

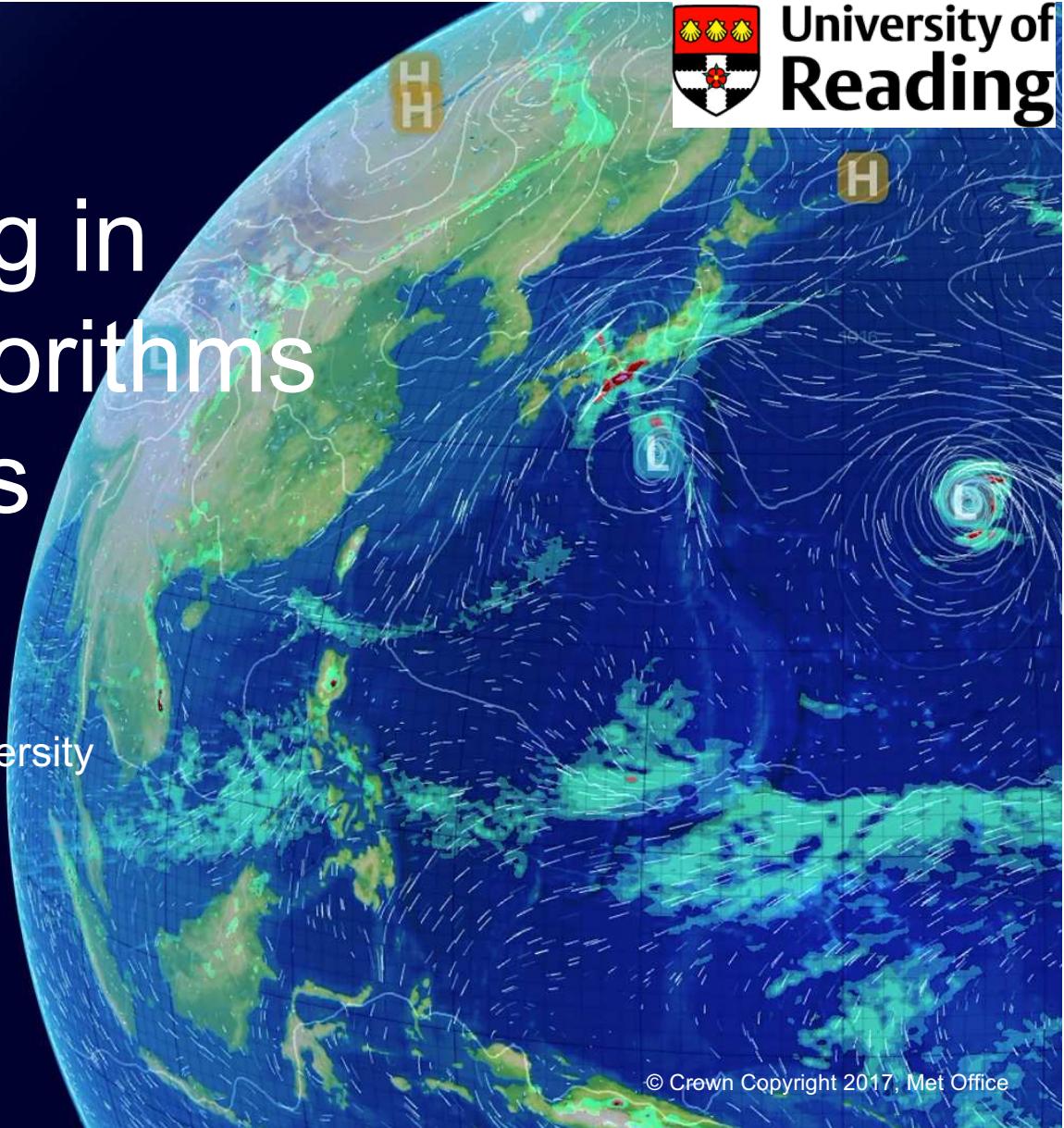
Parallel programming in practice: Scaling algorithms and Coupling models

Dr Chris Maynard

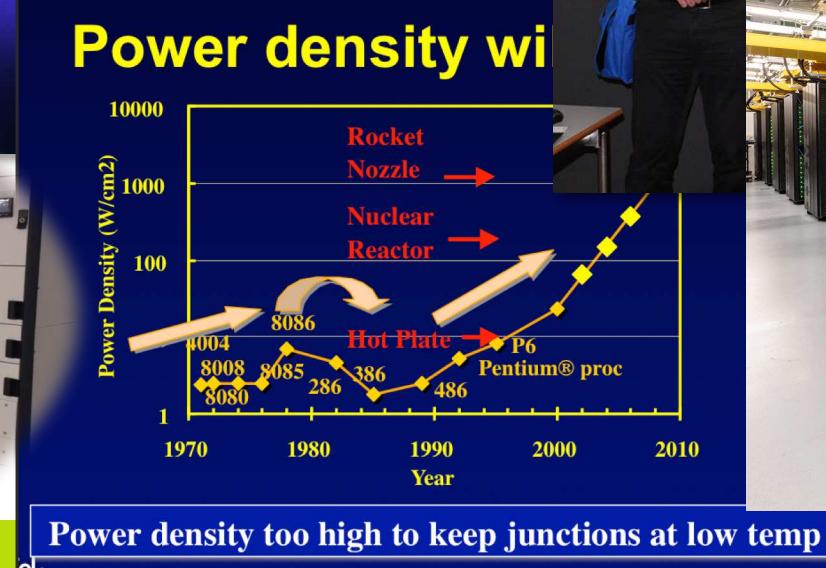
Acting HPC Optimisation Manager - Met Office

Associate Professor of Computer Science – University
of Reading

www.aces.cs.reading.uk



End of the Free Lunch



Computation

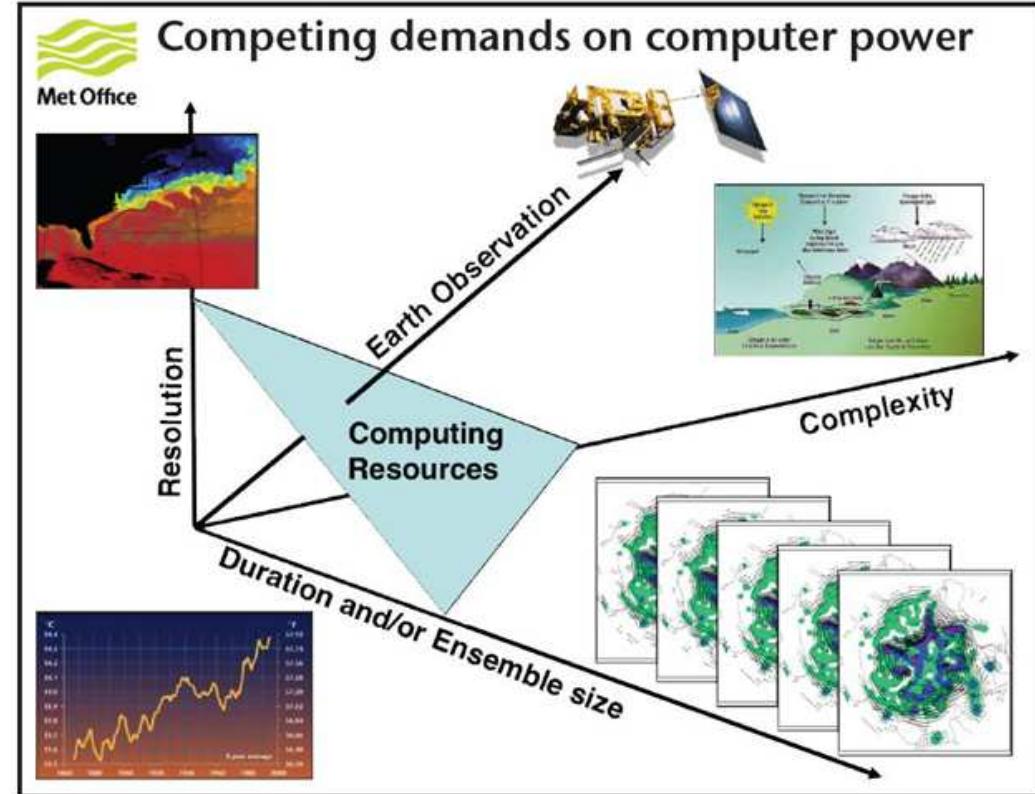
Scientific innovation is limited by computation

- Size and speed of calculation

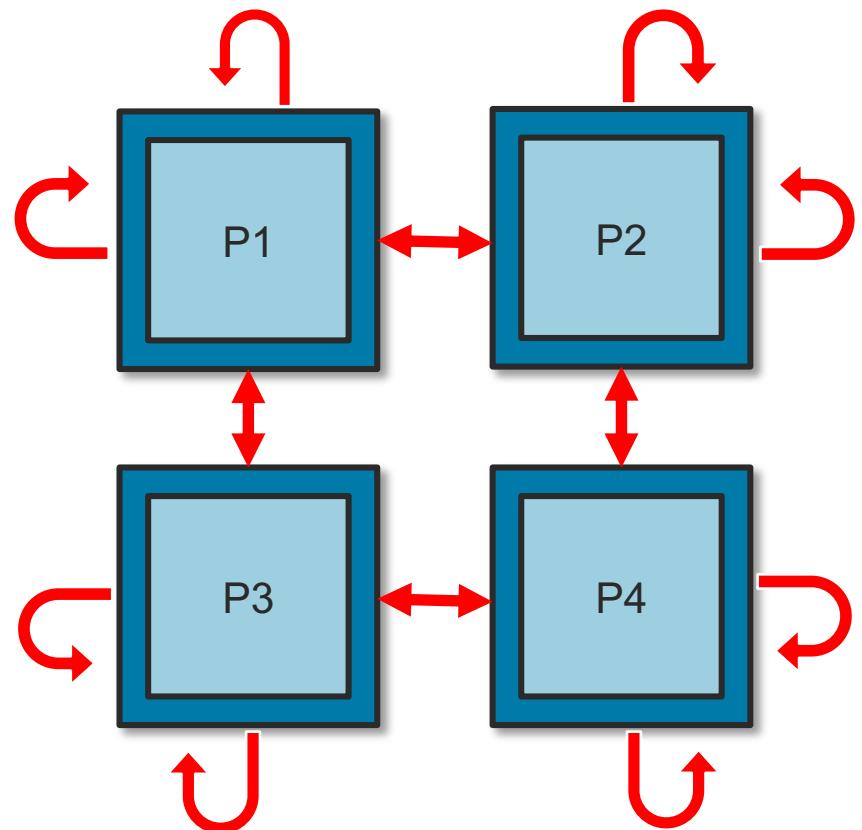
Speed of computation – End of Moore's Law

If we cannot compute X and Y faster
Can we compute X and Y simultaneously?

Dependency Y depends on X –
cannot compute simultaneously



Parallelism



Data Parallelism

Data decomposed across parallel elements (PEs)
PEs perform same action on different data
Single Program Multiple Data (SPMD)
over MPI
- this example includes data movement – halo exchange

Task parallelism

Also known as functional parallelism

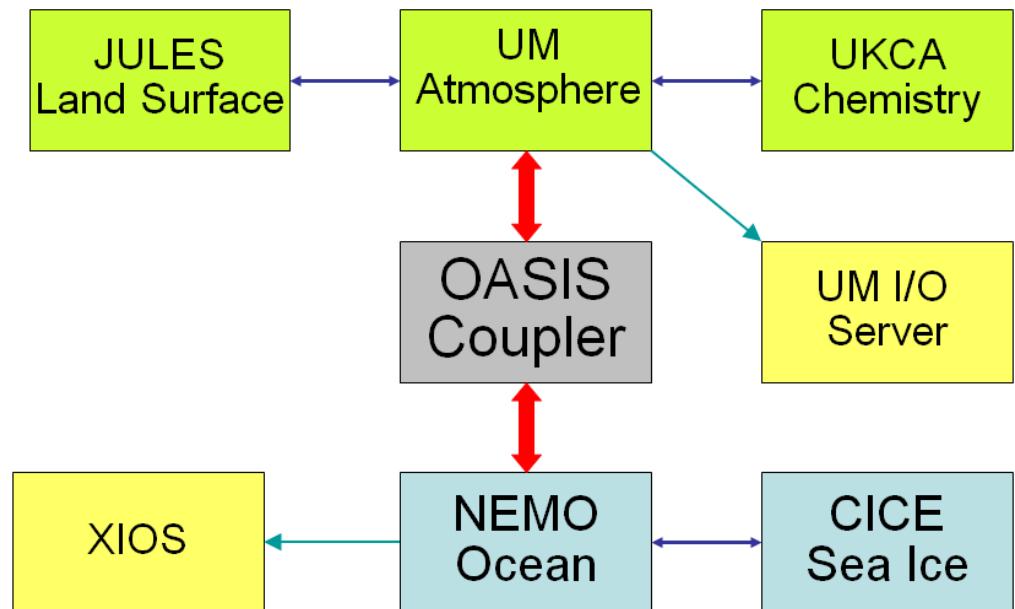
Decompose problem into
independent pieces

Coupled models (ATM + Ocean)

Ensembles (many models on
perturbed data)

I/O server – Asynchronous offload

CPU + GPU -- kernel offload



Scalar

x

*

y

$x * y$

SIMD

x3	x2	x1	x0
----	----	----	----

*

y3	y2	y1	y0
----	----	----	----

$x3 * y3$	$x2 * y2$	$x1 * y1$	$x0 * y0$
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Instruction Level Parallelism (ILP)

Fused multiply add (single op)

$$\mathbf{F} = \mathbf{a} * \mathbf{x} + \mathbf{y}$$

Single Instruction Multiple Data (SIMD) – combined with data parallelism

Vectorisation on modern CPU

Distinct from *pipeline vectorisation* on true vector processors

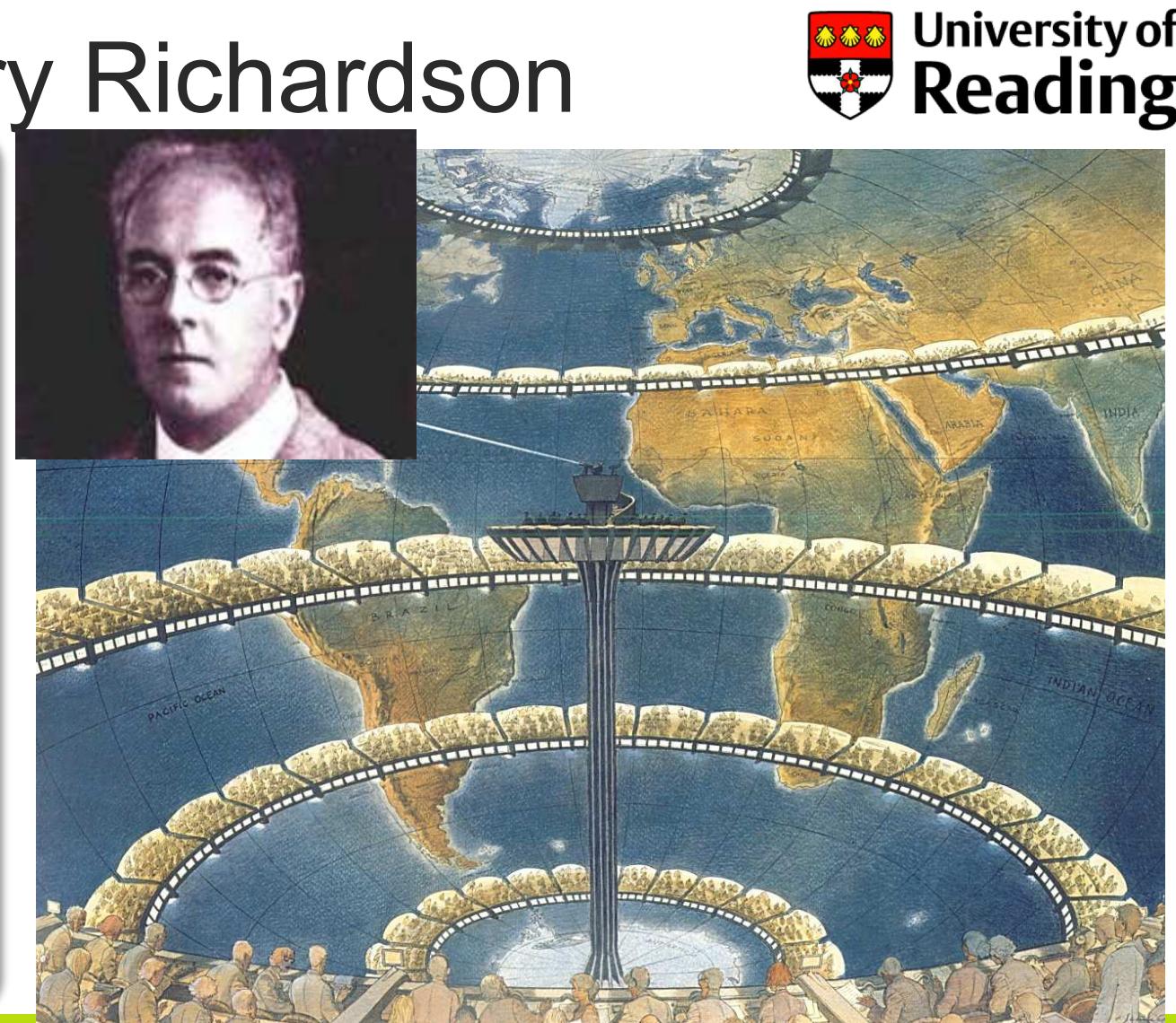
c.f. coalesced memory access on GPU Single Instruction Multiple Thread (SIMT)

LFR attempted first NWP calculation 1916-1917. Volunteer ambulance unit on the Western front.

7x7x5 grid, (250km resolution over Europe) two 3-hour timesteps.

Completely wrong – bad input data and CFL condition violation (not known).

