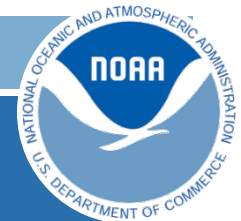


## US Climate Modeling Summit

# Challenges in High Performance Computing (HPC) for Climate Prediction and Projection

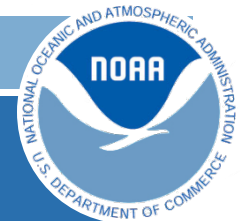
Frank Indiviglio  
Production Lead, High Performance Computing  
June 27, 2016

Acknowledgements: Bill Collins (DOE), Bill Putman (NASA), John Michalakes (DOD), V.Balaji (NOAA)



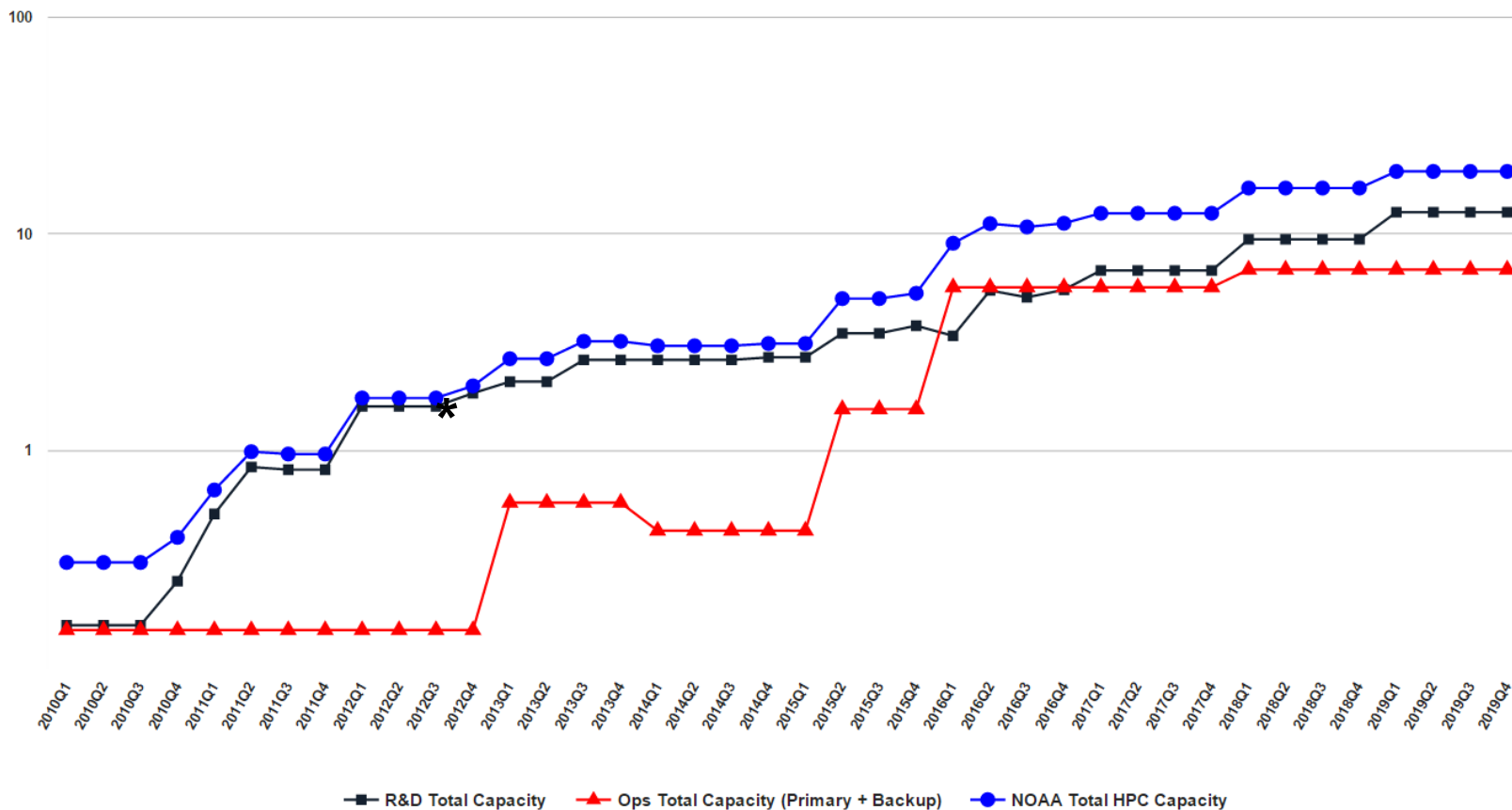
## Driving Questions

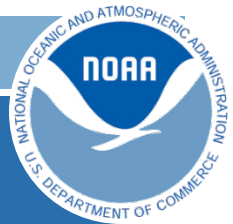
- What horizontal and vertical resolutions are necessary to adequately resolve processes in the coupled system that drive both prediction error in short term forecasts and climate simulation bias?
- What is the computational cost of the key biogeochemical/physical processes must be included in models to address mission requirements?
- What is the ideal size of the ensemble needed for this effort both for prediction, for understanding coupled processes and biases, and quantifying uncertainty?
- What modeling improvements will most significantly impact computing and storage requirements (e.g., resolution, processes/complexity, ensemble members, etc) and system balance (between compute, networking, storage, etc)?



# NOAA HPC Growth

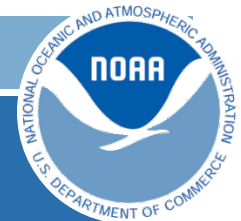
NOAA's High Performance Computing Capacity





