

xGCM: Staggered grids, topologies, and grid ufuncs in python

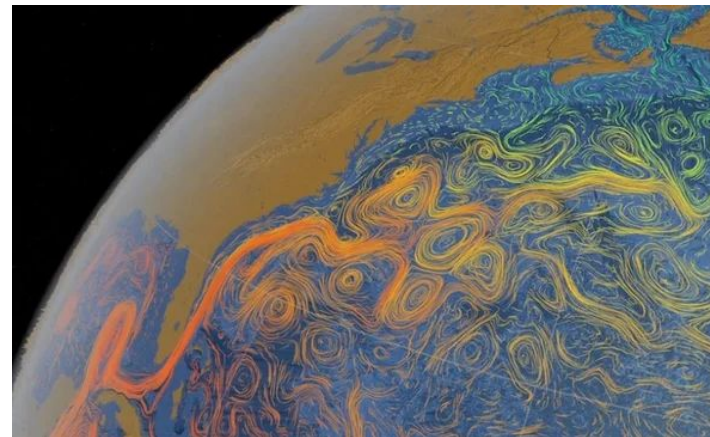
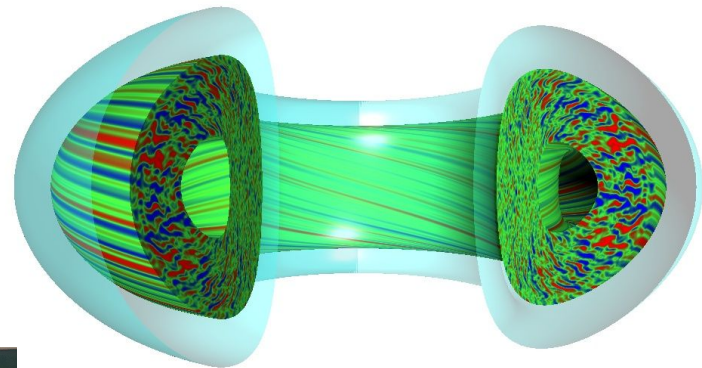
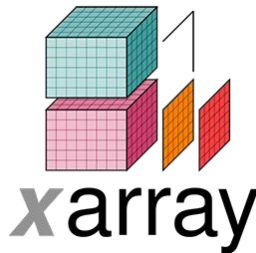


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Who am I?

- Oceanographer (ex-plasma physicist)
- Xarray core dev & Pangeo user
- Both these fields have variety of big turbulence models producing gridded data



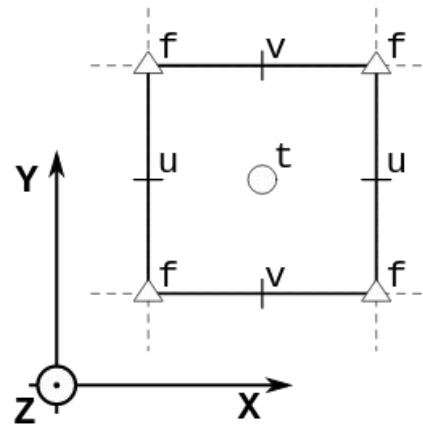
PANGEO

Talk overview

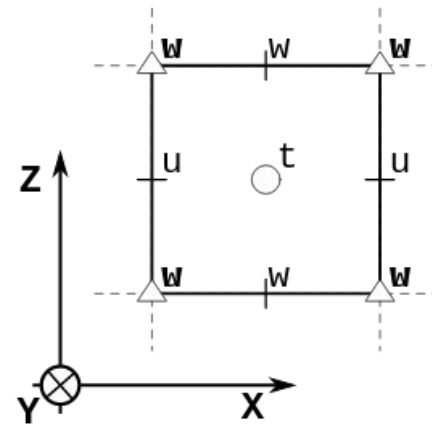
- 3 problems with model grids
 - How xGCM concepts help
- Grid ufuncs for arbitrary operations
- Will it scale?