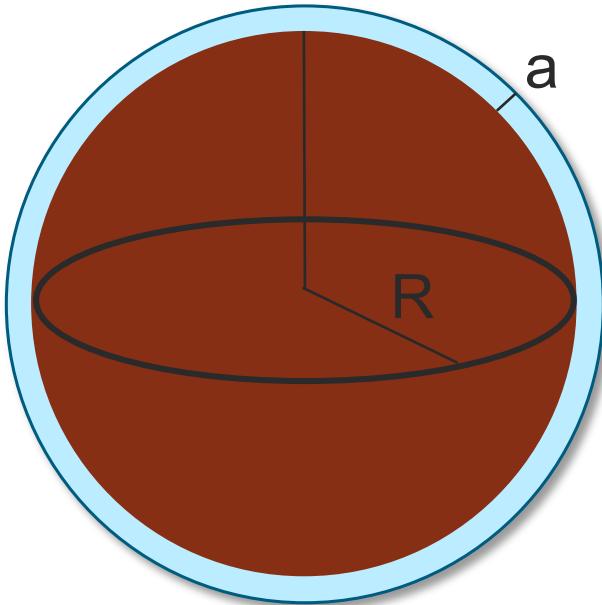
1922 paper *Weather Prediction by Numerical Process*
Recognised problem was parallel (64,000 computers)



Building a model

Designing a dynamical core

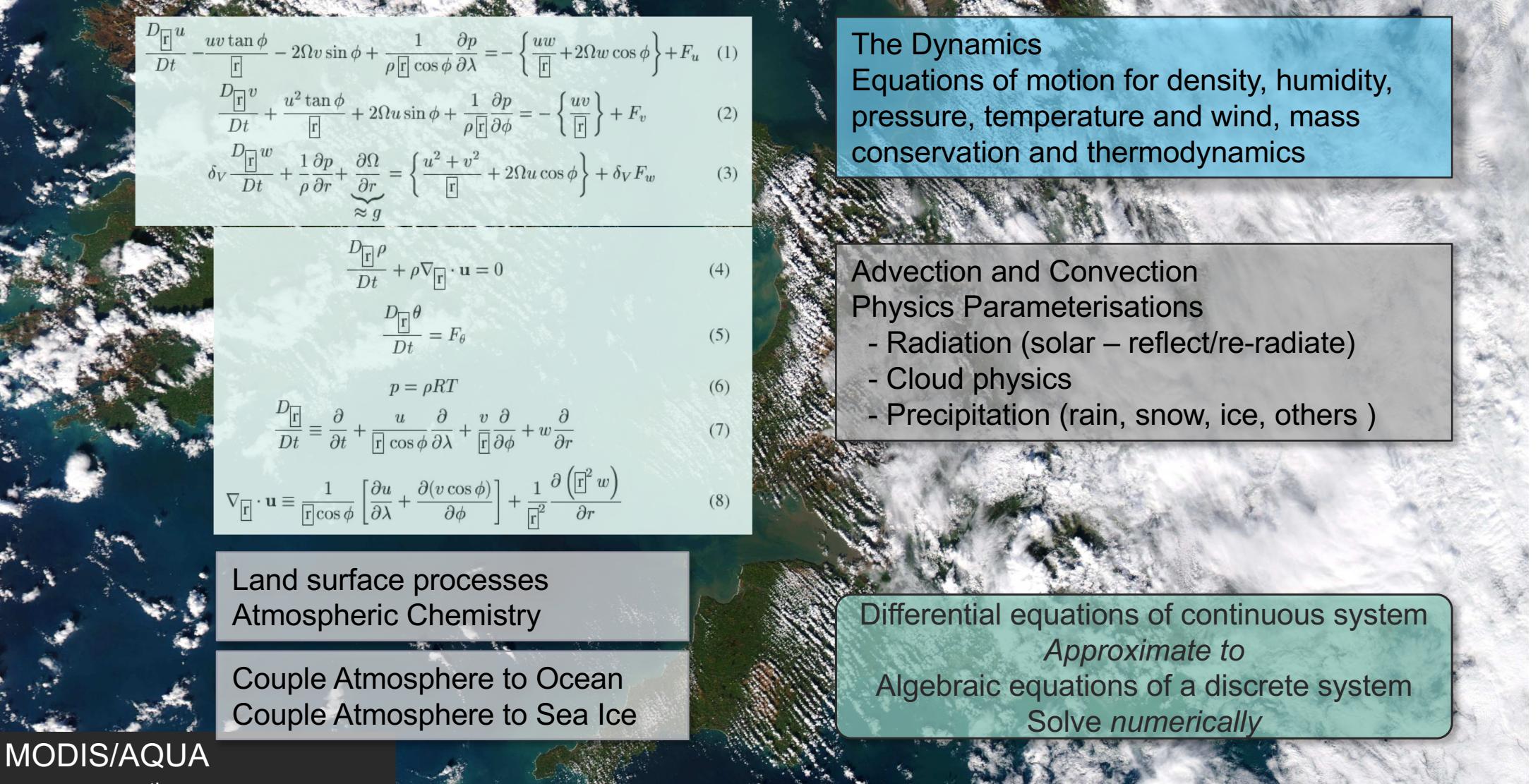
The domain specifics



Very complex domain
multi-component
multi-scale

Geometry: Spherical, Orography (Mountains)
Atmosphere is thin and vertically stratified
Kármán line $\sim 100\text{Km}$ (99.99997% atm)
Diagram drawn to scale $\sim 600\text{Km}$ atm. $R \gg a$
Rotation, not in thermal equilibrium
Atmosphere, Ocean, (Sea) Ice, land surface
Moist, Chemical and Biological processes

Models:
Large, $O(10^5) - O(10^6)$ LoC
Legacy, 10+ years to develop, lifetime 25+ years
Continuous development (e.g. UM 20% PA)
Operations (and production) – Conservative,
scientifically prudent!



$$\frac{D_{\mathbf{r}}}{Dt} u - \frac{uv \tan \phi}{\mathbf{r}} - 2\Omega v \sin \phi + \frac{1}{\rho \mathbf{r} \cos \phi} \frac{\partial p}{\partial \lambda} = - \left\{ \frac{uw}{\mathbf{r}} + 2\Omega w \cos \phi \right\} + F_u \quad (1)$$

$$\frac{D_{\mathbf{r}}}{Dt} v + \frac{u^2 \tan \phi}{\mathbf{r}} + 2\Omega u \sin \phi + \frac{1}{\rho \mathbf{r}} \frac{\partial p}{\partial \phi} = - \left\{ \frac{uv}{\mathbf{r}} \right\} + F_v \quad (2)$$

$$\delta_V \frac{D_{\mathbf{r}}}{Dt} w + \frac{1}{\rho} \frac{\partial p}{\partial r} + \underbrace{\frac{\partial \Omega}{\partial r}}_{\approx g} = \left\{ \frac{u^2 + v^2}{\mathbf{r}} + 2\Omega u \cos \phi \right\} + \delta_V F_w \quad (3)$$

$$\frac{D_{\mathbf{r}}}{Dt} \rho + \rho \nabla_{\mathbf{r}} \cdot \mathbf{u} = 0 \quad (4)$$

$$\frac{D_{\mathbf{r}}}{Dt} \theta = F_{\theta} \quad (5)$$

$$p = \rho R T \quad (6)$$

$$\frac{D_{\mathbf{r}}}{Dt} \equiv \frac{\partial}{\partial t} + \frac{u}{\mathbf{r} \cos \phi} \frac{\partial}{\partial \lambda} + \frac{v}{\mathbf{r}} \frac{\partial}{\partial \phi} + w \frac{\partial}{\partial r} \quad (7)$$

$$\nabla_{\mathbf{r}} \cdot \mathbf{u} \equiv \frac{1}{\mathbf{r} \cos \phi} \left[\frac{\partial u}{\partial \lambda} + \frac{\partial (v \cos \phi)}{\partial \phi} \right] + \frac{1}{\mathbf{r}^2} \frac{\partial (\mathbf{r}^2 w)}{\partial r} \quad (8)$$

Land surface processes
Atmospheric Chemistry

Couple Atmosphere to Ocean
Couple Atmosphere to Sea Ice

The Dynamics

Equations of motion for density, humidity, pressure, temperature and wind, mass conservation and thermodynamics

Advection and Convection

Physics Parameterisations

- Radiation (solar – reflect/re-radiate)
- Cloud physics
- Precipitation (rain, snow, ice, others)

Differential equations of continuous system

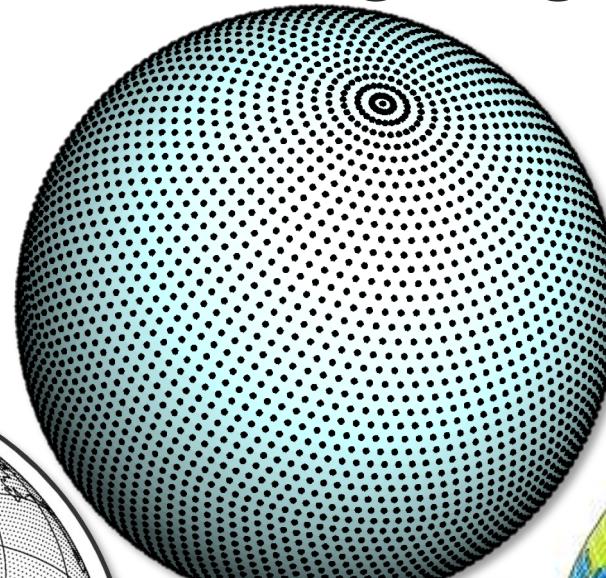
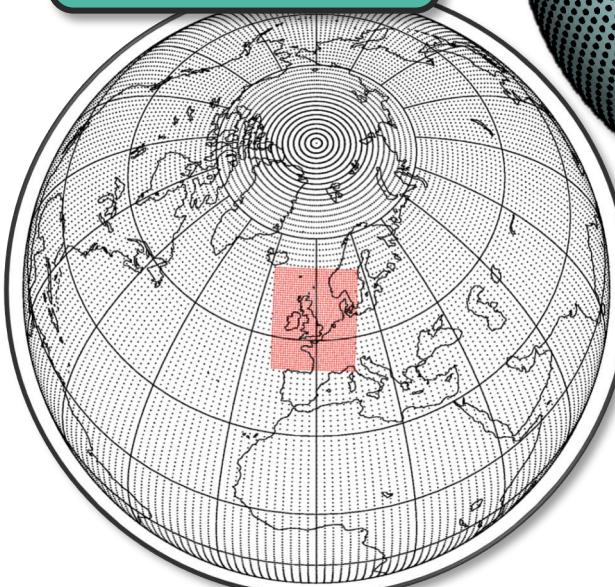
Approximate to

Algebraic equations of a discrete system

Solve numerically

Choosing a grid

Lat-Lon grid
Structured
Unified Model
(Met Office)

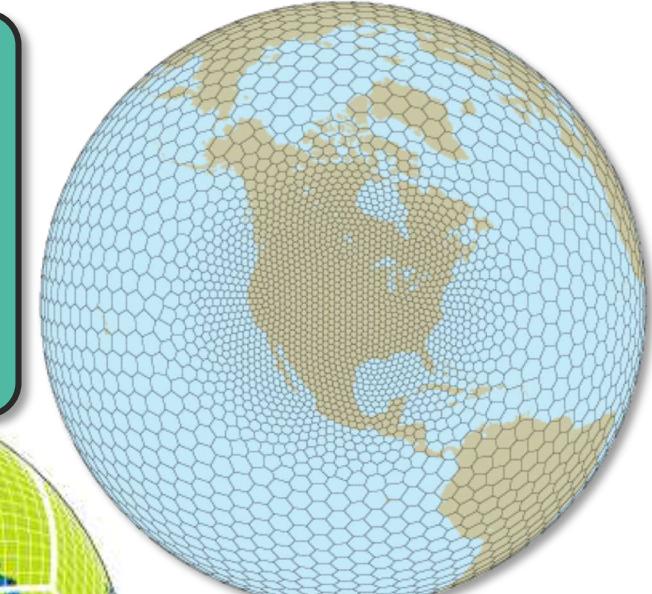


Octahedral
Gaussian grid
Structured
IFS (ECMWF)

Cubed Sphere
GungHo/LFRic
unstructured
(MO)
FV3 structured
(NOAA)



Icosahedral mesh
Unstructured
ICON (DWD)
MPAS Voronoi
stretch meshes

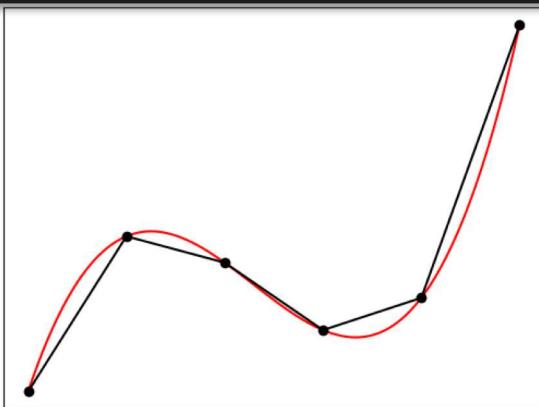




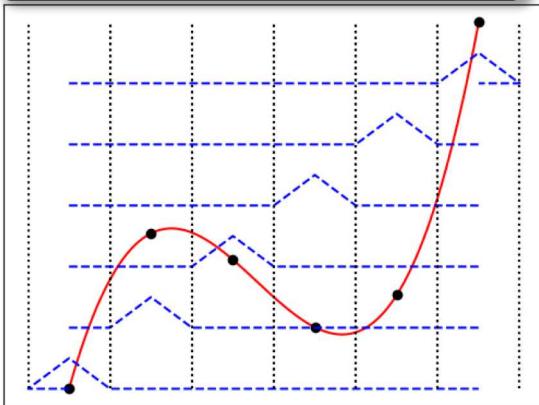
Choice of grid based
on Numerical Analysis
Symmetry properties
Consequences for
1. Accuracy
2. Stability
of numerical method

Structured
Neighbouring grid-points/cells known
Direct memory access
 $u(i) = u(i-1) + u(i+1)$
Good for data locality and caching
Geometry of sphere → problematic
communication patterns

Unstructured
Neighbour grid-points/cells not known. Use look up table → indirect
memory access
 $u(m(cell)) = u(st(cell,1)) + u(st(cell,2))$
Bad for data locality, can avoid problematic communication patterns

i) 1st order forward difference

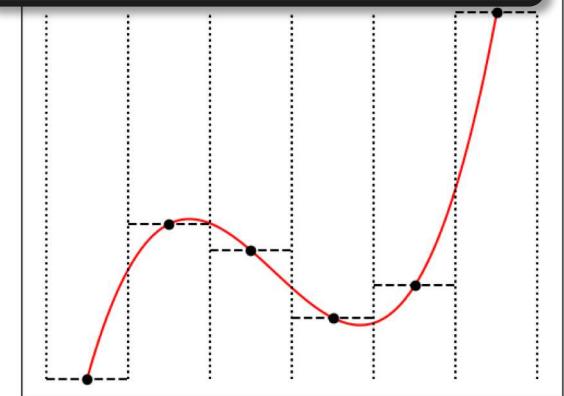
iii) Linear Finite Element



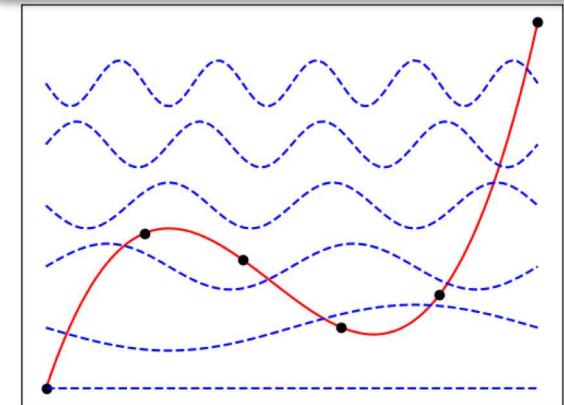
- i) Finite Difference - values at grid-points
- ii) Finite Volume - values over volume
- iii) Finite Element - functions over cell
- iv) Spectral - periodic (trigonometric/hyperbolic) functions over whole domain

All can be at higher order (lowest shown)

i) Constant Finite Volume



iv) Spectral method



Also need to discretise time

– again choices of “grid” and time-stepping scheme.

Courant-Friedrichs-Lowy (CFL) condition (stability)

$$\frac{u\Delta t}{\Delta x} \leq C$$

1-d advection where u is wave-velocity

C depends on u and discretisation scheme $C \sim \mathcal{O}(1)$

Different atmospheric waves

Acoustic, Gravity, Rossby different wavelengths and can have different treatments

i) Explicit $u(t_n) \sim f(u(t_{n-1}))$

ii) Implicit $u(t_n) \sim g(u(t_{n-1}), u(t_n))$

Also advection, i) Eulerian versus ii)
Semi-Lagrangian

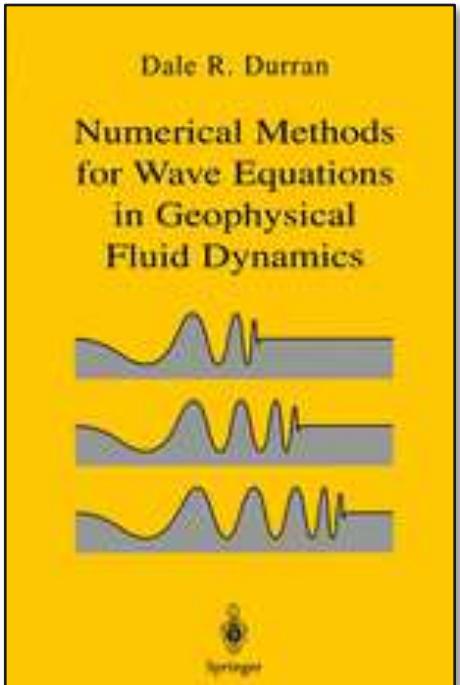
i) Cheap to compute, small time-step
ii) Costly to compute, large time-step

This is not a course on the dynamics nor numerical analysis

Choices of

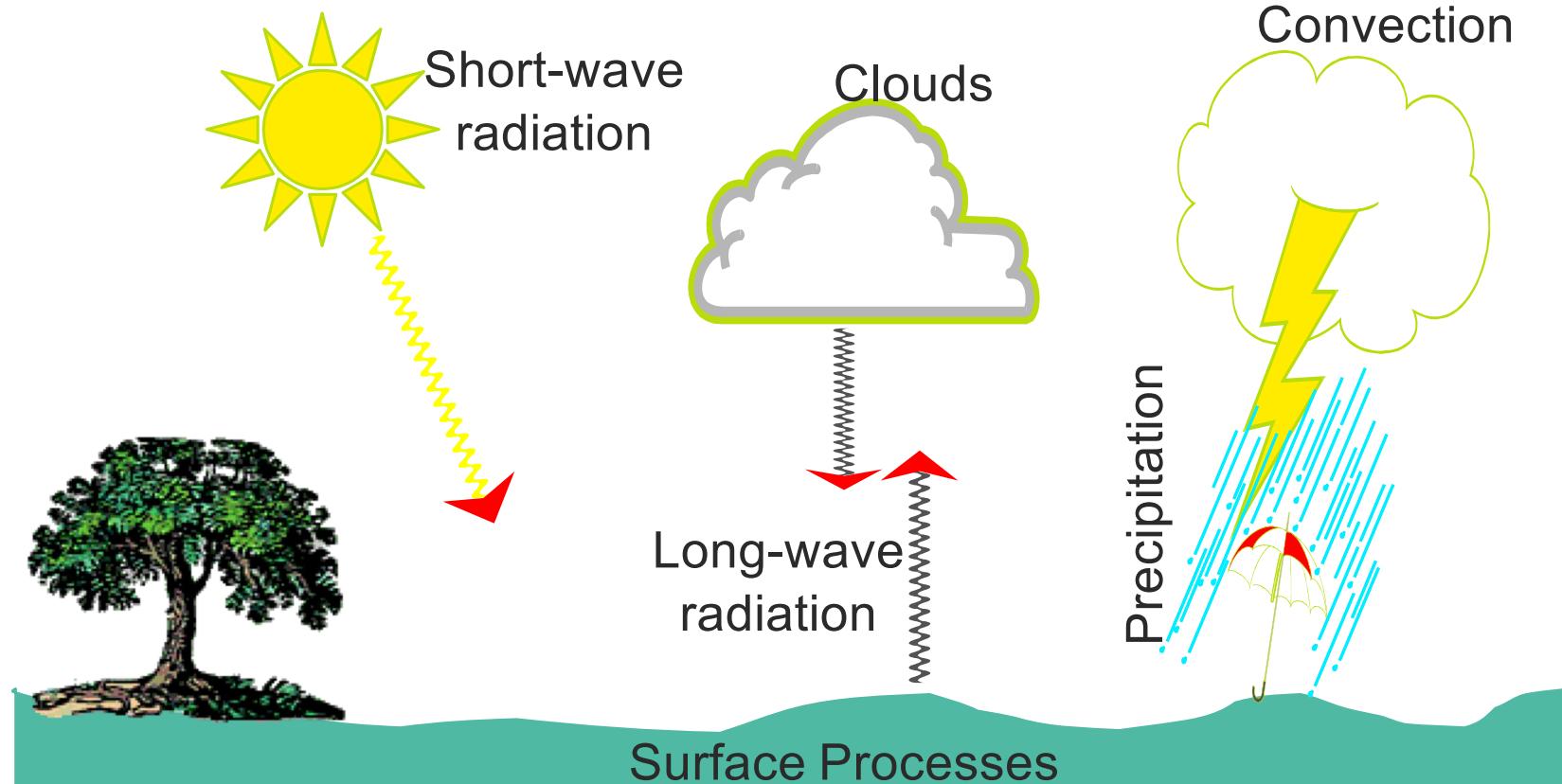
- i) Discretisation (space - grid, and time)
- ii) Method
- iii) Order
- iv) Time-stepping scheme
- v) Solution method

