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What is xarray?

For Python / Numpy users

xarray handles N-dimensional arrays with labels (dimension names & coordinates) and metadata.

For Python / Pandas users

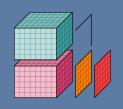
 xarray is a powerful, pandas-like toolkit for analytics on multi-dimensional arrays.

For scientists familiar with the netCDF format

xarray implements the netCDF data model with a high level Python API.

For scientists working with big datasets

 xarray (with dask) supports efficient, out-of-core computing for datasets that don't fit in memory.



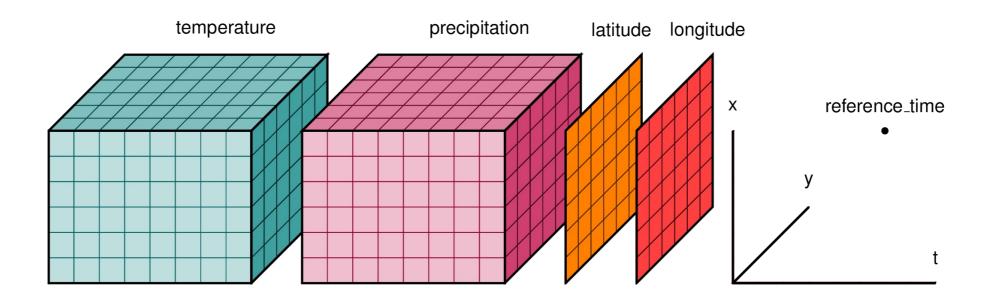
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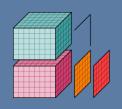
What is xarray?

If you are dealing with data that...

- is multi-dimensional;
- is labelled;
- has (lots of) metadata;
- is sometimes (very) large;



...then you may find xarray very useful!



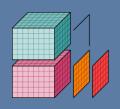
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xarray

- Open source
- Very good integration with other Python libraries for scientific computing (SciPy / PyData Stack)
- Documentation: http://xarray.pydata.org
- Repository: https://github.com/pydata/xarray
- 60 contributors (still growing)
- Latest release: v0.9.2 (02.04.2017)
- Umbrellas (no funding): Python for Data & NumFOCUS





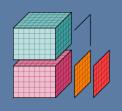


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numpy.array

Not well supported by numpy:

• array dimensions and indexes often have a meaning, e.g., latitude / longitude and their coordinates.

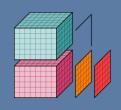


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xarray.DataArray

```
>>> import xarray as xr
>>> da = xr.DataArray(a, dims=['latitude', 'longitude'],
                      coords={'longitude':[11, 12, 13], 'latitude':[1, 2]})
>>> da
<xarray.DataArray (latitude: 2, longitude: 3)>
array([[1, 3, 9],
      [2, 8, 4]])
Coordinates:
  * longitude
              (longitude) int64 11 12 13
  * latitude
               (latitude) int64 1 2
>>> da.sel(longitude=13, latitude=2) # easier to work with coordinate values!
<xarray.DataArray ()>
array(4)
Coordinates:
   longitude int64 13
   latitude
                int64 2
>>> da.mean(dim='latitude') # easier to remember dimension names!
<xarray.DataArray (longitude: 3)>
array([ 1.5, 5.5, 6.5])
Coordinates:
               (longitude) int64 11 12 13
  * longitude
```



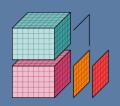
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xarray.Dataset

A collection of xarray. DataArray, a netCDF file...

```
>>> ds = xr.open_dataset('ERA-Int-MonthlyAvg-4D-TUVWZ.nc')
>>> ds
<xarray.Dataset>
               (latitude: 241, level: 15, longitude: 480, month: 12)
Dimensions:
Coordinates:
  * latitude
               (latitude) float32 90.0 89.25 88.5 87.75 87.0 ...
  * level
               (level) int32 50 100 150 200 300 400 500 600 ...
  * longitude
               (longitude) float32 -180.0 -179.25 -178.5 ...
  * month
               (month) int64 1 2 3 4 5 6 7 8 9 10 11 12
Data variables:
               (month, level, latitude, longitude) float64 10.38 ...
    U
               (month, level, latitude, longitude) float64 5.594 ...
    V
               (month, level, latitude, longitude) float64 -0.0003052 ...
    W
               (month, level, latitude, longitude) float64 1.888e+05 ...
    Z
               (month, level, latitude, longitude) float64 201.1 ...
Attributes:
    Conventions: CF-1.0
    Info: Monthly ERA-Interim data.
```



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Powerful analytics

Advanced selection

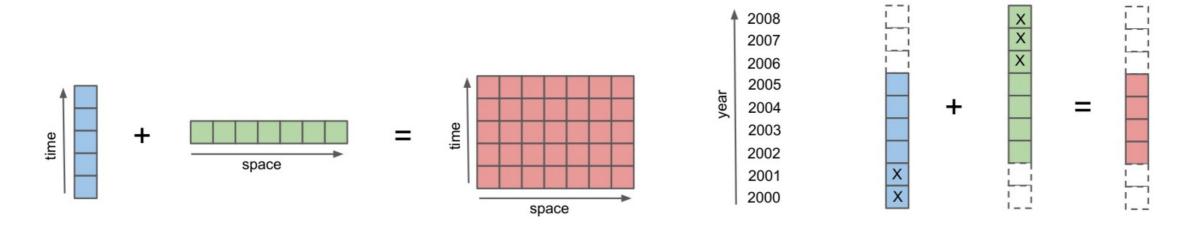
```
ds.sel(latitude=47.26, method='nearest') # no need to provide exact coordinate values
```

Aggregation

```
ds.mean(dim=['month', 'longitude'])
```

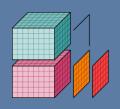
Arithmetics

ds1 + ds2 # supports automatic broadcasting and alignment!



EGU Vienna 04-2017 (PICO)

summary data structures data analytics



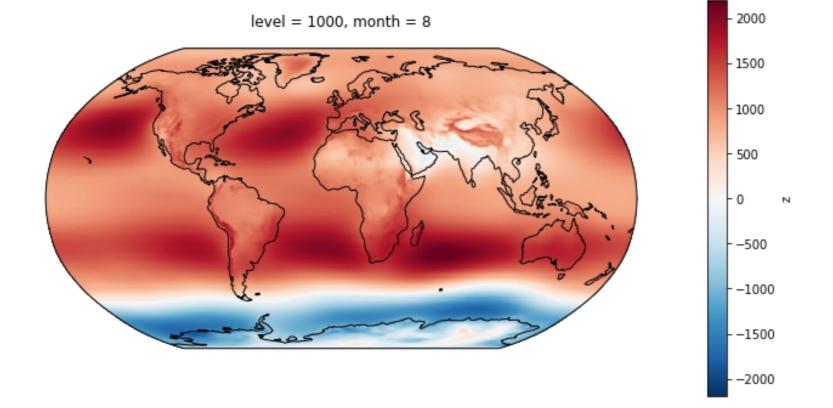
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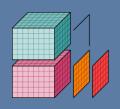
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Powerful analytics (2)

Plotting

```
import cartopy.crs as ccrs
ax = plt.axes(projection=ccrs.Robinson())
ds.z.sel(level=1000, month=8).plot(ax=ax, transform=ccrs.PlateCarree());
ax.coastlines();
```





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Powerful analytics (3)

• Out-of-core computing

• ... and more!