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# **tropy Documentation**

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## API Reference:

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*tropy* contains a set of tools which were developed at TROPOS to support scientific work.

Main task are:

- input of datasets
- regridding & filtering
- segmentation & optical flow applications
- calculation of derived variables

Tutorial can be found here: <<https://trophy-tutorials.readthedocs.io/en/latest/>>



## Functions Overview

### 1.1 tropy.analysis\_tools

#### 1.1.1 tropy.analysis\_tools.derived\_variables

<i>calc_lwp</i>	Calculates the liquid water path.
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#### 1.1.2 tropy.analysis\_tools.grid\_and\_interpolation

<i>create_interpolation_index</i>	Given georeference of two grids in lon and lat, an index is built to map a field on grid 1 to grid 2.
<i>curve_flow_filter</i>	Smoothing filter depending on isoline curvature.
<i>cutout_cluster</i>	Makes a cutout of a categorial field c for an object of class / number nc.
<i>cutout_field4box</i>	Cuts out a field based on center pix index and box size.
<i>cutout_fields</i>	Cuts out a field or a list of similar fields given row and column slices.
<i>get_index</i>	Given a point p the nearest grid point is returned.
<i>i2iset</i>	Returns an index set for a n-dim numpy array given an index i which points to a position within the same but flattened array.
<i>interpolate_field_to_hor_pos</i>	Given a sequence of points (or just one) a field (2d or 3d) given a grid is interpolated to the new positions.
<i>ldiff</i>	Calculates difference of a field at adjacent levels.
<i>ll2xy</i>	Transformation between longitude and latitude and local Cartesian coordinates in west-east direction (x) and south-north direction (y).
<i>ll2xyc</i>	Applies a centered version of ll2xy.
<i>lmean</i>	Calculates the layer mean of a field.

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Table 2 – continued from previous page

<i>make_add_edge</i>	Adds egde region to 2d data field.
<i>make_index_set</i>	This simple routine makes an 2d index set for given row and column number of a 2d field.
<i>make_vert_cut</i>	Makes a horizontal - vertical cut through a 3d field.
<i>mid2edge_gridvector</i>	Converts an mid-point based grid vector into an edge-based grid vector.
<i>region2slice</i>	Given a region of ((lon1,lon2),(lat1, lat2)) the routine provides a slicing ((row1, row2), (col1, col2)) than accomodates the region.
<i>remap_field</i>	Do nearest nearbor remapping on a simple equi-distant local cartesian coordiante grid.
<i>simple_pixel_area</i>	Approximates the grid box area of a lon - lat grid.
<i>spline_smoothing</i>	Smoothes a curve with smoothing spline.
<i>tube_cutout4box</i>	Performs tube cutout with given index set.
<i>xy2ll</i>	Transformation between local Cartesian coordinates in west-east direction (x) and south-north direction (y) and longitude and latitude.

### 1.1.3 trophy.analysis\_tools.optical\_flow

<i>displacement_from_opt_flow</i>	Derives displacement vector between to fields f1 and f2 using optical flow method (either farnebaeck, or tvl1).
<i>displacement_from_opt_flow_farneback</i>	Derives displacement vector between to fields f1 and f2 using optical flow method after Farneback (2003).
<i>displacement_from_opt_flow_tvl1</i>	Derives displacement vector between to fields f1 and f2 using optical flow method after Zach et al (2007).
<i>displacement_vector</i>	Calculates displacement vector for several times.
<i>flow_velocity</i>	Calculates flow field for Lagrangian displacements of tracers in the field.
<i>Lagrangian_change</i>	Calculates Lagrangian change of a field using possibly another field to generate optical flow.
<i>morph_trans_opt_flow</i>	Applies morphological transformation of field f given a displacement field.

