

Python_course_5_oceanography

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1 Python course 5: Introduction to tools for oceanography

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In this course, we use climate datasets from Copernicus, that can be downloaded for free from their website at <https://cds.climate.copernicus.eu/#!/home>.

For datasets that are too big to fit into memory, you can use Dask: <https://xarray.pydata.org/en/stable/user-guide/dask.html>

1.1 Import the necessary modules

These modules, we already know:

```
[1]: import numpy as np
import matplotlib.pyplot as plt
import cmoclean
```

These modules are new today:

```
[2]: import gsw

import xarray as xr

import cartopy.crs as ccrs
import cartopy.feature as cfeature
```

crs stands for Coordinate Reference System.

1.2 Configure the visualization

```
[3]: plt.rcParams["font.size"] = 14
```

1.3 Example 1: sea surface temperature

1.3.1 Load the dataset

```
[4]: ds_SST = xr.load_dataset("../Data/  
→dataset-satellite-sea-surface-temperature-f9354174-5df6-438e-913a-63ae8bc7fc02/  
→20170101120000-C3S-L4_GHRSST-SSTdepth-OSTIA-GLOB_ICDR2.0-v02.0-fv01.0.nc")  
ds_SST
```

```
[4]: <xarray.Dataset>  
Dimensions:                (lat: 3600, lon: 7200, time: 1)  
Coordinates:               (time) datetime64[ns] 2017-01-01T12:00:00  
                            * lat                (lat) float32 -89.975 -89.925 ... 89.925 89.975  
                            * lon                (lon) float32 -179.975 -179.925 ... 179.925 179.975  
Data variables:            analysed_sst          (time, lat, lon) float32 nan nan nan ... 271.35 271.35  
                            analysis_uncertainty (time, lat, lon) float32 nan nan ... 0.72999996  
                            sea_ice_fraction     (time, lat, lon) float32 nan nan nan ... 0.94 0.94  
                            mask                 (time, lat, lon) float32 2.0 2.0 2.0 ... 9.0 9.0 9.0  
Attributes:               Conventions:          CF-1.5, Unidata Observation Dataset v1.0  
                            title:              C3S SST L4 product  
                            summary:            OSTIA L4 product from the C3S project, p...  
                            references:         http://www.esa-sst-cci.org  
                            institution:        C3S  
                            history:            Created using OSTIA reanalysis system v3.0  
                            comment:            These data were produced by the Met Offi...  
                            license:            Creative Commons Licence by attribution ...  
                            id:                 OSTIA-C3S-L4-GLOB-v2.0  
                            naming_authority:   org.ghrsst  
                            product_version:    2.0  
                            uuid:               7fdf2639-26e5-4d4f-a60e-0bcfc9744204  
                            tracking_id:         7fdf2639-26e5-4d4f-a60e-0bcfc9744204  
                            gds_version_id:     2.0  
                            netcdf_version_id:  4.2.1.1 of Oct 19 2012 14:25:16  
                            date_created:       20180620T065809Z  
                            creation_date:      2018-06-20T06:58:09Z  
                            file_quality_level:  3  
                            spatial_resolution: 0.05 degree  
                            start_time:         20170101T000000Z  
                            time_coverage_start: 20170101T000000Z  
                            stop_time:          20170102T000000Z  
                            time_coverage_end:   20170102T000000Z  
                            time_coverage_duration: P1D  
                            time_coverage_resolution: P1D  
                            northernmost_latitude: 90.0  
                            geospatial_lat_max: 90.0
```

```

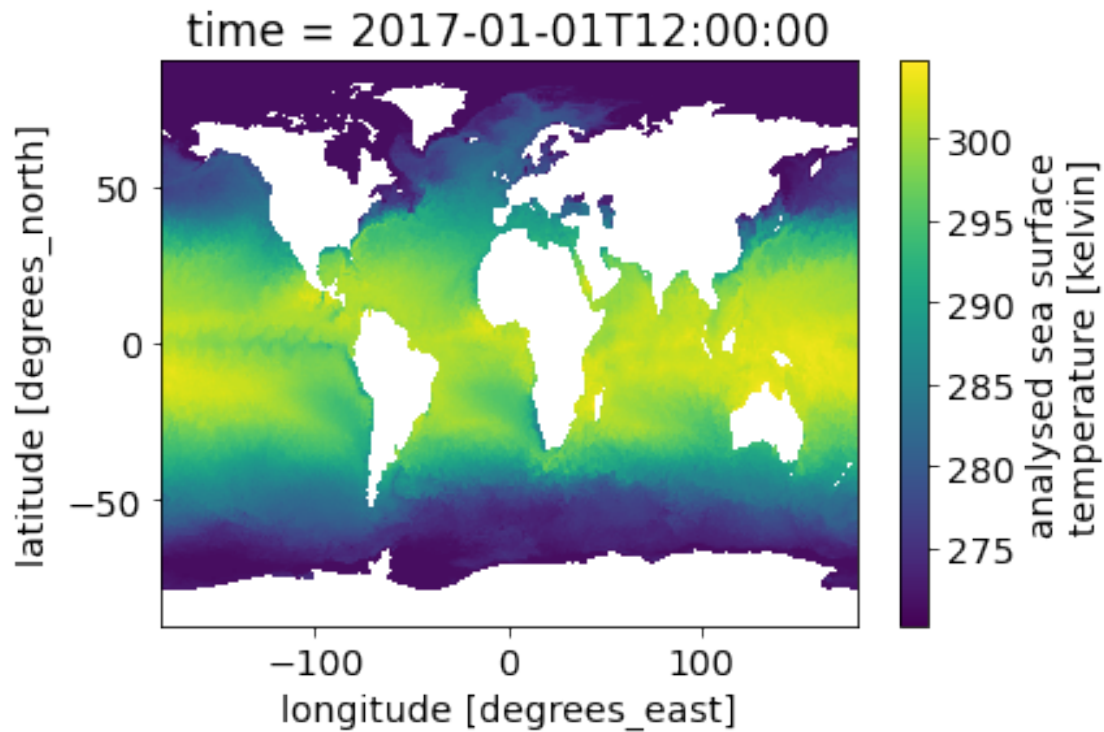
southernmost_latitude: -90.0
geospatial_lat_min: -90.0
easternmost_longitude: 180.00002
geospatial_lon_max: 180.00002
westernmost_longitude: -180.0
geospatial_lon_min: -180.0
geospatial_vertical_min: -0.2
geospatial_vertical_max: -0.2
source: AVHRR19_G-C3S-L3U-ICDR-v2.0 AVHRRMTA_G-C...
platform: NOAA-19, MetOpA, Sentinel-3A
sensor: AVHRR_GAC, SLSTR
Metadata_Conventions: Unidata Dataset Discovery v1.0
metadata_link: http://www.esa-cci.org
keywords: Oceans > Ocean Temperature > Sea Surface...
keywords_vocabulary: NASA Global Change Master Directory (GCM...
standard_name_vocabulary: NetCDF Climate and Forecast (CF) Metadat...
geospatial_lat_units: degrees_north
geospatial_lat_resolution: 0.05
geospatial_lon_units: degrees_east
geospatial_lon_resolution: 0.05
acknowledgment: Funded by the Copernicus Climate Change ...
creator_name: Copernicus Climate Change Service (C3S)
creator_email: science.leader@esa-sst-cci.org
creator_processing_institution: These data were produced on the JASMIN i...
creator_url: https://climate.copernicus.eu/
project: Copernicus Climate Change Service (C3S)
publisher_name: Copernicus Climate Data Store
publisher_url: https://climate.copernicus.eu/
publisher_email: copernicus-support@ecmwf.int
processing_level: L4
cdm_data_type: grid
product_specification_version: SST_CCI-PSD-UKMO-201-Issue-H
contact: http://copernicus-support.ecmwf.int

