The Impact of the Changing Nature of Computing on Climate Science

Bryan Lawrence & a cast of thousands

NCAS &

University of Reading: Departments of Meteorology and Computer Science

UoR, 21 Jan 2020



- An introduction to climate modelling ...
- and the data handling workflow.
- The JASMIN super data computer, and some examples of JASMIN cloud usage.
- The end of Moore's Law
- What next? Maths, computer science, and some of our research directions.





The reality of climate change

Context •oo



A little more than 1C global warming is not manifesting as slightly nicer summers and slightly warmer winters!





The Greenhouse Effect: How did we get here?

Not a new idea: Arrhenius 1896



On the Influence of Carbonic Acid in the Air upon the Temperature of the Ground, Philosophical Magazine and Journal of Science, 1896: Doubling atmospheric carbon dioxide will lead to an increase in surface temperatures of 5-6K.

Building on Tyndall, c 1859

The Earth's atmosphere is warmer than it should be (in terms of the radiative heat input from the sun).



Tyndall explained the heat in the Earth's atmosphere in terms of the capacities of the various gases in the air to absorb radiant heat, in the form of infrared radiation (from the earth radiating outward).



Context 000



We have not wasted the following hundred years

IPCC reports

Context

Five assessment reports (1990, 1995, 2001, 2007, 2013-14)

1992 supplementary report and 1994 special report

Nine special reports (1997, 1999, 2000, 2005, 2011, 2012)

Guidelines for national GHG inventories, good practice

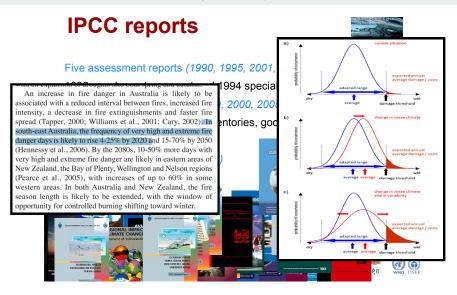
guidance (1995, 2006, 2013) Six technical papers (1996-2008)



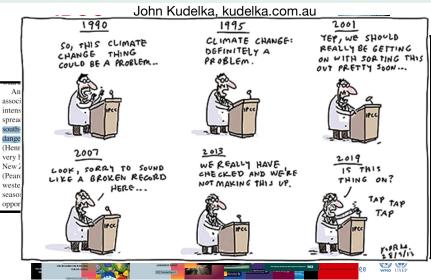


We have not wasted the following hundred years

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We have not wasted the following hundred years





Context



We want to simulate our world

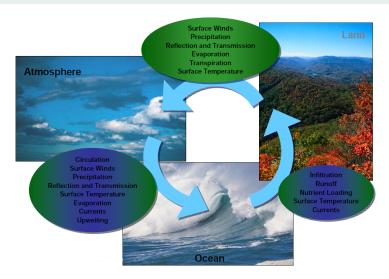


Image: from J. Lafeuille, 2006





Basic Fluid Equations (for the atmosphere)

State Variables:

u, v, w - wind

□ Exner function

(non-dimensional pressure)

Climate Modelling 0000000000000

Coordinates:

 r, ϕ, λ — radial position, latitude. Iongitude

Things that cause change: $\frac{D}{Dt}$ – time derivative following

motion S — External Forcing (radiative)

heating etc)

Newton's second law

$$\frac{D_r u}{Dt} - \frac{uv\tan\phi}{r} - 2\Omega\sin\phi v + \frac{c_{\rm pd}\theta}{r\cos\phi} \frac{\partial\Pi}{\partial\lambda} = -\left(\frac{uw}{r} + 2\Omega\cos\phi w\right) + {\rm S}^u$$

$$\frac{D_r v}{Dt} + \frac{u^2 \tan\phi}{r} + 2\Omega \sin\phi u + \frac{c_{\rm pd}\theta}{r} \frac{\partial \Pi}{\partial \phi} = -\left(\frac{vw}{r}\right) + S^v$$

$$\frac{D_r w}{Dt} + c_{\rm pd} \theta \ \frac{\partial \Pi}{\partial r} + \frac{\partial \Pi}{\partial r} = \left(\frac{u^2 + v^2}{r}\right) + 2\Omega {\rm cos} \phi u + {\rm S}^w$$

