

# Accessing GLOFAS data using the Climate Data Store

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Climate Change











## The Climate Data Store Objectives

# The Copernicus Climate Change Service (C3S) provides information to support adaptation and mitigation policies



Make data discovery and access simple and relevant



Provide online capabilities to process the data and develop easy-to-use applications



Enable reproducible research



Spend less time handling the data







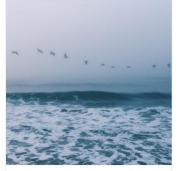


### What kind of data?









#### Observations

Observations are key to understanding the climate system. C3S users can access a vast variety of instrumental data records, ranging from historic weather observations to the latest measurements from space.

Read more



Climate reanalyses combine past observations with models to generate consistent time series for a large set of climate variables.

Reanalyses are among the mostused datasets in the geophysical sciences.

Read more >

Reanalysis data on the >
CDS

#### Seasonal forecasts

C3S seasonal forecasts combine outputs from several state-of-theart seasonal prediction systems from providers in Europe and elsewhere. The latest data and products are published monthly on the Climate Data Store.

Read more >

Seasonal forecast data >
on the CDS

#### Climate projections

Projections of future climate change are available for different scenarios for concentrations of greenhouse gases and aerosols, based on outputs from multiple global and regional climate models.

Read more >

Climate projection data >
on the CDS







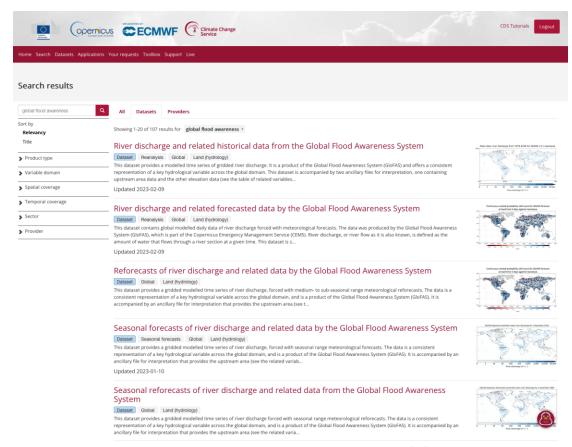
# Concept

The CDS is a **distributed system** which provides access to externally hosted datasets through a unified web interfaces DATA SUPPLIERS INTEROPERABILITY Climate Data Store Infrastructure DATA Petabytes INFORMATION





### Example catalogue entries



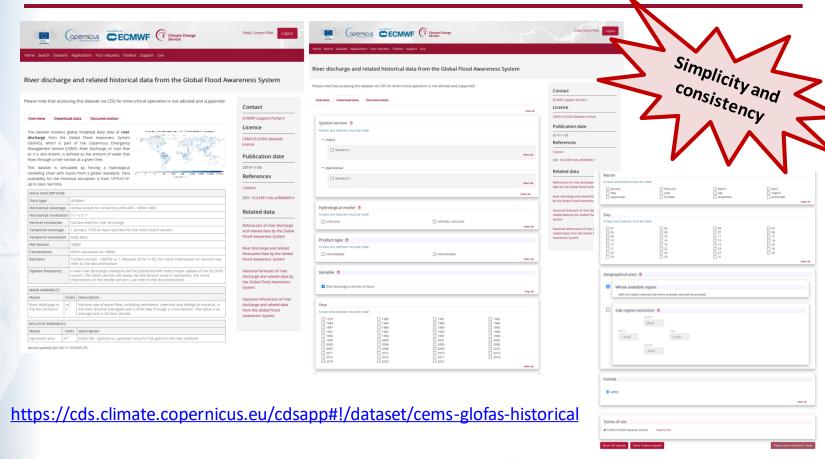








Example catalogue entry (GLOFAS Historical)











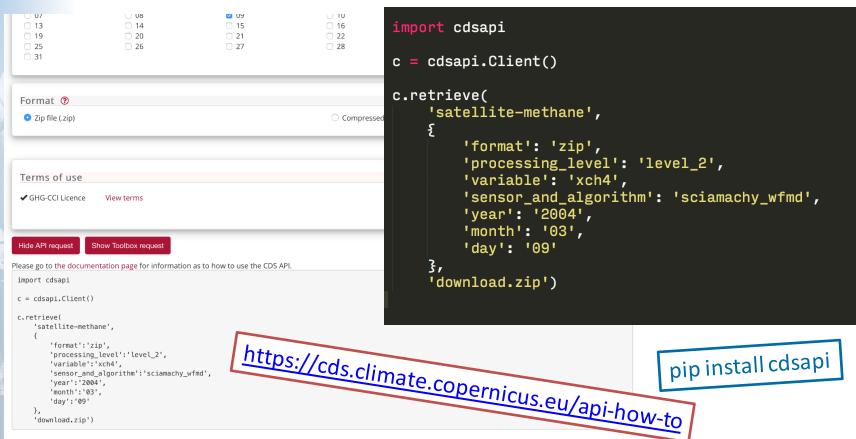
Change

'variable':'xch4',

'year':'2004', 'month':'03', 'day':'09' 'download.zip')