Grid problem #1: Staggered grids

- Fluid variables live on “**Arakawa Grids**”
- Variables’ positions are **offset**
- **Finite-volume calculations** must account for this to get correct results



C-grid — horizontal view

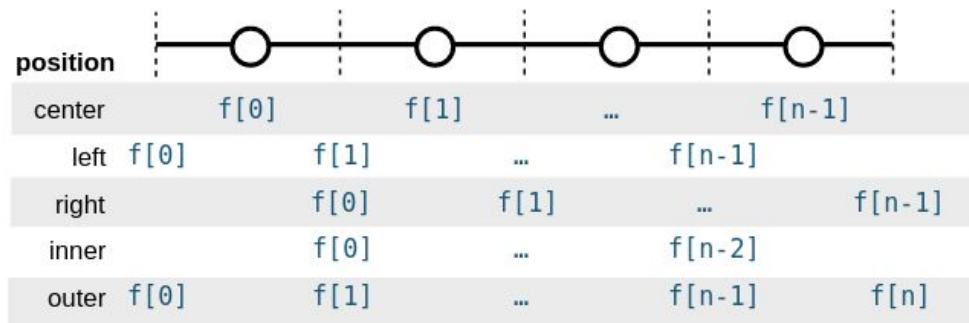


C-grid — vertical view



Staggered grids in xGCM

- xGCM handles **staggered variables**
- Extends xarray's data model** with **Axes** and **Grid** objects
- Variables may live on **different positions** along **xgcm.Axes**
- Axes stored in **Grid** object



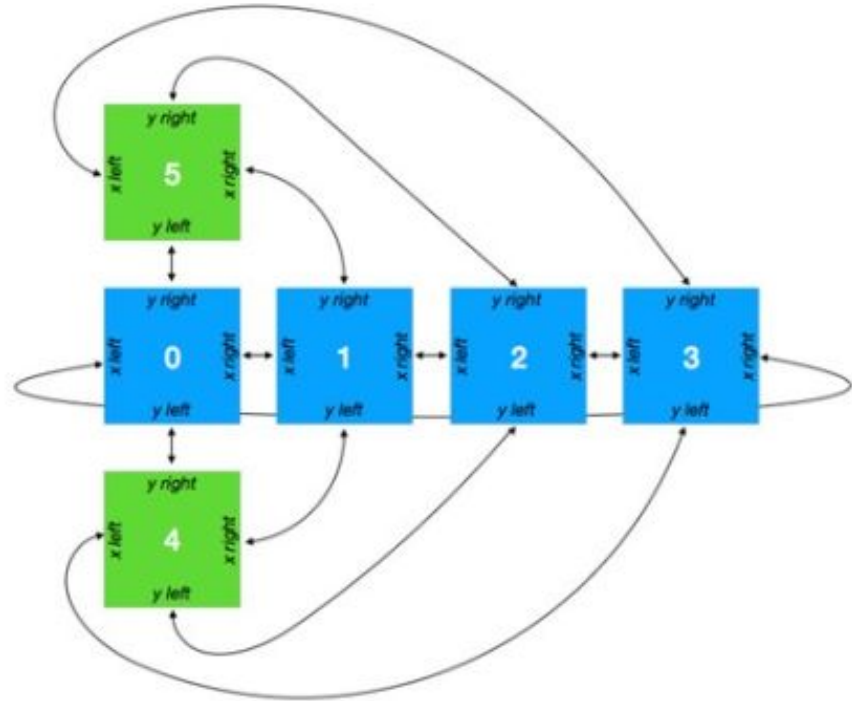
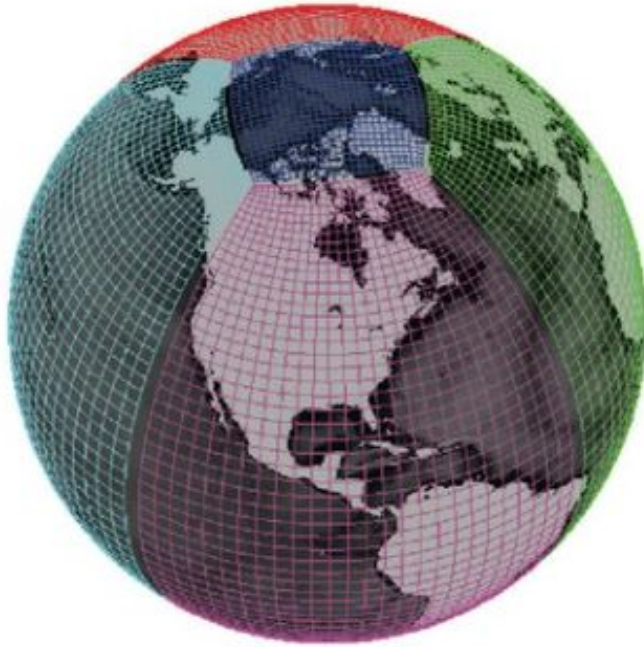
```
In [5]: from xgcm import Grid

In [6]: grid = Grid(ds, coords={"X": {"center": "x_c", "left": "x_g"}})

In [7]: grid
Out[7]:
<xgcm.Grid>
X Axis (periodic, boundary=None):
* center  x_c --> left
* left    x_g --> center
```