mass continuity

$$\frac{D_r}{Dt} \left(\rho_{\rm d} r^2 \cos \phi \right) + \rho_{\rm d} r^2 \cos \phi \left[\frac{\partial}{\partial \lambda} \left(\frac{u}{r \cos \phi} \right) + \frac{\partial}{\partial \phi} \left(\frac{v}{r} \right) + \frac{\partial w}{\partial r} \right] = 0$$

thermodynamics

$$\frac{D_r \theta}{Dt} = S^{\theta}$$

Objective is given knowledge of the external forcing S and the state (u, v, w, Π, Θ) at time t, to advance knowledge of the state variables to time $t + \Delta t$, where Δt is the **timestep**.

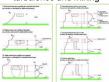




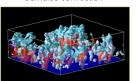
Many of the forcing terms come from parameterisations

Slide Images from Slingo, 2013

Boundary layer turbulence and mixing



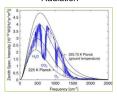
Cumulus convection



Effects of mountains



Radiation



Precipitation



Clouds and microphysics



Atmospheric composition

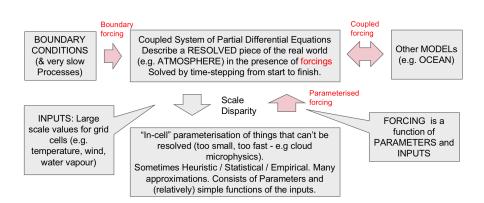


Many sub-grid scale processes which have to be parameterised (that is, approximated, and their "grid-scale" affect is represented by functions of the grid-scale variables and some knowledge of the sub-grid, e.g. orography).



One slide introduction to numerical modelling

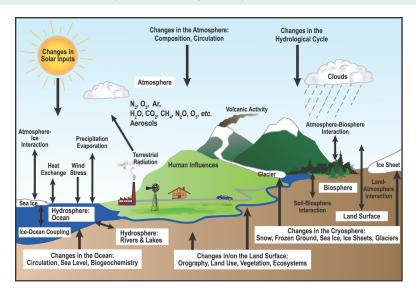
Climate Modelling 00000000000000







beyond the fluid atmosphere - Adding more processes







Everything is solved on a grid



Horizontal Grid (Latitude-Longitude)

Vertical Grid (Height or Pressure)



Given knowledge of state at every grid point at time t, calculate at every grid point state at $t + \Delta t$.

Many points, integrated for years with timestep of o(minutes)!





The Changing World in Climate Models









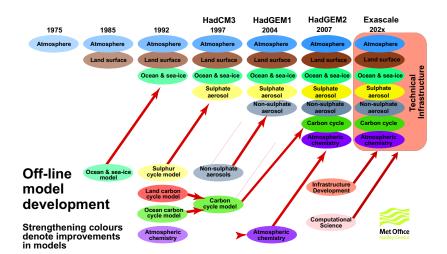






Evolution of Complexity

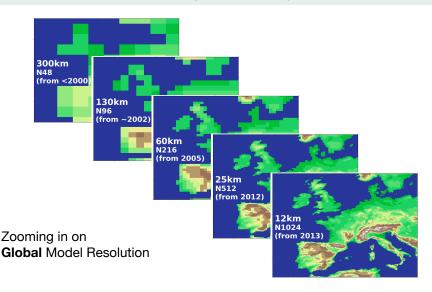
Climate Modelling 00000000000000







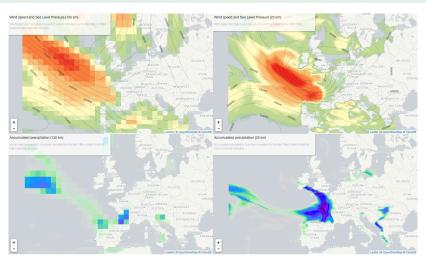
The Evolution of Resolution: A better global microscope!







The influence of resolutioon on simulations of extratropical cyclones



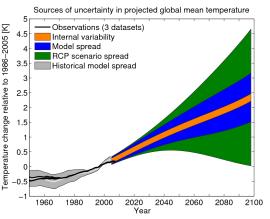
As simulated by the Met Office

https://uip.primavera-h2020.eu/storymaps/extra-tropical-cyclones



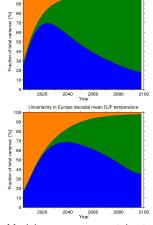






For the global big picture: model uncertainty is not the biggest problem: humanity chooses the pathway! Source: Kirtman et.al., 2013: Near-term Climate Change: Projections and Predictability.

