## CESM CMIP6 Data Workflow

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2015, September 14



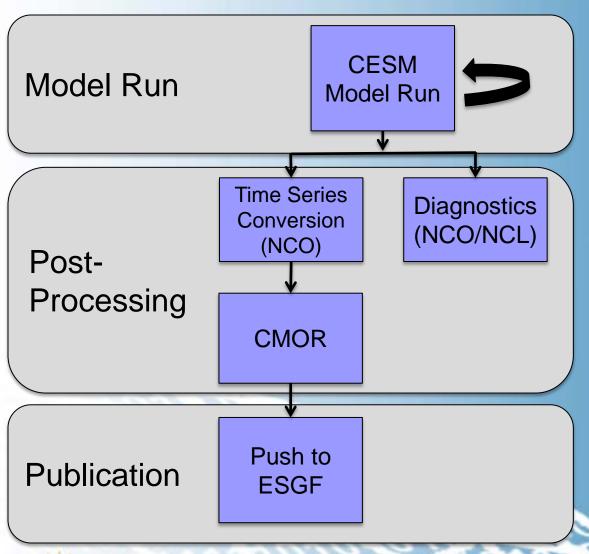
### **CESM Workflow Refactoring Team**

- Ben Andre
- Alice Bertini \*
- John Dennis
- Jim Edwards
- Mary Haley
- Jean-Francois Lamarque
- Michael Levy
- Sheri Mickelson \*
- Kevin Paul \*
- Sean Santos
- Jay Shollenberger \*
- Gary Strand
- Mariana Vertenstein (\* Developer)





### **CMIP5 Workflow**







#### Lessons We Learned From CMIP5

- CESM was the first model to complete their simulations, but the last to complete publication.
   Why?
  - All post-processing was serial
  - Workflow was error prone and took time to debug
  - Too much human intervention was needed between post-processing steps



### Plans for CMIP6 (Preliminary)

- Participate in 26 MIPS
- Low Resolution: 1° atm, 1° ocn
  - Throughput: 14.5 simulated years per wall clock day
  - Cost: ~2,000 core hours per simulated year
  - Total estimated cost: 150 million core hours
- High Resolution: 25km atm, 1° ocn
  - Throughput: 1.84 simulated years per wall clock day
  - Cost: ~215,000 core hours per simulated year
  - Total estimated cost: 350 million core hours
- Data sizes:
  - Raw size: ~12 PB
    - Published: ~6 PB = we will have to process 5TB a day for 3 years





## To quote Jim Kinter, how are we going to post – process the flood of data from CMIP6?



Our current boat won't be able to process the flood of data

CMIP6

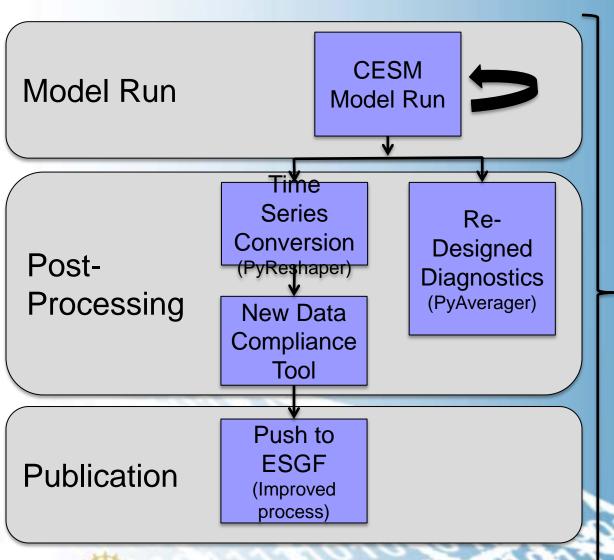
6 PB Current Prediction

CMIP5 2 TB



- We have been examining the individual pieces of the workflow and improving it where necessary
  - Increasing performance: Adding parallelization into the workflow
  - Reducing Human Intervention: Adding in automation
  - Project Management: more formal approach

