

Data Production at DKRZ

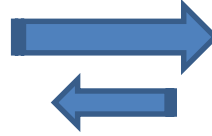
Deutsches Klimarechenzentrum (DKRZ) = German National HPC Facility for Earth System Research



Supercomputer &
Disk Storage

3.6 PFlops
266 TByte Mem.
Top500: #55 (06/2018)

54 PByte
Lustre File System



HPSS Archive System

8 + 1 Silos
70 Drives
up to 500 PByte

Visualization Server

- Integral part of HLRE-3 supercomputer
- FDR Infiniband Interconnect
- 12 Nodes Atos/Bull
 - 2 CPUs, 12 Core Haswell
 - 256 GB Main Memory
 - NVidia Kepler GPUs: Tesla K80
- 9 Nodes Atos/Bull
 - 2 CPUs, 18 Core Broadwell
 - 512 / 1024 GB Main Memory
 - NVidia Maxwell GPUs: M40 / M6000
- Remote 3 D-Rendering: TurboVNC/VirtualGL



Visualization Tools employed in Climate Sciences

Type	Name	URL	Properties	
Domain-specific	NCL	http://www.ncl.ucar.edu/	2D script-based	free
	IDV	http://www.unidata.ucar.edu/software/idv/	2D/3D interactive GUI	free
	Vapor	https://www.vapor.ucar.edu/	3D interactive GUI	free
	UV-CDAT	http://uvcdat.llnl.gov/	Collection: 2D /3D tools	free
	GrADS	http://cola.gmu.edu/grads/	2D script-based	free
	Ferret	http://www.ferret.noaa.gov/Ferret/	2D script-based	free
	GMT	http://gmt.soest.hawaii.edu/	2D script-based	free
General-purpose	ParaView	http://www.paraview.org/	3D interactive GUI	free
	Avizo	https://www.fei.com/software/avizo3d/	3D interactive GUI	\$\$ †
	IDL	http://www.harrisgeospatial.com/	2D script-based	\$\$
	Python / matplotlib	http://matplotlib.org/	2D script-based	free

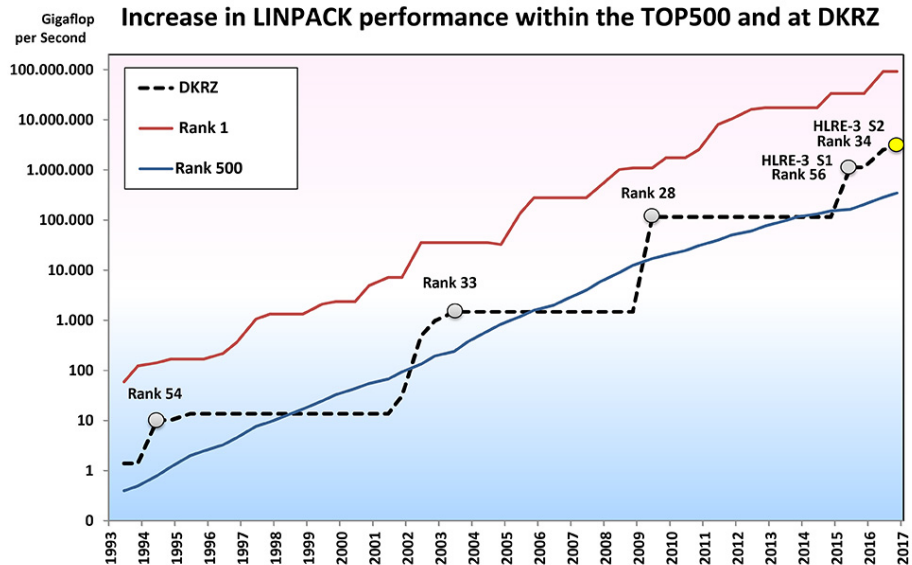
Recommended Literature:

Rautenhaus et al. (2017): Visualization in Meteorology --- A Survey of Techniques and Tools for Data Analysis Tasks, IEEE Transactions on Visualization and Computer Graphics, DOI: 10.1109/TVCG.2017.2779501

Scientific Use of Performance Increases

- 1 ■ Higher spatial / temporal resolution of climate models
 - Direct simulation instead of parameterization
- Increased model complexity (more processes / variables)
- 2 ■ Ensemble simulations -> uncertainty
- Very long simulations
 - Ice age cycles (110,000 years)

-> Challenges for visualization



Challenge 1: High Resolution



ICON, HD(CP)² and ESiWACE

- **ICON** - **ICO**sahedral **N**on-hydrostatic grid

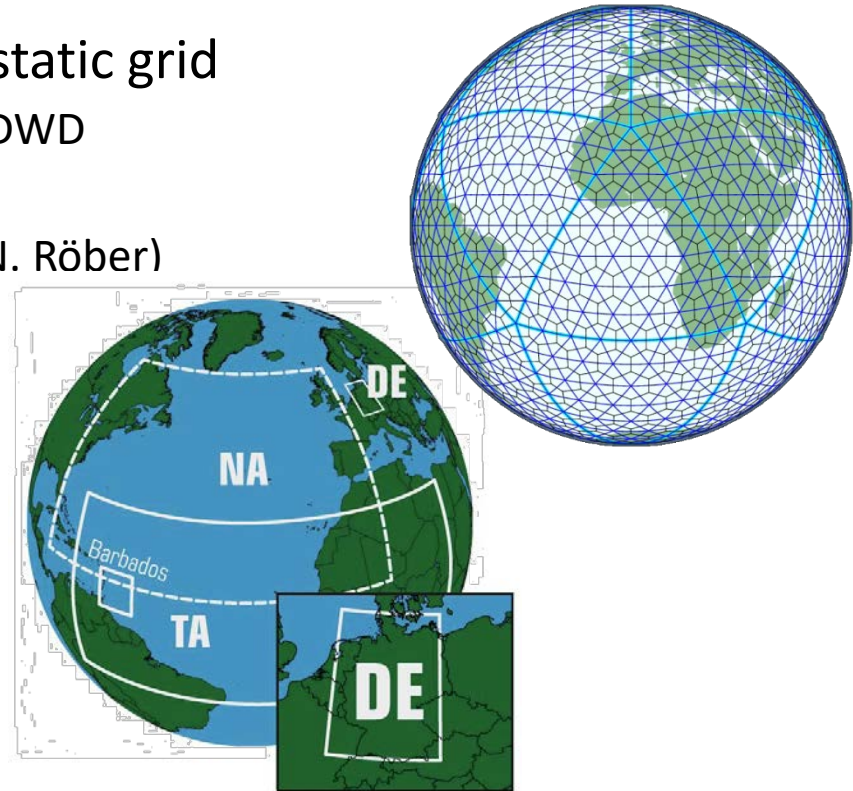
- Joint Development by MPI-M and DWD
- Unstructured Grid
- ParaView U-Grid NetCDF-Reader (N. Röber)

- **Global Setup**

- Ocean – 5 km resolution
(15 million cells * 64 levels)
- Atmosphere – 2.5 km resolution
(84 million cells * 138 levels)

- **Regional Setup (ICON-LES)**

- HD(CP)² – 120 m resolution
(22.5 million cells * 150 levels)



esiwace
CENTRE OF EXCELLENCE IN SIMULATION OF WEATHER
AND CLIMATE IN EUROPE



HD(CP)²
High definition clouds and precipitation
for advancing climate prediction

