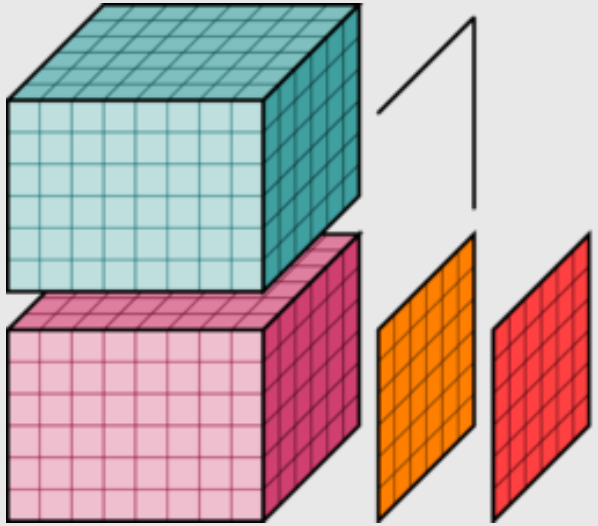
Unlocking the power of



xarray

&

dask

Summary – Aim is to write faster/better code

- Why should you listen to me?
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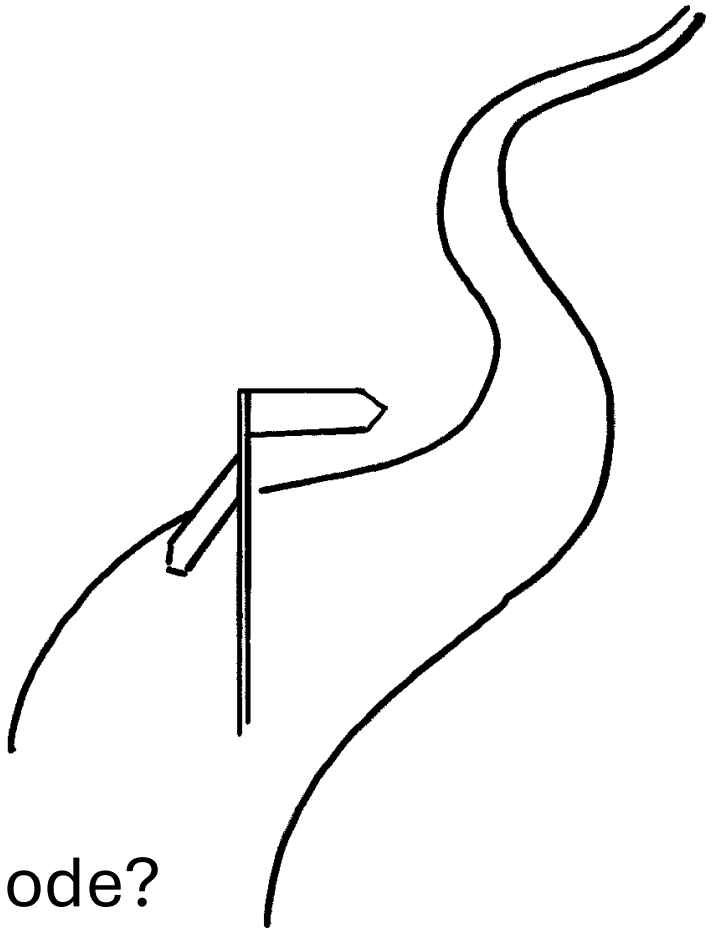
- How can xarray/dask help?
  - **“Lazy”** computing
  - Thinking about your **underlying** code
  - The voodoo that is **chunking**
- But how do I **translate** my function into xarray code?



3 key moments in my journey

- Young ICT explorers (grade 6)

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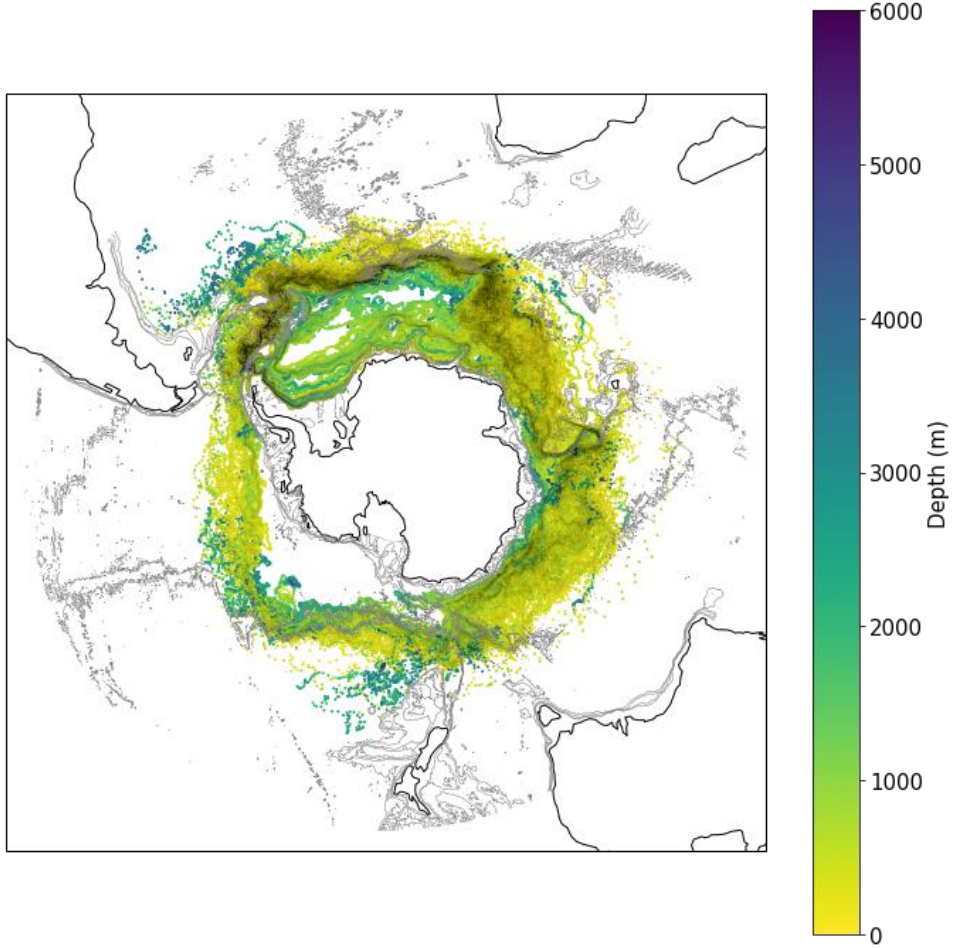
- Australian Informatics summer school (grade 11)
- AABW project in undergraduate (2^{nd} year)