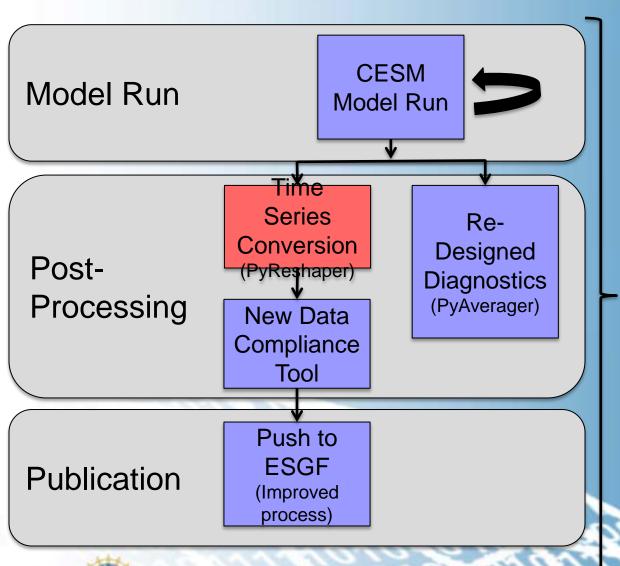




Experiments
Update
Their Status
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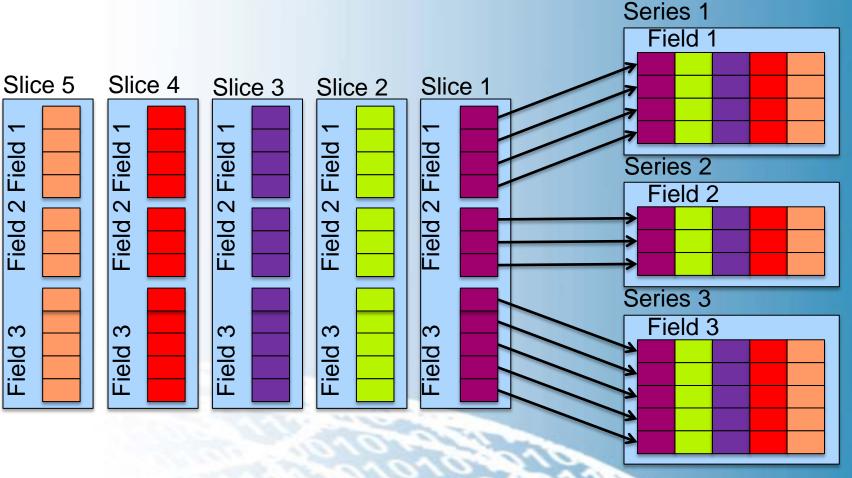


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### Time Slice to Time Series Conversion



- The previous method worked in serial using NCO
- This was one of the most expensive CMIP5 post-processing steps



## PyReshaper Details

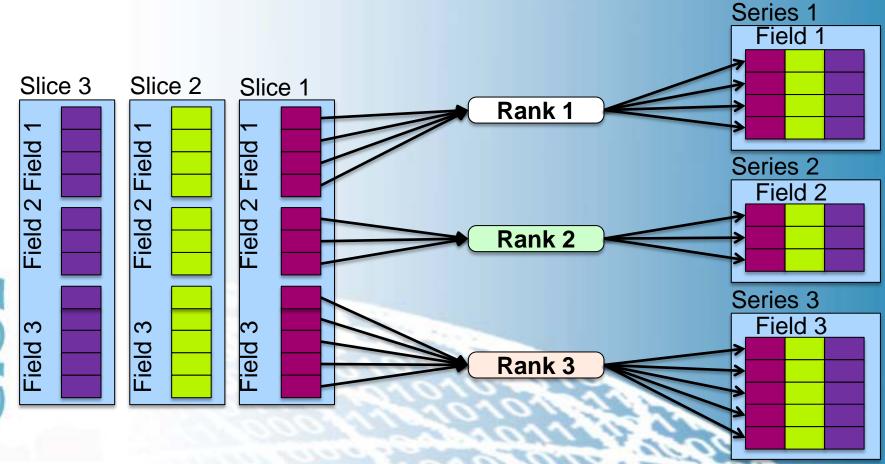
The PyReshaper is a light weight custom Python concatenation tool

- We chose Python for its flexibility and its fast development rate.
- For easier portability, we rely on only 3 packages
  - PyNIO (Python version of the NCL I/O utilities)
  - mpi4py (Package used for parallelization)
  - NumPy (Used for data storage)



### PyReshaper Parallelization Scheme

Each rank is responsible for writing one (or more) time-series variables to a file