Grid Resolution – Global Model



ICON R2B9

5 km

21 Mio Cells / Level

Several weeks

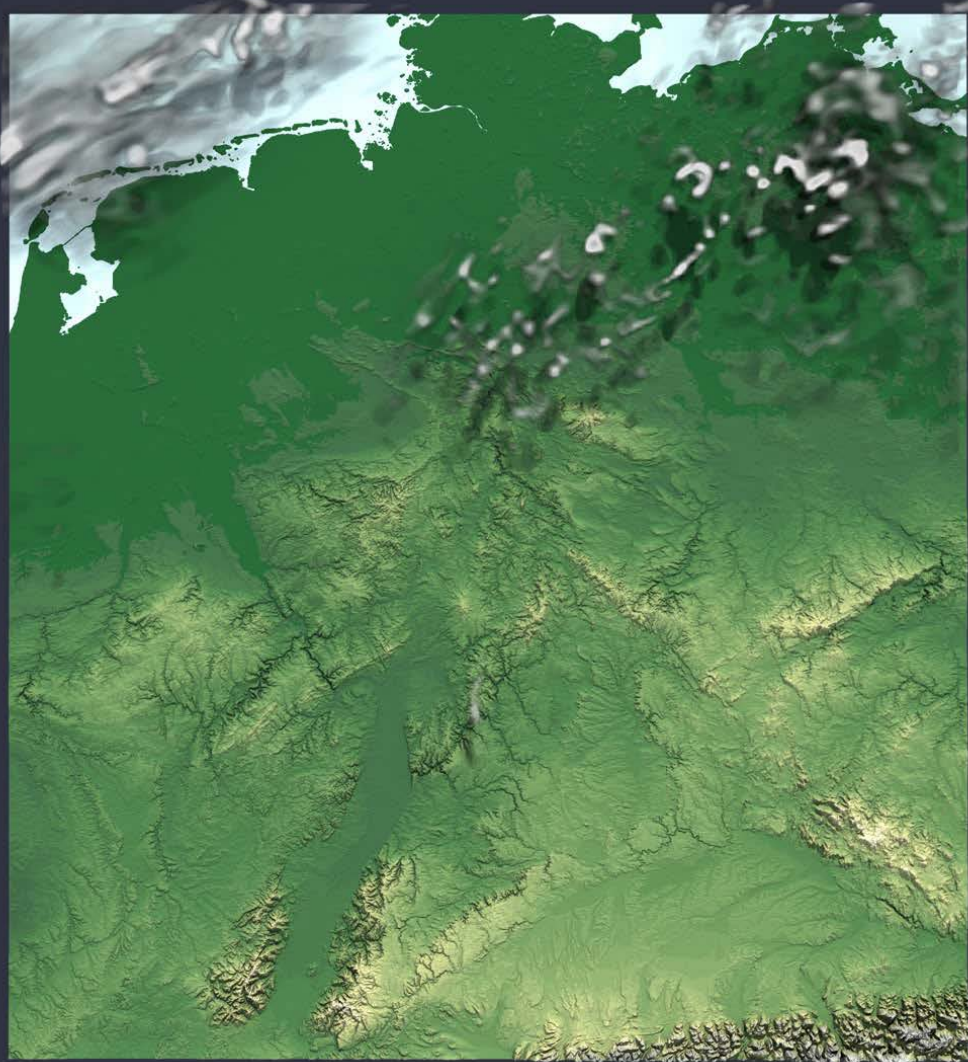
Grid Resolution – Global Model



ICON R2B10
2.5 km
84 Mio Cells / Level
Several Days

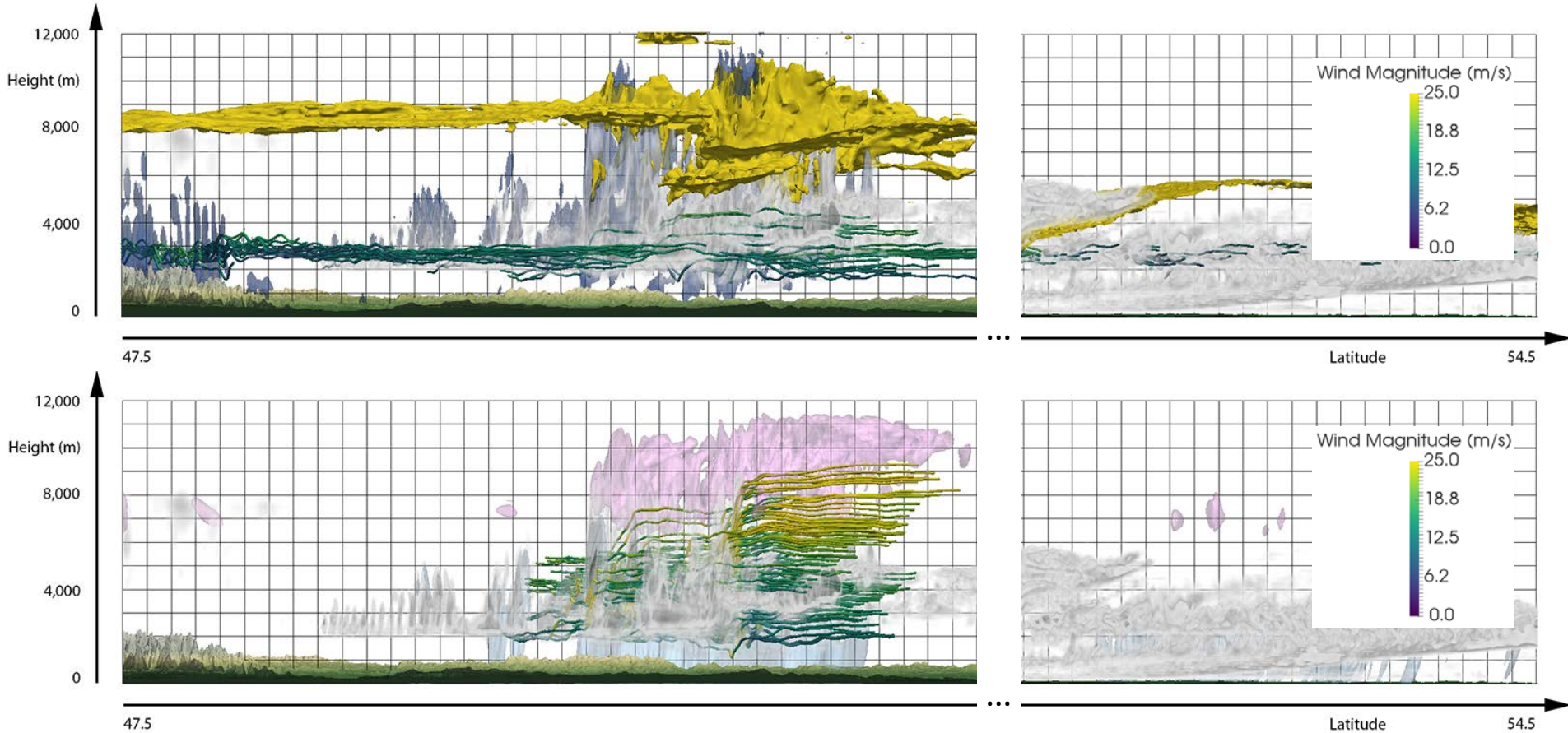


ICON-LES - DE Domain
Cloud Water
 $\Delta t = 1 \text{ min (storage)}$
 $\Delta x = 312 \text{ m}$

