

Refactoring CESM for Exascale

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Outline

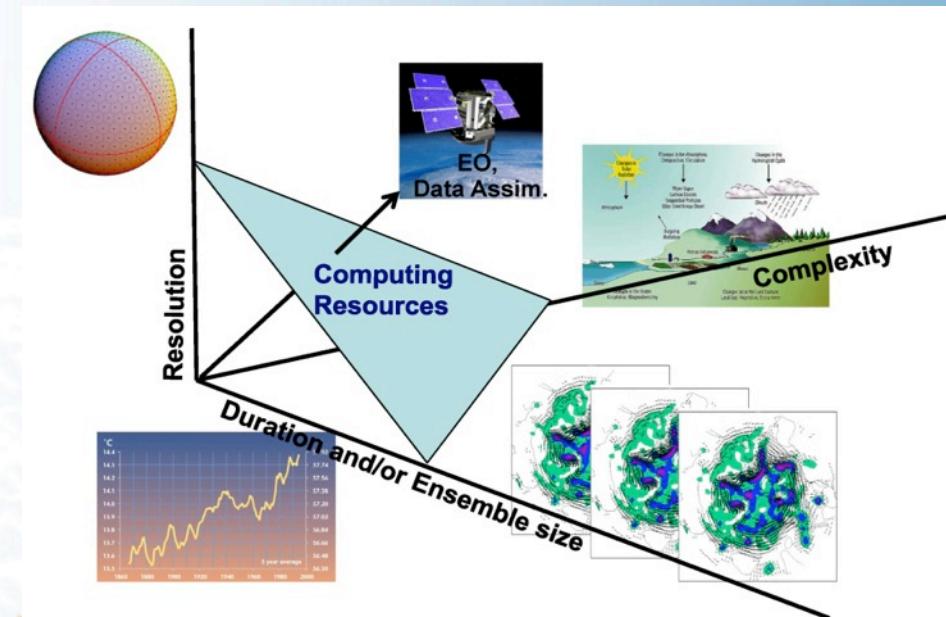
- **What will we run at exascale?**
 - Ultra-high resolution models
 - Data assimilation
 - Very large ensembles of low-res models
- **NCAR Exascale strategy**
 - Computational optimizations
 - Data optimization
- **Summary**

What will we run at exascale?

- **Single ultra high-resolution models**
 - Capture convective-scale processes
 - Non-hydrostatic; 1 km resolution
 - Most vulnerable to a system component failure
- **Climate system data assimilation**
 - Study predictability on seasonal to decadal scales
 - EKF scalable but very data-intensive: will stress memory – I/O subsystem
 - EKF can be made fault resilient

What will we run at exascale?

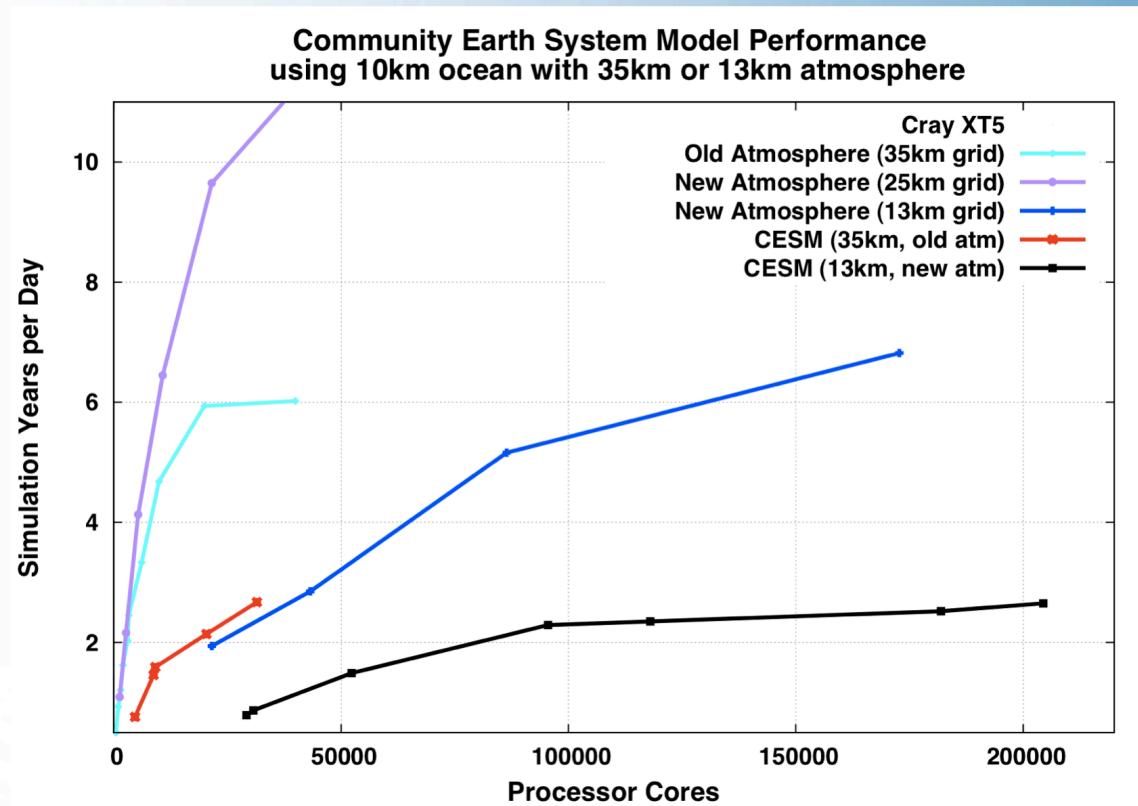
- Very large ensembles of “low-res” models
 - Study natural variability and extremes
 - High I/O volumes – I/O byte/flop goes up for low-res
 - Fault tolerant – resubmit ensemble members that fail.



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CESM Computational Performance (courtesy of Pat Worley)



