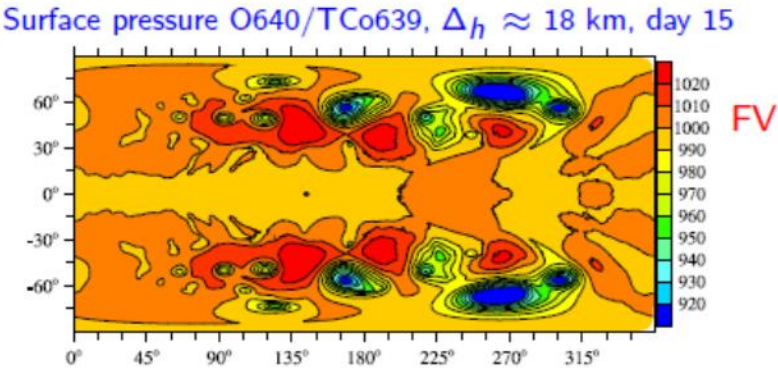
Algorithmic flexibility equally applicable to:

Data Assimilation

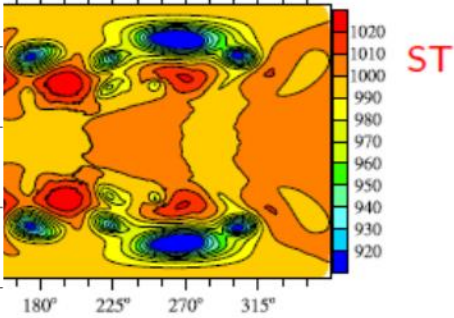
Forecasting



[Too many to single out a few]



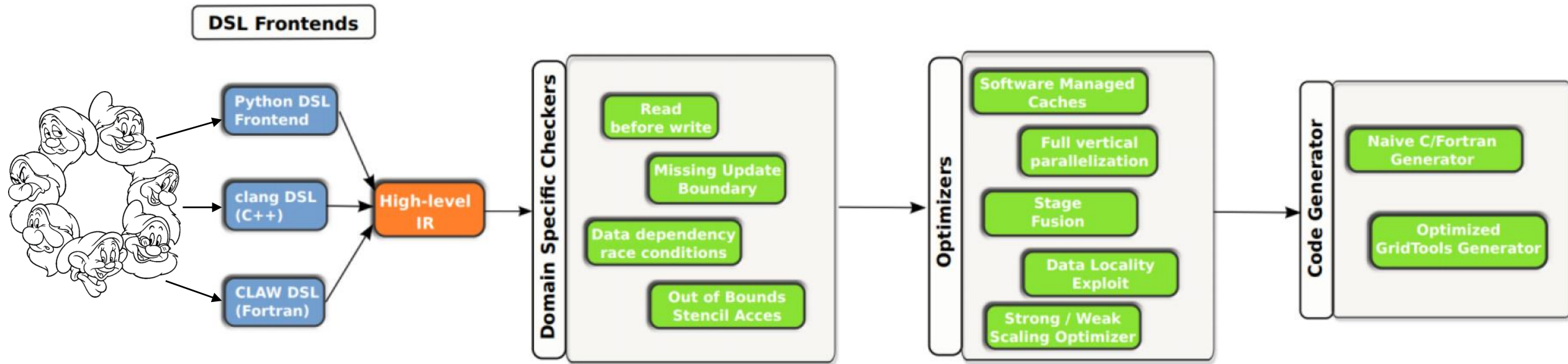
O1280/TCO1279 and L137 using dry dycore on 350 nodes of ECMWF's Cray XC40



[Christoph Kühnlein, Piotr Smolarkiewicz]