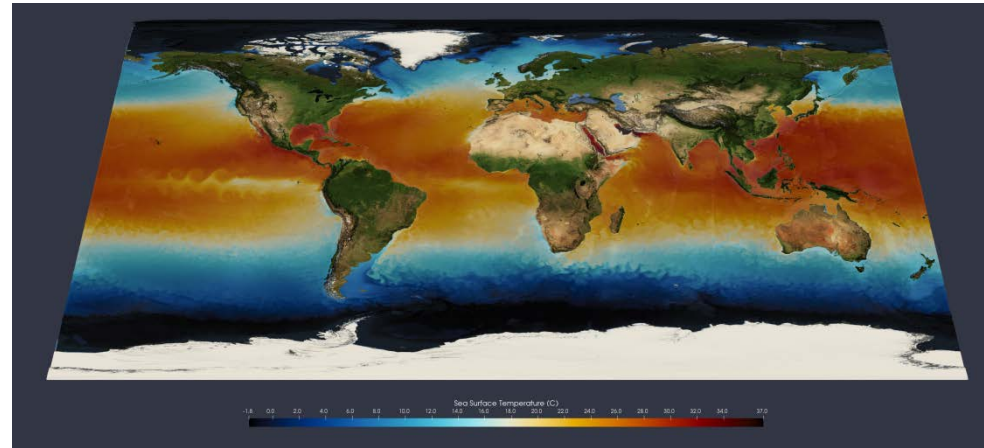
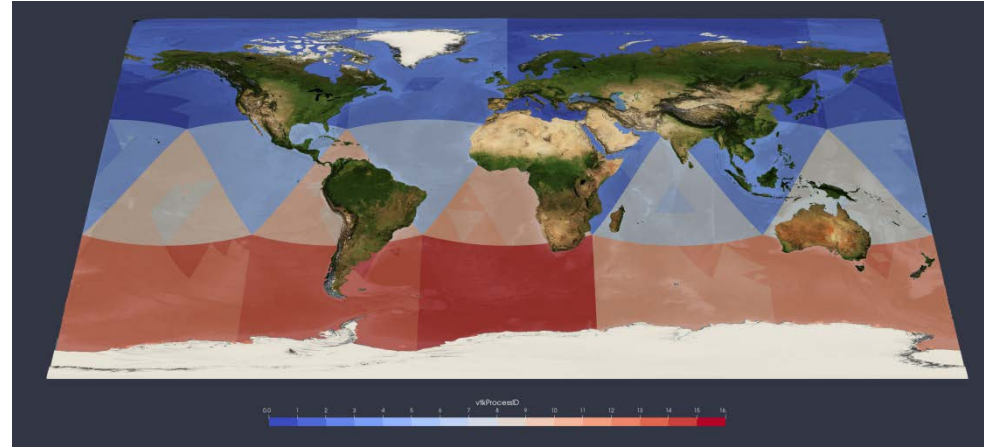
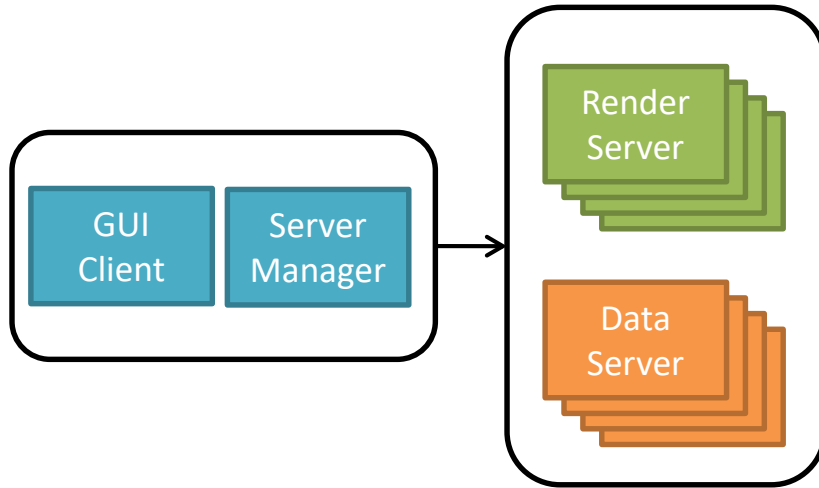


26.April 2013 0 Uhr

HD(CP)² Data

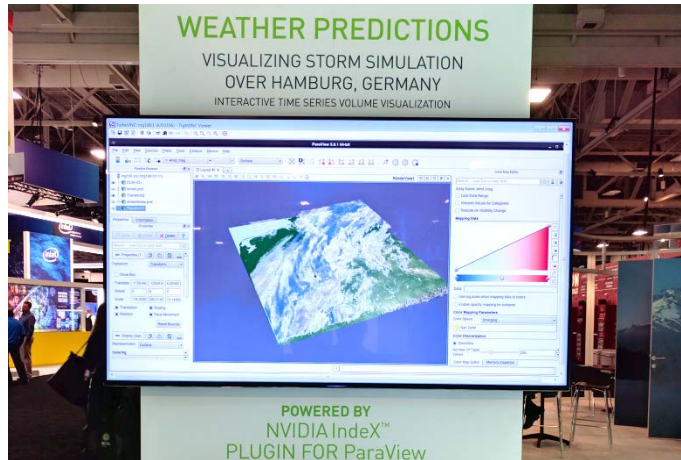


Parallel Processing and Visualization with ParaView

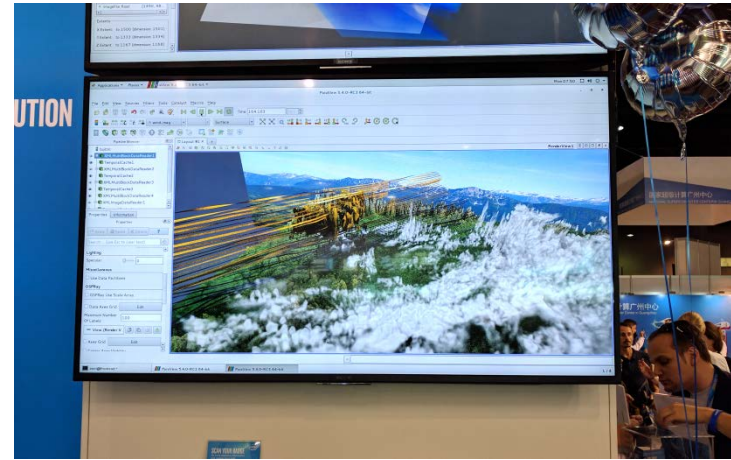


NVIDIA IndeX and INTEL OSPRay

- Two different approaches for (volume) rendering of large data
 - Using GPUs and texture based volume rendering
 - Using CPUs and software based ray-tracing
- We (of course) collaborate with both 😊