```

1.3.2 Look at the data

```
[5]: ds_SST.analysed_sst.plot()
```

```
[5]: <matplotlib.collections.QuadMesh at 0x7f460bc75dd0>
```



1.3.3 Analyse the data

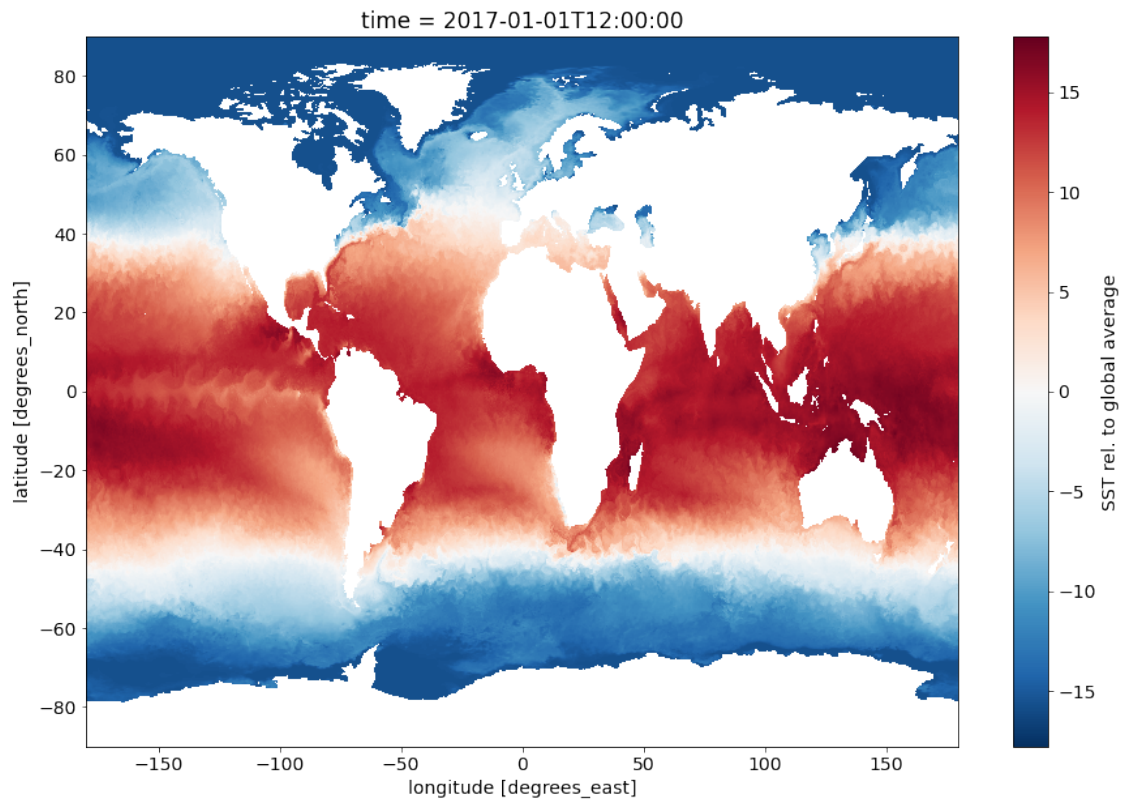
```
[6]: print(f"The global mean SST is {ds_SST.analysed_sst.mean().values - 273.15:.2f}K  
      ↪ °C")
```