In: Climate Change 2013: The Physical Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Stocker, T.F. et.al. (eds.)]. Cambridge University Press.



Uncertainty in Global decadal mean ANN temperature

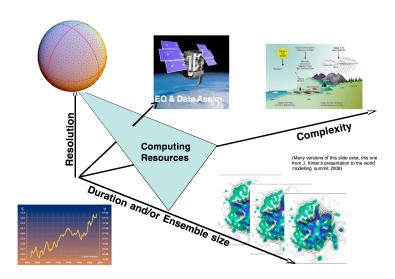
Models are more uncertain at regional scales.





00000000000000 Give me more computing?

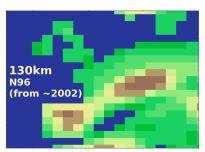
Climate Modelling





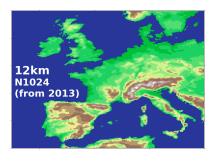


A modest (?) step ...





1 field, 1 year, 6 hourly, 80 levels 1 x 1440 x 80 x 148 x 192



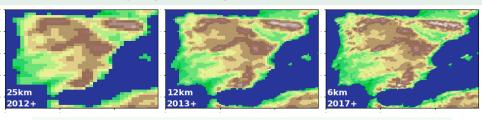
One "field-year" ->6 TB

1 field, 1 year, 6 hourly, 180 levels 1 x 1440 x 180 x 1536 x 2048





Volume — the reality of global 1km grids



What about 1km? That's the current European Network for Earth System Modelling (ENES) goal!

Consider N13256 (1.01km, 26512x19884)):

- ► 1 field, 1 year, 6 hourly, 180 levels
- ► 1 x 1440 x 180 x 26512 x 19884 = 1.09 PB

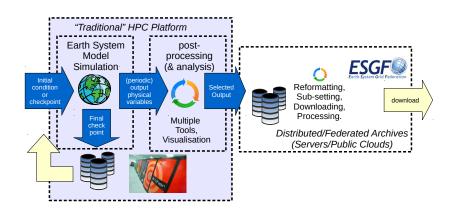
- 760 seconds to read one 760 GB (xy) grid at 1 GB/s
- but it's worse that that: 10 variables hourly, > 220 TB/day!

Can no longer consider serial diagnostics, and even parallelised is a challenge for the I/O system!





How we used to do it: from supercomputer to download







The consequences of data at scale — download doesn't work!

Earth System Grid Experience



Slide content courtesy of Stephan Kindermann, DKRZ and IS-ENES2





Started with Individual End Users

Limited resources (bandwidth, storage)

Moved to **Organised User Groups**

- Organize a local cache of files
- Most of the group don't access ESGF, but access cache.

Then Data Centre Services

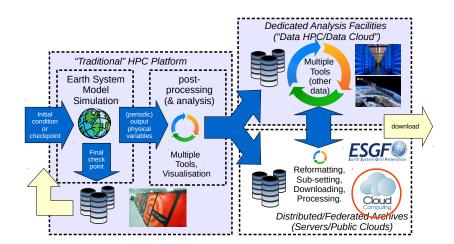
- Provide access to a replica cache
- May also provide compute by data
- CEDA, DKRZ, etc

Trend from download at home, to exploit a cache, to exploit a managed cache with compute!





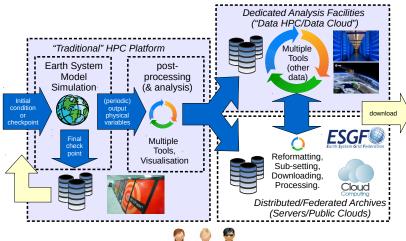
Many different supercomputing environments







Many different supercomputing environments



Multiple Roles, at least:

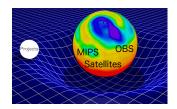


Model Developer, Model Tinkerer, Runner, Expert Data Analyst, Service Provider, Data Manager, Data User





JASMIN — 4 steps in exploiting data gravity to deliver a data commons

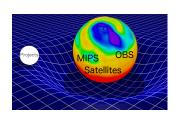


- 1. Provide and populate a managed data environment with key datasets (the "archive").
- Encourage and facilitate the bringing of data and/or computation alongside/to the archive!