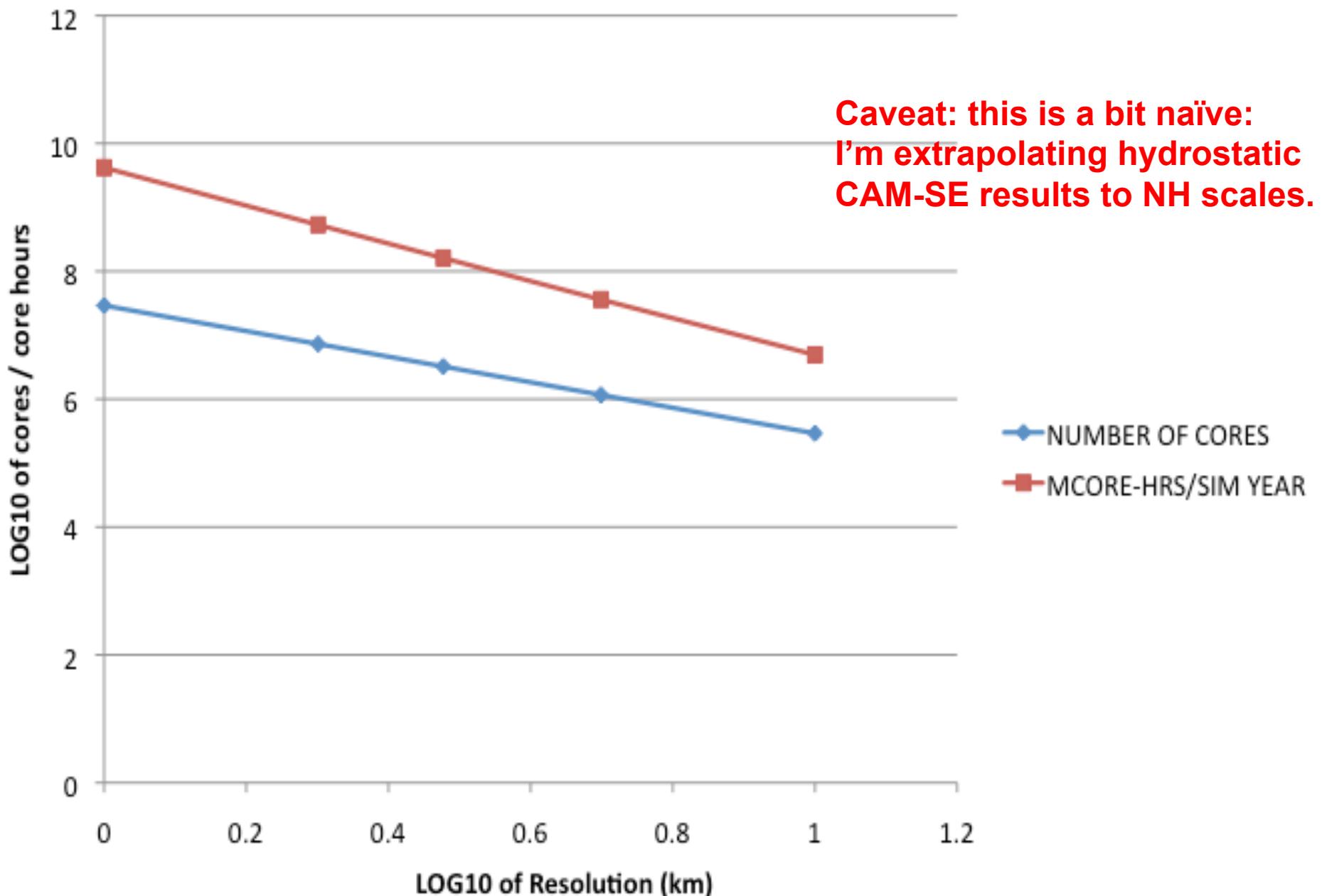
JaguarPE

For 35km CAM / 10km POP/CICE, CESM constrained by CAM, CICE and POP

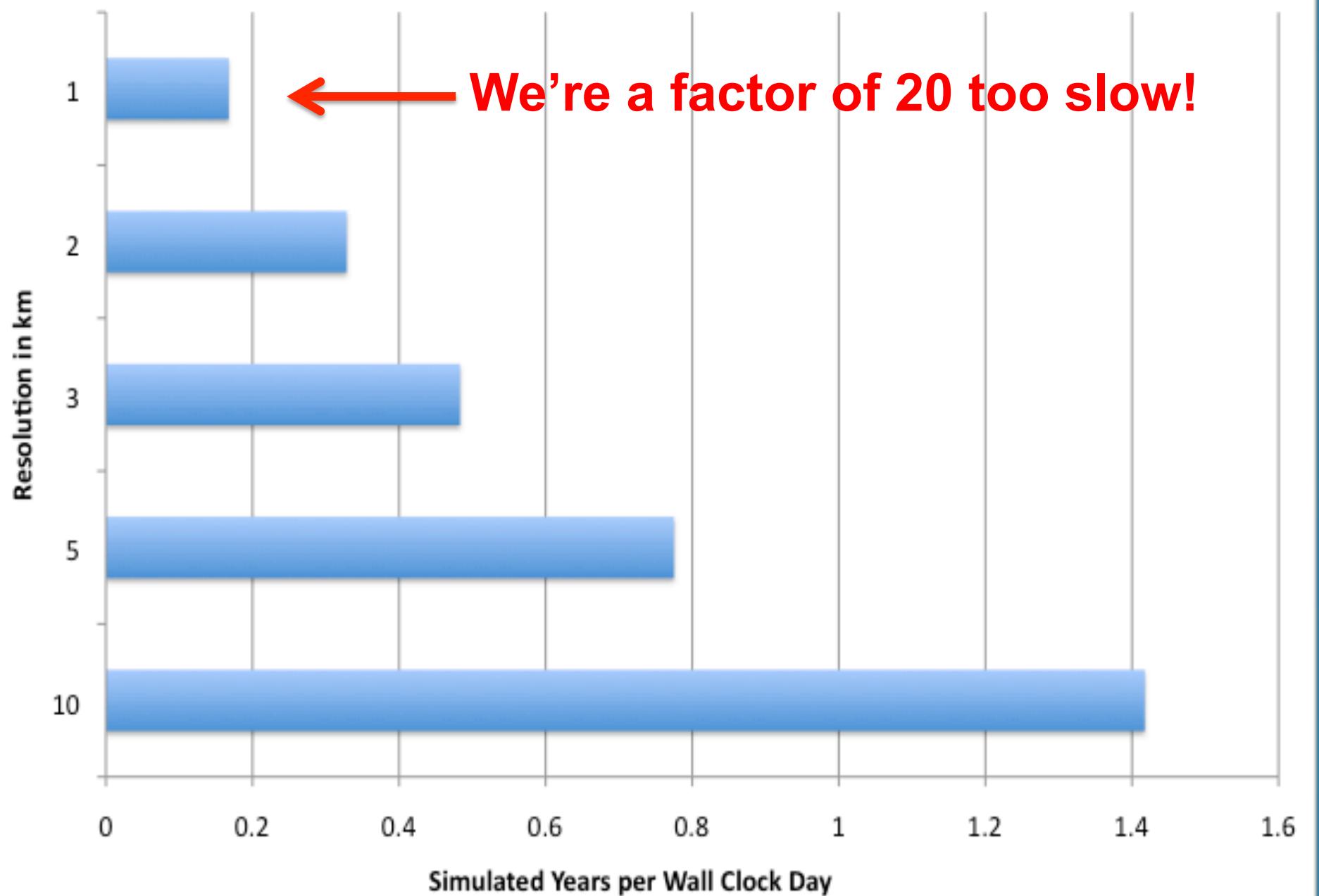
For 13km CAM / 10KM POP/CICE, CESM constrained by CAM

Spectral element-based atmospheric dynamics permits scalable CESM performance at high resolution.

Extrapolation of cost and parallelism: 10 km - 1 km

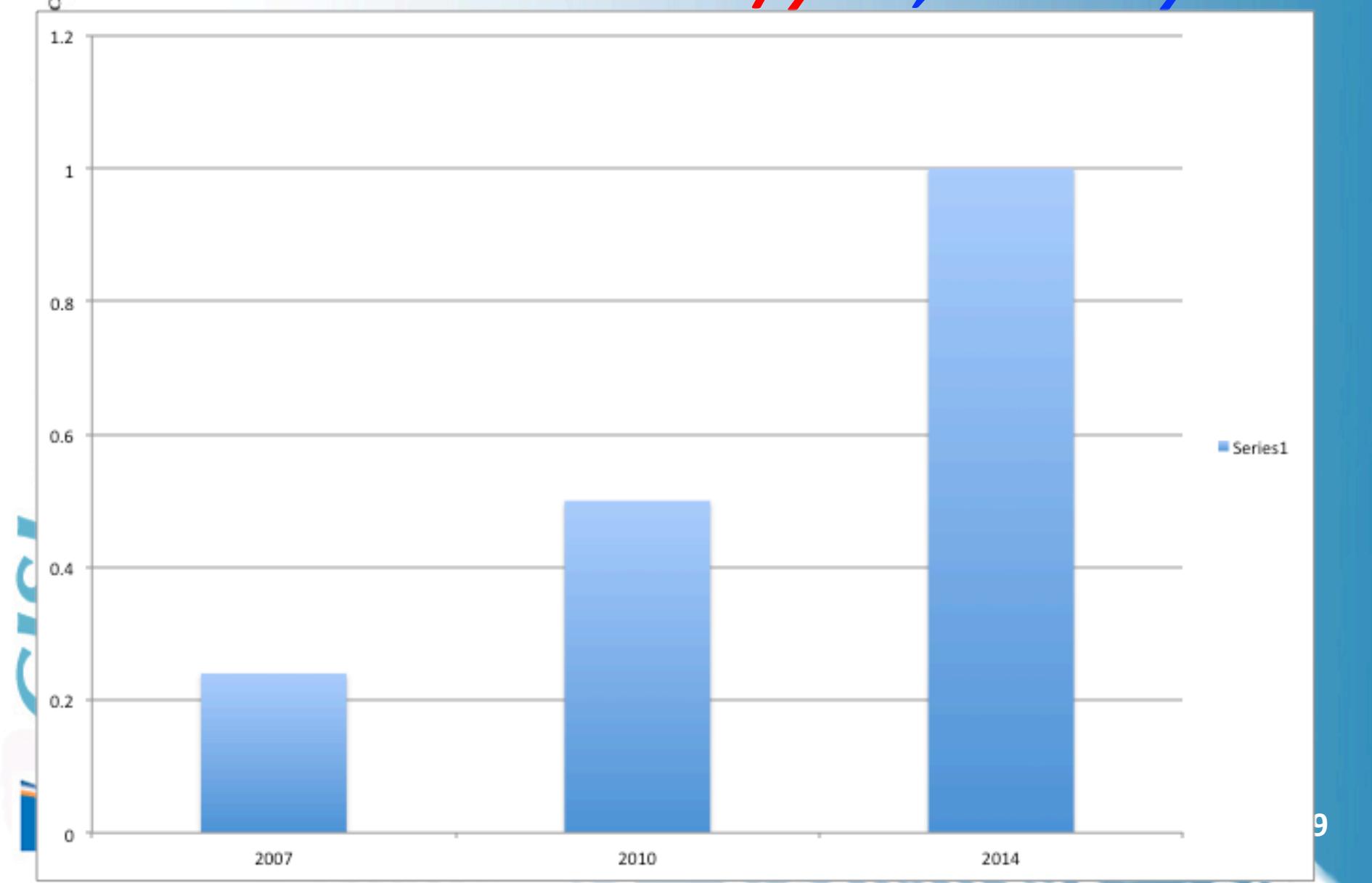


Extrapolated Integration Rate vs Resolution

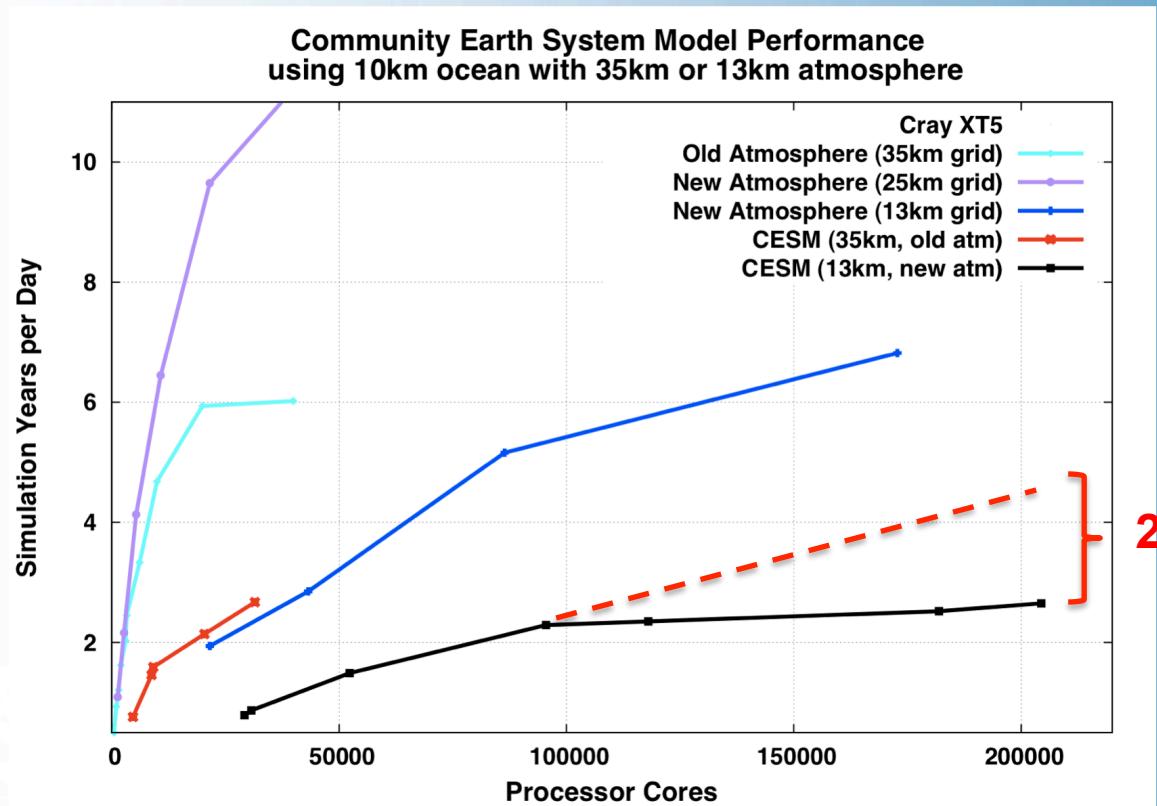


Rate of improvement of memory bandwidth

IBM POWER series: 22.6%/year; ~3.4x by 2020



CESM Computational Performance (courtesy of Pat Worley)



Spectral element-based atmospheric dynamics permits scalable CESM performance at high resolution.