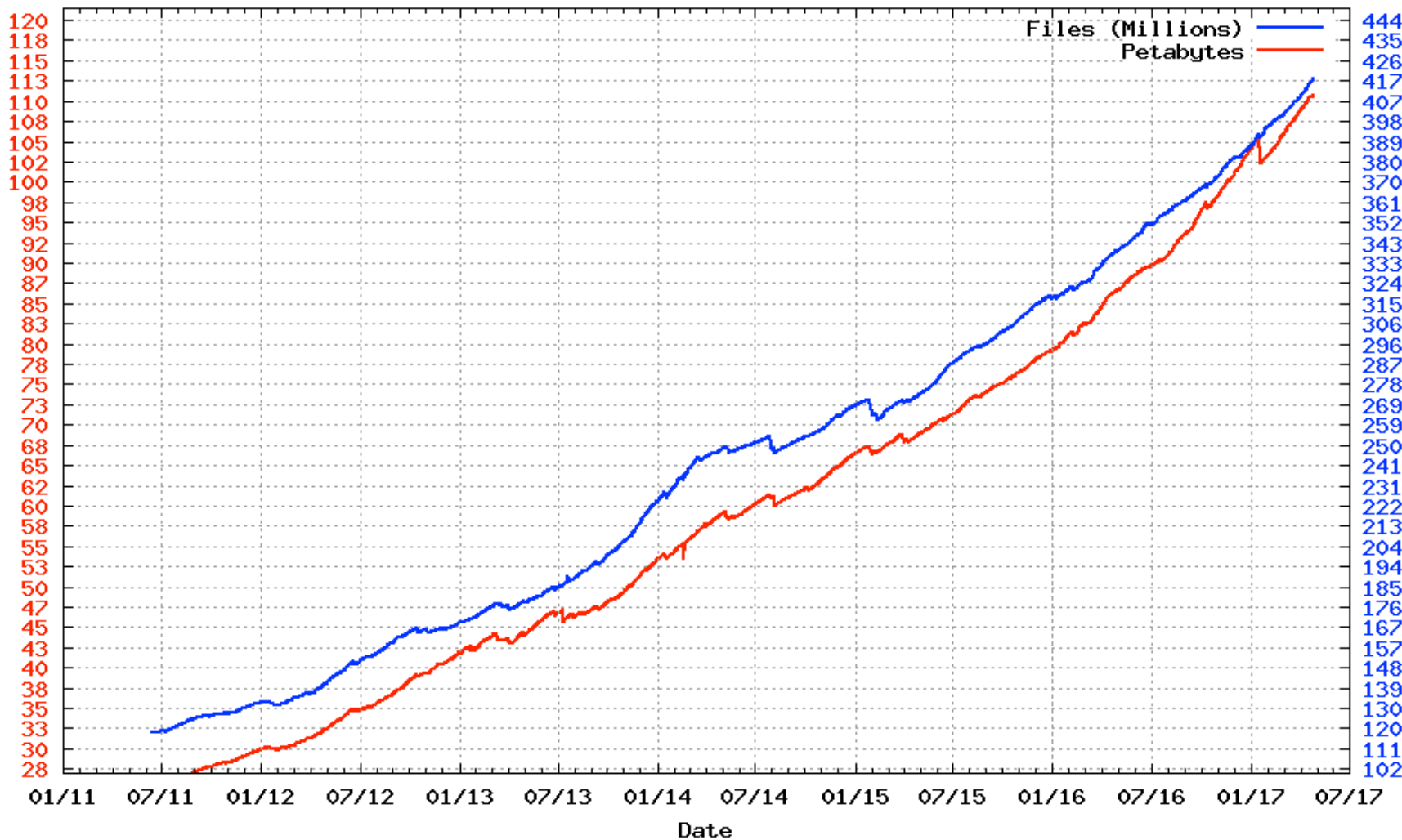
# Data Challenges

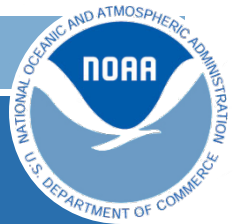
- How will increasingly high-resolution data be stored and shared for community research?
- As resolution increases, it becomes more difficult to save every bit to disk. How can we reduce the storage burden from coupled hi-res integrations?
- What must be analyzed at full resolution, and what can be evaluated at coarser spatial resolution?
- What aspects of your analysis can be in-lined during computation to reduce the required storage?
- How does increased horizontal resolution impact the necessary temporal resolution of your analysis and data storage?
