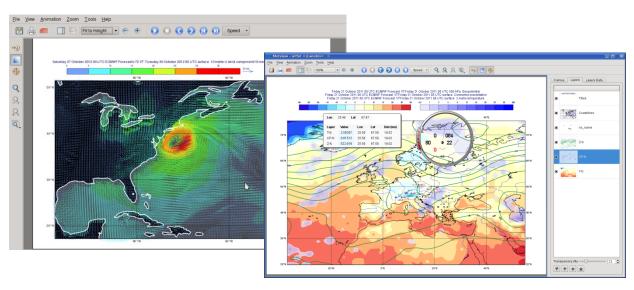
Data Handling with Metview



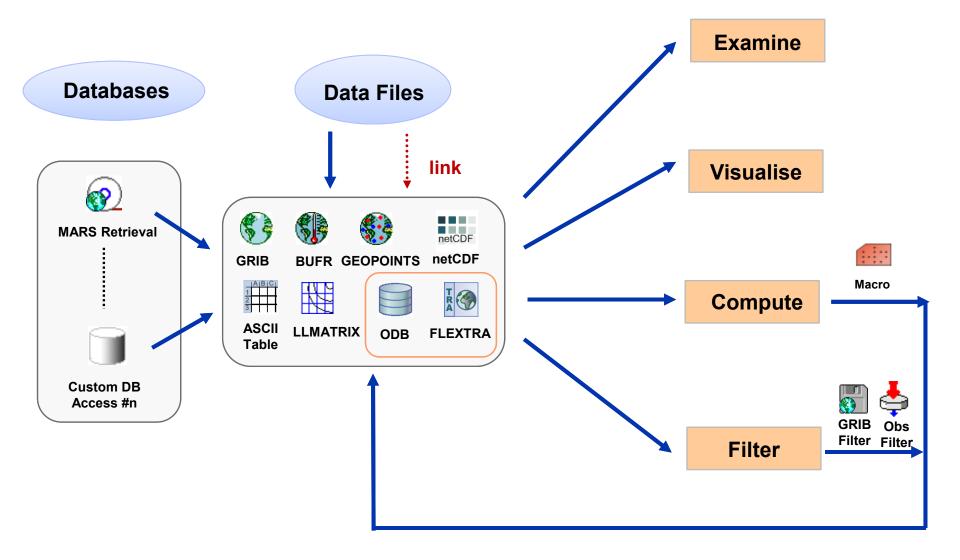


Development Section **ECMWF**



Data handling in Metview













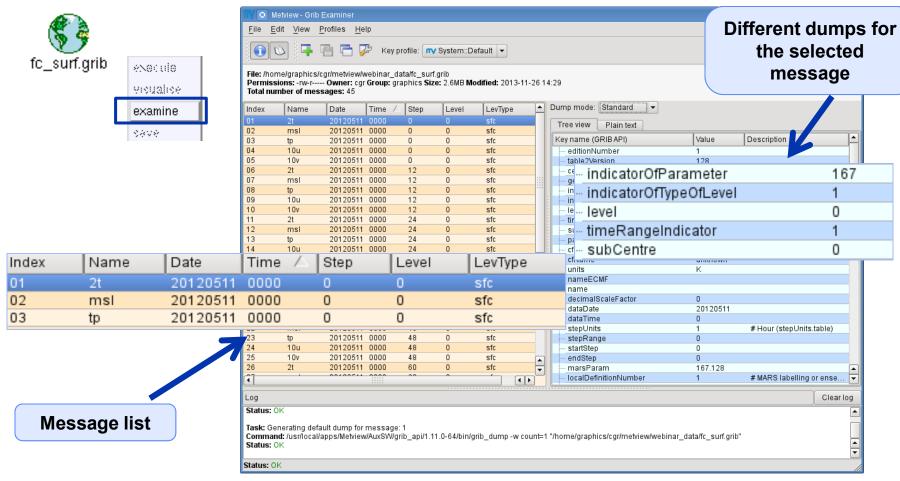
- WMO's binary format for gridded data
- The Metview interface is based on GRIB API
- Access to both Edition 1 and 2 files





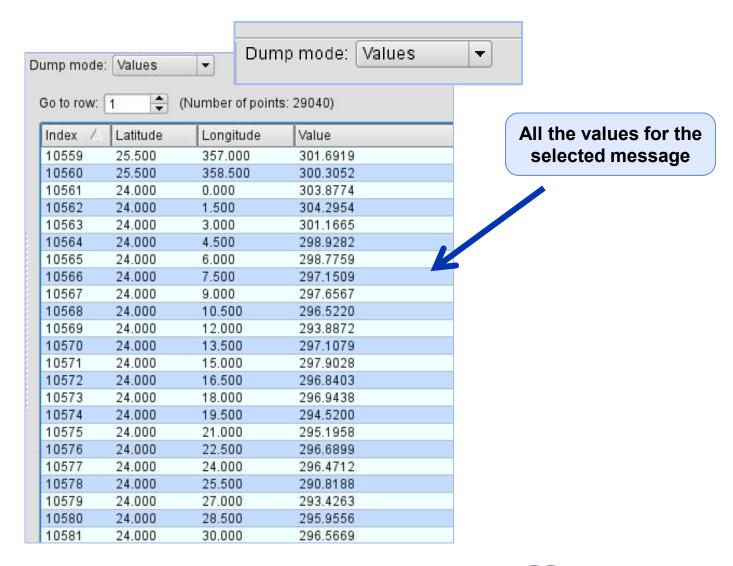


GRIBs contents can be checked with the GRIB Examiner





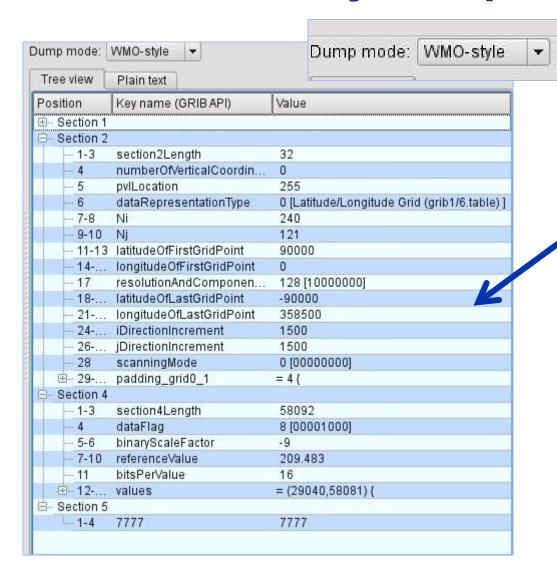
GRIB Examiner - Values dump







GRIB Examiner – WMO-style dump



Each section of the GRIB message is shown in a tree view





GRIB Examiner – Namespace dump

geography.longitudeOfLastGridPointInDegrees

Namespace is a **GRIB API concept** to define GRIB API Dump mode: Namespace ▼ key categories GRIB API Namespace: | geography | Dump mode: Namespace ▼ GRIB API Namespace: geography -Value Key name (GRIB API) Key type geography.bitmapPresent long 0 geography.gridType regular_ll string geography.iDirectionIncrementInDegrees double 1.5 geography.iScansNegatively 0 long geography.jDirectionIncrementInDegrees double 1.5 geography.jPointsAreConsecutive long 0 geography.jScansPositively long 0 90 geography.latitudeOfFirstGridPointInDegrees double **Namespace** -90 geography.latitudeOfLastGridPointInDegrees double geography groups geography.longitudeOfFirstGridPointInDegrees double 0 keys describing