### 1.1.4 trophy.analysis\_tools.segmentation

<i>clustering</i>	General interface to aplly either connectivity or watershed clustering.
<i>combine_object_stack</i>	Sequentially combines object label fields by searching for new objects that have no predecessor in the higher hierarchy level.
<i>connected_sequences2pairlists</i>	The routine gets a set of paired numbers which show connection between numbers and collects all connections in a list of pairs.
<i>connectivity_clustering</i>	Applies connectivity clustering to a field.
<i>markers_from_iterative_shrinking</i>	The routine tries to find core parts of objects via iterative shrinking.
<i>multithreshold_clustering</i>	Performs sequential segmentation and returns the combine result only keeping the predecessor object alive.

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Table 4 – continued from previous page

<i>percentiles_from_cluster</i>	Calculates percentiles of cells in an segmented (labeled) field.
<i>remove_clustersize_outside</i>	Counts numbers of pixels per cluster cell and removes clusters outside a size range defined by a minimum and maximum size.
<i>remove_clustersize_outside_slow</i>	OLD VERSION -> much slower ...
<i>remove_small_clusters</i>	Counts numbers of pixels per cluster cell and removes clusters smaller than minimum size.
<i>sequential_segmentation</i>	Performs sequential segmentation.
<i>set_connectivity_footprint</i>	Return connectivity footprint for either 4-connectivity or 8-connectivity.
<i>sort_clusters</i>	It sorts clusters by size (number of pixels) in decreasing order.
<i>watershed_clustering</i>	Applies watershed clustering to a field.
<i>watershed_merge_clustering</i>	Applies watershed clustering to a field and then tries to merge them again.

### 1.1.5 trophy.analysis\_tools.statistics

<i>autocorr</i>	Calculates autocorrelation function of masked array.
<i>cond_perc_simple</i>	This is a simple routine to emulate the Rosenfeld T-Re plots.
<i>correlation_bootstrap</i>	Calculates percentile values of possible correlation using a bootstrap approach.
<i>crosscorr</i>	Calculates time-lagged crosscorrelation function of masked array.
<i>cumsum_data_fraction</i>	The function uses iso-lines of equal density and maps the fraction of data enclosed by these lines onto it.
<i>draw_from_empirical_1ddist</i>	Draw random number from an empirical distribution.
<i>draw_from_empirical_dist</i>	Draw random number from an empirical distribution.
<i>fdistrib_mapping</i>	Performs a transformation of field fin to have the same distribution as fmap.
<i>get_outlier_from_residuals</i>	Calculate outliers from residuals using percentile-based standardization.
<i>KStest</i>	Calculates Kolmogorov Smirnov test for the distributions of two samples in a standard way, but allows for input of effective degrees of freedom.
<i>normalize_field</i>	Normalization of a field is applied.
<i>odrfit_with_outlier_removal</i>	Calculates fitting parameter object using a orthogonal distance regression.
<i>rank_transformation</i>	The routine performs rank transformation of a field, meaning that the ranks of the individual field values are returned.

### 1.1.6 trophy.analysis\_tools.thermodynamic\_variables

<i>absolute_humidity</i>	Calculates absolute humidity.
<i>both_mass_fractions2mixing_ratios</i>	Converts water and the vapor mass fraction to condensed water and vapor mixing ratio.

Continued on next page

Table 6 – continued from previous page

<i>dew_point</i>	Calculates dew point temperature after Markowski book p 13 eq (2.25)
<i>dry_air_density</i>	Calculates dry air density.
<i>dry_air_potential_temperature</i>	Calculates dry air potential temperature.
<i>equivalent_potential_temperature</i>	Calculates equivalent potential temperature.
<i>H2r</i>	Converts relative humidity H into mixing ratio r.
<i>lifting_condensation_level_temperature</i>	Calculates lifting condensation level temperature.
<i>moist_gas_constant</i>	Calculates gas constant of moist air.
<i>moist_potential_temperature</i>	Calculates moist air potential temperature.
<i>moist_specific_heat</i>	Calculates specific heat at constant pressure of moist air.
<i>relative_humidity</i>	Calculates relative humidity.
<i>saturation_over_ice</i>	Calculates saturation water vapor pressure over ice.
<i>saturation_pressure</i>	Calculates saturation water vapor pressure after Bolton (1980) MWR 108, p.1046, eq.
<i>specific_humidity</i>	Calculates specific humidity.
<i>thermodynamic_constants</i>	Sets some thermodynamic constants.
<i>total_density</i>	Calculates total density of moist air with hydrometeors.
<i>watermass_fraction2rw</i>	Converts water mass fraction to condensed water mixing ratio.
<i>water_vapor_pressure</i>	Calculates water vapor pressure.