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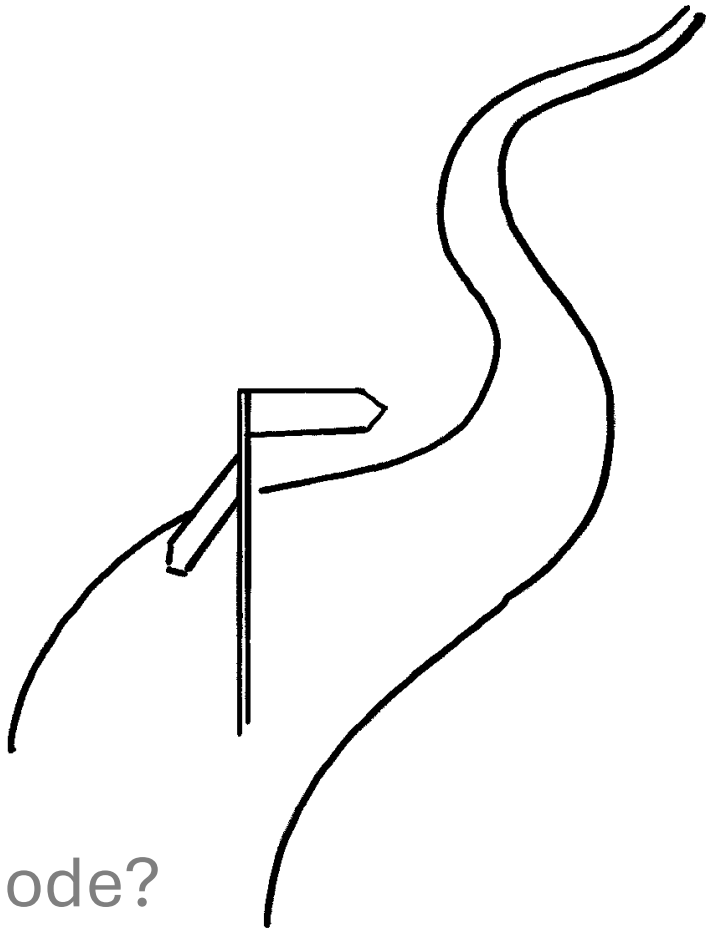
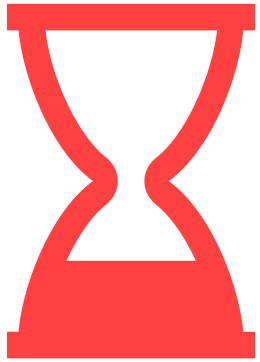


Why bother writing good code?



Save time

Do more

Make something work at all

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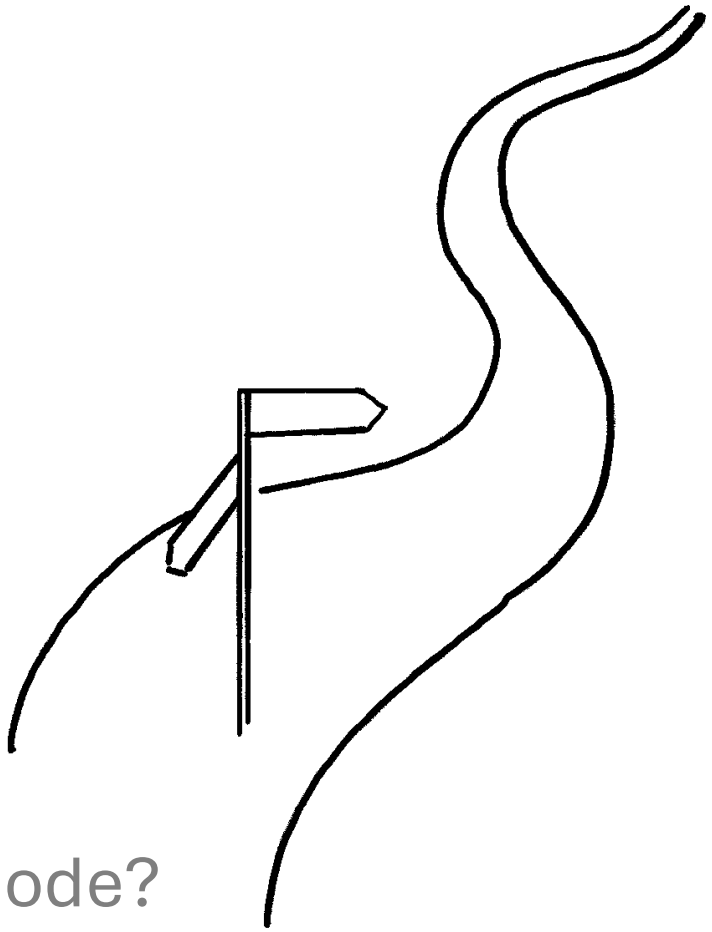

Timing examples [switch to notebook]

[1]:
```
import numpy as np
import time
```

[13]:
```
t0 = time.time()
print('hi')
print(time.time()-t0)
```

```
hi
0.00023674964904785156
```

[15]:
```
t0 = time.time()
x=1+1
print(time.time()-t0)
```

```
0.00012826919555664062
```

[16]:
```
t0 = time.time()
for i in range(100):
    x=1+1
print(time.time()-t0)
```

```
0.0002048015594482422
```

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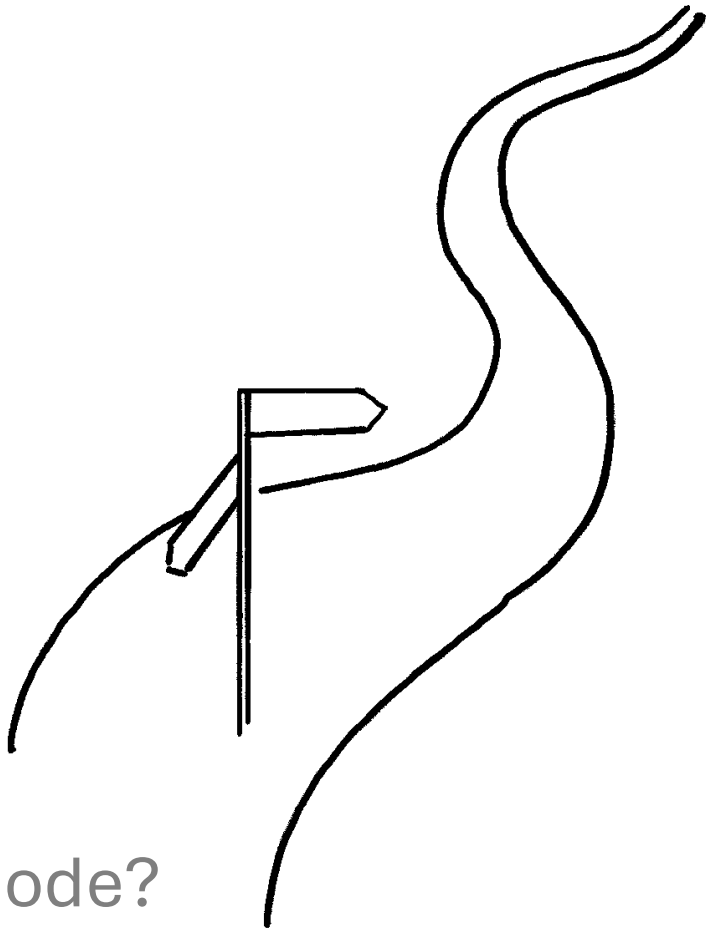
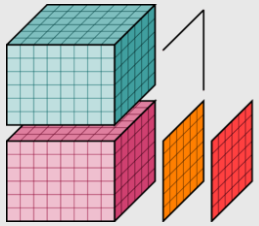