the grid resolution, area etc. GRIB API concept

358.5

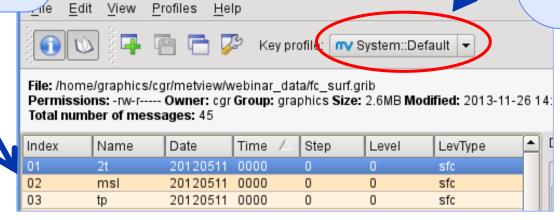
double



GRIB Examiner – Key profiles

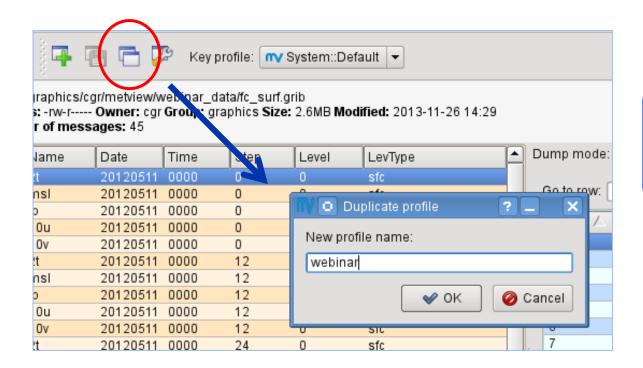
The message list is presented using a set of GRIB API keys. A group of these keys is called a key profile.

The default profile is called System::Default. It is read-only! However you can create any number of additional profiles.





GRIB Examiner – Create a new key profile

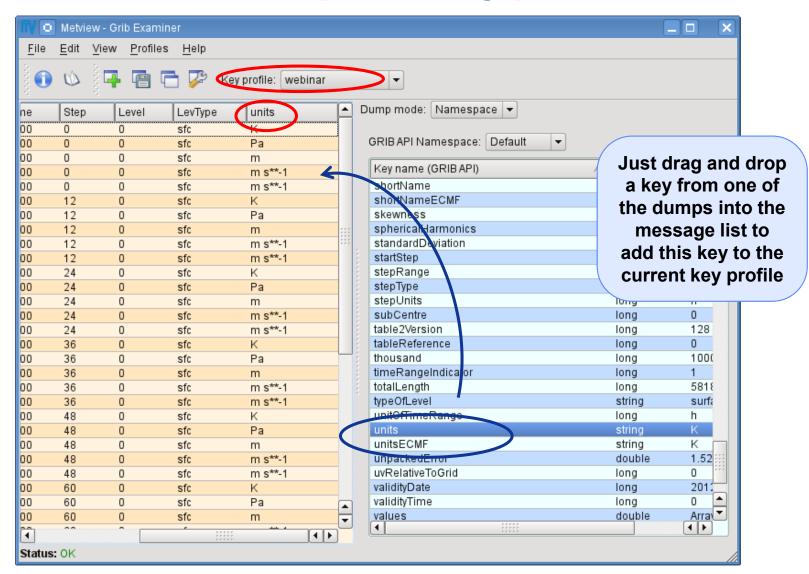


The easiest way to create a new key profile is to duplicate an existing one





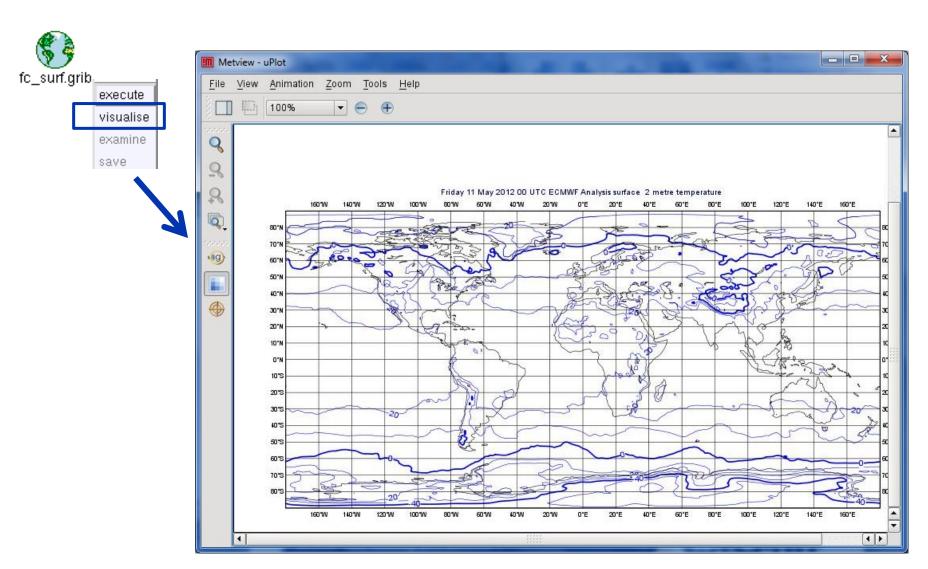
GRIB Examiner – Populate key profiles







GRIB plotting







Overlaying fields from the same GRIB file

Example: overlay T2 and MSLP forecasts from file fc_surf.grib

- We need to filter out each parameter into a separate file
- We will use the GRIB Filter icon

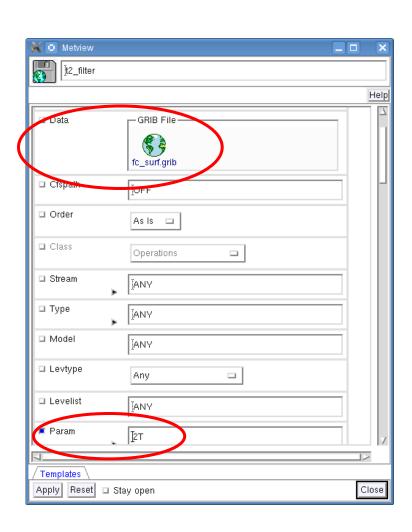


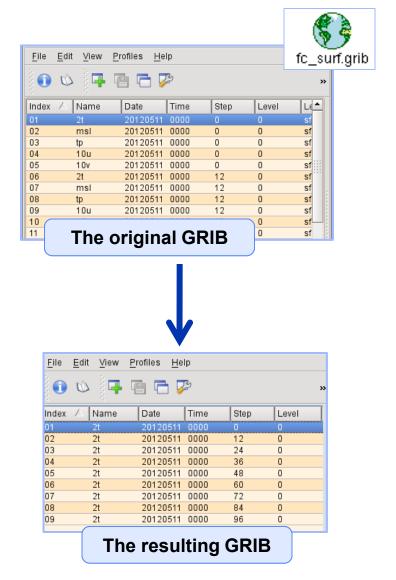
- It allows filtering according to parameter, date, time, level etc.
- It caches the results (name turns green) and can be used directly in the same way as GRIB icon