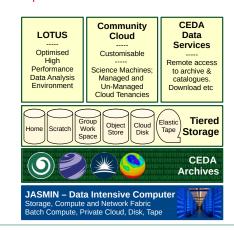


JASMIN — 4 steps in exploiting data gravity to deliver a data commons



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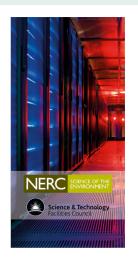
- 3. Provide a state-of-the art storage and computational environment
- Provide FLEXIBLE methods of exploiting the computational environment.







JASMIN: A Data Intensive Computing System





- Customised Fast Network.
- 44 PB Disk Storage.
- Tape Robot and "Elastic Tape Service".
- 12000 compute cores:
 The "Lotus" batch cluster; hosted compute; cloud.
- Some high memory nodes. Some GPU systems from Q2 2019.



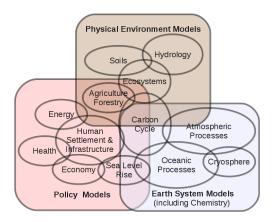








Communities



Many interacting communities, each with their own software, compute environments, observations etc.

Figure adapted from Moss et al, 2010





Virtual Compute and Virtual Organisations







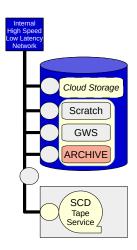
Platform as a Service → Infrastructure as a Service

Example: NCAS as as a big organisation can run a semi-managed virtual organisation (with multiple group work spaces), but large groups within NCAS can themselves setup a virtual organisation to run their own clusters in the un-managed environment.





JASMIN Tiered Storage Requirements

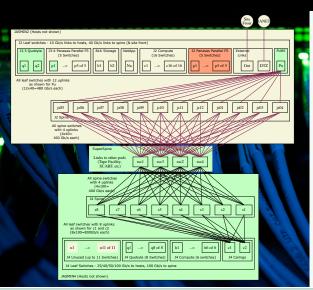


There is not one storage system to rule them all

- Tape is (relatively) cheap. Tape is faster than you think. But tape latency is bad.
- ► Filesystems come with constraints: bandwidth, reliability, scalability, consistency, access control issues. You can't have it all!
 - Cloud Storage:
 - Block storage: build their own file systems.
 - Object Storage: Scalable, simple, flexible access control.
 - Shared file system requirements:
 - Scratch: fast, but trade-off between fast for large volume, and fast for small files.
 - Group Work Spaces: Community shared storage; not necessarily high performance.
 - Archive: long-term persistent, shared access, reliable.







- Pod design with five layer CLOS network connecting pods via a superspine.
- Some blocking into the superspine.
- Evolving:
 - ► JASMIN 2 injection bandwidth into superspine ≈ 2 Tbit/s;
 - >6 Tbit/s.
- More pods possible.
- Designed by Jonathan Churchill, STFC, Inspired by Facebook.





Uncommon (and inappropriate?) software solutions

Multiple tools

Contrast between two very types of workflow:

- Build Once: Many analysis tasks are build once, use once, throwaway. No room for optimisation (or MPI). Need efficient libraries.
- Repeatable: "build", "run", "move", "reduce/reformat", "analyse". Much room for automation...

What to use? Plethora of architectures and tools out there













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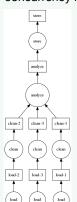






Exploiting Concurrency

Whatever tools, need to get used to generating, understanding, and exploiting concurrency in more complicated ways:





Much to do to harness tools to accelerate workflows!

(These two examples: dask, and cylc, representing bespoke analysis and scheduling, reduction and proliferation.)