github.com/xgcm/xgcm

Grid problem #2: Topologies



Topologies in xGCM

```
In [15]: face_connections = {'face':
.....:     {0: {'X': ((3, 'X', False), (1, 'X', False)),
.....:          'Y': ((4, 'Y', False), (5, 'Y', False))},
.....:     1: {'X': ((0, 'X', False), (2, 'X', False)),
.....:          'Y': ((4, 'X', False), (5, 'X', True))},
.....:     2: {'X': ((1, 'X', False), (3, 'X', False)),
.....:          'Y': ((4, 'Y', True), (5, 'Y', True))},
.....:     3: {'X': ((2, 'X', False), (0, 'X', False)),
.....:          'Y': ((4, 'X', True), (5, 'X', False))},
.....:     4: {'X': ((3, 'Y', True), (1, 'Y', False)),
.....:          'Y': ((2, 'Y', True), (0, 'Y', False))},
.....:     5: {'X': ((3, 'Y', False), (1, 'Y', True)),
.....:          'Y': ((0, 'Y', False), (2, 'Y', True))}}
.....:
```

```
In [16]: grid = xgcm.Grid(ds, face_connections=face_connections)
```

```
In [17]: grid
```

```
Out[17]:
```

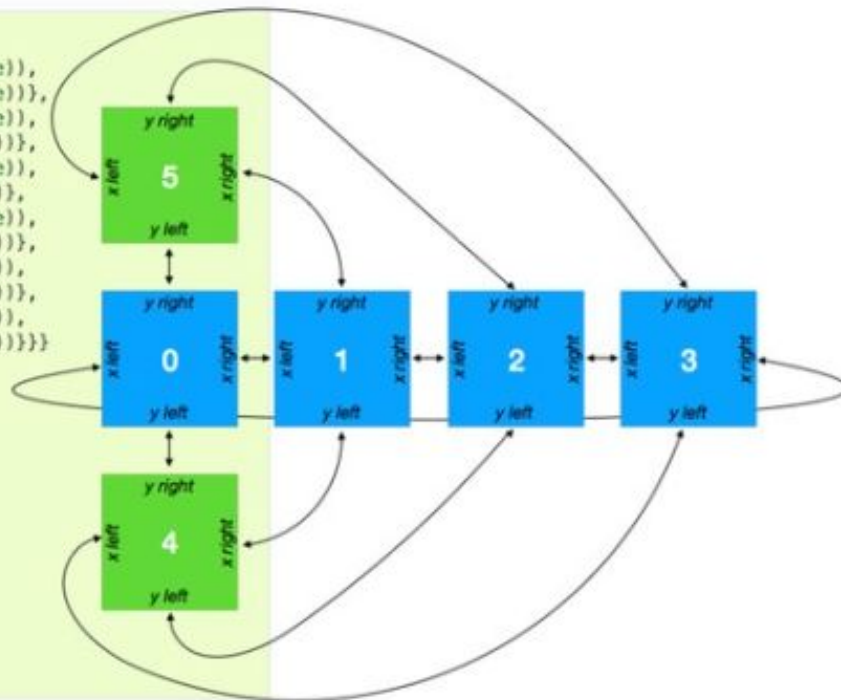
```
<xgcm.Grid>
```

```
Y Axis (periodic):
```

```
* center  y --> left
* left    yl --> center
```

```
X Axis (periodic):
```

```
* center  x --> left
* left    xl --> center
```