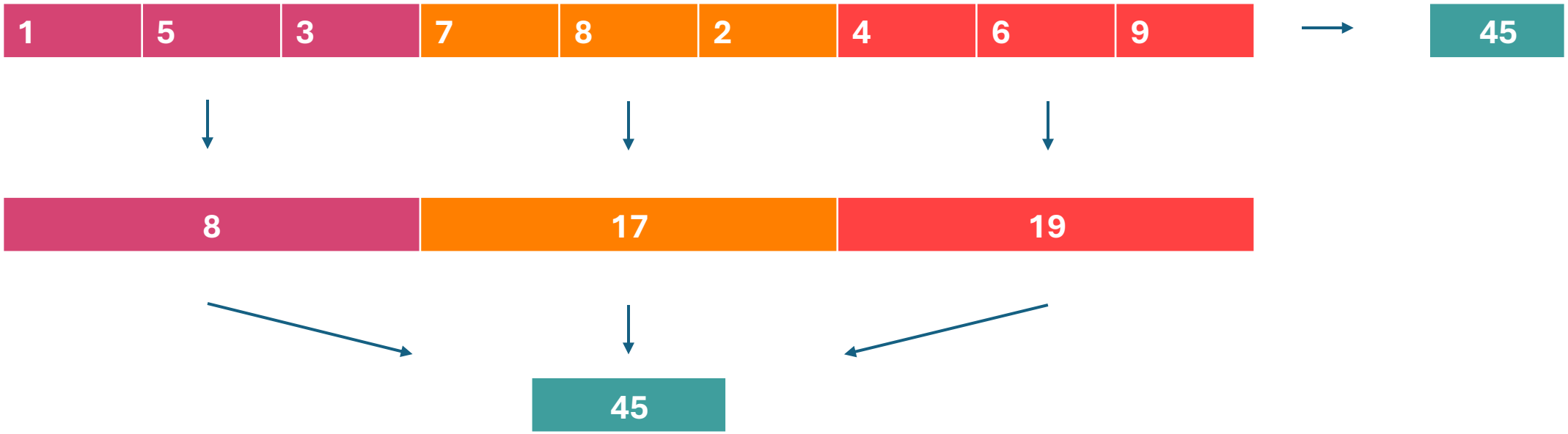


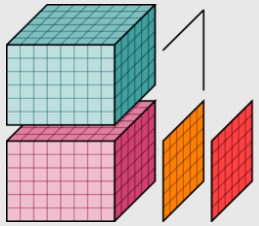
Lazy computing

dask

- Lazy ≈ “do it later”
- Smaller pieces

- Example: **summing** in different chunks



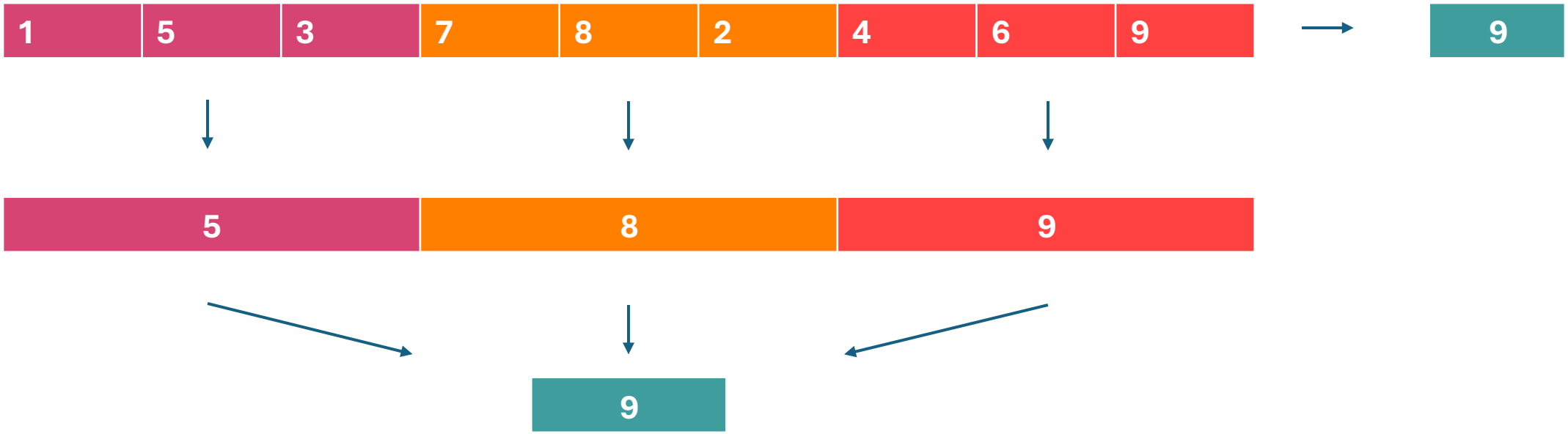


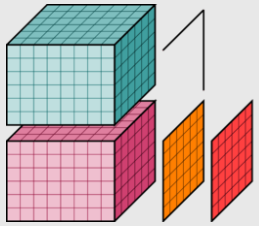
Lazy computing

dask

- Lazy ≈ “do it later”
- Smaller pieces

- Example: **finding maximum** in different chunks





[switch to notebook]

dask

[2]:
```
from dask.distributed import Client
client = Client(threads_per_worker=1,memory_limit=0)
client.amm.start()
client
```

[2]:

Client

Client-2dbbae3b-264f-11f0-b86d-000007a8fe80

Connection method: Cluster object **Cluster type:** distributed.LocalCluster

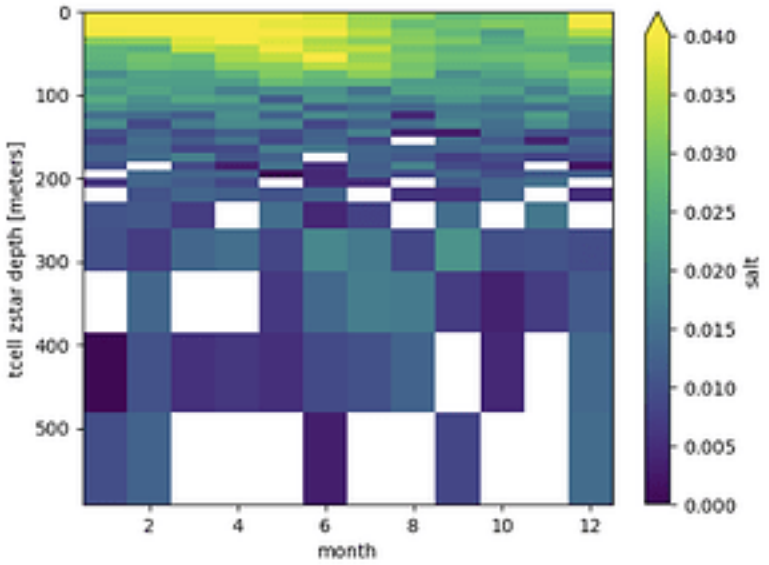
Dashboard: /proxy/8787/status

Launch dashboard in JupyterLab

▸ Cluster Info

Underlying code: an example

Salinity standard deviation



"the xarray code is wrong" – an interpretation I heard many times

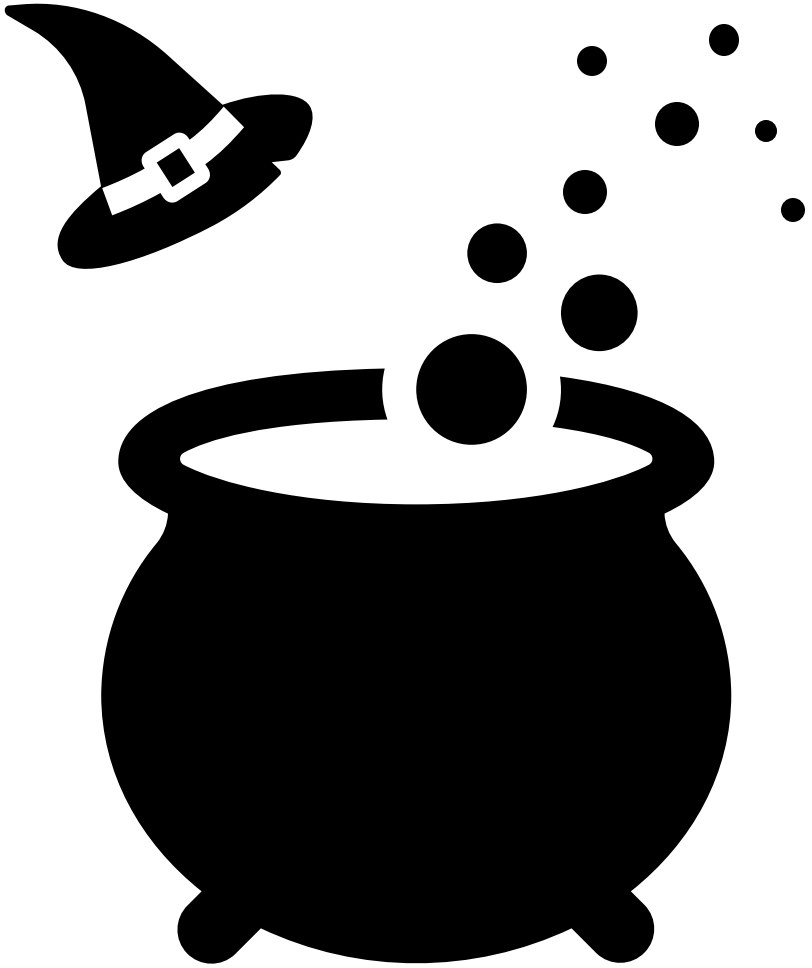
Variance:

$$\sigma^2 = \frac{1}{n}\sum_i (x_i - \bar{x})^2$$

or

$$\sigma^2 = \frac{1}{n}\sum_i \left(x_i^2\right) - \frac{1}{n}\sum_i \left(\bar{x}^2\right)$$

The black magic of chunking



- Match the chunks on file
- Either chunk *when opening* or *after loading*
- Double check your assumptions
  - Test in small batches
  - Keep the task stream open
- Dask uses ~2x the RAM of a dataset
- Sometimes dask overhead is too much to help
- Sometimes it’s faster to just shove small pieces into a for loop

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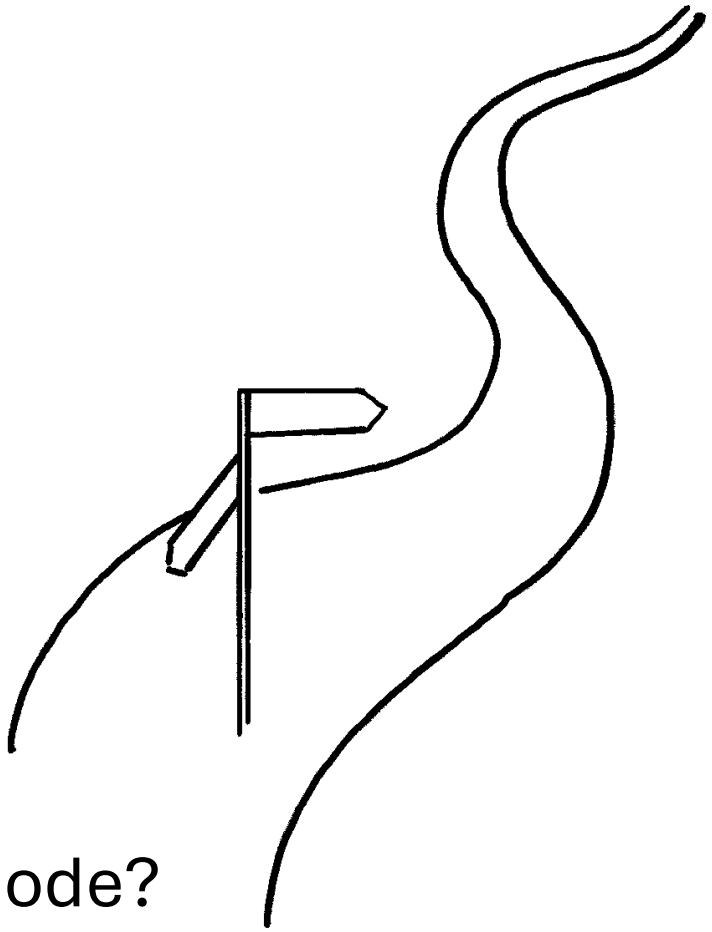
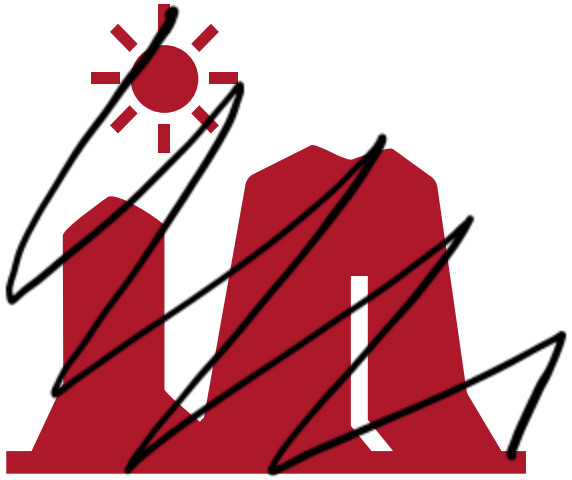
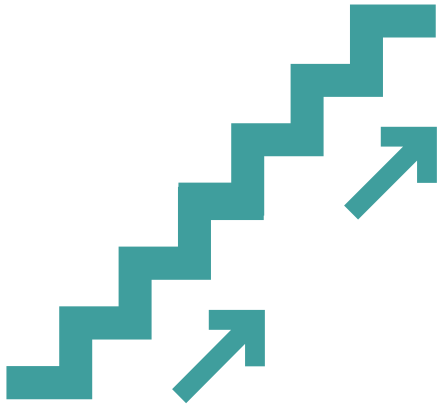