Virtual Research Environments on JASMIN hosted cloud



Thematic Exploitation
Platforms for ESA



EOS Cloud —
Desktop-as-a-Service
for Environmental



CCI Open Data Portal for ESA



Hosted Ipython Notebooks



MAJIC interface to JULES model

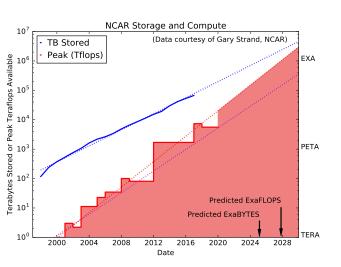


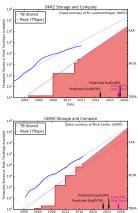
NERC Environmental





History has given us exponential compute linked to exponential data ...









Faster Compute

1981: ICL Dist.Array.Proc. (20 MFlops)



2014: Archer (then 1.4 PFlops)





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EPCC Advanced Computing Facility, 2014





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Slide content courtesy of Arthur Trew:

EPCC Advanced Computing Facility, 2014



From 1981, without Moore's Law











Moore's Law

More often misquoted and misunderstood:

- Original, Moore, 1965: The complexity for minimum component costs has increased at a rate of roughly a factor of two per year.
- House (Intel) modified it to note that: The changes would cause computer performance to double every 18 months
- Moore (Modified 1975): The number of transistors in a dense integrated circuit doubles about every two years

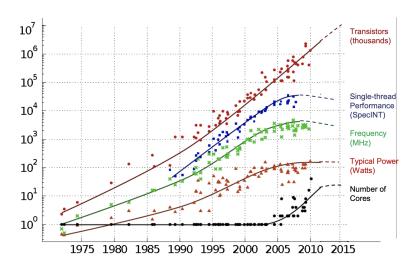
Dennard Scaling

- ▶ The performance per watt of computing is growing exponentially at roughly the same rate (doubling every two years).
- (Increasing clock frequency as circuits get smaller, but this stopped working around 2006, too much power too small, means meltdown!)





The end of Dennard Scaling

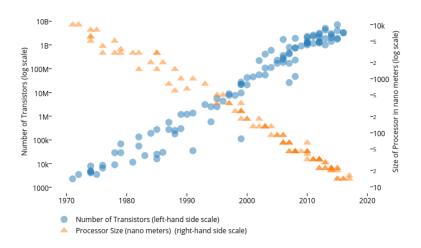


Original data collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond and C. Batten





Moores's Law



https://www.yaabot.com/31345/quantum-computing-neural-chips-moores-law-future-computing/





Moore's 2nd Law aka Rock's Law

- The cost of a semiconductor chip fabrication plant doubles every four years.
- Noyce, 1977: "...further miniaturization is less likely to be limited by the laws of physics than by the laws of economics."



By Shaun Nichols in San Francisco 27 Aug 2018 at 23:55

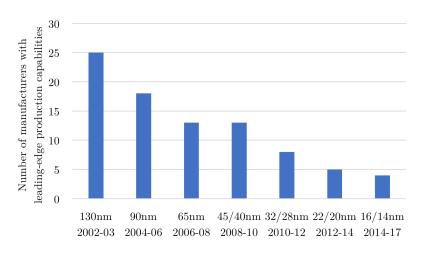
...to shift resources (including R&D). to the 14 and 12nm efforts where ...most of their chip customers ...are planning to stay with the current-gen architectures and squeeze performance out by other means.

7nm is expensive, it's cheaper and easier to improve the performance and density of 12nm, and hardware accelerators and custom chips ...





18 ☐ SHARE ▲



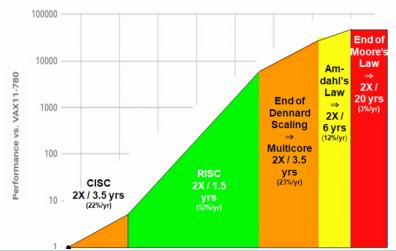
https://www.nextplatform.com/2019/02/05/the-era-of-general-purpose-computers-is-ending/





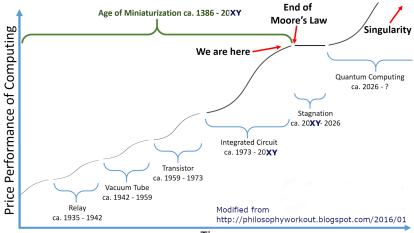
The Evolving Moore's Law

40 years of Processor Performance















What now then?

No more advances for free on the back of computer hardware improvements and relatively little pain! Need to "resort" to

Maths

Algorithms

Customised Hardware

Software Solutions for performance, portability, and productivity.