## 1.2 tropy.io\_tools

### 1.2.1 tropy.io\_tools.data\_collection

<i>Dataset</i>	Description:
<i>DataCollection</i>	

### 2.1 `tropy.analysis_tools`

The `tropy.analysis` package is a collection of tools for dedicated analysis of scientific data. It has been especially used to explore atmospheric data, but the application to other types of data might be rather straight forward and meaningful.

#### 2.1.1 `tropy.analysis_tools.derived_variables`

This module contains / will contain variables derived from some basic input fields. The name “basic” refers to the typical set of variables stored in climate or weather model output.

`tropy.analysis_tools.derived_variables.calc_lwp` ( $p, T, qv, qc, g=9.8, axis=-1$ )

Calculates the liquid water path.

##### Parameters

- **p** (*numpy array*) – atmospheric pressure [hPa]
- **T** (*numpy array*) – temperature [K]
- **qv** (*numpy array*) – water vapor mixing ratio [kg / kg]
- **qc** (*numpy array*) – liquid water mixing ratio [kg / kg]

**Returns** **lwp** – liquid water path [g / m\*\*2]

**Return type** `numpy array`

##### Notes

- hydrostatic approximation is used to convert vertical integration from pressure to height levels
- impact of condensate mass on air density is ignored.

## 2.1.2 trophy.analysis\_tools.grid\_and\_interpolation

`trophy.analysis_tools.grid_and_interpolation.COSMO_mean_vertical_levels` (*Nlev*)

Outputs the mean vertical levels of COSMO-DE taken from the model description of COSMO (see reference below).

**Parameters** *Nlev* (*int*) – number of vertical levels, i.e.

*Nlev* = 51 for intermediate half-levels (e.g. for *w*) *Nlev* = 50 for main levels (e.g. for *u*)

**Returns** *lev* – mean vertical COSMO-DE levels

**Return type** numpy array (1-dim)

### References

M. Baldauf, J. Foerstner, S. Klink, T. Reinhardt, C. Schraff, A. Seifert und K. Stephan “Kurze Beschreibung des Lokal-Modells Kuerzestfrist COSMO-DE (LMK) und seiner Datenbanken auf dem Datenserver des DWD”, Version 1.6, Stand: 31. 03. 2011

Tabelle 1, Seite 29

`trophy.analysis_tools.grid_and_interpolation.create_interpolation_index` (*lon1*,  
*lat1*,  
*lon2*,  
*lat2*,  
*xy=False*)

Given georeference of two grids in lon and lat, an index is built to map a field on grid 1 to grid 2.

#### Parameters

- **lon1** (*numpy array (2-dim)*) – longitude of input grid 1
- **lat1** (*numpy array (2-dim)*) – latitude of input grid 1
- **lon2** (*numpy array (2-dim)*) – longitude of target grid 2
- **lat2** (*numpy array (2-dim)*) – latitude of target grid 2
- **xy** (*bool, optional, default = False*) – switch if georef field are interpreted as local Cartesian coordinates: Then no internal transformation is used.

#### Returns

- **ir** (*numpy array, 2-dim, dtype = int*) – index field for transforming rows
- **ic** (*numpy array, 2-dim, dtype = int*) – index field for transforming columns

`trophy.analysis_tools.grid_and_interpolation.curve_flow_filter` (*f*, *numberOfIterations=5*)

Smoothing filter depending on isoline curvature. Interface for curvature flow filter from simpleITK toolkit.

