

Parallel Python Tools for Handling Big Climate Data

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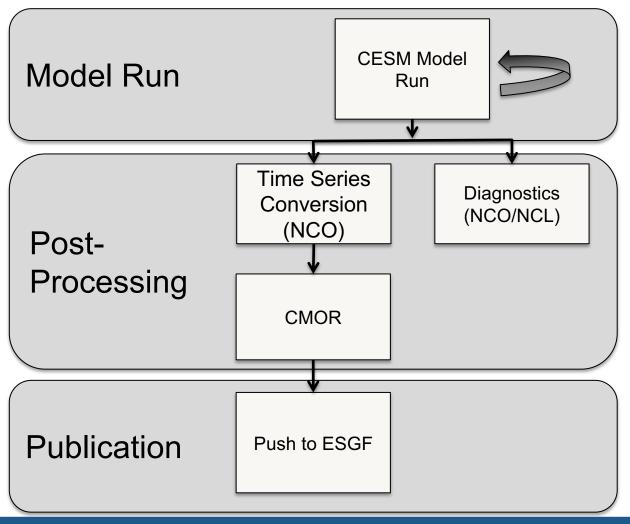


2018 Workshop on Developing Python Frameworks for Earth System Sciences



October 30, 2018

CESM's CMIP5 Workflow



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Lessons We Learned From CMIP5

CESM was the first model to complete their simulations, but the last to complete publication.

Why?

- All of the post-processing was serial and it took a long time to run
- Workflow was error prone and was time consuming to debug
- Too much human intervention was needed between post-processing steps and time was wasted
- There was only one person who knew the status of all of the experiments

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Motivating Factor CMIP5 vs CMIP6

CMIP5

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CMIP6

- 25 Experiments
- Timeline: 3 years
- Output size: 800TB
- Published size: 200TB

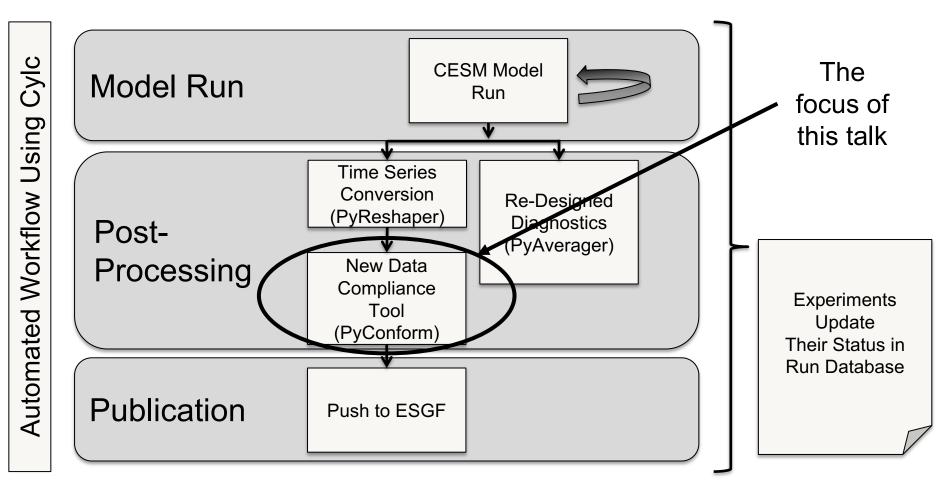
- **102 Experiments**
- Timeline: 1 year
- Output size: 8PB (estimate)
- Published size: 2PB (estimate)

http://www.bbc.com/earth/story/20170510-terrifying-20m-tall-rogue-waves-are-actually-real