- What new technologies, such as non-volatile random-access memory (NVRAM), provide the greatest potential to improve the scalability and efficiency of your coupled systems and particularly IO bottlenecks that are inevitable at high resolution?



# NOAA Data Archive Growth

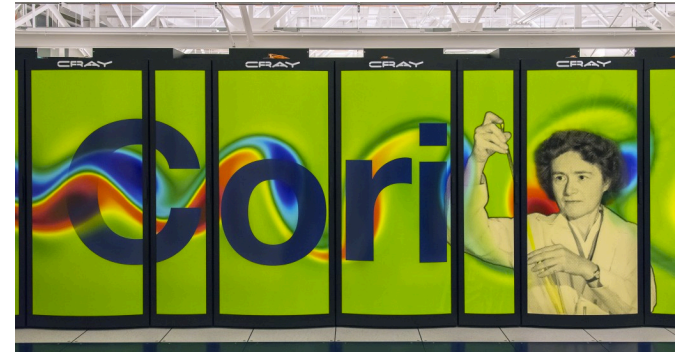
Archive Usage



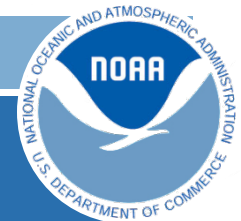


# Today's DOE Leadership Systems

- NERSC Cori (Phase II)
  - >31.4 Pflops
  - 29+ Pflops Xeon Phi
  - 32 Core Hazwell and 68 Core Xeon Phi
- OLCF Titan
  - 27 Pflops
  - 16 core AMD Opteron + NVIDIA GPU
- ALCF Mira
  - 10 PFlops
  - 16 core PowerPC