Writing code to work well with xarray/dask

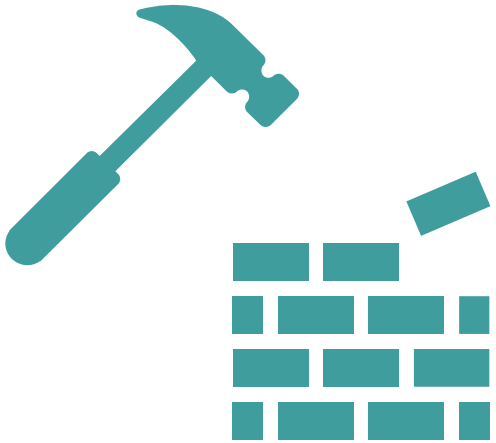
A completely different way of thinking





Rethinking your problem into these pieces?

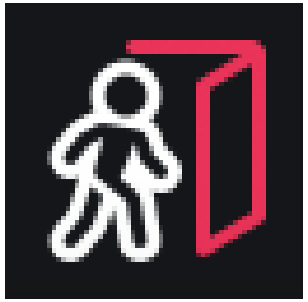
- Any maths
  `(multiply, add, np.log(), ds.cumsum(), …)`
- Dimension reduction by maths
  `(ds.sum(), ds.std(), …)`
- Dimension reduction by picking one
  `(ds.max(), ds.argmin(),ds.sel(), …)`
- Moving stuff
  `(ds.shift(), ds.coarsen(), …)`
- Grabbing strange pieces
  `(ds.where(), ds.groupby(), …)`



Full list here: https://docs.xarray.dev/en/stable/generated/xarray.DataArray.html

Rethinking code: examples

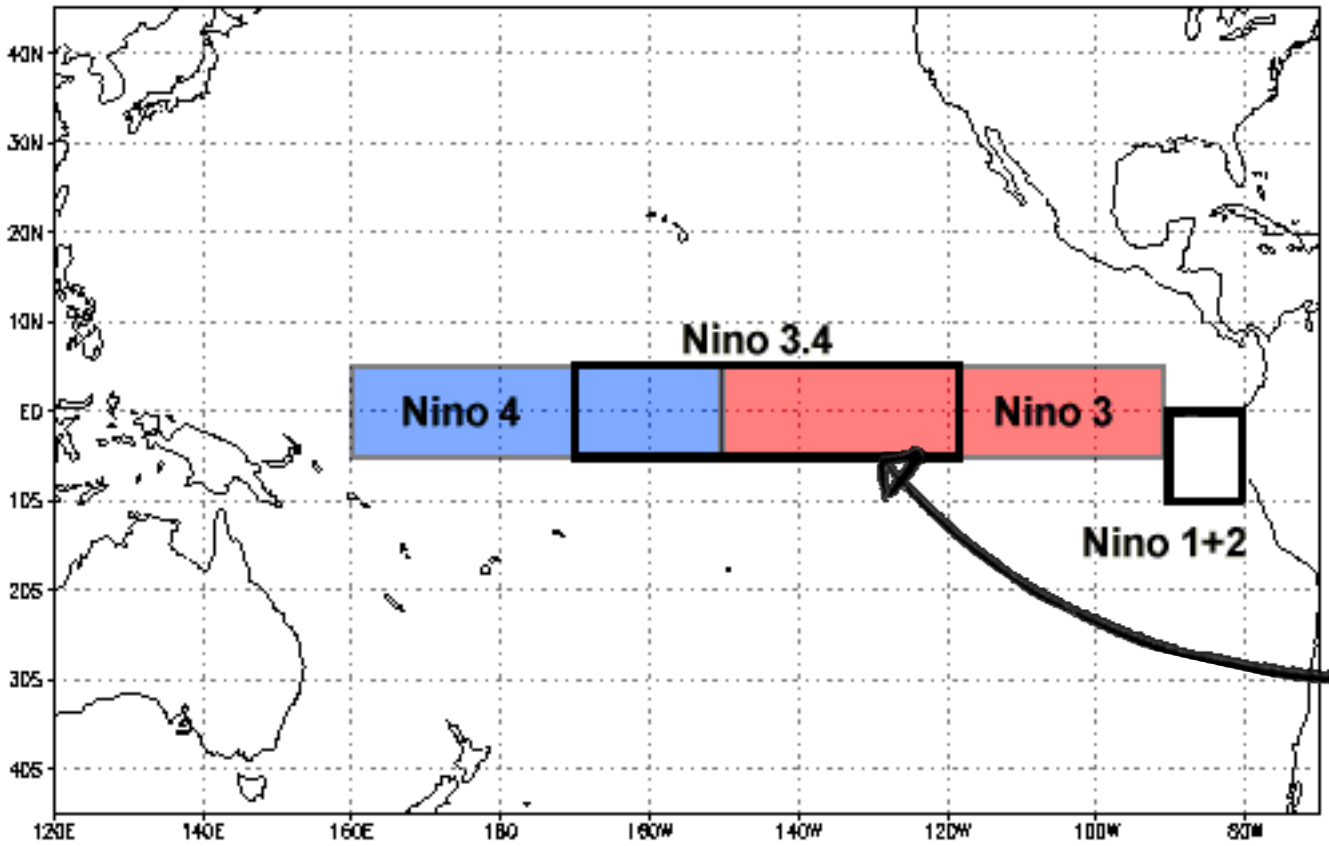
1. NINO34 index
2. Weighted rolling average
3. 20°C isotherm
4. Model analogue forecasts



?