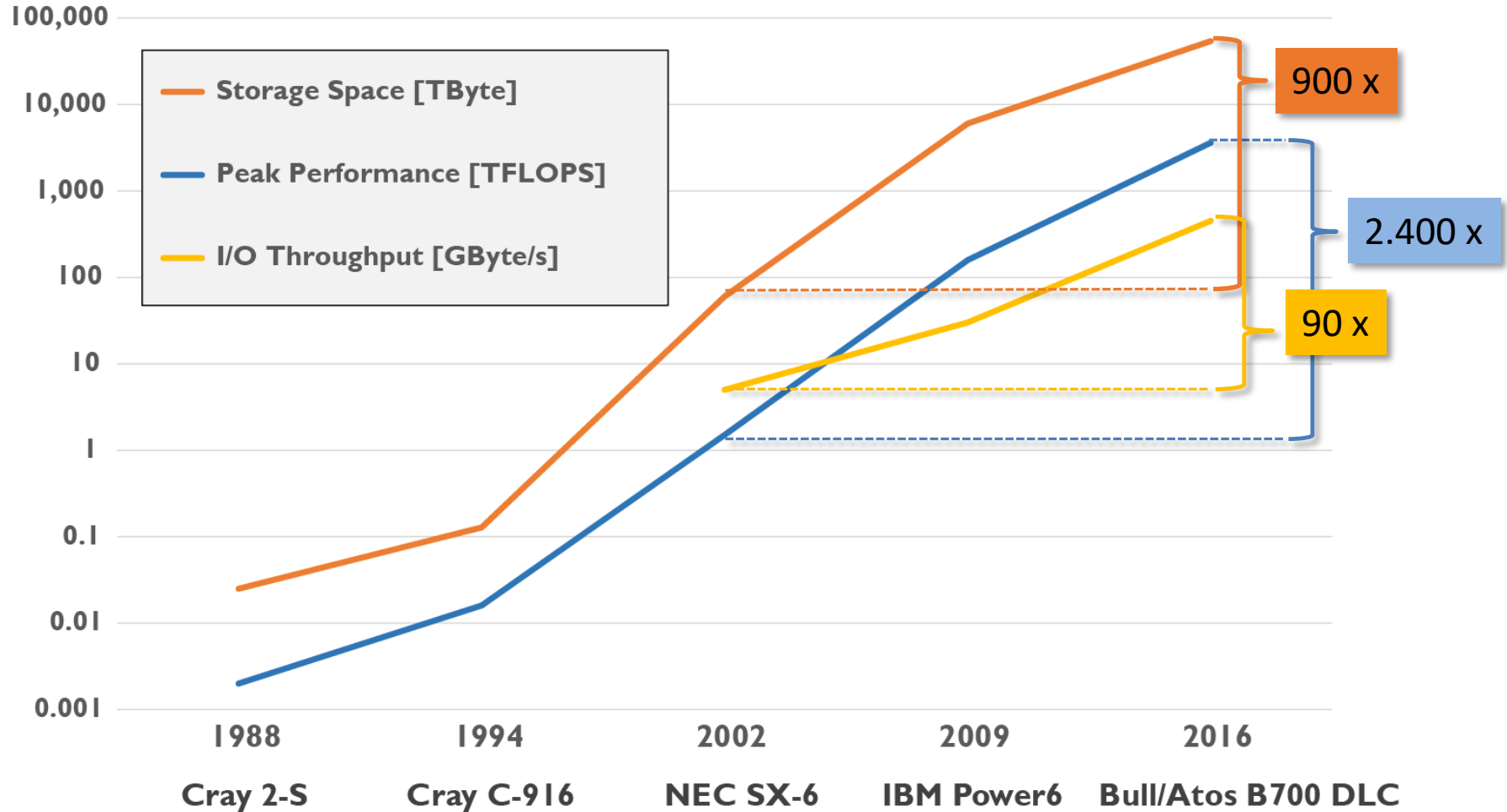


SC 16



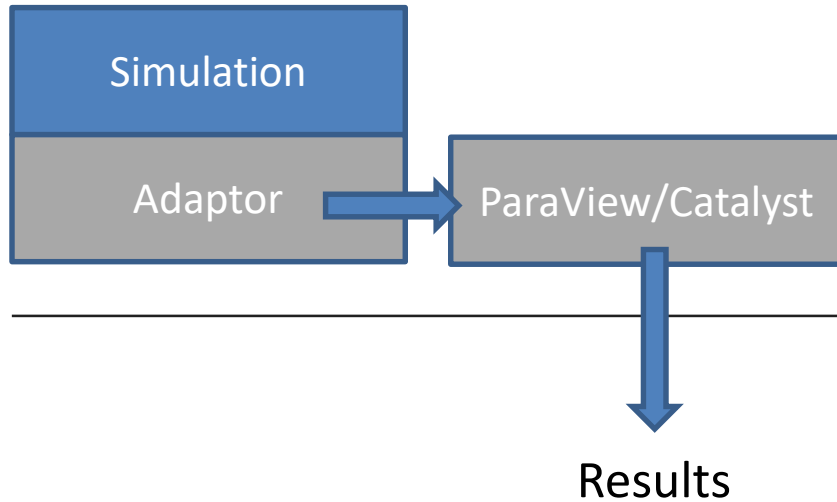
ISC 17

Storage Space, Peak Performance and I/O Throughput at DKRZ

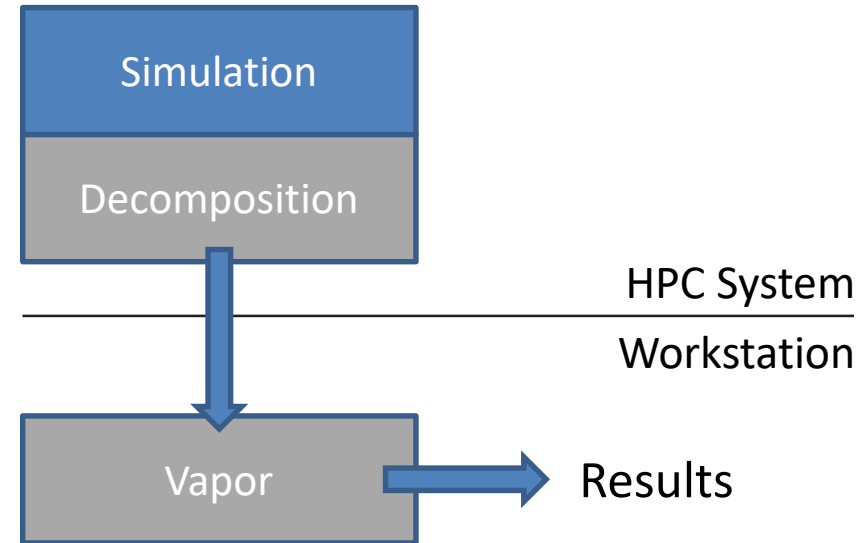


Two Options

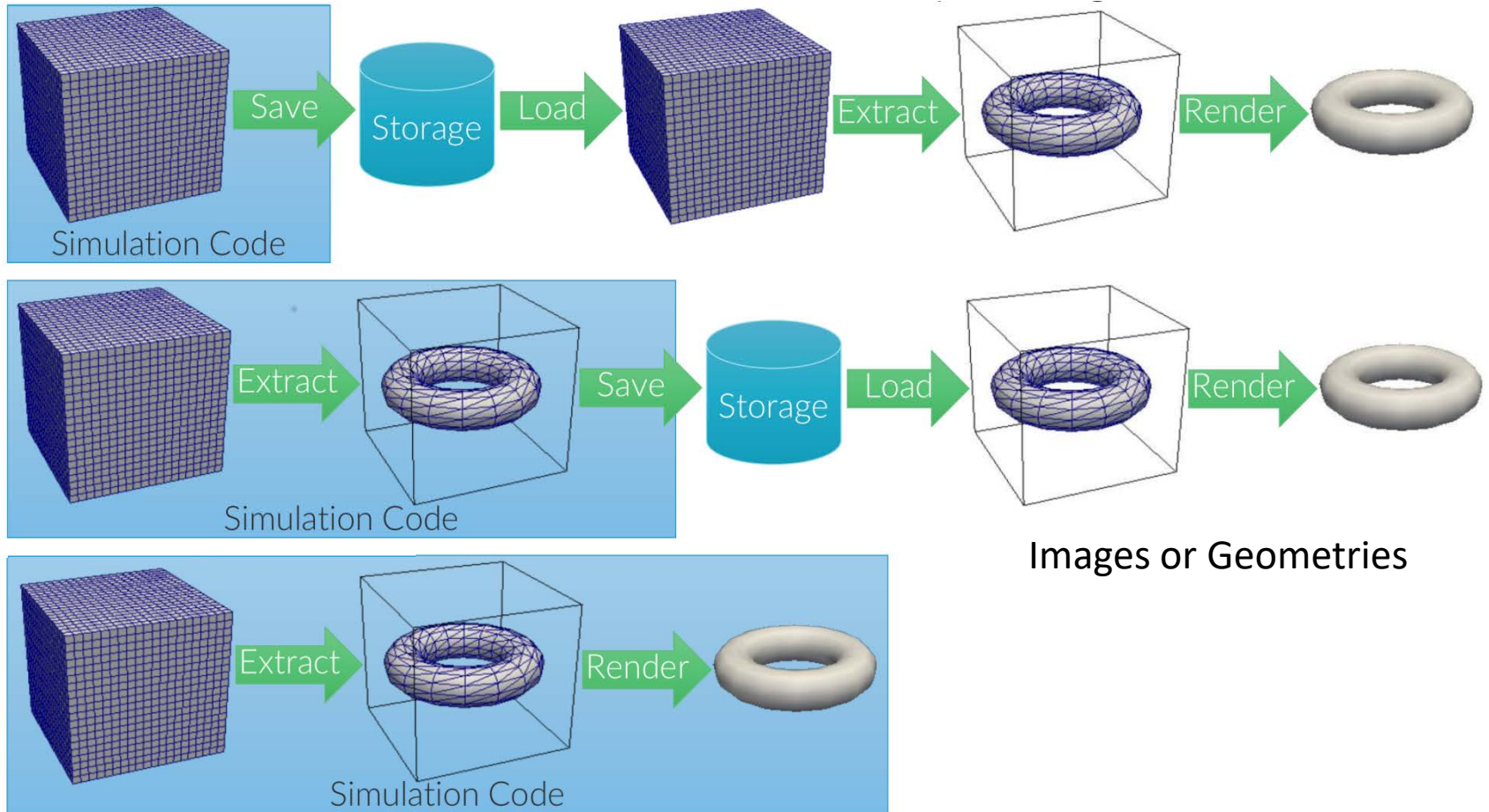
in-situ Visualization (ParaView/Catalyst/Cinema)