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New CESM/CMIP6 Workflow

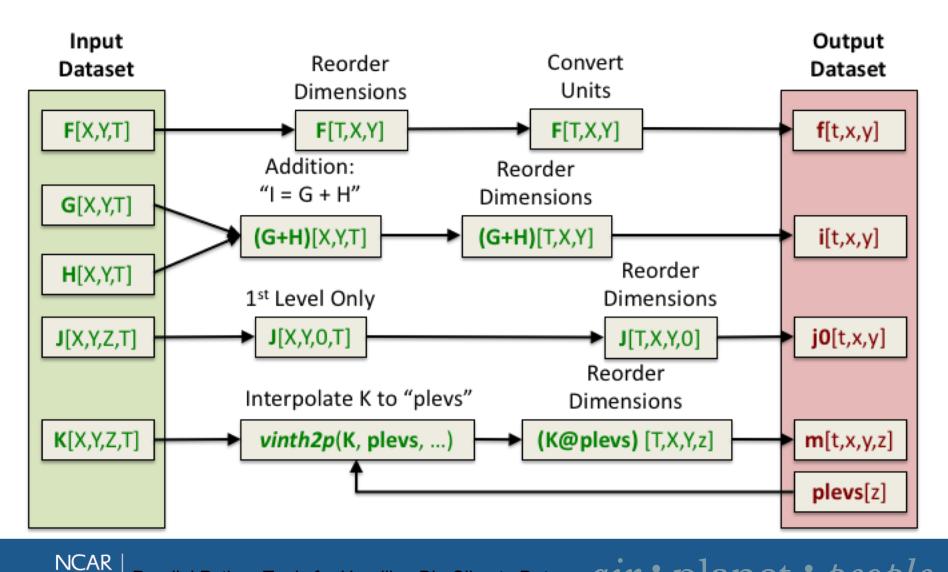


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Data Compliance

Taking model output and processing it into experiment compliant data



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Previous Version Used for CMIP5

- Used Fortran and CMOR to read in the raw output, do all conversions, and write out complaint files
- Serial, no parallelization

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PyConform: New Version Used for CMIP6

- Uses Python netCDF4, numpy, dreqPy, cf_units, pyNGL
- Parallelization done with MPI4Py
- A three step process

Faster (16x to 38x speedup over Fortran method) Flexible User Interface



First Step

Users need to create a text file with definitions that describe how to map model variables to requested variables

Examples:

cfcllglobal=fllvmr ch4=vinth2p(CH4, hyam, hybm, plev, PS, P0) mc=CMFMC+CMFMCDZM siage=siage

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Second Step

Then users run the iconform tool that matches the definitions to its variable information within the CMIP6 Data Request

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The Data Request lists variable requirements:

• Units

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- Dimensions
- Descriptions
- Positive Attribute on Vertical Dimensions
- And a lot more ...

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Sample Portion of a PyConform Input File

```
"ua": {
        "attributes": {
            " FillValue": "1e+20",
            "cell measures": "area: areacella",
            "cell methods": "time: mean",
            "comment": "\"Eastward\" indicates a
vector component which is positive when directed
eastward (negative westward). Wind is defined as a
two-dimensional (horizontal) air velocity vector,
with no vertical component. (Vertical motion in
the atmosphere has the standard name
upward air velocity.)",
            "description": "\"Eastward\" indicates
a vector component which is positive when directed
eastward (negative westward). Wind is defined as a
two-dimensional (horizontal) air velocity vector,
with no vertical component. (Vertical motion in
the atmosphere has the standard name
upward air velocity.)",
            "frequency": "mon",
```

```
"id": "ua",
"long_name": "Eastward Wind",
"mipTable": "Amon",
```

"out_name": "ua",

```
"prov": "Amon ((isd.003))",
```

"realm": "atmos",

"standard_name": "eastward_wind",

"time": "time",

"time_label": "time-mean",

"time_title": "Temporal mean",

"title": "Eastward Wind",

"type": "real",

"units": "m s-1",

"**variable_id":** "ua"