#### Parameters

- **f** (*numpy array (2-dim)*) – 2d field to be filtered (smoothed)
- **numberOfIterations** (*int, optional, default = 5*) – number of iterations, increases smoothing effect

**Returns** *f\_sm* – smoothed 2d field

**Return type** numpy array (2-dim)

## Notes

Only works if SimpleITK is installed !!!

`trophy.analysis_tools.grid_and_interpolation.cutout_cluster(c, nc, nedge=30)`

Makes a cutout of a categorical field *c* for an object of class / number *nc*.

### Parameters

- **c** (*numpy array, 2-dim, dtype = int*) – 2d categorical field
- **nc** (*int*) – category number / class which is chosen
- **nedge** (*int, optional, default = 30*) – number of edge pixels added / included

**Returns** *ccut* – cutout of categorical field *c*

**Return type** *numpy array, 2-dim, dtype = int*

`trophy.analysis_tools.grid_and_interpolation.cutout_field4box(f, ind, bsize, **kwargs)`

Cuts out a field based on center pix index and box size.

### Parameters

- **f** (*2d or 3d numpy array*) – the field to be cut out
- **ind** (*tuple or list*) – center index of horizontal cutout
- **bsize** (*int*) – size of the quadratic cutout box

**Returns** *fcut* – resulting quadratic cutout

**Return type** *2d or 3d numpy array*

`trophy.analysis_tools.grid_and_interpolation.cutout_fields(fin, slices, vaxis=0)`

Cuts out a field or a list of similar fields given row and column slices.

### Parameters

- **fin** (*2d or 3d numpy array or list of fields*) – the fields to be cut out
- **slices** (*tuple or list*) – region slice such that the cutout is done as ((row1, row2), (col1, col2))
- **vaxis** (*int, optional, default = 0*) – for 3d fields, axis which is still varying

**Returns** *fout* – resulting cutout (possibly a list of cutted fields)

**Return type** *2d or 3d numpy array or list of fields*

`trophy.analysis_tools.grid_and_interpolation.get_index(p, lon, lat)`

Given a point *p* the nearest grid point is returned.

### Parameters

- **p** (*numpy array*) – point in the 2d grid
- **lon** (*numpy array (2-dim)*) – grid values of the 1st dimension (can be longitude, x, etc.)
- **lat** (*numpy array (2-dim)*) – grid values of the 2nd dimension (can be latitude, y, etc.)

### Returns

- **ir** (*numpy array (2-dim)*) – row index (the 1st)
- **ic** (*numpy array (2-dim)*) – column index (the 2nd)

`trophy.analysis_tools.grid_and_interpolation.i2iset(v, i)`

Returns an index set for a n-dim numpy array given an index i which points to a position within the same but flattened array.

It is assumed that the array was flattened in C-mode.

**Parameters**

- **v** (*numpy array (n-dim)*) – field which determines the rank and shape
- **i** (*int*) – input index which points to the position in `v.flatten()`

**Returns** **iset** – index set which points to the position in **v**

**Return type** `tuple`

**Notes**

This routine helps you to locate a maximum in a n-dimensional array using e.g. the `np.argmax` function.

`trophy.analysis_tools.grid_and_interpolation.interpolate_field_to_hor_pos(pos, lon, lat, field, **kwargs)`

Given a sequence of points (or just one) a field (2d or 3d) given a grid is interpolated to the new positions.

**Parameters**

- **pos** (*numpy array (2-dim)*) – sequence of points given as `np.array((x,y))` where x and y can be n-dimensional
- **lon** (*numpy array (2-dim)*) – grid values of the 1st dimension (can be longitude, x, etc.)
- **lat** (*numpy array (2-dim)*) – grid values of the 2nd dimension (can be latitude, y, etc.)
- **field** (*numpy array*) – 2d or 3d field to be interpolated to pos
- **\*\*kwargs** (*dict content*) – keyword arguments for the `griddata` routine of the `scipy.interpolate` package can be passed through

**Returns** **f** – field interpolated to the position sequence pos, f has the dimension n x Nlev, where n is the dimension of pos, Nlev the vertical dimension

**Return type** `numpy array`

`trophy.analysis_tools.grid_and_interpolation.ldiff(var, axis=-1)`

Calculates difference of a field at adjacent levels. Now, wrapper for `numpy.diff` with default on last axis.