Grid problem #3: Vertical coordinates

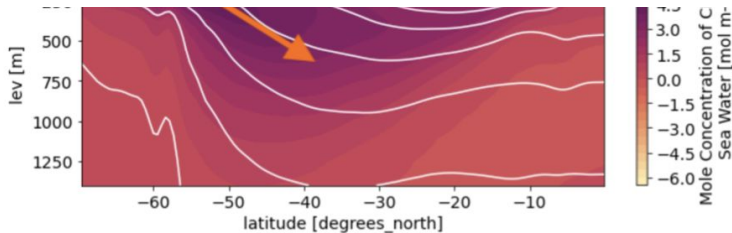
```
so_sigma = grid.transform(
    ds.so,
    'Z',
    target_values,
```

Published in **pangeo** · Pinned

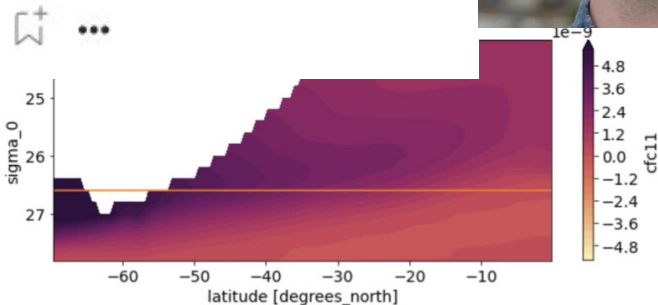
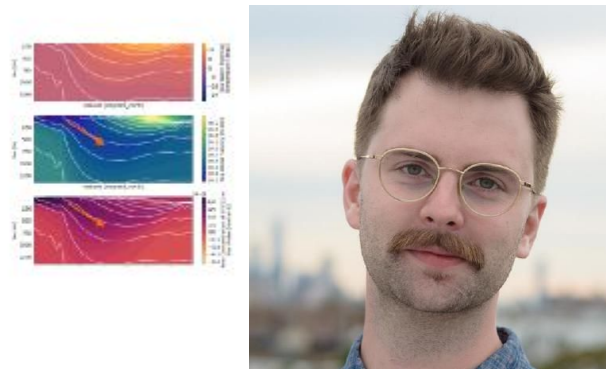
Vertical coordinate transformation with Pangeo — Have some pancakes and let Xgcm do the work

The ocean, a vast, mysterious, and beautiful place, but did it ever make you think of pancakes? Well if you are an oceanographer, it makes a l...

Xgcm 9 min read



Latitude-Depth sections along 130W in the Southern Ocean. From top to bottom: Potential temperature, Salinity, CFC11 concentrations



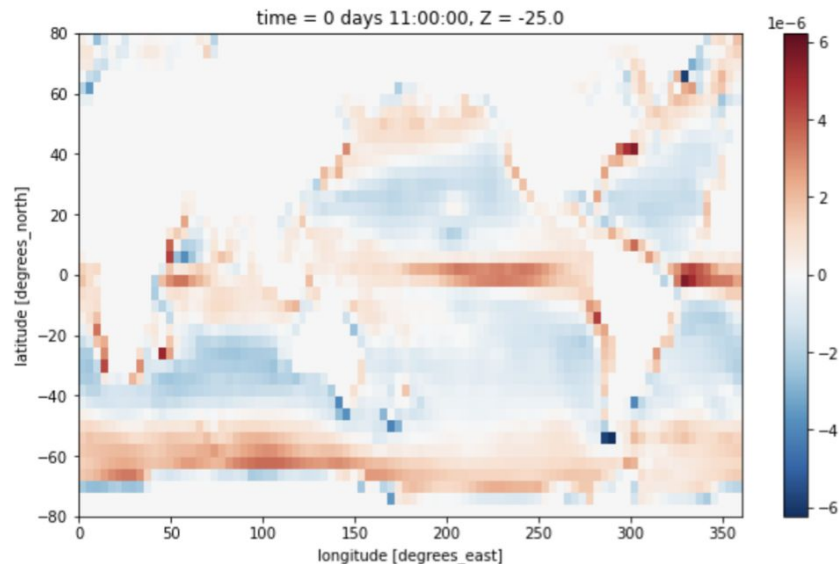
Latitude-Density sections along 130W in the Southern Ocean. Top: salinity; Bottom: CFC11 concentrations