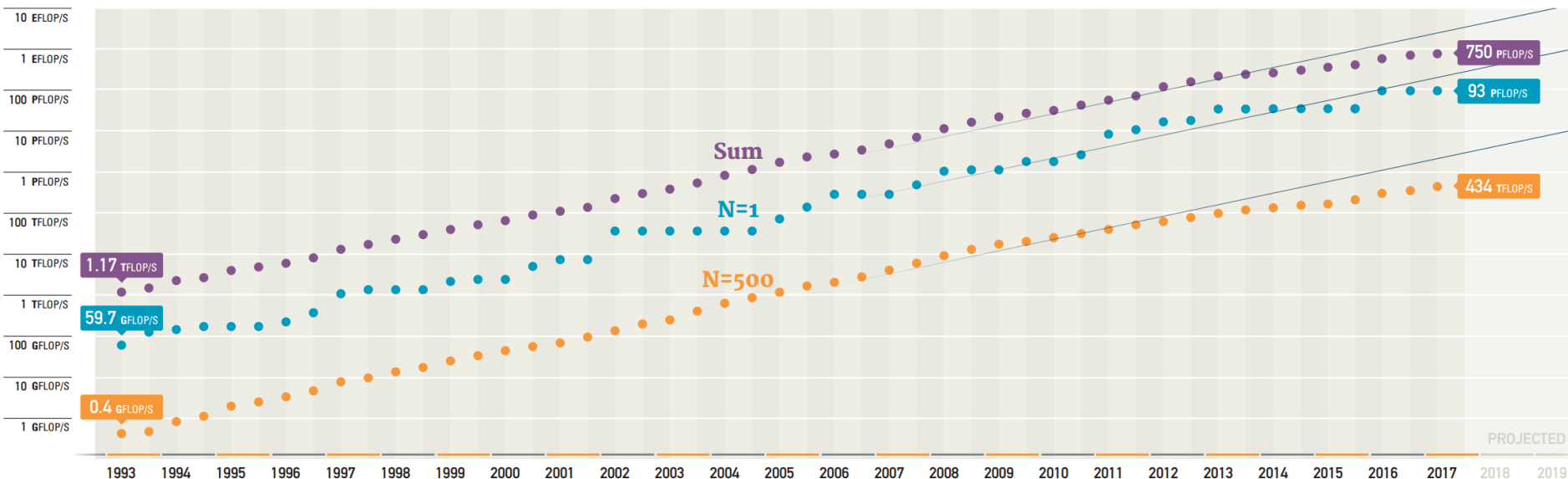
# DOE Next Generation Machines

- Key goal of ACME: Run on next generation DOE machines:
- OLCF Summit 2018
  - ~200 PFlops
  - Multiple IBM power9 and NVIDIA GPUs
- ALCF Aurora 2018
  - ~180 Pflops
  - 50K Nodes, 3<sup>rd</sup> gen Intel Phi



# A Look at The HPC Industry

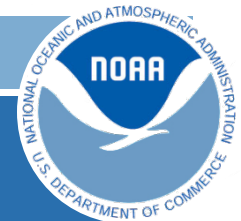
## PERFORMANCE DEVELOPMENT



- The top 500 systems performance continues to flatten
- Accelerated platforms now occupy almost 20% of the list
- The benchmarks in which the industry is using to evaluate performance are changing, HPCG is now being incorporated into the evaluation

Graphic – HPCwire Top 500 Results June 19, 2017 - <https://www.hpcwire.com/2017/06/19/49th-top500-list-announced-isc/>





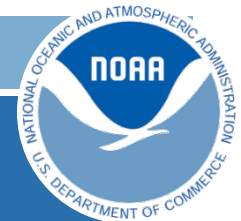
# JPSY Comparison Across ESMs

Model	Machine	Resolution	SYPD	CHSY	JPSY
CM4	Gaea/c2	$1.2 \times 10^8$	4.5	16000	$8.92 \times 10^8$
CM4	Gaea/c3	$1.2 \times 10^8$	10	7000	$3.40 \times 10^8$

- Comparative measures of capability (SYPD), capacity (CHSY), and energy cost (JPSY) per “unit of science”.
- Can you have codes that are “slower but greener”? Algorithms that are “less accurate but more eco-friendly”?
- From Balaji et al (2016), in review at GMDD.

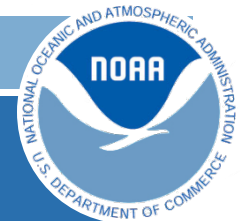


<http://goo.gl/Nj1c2N>



## On software development ....

- Experience to date with fine-grained architectures: kernels can sing ( $\sim 40X$ ), but complex multi-physics codes croak ( $\sim < 2X$ )
- Approach: code revisions for performance on conventional architectures will get us a significant way toward performance on fine-grained systems.
  - Component Concurrency
  - Offload I/O, Diagnostics
  - Performance analysis tools
  - vectorization (requires interaction with compiler vendors)
  - wide halos (to reduce comms)
  - nonmalleable executables (aka static memory)
  - direct use of coarray