GRIB Filter: Parameter selection

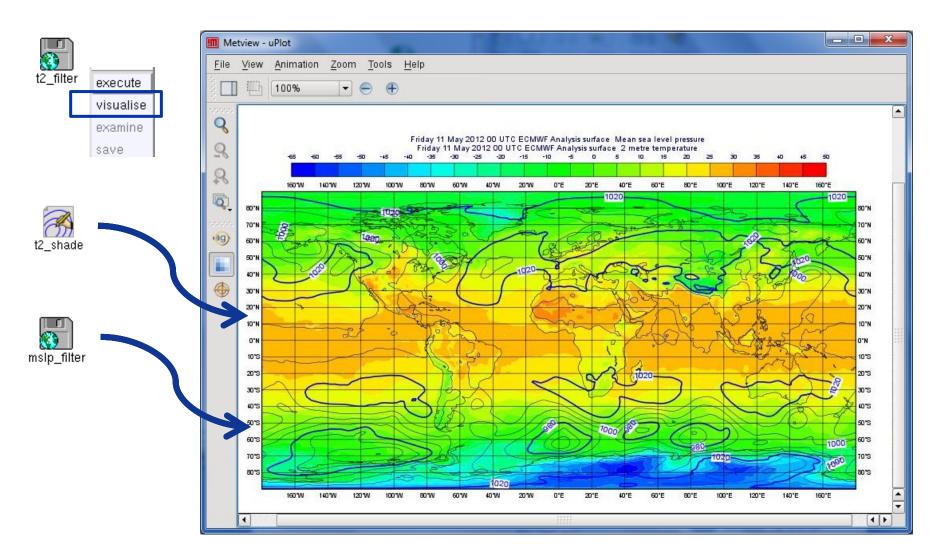








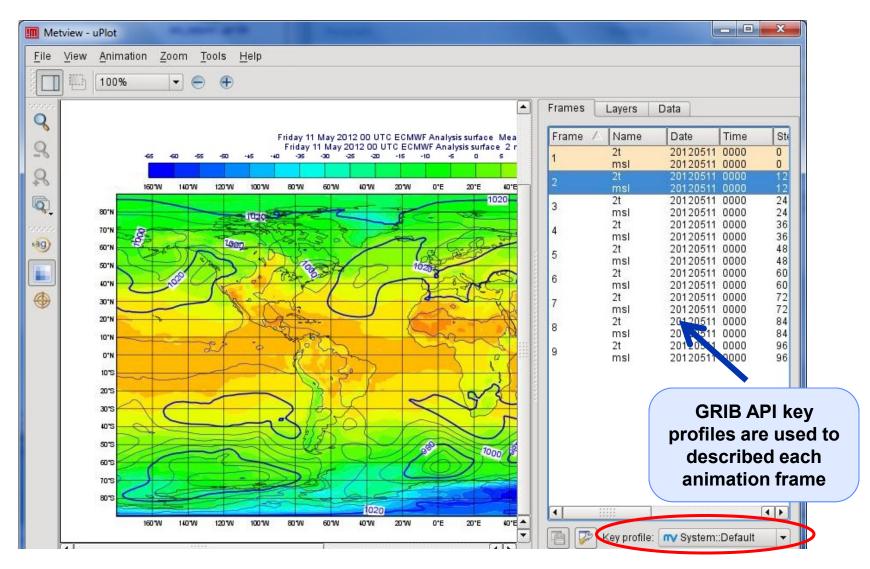
Overlaying GRIB fields







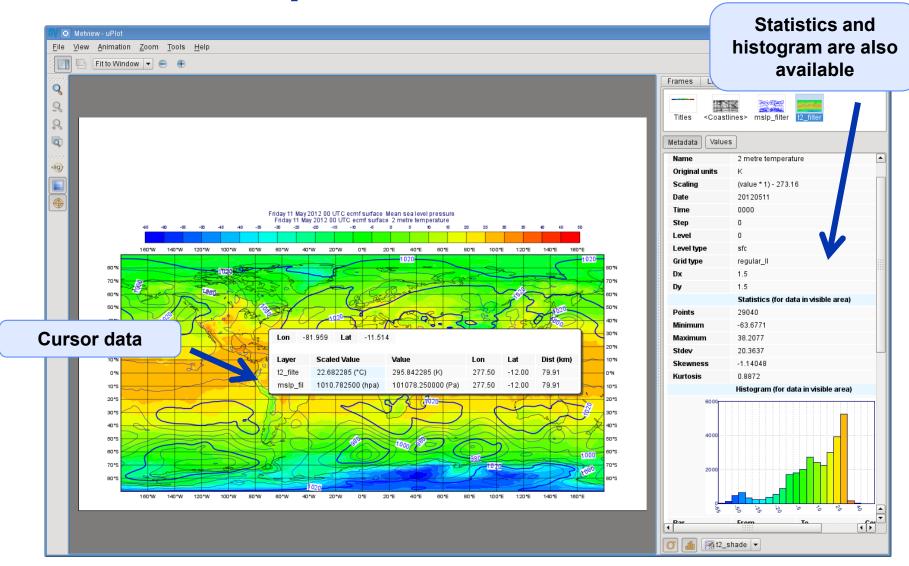
Overlaying GRIB fields







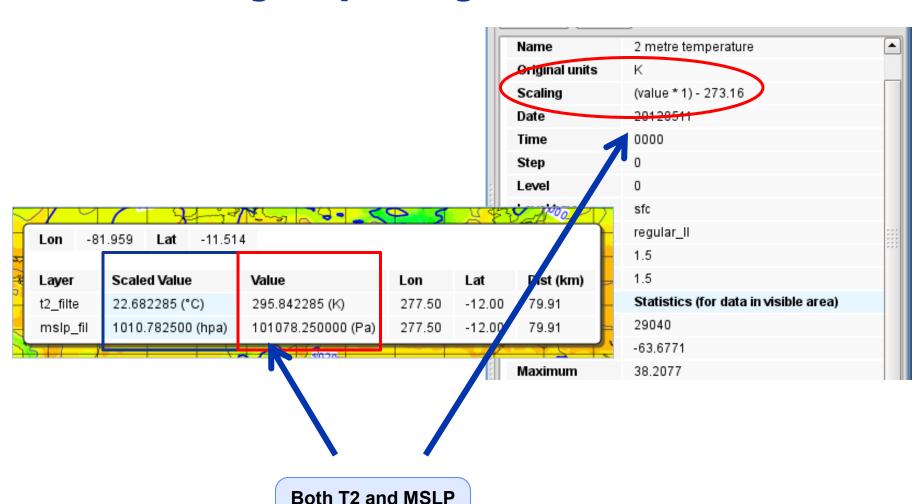
GRIB data inspection







GRIB scaling for plotting





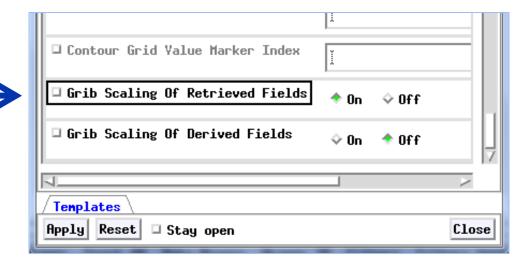
are scaled



GRIB scaling for plotting



This parameter tells Metview to apply scaling for certain fields for contouring



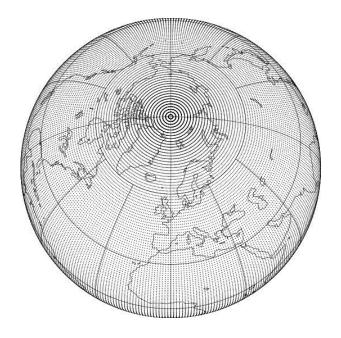




Other usage of GRIB Filter: interpolation

- Spherical harmonics to gridpoint transformation
- Interpolation between different grids
 - Regular Gaussian grid
 - Reduced Gaussian grid
 - lat-lon grids etc.
- Currently it is based on EMOS lib









How to use the interpolation?

Example: compute the difference between two different resolution T500 fields



Forecast 1.5x1.5 global grid



Analysis 0.25x0.25 limited area grid

- The steps involved:
 - 1. Filter T500 for the matching date and time
 - 2. Interpolate the global field to the LAM grid
 - 3. Compute the difference



We will write a macro!