Different patterns of computation, computational and data dependency, and communication





Processes not resolved at grid scale



Dynamics

- Advection
- Solver

Physics

- Fast physics
 - cheap to compute,
varies quickly
- Slow physics
 - costly to compute, varies
slowly

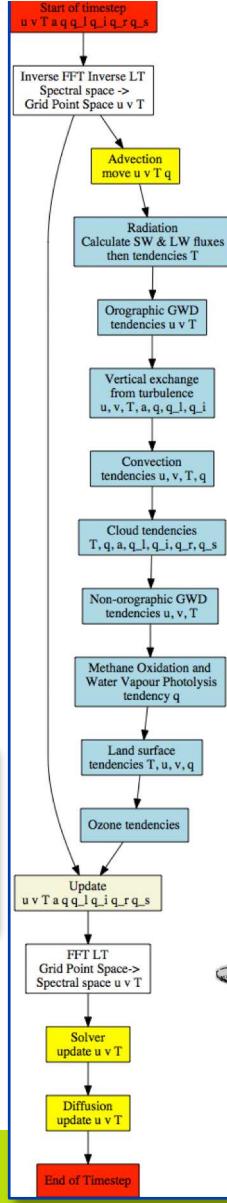
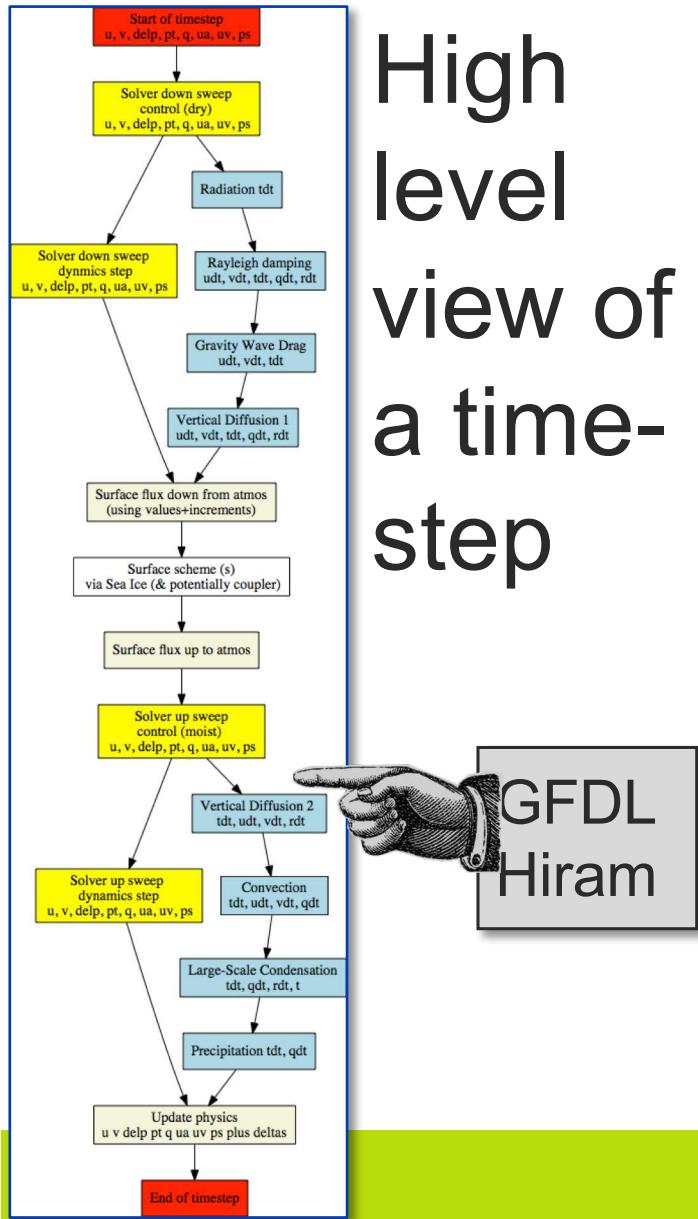
Different methods and algorithms for solving the problems.

Crossing the Chasm: how to develop weather and climate models for next generation computers? B.N. Lawrence *et al.*

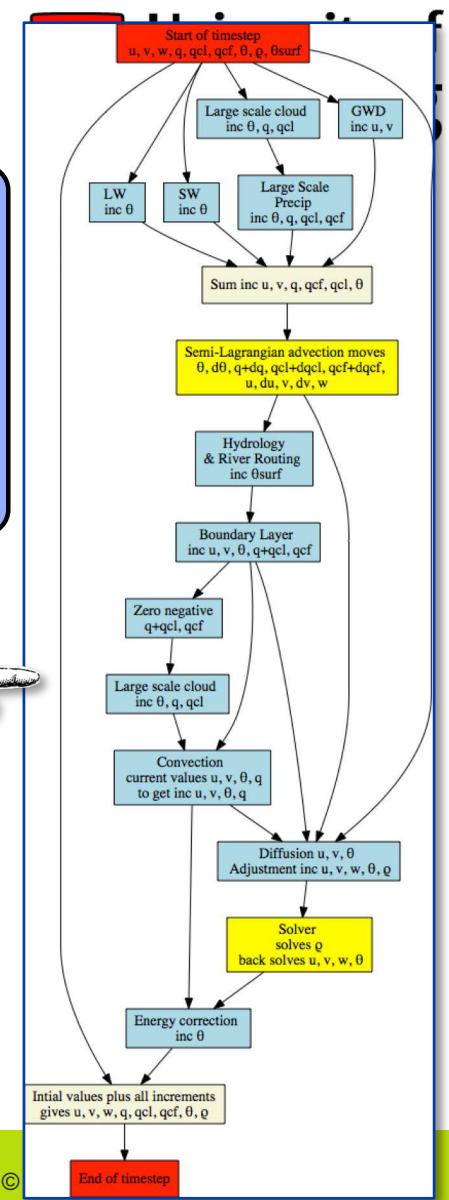
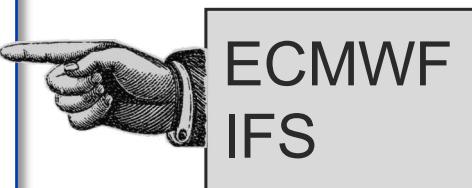
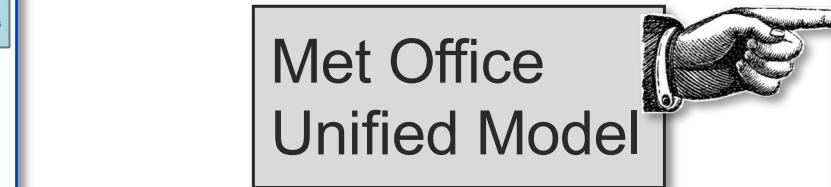
(Geosci. Model Dev., 11, 1799–1821, 2018

<https://doi.org/10.5194/gmd-11-1799-2018>)

High level view of a time-step



Red: Start/end
Yellow: Dynamics
Blue: Increment Physics
Beige: Physics, full values
White: Coupling/library



Parallel scaling

P is proportion of program which is parallelisable

S, maximum speed up achievable compared to serial code is

$$S_{\max} = \frac{1}{1 - P}$$

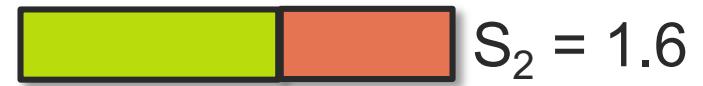
Even if all parts of program are parallelised, they have different scaling behaviour due to *communication* between parallel elements

$$P = 0.75$$

$$F = 0.25$$



$$n_t = 2$$

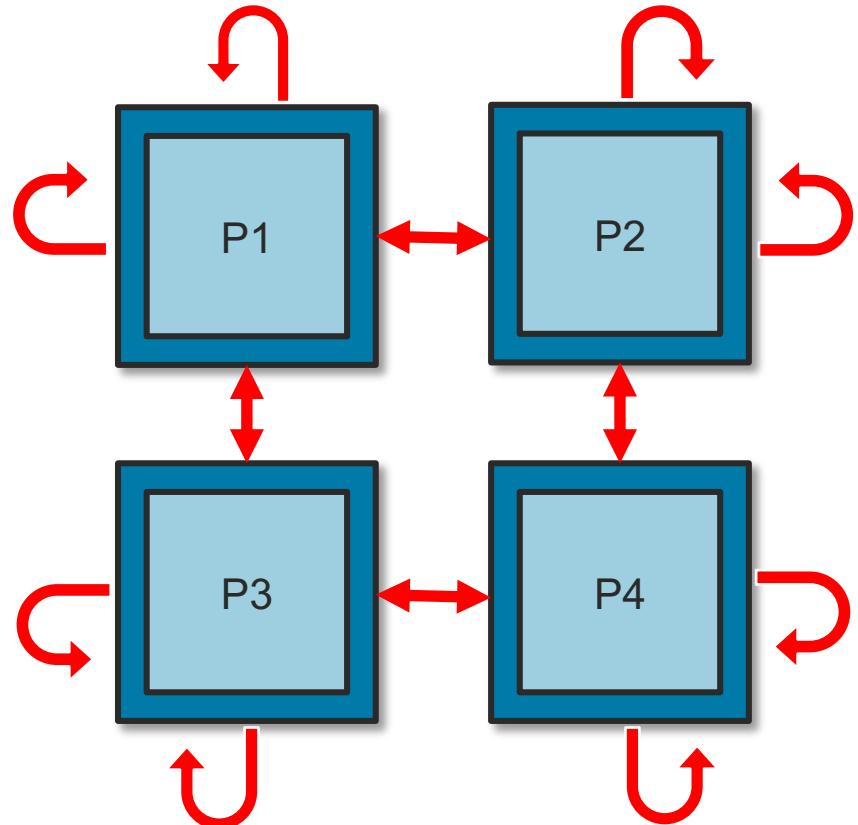


$$n_t = 4$$



$$n_t \rightarrow \infty$$





Local communication
Stencil-type calculations require data from neighbour
Halo exchange
Stencil size → halo depth
(e.g. Semi-lagrangian → large halos)
Point-to-point
Bandwidth limited



Global communication

All parallel elements take part

- Reductions – global sum (iterative solvers)
 - latency bound
- All-to-all – spectral transforms
 - latency and bandwidth bound
- I/O -- Serial data to parallel memory and vice versa
 - latency, bandwidth and raw data rate bound

Supercomputer turns a compute bound problem into an I/O bound problem. *Ken Kennedy*

Weak

Keep local problem size fixed
-- data size per parallel element is fixed – work rate constant
Local communication increases across whole problem, but not per PE
Global communication increases

Strong

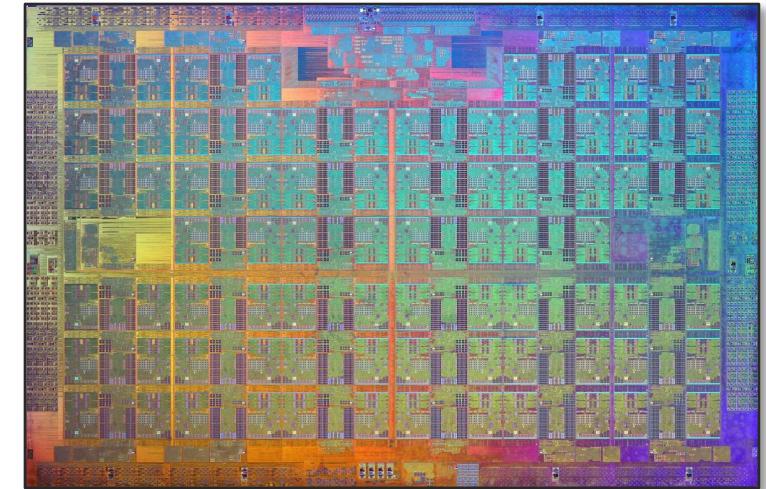
Keep problem sized fixed

-- Size of globe is fixed, but resolution is not

- Amount of work per parallel element decreases (solve faster)
 - Local communication decreases (but slower)
 - Global communication increases
- Strong scaling regime -- communication dominates

Levels of parallelism

This simple model of parallelism doesn't map onto modern, complex processors. Typically exhibiting multiple levels of parallelism and requiring multiple programming models to exploit them.



MPI + X

Where MPI is used for inter-node
distributed memory

X is intra-node parallelism

Usually OpenMP/OpenACC

Node Comparison

Met Office Cray XC40 (32 on top500)

Dual socket 18-core Intel Xeon

Non-Uniform Memory Access

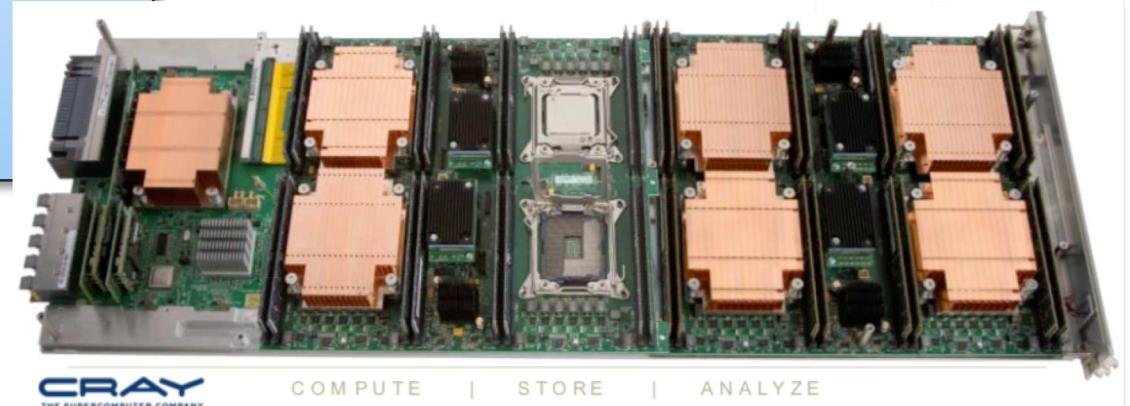
256 bit AVX SIMD ILP

6000+ nodes

Can Program MPI only

MPI + OpenMP is common

Whole machine
3 MW



Met Office Summit

Oak Ridge National Lab (ORNL), USA

2 on Top 500

dual IBM Power 9 22-core CPU

+ 6 NVIDIA VOLTA GPU (4000+
nodes)

Host and device memory

NV-link connections

84 streaming multiprocessors

Each SM has 64/32 32/64-bit cores

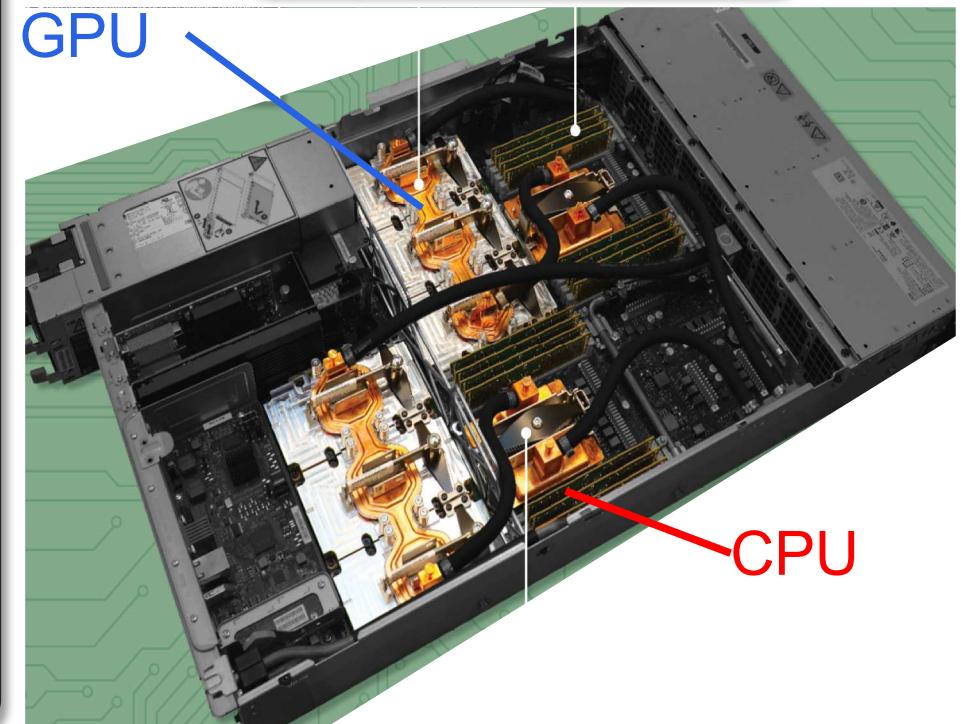
Hierarchy blocks, warps and threads

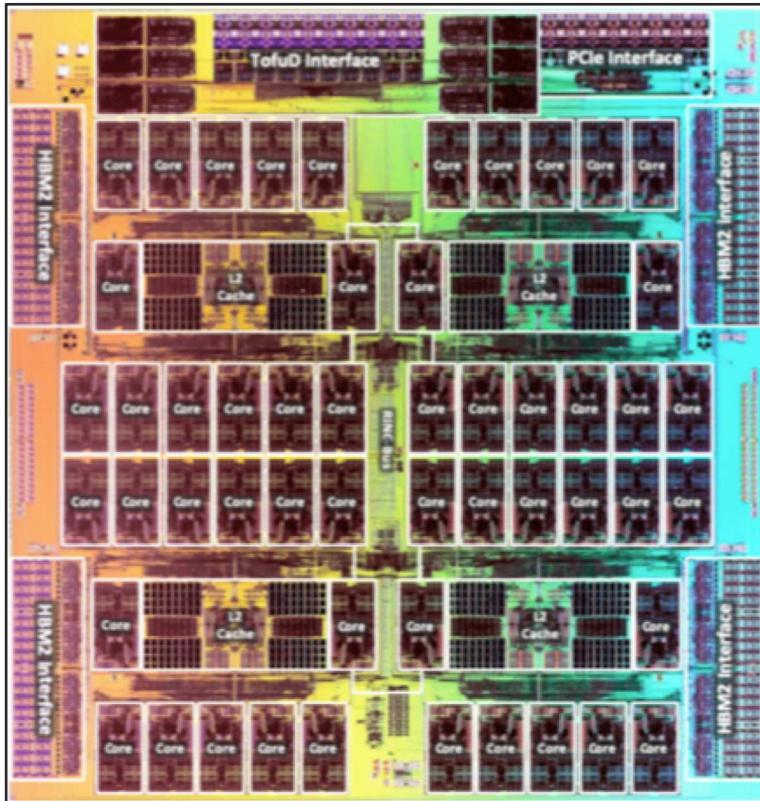
Oversubscribed concurrency

Tens of thousands of SIMT threads per
GPU



Whole machine
15 MW





Riken, Japan, 1 Top 500
Fujitsu 64-bit ARM processor
48 cores, 4*12 mini-NUMA
Each has 512-bit Scalable Vector
Extension SIMD (ILP)
150,000+ nodes
7M+ cores