in-situ Compression (Vapor)

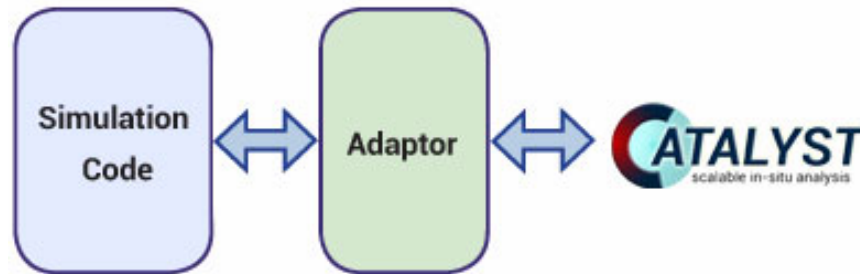


From Post Visualization to In-Situ



In-situ Visualization with ParaView

- In-situ (co)processing/visualization using Paraview/Catalyst
- Adaptor required that connects ICON (model) and Catalyst
- Two possibilities for co-processing:
 - Batch-visualization using pre-defined Python scripts
 - Live visualization within a client/server setting



- First successful tests with prototype for ICON

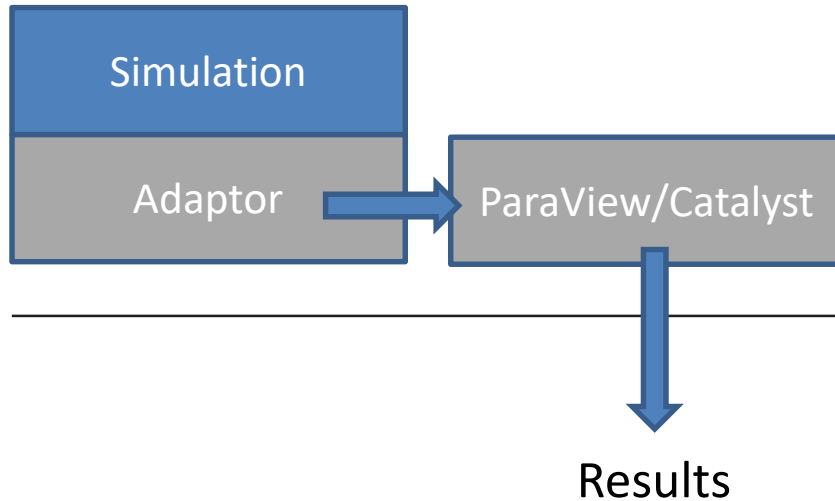
In-Situ Visualization - Challenges

- Paradigm change in scientific and data workflow
- Research/analysis questions need to be better known ahead of the simulation
- Climate = statistics of weather. How to derive and visualize statistical information in-situ?

Large Data Visualization

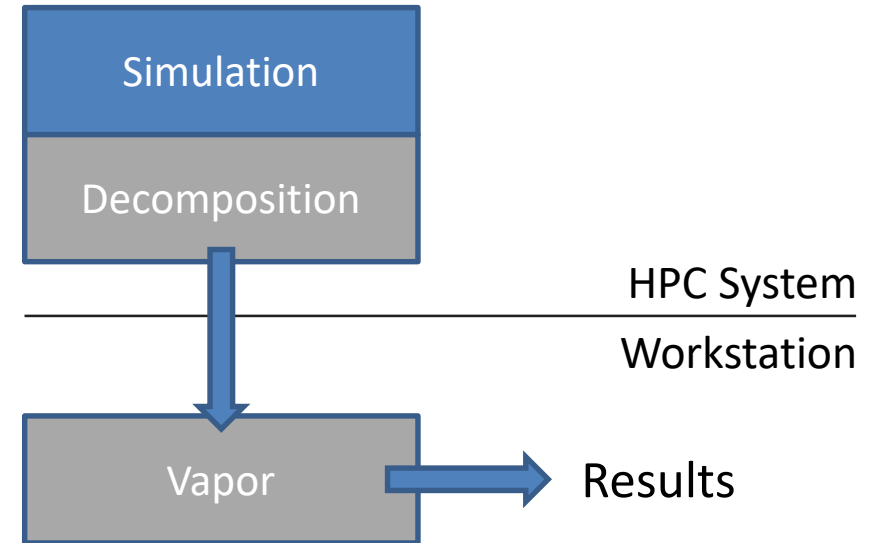
in-situ Visualization

(ParaView/Catalyst/Cinema)



in-situ Compression

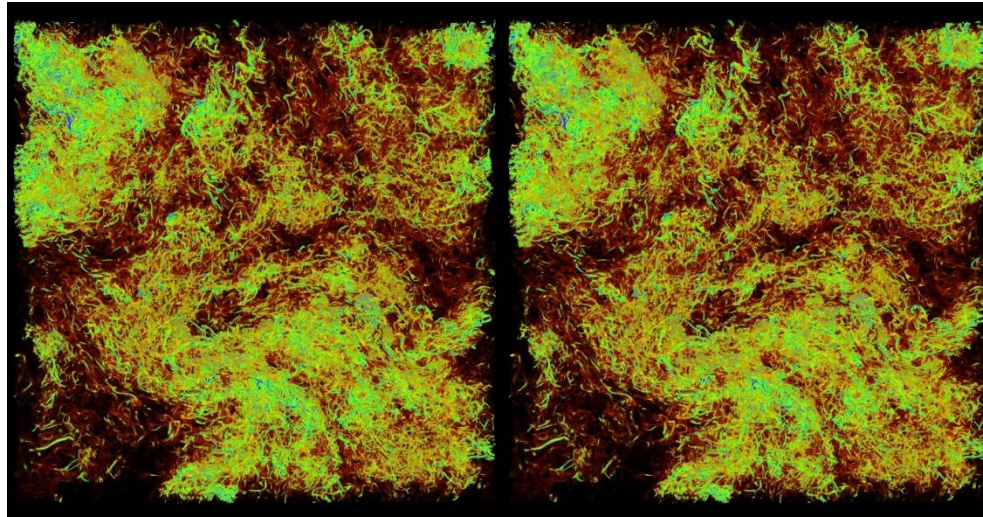
(Vapor)



VAPOR and Compression

- Wavelet-based compression and data storage
- Progressive data access with multiresolution rendering
- Coefficients are sorted and prioritized