Macro: Compute difference #1

Filter parameter and level from analysis. In Macro the GRIB Filter is invoked via command: read

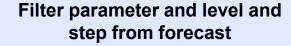
```
1 #Metview Macro
3 #Read T500 analysis
4 g an=read(source: "an_upper.grib",
            param: "t",
            level: 500)
8 #Read target area borders
 9 lat1=grib_get_double(g_an, "latitudeOfFirstGridPointInDegrees") #north
10 lat2=grib_get_double(g_an, "latitudeOfLastGridPointInDegrees") #south
  lon1=grib_get_double(g_an, "longitudeOfFirstGridPointInDegrees") #west
 lon2=grib_get_double(g_an,"longitudeOfLastGridPointInDegrees") #east
4 #Read target grid resolution
5 dx=grib_get_double(g_an, "iDirectionIncrementInDegrees")
6 dy=grib get double(g an, "jDirectionIncrementInDegrees")
```

Here we read a set of GRIB API keys





Macro: Compute difference #2



```
18 #Read T500 forecast for the analysis date (+96h)
19 #and interpolate it to the target grid
20 q_fc=read(source: "fc_upper.grib",
21
             param: "t",
22
             level: 500.
23
             step: 96,
24
             area: [lat2,lon1,lat1,lon2], #[s,w,n,e]
25
             grid: [dx,dy] )
26
27 #Compute difference
28 g_res=g_fc-g_an
30 #Return results
31 return q_res
```

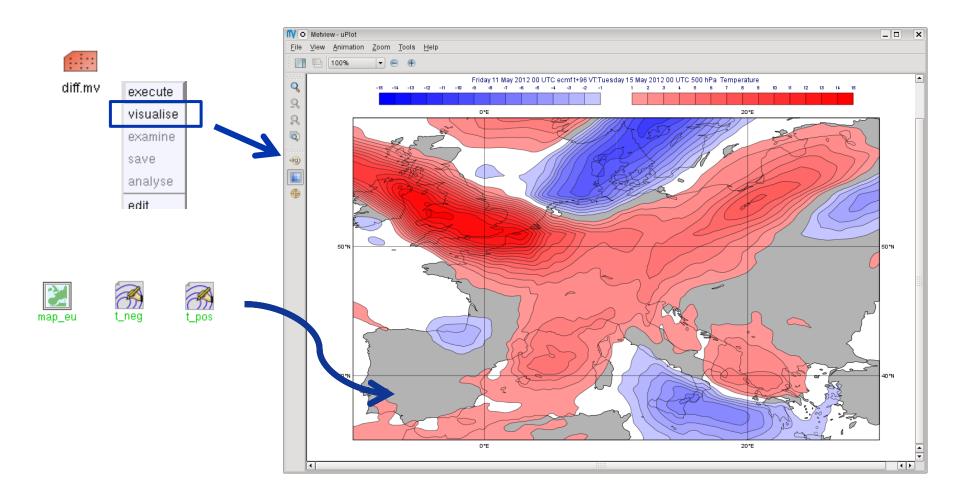
Difference operator only works between grids with the same number of points

Interpolation





Macro: Compute difference #2







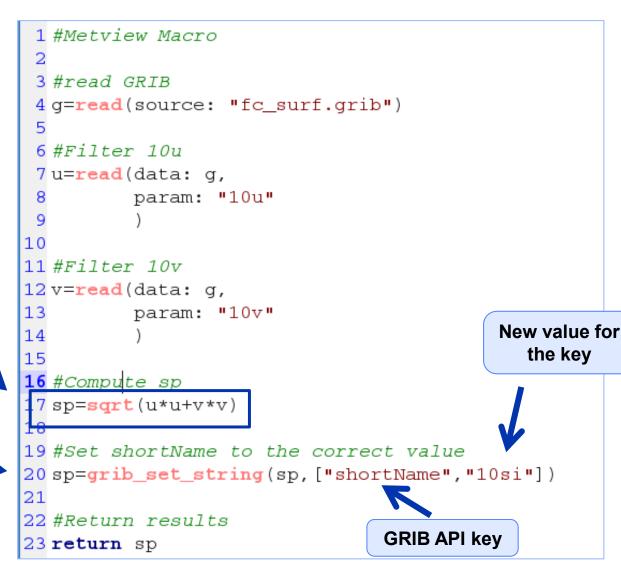
Macro usage: compute wind speed

Example:

compute 10
windspeed from u
and v
components

Fieldset operation

Here we set the GRIB header



Macro usage: compute precipitation for intervals

- Precipitation is often stored as an accumulated quantity
- We want to see precipitation for a given interval (e.g. 12h, 24h)

```
Example: compute
                               1 #Metview Macro
precipitation for 12 h
                                                                         tp.mv
 intervals from file
                               3 #Read tp fields from the GRIB file
     fc_surf.grib
                               4 g=read(
                                        source: "fc_surf.grib",
                                        param: "tp"
        We filter tp
                               9 #Number of tp mesages
                              10 num=count (q)
        Operation
      between slices
                             12 #Compute tp for 12h intervals
        of fieldsets:
                              13 gr = g[2, num] - g[1, num - 1]
       we compute
                              14
       the difference
                             15 #Return results
        of adjacent
                             16 return gr
          fields
```





Precipitation computation explained

g

f1 f2

f3

.....

fnum



Our original fieldset has num fields

```
12 #Compute tp for 12h intervals
13 gr=g[2,num]-g[1,num-1]
14
```

g[2,num]

f2

f3

.....

fnum

g[1,num-1]

f1

f2

.....

fnum-1

g[2,num]g[1,num-1] f2-f1

f3-f2

......

fnum-fnum-1



Macro usage: more functions

- A rich set of macro functions exists for GRIB. A few examples:
 - latitudes(), longitudes(), values(): read the latitudes, longitudes and values of a field into vectors (in-memory arrays)
 - average(): compute average
 - mask(): set field values to 0 or 1 using an area mask
 - bitmap(): assign missing values to a field using a mask
 - nobitmap(): replace missing values
- See Macro Tutorial 3 for some elaborated examples, such as masking one field based on the values of another (e.g. apply a land sea mask to a field to remove (i.e. to bitmap) points over sea)





Complex plot types for GRIB

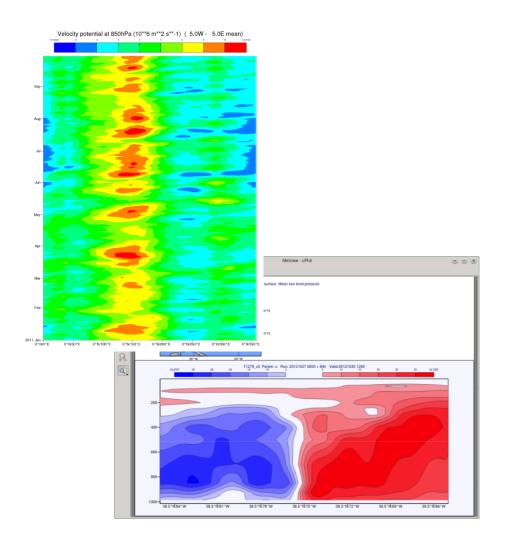
- These plots require data extraction from multiple fields and some computations as well
- There are a set of GRIB specific icons to generate:
 - Cross sections
 - Hovmøller diagrams
 - Zonal mean plots
 - Vertical profiles