Whole machine
28 MW

Real model scaling

Unified Model

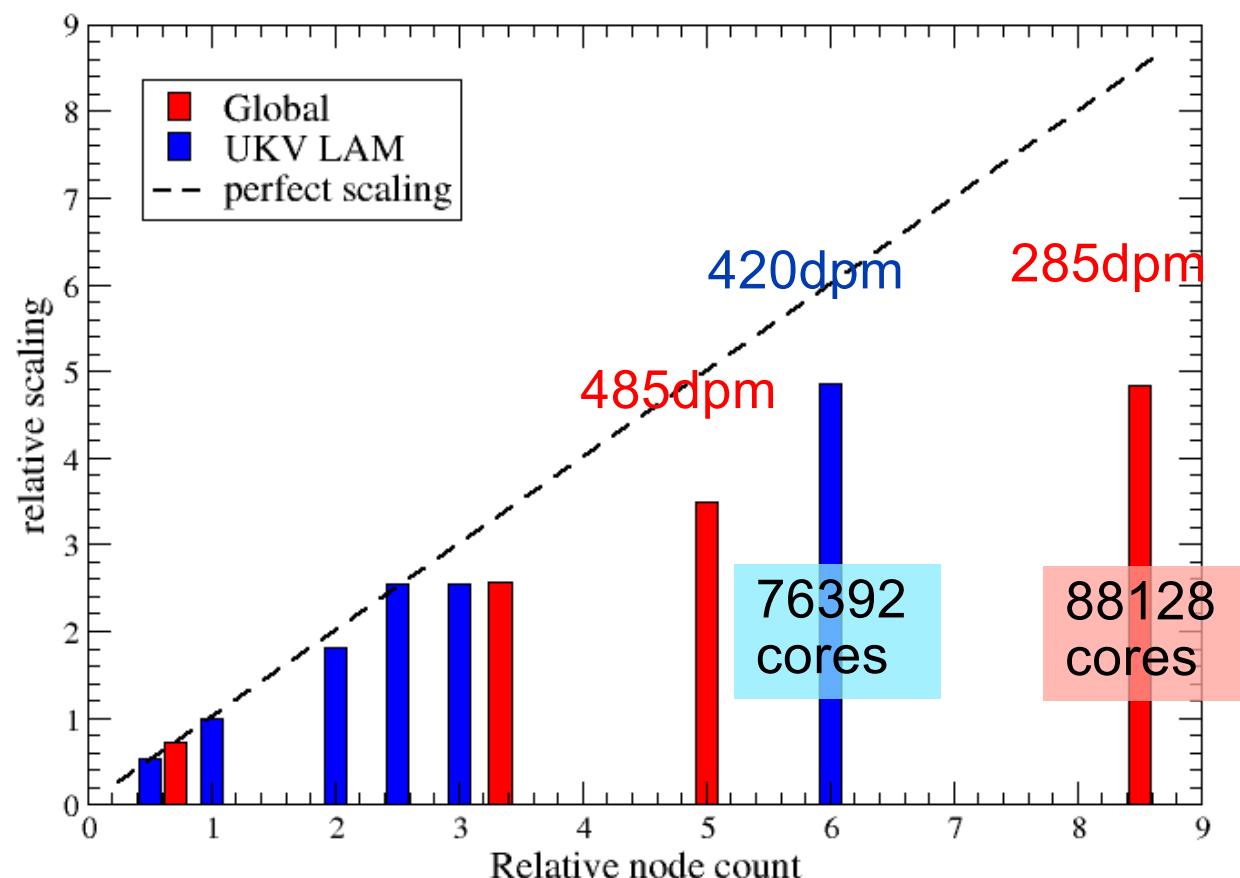
Other models are available!

Comparing relative scaling of Global Model (N2048 ~6km) and Limited Area Model (LAM)

Normalise relative to 2nd datum

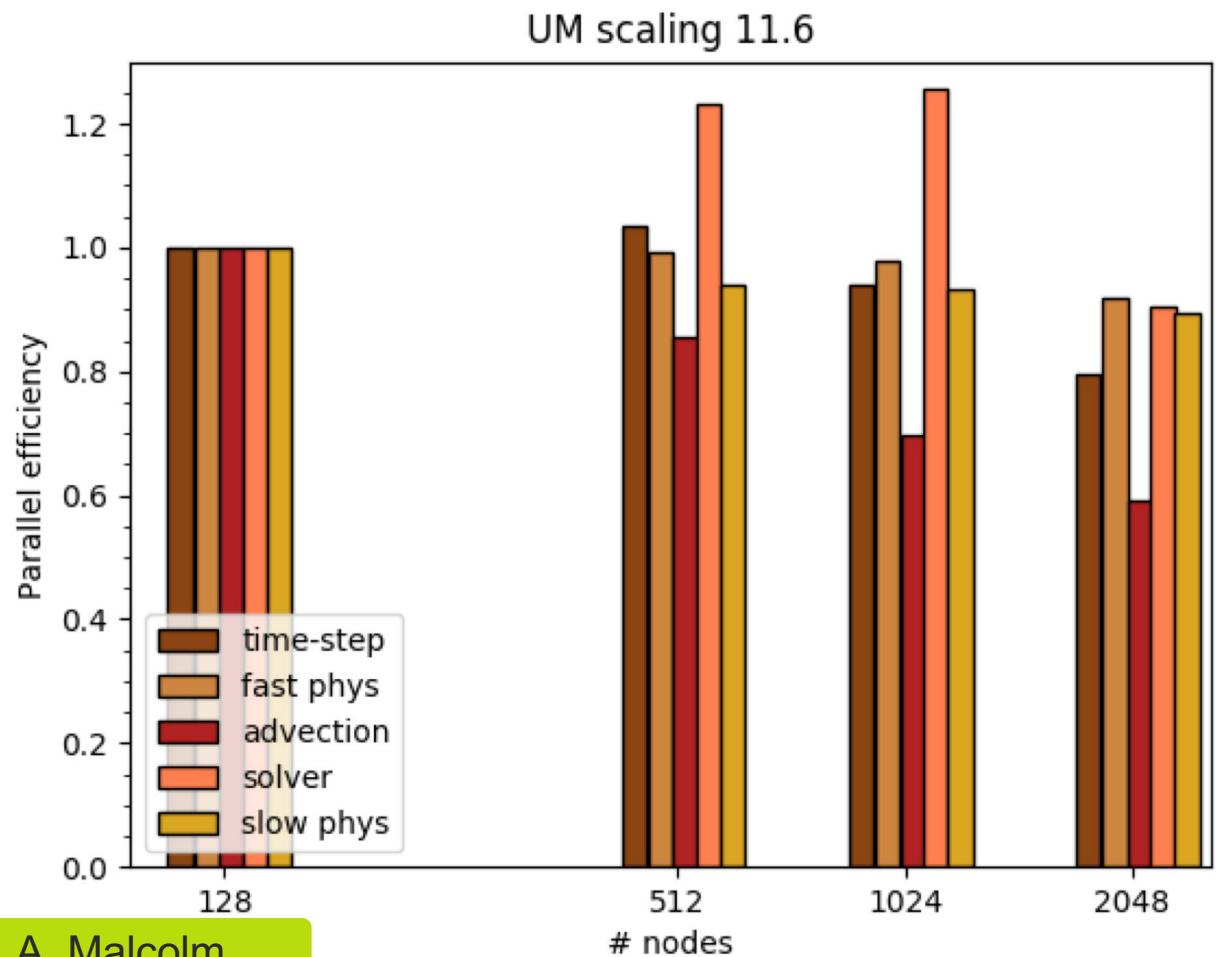
Both have similar points per MPI task ~ 2500
LAM is scaling much better than Global

(Selwood & Malcolm)



$$PE = \frac{T_n/T_0}{N_n/N_0}$$

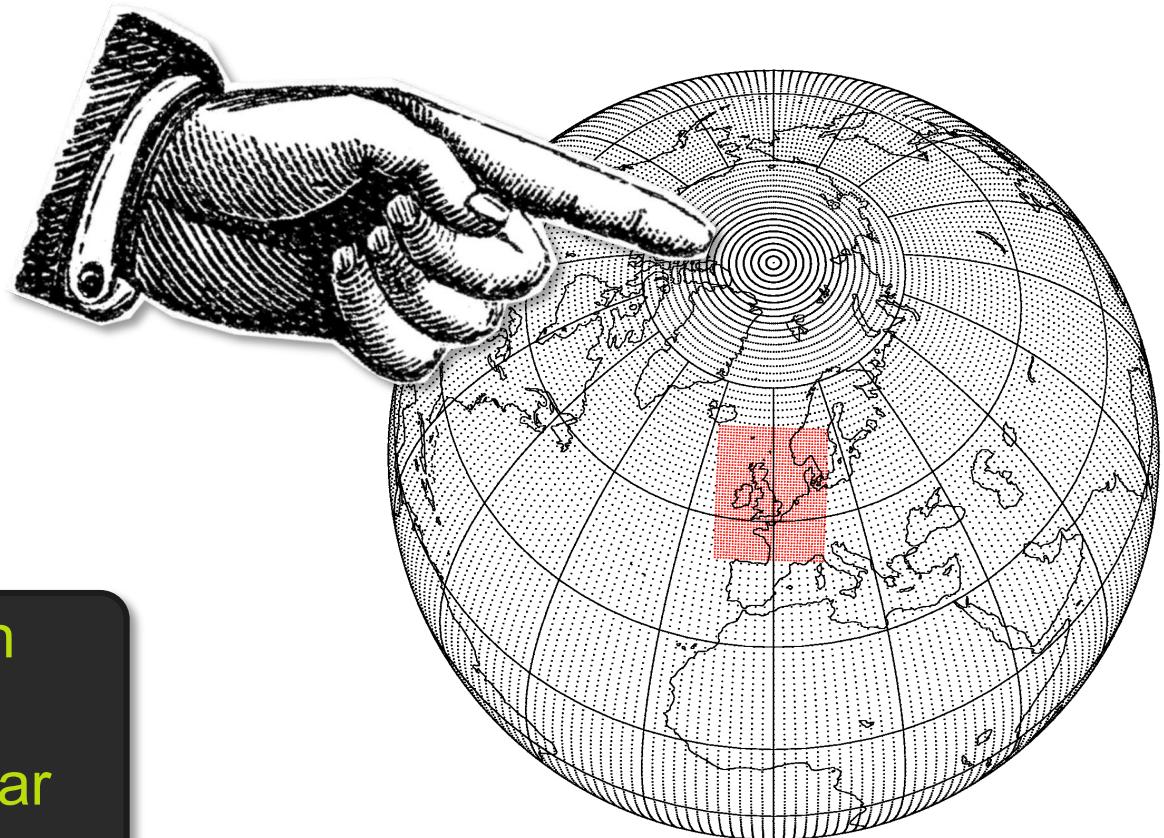
UM11.6 Global N2048 ~ 6km resolution
Met Office XC40
MPI+OMP – 3 (6) threads
Time-step scales OK.
Physics scales well (no comms)
Solver –super-linear (memory)
Advection scales poorly



At 25km resolution,
grid spacing near
poles = 75m

At 10km reduces to
12m!

Semi-Lagrangian Advection
→ Large halos
→ Lots of communication near
poles



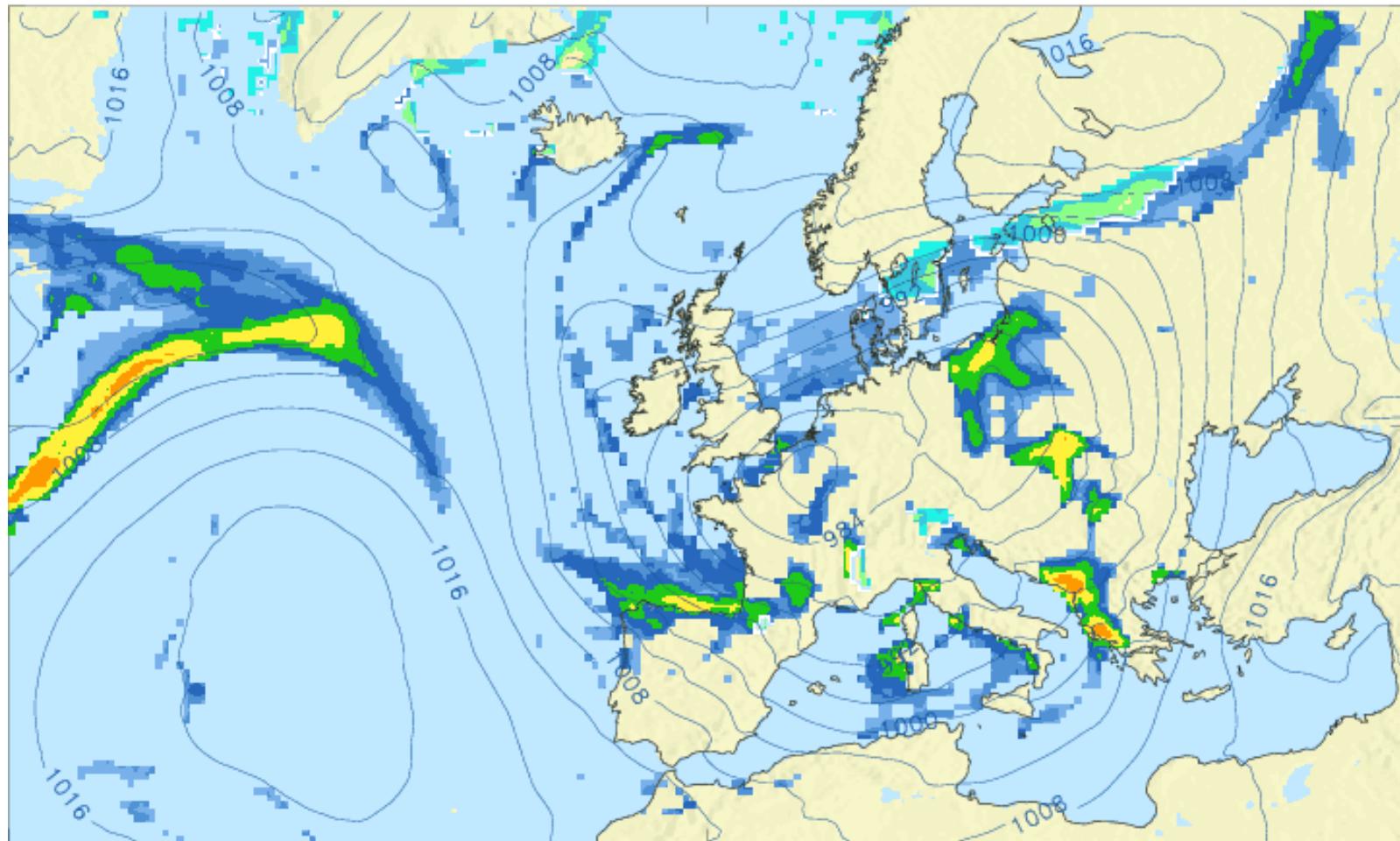


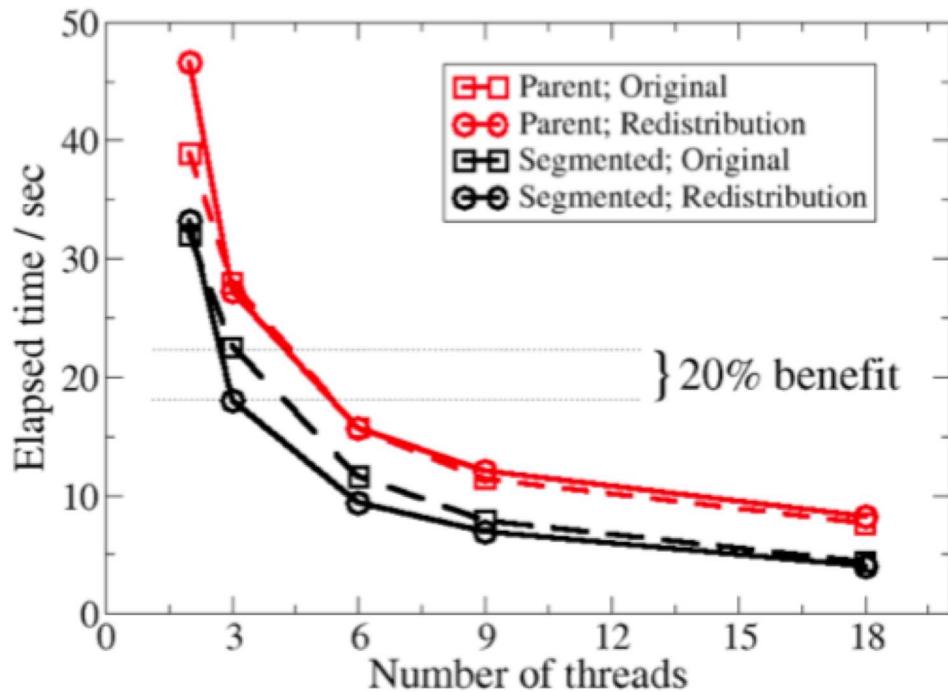
MPI on-node communication is efficient
However, OpenMP reduces the amount of OpenMP required and balance of computation
Glover et al CUG 2016

Code Section	36x60_2 Baseline	72x90_2 MPI	36x60_6 Threads
U_MODEL4A	1792	829	825
ATM_STEP_4A	1571	624	599
AS SOLVER	510	176	158
AS S-L Advect	356	183	153
AS Atmos_phys2	344	121	135
AS Atmos_phys1	283	105	114
INITIAL	209	197	218

Glover et al CUG 2016 time in seconds (MPI E-WxN-S_NOMP)

Solver and Advection both have lots of communication → OMP benefit
Physics has not much, but lots of loops to thread (2016) and potentially poor load balance





Lower is better

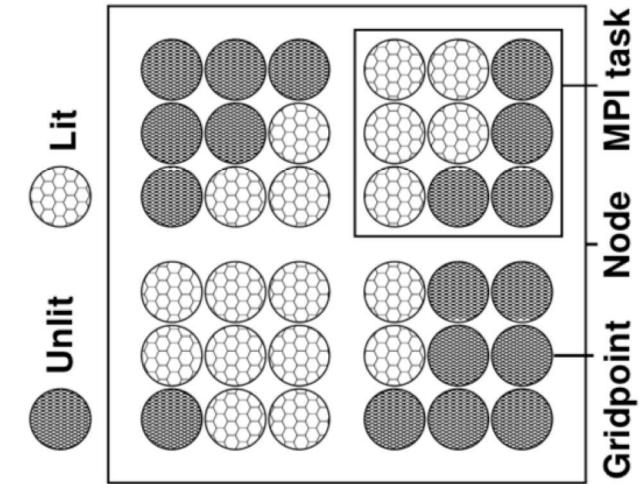


Fig. 6. Illustration of on-node load imbalance. Only lit points require computation.