'sensor and algorithm': 'sciamachy wfmd',

## Example catalogue entry (GLOFAS Historical)



pip install cdsapi

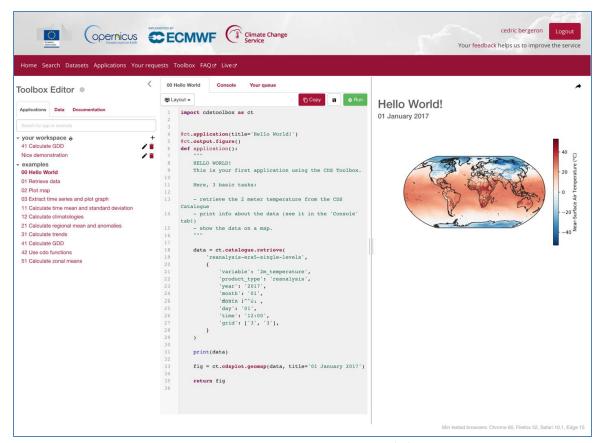








# Online processing











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## Training Material

## https://ecmwf-projects.github.io/copernicus-training-c3s/intro.html



#### 3. Anomaly calculation

The next step is now to calculate the anomaly of a specific year with respect to the climate normal. The term anomaly refers to the deviation of a value from the long-term average. Positive or negative anomalies indicate that the average temperatures of a particular year were respectively warmer or cooler than the reference value.

Let us calculate the near-surface air temperature anomaly for the year 2016. In a first step, we select the average near-surface temperature values for the year 2016 from the xarray DataArray object year ty near.

With the xarray function set (1), you can select a data array based on coordinate labels. The coordinate label of interest is xero 1016.

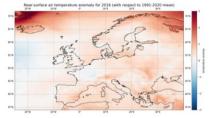
t2m\_2016 = yearly\_mean.sel(year=2016)

Next, we calculate the near-surface air temperature anomaly for 2016 by subtracting the climate normal (i.e. the reference near-surface air temperature values) from the average near-surface air temperature for 2016.

anon\_2016 = t2n\_2016 - ref\_mean

Let's visualize the global near-unface air temperature anomaly for 2016 to see which regions were warmer or cooler compared to the reference period. This time we will make use of a combination of the plotting libraries matglotidh and Cartory to create a more customised figure. One of Cartory's key features is fits ability to smarform array data into different geographic projections. In combination with matglotib, it is a very powerful were orates high-quality visualisations and administions.

s create the figure panel and the map using the Cortagy PlateCorres projection fig. as = pit.ngbistit. 1, figure (fig. 8), suppl.-be("pig)climin" carriplateGarree(1)) a Fig. to deta in = pit.spc.imree(1) as a pit.spc.imree(1) as pit.spc.imree(1)



#### Seasonal analysis of Arctic near-surface air temperature

In the final part of this totorial we will compare seasonal trends in Arctic near-surface air temperature. To do this we return to our monthly geographically were apple dataset, and we will downsample the monthly averages to seasonal averages using the function resumple;1. By specifying 11se 165-66°C, the data is split into consecutive three-month periods, sunchored at December; if we add additionally the function mean? In we obtained the average of the three-month period.

Arc\_seasonal = Arc\_mean.resample(time='QS-DEC').mean()

In the code below, before visualising the seasonal average air temperature data for the Arctic, we create for each season (vinter, spring, summer and autum) a pandas. DataFrame. We then create a for loop to visualise each seasonal time series in a successive plot.



Arctic seasonal averages of 2m air temp - 1979-2020



Note the difference in variability of the seasonal average of air temperature in the Arctic: mean summer temperatures seem to be more constant compared to the other seasons.



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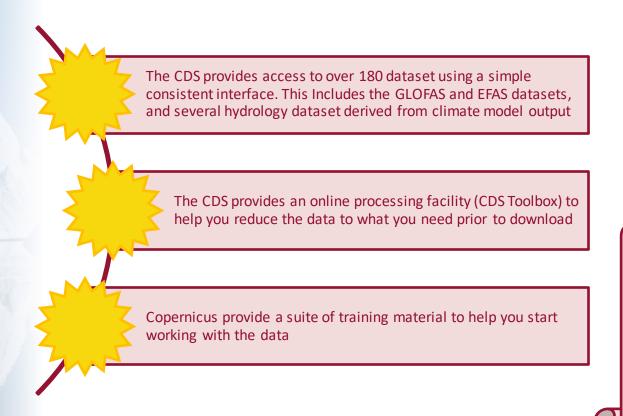








## Summery Summary



Please visit me
at the CDS
stand in
Gathertown for
some hands on
demonstrations