The global mean SST is 13.81 °C

```
[7]: T_anom = ds_SST.analysed_sst - ds_SST.analysed_sst.mean()
      T_anom.name = "SST rel. to global average"

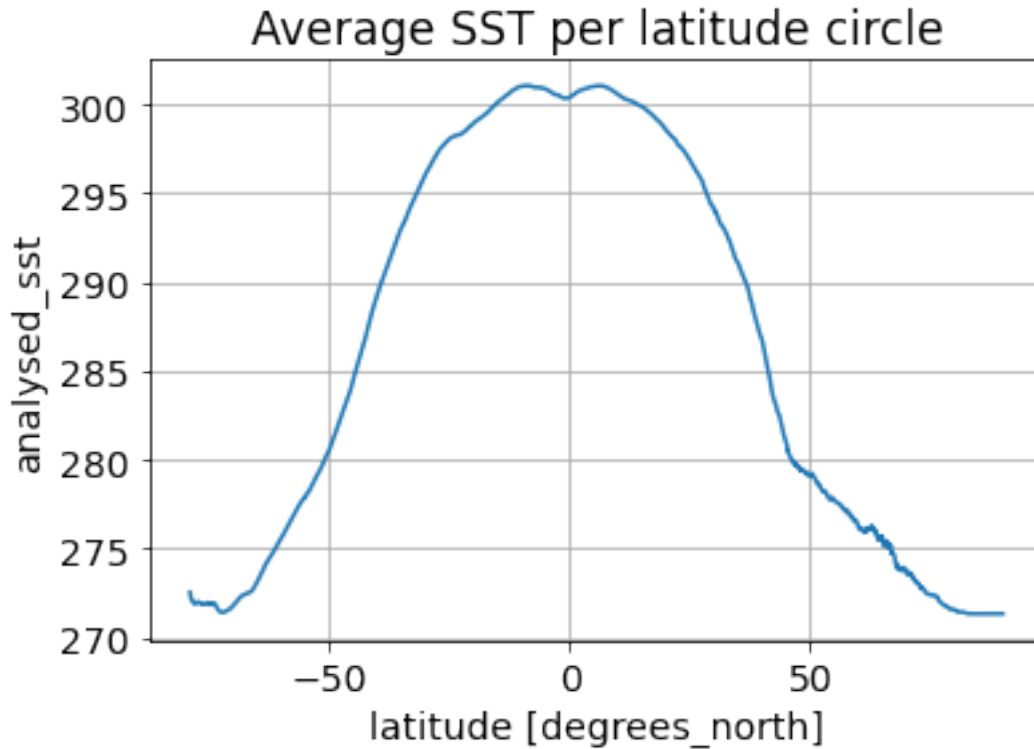
      plt.figure(figsize=(15, 10))
      T_anom.plot()
```

```
[7]: <matplotlib.collections.QuadMesh at 0x7f45fd550dd0>
```



```
[8]: ds_SST.analysed_sst.mean("lon").plot()  
plt.grid()  
plt.title("Average SST per latitude circle")
```

```
[8]: Text(0.5, 1.0, 'Average SST per latitude circle')
```