Further improvements will require optimization of ocean and sea ice models

Can accelerators fill in the missing performance 3-4 x?



Discontinuous Galerkin Gradient Kernel

- Analogous to vector calculus calculation in the atmospheric dynamics of CAM-SE component of CESM

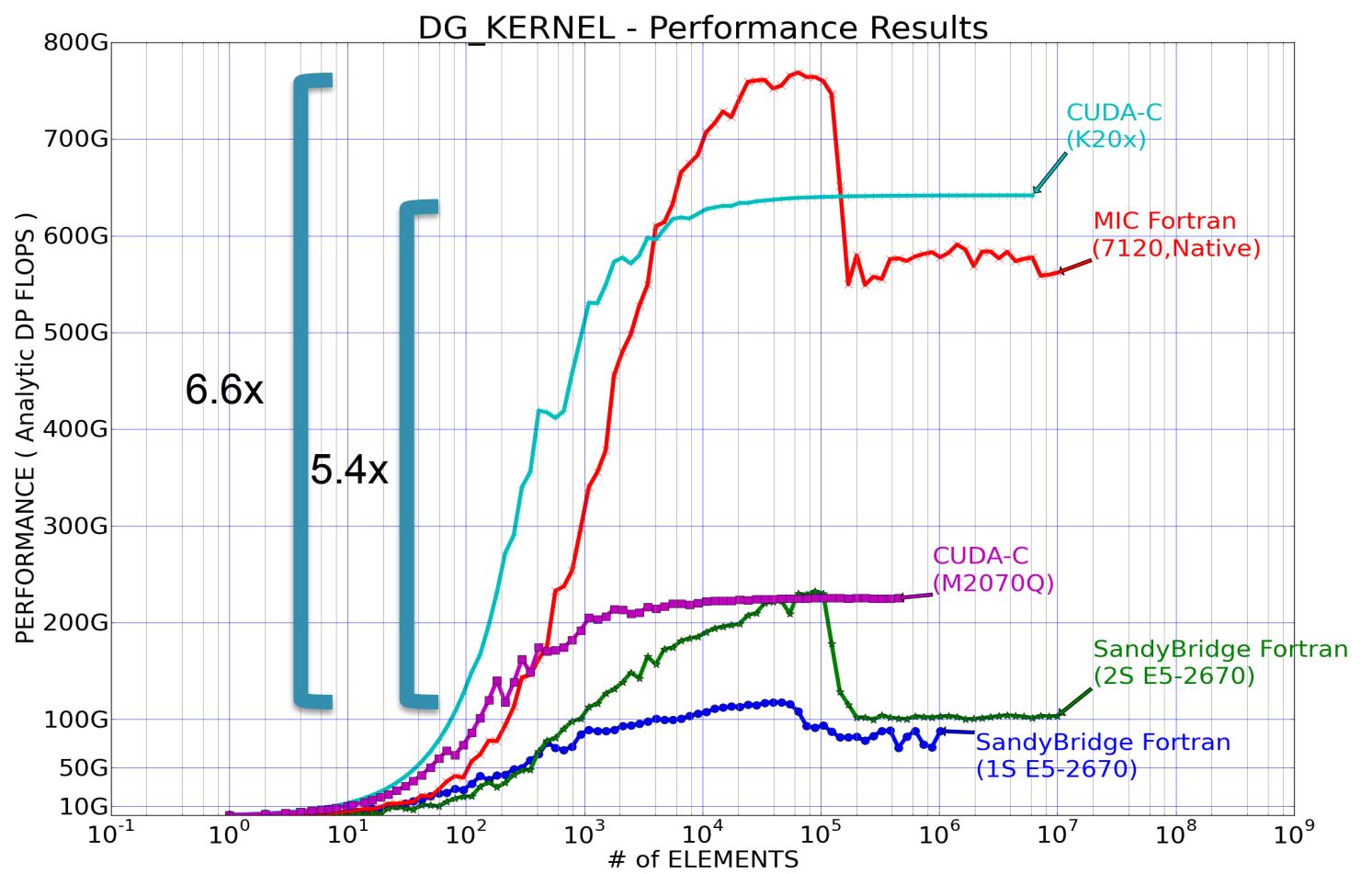
DG and CG both highly scalable, element-based method*

Procedure for N elements:

- Each 4x4 element contains a scalar quantity
- Do two 4x4 double precision matrix multiplies of a derivative matrix with the element and its transpose to get x and y components of gradient
- Use gradients to update flux vectors
- Repeat

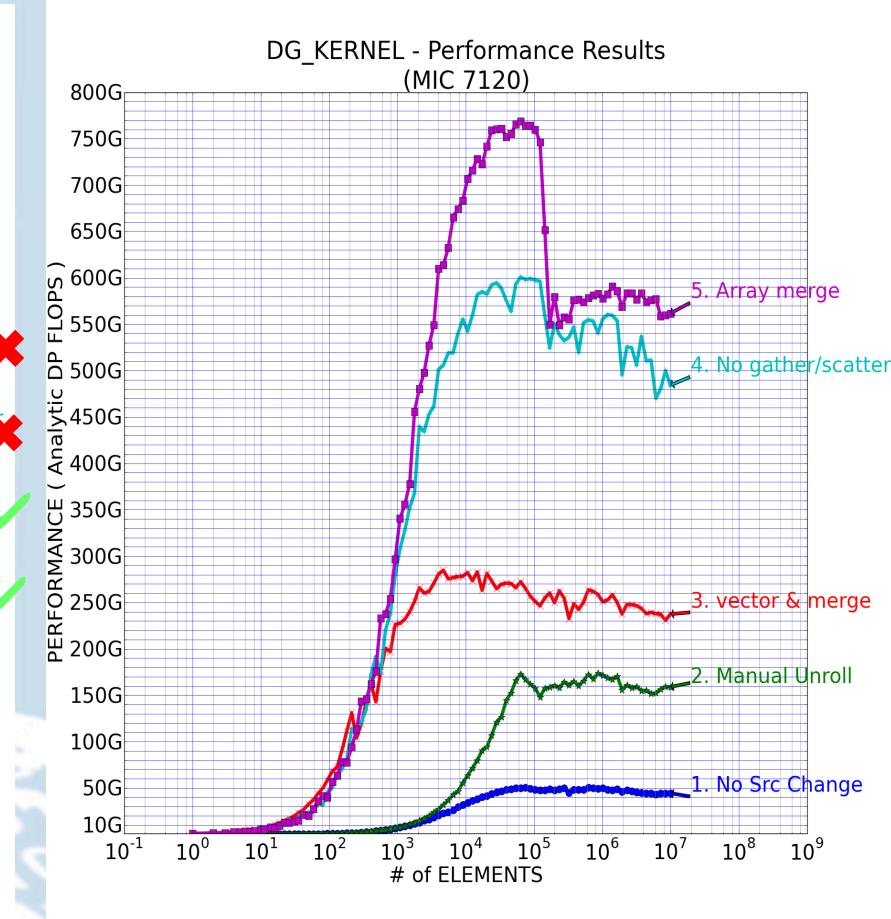
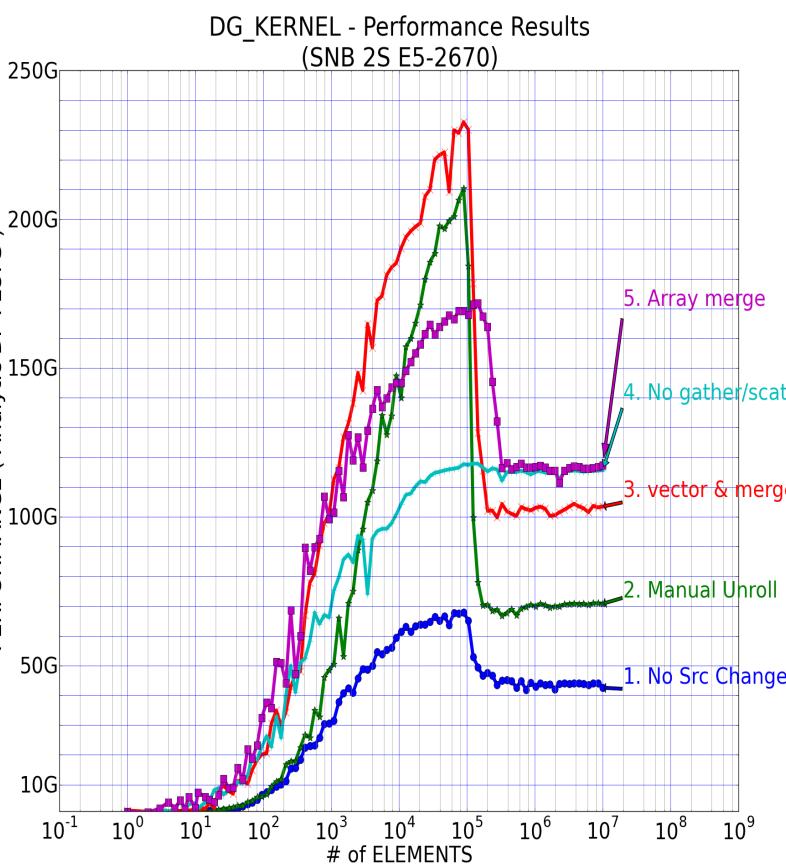
* R. D. Nair, Stephen J. Thomas, and Richard D. Loft: A discontinuous Galerkin global shallow water model, *Monthly Weather Review*, Vol. 133, pp 876-888

The best performance results from CPU, GPU, and MIC



Optimization portability between Xeon and Xeon Phi

CICE Climate Laboratory



Generally, performance tuning on a micro-architecture also helps to improve performance on another micro-architecture. However, it is not always true.