Far-hanging fruits: Algorithms – programming - hardware

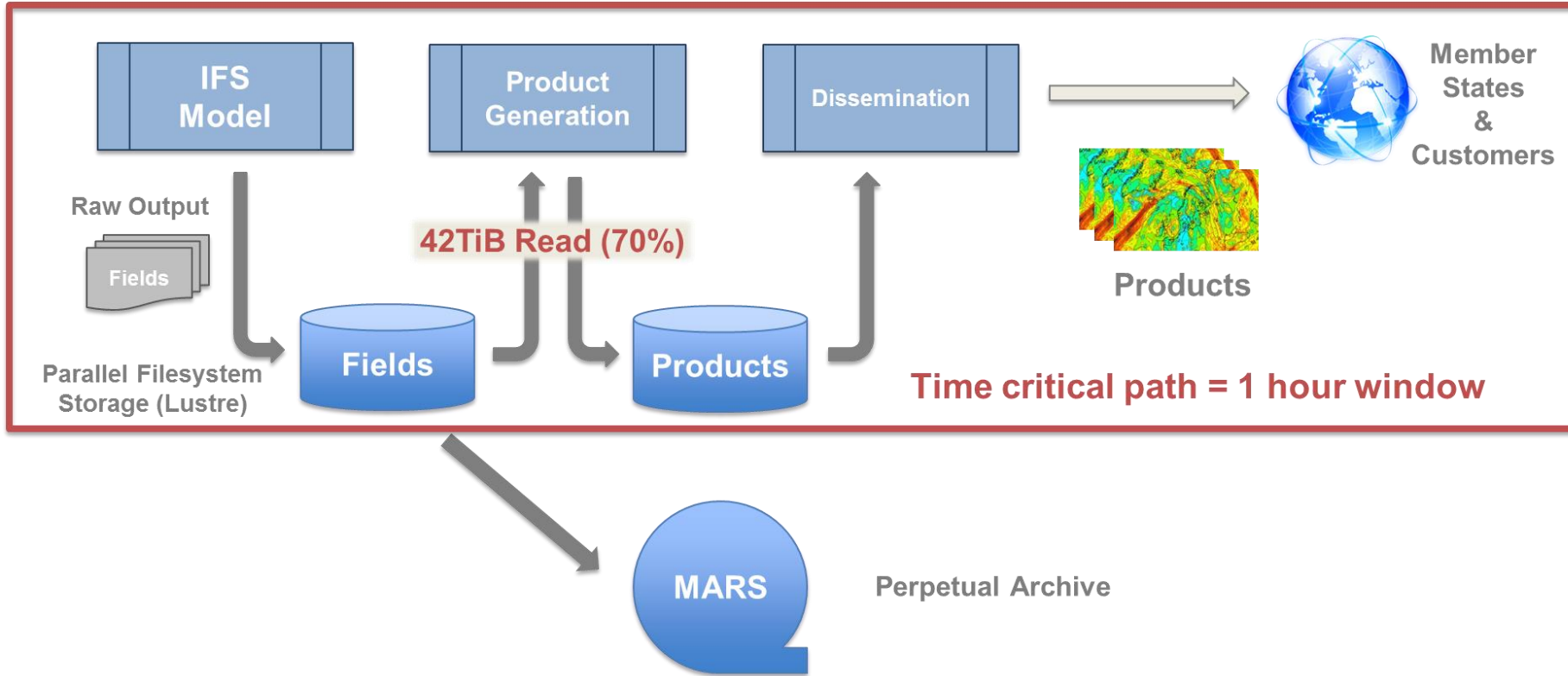
The **ESCAPE2** DSL tool-chain ...



... supported by the  toolchain:

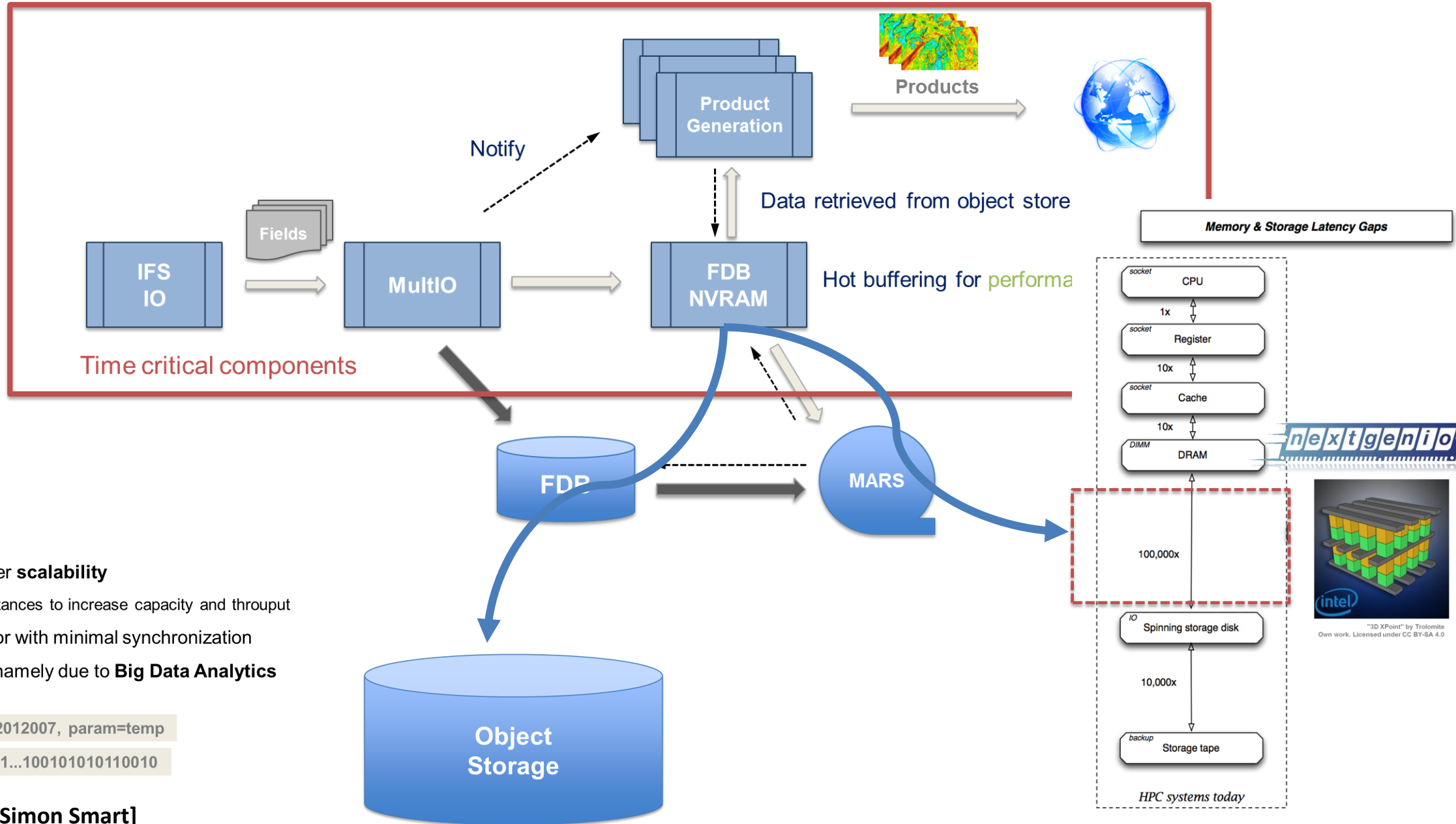
1. Generate single-column abstraction code for physics dwarfs using Loki (=Python code transformation)
2. Generate GPU-code using CLAW
3. Generate prototype C-kernel for initial FPGA porting

Examples of far-hanging fruit: Post-processing



	Model	Model + I/O	Model + I/O + PGen
Nodes	2440	2776	2926
Run time [s]	5765	6749	7260
Relative	-	+ 17%	+ 26%

Examples of far-hanging fruit: Post-processing



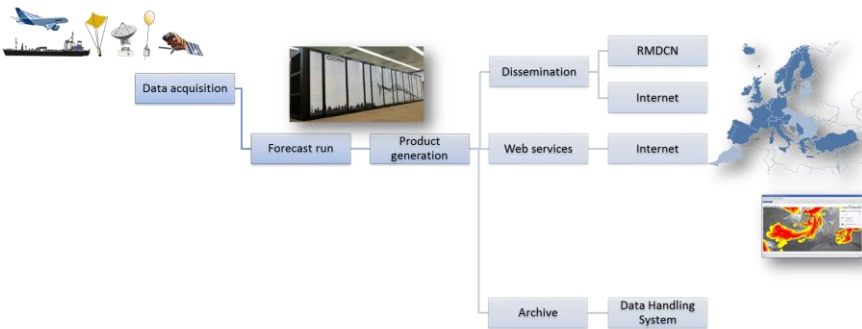
- Key-Value stores offer **scalability**
 - Just add more instances to increase capacity and throuput
- **Transaction** behavior with minimal synchronization
- Growing popularity, namely due to **Big Data Analytics**