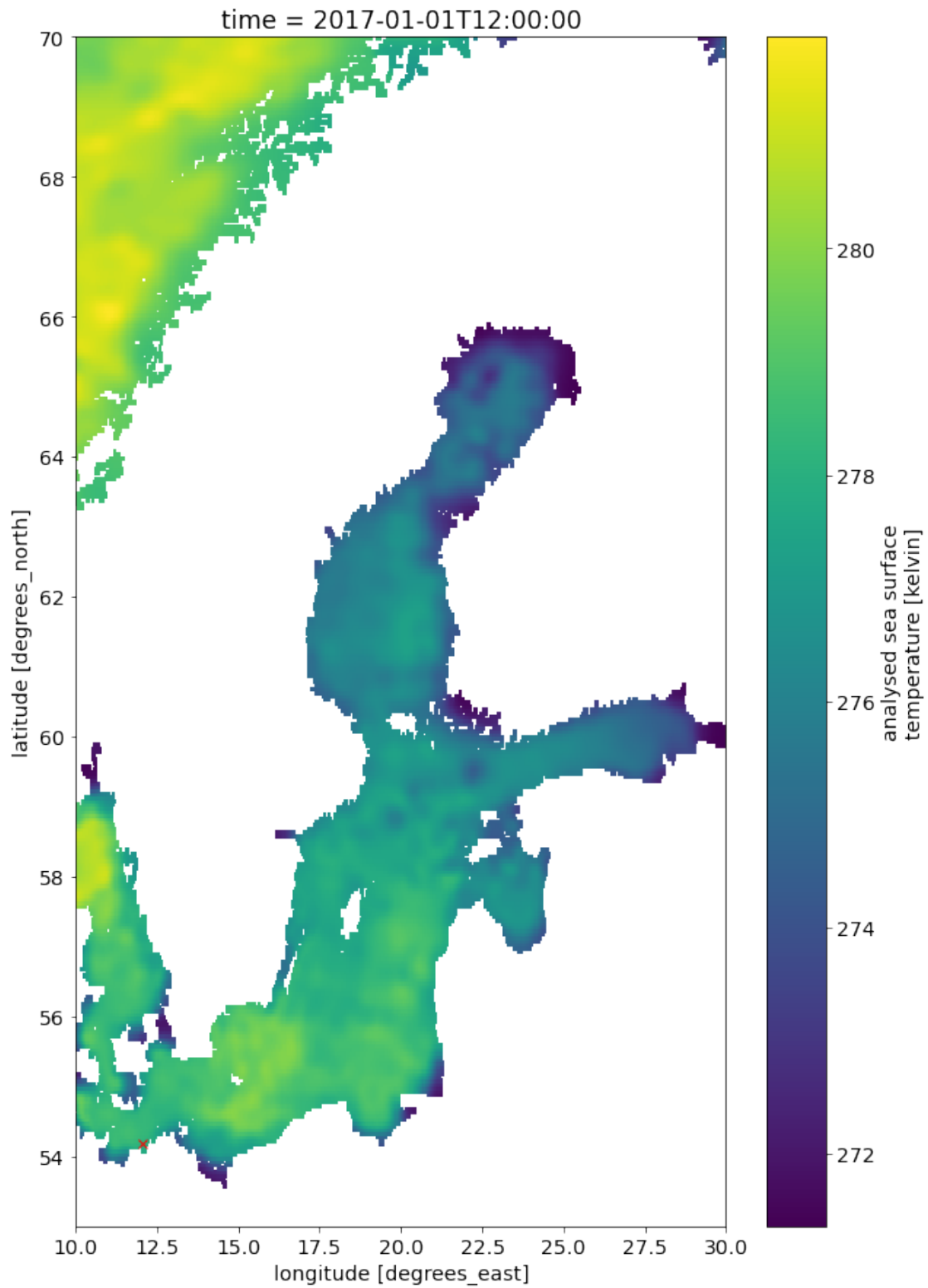


1.3.4 Select specific data

```
[9]: plt.figure(figsize=(10, 15))
# Zoom to the Baltic Sea
ds_SST.analysed_sst.sel(lat=slice(53, 70), lon=slice(10, 30)).plot()
# Mark the mouth of the Warnow
lat_Warnemünde = 54.175
lon_Warnemünde = 12.075
plt.plot(lon_Warnemünde, lat_Warnemünde, "rx")
# Get the temperature in the mouth of the Warnow
ds_SST.analysed_sst.sel(lat=lat_Warnemünde, lon=lon_Warnemünde)
```

```
[9]: <xarray.DataArray 'analysed_sst' (time: 1)>
array([278.04], dtype=float32)
Coordinates:
  * time      (time) datetime64[ns] 2017-01-01T12:00:00
    lat       float32 54.175
    lon       float32 12.075
Attributes:
  long_name:    analysed sea surface temperature
  standard_name: sea_water_temperature
  units:       kelvin
```

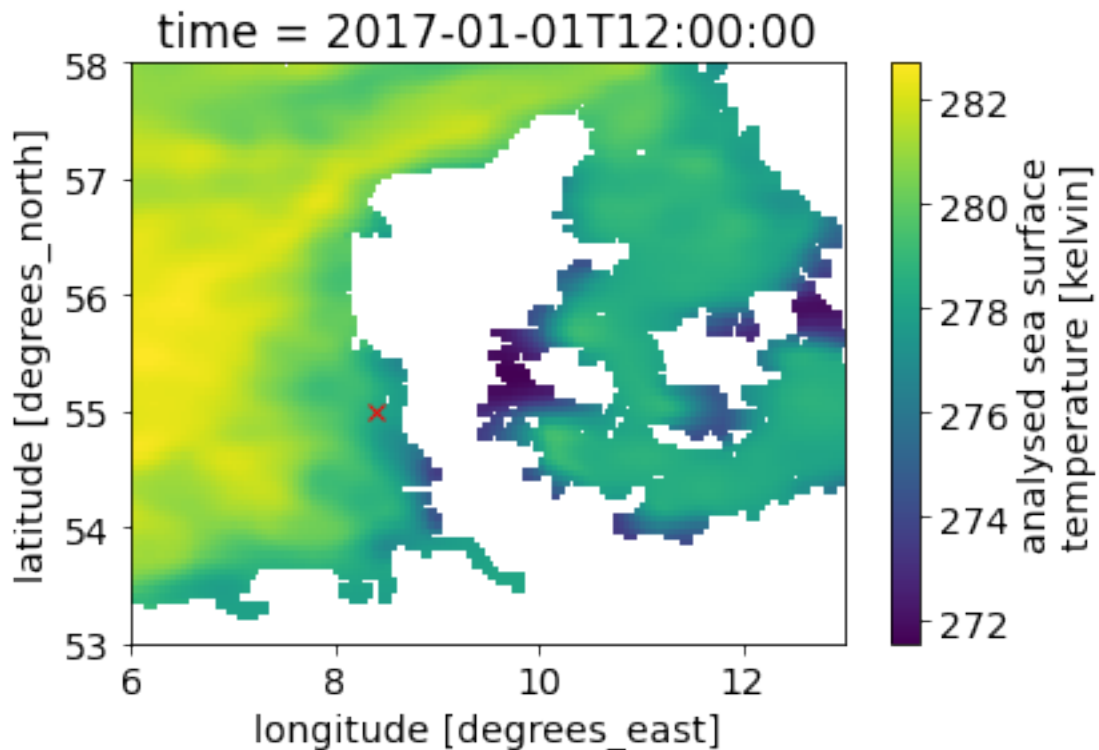
valid_min: -300
valid_max: 4500
source: AVHRR19_G-C3S-L3U-ICDR-v2.0 AVHRRMTA_G-C3S-L3U-ICDR-v2.0 ...



```
[10]: # Zoom to the German and Danish coasts
ds_SST.analysed_sst.sel(lat=slice(53, 58), lon=slice(6, 13)).plot()
# Mark Sylt
plt.plot(8.4, 55, "rx")

# Get the temperature near Sylt
ds_SST.analysed_sst.sel(lat=55, lon=8.4, method="nearest")
```

```
[10]: <xarray.DataArray 'analysed_sst' (time: 1)>
array([278.46], dtype=float32)
Coordinates:
  * time      (time) datetime64[ns] 2017-01-01T12:00:00
    lat       float32 55.025
    lon       float32 8.375
Attributes:
  long_name:      analysed sea surface temperature
  standard_name:  sea_water_temperature
  units:          kelvin
  valid_min:      -300
  valid_max:      4500
  source:         AVHRR19_G-C3S-L3U-ICDR-v2.0 AVHRRMTA_G-C3S-L3U-ICDR-v2.0 ...
```