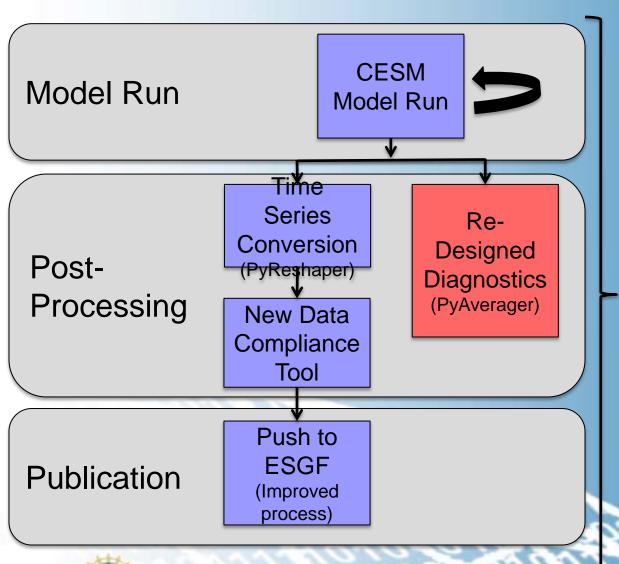
# Time-Slice to Time-Series Conversion Timing Statistics

Existing Method (NCO)	Time (per MIP per Year)	Average Throughput (per run)	Throughput per Calendar Day
1° atm, 1° ocn	225 minutes	1.85 MB/sec	0.15 TB
25km atm, 1° ocn	478 minutes	4.85 MB/sec	0.40 TB

New Method (PyReshaper)	Time (per MIP per Year)	Average Throughput (per run)	Throughput per Calendar Day
1° atm, 1° ocn	4 minutes	104 MB/sec	8 TB
25km atm, 1° ocn	8 minutes	290 MB/sec	23 TB

- Conversions were ran on Yellowstone using 4 nodes/4 cores (16 cores total)
- The PyReshaper increases performance by 50-60x and achieves better machine utilization





**Experiments** Update Their Status in Run **Database** 



Automated



### Original Diagnostic Packages

### Work focusing on redesigning

- Our 4 main component Diagnostic Packages
- ILAMB Package

#### The Original packages

- Contain top level control scripts
- Create climatology files with NCO tools
- Create hundreds of plots with NCL scripts
- Create web pages that allow users to browse through plots
- The ILAMB Package creates climos and plots with NCL

#### • Problems:

- Contain no parallelization
- They often break at high resolution
- They do not work with time series files



## Re-Design of Diagnostic Packages

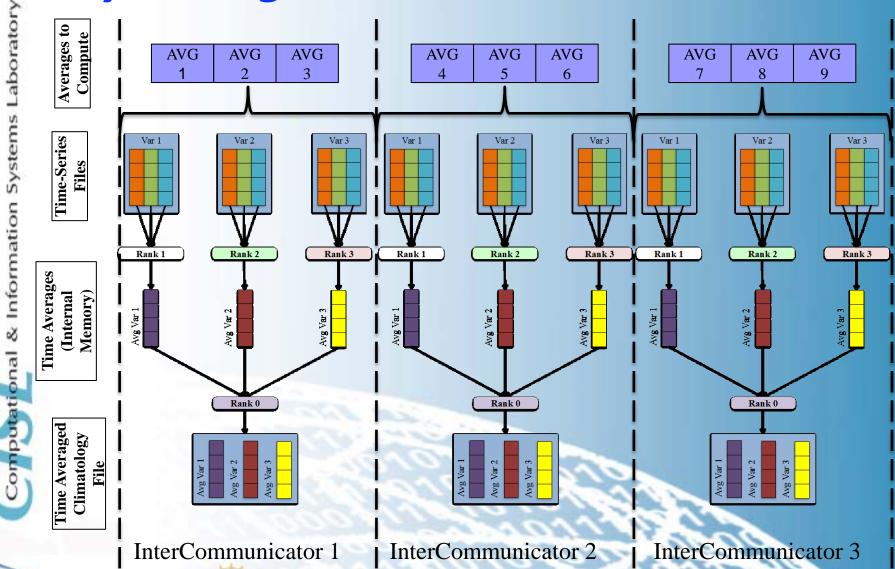
- Add in Parallelization
  - Instead of NCO, use the PyAverager to create the climatology files in parallel
  - Run the NCL plotting scripts in parallel
- Allow the packages to work with either time slice or time series files