**Parameters**

- **var** (*numpy array (n-dim where n = [1,2,3])*) – field; 1d, 2d or 3d array
- **axis** (*int, optional, default = -1*) – axis used for layer averaging

**Returns** **dvar** – level difference

**Return type** `numpy array (n-dim where n = [1,2,3])`

`trophy.analysis_tools.grid_and_interpolation.ll2xy(lon, lat, lon0=10.0, lat0=50.0, R=6380.0)`

Transformation between longitude and latitude and local Cartesian coordinates in west-east direction (x) and south-north direction (y). The local projection is called “sinusoidal”.

### Parameters

- **lon** (*numpy array*) – longitude
- **lat** (*numpy array*) – latitude
- **lon0** (*float, optional, default = 10.*) – arbitrary initial longitude for  $x = 0$
- **lat0** (*float, optional, default = 50.*) – arbitrary initial latitude for  $y = 0$
- **R** (*float, optional, default = 6380.*) – approximation for the radius of Earth

### Returns

- **x** (*numpy array*) – Cartesian coordinate in west-east direction (increasing towards the East)
- **y** (*numpy array*) – Cartesian coordinate in south-north direction (increasing towards the North)

### Notes

The unit of Earth's radius determines the units of  $x$  and  $y$ . Default (km).

### References

See [https://en.wikipedia.org/wiki/Sinusoidal\\_projection](https://en.wikipedia.org/wiki/Sinusoidal_projection) for further info.

`trophy.analysis_tools.grid_and_interpolation.ll2xyc(lon, lat, mlon=None, mlat=None, lon0=0, lat0=0.0)`

Applies a centered version of ll2xy. The centering is around the mean lon/lat value and lon0, lat0 only define the x/y offset.

### Parameters

- **lon** (*numpy array*) – longitude
- **lat** (*numpy array*) – latitude
- **mlon** (*float, optional, default = None*) – longitude center point of projection  
if None then average of lon is used
- **mlat** (*float, optional, default = None*) – latitude center point of projection  
if None then average of lat is used
- **lon0** (*float, optional, default = 0.*) – zero longitude (for x-offset)
- **lat0** (*float, optional, default = 0.*) – zero latitude (for y-offset)

### Returns

- **x** (*numpy array*) – x-coordinate
- **y** (*numpy array*) – y-coordinate

`trophy.analysis_tools.grid_and_interpolation.lmean(var, axis=-1)`

Calculates the layer mean of a field.

### Parameters

- **var** (*numpy array (n-dim where  $n = [1, 2, 3]$ )*) – field; 1d, 2d or 3d array

- **axis** (*int*, *optional*, *default* = -1) – axis used for layer averaging

**Returns** **varm** – layer mean

**Return type** numpy array (n-dim where n = [1,2,3])

`trophy.analysis_tools.grid_and_interpolation.low2hres_region(region)`

Based on low-res. region of MSG channels an corresponding HRV region is calculated.

**Parameters** **region** (*tuple of two tuples*) – region slice for cutting a field ((row1, row2), (col1, col2))

**Returns** **hrv\_region**

**Return type** corresponding HRV region as ((row1, row2), (col1, col2))

## Notes

### Calculation

It is assumed that the hrv region is a dependent variable which depends on the low res region attribute That means: Given the low res region cutout hrv region is determined as corresponding cutout which

- (i) exactly fits with the low res region and
- (ii) refers to the artificial hrv full disk of 11136 x 11136

low res pixel with the index (I,J) = (0,0) has a high res pixel with index (i\_m, j\_m) = (2,2) in the middle and (i\_u, j\_u) = (1,1) in the upper corner see doc “MSG Level 1.5 Image Data Format Description”, Fig.8 which then leads to the relation between the low res pixel (I,J) and its corresponding upper corner high res pixel (i\_u, j\_u): (i\_u, j\_u) = 3 \* (I, J) + 1

`trophy.analysis_tools.grid_and_interpolation.make_add_edge(var)`

Adds egde region to 2d data field. Values in added left and right column, and lower and upper row are linearly extrapolated from their neighbors.