1.4 Example 2: unknown dataset

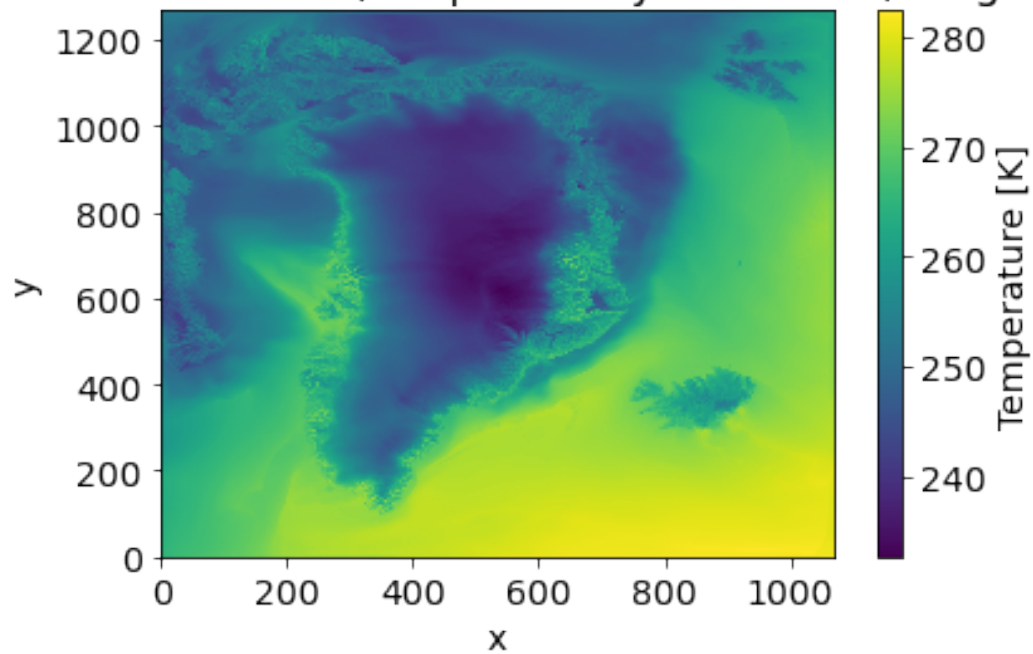
```
[4]: ds = xr.load_dataset("../Data/adaptor.uerra_mars.external-1620771977.  
    ↪1561432-27059-25-89c03805-2d54-4fc9-a4de-549d70380e0a.nc")  
ds
```

```
[4]: <xarray.Dataset>  
Dimensions:                (heightAboveGround: 11, time: 12, x: 1069, y: 1269)  
Coordinates:               (time) datetime64[ns] 2019-01-01 ... 2019-12-01  
    step                   timedelta64[ns] 00:00:00  
    * heightAboveGround    (heightAboveGround) int64 15 30 50 75 ... 250 300 400 500  
    latitude              (y, x) float64 55.81 55.82 55.82 ... 77.87 77.85 77.83  
    longitude             (y, x) float64 302.9 302.9 303.0 ... 37.56 37.6 37.63  
    valid_time            (time) datetime64[ns] 2019-01-01 ... 2019-12-01  
Dimensions without coordinates: x, y  
Data variables:            (time, heightAboveGround, y, x) float32 101643.31 ... ...  
    pres                  (time, heightAboveGround, y, x) float32 264.41602 ... ...  
    t                    (time, heightAboveGround, y, x) float32 320.44083 ... ...  
    wdir                 (time, heightAboveGround, y, x) float32 8.533268 ... 2...  
    ws                  (time, heightAboveGround, y, x) float32 8.533268 ... 2...  
Attributes:               (time, heightAboveGround, y, x) float32 8.533268 ... 2...  
    GRIB_edition:        2  
    GRIB_centre:         enmi  
    GRIB_centreDescription: Oslo  
    GRIB_subCentre:      255  
    Conventions:         CF-1.7  
    institution:         Oslo  
    history:             2021-05-12T06:59:36 GRIB to CDM+CF via cfgrib-0...
```

```
[5]: # Let us look at the dataset to get an idea for it  
ds.t.isel(time=0, heightAboveGround=0).plot()
```