Lit points have to be determined. Redistribution for load balance (cost), all threads have similar amount of work.

Models produce lots of data.

Higher-resolution means more data.

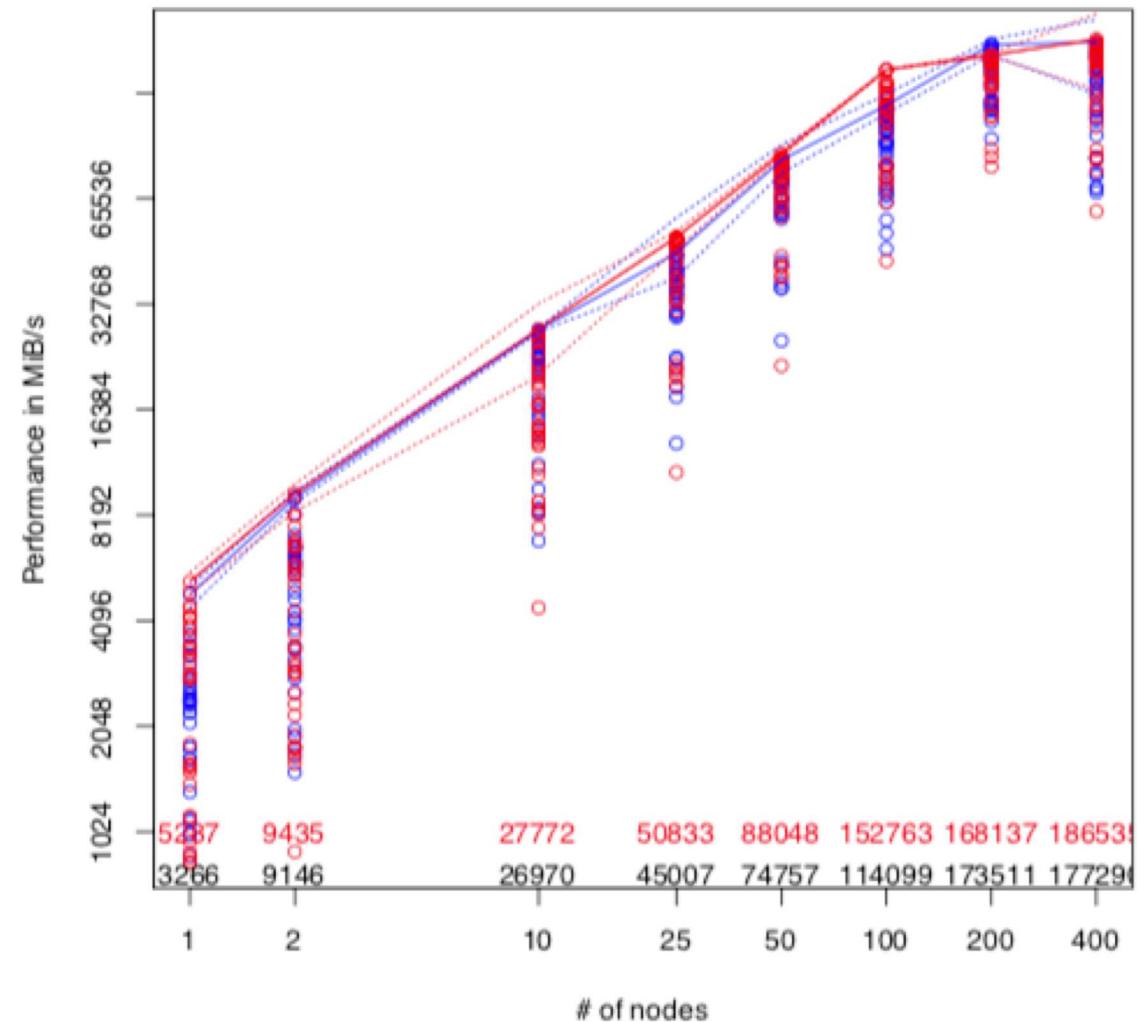
IO Server avoids model computation waiting whilst output is done

Dedicated (MPI) resource to do output only

Most PEs do computation, asynchronous offload of data to IO server resources which write data, whilst computation continues.

How many IO servers to compute PEs depends on machine characteristics, problem size, diagnostics selected, compute speed compared to IO speed.

Plot from JM Kunkel
DKRZ supercomputer
Showing different numbers IO
throughput for different
numbers of clients and servers,
processes per node, tunable
IO parameters, **read/write**
Best performance gives
~6GiB/s per node (small cfg)
~1.5 GiB/s per node (large cfg)

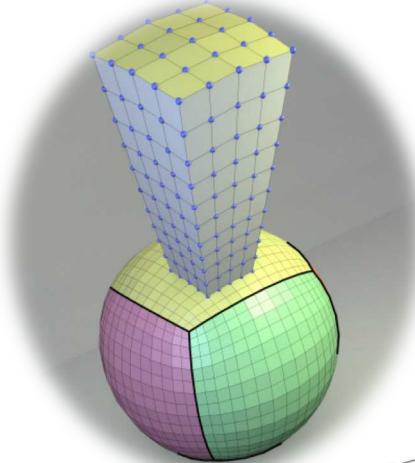


Gung Ho and LFRic

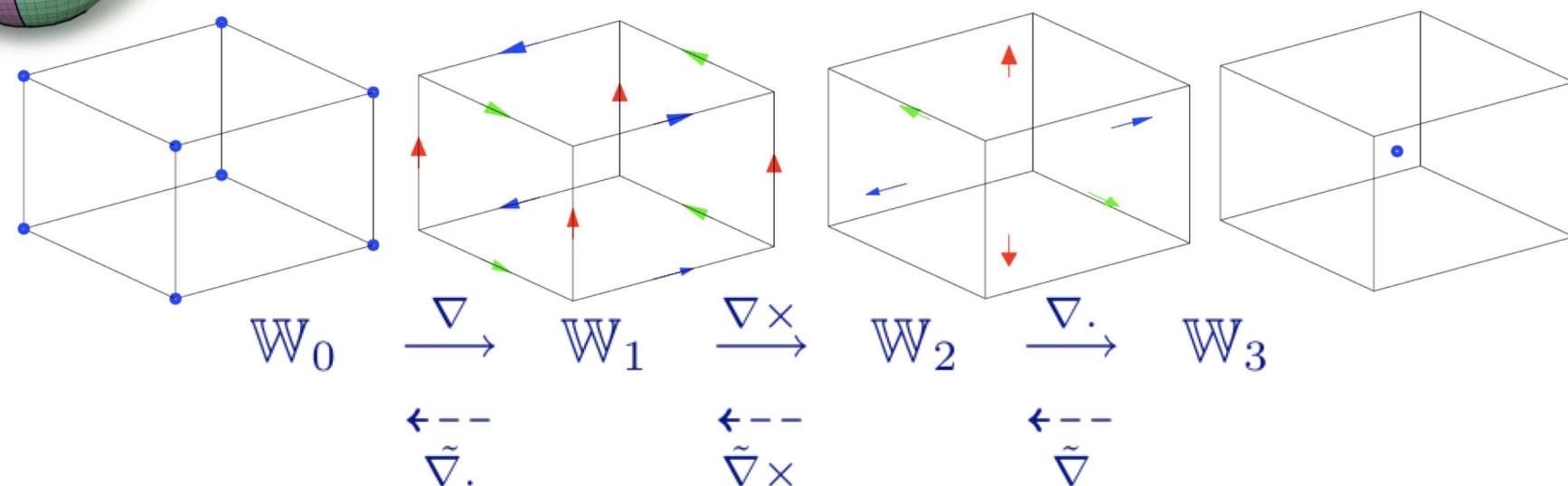
Lon-lat grid will ultimately prevent UM scaling → changing the grid changes everything

LFRic: Adams *et al* JDPC V132 (2019) 383-396 DOI:[10.1016/j.jpdc.2019.02.007](https://doi.org/10.1016/j.jpdc.2019.02.007)

Gung Ho: Melvin *et al* Q J R Meteorol Soc. 2019; 145: 2835- 2853. <https://doi.org/10.1002/qj.3501>



Cubed Sphere → no singular poles lon-lat
Unstructured mesh → can use other meshes
Mixed finite element scheme – C-Grid
Exterior calculus *mimetic* properties
Semi-implicit in time



$$\mathbf{A} \cdot \vec{x} = \vec{b}$$

Many different types

Build an improved solution based on previous

Compute intensive Matrix \times Vector (linear operator – part of the subspace)

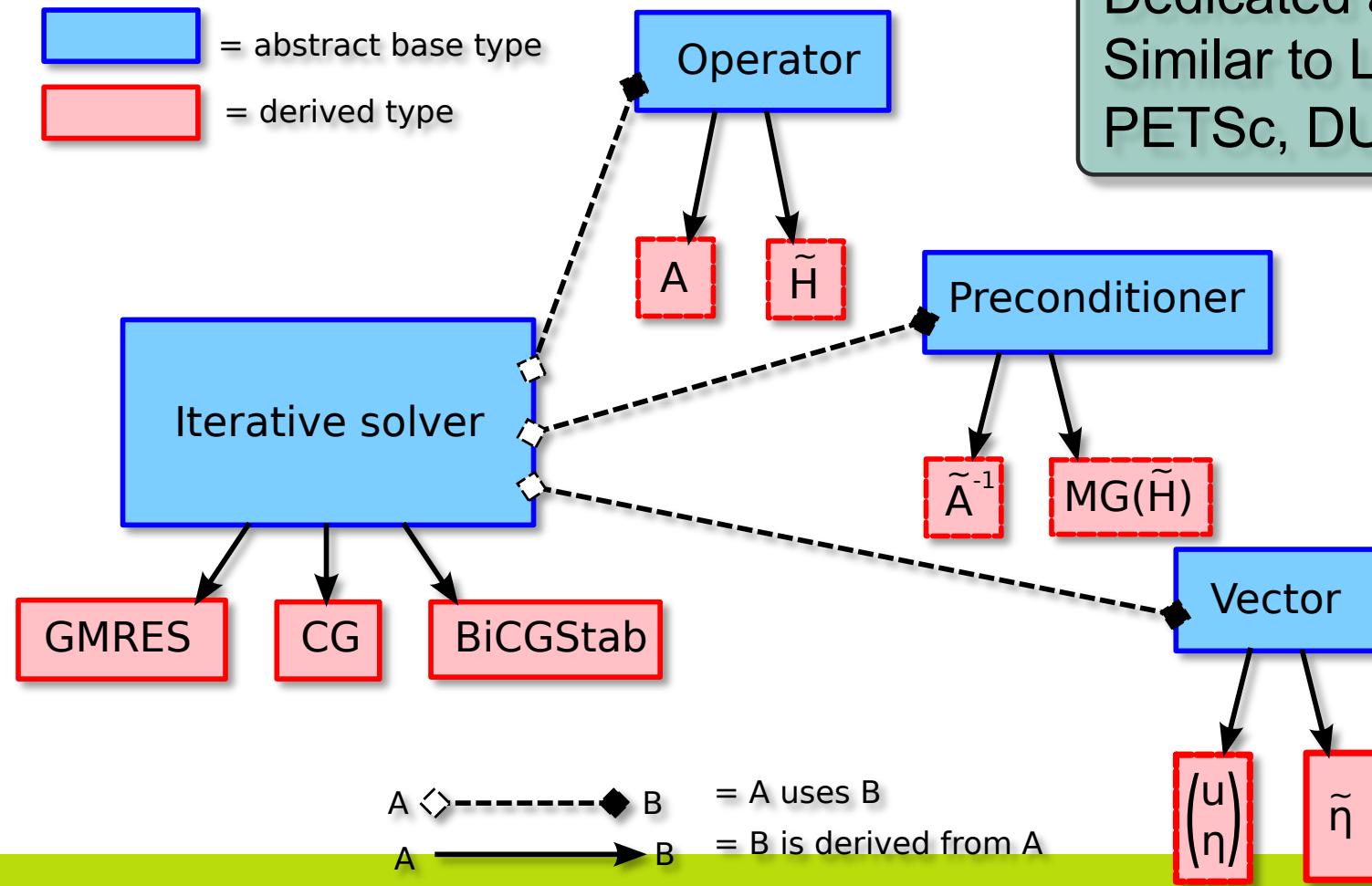
Scalars computed from norms of vectors

Global sum (MPI_Allreduce)

Fewer iterations \rightarrow fewer global sums
preconditioners

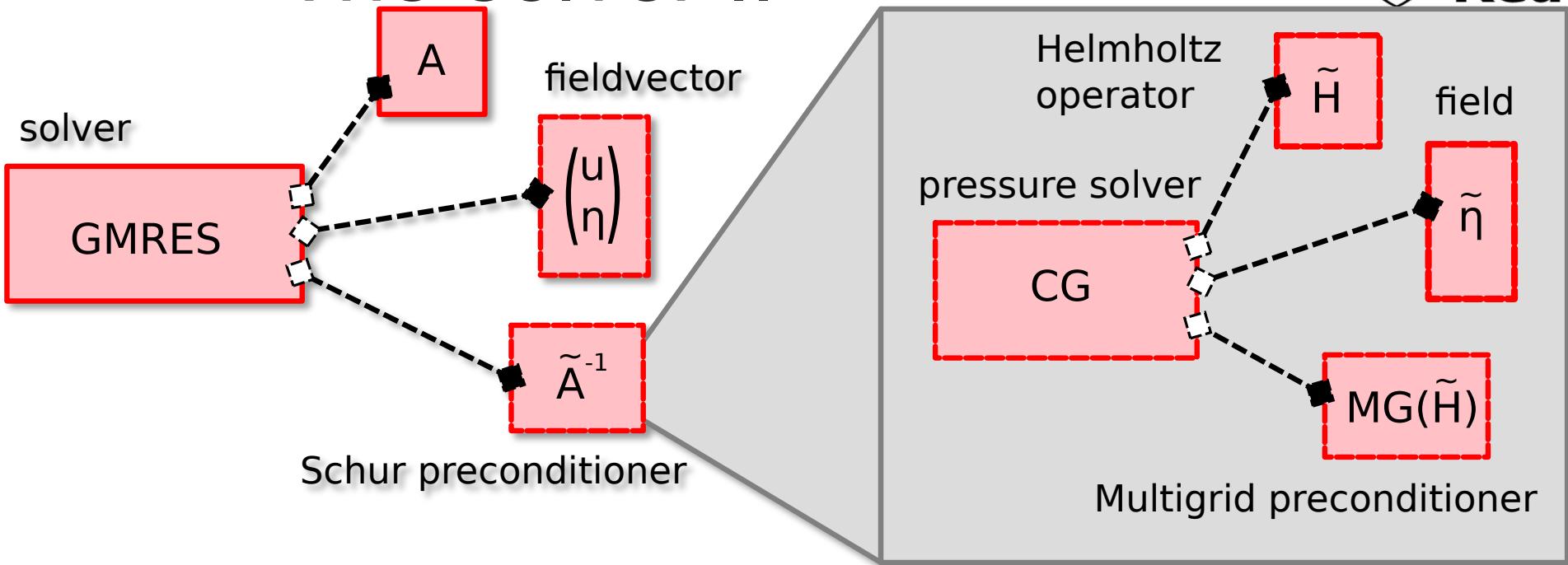
The solver

 = abstract base type
 = derived type



Dedicated abstraction in F2K3 OO
 Similar to Lin. Alg Libs e.g.
 PETSc, DUNE-ISTL, Trillinos

The solver II



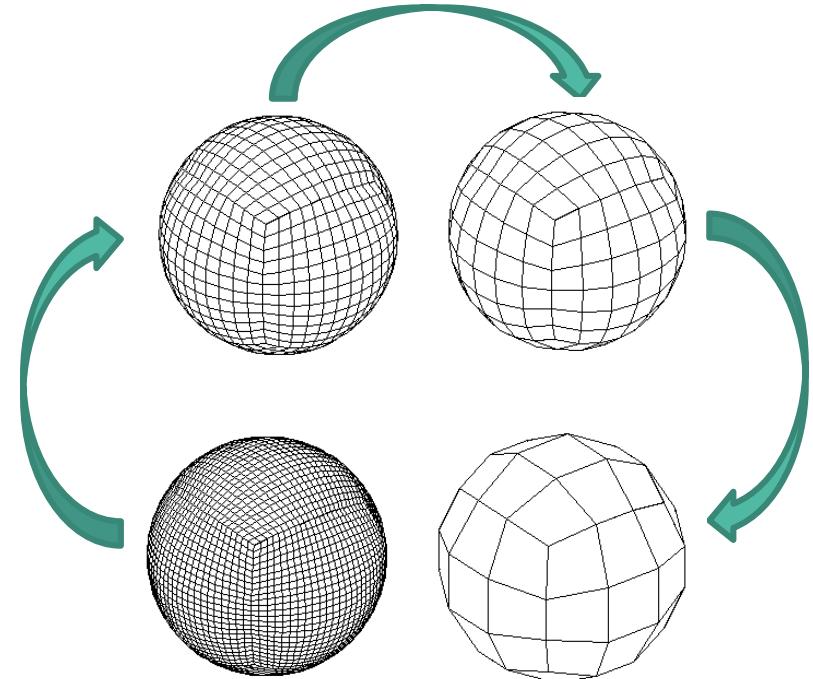
Allows for easy implementation of sophisticated nested solver
Multigrid preconditioner - reduce work for iterative solver
- faster and less global sums (better scaling)