## **PyAverager Details**

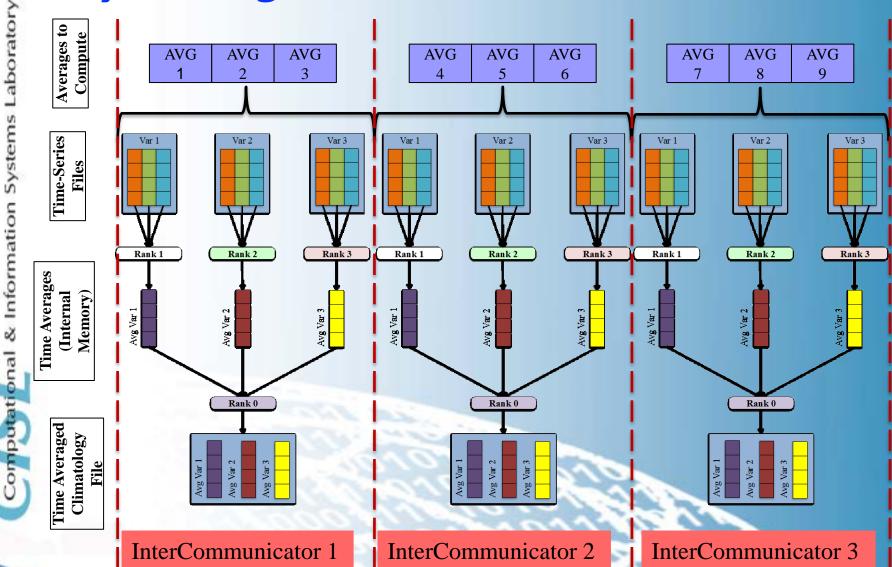
- A light weight custom Python averaging tool
  - Parallelizes over variables and averages
  - Depends on PyNIO, mpi4py, and NumPy
- Computes temporal averages
  - Seasonal, Yearly, Annual, Monthly (weighted optional)
- Can concatenate in parallel