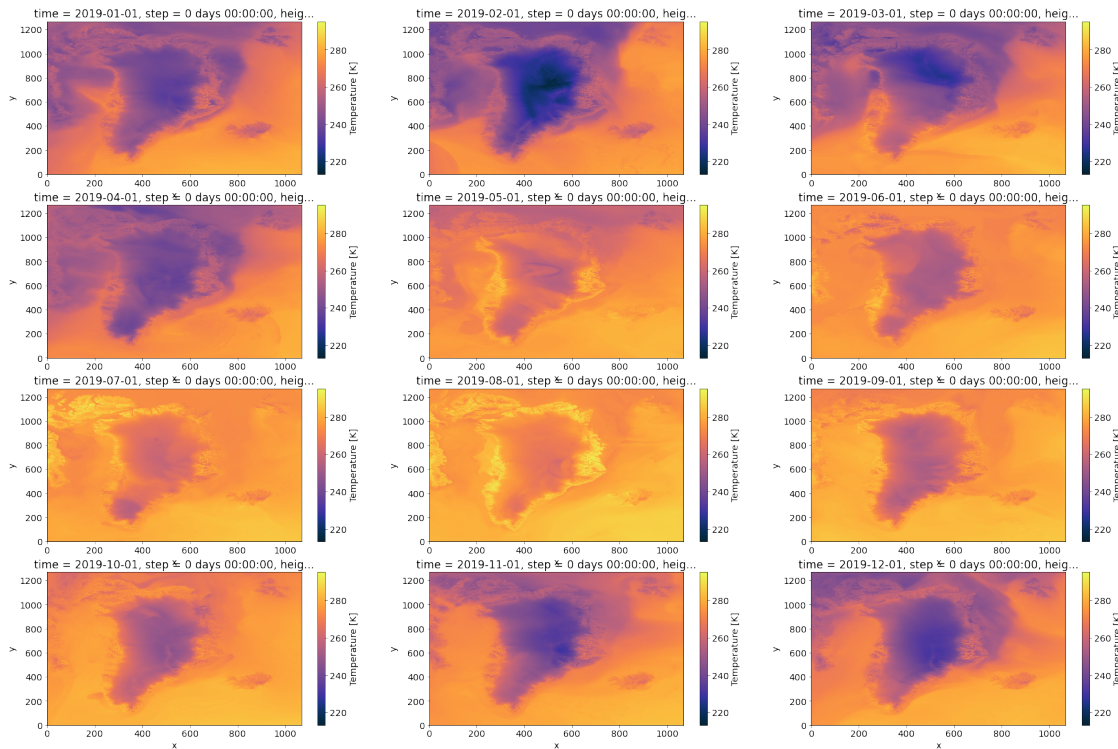
```
[5]: <matplotlib.collections.QuadMesh at 0x7fd858f6b810>
```

time = 2019-01-01, step = 0 days 00:00:00, heig...



```
[6]: # Let us plot a time series of the near ground temperature
ground_temperature = ds.t.sel(heightAboveGround=0, method="nearest")

fig, axs = plt.subplots(ncols=3, nrows=4, figsize=(30, 20))
for i, ax in enumerate(axs.flatten()):
    ground_temperature.isel(time=i).plot(
        ax=ax, vmin=ground_temperature.min(), vmax=ground_temperature.max(),
        cmap=cmocean.cm.thermal
    )
```



1.5 Example 3: Argo CTD profile

This dataset has been downloaded from Ifremer's FTP-server. For the link to the server and information on data access, see: <http://www.argodatamgt.org/Access-to-data/Access-via-FTP-or-HTTPS-on-GDAC>

For more information on the ARGO project, see: <https://argo.ucsd.edu/>

```
[7]: ds_ARGO = xr.load_dataset("../Data/ARGO_20210512_prof.nc")
ds_ARGO = ds_ARGO.sel(N_PROF=136) # select a single profile from the dataset
ds_ARGO
```

```
[7]: <xarray.Dataset>
Dimensions:                      (N_CALIB: 2, N_HISTORY: 0, N_LEVELS: 2027,
N_PARAM: 3)
Dimensions without coordinates: N_CALIB, N_HISTORY, N_LEVELS, N_PARAM
Data variables:
    DATA_TYPE                    object b'Argo profile'
    FORMAT_VERSION                object b'3.1'
    HANDBOOK_VERSION              object b'1.2'
    REFERENCE_DATE_TIME           object b'19500101000000'
    DATE_CREATION                 object b'20210513142752'
    DATE_UPDATE                   object b'20210513142752'
    PLATFORM_NUMBER               object b'4902106'
```