**Parameters** **var** (*numpy array (2-dim)*) – 2d variable field (Nrow, Ncols)

**Returns** **var\_new** – 2d variable field with added edge values (Nrows + 2, Ncols + 2)

**Return type** numpy array (2-dim)

`trophy.analysis_tools.grid_and_interpolation.make_hrv_upscaling(v)`

A 2d field is upscaled to 3-fold higher resolution using linear interpolation. Edge values are determined by linear extrapolation.

**Parameters** **v** (*numpy array (2-dim)*) – 2d variable field (Nrow, Ncols)

**Returns** **vhigh** – variable field with 3-times higher resolution (3\*Nrows, 3\*Ncols)

**Return type** numpy array (2-dim)

`trophy.analysis_tools.grid_and_interpolation.make_index_set(nrows, ncols)`

This simple routine makes an 2d index set for given row and column number of a 2d field.

**Parameters**

- **nrows** (*int*) – number of rows
- **ncols** (*int*) – number of columns

**Returns**

- **ii** (*numpy array, dtype = int*) – 2d field of rows indices
- **jj** (*numpy array, dtype = int*) – 2d field of column indices



`trophy.analysis_tools.grid_and_interpolation.make_vert_cut` (*p1*, *p2*, *lon*, *lat*, *vg*,  
\*\**kwargs*)

Makes a horizontal - vertical cut through a 3d field.

#### Parameters

- **p1** (*list* or *numpy array*) – 1st end of the cutting line (lon, lat)
- **p2** (*list* or *numpy array*) – 2nd end of the cutting line (lon, lat)
- **lon** (*numpy array (2-dim)*) – longitude
- **lat** (*numpy array (2-dim)*) – latitude
- **vg** (*numpy array (3-dim)*) – 3d field to be interpolated
- **\*\*kwargs** (*dict content*) – keyword arguments for the griddata routine of the `scipy.interpolate` package can be passed through

#### Returns

- **s** (*numpy array (1-dim)*) – distance of km from one to the other end of the cutting line
- **lo** (*numpy array (1-dim)*) – longitude along the cut
- **la** (*numpy array (1-dim)*) – latitude along the cut
- **v** (*numpy array (2-dim)*) – field interpolated to the cutting line

Vertical levels are kept.

`trophy.analysis_tools.grid_and_interpolation.mid2edge_gridvector` (*x\_mid*,  
*x\_edge0*)

Converts an mid-point based grid vector into an edge-based grid vector.

#### Parameters

- **x\_mid** (*numpy array (1-dim)*) – mid-point-based grid vector (length N)
- **x\_edge0** (*float*) – left edge

**Returns** **x\_edge** – edge-based grid vector (length N + 1)

**Return type** *numpy array (1-dim)*

`trophy.analysis_tools.grid_and_interpolation.region2slice` (*lon*, *lat*, *region*)

Given a region of ((lon1,lon2),(lat1, lat2)) the routine provides a slicing ((row1, row2), (col1, col2)) than accommodates the region.

#### Parameters

- **lon** (*numpy array (2-dim)*) – longitude
- **lat** (*numpy array (2-dim)*) – latitude
- **region** (*tuple of two tuples*) – ((lon1, lon2), (lat1, lat2))

**Returns** **slice** – region slice for cuuting a field ((row1, row2), (col1, col2))

**Return type** *tuple of two tuples*

`trophy.analysis_tools.grid_and_interpolation.remap_field` (*lon*, *lat*, *f*, *dr=0.5*)

Do nearest nearbor remapping on a simple equi-distant local cartesian coordiante grid.