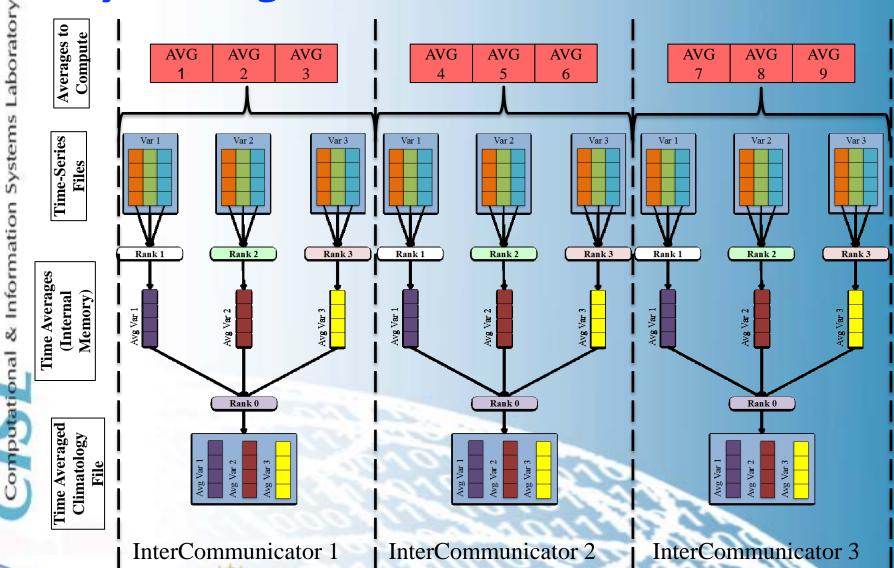






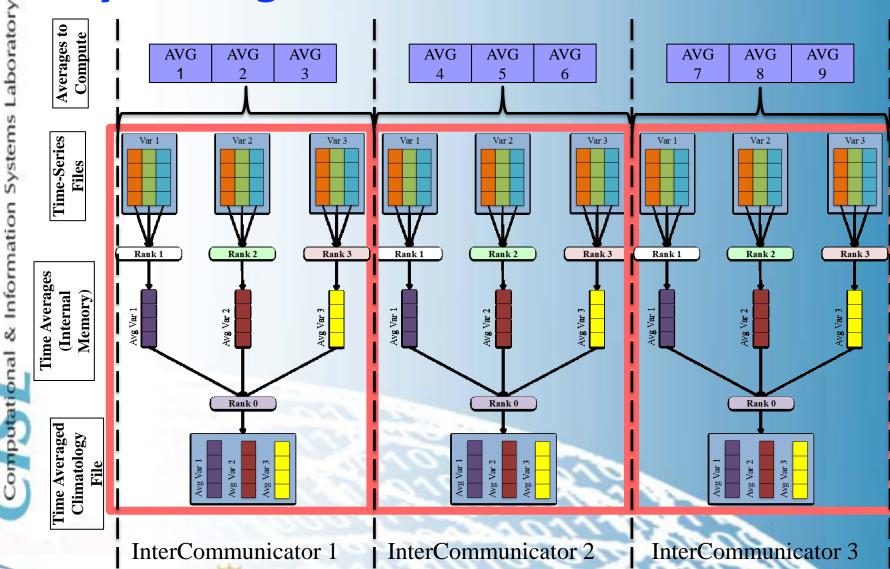






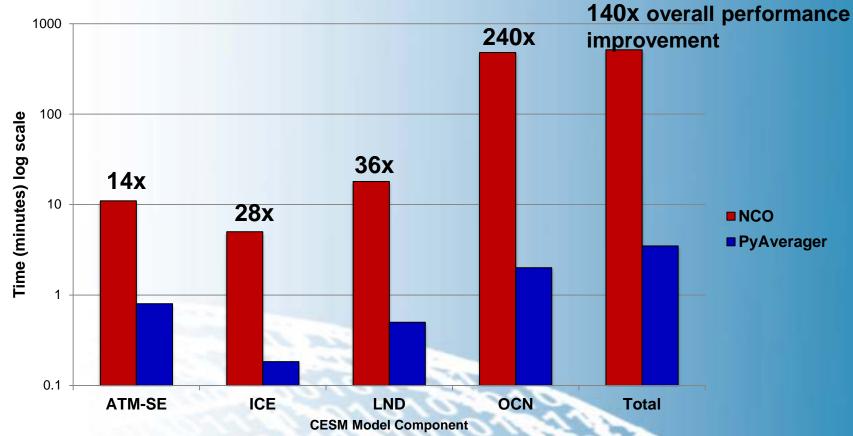








# PyAverager Performance Using CESM Data



Time to compute climatology files for 10 years of CESM monthly time slice files.

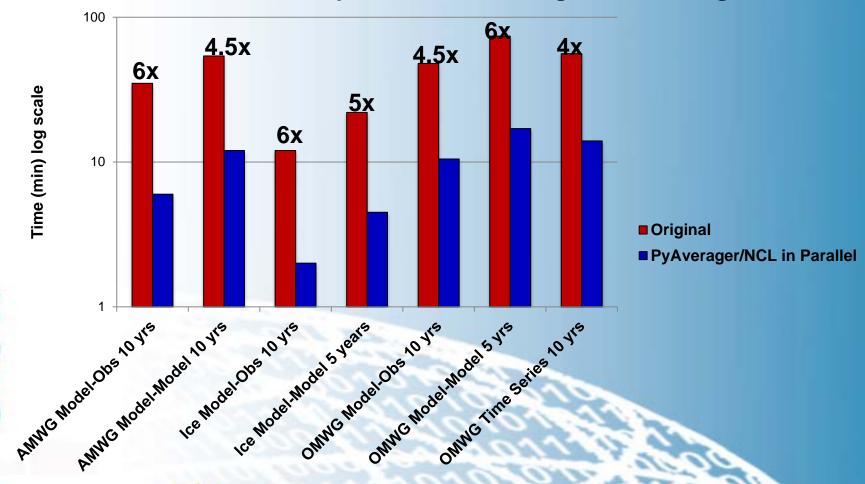
The PyAverager ran on 120 cores on yellowstone.