| | | |
|------------------------------|--|-------------|
| PROJECT_NAME | object b'US ARGO PROJECT | ... |
| PI_NAME | object b'BRECK OWENS, STEVEN JAYNE, P.E. RO... | |
| STATION_PARAMETERS | (N_PARAM) object b'PRES | ' ... b'... |
| CYCLE_NUMBER | float64 207.0 | |
| DIRECTION | object b'A' | |
| DATA_CENTRE | object b'AO' | |
| DC_REFERENCE | object b'6463 | ' |
| DATA_STATE_INDICATOR | object b'2B | ' |
| DATA_MODE | object b'A' | |
| PLATFORM_TYPE | object b'S2A | ' |
| FLOAT_SERIAL_NO | object b'7318 | ' |
| FIRMWARE_VERSION | object b'SBE602 ARM_v2.0_xmsg_ve | ' |
| WMO_INST_TYPE | object b'854 | ' |
| JULD | datetime64[ns] 2021-05-12T01:48:08 | |
| JULD_QC | object b'1' | |
| JULD_LOCATION | datetime64[ns] 2021-05-12T01:52:00 | |
| LATITUDE | float64 44.4 | |
| LONGITUDE | float64 -34.1 | |
| POSITION_QC | object b'1' | |
| POSITIONING_SYSTEM | object b'GPS | ' |
| PROFILE_PRES_QC | object b'A' | |
| PROFILE_TEMP_QC | object b'A' | |
| PROFILE_Psal_QC | object b'A' | |
| VERTICAL_SAMPLING_SCHEME | object b'Primary sampling: averaged [nomina... | |
| CONFIG_MISSION_NUMBER | float64 4.0 | |
| PRES | (N_LEVELS) float32 1.2 2.0 3.04 ... nan nan | |
| PRES_QC | (N_LEVELS) object b'1' b'1' b'1' ... nan nan | |
| PRES_ADJUSTED | (N_LEVELS) float32 1.2 2.0 3.04 ... nan nan | |
| PRES_ADJUSTED_QC | (N_LEVELS) object b'1' b'1' b'1' ... nan nan | |
| PRES_ADJUSTED_ERROR | (N_LEVELS) float32 nan nan nan ... nan nan nan | |
| TEMP | (N_LEVELS) float32 15.778 15.777 ... nan nan | |
| TEMP_QC | (N_LEVELS) object b'1' b'1' b'1' ... nan nan | |
| TEMP_ADJUSTED | (N_LEVELS) float32 15.778 15.777 ... nan nan | |
| TEMP_ADJUSTED_QC | (N_LEVELS) object b'1' b'1' b'1' ... nan nan | |
| TEMP_ADJUSTED_ERROR | (N_LEVELS) float32 nan nan nan ... nan nan nan | |
| PSAL | (N_LEVELS) float32 36.096 36.096 ... nan nan | |
| PSAL_QC | (N_LEVELS) object b'2' b'2' b'2' ... nan nan | |
| PSAL_ADJUSTED | (N_LEVELS) float32 36.04031 36.04031 ... nan | |
| PSAL_ADJUSTED_QC | (N_LEVELS) object b'2' b'2' b'2' ... nan nan | |
| PSAL_ADJUSTED_ERROR | (N_LEVELS) float32 nan nan nan ... nan nan nan | |
| PARAMETER | (N_CALIB, N_PARAM) object b'PRES | ... |
| SCIENTIFIC_CALIB_EQUATION | (N_CALIB, N_PARAM) object b'none | ... |
| SCIENTIFIC_CALIB_COEFFICIENT | (N_CALIB, N_PARAM) object b' | ... |
| SCIENTIFIC_CALIB_COMMENT | (N_CALIB, N_PARAM) object b' | ... |
| SCIENTIFIC_CALIB_DATE | (N_CALIB, N_PARAM) object b' | '... |
| HISTORY_INSTITUTION | (N_HISTORY) object | |
| HISTORY_STEP | (N_HISTORY) object | |

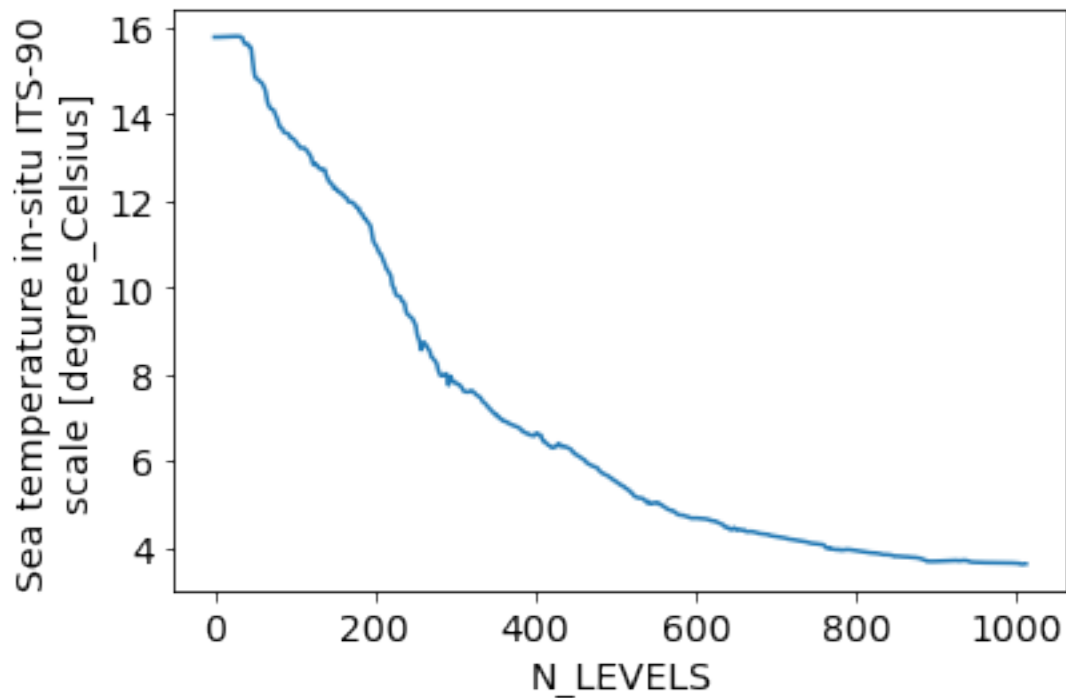
| | | |
|--------------------------|-------------|---------|
| HISTORY_SOFTWARE | (N_HISTORY) | object |
| HISTORY_SOFTWARE_RELEASE | (N_HISTORY) | object |
| HISTORY_REFERENCE | (N_HISTORY) | object |
| HISTORY_DATE | (N_HISTORY) | object |
| HISTORY_ACTION | (N_HISTORY) | object |
| HISTORY_PARAMETER | (N_HISTORY) | object |
| HISTORY_START_PRES | (N_HISTORY) | float32 |
| HISTORY_STOP_PRES | (N_HISTORY) | float32 |
| HISTORY_PREVIOUS_VALUE | (N_HISTORY) | float32 |
| HISTORY_QCTEST | (N_HISTORY) | object |