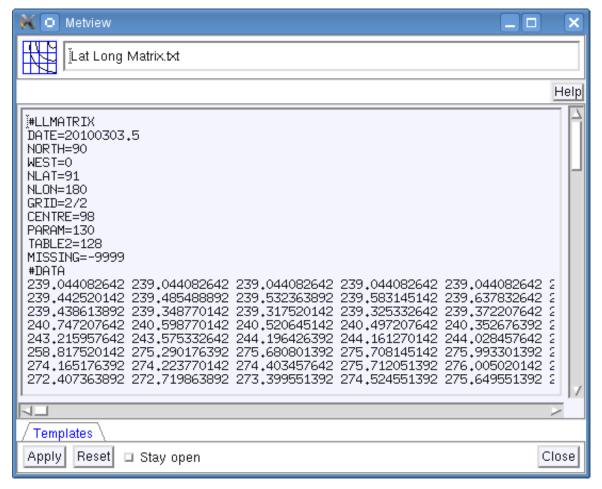




Lat Long Matrix



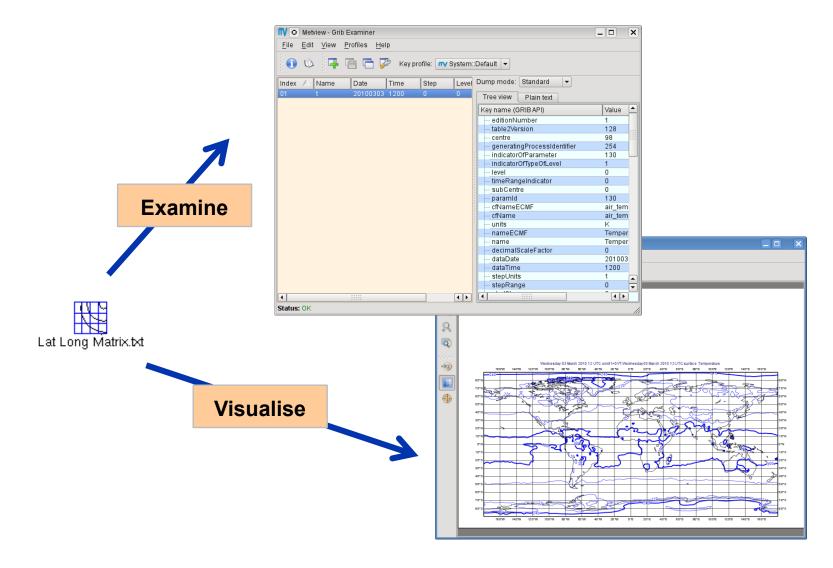
- Metview's ASCII format for gridded data
- Turned into GRIB internally
- Can be edited as a text file







Lat Long Matrix - Behaves like a GRIB





BUFR





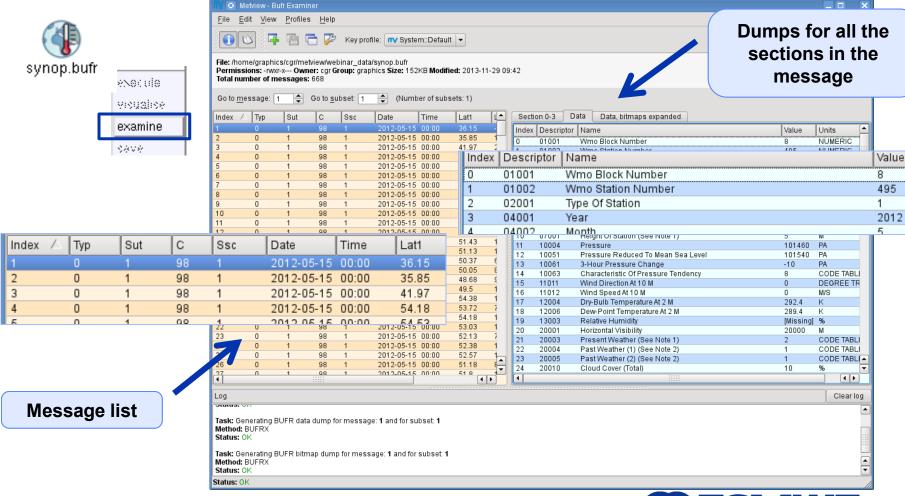
- WMO's binary format for observation data
- Metview offers a high level interface to work with BUFR
- Internally we use BUFRDC (part of EMOS lib) to decode BUFR messages
- There is a BUFR tutorial available on the Metview web page





BUFR Examiner

BUFRs contents can be checked with the BUFR Examiner

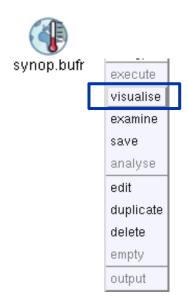


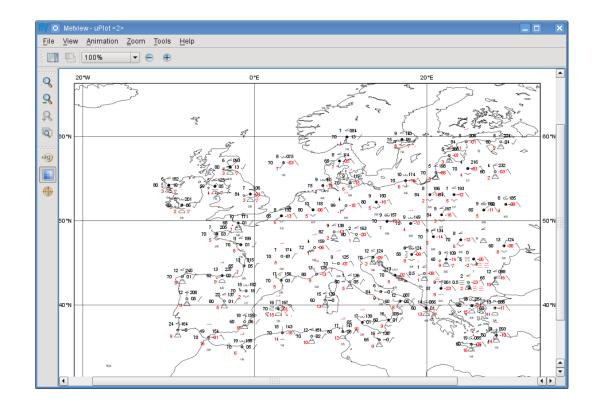






 We can directly visualise BUFR files with conventional observations (e.g. SYNOP)









BUFR: Accessing data

Example: extract and plot T2 with symbol plotting from file synop.bufr

We need to use the Observation Filter icon

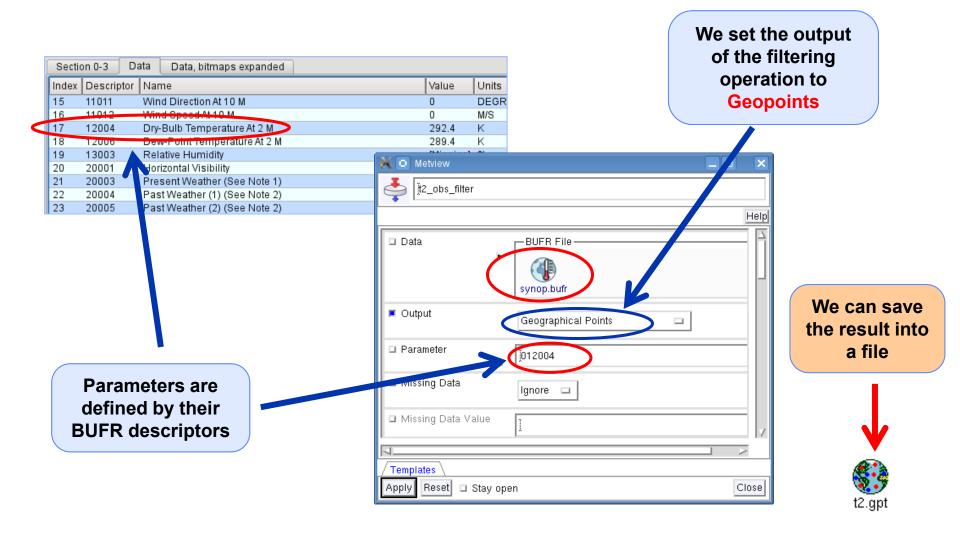


 It can perform filtering according to parameter, level, area, time, channel etc.