## Diagnostic Performance

**Performance Comparison Across Diagnostic Packages** 







## Making Our Tools General Enough to Handle Other Modeling Data

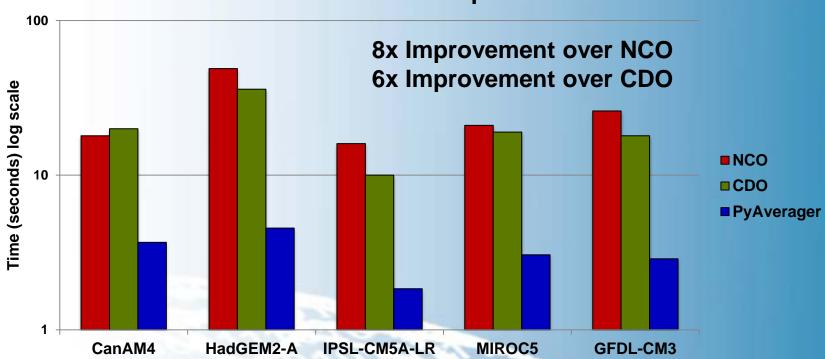
 Both the PyReshaper and the PyAverager can operate on non-CESM data

 Because we chose PyNIO for our I/O library, we can read in any data type that NCL can handle



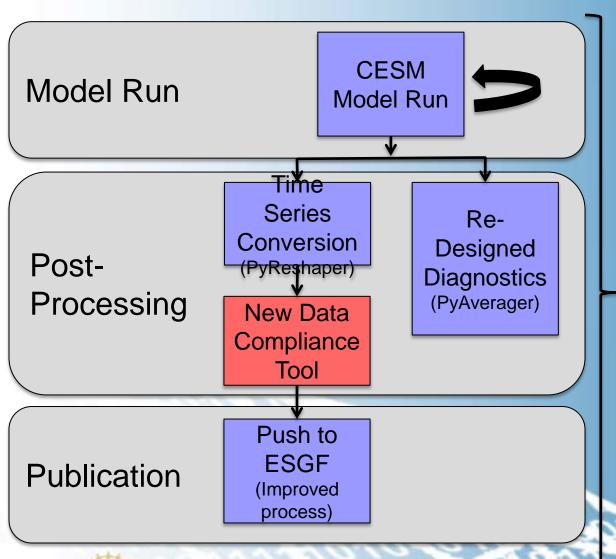
# PyAverager Performance Using Data From Other Models

#### **Performance Comparison**



CMIP5/AMIP monthly data. 5 seasonal and 12 monthly averages were computed over 29 years for 5 variables. The PyAverager was ran on 36 cores on yellowstone. The combined operation option was used for CDO.





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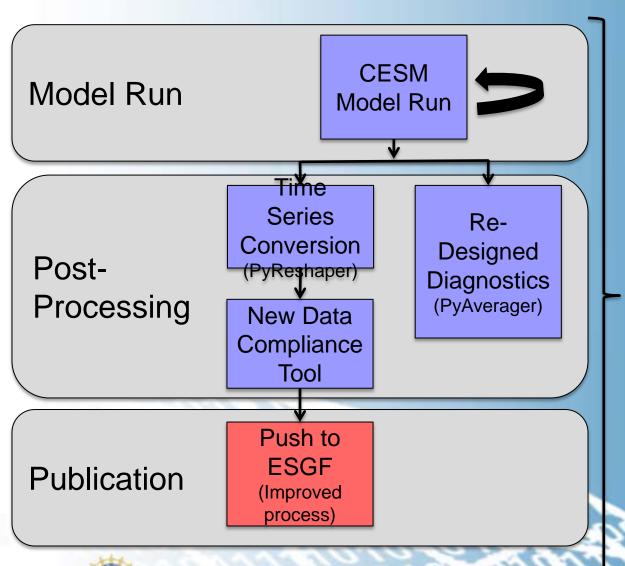




### New Data Compliance Tool

- Main Goals
  - Provide a simple user interface
  - Add data transformation and calculator abilities
  - Add parallelization to increase performance (this was another bottleneck in CMIP5)
- We are using similar techniques that were used by the PyReshaper and PyAverager