How do you get beyond kernels?

- **Lots of code (1.5M lines in CESM)**
 - Single source multiple architectures
 - portability & performance resilience
- **Code is rapidly changing**
 - Optimization cycle of 1 month – OK
 - Optimization cycle of 6 months – start over
- **Verification/validation requirements**
 - Most restrictive in climate but some sort of software verification/science validation is always required.

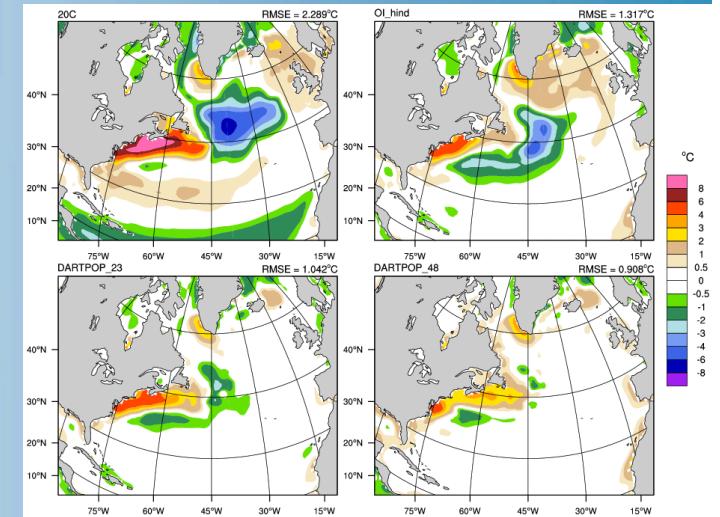
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What about Data Assimilation Using Ensemble Kalman Filter Methods?

Decadal prediction problem

- Fully coupled CESM
- 10 km resolution
- 50 Ensembles members
- 10 Different Start dates
- 10 Years in length
- State initialized using the Data Assimilation Research Testbed (DART)

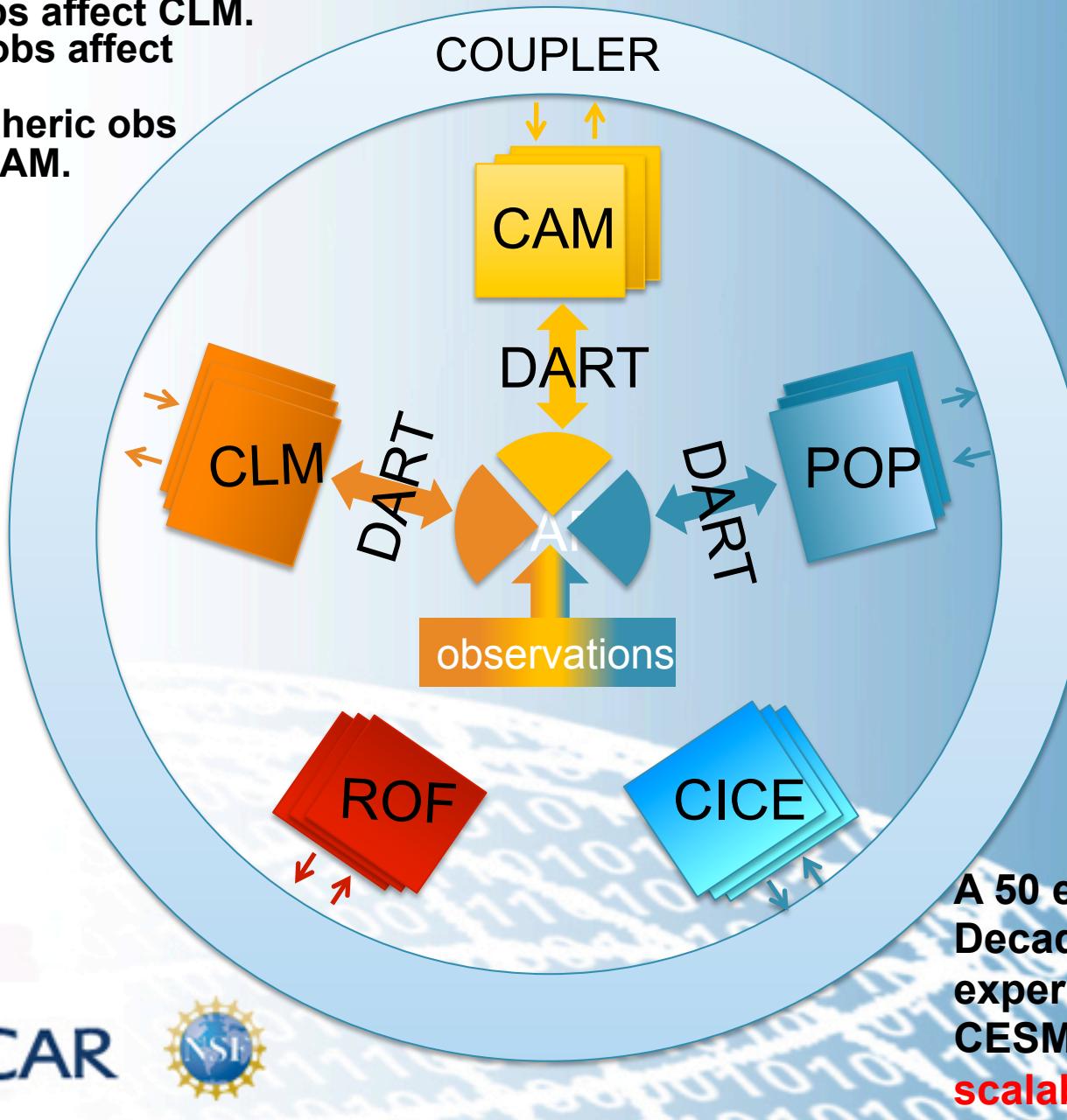


Ensemble Ocean Reanalysis
IPCC Decadal Forecast ICs

50 members x 200,000 cores = 10 million cores

Climate Data Assimilation with DART...

- Land obs affect CLM.
- Ocean obs affect POP.
- Atmospheric obs affect CAM.



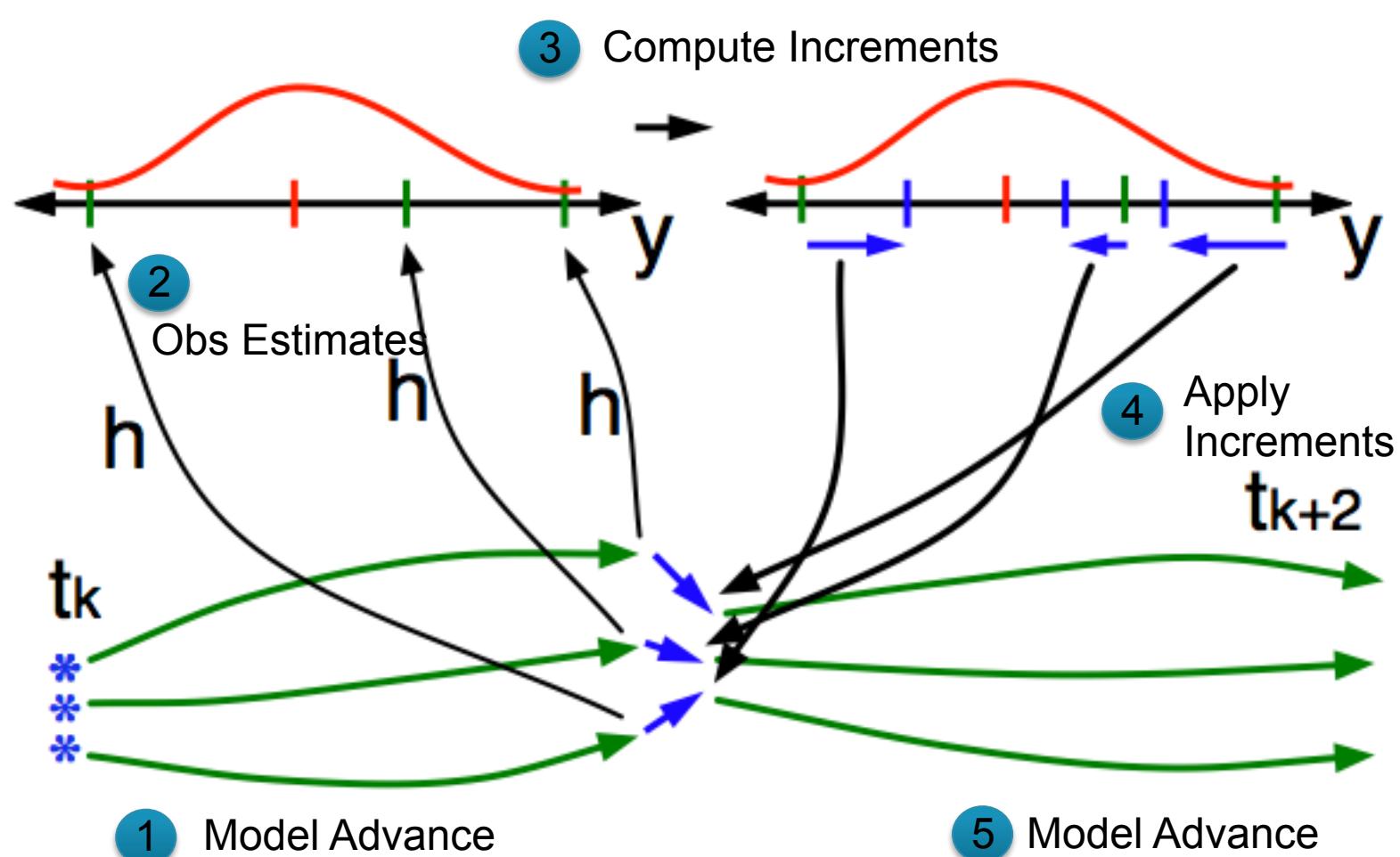
A 50 ensemble member
Decadal prediction
experiment with
CESM 0.1° needs a
scalable DART