- Helmholtz system $H\Pi' = R$ solved using a single Geometric-Multi-Grid V-cycle with block-Jacobi smoother

$$H = M_3^{\Pi^*} + \left(P_{3\theta}^* \mathring{M}_{\theta}^{-1} P_{\theta 2}^* , z + M_3^{\rho^*} M_3^{-1} D^{\rho^*} \right) \left(\mathring{M}_2^{\mu, C} \right)^{-1} G^{\theta^*}.$$

- Block-Jacobi smoother with small number (2) of iterations on each level
- Exact (tridiagonal) vertical solve: \hat{H}_z^{-1}

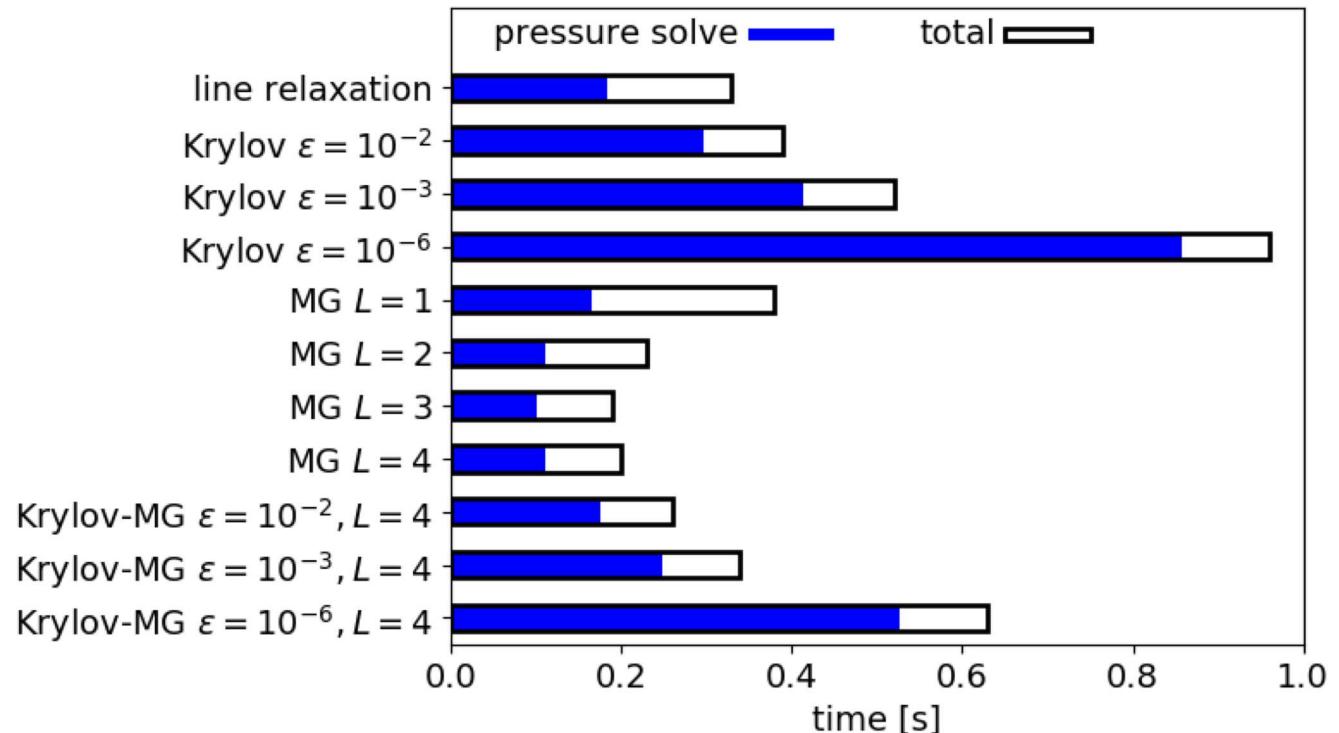
$$\tilde{\Pi}' \leftarrow \tilde{\Pi}' + \omega \hat{H}_z^{-1} \left(\mathcal{B} - H \tilde{\Pi}' \right)$$



Maynard, Melvin and Mueller,
QJR MetS
<https://rmets.onlinelibrary.wiley.com/doi/10.1002/qj.3880>

C192 cubed sphere
with 30 L (~50Km)
Baroclinic wave test
Met Office Cray
XC40 64 nodes
(2304 cores) Mixed
mode 6 MPI/6 OMP
threads

c.f. $\|r\| = \|\mathbf{A}x - b\|$ Of
Krylov 10^{-2}
Before and after MG
3-level V-cycle



C1152 mesh → 1152 X 1152 X 6 mesh with 30 levels – 9Km resolution

Dynamics only, (Baroclinic Wave)

400 time-steps. $\Delta t = 205$ s CFL_H (acoustic) ~ 8

Intel 17 compiler

6 MPI ranks per node, 6 OpenMP threads per rank

384 nodes (13824 cores) – 3456 nodes (124416 cores)

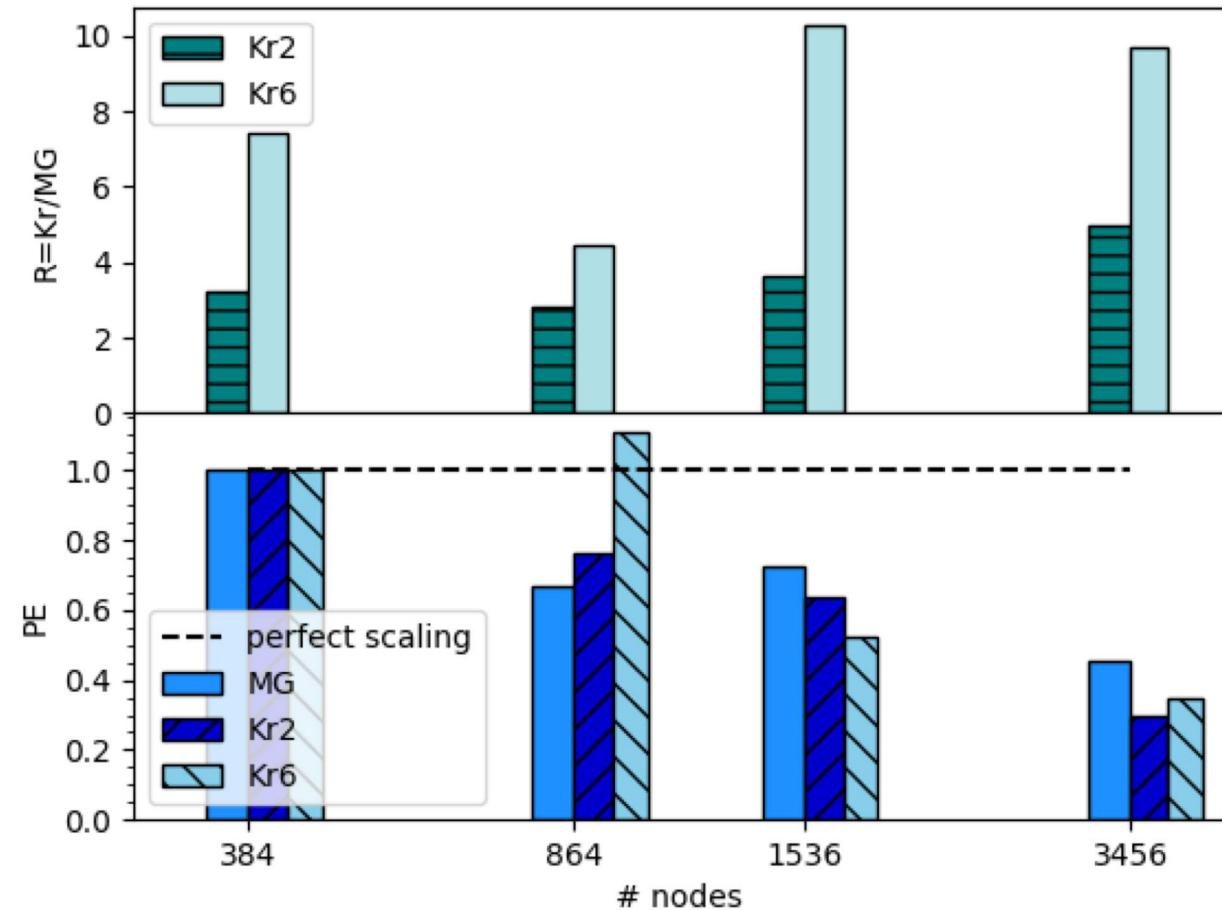
Data per PE 24x24, 16x16, 12x12, 8x8

MG is 3-levels of MG inner solve is preconditioner only

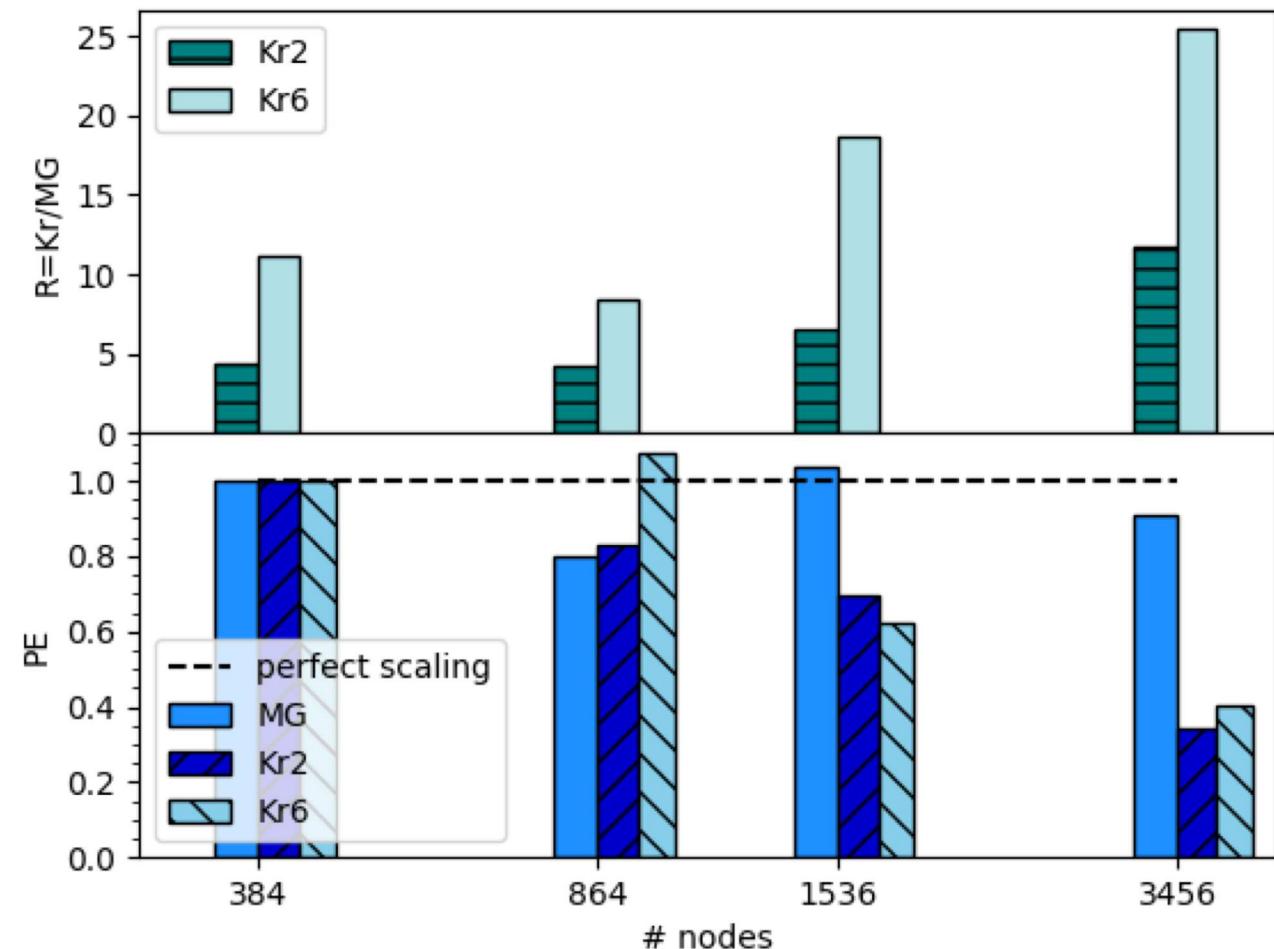
KR2 is $\|r\|=10^{-2}$

KR6 is $\|r\|=10^{-6}$

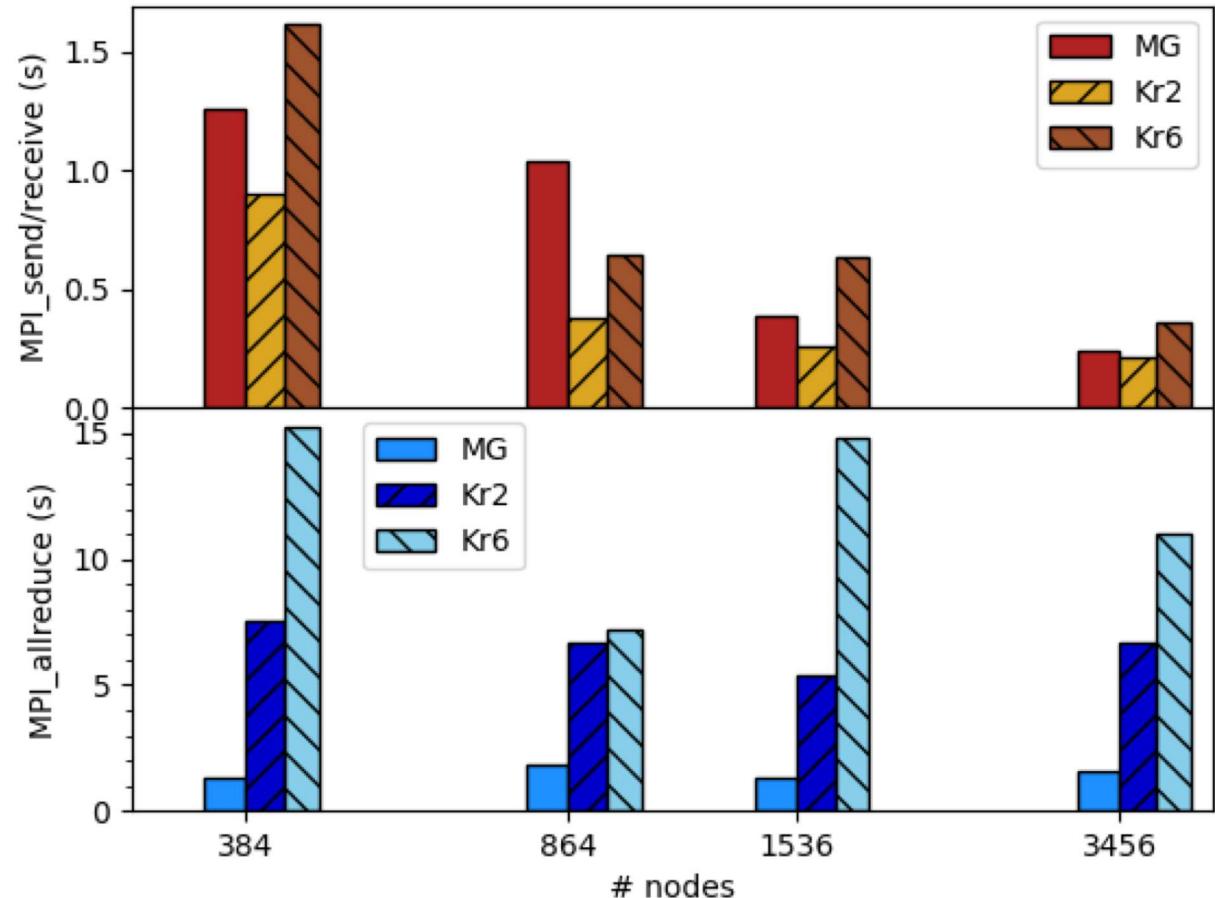
Bottom panel is parallel efficiency
1 is perfect scaling
Top panel shows relative cost of KR solvers *c.f.* MG
(higher is better)



Bottom panel is parallel efficiency
1 is perfect scaling
Top panel shows relative cost of KR solvers *c.f.* MG
(higher is better)
MG is much faster and scales much better



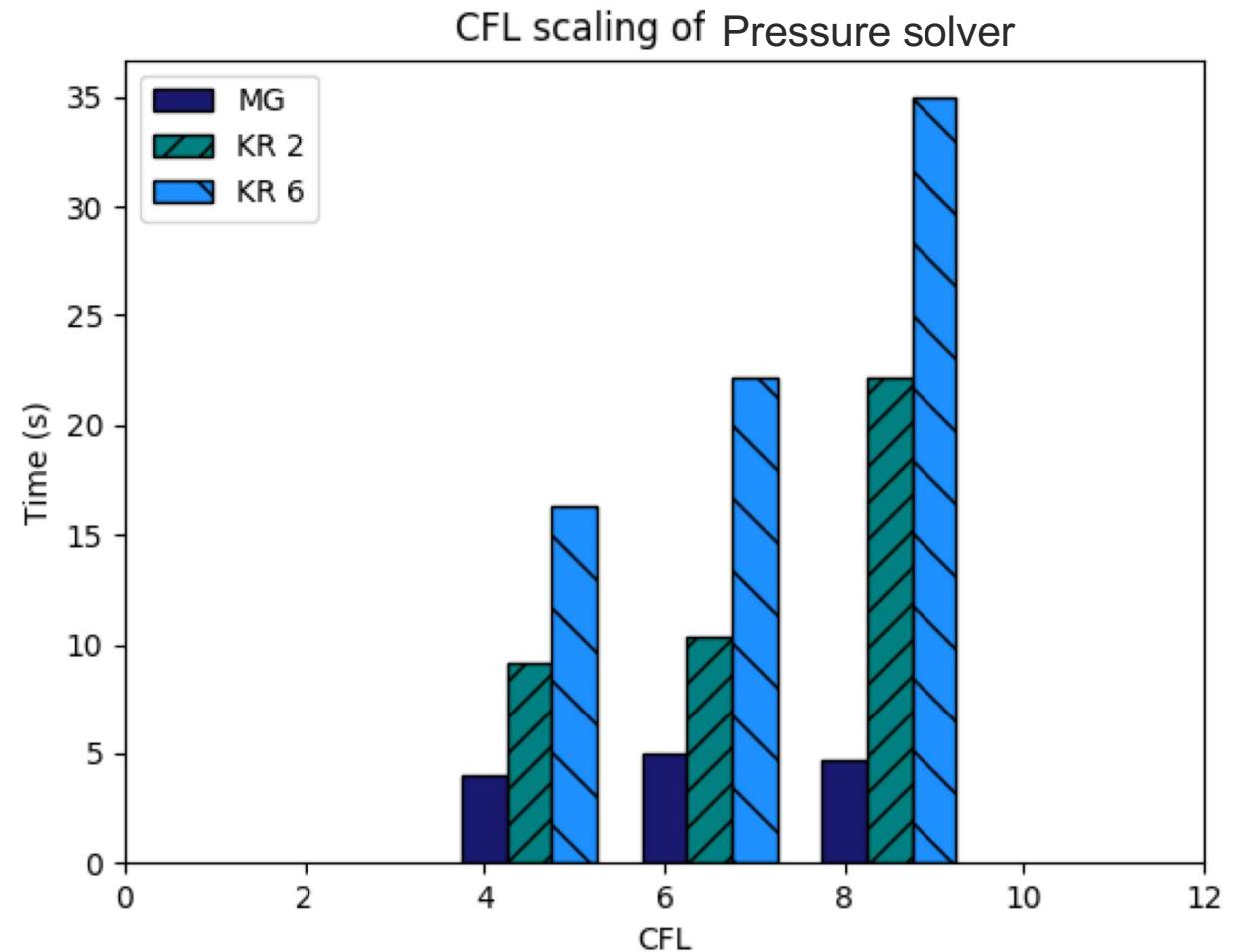
Time per time-step in communication
Bottom panel is MPI allreduce (global sum) cost
Massive reduction for MG
Upper panel shows local comms scale with data size