Features: calculus operations

- Basic **calculus operations**
.diff, .interp etc.
- Uses **correct numerical scheme** for grid position!
- Consumes and produces **xarray objects**

```
u_transport = ds.UVEL * ds.dyG * ds.hFacW * ds.drF  
v_transport = ds.VVEL * ds.dxG * ds.hFacS * ds.drF
```

```
div_uv = (  
    grid.diff(u_transport, 'X') + grid.diff(v_transport, 'Y')  
    ) / ds.rA
```



New idea: “grid ufuncs”

- Wrap numpy ufuncs to be grid-aware
- Positions specified through “signature”
 - “(X:left) -> (X:center)”
- Signature is property of computational function
 - i.e. language-agnostic idea

```
from xgcm import as_grid_ufunc

@as_grid_ufunc(signature="(ax1:center)->(ax1:left)")
def diff_center_to_left(a):
    return a - np.roll(a, -1, axis=-1)
```

<code>interp_left_to_center(a)</code>	<code>"(X:left)->(X:center)"</code>	Forward interpolation
<code>diff_center_to_center(a)</code>	<code>"(X:center)->(X:center)"</code>	Second order central difference
<code>mean_depth(w)</code>	<code>"(depth:center)->()"</code>	Reduction

@as_grid_ufunc decorator

- Allows custom ufuncs
 - User-specific algorithms (e.g. from climate model)
 - Can auto-dispatch to correct ufunc for data
- Can specify grid positions via annotated type hints
- Could chain with other decorators, e.g. `numba.jit`

```
from typing import Annotated
from xgcm import as_grid_ufunc

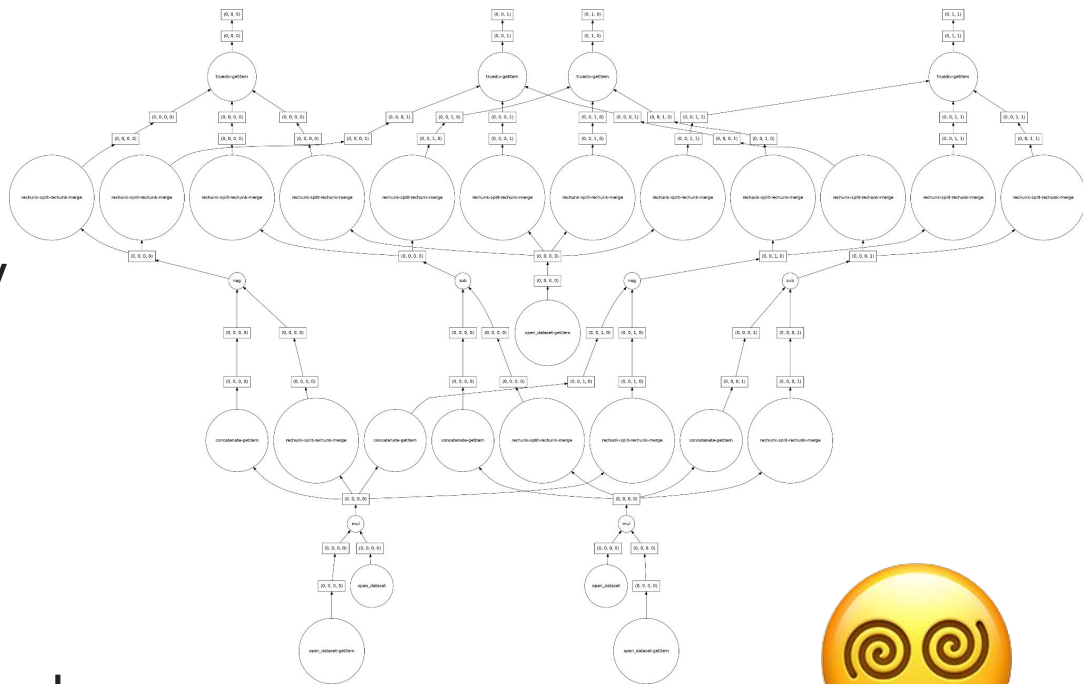
@as_grid_ufunc(
    boundary_width={"X": (0, 1), "Y": (0, 1)},
)
def divergence(
    u: Annotated[np.ndarray, "(X:left,Y:center)"],
    v: Annotated[np.ndarray, "(X:center,Y:left)"],
) -> Annotated[np.ndarray, "(X:center,Y:center)"]:

    du_dx = u[..., 1:, :] - u[..., :-1, :]
    dv_dy = v[..., 1:] - v[..., :-1]

    return du_dx + dv_dy
```