CISL
Computational & Information Systems Laboratory

NCAR



DART's EKF Data Assimilation



EKF Scalability Challenges

- **Data Distribution**
 - Observations non-uniformly distributed on globe
 - Load imbalances
- **Data Movement**
 - At least 3 distinct data decompositions in the current software implementation
 - Message passing costs
- **Data Volume**
 - Running 50-100 copies of a large model means 50-100x the data input and output
 - Each 10 km instance = $O(10)$ TB/simulated year
 - 10 simulated years x 50 member ensembles = $O(5)$ PB/decadal forecast

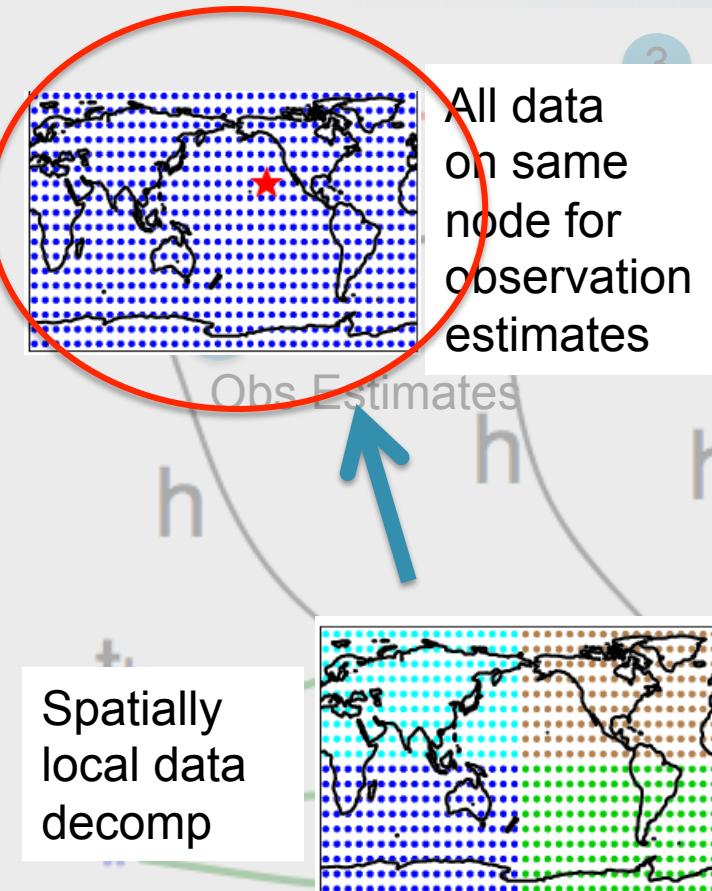
Current DART Data Decompositions

aboratory

CICE



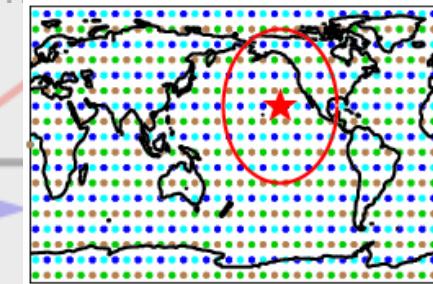
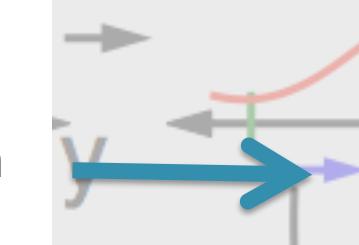
Spatially local data decomp



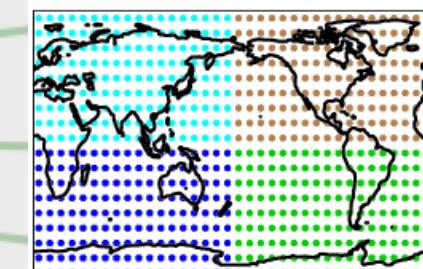
1 Model Advance

2

Compute Increments



4 Apply Increments
 t_{k+2}



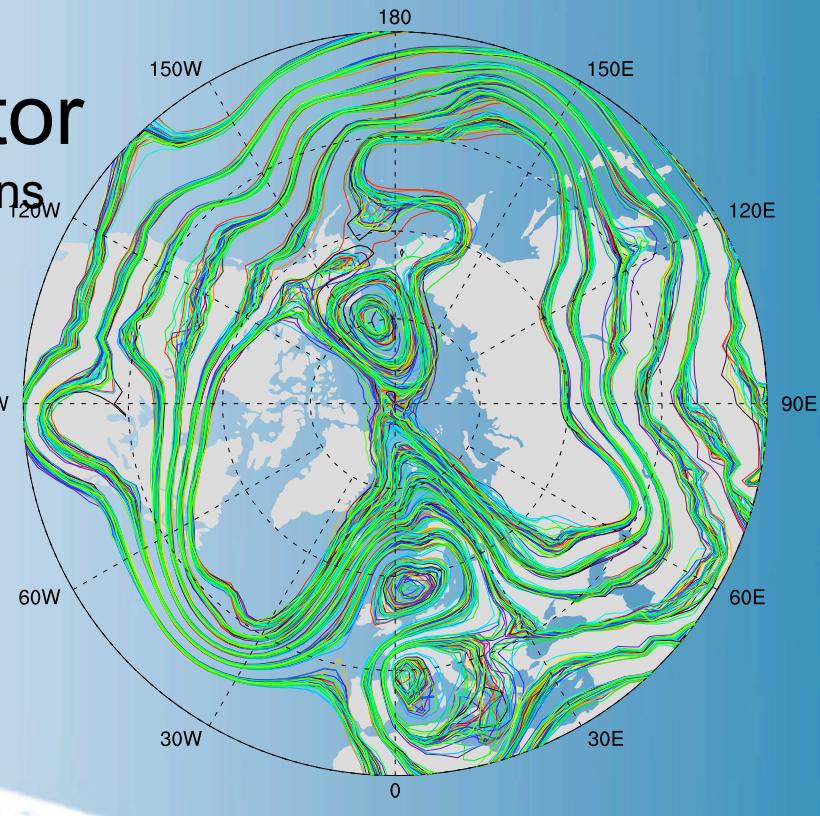
5 Model Advance

DART scalability needs work

CAM FV forward operator

Specific humidity only : 23 090 observations

processors	512	4096
DART (original)	1.01s	0.96s



CONTOUR FROM 5200 TO 5700 BY 100

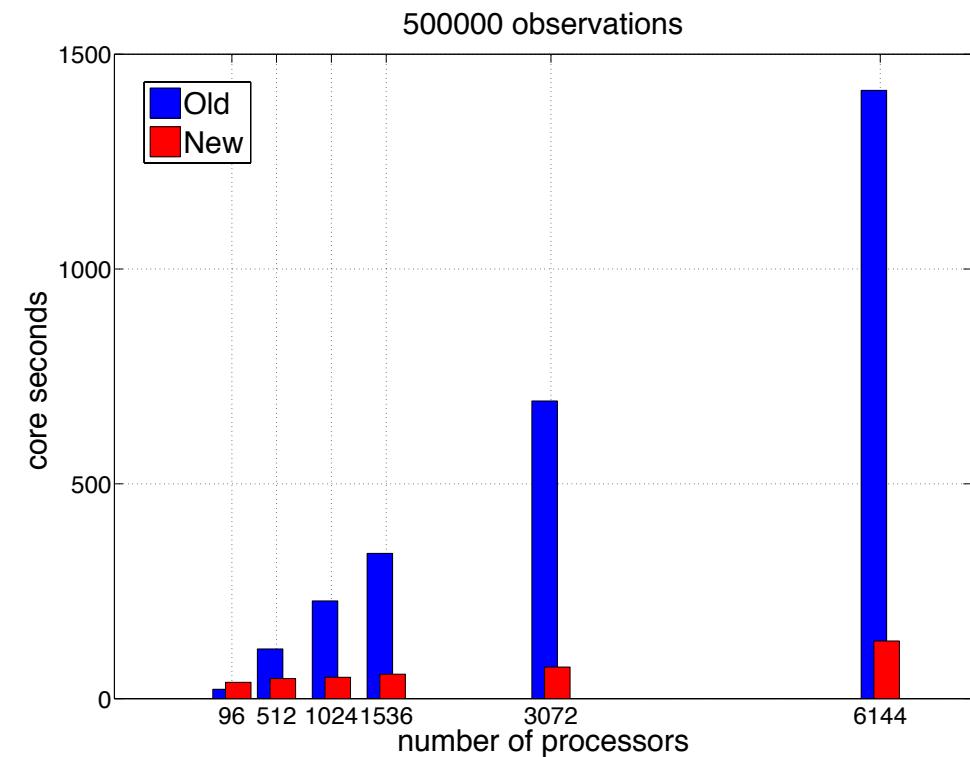
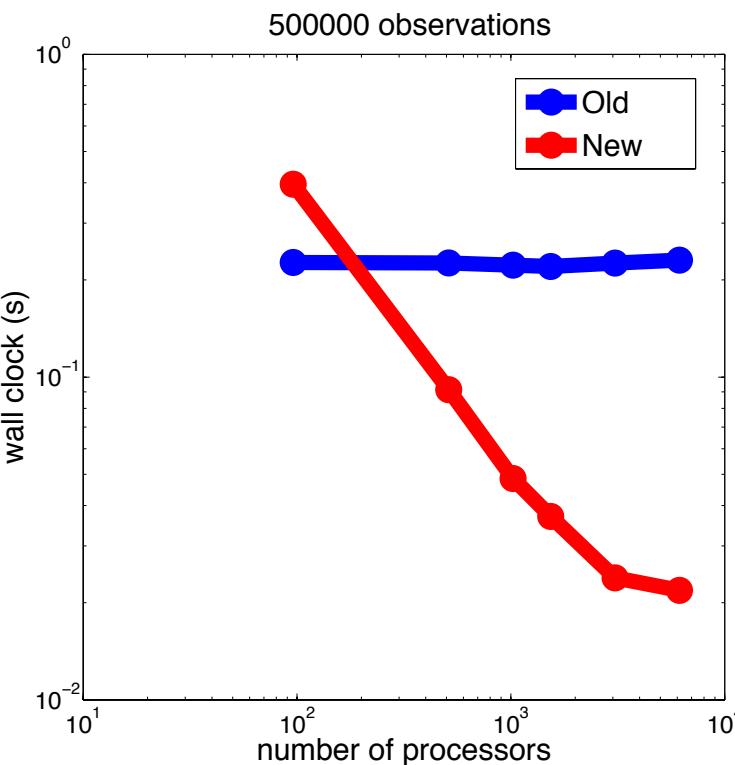
WRF forward operator