Example 1: NINO34 index

- I have sea surface temperature (SST) from a model



Average that box

Example 1: NINO34 index

- I have sea surface temperature (SST) from a model
- I want to calculate the average SST within the NINO34 region (5°S-5°N, 190°-240°E) for every time

Example 1: NINO34 index

- I have sea surface temperature (SST) from a model
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1. Get data within the NINO34 region

```
nino34_sst = sst.sel(lat=slice(-5,5),lon=slice(190,240))
```

```
nino34_sst = sst.where((sst.lat <-5)&(sst.lat <-5)
        &(sst.lon >190)&(sst.lon<240))
```

2. Average in latitude and longitude space

```
nino34_sst.mean(("lat","lon"))
```

Example 2: Rolling weighted average through time

$$\text{Weighted rolling average} = \frac{rolling\ sum\ of\ (weights \times data)}{rolling\ sum\ of\ weights}$$

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$$\text{Weighted rolling average} = \frac{rolling\ sum\ of\ (weights \times data)}{rolling\ sum\ of\ weights}$$

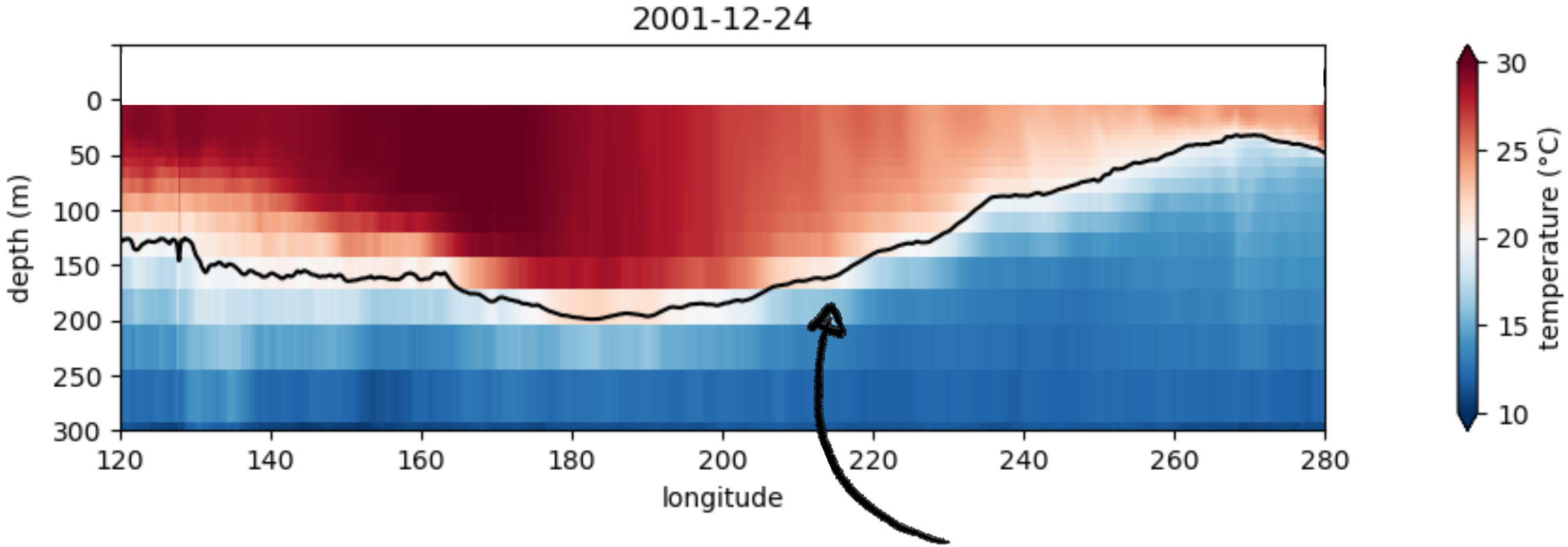
1. Multiply data by weights
2. Find rolling sum

```
numerator = (weights*data).rolling({'time':3}).sum()
```