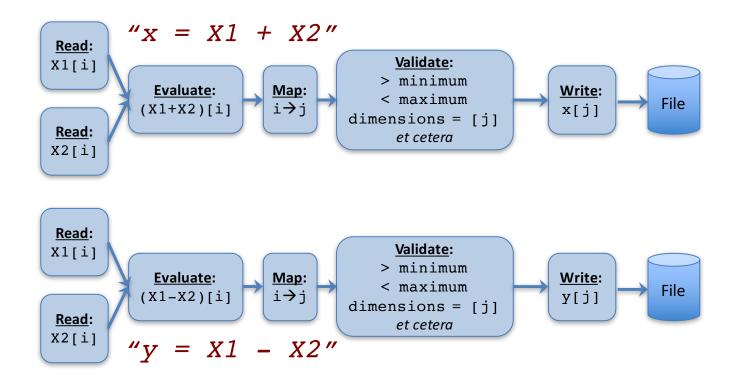
"datatype": "real",

},

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Third Step

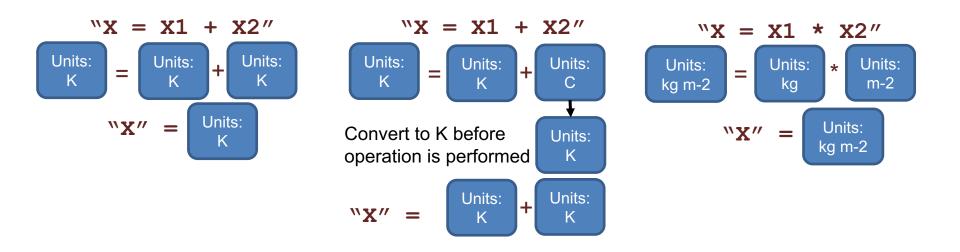
Then users run the xconform tool that generates requested variables based on the input specifications



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- Is a subclass of the maskedArray in NumPy
- Additional features that were needed above the masked array class:
 - Automatic Unit Conversion
 - Automatic Dimension Handling
 - Automatic Handling of the Positive Attribute

Extra features we needed to generate the data correctly: Automatic Unit Handling



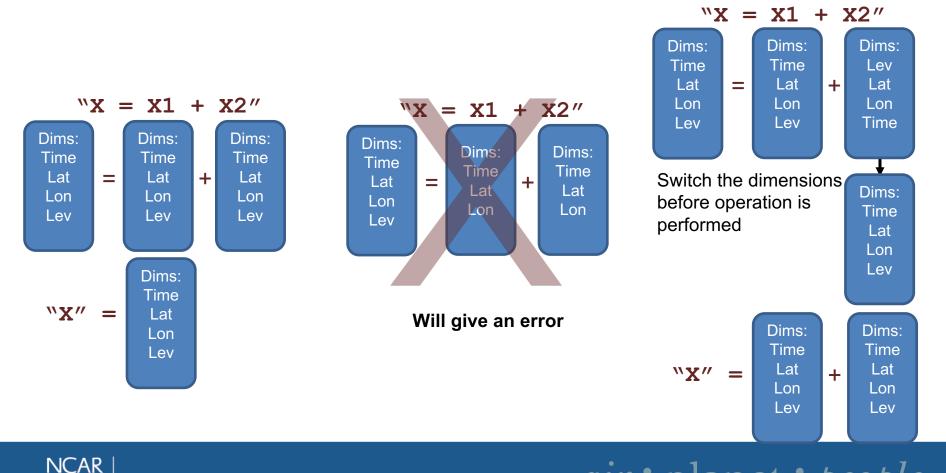
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* Must be cf compliant units

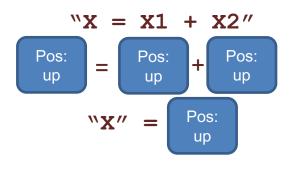
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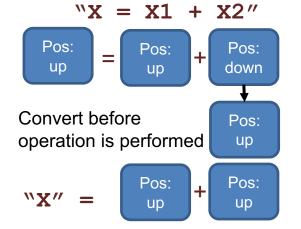
Extra features we needed to generate the data correctly: Automatic Dimension Handling



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Extra features we needed to generate the data correctly: Automatic Handling of the Positive Attribute (flipping the vertical dimension)







Switching to Xarray/Dask

- We are working on a new version that uses xarray
- While we no longer need the ability to handle dimension reordering, we still need functionality to handle the unit conversion and the flipping of the vertical dimension
- We will also need to evaluate the performance

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Moving Forward

- We are currently using PyConform in its current form for our CMIP6 output
- We are looking at a redesign of the internal data structures to use new capabilities that didn't exist when we started the project
- Performance and usability are key for this tool and we will move in those directions







Questions

Contact: mickelso .at. ucar.edu https://github.com/NCAR/PyConform



http://m.sweetclipart.com/halloween-pumpkin-pail-with-candy/