54, 400 observations

processors	1024	4096
DART (original)	0.6s	0.6s

DART: Parallelization of fwd operator: Lorenz_96 test case

latory



Time

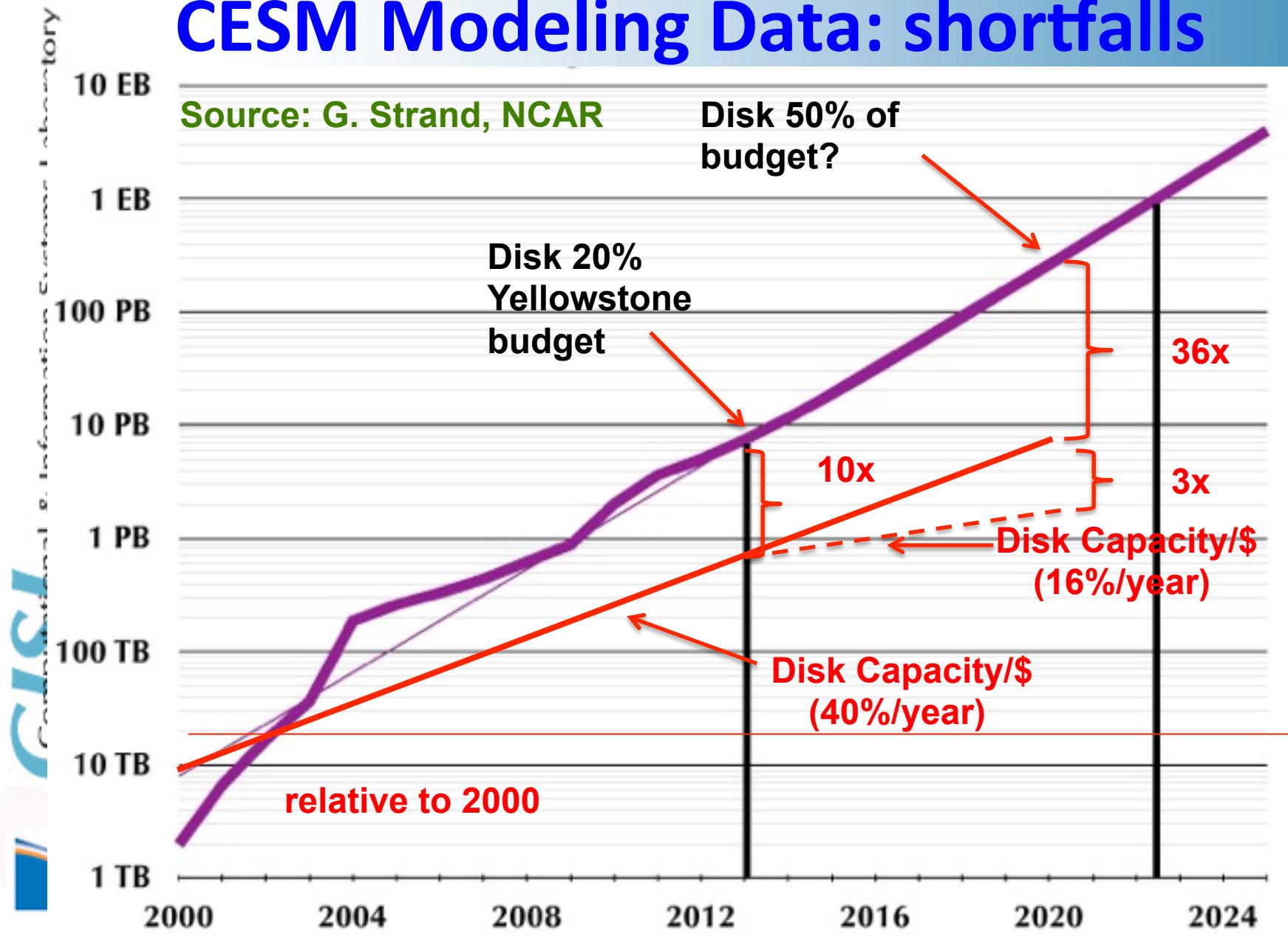


Cost

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Historical and Projected Volume of CESM Modeling Data: shortfalls



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Strategy Overview

- **Investments**

- 4 new staff in last 18 months. More needed!

- **Partners**

- DoE (long-standing)
 - **G8 ECS project** (3 years)
 - **Intel Parallel Computing Center** (new)

- **Workshops/Meetings**

- NCAR workshop on “Programming weather, climate, and earth-system models on heterogeneous multi-core platforms”

- **Platforms**

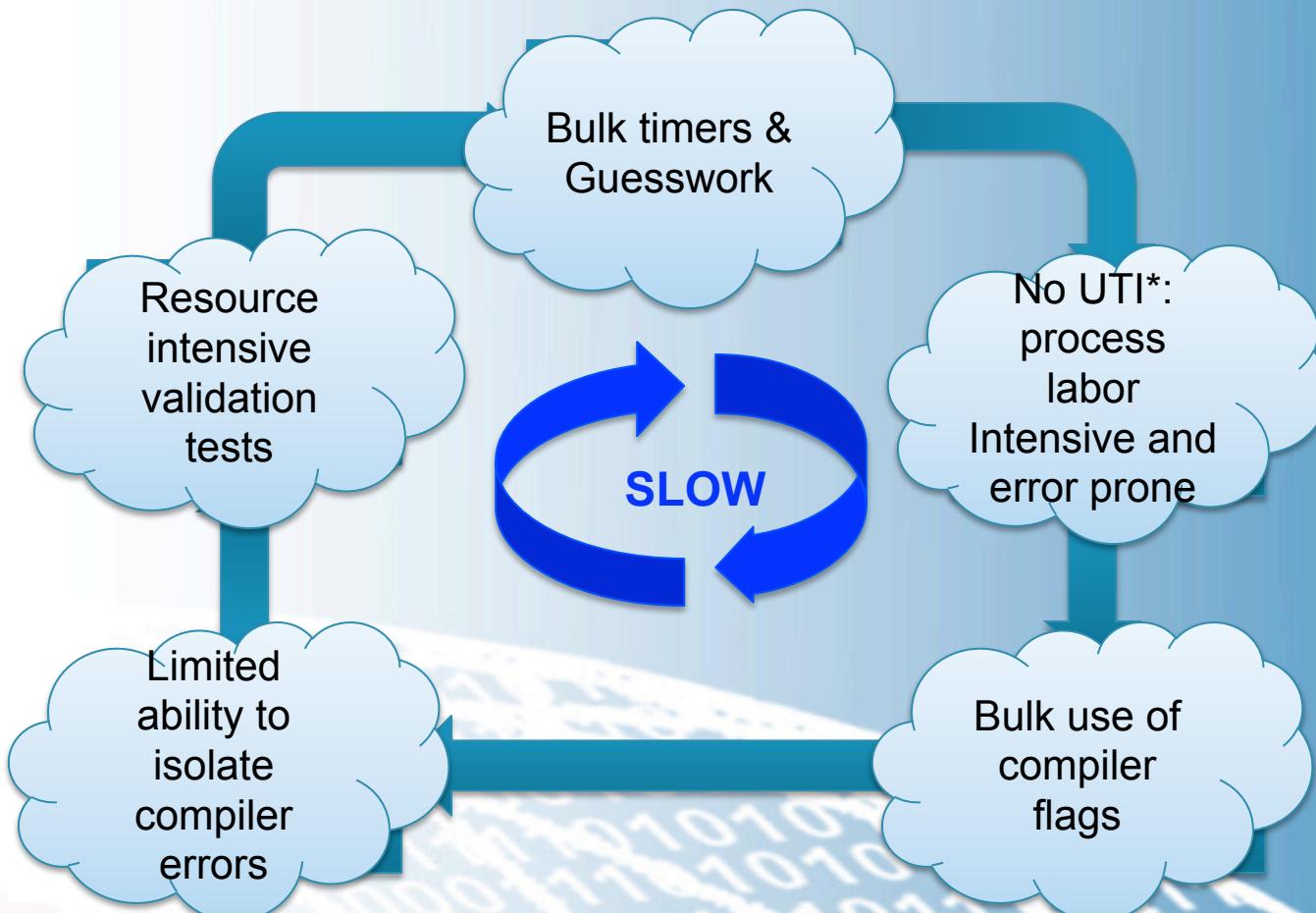
- Locally: 16-node Caldera (NV), Pronghorn (Xeon Phi)
 - U.S.: Titan, Stampede, etc...