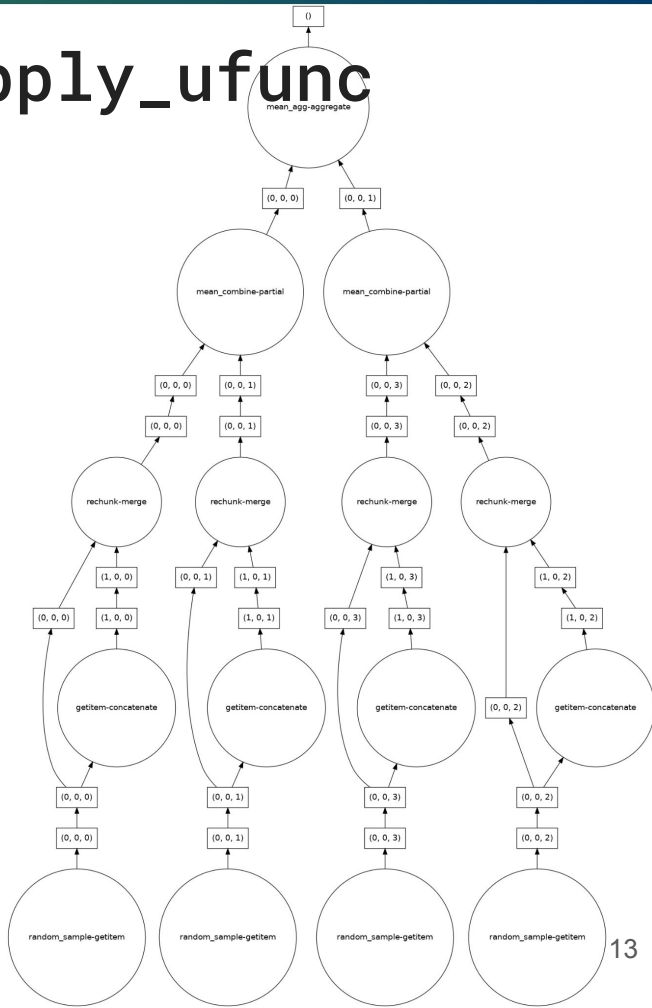
Will it scale? Originally no...

- xGCM's **finite-volume functions**, e.g. diff, interp
- **Require padding** to apply boundary conditions
- Previously used **custom reduction code**
- Chaining diff, interp etc. led to **explosion of dask tasks**



Dask-optimised xGCM via `xarray.apply_ufunc`

- Apply all grid ufuncs through `xarray.apply_ufunc`
 - Common code path for all functions
- Only pad once
 - Avoids task explosion
- Creates minimal dask graph
- Ex. reduction: Almost blockwise (+ a rechunk-merge operation after padding)

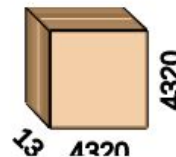


Now does it scale?