3. Find rolling sum of weights
4. Divide

```
numerator = (weights).rolling({'time':3}).sum()
```

Example 3: 20° isotherm



Find that depth

Example 3: 20° isotherm

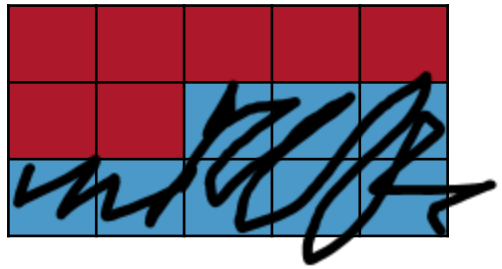
- Find 20° C isotherm

= find depth that is 20° C

≈ find deepest depth that is warmer than 20° C

1 Depth warmer than 20°C

```
warm_depth = depth.where(ds>20)
```



2 Deepest depth

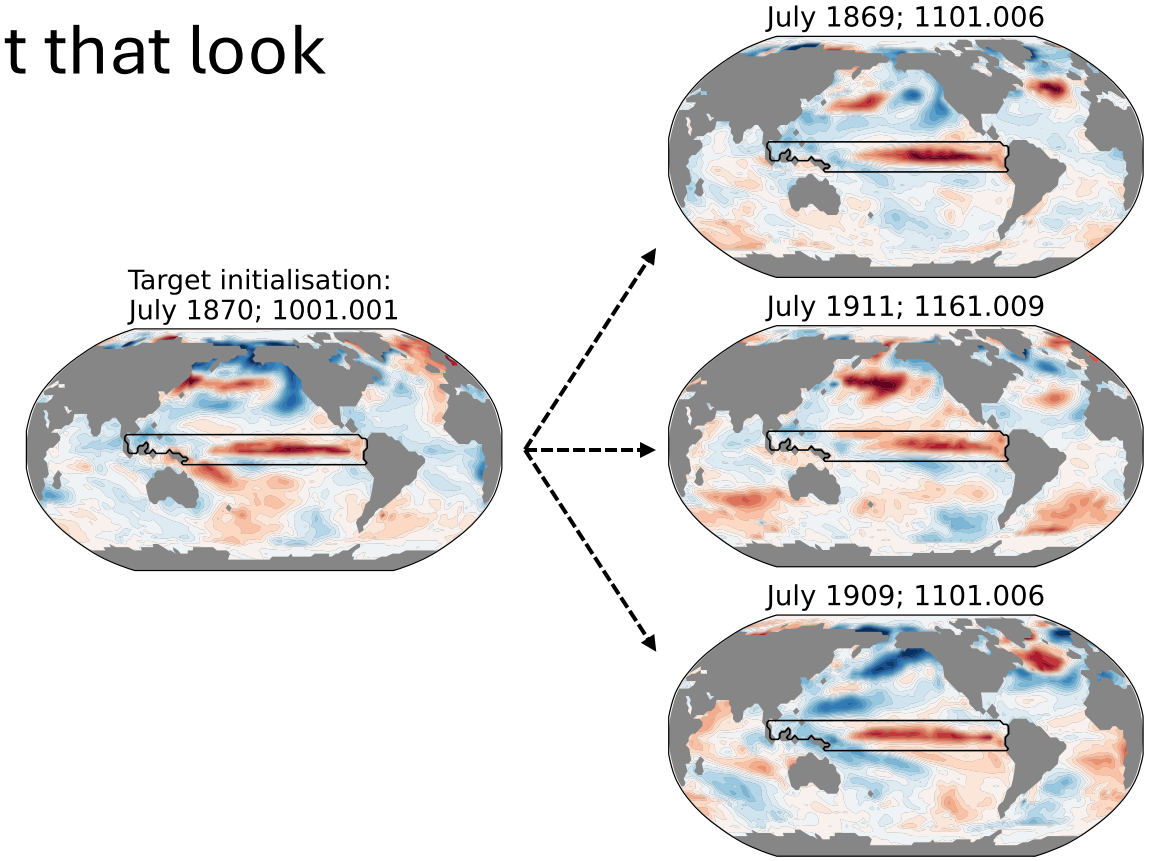
```
warm_depth.argmax("depth")
```

or

```
warm_depth.idxmax("depth")
```

Example 4: analogue forecasts

- Find the months of model output that look most similar to my target month



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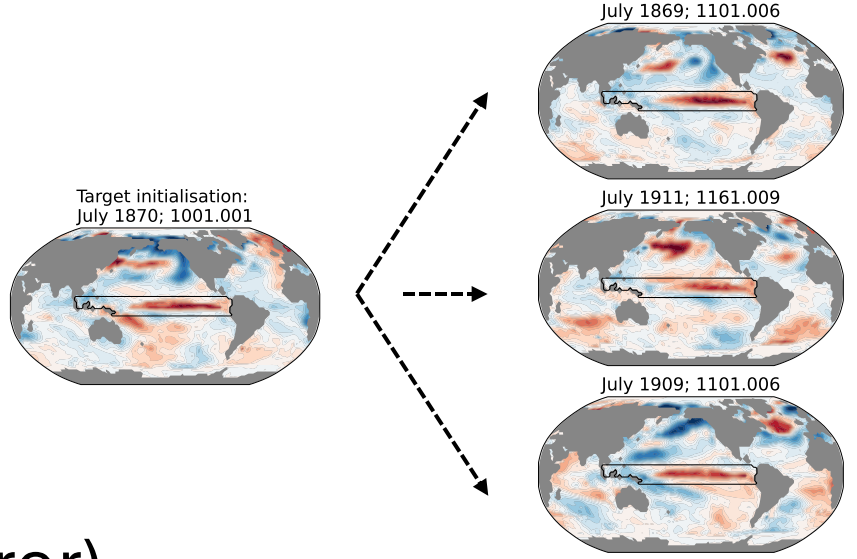

