

BER/EESM/RGMA

New Tools for Analysis of Data on Unstructured Grids Robert Jacob¹, Paul Ullrich², Jill Zhang², Rajeev Jain¹, Hongyu Chen³, Tom Vo², Stephen Po-Chedley²

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- Some of the new standard tools SEATS is developing are described below.





- User opens two files to instantiate UXarray: a NetCDF file with output fields and second NetCDF file with the mesh description.
- UXarray can read mesh descriptions in several formats (SCRIP, ESMF, MPAS, Exodus) but internally stores it following UGRID. (UGRID has been added to CF to provide a standard for describing unstructured climate and weather grids in NetCDF).
- Supports operations on unstructured grids
- See poster #115 in this session for more info
- UXarray is developed in collaboration with NSF Project Raijin (NCAR)

• Climate models such as E3SM increasingly have components that use unstructured grids for their numerical methods. • Because many of the current analysis tools assume data is on structured grids, additional steps such as interpolation are required before analysis can be done. • One goal of the **SEATS** (Simplifying ESM Analysis Through Standards) project is to create new tools that can be used to analyze unstructured data directly.



in the file, not a full mesh description.

- Simple point-and-click interface
- Written in C++ with the WxWidgets GUI library

All tools available in e3sm-unified and individually from conda-forge



More ESMs are using unstructured meshes

- NCVIS is a new standalone GUI application for quickly viewing fields stored in CF-compliant NetCDF files.
- Inspired by *Ncview* (by David Pierce) but allows viewing of model data on unstructured grids.
- Only requires that latitude and longitude arrays be included



Calculate monthly departure s_monthly = ds.ts.groupby("time.month") # group by m s_monthly_clim = ts_monthly.mean(dim="time") # ca ts_anom = ts_monthly - ts_monthly_clim # d oslat = np.cos(np.deg2rad(ds.lat)) ts_anom_weighted = ts_anom.weighted(coslat) ts_anom_global = ts_anom_weighted.mean(dim="lat").mean(dim="lon") th_len = ts_anom_global.time.dt.days_in_month ith_len_by_year = month_len.groupby("time.year") gts = month_len_by_year / month_len_by_year.sum() emp_sum = (ts_anom_global * wgts).resample(time="AS").sum(dim="time") enominator_sum = (wgts).resample(*time*="AS").sum(*dim*="time") ts_anom_global_ann = temp_sum / denominator_sum

Above: Code comparison for calculating monthly departures, global mean time series, and annual averages with plain Xarray (upper

left) and with xCDAT (upper right). Plots showing the results of these calculations are shown on the right. **Right**: A) Map showing the monthly surface temperature departures departures (September 1850) and B) the corresponding monthly (gray) and annual mean (black) global mean surface temperature time series. Data from E3SM v2 historical simulation from CMIP archive.

- xCDAT is a successor to CDAT and is designed to perform routine geospatial analysis operations in python.
- xGCM, cf_xarray)
- and regridding.

- **PCMDI** projects

• Can operate on UXarray objects. Further integration TBD. • See talk by Tom Vo in Metrics session, 2:45pm, Thursday. • xCDAT is developed in collaboration with the E3SM and

• Key features include: spatial averaging, temporal averaging,

• xCDAT extends Xarray, is interoperable with dask (parallelism) and matplotlib (plotting) and leverages other packages for key functionality/interoperability (e.g., xESMF,

▽ 1.00

0.50

-0.75







4+ ds_anom_global = ds_anom.spatial.average("ts")

17+ ds_anom_global_ann = ds_anom_global.temporal.group_average("ts", *freq*="year"

A. Surface Temperature Anomalies (September 1850)

-4 -3 -2 -1 0 1 2 3 4

B. Global Mean Surface Temperature Anomalies







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