Outline

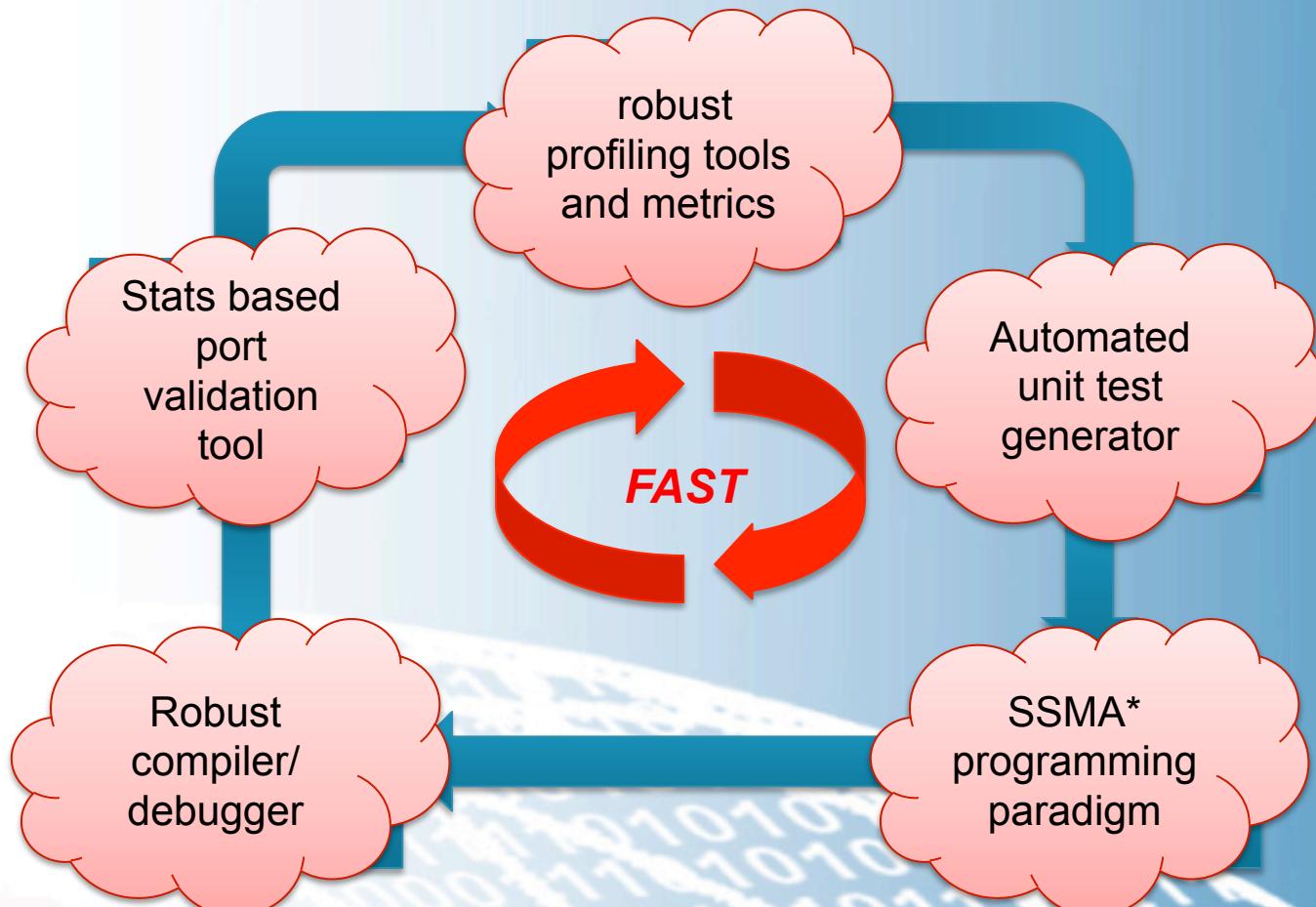
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Life without a Performance Enhancement Methodology

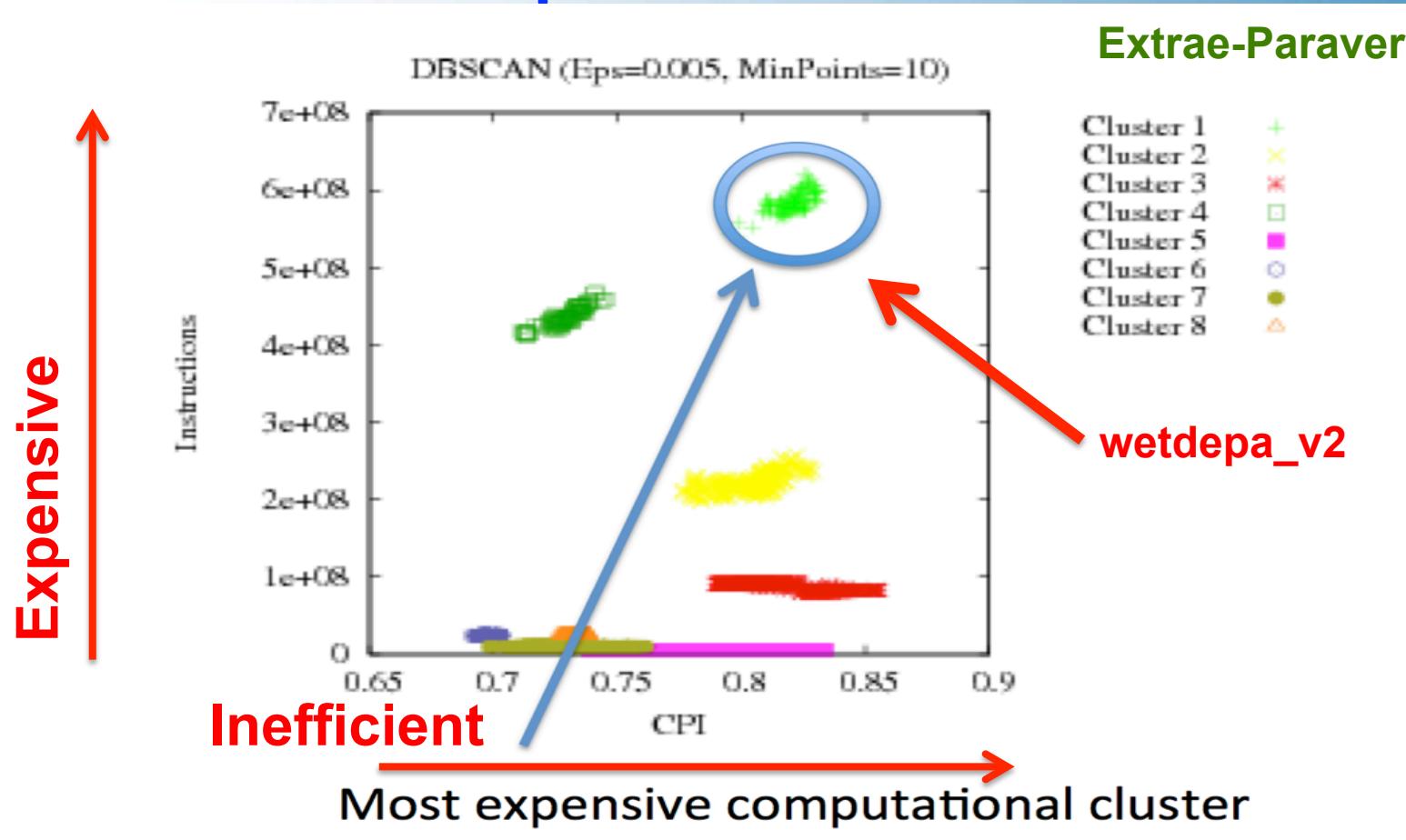


* unit test infrastructure

Performance Enhancement Methodology: a virtuous cycle for code improvement



BSC tools helping to find high priority sections that are expensive *and* inefficient.



- Result of an Exrae trace of CESM on Yellowstone.
- Vertical ~ time; horizontal ~ 1/flops

Now use what we've learned from dg_kernel to speed up wetdepa_v2... to get fast and right!