Original
275 GBs / 3D field



800:1
0.34 GBs / 3D field

(c) John Clyne, NCAR

File Edit Data Capture Help

Modes: * Navigation 0 < > < > < > Visualizer No. 0 Align View Interactive Refinement: < >

ntours 2D Image Probe Iso Flow DVR < >

Renderer Control

Instance View New Delete

Insta... Duplicate In: < >

Set Default Fidelity Low .. Fidelity .. high

LOD 1:1 Refinement 0

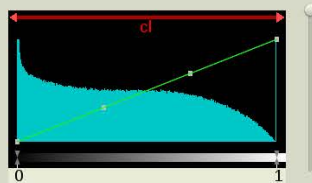
Variable cl < > Under Type 3DTexture < >

Color Selector

Hue: Red: Sat: Green: Val: Blue: < >

Transfer Function Editor

Edit Zoom/Pan Fit to View Histo

☐ Discrete Color Map

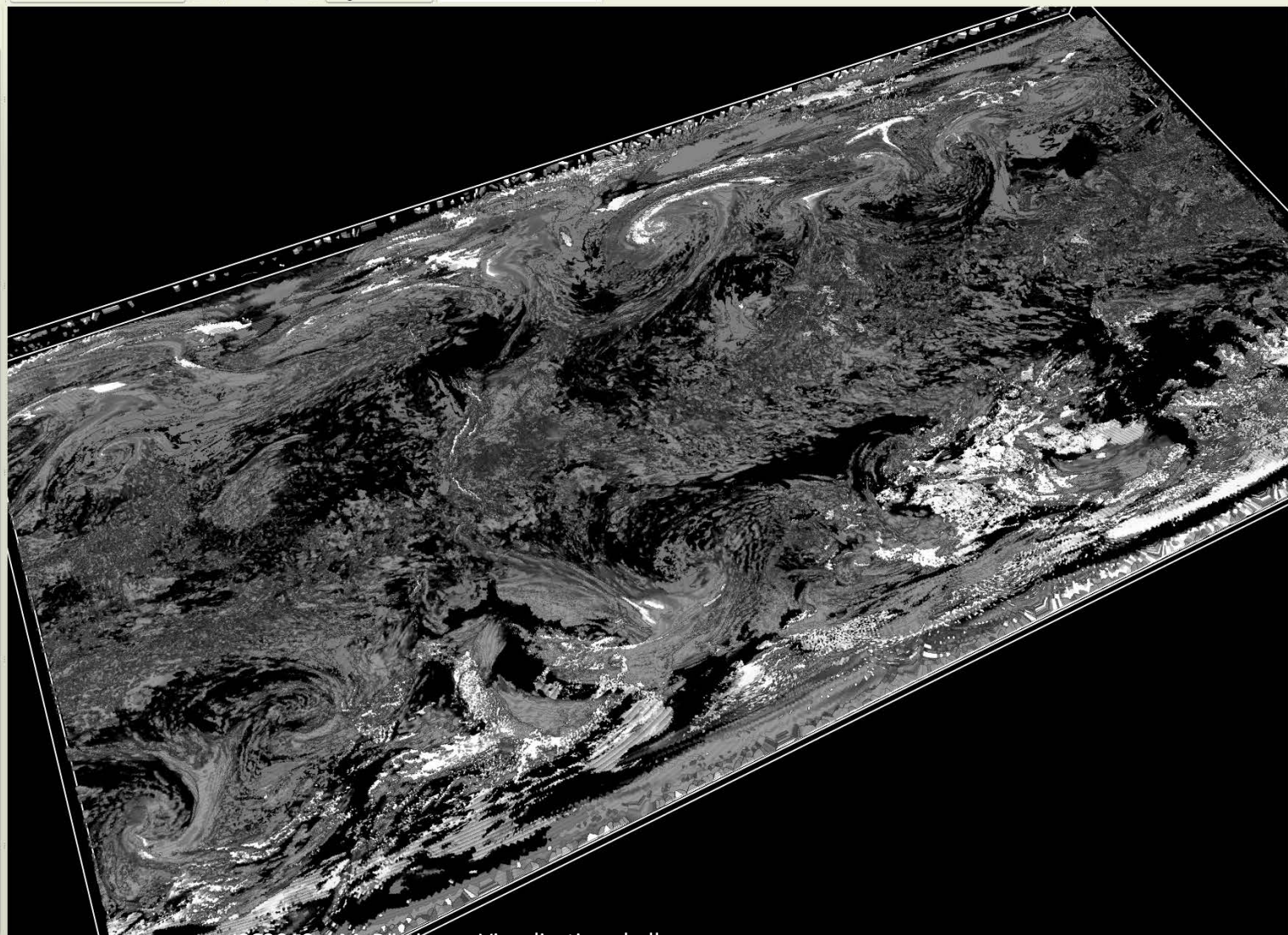
0 TF Domain Bounds 1

0 Data Bounds 1

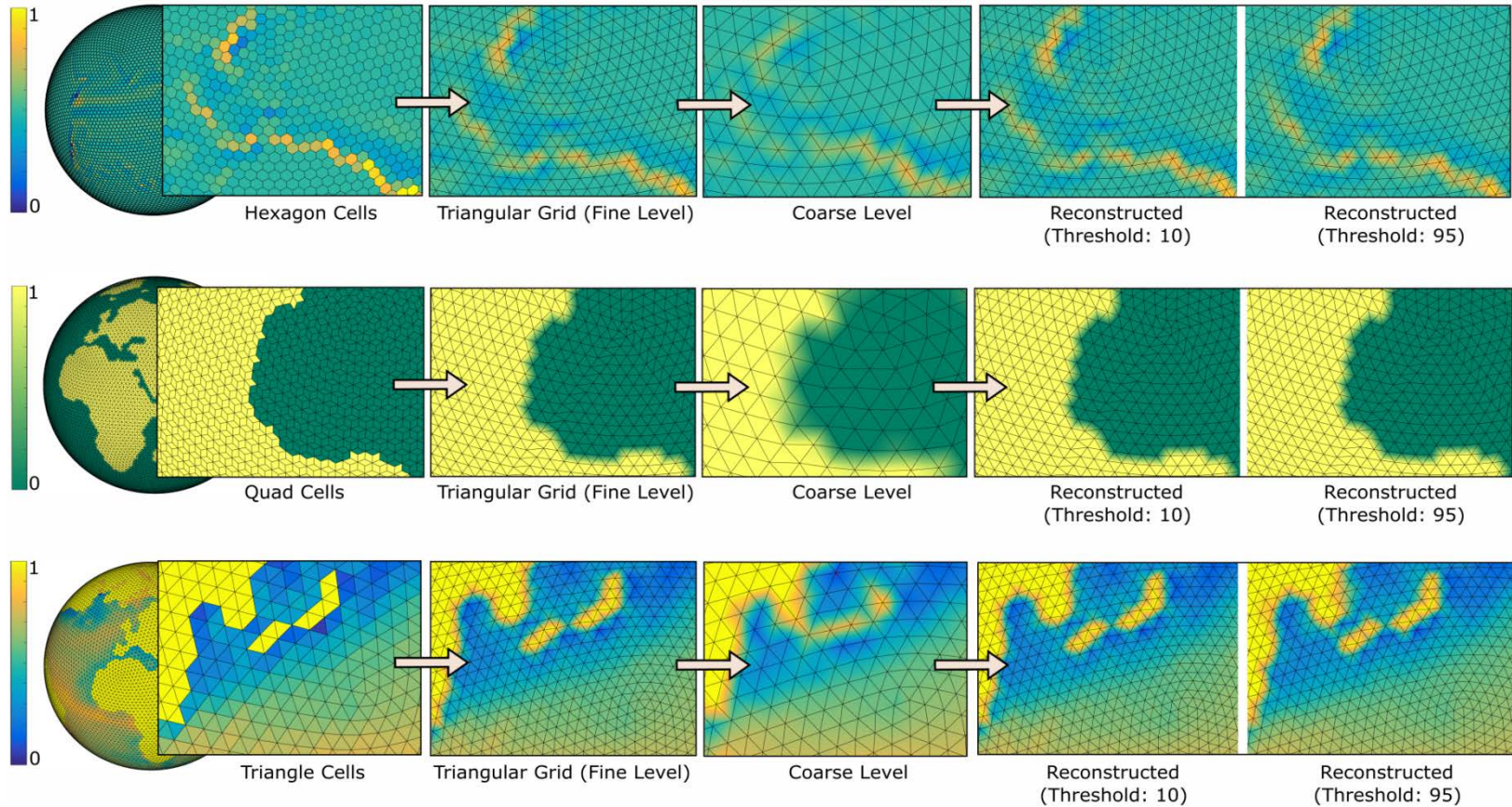
Find Color->Op Find Opac->Col Fit Data

Load TF Load Installed TF Save TF

Bits per voxel 8 Histo scale 1

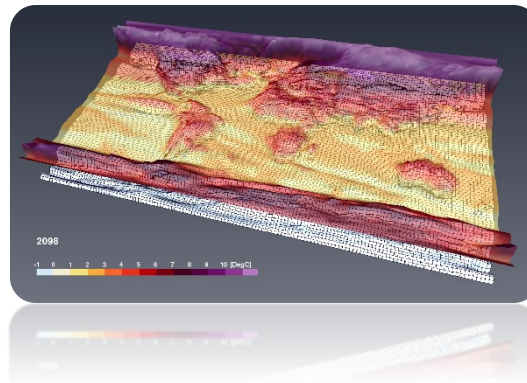
☐ Lighting On ☒ Pre-integration On

Multiresolution with Icosahedral Maps



[1] Jubair et.al. "Icosahedral Maps for a Multiresolution Representation of Earth Data", VMV 2016

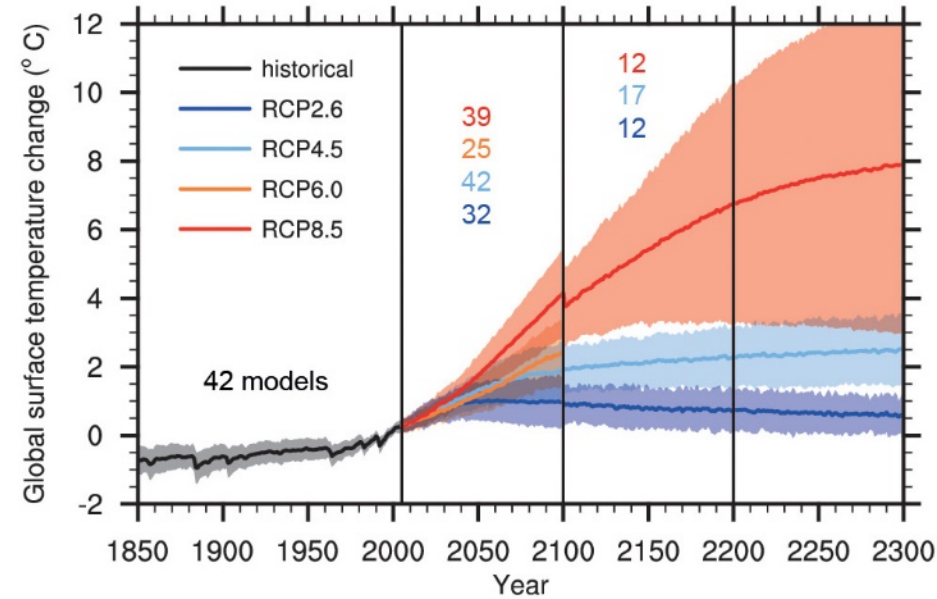
Challenge 2: Ensembles and Uncertainty



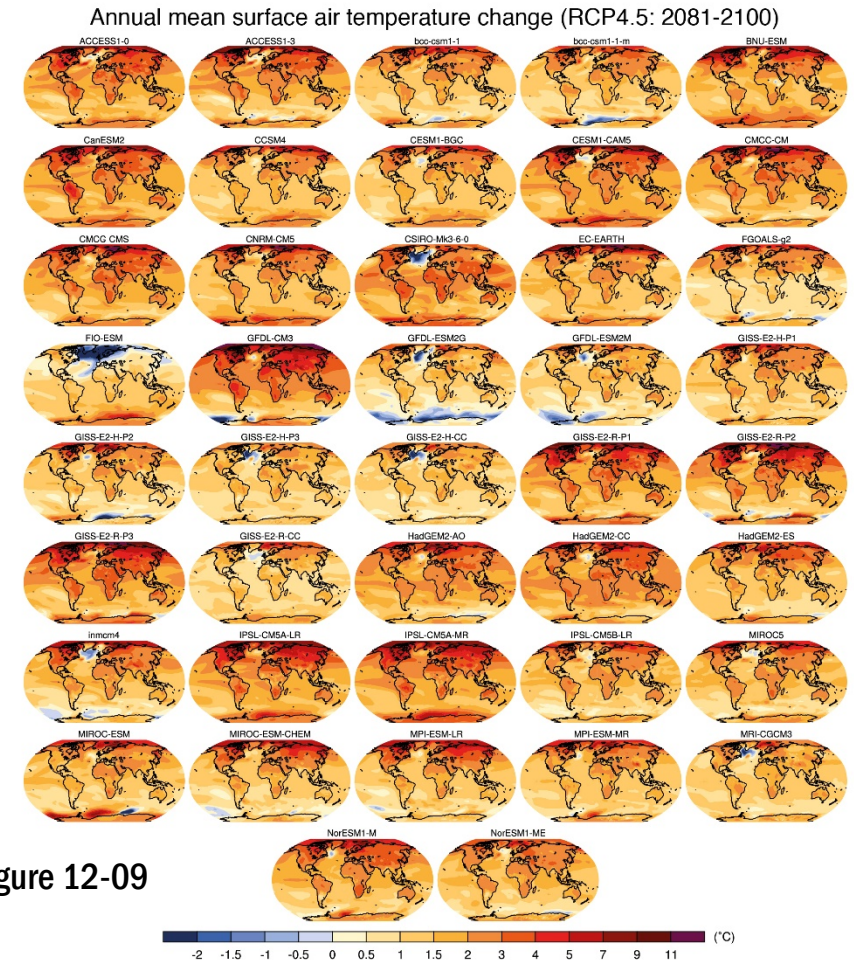
Uncertainty or Robustness?