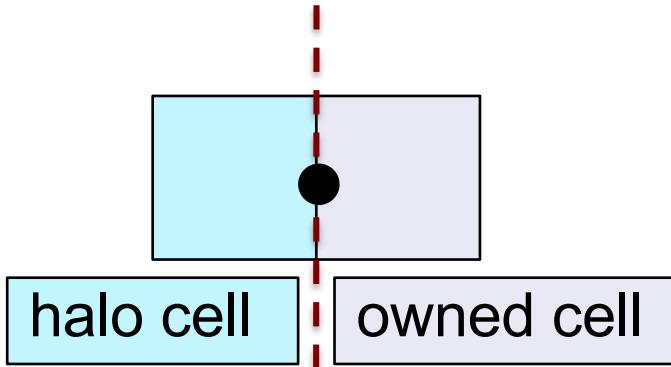
Size of time-step

CFL _h	Solver	# iter		$T_{\text{solve}}^{(m)}$
		mixed	pressure	
4	MG	12.9	—	3.96
	Kr2	13.2	9.4	9.13
	Kr6	12.0	26.2	16.29
6	MG	13.6	—	4.98
	Kr2	13.3	13.2	10.31
	Kr6	12.3	40.4	22.16
8	MG	14.0	—	4.70
	Kr2	13.3	17.3	15.15
	Kr6	12.6	54.2	34.96

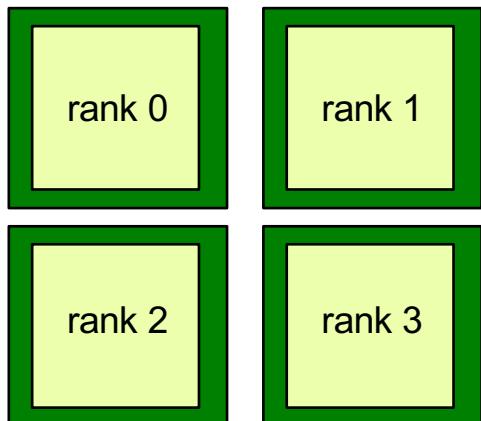
As time-step increases, condition number of matrix increases
→ more iterations
Multigrid V-cycle is fixed cost
As long as solution is good enough, no extra cost to increase time-step



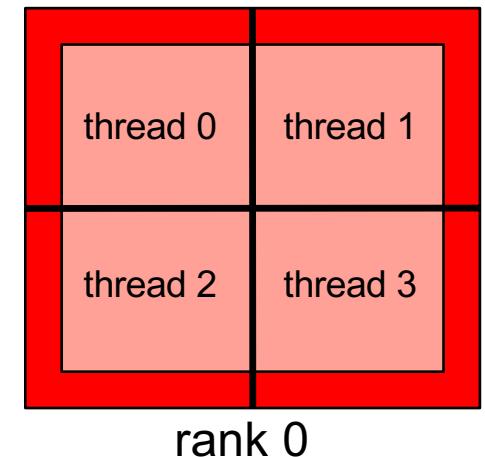
Redundant computation

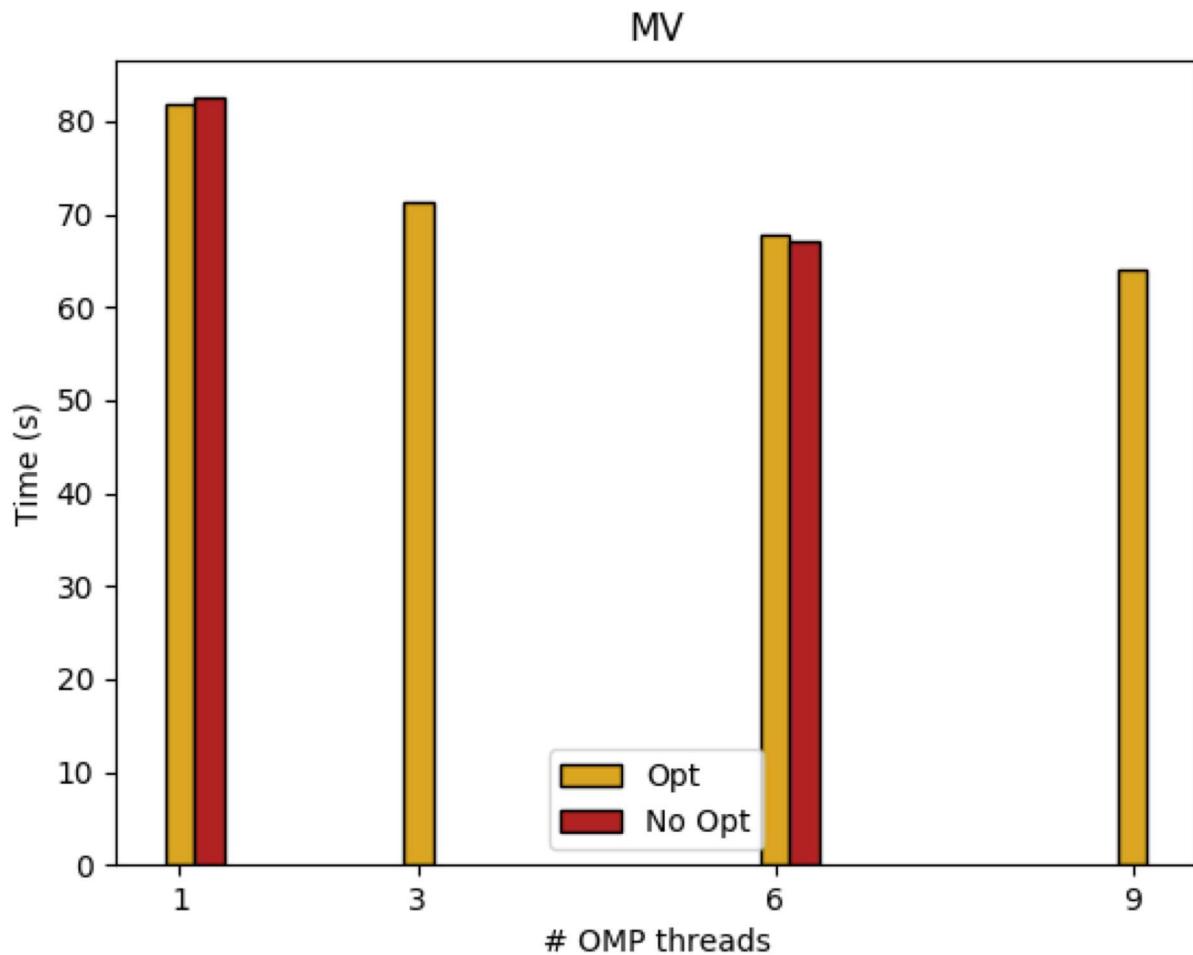


Dof living on shared (partitioned) entity (edge).
Receive contribution from owned and halo cell.
Redundant compute contribution in halo to shared dof.
Less communication

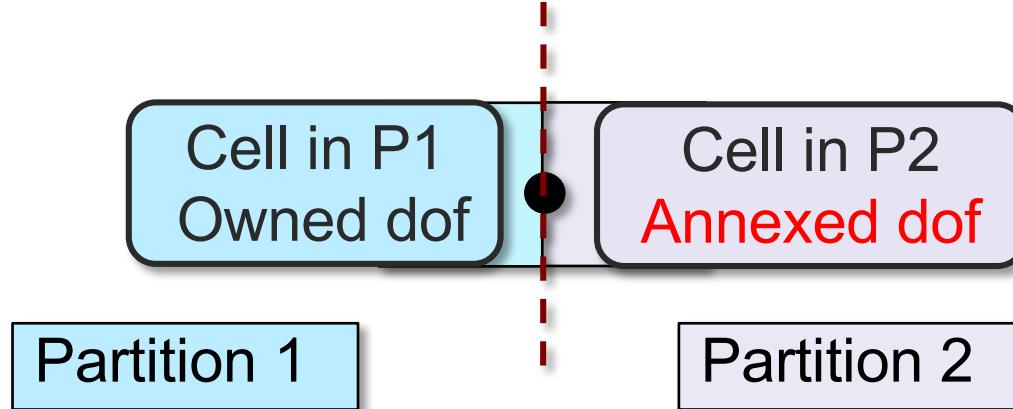


MPI only, 4 MPI ranks all have halos
Hybrid, 1 MPI task has a halo, 4 OpenMP threads share halo
boundary-to-area scaling
→ Less work for OpenMP threads





C288 mesh, 96 nodes
1 OMP thread is 36 MPI ranks per node
9 OMP threads is 4 MPI ranks per node
Redundant computation favours more threads
(opt/No Opt is MPI comms env variable – not relevant here)



Point-wise computations (e.g. set field to a scalar) loop over dofs
Looping to owned dofs → halo exchange required for P2
Looping to annexed dofs is now transformation in Psyclone
Small increase in redundant computation
Large reduction in number of halo exchanges required

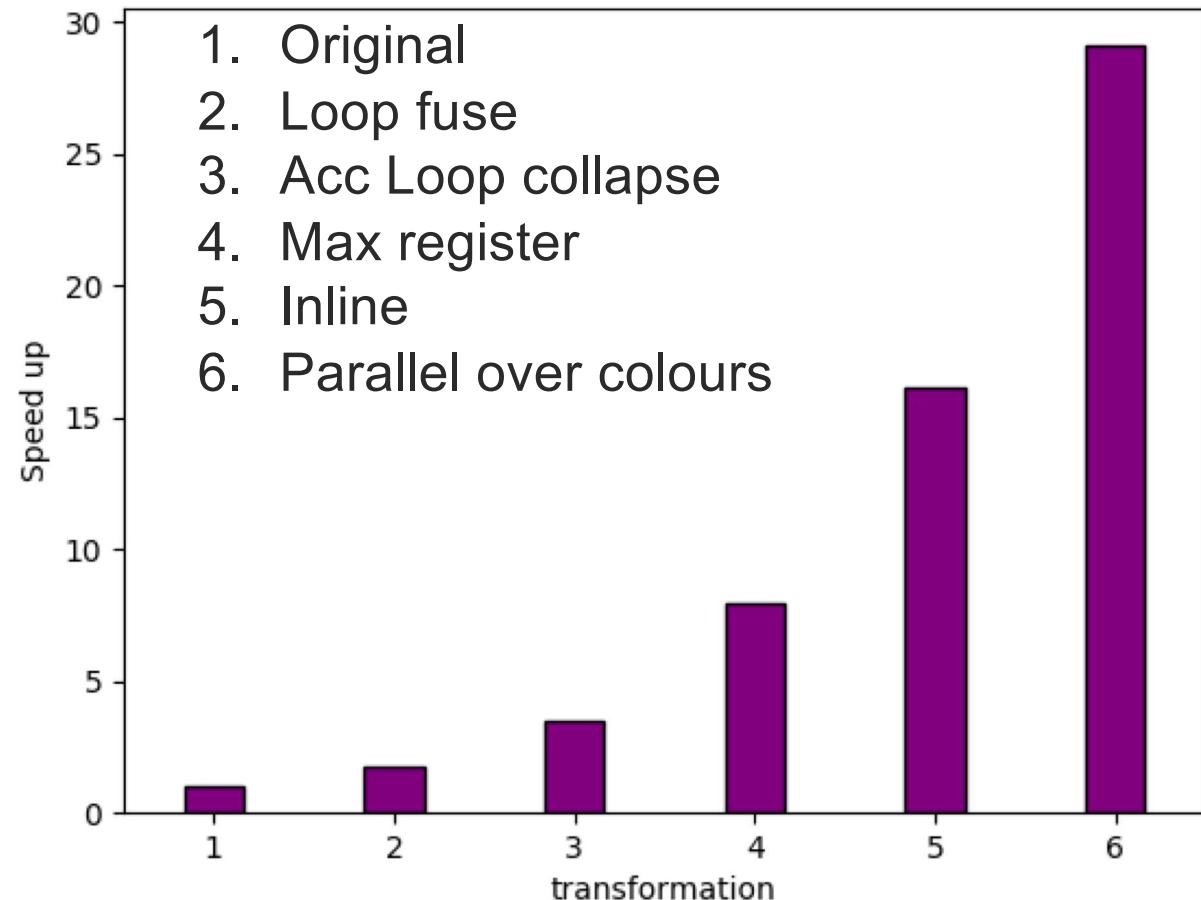
GPUs and other animals

Distinct memory spaces
Data transfer/synchronisation
ILP

Exploited data parallelism in horizontal for CPUs
Dynamics kernels tend to have limited data dependency in vertical
SIMD/SIMT GPU vector across vertical dofs – 128+ levels
Physics kernels often have dependency in vertical ...
But have extra dofs, e.g. radiation bands
Exploiting CPU and GPU together with hard to synchronise
Simpler to reduce data movement and compute on GPU only



A. Gray (NVIDIA)
Cumulative speed up
against original
OpenACC code.
Problem size is too
small for GPU.
Amortise cost of data
movement by
offloading multiple
kernels

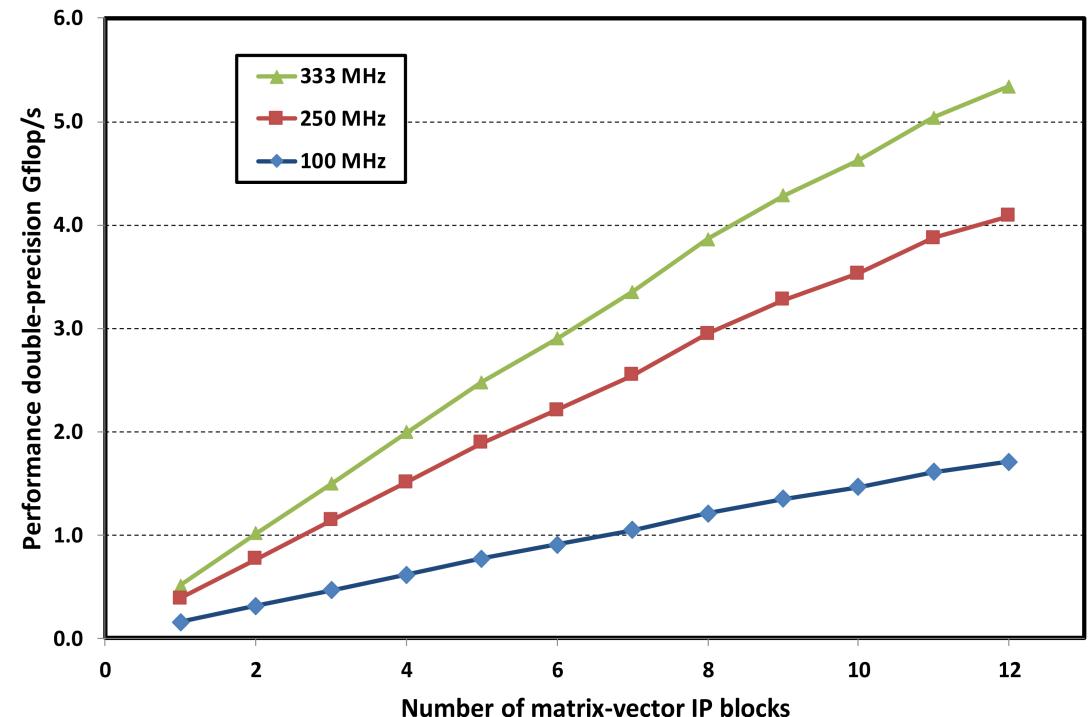


EuroExa project: ARM CPUs + FPGA accelerator prototype → low power
LFRic one of several applications. Mike Ashworth Uni of Manchester

Ported using High-level
Synthesis tool from Xilinx
Vivado.

Graph shows scaling versus IP
block and clock speed.
Max 5.3 GFlop/s in double
precision.

Comparable to CPU and GPU.
Significant benefits considering
power.