Computational & I

Single core

	Intel Phi (Intel 13.1.1)			Intel Sandybridge (Intel 13.1.2)		
	-O2	-O3	-O3 -fast	-O2	-O3	-O3 -fast
orig	42.85	41.24	3.74 	3.43	3.32	0.97
mod	6.50	6.61	4.58	1.09	1.12	1.04

- wetdepa is small only ~600 lines
- Restructured branched loops + promoted scalars to vectors.
- -O3 fast for original code gave incorrect results
- 2.5% to 0.7% of code execution time = \$222K savings

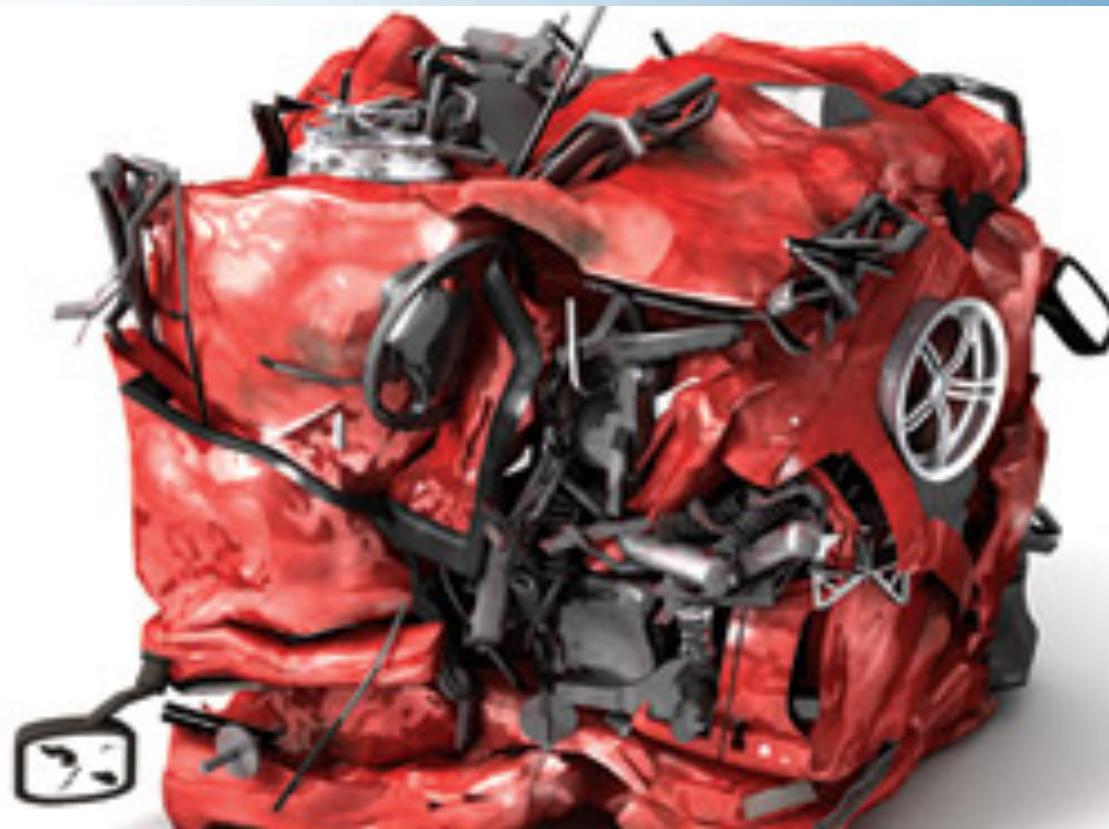


Significant gains possible from code refactoring!

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What can lossy data compression get us?



Lossy data compression evaluation metrics

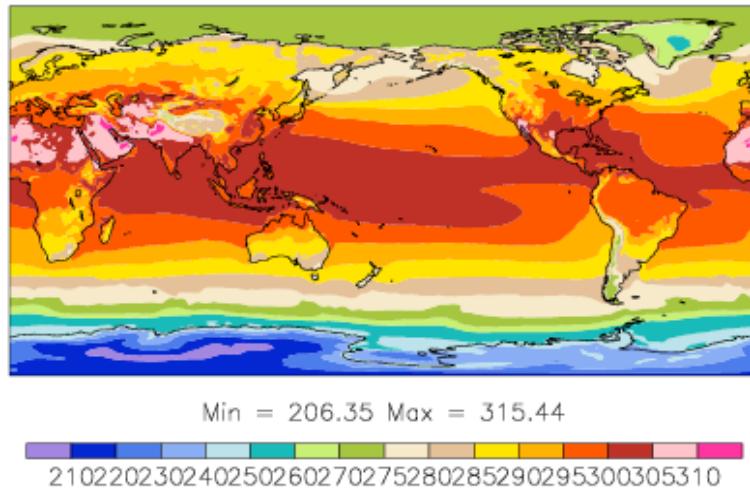
- **Pearson Correlation:** < 0.99999
- **RSMZ-ensemble test**
 - Choose single ensemble member
 - Compress/decompress member
 - Does decompressed members z-score still belong to ensemble?
- **RSMZ-bias test**
 - Compress/decompress all members
 - Calculate z-score versus uncompressed ensemble
 - Compare z-score of compressed versus original
 - Does compression/decompression introduce bias?
- **Ermax ensemble test**
 - Compress/decompress all members
 - Calculate Ermax for uncompressed ensemble
 - Calculate Ermax for original ensemble

T_s (Surface Temperature)

JJA

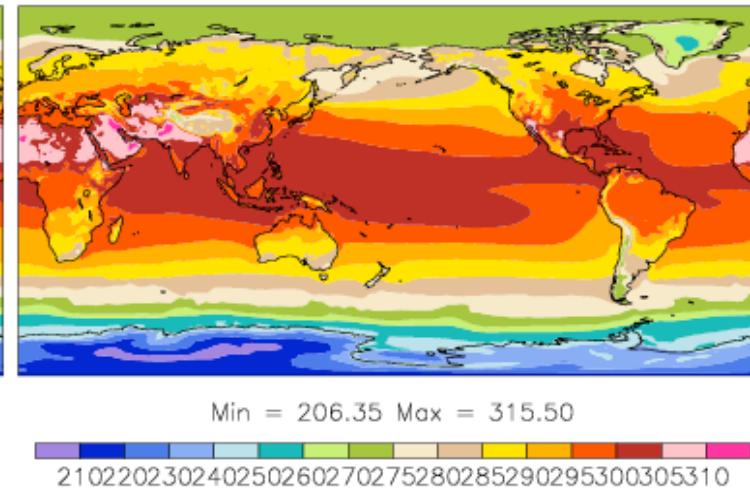
LRC01

Surf Temp (radiative) mean = 290.09



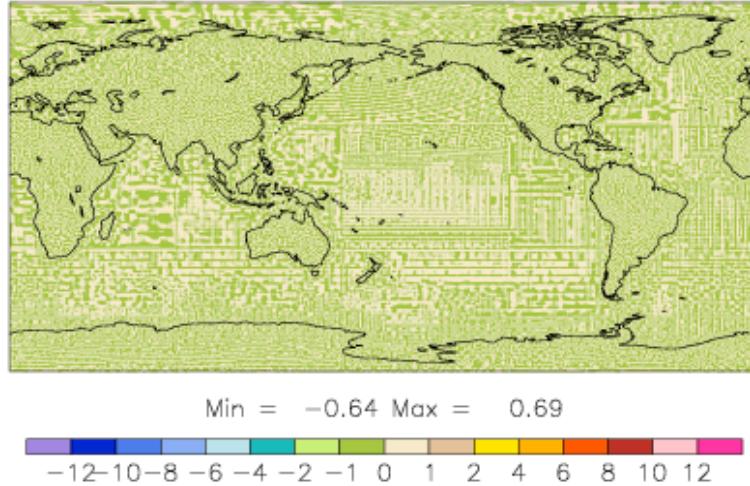
LRC01

K Surf Temp (radiative) mean = 290.09



LRC01 - LRC01

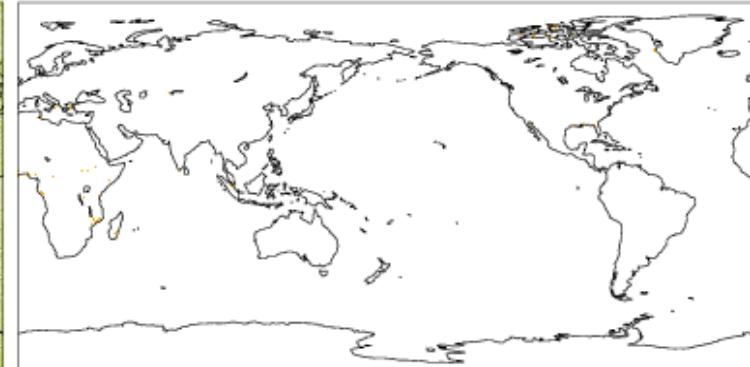
mean = -0.00 rmse = 0.07



T-test of the two means at each grid point

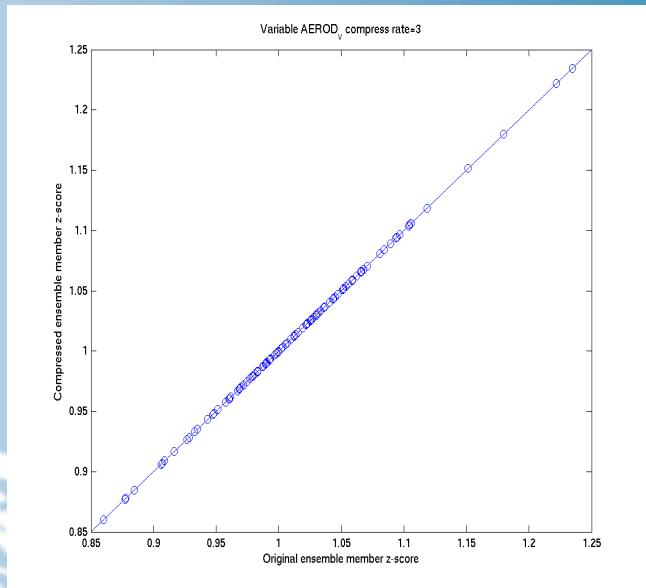
Colored cells are significant at the 0.05 level

K



Lossy Compression – 5x

- **Using fpzip**
- **For a variable (u) choose highest compression rate such that all pass**
 - Pearson Correlation ✓
 - RMSZ-ensemble test ✓
 - Ermax-ensemble test ✓
 - RMSZ-bias test ✓
- **170 variables**
 - 7 variables: fpzip-32 [68%]
 - 50 variables: fpzip-24 [39%]
 - 113 variables: fpzip-16 [15%]
- **Overall 18% of original file size**



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Conclusions for 2020

- **20x** integration rate shortfall can be addressed through improvements in scalability, memory bandwidth and node performance optimizations (with accelerators?)
- **3.6x** climate data storage shortfall can be addressed with lossy compression without sacrificing climate information
- We are ramping up significant new staff/tech investments to accelerate our rate of progress in both areas.



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Thanks!