- Uncertainty
 - What we don't know
 - Communicate some result plus some uncertainty range (model spread)
- Robustness
 - What we know (= positive)
 - Determine which part of the data (i.e. projected signals) is statistically significant

CMIP5 - Projected Changes in the 2m-Temperature

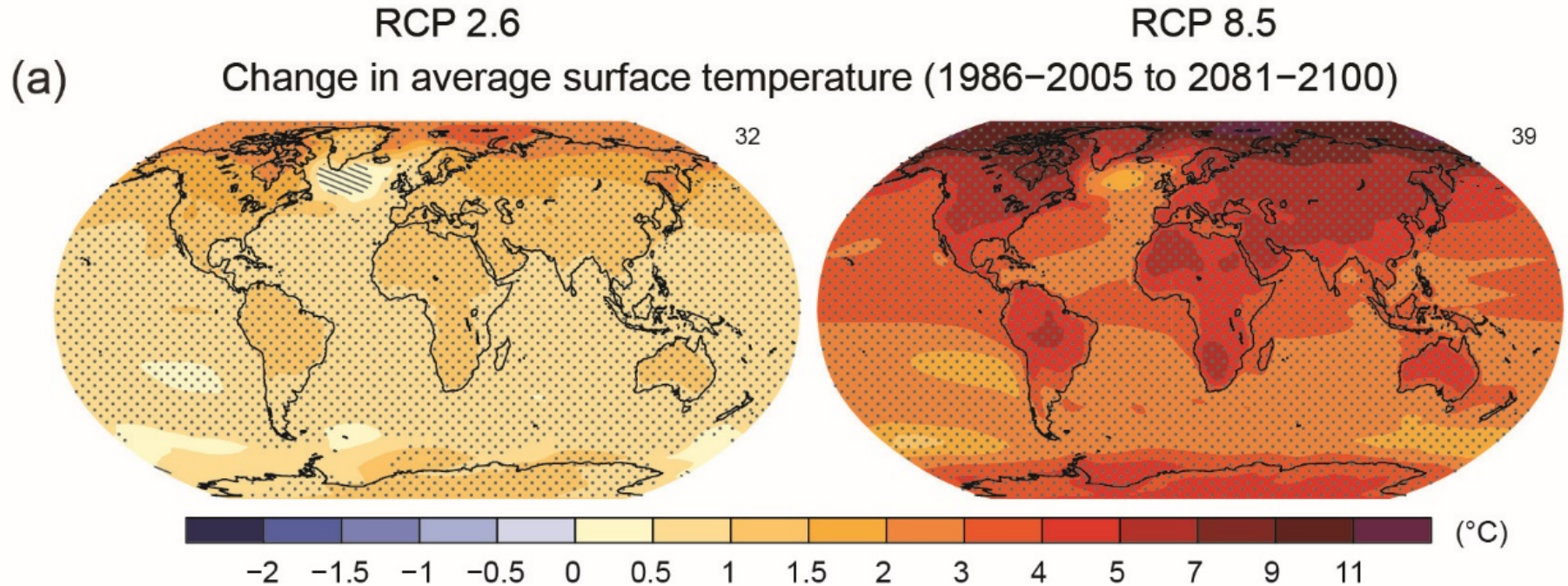


Source: IPCC AR5, TS Figure 15



Source: IPCC AR5, Figure 12-09

Overlaid Stippling and Hatching indicating Robustness



Stippling indicates multi-model mean is more than two standard deviations of natural internal variability in 20-yr means; hatching indicates multi-model mean is less than one standard deviation of natural internal variability in 20-yr means, and where at least 90% of models agree on the sign of change.

Source: IPCC AR5, SPM Figure 8



Sources of uncertainty

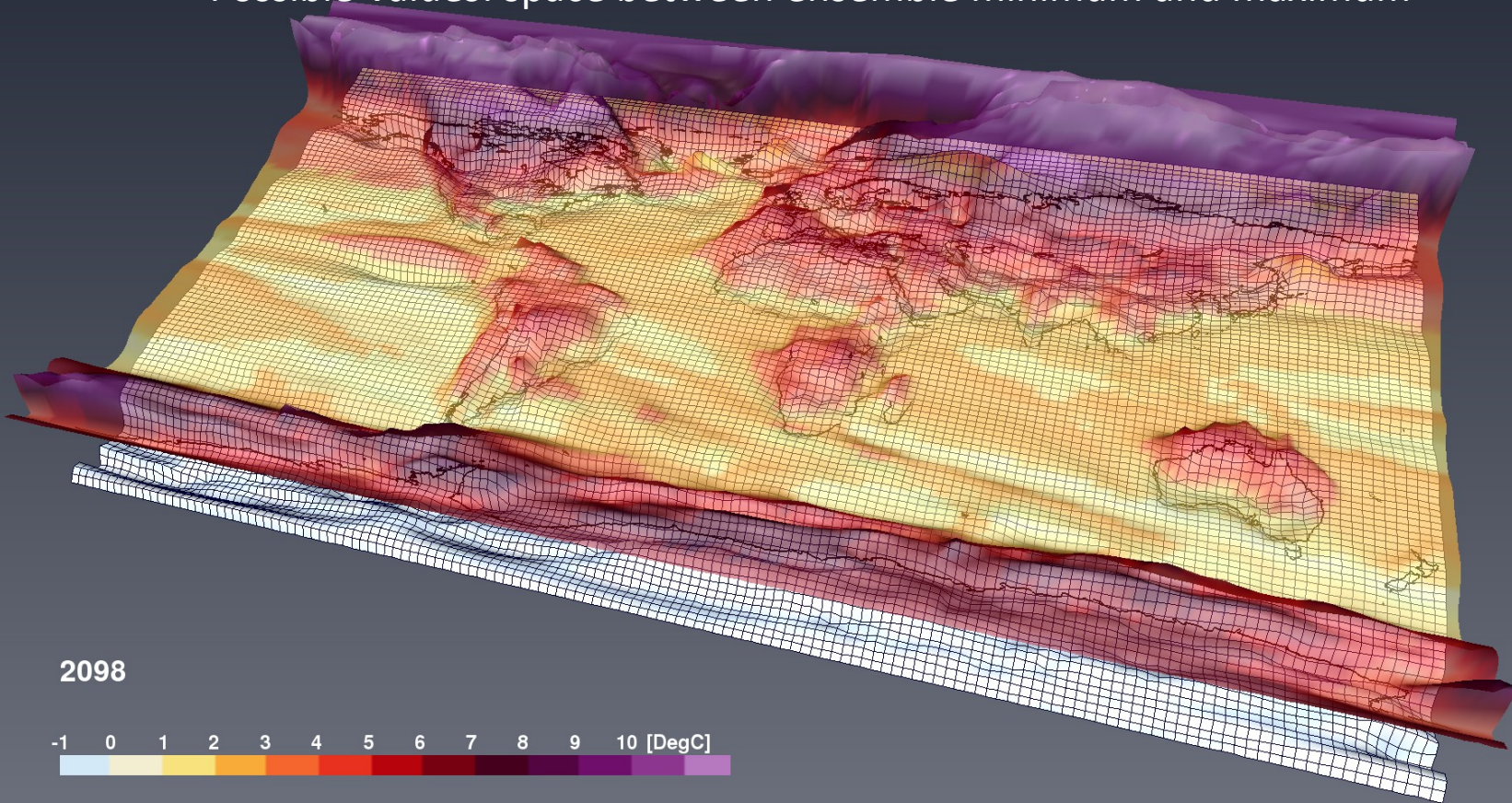
- Internal climate variability
- Model uncertainty:
 - representation of processes
 - climate sensitivity
- Scenario uncertainty (following example: only RCP4.5)
- **Unknown volcanic activity (space, time, magnitude)**

Model and experiments described in

Ingo Bethke, Stephen Outten, Odd Helge Otterå, Ed Hawkins, Sebastian Wagner, Michael Sigl & Peter Thorne: Potential volcanic impacts on future climate variability, Nature Climate Change volume 7, pages 799–805 (2017), doi:10.1038/nclimate3394

Uncertainty range of temperature anomaly

Possible Values: space between ensemble minimum and maximum



Development of ensemble mean, minimum and maximum



Conclusions

- Even with moderately parallel vis resources, visualization of current ESM simulation data sizes is scratching the technical limits
- In-situ visualization, compression and analysis are promising technical solutions
 - Require paradigm and workflow changes in scientific practice
- Ensemble Simulations add further dimension
 - Robustness versus uncertainty
 - Stippling/hatching versus semi-transparent shading
 - Combined heightfield displays of ensemble minimum and maximum visualize the uncertainty space

The CMIP5 visualization shown in Challenge 2 was used in a recent video produced by the WCRP (World Climate Research Programme) to promote the CMIP project.

<https://www.youtube.com/watch?v=WdRiYPJLt4o>

Questions?

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