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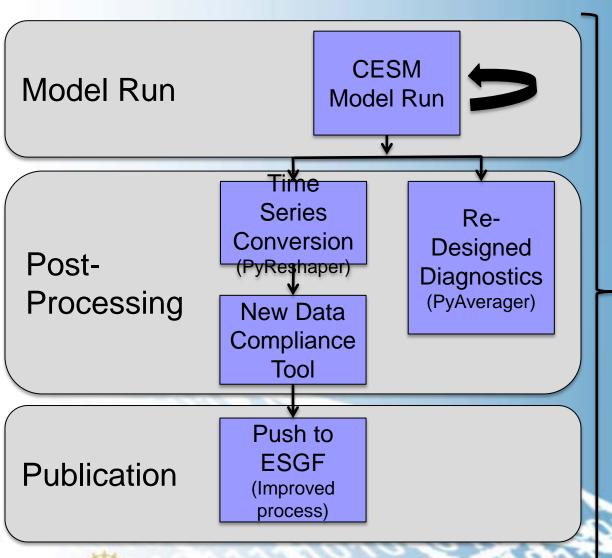
### **ESGF** Publication

- Move the data staging, directory structuring, and versioning responsibilities into the new compliance tool
- Streamline the submission process
- ESGF currently undergoing a major overhaul to address performance and reliability concerns
- New version of TDS (Unidata)
  - Better memory and resource management
- We will be setting up a test ESGF node to test new features and to access this portion of the workflow



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## New CESM/CMIP6 Workflow



**Experiments** Update Their Status in Run **Database** 



Automated



### Requirements for an Automated Workflow Management System

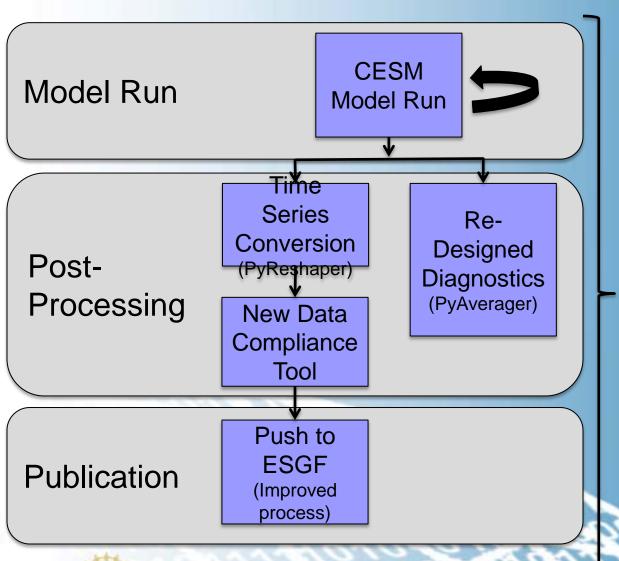
- We need a light weight and portable option
  - Needs to run on everything from laptops to shared supers per CESM user community requirements
- We evaluated Cylc and Rocoto
  - Both are very impressive, but lacked the portability that CESM required
- Since we have already developed an experiment database that we can extend, we only lacked a task scheduler



### Our New Workflow Management System

- The system evaluates user options and submits workflow to the machine's queue as dependency jobs
- The management system will automatically:
  - Run the PyReshaper after a select number of years has passed
  - Detect if and when to run diagnostics based on user selected date ranges
  - Run the data compliance tool after the PyReshaper successfully finishes





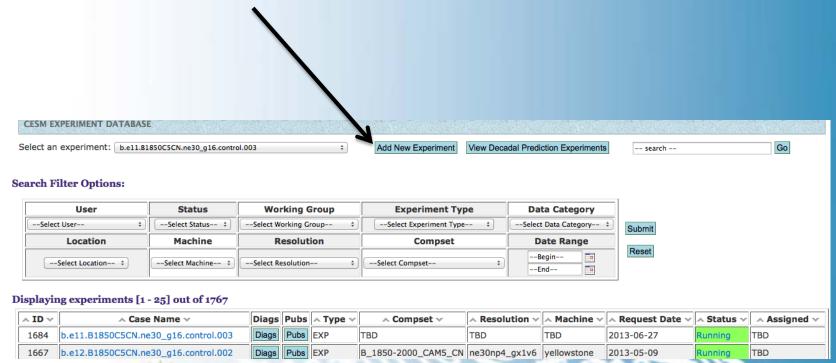
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### Current Run Database