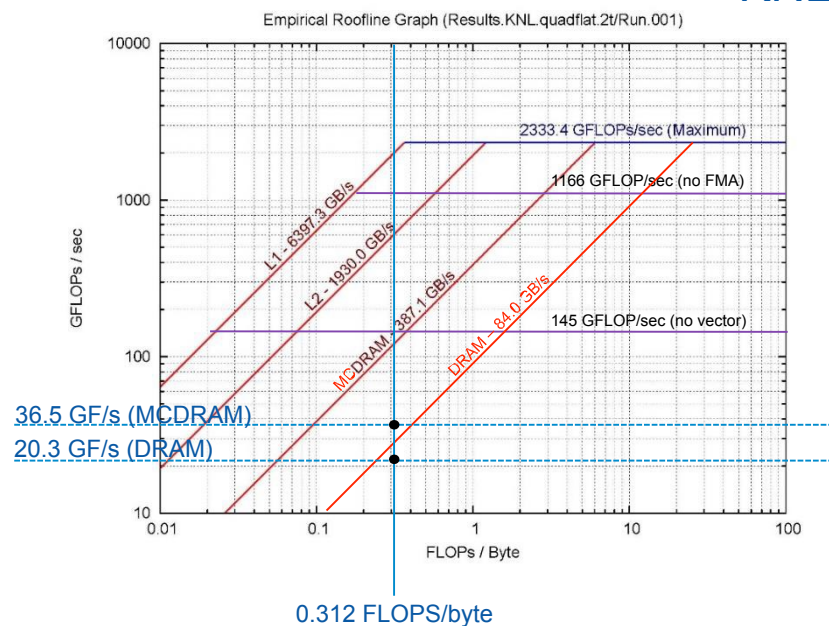
# ACME Mini-App Strategy

- Recommendation of ACME/Exascale study group:  
*Identify key kernels/modules that are small enough so a single person can understand/refactor/rewrite to test new approaches, but that are large enough that successful results are meaningful for ACME.*
- Target: Transport mini-apps for both atmosphere and ocean to cover finite element and finite volume approaches used in ACME
- Atmosphere tracer transport is single most expensive ACME component. Ocean tracer transport is 30% of the ocean model

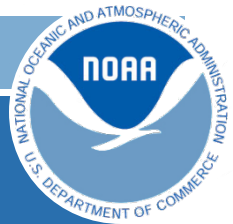
# Neptune Example

- Roofline Model of Processor Performance
  - Bounds application performance as a function of **computational intensity**
  - If intensity is high enough, application is “compute bound” by floating point capability
  - If intensity is not enough to satisfy demand for data by the processor’s floating point units, the application is “memory bound”
    - 128 GB main memory (DRAM)
    - 16 GB High Bandwidth memory (MCDRAM)
  - KNL is nominally 3 TFLOP/sec but to saturate full floating point capability, need:
    - 0.35 flops per byte from L1 cache
    - 1 flop per byte from L2 cache
    - 6 flops per byte from high bandwidth memory
    - 25 flops per byte from main memory
  - Hard to come by in real applications!
    - NEPTUNE benefits from MCDRAM (breaks through the DRAM ceiling) but realizing only a fraction of the MCDRAM ceiling
- John Michalakes – 17<sup>th</sup> Workshop on High Performance Computing in Meteorology

KNL

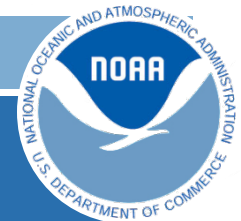


NEPTUNE E14P3L40  
(U.S. Navy Global Model Prototype)



## Other Questions to Consider

- What do workflow and machine policy add to the cost of science?
- What is the scaling “data intensity” of data with compute and how does that change with model resolution?
- How will future architectures effect how we enact these workflows?



## Summary/Discussion

- Driving questions center on what HPC capability and configuration will be need needed to address science priorities.
- Climate agency HPC is growing, but it's unlikely that it's growing fast enough.
- The interaction between simulation, data analytics, and storage needs to be constantly assessed.
- Software innovations to leverage anticipated HPC architectures are hard to implement. And necessary.
- Partnerships on hardware and software have accelerated us toward our goals.



# Questions?



# NOAA's Science Network