BUFR: Filtering





Geopoints





- Metview's custom format to store scattered geo-referenced data
- ASCII files with 4 different types: The default is shown here:

#GEO					
PARAMETER = 12004					
lat	long	level	date ti	ime	value
#DATA					
36.15	-5.35	0	20120515 0	0000	292.4
35.85	14.48	0	20120515	0000	288.8
41.97	21.65	0	20120515	0000	282.4

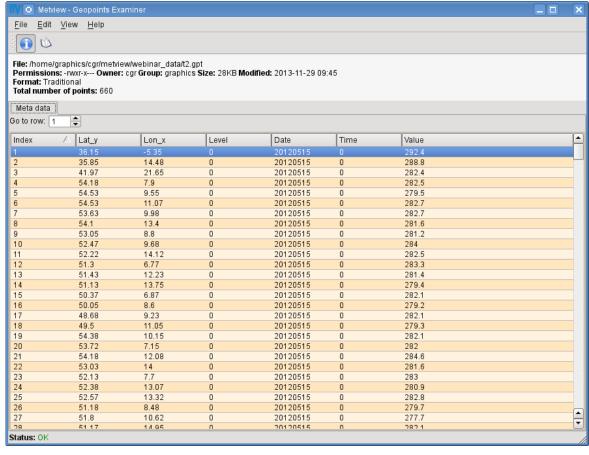


Geopoints Examiner

- Geopoints contents can be checked with the Geopoints Examiner
- This is how the result of the BUFR filtering looks like











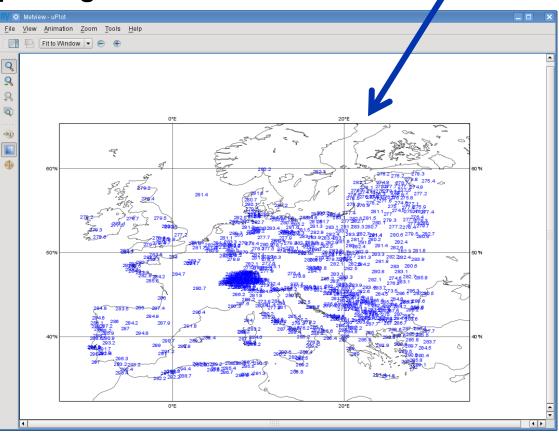
Geopoints Plotting

We can directly visualise Geopoints

It is based on symbol plotting

By default the numbers are plotted to the map



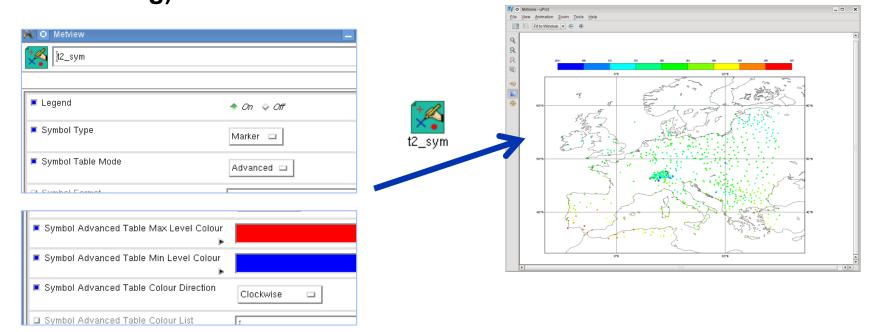






Customisation with Symbol Plotting

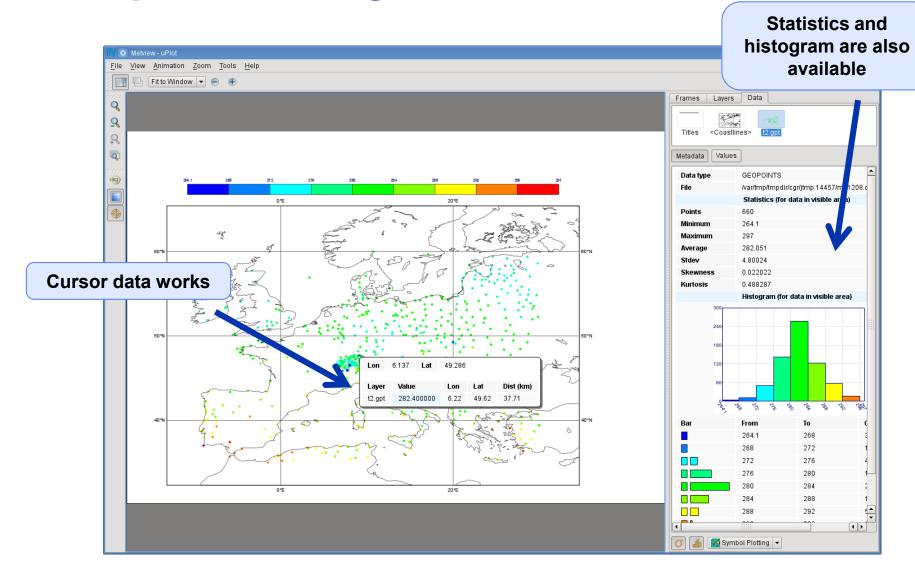
- The Symbol Plotting icon offers a large number of options for plot customisation
- We can use the Advanced Table Mode to define a nice colour palette between the min and max colours (just like for Contouring)







Geopoints Plotting





Macro: difference between GRIB and Geopoints

Example:

compute the difference between the T2 forecast and observations

This step involves interpolation of the GRIB data to the Geopoints locations

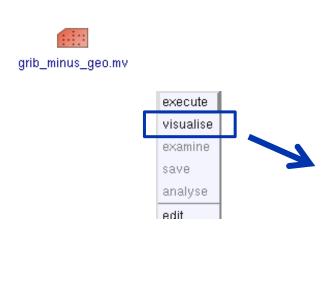
The result is another Geopoints

```
1 #Metview Macro
 3 #Read t2 field forecast (96h) from the GRIB file
 4 g=read(
          source: "fc_surf.grib",
         param: "2t",
          step: 96
10 #Read observations from gepoints
11 gpt=read("t2.gpt")
12
13 #Compute the difference
14 res=q-apt
15
16 #Return results
17 return res
```

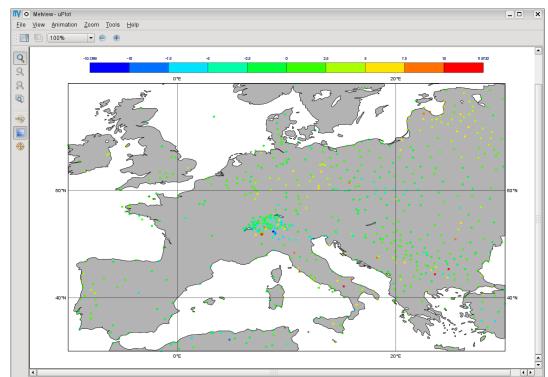




Compute difference between GRIB and Geopoints



Forecast minus observation differences







Geopoints to GRIB

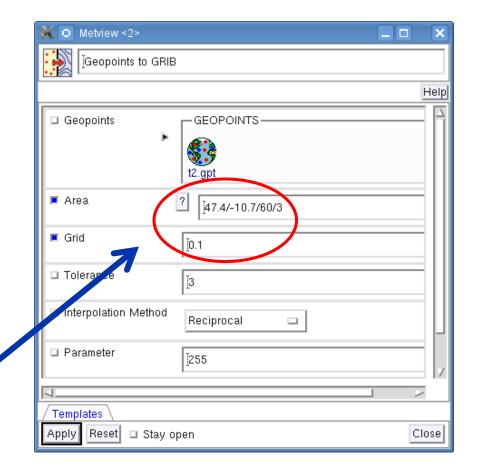
Example: interpolate T2 observations onto a grid then apply contouring