Uses curvature flow filter for smoothing and pixel edge correction.

#### Parameters

- **lon** (*numpy array, (2-dim)*) – longitude

- **lat** (*numpy array, (2-dim)*) – latitude
- **f** (*numpy array, (2-dim)*) – 2d field to be interpolated
- **dr** (*float, optional, default = 0.5*) – grid spacing in km

#### Returns

- **xnew** (*numpy array, (2-dim)*) – regridded, equi-distant x-coordinate in km
- **ynew** (*numpy array, (2-dim)*) – regridded, equi-distant y-coordinate in km
- **fint** (*numpy array, (2-dim)*) – remapped field

`trophy.analysis_tools.grid_and_interpolation.simple_pixel_area(lon, lat, xy=False, uncertainty=False)`

Approximates the grid box area of a lon - lat grid.

The grid is treated as if it is an equidistant grid in local Cartesian coordinates on Earth's surface (i.e. in x and y). The grid point is placed at each of the four corners of the grid box and the corresponding average area is calculated.

#### Parameters

- **lon** (*numpy array (2-dim)*) – longitude
- **lat** (*numpy array (2-dim)*) – latitude
- **xy** (*bool, optional, default = False*) – switch if lon & lat are interpreted as local Cartesian coordinates
- **uncertainty** (*bool, optional, default = False*) – switch if std of pixel area is output

Deviation are caused by effects of the not-rectangular grid.

#### Returns

- **a** (*numpy array (2-dim)*) – grid box area
- **da** (*numpy array (2-dim), optional*) – standard deviation of grid box area, if uncertainty == True

`trophy.analysis_tools.grid_and_interpolation.spline_smoothing(x, y, s=0.1, fixed_endpoints=True)`

Smooths a curve with smoothing spline.

#### Parameters

- **x** (*numpy array (1-dim)*) – abscissa values
- **y** (*numpy array (1-dim)*) – ordinate values
- **s** (*float, optional, default = 0.1*) – smoothing parameter
- **fixed\_endpoints** (*bool, optional, default = True*) – switch, if endpoints should be hold fixed

**Returns** **y\_smooth** – smoothed version of y

**Return type** *numpy array (1-dim)*

## Notes

Uses the function `scipy.interpolate.UnivariateSpline`.

`trophy.analysis_tools.grid_and_interpolation.tube_cutout4box` (*v3d, i0, boxsize*)

Performs tube cutout with given index set. The center index might not be constant.

### Parameters

- **v3d** (*numpy array (3-dim)*) – 3d field, 1st dimension is time (or variable)
- **i0** (*list or numpy array (2-dim)*) – center indices (irvec, icvec) index tuple, (row index vector, co index vector)
- **boxsize** (*int*) – size of the quadratic cutout box

**Returns** **vtube** – 3d tube, quadratic cutout in the last two dimensions

**Return type** `numpy array (3-dim)`

## Notes

This function can be used to make vertical cutouts, but a transpose command has to be applied before and after.

`trophy.analysis_tools.grid_and_interpolation.xy2ll` (*x, y, lon0=10.0, lat0=50.0, R=6380*)

Transformation between local Cartesian coordinates in west-east direction (x) and south-north direction (y) and longitude and latitude. This assumes a sinusoidal projection. Inversion of the function `ll2xy`.

### Parameters

- **x** (*numpy array*) – Cartesian coordinate in west-east direction (increasing towards the East)
- **y** (*numpy array*) – Cartesian coordinate in south-north direction (increasing towards the North)
- **lon0** (*float, optional, default = 10.*) – arbitrary initial longitude for  $x = 0$
- **lat0** (*float, optional, default = 50.*) – arbitrary initial latitude for  $y = 0$
- **R** (*float, optional, default = 6380.*) – approximation for the radius of Earth

### Returns

- **lon** (*numpy array*) – longitude
- **lat** (*numpy array*) – latitude

## Notes

The unit of Earth's radius determines the units of x and y. Default (km).