### **Basic Catalog of Experiment Information**

Users manually enter experiments into the DB



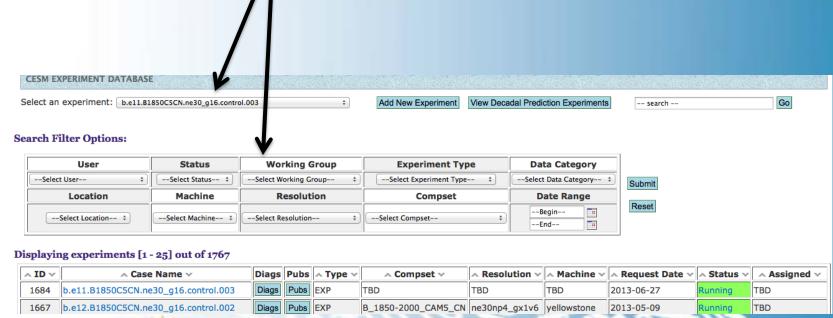




### Current Run Database

### **Basic Catalog of Experiment Information**

- Users manually enter experiments into the DB
- Contains basic search utilities and search filters



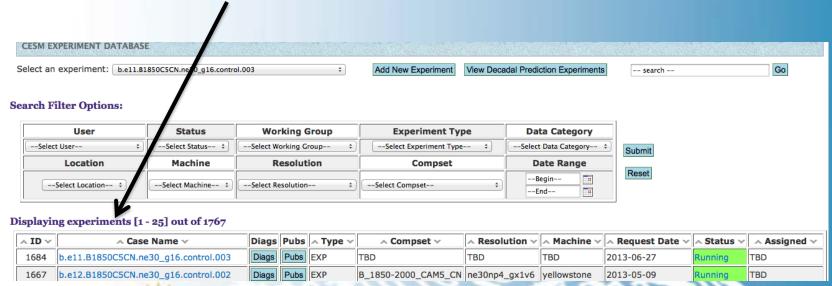




### Current Run Database

### **Basic Catalog of Experiment Information**

- Users manually enter experiments into the DB
- Contains basic search utilities and search filters
- Lists available experiments





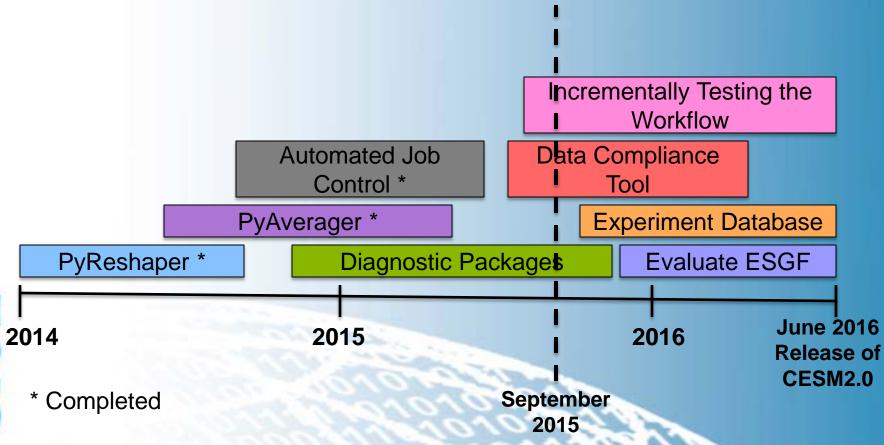


### Enhancements to the Run Database

- Automating the process of adding an experiment to the database
- Adding a separate section for CMIP6 experiments
- All experiments will update their run status to the database (simulation progress/color coded run status)
- CMIP6 timeline views
- Resource tracking (i.e. available disk space)
- Optional link to diagnostic web pages



# Project Timeline January 2014 – June 2016





### **Conclusions**

- CMIP5 stressed our workflow and showed us where we needed improvements
- We are introducing incremental changes and adding them into our current workflow and testing them before CMIP6 starts
- Our new tools provide significant performance improvements
  - PyReshaper: 50-60x speedup
  - PyAverager: 14-240x speedup



### **Conclusions Continued**

- Our new Python tools are general across all models, do not have many dependencies, and can be easily integrated into a workflow
- We continue to building additional tools that will improve our workflow



### Questions?

## Python Tool Availability Github

- https://github.com/NCAR-CISL-ASAP/PyReshaper
- https://github.com/NCAR-CISL-ASAP/pyAverager
- https://github.com/NCAR-CISL-ASAP/ASAPPyTools

#### PIP

- pip install PyReshaper
- pip install pyAverager
- pip install ASAPTools

#### **Contact Information**

mickelso.at.ucar.edu

https://www2.cisl.ucar.edu/tdd/asap/parallel-python-tools-post-processing-climate-data