 We need to use the Geopoints to GRIB icon



 This icon interpolates
 Geopoints data onto a regular lat-lon grid and encodes it into GRIB

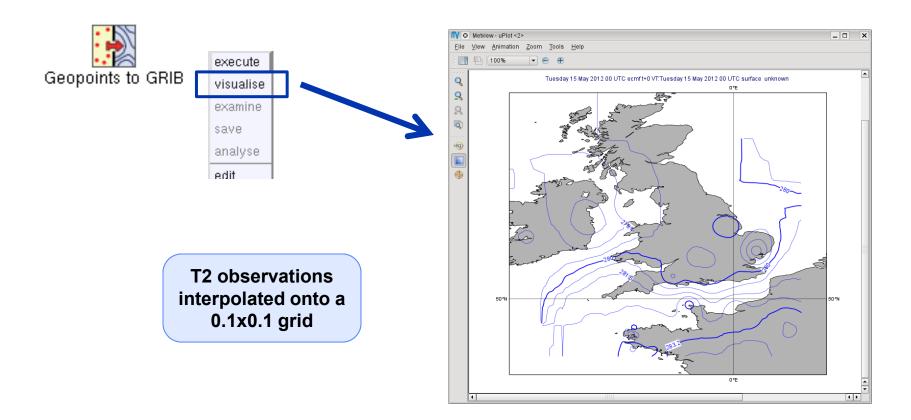
The grid definition













NetCDF



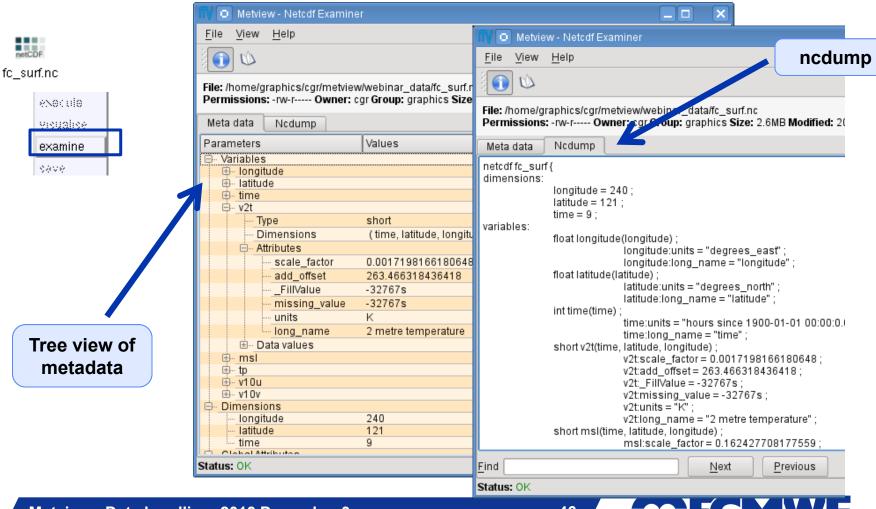


- UNIDATA's binary format for multidimensional arrays
- Metview's NetCDF plotting interface was added a few years ago





NetCDF contents can be checked with the NetCDF Examiner





NetCDF: How to plot it?

- NetCDF is so flexible it can contain almost any kind of data
- We need to use the NetCDF Visualiser icon



It defines the way variables/dimensions are used for plotting





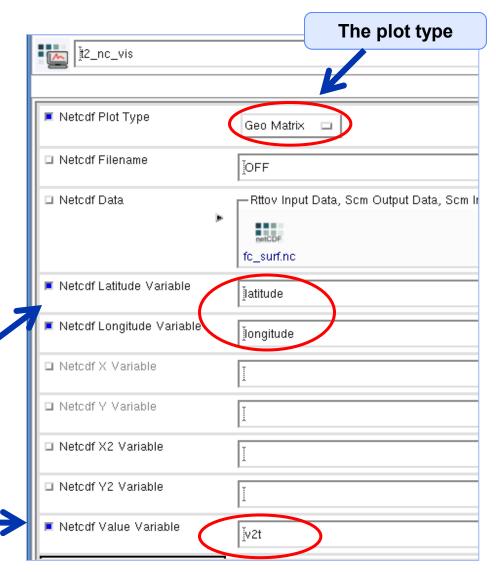
Plotting NetCDF data

Example:
plot T2 from
file
fc_surf.nc



Latitude and longitude variables

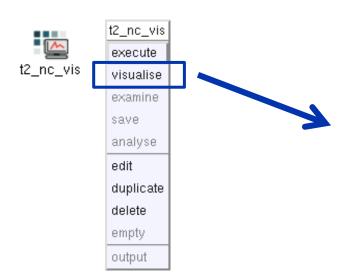
The variable to plot (T2 is called v2t in our file)



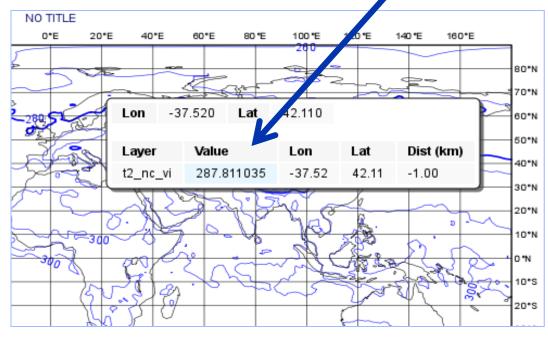




NetCDF: Plotting



No scaling applies for NetCDF values: we have values in Kelvins







NetCDF: Macro Usage

nc_K_to_C.mv

Example: convert values of T2 from Kelvin to Celsius

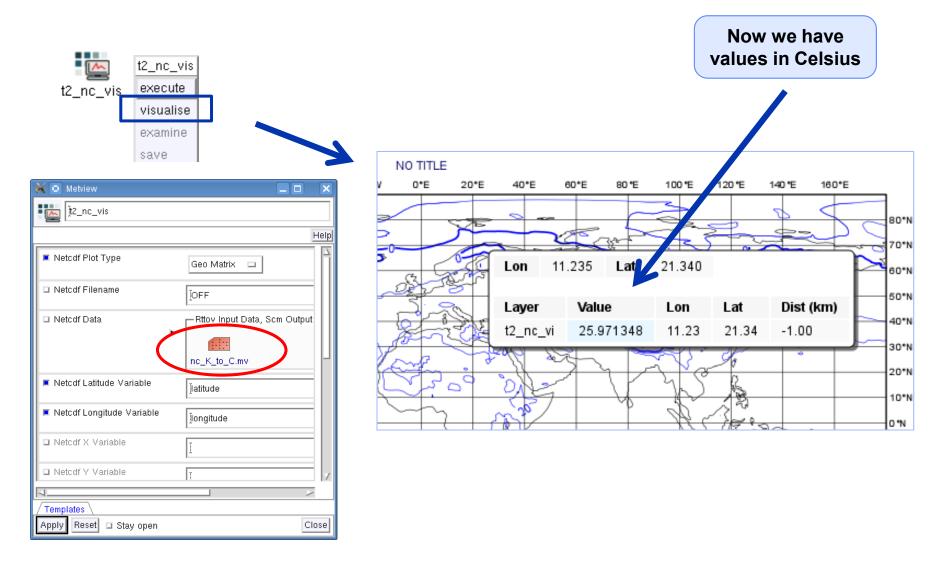
 The NetCDF macro interface is based on the current variable concept: all operations are only valid to the currently selected NetCDF variable!

```
1 #Metview Macro
 3 #Read netcdf file
 4 nc=read("fc_surf.nc")
 6 #Get the list of netcdf variables
 7 var list = variables(nc)
 9 #Find index for t2
10 idx=find(var list, "v2t")
12 #Set the current variable to t2
13 setcurrent (nc, idx)
14
15 #Change the values of the current variable
16 \, \text{nc} = \text{nc} - 273.16
18 #Return results
19 return nc
```