Key: date=12012007, param=temp

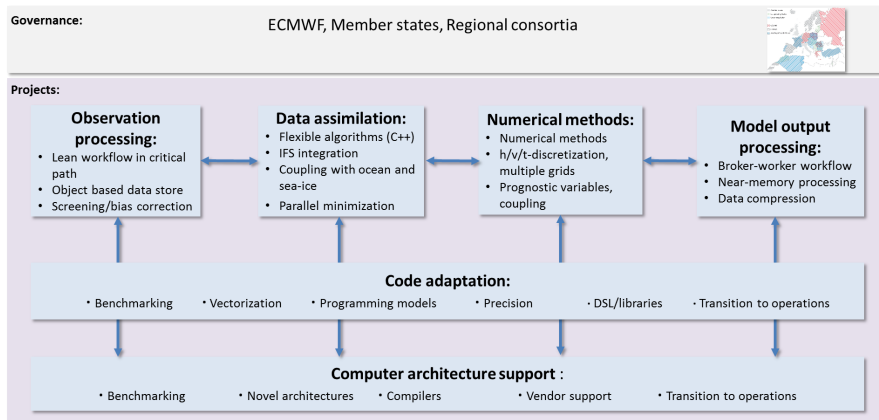
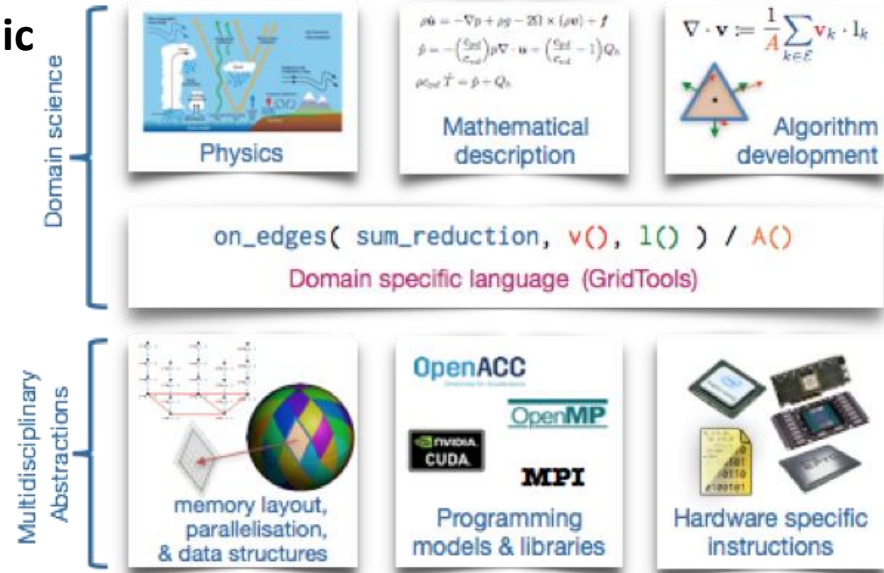
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ECMWF Scalability Programme 2.0

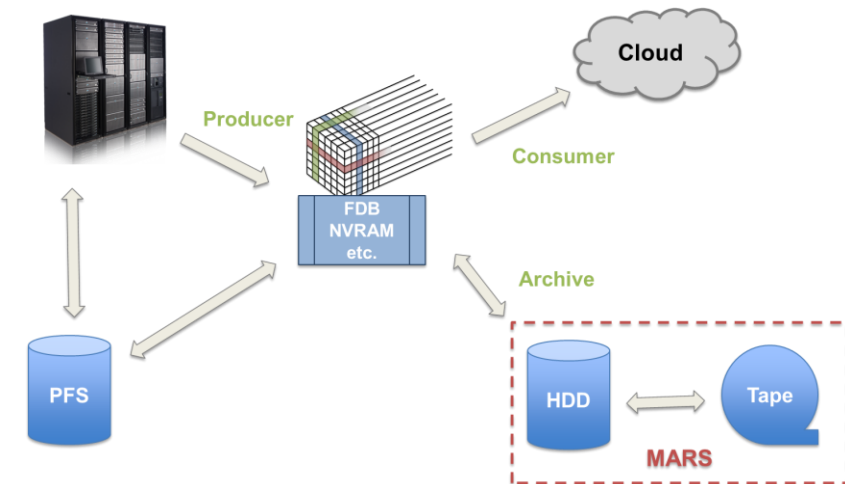
Scalability Programme 1.0



Co-development of algorithmic options, programming and hardware:



Co-development of flexible workflows, object data stores and hardware (cloud aware):



There is an opportunity to take this to the extreme!



Human Brain Project



Future and Emerging Technology Flagships are:

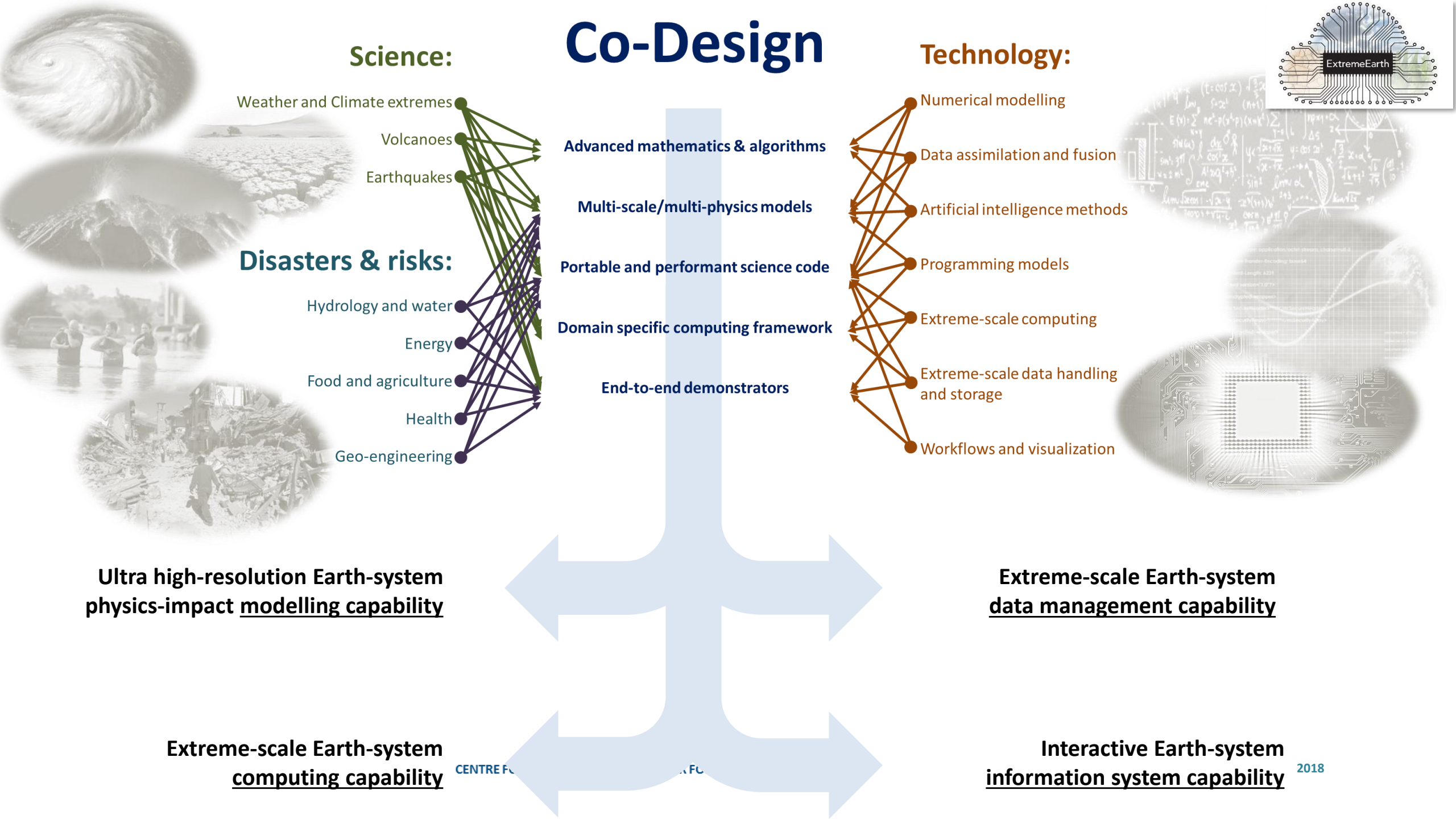
“... science- and technology-driven, large-scale, multidisciplinary research initiatives built around a visionary unifying goal ... tackle grand science and technology challenges ... strong and broad basis for future innovation and economic exploitation ... novel benefits for society of a potential high impact ... long-term and sustained effort.”

The *ExtremeEarth* proposal: www.extremeeearth.eu

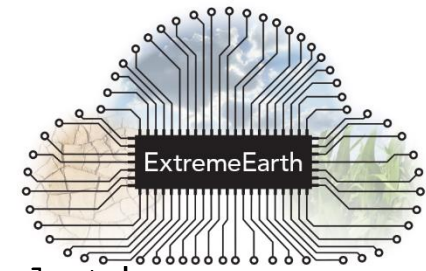


EUROPEAN CENTRE FOR MEDIUM-RANGE WEATHER FORECASTS





Future forecasting



Incremental:

```
do while (skill .ne. good_enough)
```

```
    model%resolution      = model%resolution      / model%dresolution
    model%complexity      = model%complexity      * model%dcomplexity
    ensemble%size         = ensemble%size         * ensemble%dsample
    downstream%application = downstream%application + 1
```

```
    call performance (model, ensemble, downstream, speed)
```

```
    call translate   (model, ensemble, downstream, speed, software, hardware)
```

```
do while (speed .ne. fast_enough)
```

```
    call add_funding      (bucks, software, hardware)
```

```
    call add_optimization (software, hardware, speed)
```

```
    call add_processors   (software, hardware, speed)
```

```
    if (bucks .gt. budget) abort
```

```
end do
```

```
    call science (model, ensemble, downstream, skill)
```

```
end do
```

Radical:

```
call extreme_earth
```