Attributes:

| | |
|----------------------|---|
| title: | Argo float vertical profile |
| institution: | FR GDAC |
| source: | Argo float |
| history: | 2021-05-13T14:27:52Z creation |
| references: | http://www.argodatamgt.org/Documentation |
| user_manual_version: | 3.1 |
| Conventions: | Argo-3.1 CF-1.6 |
| featureType: | trajectoryProfile |

```
[8]: ds_ARGO.TEMP.plot()
```

```
[8]: [<matplotlib.lines.Line2D at 0x7fd8506fbad0>]
```

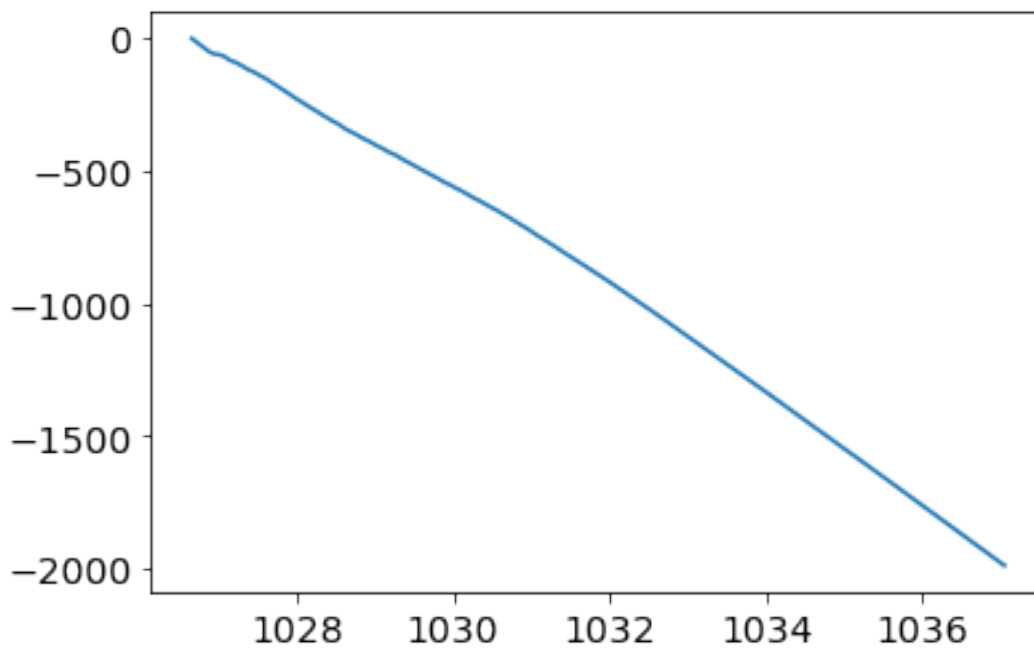


```
[9]: # Convert practical to absolute salinity
SA = gsw.SA_from_SP(ds_ARGO.PSAL, ds_ARGO.PRES, ds_ARGO.LONGITUDE, ds_ARGO.
    ↳LATITUDE)
# Convert in-situ to conservative temperature
CT = gsw.CT_from_t(SA, ds_ARGO.TEMP, ds_ARGO.PRES)
# Calculate the density
rho = gsw.rho(SA, CT, ds_ARGO.PRES)
```

```
[10]: # Calculate the vertical coordinate from pressure
z = gsw.z_from_p(ds_ARGO.PRES, ds_ARGO.LATITUDE)
```

```
[11]: # Make a plot of a density profile
plt.plot(rho, z)
```

```
[11]: [ <matplotlib.lines.Line2D at 0x7fd8506eadd0>]
```



1.5.1 Show the position of the profile on a map

Check out <https://scitools.org.uk/cartopy/docs/latest/crs/projections.html> for a list of available map projections.

```
[12]: ax = plt.axes(projection=ccrs.Mercator())
ax.gridlines()
ax.coastlines()
```

```
ax.set_global()

ax.plot(ds_ARGO.LONGITUDE, ds_ARGO.LATITUDE, "ro", transform=ccrs.PlateCarree())
```

[12]: [<matplotlib.lines.Line2D at 0x7fd83d59ea10>]



1.5.2 Add features to the map

```
[13]: fig = plt.figure(figsize=(10, 10))
ax = fig.add_subplot(projection=ccrs.EuroPP())

ax.coastlines(resolution='50m')
lakes = cfeature.NaturalEarthFeature(
    category='physical',
    name='lakes',
    scale='50m',
    edgecolor='none',
    facecolor=cfeature.COLORS['water'],
)
rivers = cfeature.NaturalEarthFeature(
    category='physical',
    name='rivers_lake_centerlines',
    scale='50m', # can also be '10m' (more details) or '110m' (less details)
    edgecolor=cfeature.COLORS['water'],
    facecolor='none',
)
ax.add_feature(lakes)
```

```
ax.add_feature(rivers)
```

```
[13]: <cartopy.mpl.feature_artist.FeatureArtist at 0x7fd83d461250>
```



```
[ ]:
```