Conclusions

End of the *free lunch* – no faster processors → exploit ever more parallelism

Mathematics of problem dictates what can be computed in parallel

Choice of how to solve mathematics for weather and climate

Leading to different parallel algorithms and implementations

Interplay between implementations and parallel algorithms

Scaling of algorithmic components of time-steps

Newer architectures require exploitation of more parallelism

Extra slides

If there is time



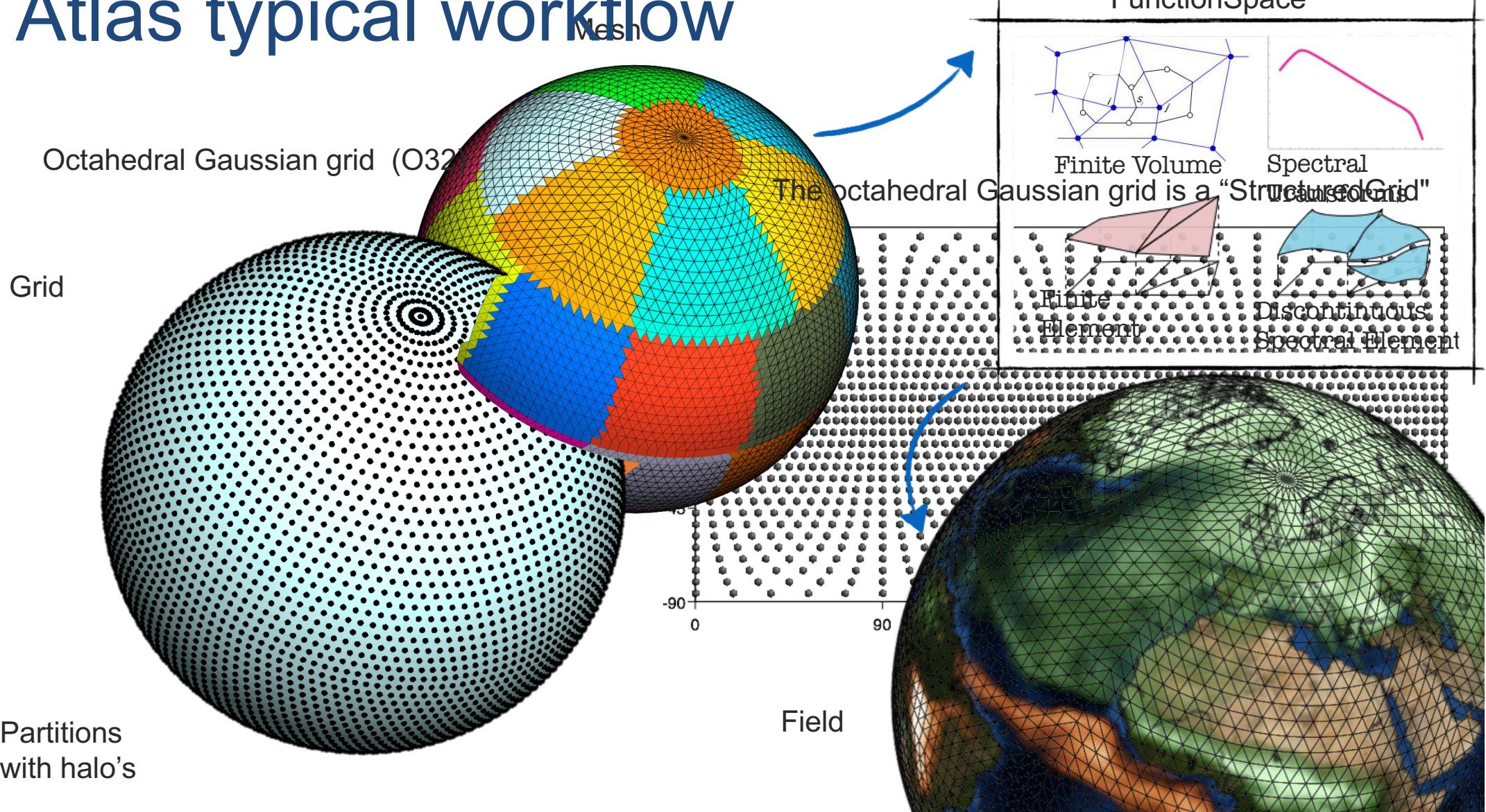
Gung Ho/LFRic/Psyclone **Replace UM**
UK Met Office + STFC + NERC
New FEM dynamical core +
infrastructure
Code generator (automatic parallelism)
DSEL (Fortran)

Gridtools/Stella **Rewrite**
MeteoSwiss/CSCS (Cosmo/
ICON)
Finite difference/structured mesh
Initially GPU code
DSEL (C++)

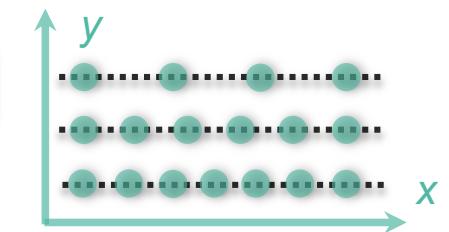
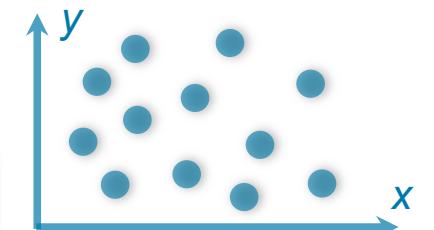
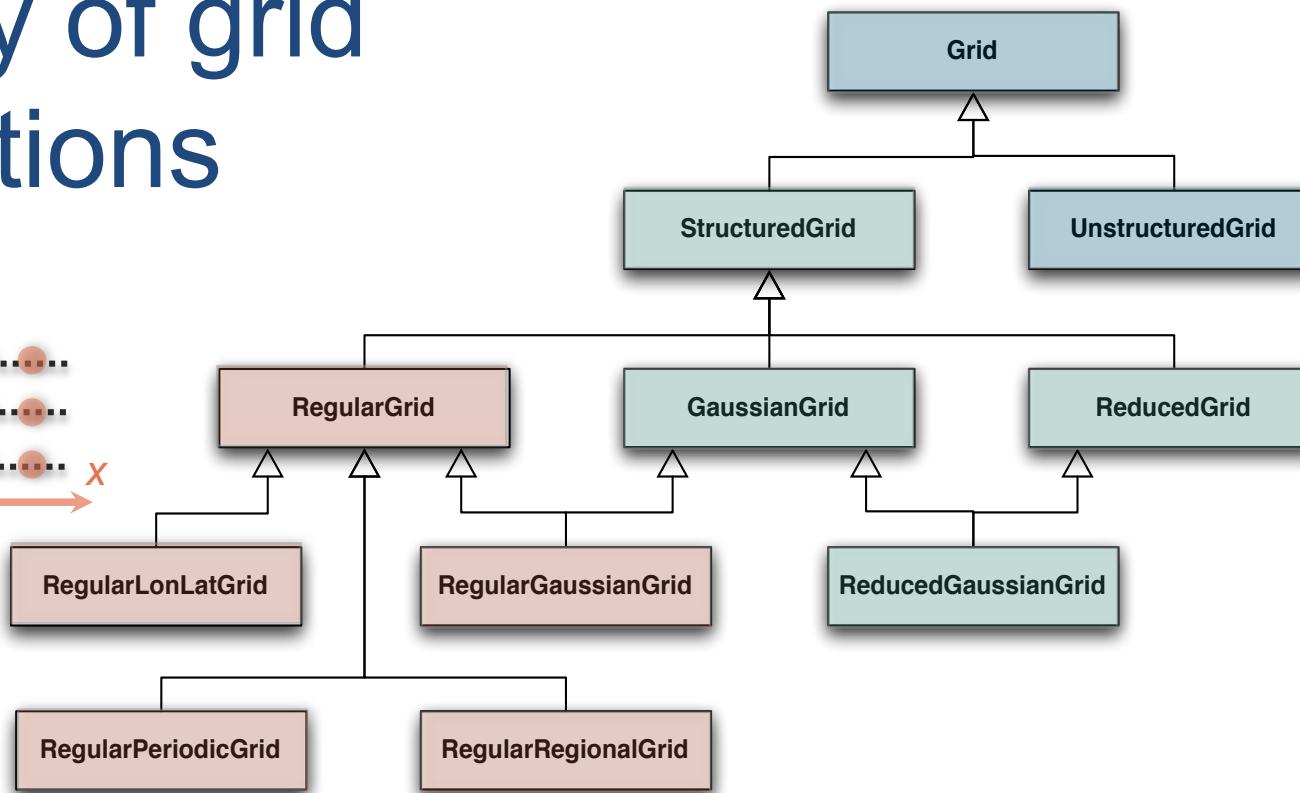
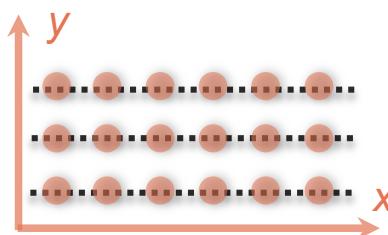
Atlas Library/Framework **Change Alg**
ECMWF
C++ with Fortran 2008 wrappers
Support for different grids structured
and unstructured
Methods FD/FV/FEM

ESCAPE (ECMWF + lots of
partners weather/climate and
vendors)
Extract computational patterns
(Dwarfs or mini-apps)
Explore optimisation space

Atlas typical workflow



A library of grid abstractions



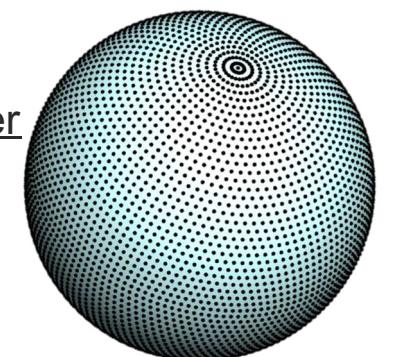
Example creation of operational octahedral reduced Gaussian grid using unique identifier

C++

```
atlas::Grid grid;
grid = atlas::Grid( "O1280" )
```

Fortran

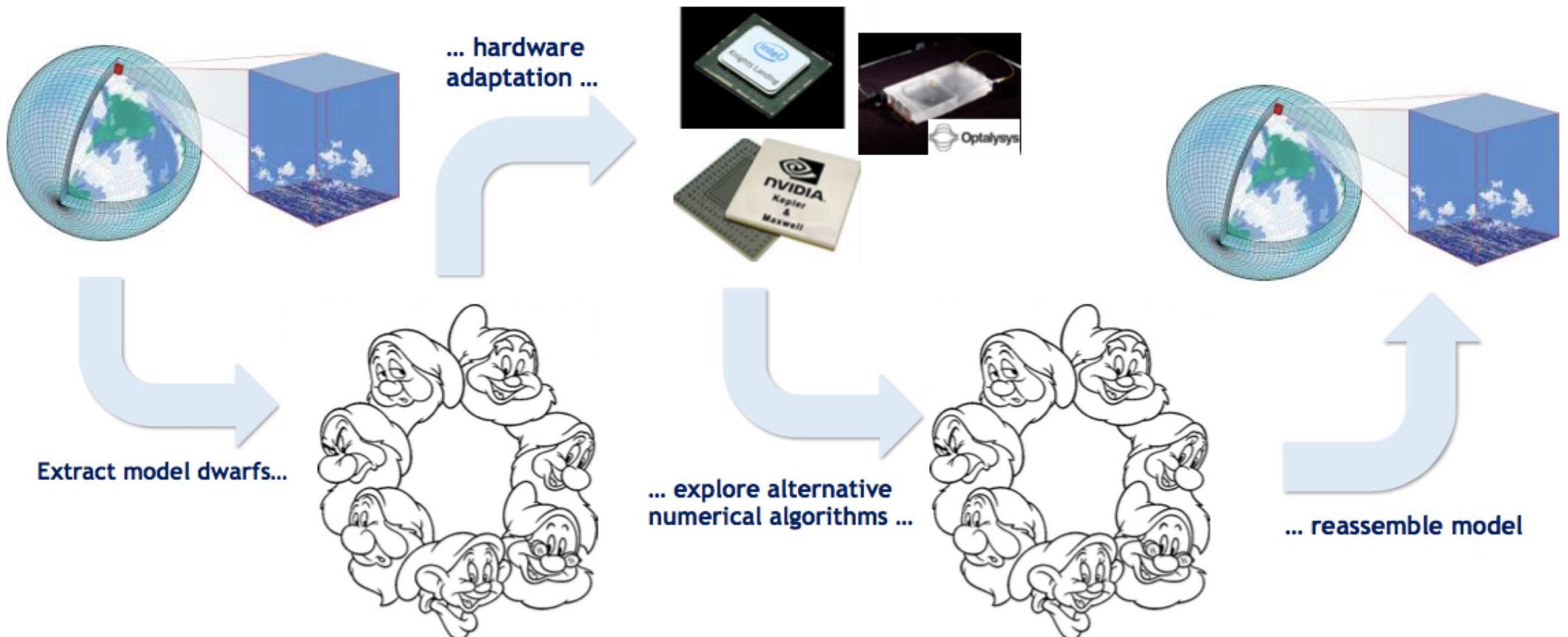
```
type(atlas_Grid) :: grid
grid = atlas_Grid( "O1280" )
```





Funded by the
European Union

Idea behind ESCAPE



Optimization of spectral transform dwarf

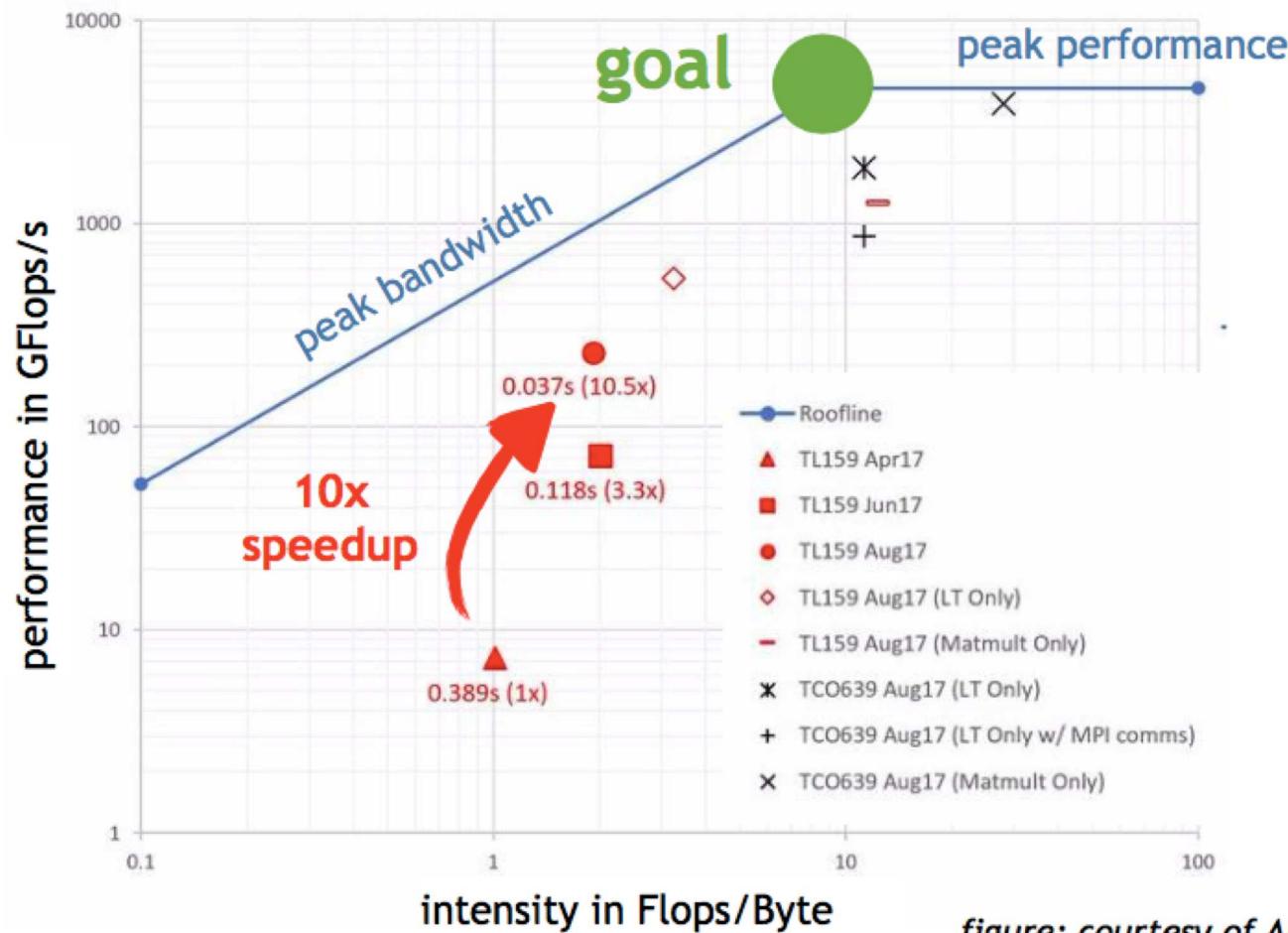
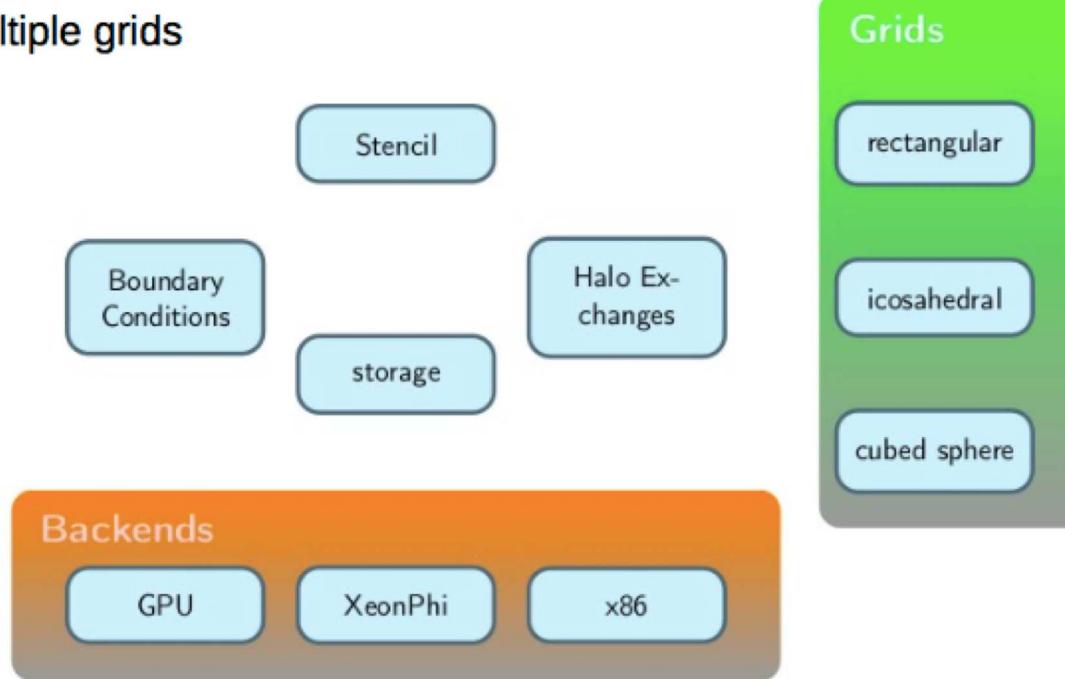


figure: courtesy of Alan Gray, Peter Messmer (NVIDIA)



GridTools

- Set of grid tools, including DSL for stencil codes, for solving PDEs on multiple grids



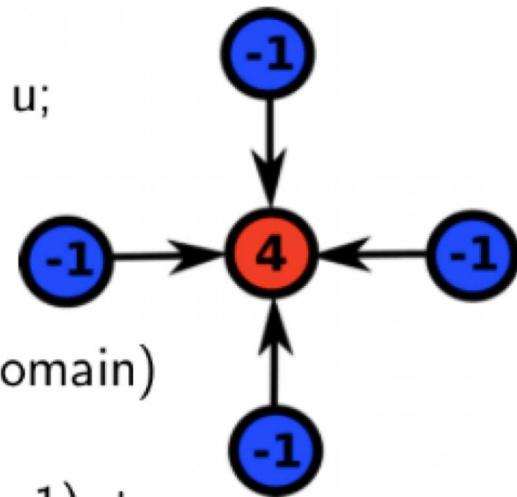
- Provides separation of concerns: Separates model and algorithm from hardware specific implementation and optimization
- Supports multiple hardware and grid backends.



Encoding Stencil Information in Types

```
struct Laplace
{
    typedef in_accessor<0, range<-1,1-1,1> > u;
    typedef out_accessor<1> lap;

    template<typename Evaluation>
    static void Do(Evaluation const& eval, full_domain)
    {
        eval(lap()) = eval(-4*u() + u(i+1) + u(i-1) +
                           u(j+1) + u(j-1));
    }
};
```





Co-design: Extending Collaborations