1. Calculate similarity (ie, mean squared error)

```
mse = ((all_sst-target_sst)**2).mean(('lat','lon'))
```

2. Find the indices of lowest mean squared error

```
mse.idxmin('time')
```

The last resort: xr.apply_ufunc

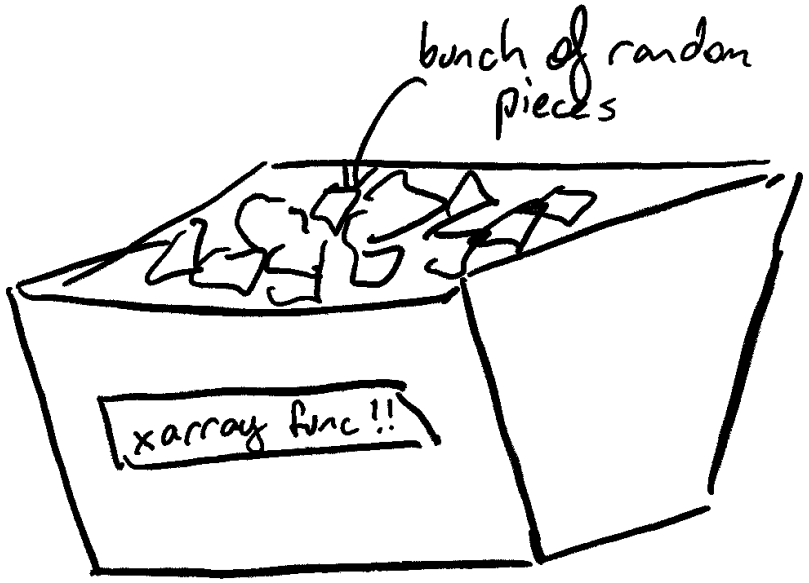

Fig 1. My nice contained function

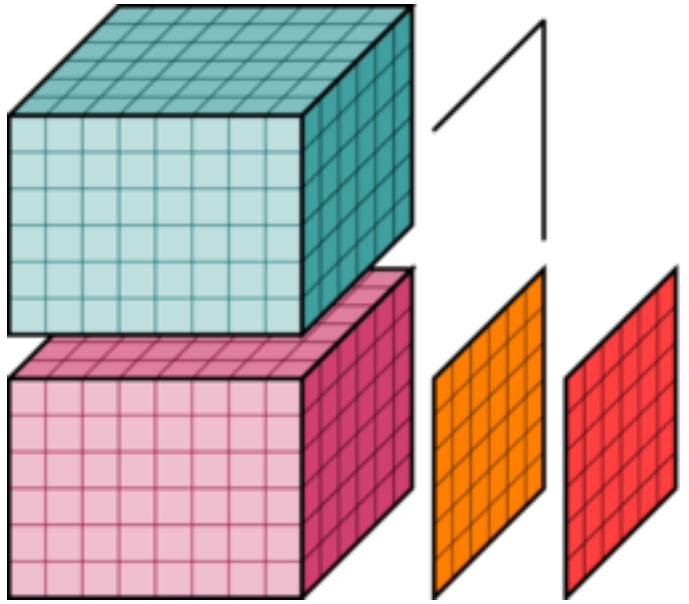

Fig 2. xarray

cosima-recipes/Tutorials/Apply_function_to_every_gridpoint.ipynb

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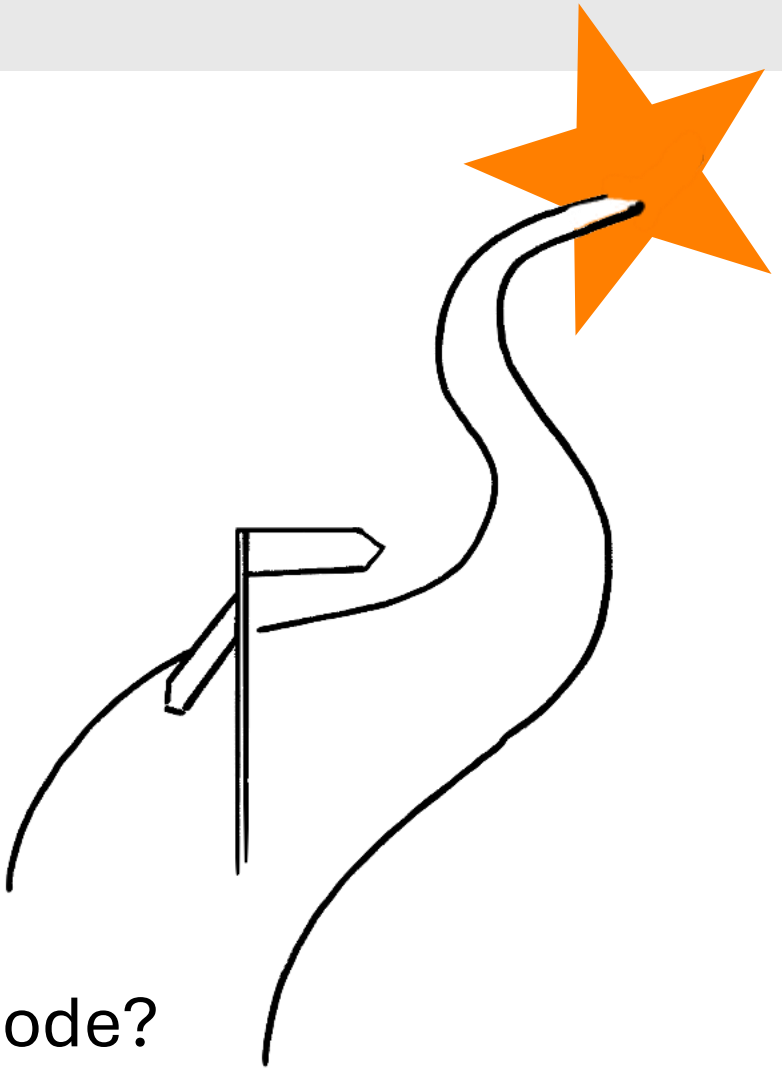
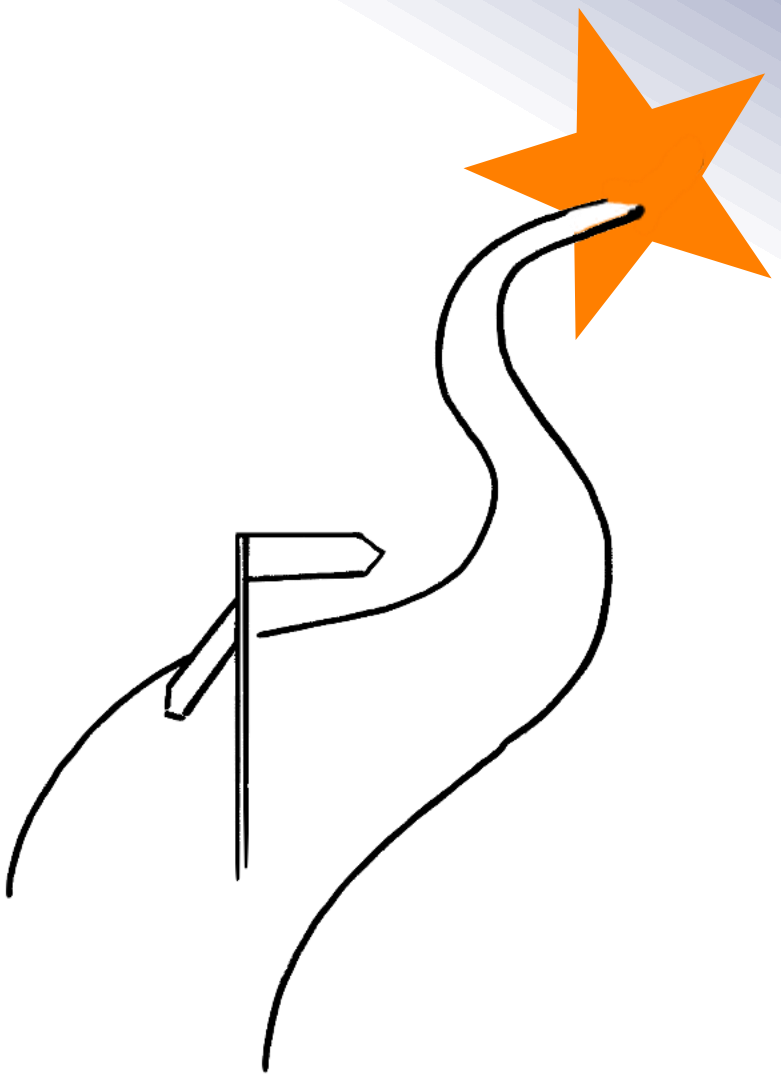
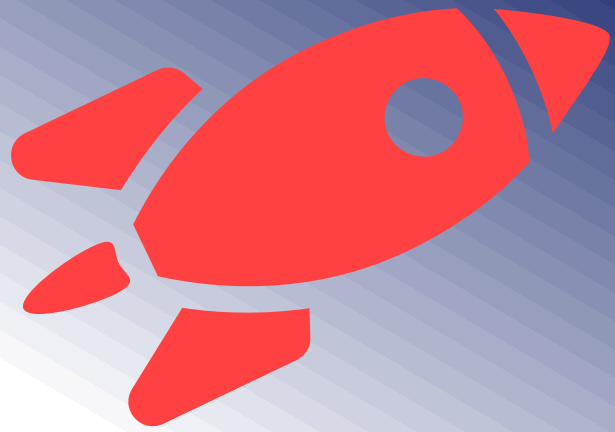


Good luck





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