(Avoidance and Sharing)

No more free lunch, a very different modelling world!





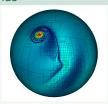
No time to talk about maths and algorithms ...these, and more!

Parallel Time-Stepping

$$\mathbf{X}_{t+1}(x, y, z, t) = f(\mathbf{X}_{t-1}, \mathbf{X}_t)$$

The function *f* could involve several steps (iterates) *carried out in parallel*.

Adaptive Grids

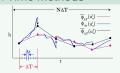


Machine Learning: Optimisation

Using AI/ML to

- Pre-condition solvers.
- Optimise/tune parameters

Parallel in Time Methods



Variable Precision

Not all variables need the same precision (number of bits) in calculation, or in output.

Machine Learning: Emulators

Replace slow "exact" (stationary?) parameterisations with fast "learnt" emulators.





From decades of the same to a Cambrian Explosion



Vector Processors on Intel Zeon



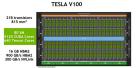
Vector Processing Units from NEC



Google's Tensor Programming Unit



Server chips based on ARM designs



GPUs from NVIDIA and AMD



FPGA from many sources

The end of Moore's Law means more specialisation: all with very different programming models!





Too many levels of parallelism

Vector Units (on chip)

Parallelism Across Cores

Shared Memory Concurrency

Distributed Memory

Numerical Method Concurrency

Internal Component Concurrency

Coupled Component Concurrency

I/O and Diagnostic Parallelism

(Storage System Parallelism)





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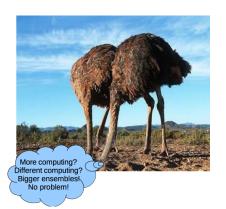
Entirely new programming models are likely to be necessary, with entirely new* constructs such as thread pools and task-based parallelism possible. Memory handling will be crucial!

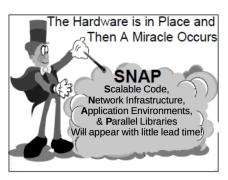
^{*}New in this context!





What about software?





Some people have a very naive idea about the relationship between the hardware and the software!





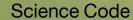


Hardware changing rapidly & accelerating!

How far is it between our scientific aspiration and our ability to develop and/or rapidly adapt our codes to the available hardware?







How do we bridge the gap?

Compilers, OpenMP, MPI etc

Hardware & Operating System





Crossing the Chasm: How to develop weather and climate models for next generation computers?

Lawrence, Rezny, Budich, Bauer, Behrens, Carter, Deconinck, Ford, Maynard, Mullerworth, Osuna, Porter, Serradell, Valcke, Wedi, and Wilson

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

IS-ENES2 Deliverable 3.2

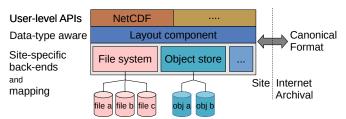






Earth System Data Middleware





Key Concepts



- Applications work through existing application interfaces (currently: NetCDF library)
- Middleware utilizes layout component to make placement decisions
- Data is then written/read efficiently avoiding file system limitations (e.g. consistency constraints)
- Potential for deploying with an active storage management system.





Work in Progress: In-Flight Parallel Data Analysis

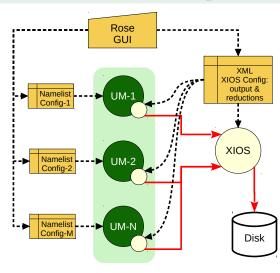


An ensemble is a set of simulations running different instances of the same numerical experiment. We do this to get information about uncertainty.

Dealing with too much ensemble data

Instead of writing out all ensemble members and doing all the analysis later:

- Calculate ensemble statistics on the fly.
- Only write out some ensemble members.
- (Which ones? A tale for another day, see Daniel Galea's Ph.D work.)



Cole, Lawrence, Lister, Meursdesoif, Nash, Weiland





► Climate modelling is one of the grand computational challenges





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- Data handling is challenging, and getting more so.



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- ► The traditional route to more computing, via Moore's Law is ending.
- ▶ New ways forward need to be found: from new maths, new techniques such as ML and AI, to new ways of programming, and new methods of data handling.





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There is a lot for Computer Scientists to do! aces.cs.reading.ac.uk