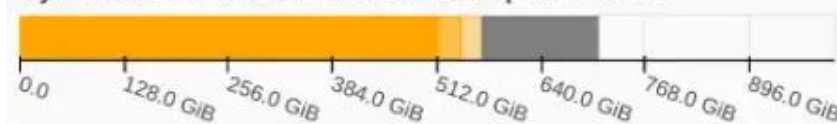
- Tried it on huge **global ocean simulation** dataset
- Single variable is **8.76 TB**
 - 117390 Zarr chunks
- But saw dreaded orange bars!
- Nice task graph, but dask **using way too much RAM!**

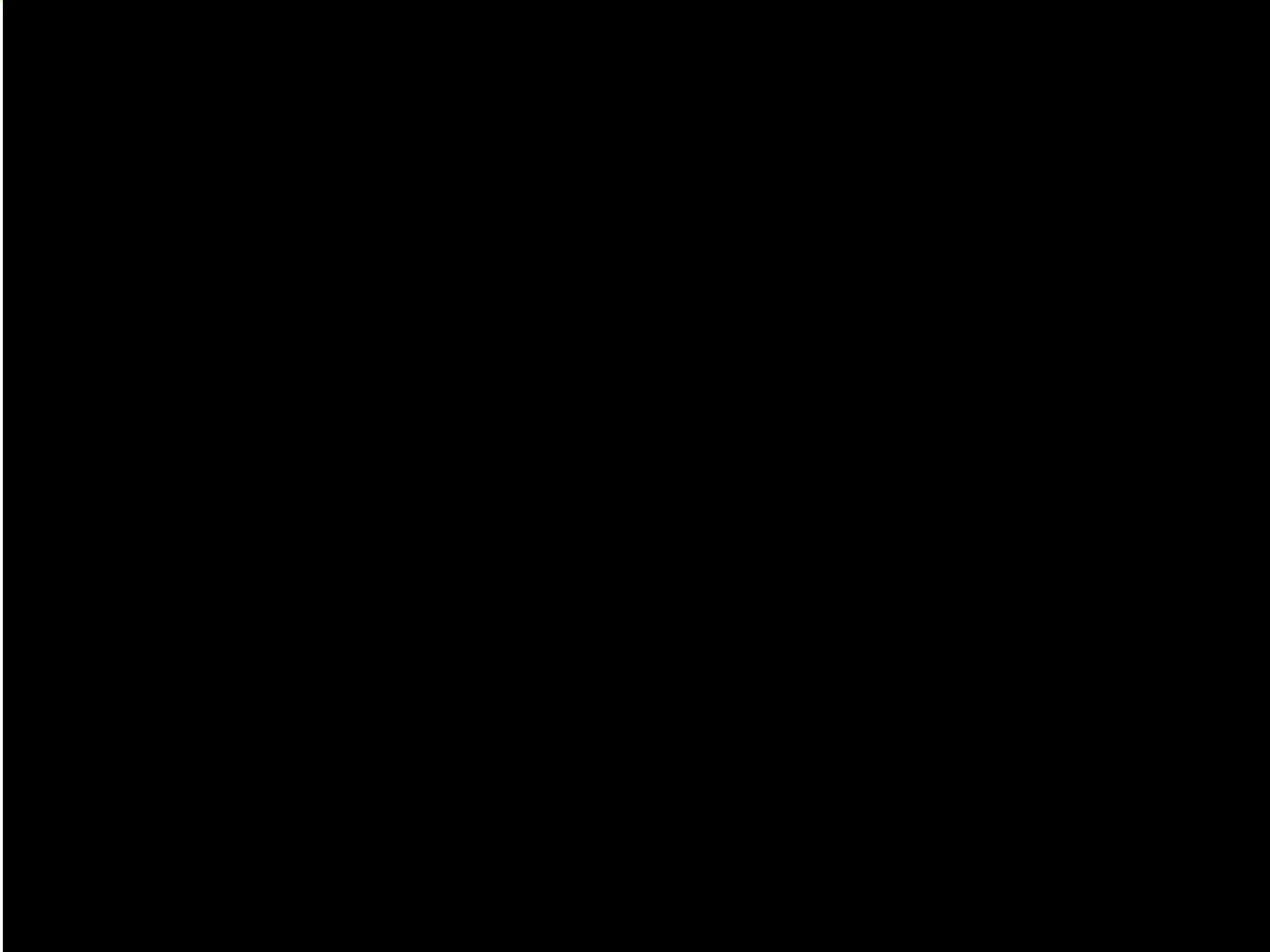
▼ Data variables:

U	(time, face, j, i_g)	float32	dask.array<chunksize=(1, 1, 4320, 4320)
	Array	Chunk	
Bytes	7.97 TiB	71.19 MiB	
Shape	(9030, 13, 4320, 4320)	(1, 1, 4320, 4320)	
Count	117390 Tasks	117390 Chunks	
Type	float32	numpy.ndarray	



Bytes stored: 567.75 GiB + 141.76 GiB spilled to disk





Dask needed fixing!

- Exposed a bug in `dask.distributed`
- Gabe Joseph of Coiled fixed it!
- See pangeo blog post

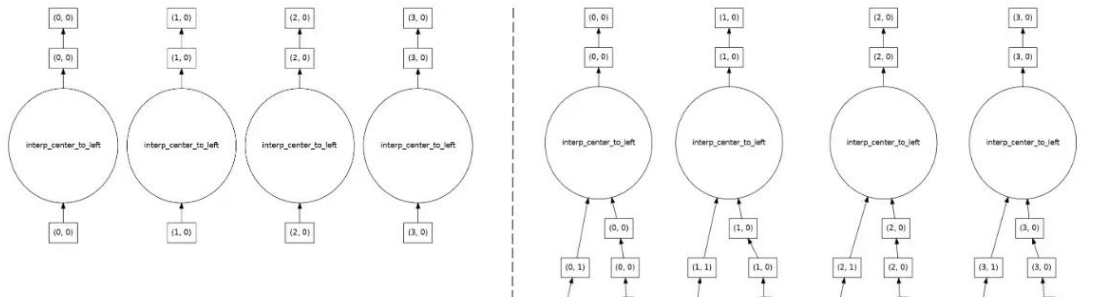
Dask.distributed and Pangeo: Better performance for everyone thanks to science / software...

Improvements to dask's performance at scale are a success story for pangeo too



Tom Nicholas
Jan 4 · 10 min read

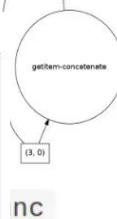
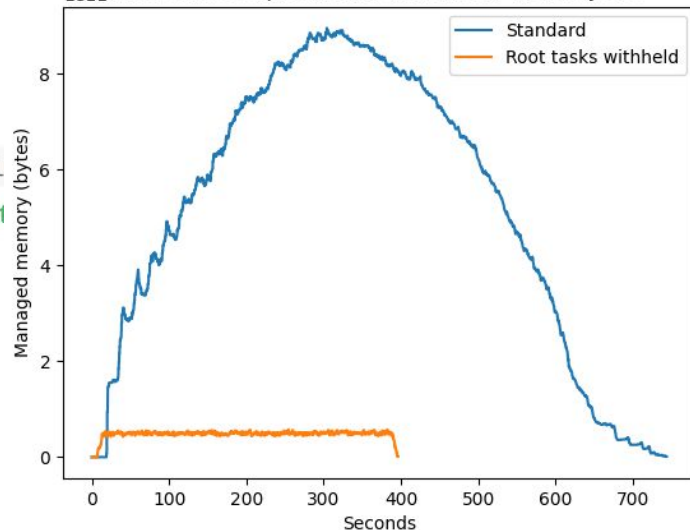
Two embarrassingly-parallel workloads:



`xarray.apply_`
Streams smooth



1e11 Root task overproduction and cluster memory use



Summary

- GCM grids are a pain
- But xGCM can help!
- Grid ufuncs are a cool idea
- We improved dask along the way!



github.com/xgcm/xgcm

**P.S. I am looking for my
next big project 😁**