NetCDF: Plotting the modified data









	ĮΑ	В	С
1			
2			
3			

- ASCII file with data arranged with one variable per column
- Can contain a header
- CSV files can be handled as Table Data
- Geopoints files can be treated as Table Data as well



```
latitude, longitude, fc, an, fc-an 90,0,-30.29,-25.81,4.48 90,4,-30.29,-25.81,4.48 90,8,-30.29,-25.81,4.48 90,12,-30.29,-25.81,4.48 90,16,-30.29,-25.81,4.48 90,20,-30.29,-25.81,4.48 90,24,-30.29,-25.81,4.48
```







Table Data plotting is based on the Table Visualiser icon



It defines the way columns are used for plotting



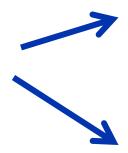


1 4

Example: plot the forecast values from file data.csv



We need to tell the visualiser which columns should be used from the file



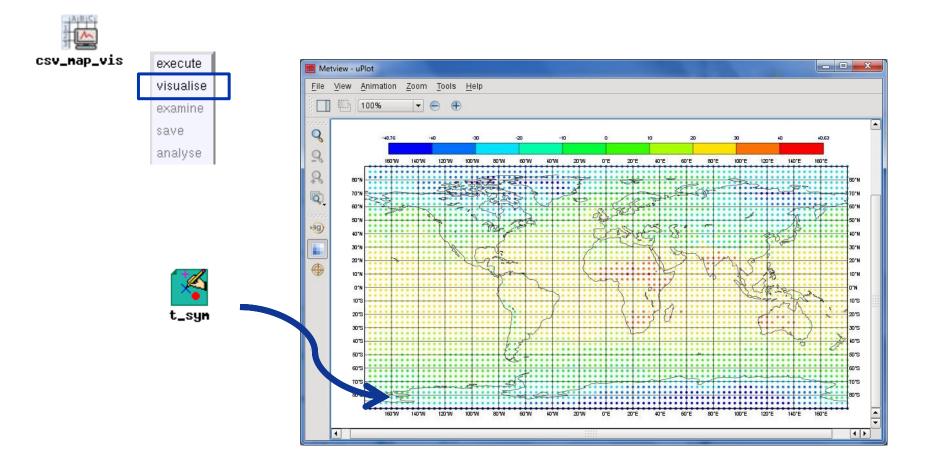
Metview	The plot type
csv_map_vis	
■ Table Plot Type	Geo Points
□ Table Filename)OFF
□ Table Data	Notes, GEOPOINTS, Table-

■ Table Yariable Identifier Type	Index 🗆
■ Table Longitude Variable	½
■ Table Latitude Yariable)ı
□ Table X Component Variable	Ĭ.
□ Table Y Component Variable	Y
■ Table Yalue Yariable	3





Plotting Table data





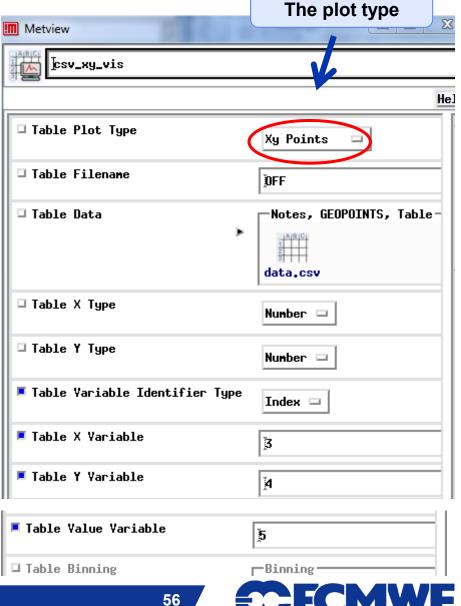


Scatterplots from Table data

Example: generate a scatterplot from file data.csv with forecast in X axis and analysis in Y axis, and values (for colouring) taken from fcan.

We need to tell the visualiser which columns should be used for X, Y and value







Scatterplots from Table data

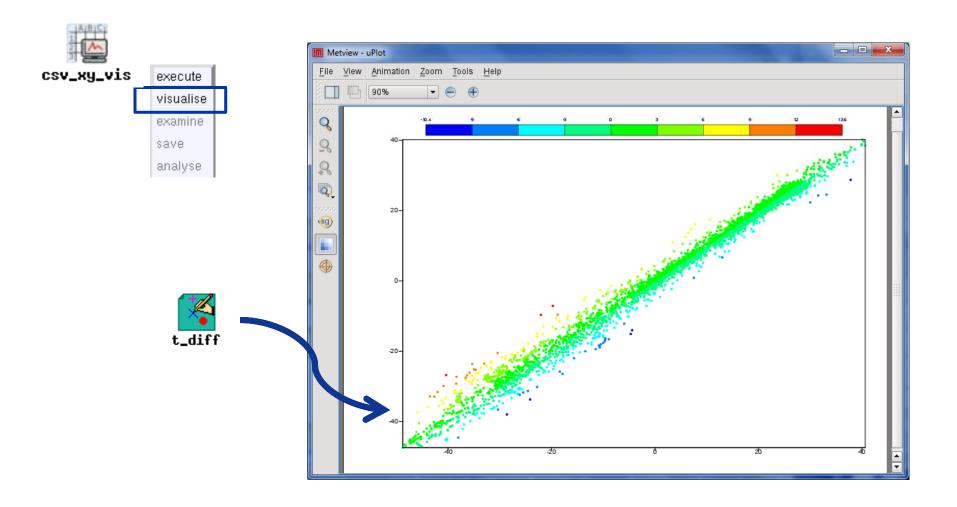






Table Data: macro usage

Example: compute the mean of the forecast-analysis values (5th column) from file data.csv

1 #Metview Macro 3 #Read netcdf file 4 t=read_table(table_filename: "data.csv") 6 #Read the fc-an column into a vector 7v = values(t, "fc-an") 8 #could be v=values(t,5) as well 9 #since fc-an is the fifth column 11 #Print mean 12 print("mean=", mean(v)) 13 14

The output of the macro



mean=-0.0265241545894

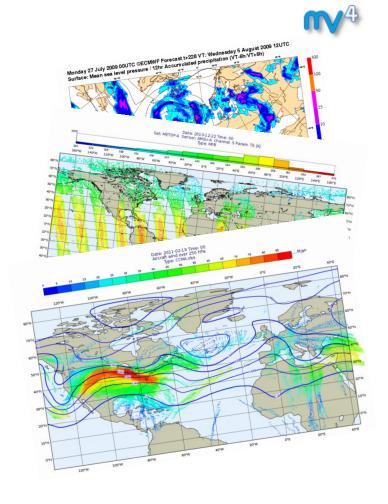


For more information ...

email us:

visit our web pages:

- https://software.ecmwf.int/metview
- Documentation and tutorials
- Download the virtual machine



Thursday, 5th December, 9:30 AM UTC: Q&A

www.hipchat.com/gRuxxenIY