## References

See [https://en.wikipedia.org/wiki/Sinusoidal\\_projection](https://en.wikipedia.org/wiki/Sinusoidal_projection) for further info.

### 2.1.3 tropy.analysis\_tools.optical\_flow

```
tropy.analysis_tools.optical_flow.Lagrangian_change(f3d,          tracer_field=None,
                                                    gauss_sigma=1.0,
                                                    flow_input=None,    vmin=0,
                                                    vmax=1.0)
```

Calculates Lagrangian change of a field using possibly another field to generate optical flow.

#### Parameters

- **f3d** (*numpy array*) – the field (time on 1st axis) for which Lagrangian change is calculated
- **tracer\_field** (*numpy array, optional, default = None*) – the field (time on 1st axis) for which displacement is derived if None: displacement is derived from f3d
- **sigma\_gauss** (*float, optional, default = 1.*) – sigma for Gaussian smoothing for Lagrangian change fields
- **flow\_input** (*numpy array, optional, default = None*) – predefined flow field for u/v calculation  
if None: flow is derived from either f3d or tracer\_field
- **vmin** (*float, optional, default = 0.*) – lower bound of fields for rescaling needed for flow calculation
- **vmax** (*float, optional, default = 1.*) – upper bound of fields for rescaling needed for flow calculation

**Returns** **df** – Lagrangian change of field f3d

**Return type** numpy array

#### Notes

Lagrangian changes are derived from displacements between subsequent time slots and hence, time dimension is `ntimes - 1`.

```
tropy.analysis_tools.optical_flow.displacement_from_opt_flow(f1,          f2,
                                                            method='farneback',
                                                            **kwargs)
```

Derives displacement vector between to fields f1 and f2 using optical flow method (either farnebaeck, or tvl1).

#### Parameters

- **f1** (*numpy array, 2dim*) – field for past time step
- **f2** (*numpy array, 2dim*) – field for actual time step
- **method** (*str, optional, default = 'farneback'*) – method selected for optical flow calculations possible options: 'farneback' and 'tvl1'
- **kwargs** (*dict*) – parameters for opencv optical flow algorithm if not given, default is taken from `_farnebaeck_default_parameters`

**Returns** **flow** – displacement vector as index shift 1st component is related to u, i.e. displacement along row, 2nd to v, i.e. along column

**Return type** numpy array, 3dim

```
trophy.analysis_tools.optical_flow.displacement_from_opt_flow_farneback(f1,
                                                                    f2,
                                                                    vmin=None,
                                                                    vmax=None,
                                                                    **kwargs)
```

Derives displacement vector between to fields f1 and f2 using optical flow method after Farneback (2003).

#### Parameters

- **f1** (*numpy array, 2dim*) – field for past time step
- **f2** (*numpy array, 2dim*) – field for actual time step
- **vmin** (*float, optional, default = None*) – lower bound of fields for rescaling
- **vmax** (*float, optional, default = None*) – upper bound of fields for rescaling
- **kwargs** (*dict*) – parameters for opencv optical flow algorithm if not given, default is taken from `_farnebaeck_default_parameters`

**Returns flow** – displacement vector as index shift 1st component is related to u, i.e. displacement along row, 2nd to v, i.e. along column

**Return type** *numpy array, 3dim*

```
trophy.analysis_tools.optical_flow.displacement_from_opt_flow_tv11(f1, f2,
                                                                    vmin=None,
                                                                    vmax=None,
                                                                    **kwargs)
```

Derives displacement vector between to fields f1 and f2 using optical flow method after Zach et al (2007).

#### Parameters

- **f1** (*numpy array, 2dim*) – field for past time step
- **f2** (*numpy array, 2dim*) – field for actual time step
- **vmin** (*float, optional, default = None*) – lower bound of fields for rescaling
- **vmax** (*float, optional, default = None*) – upper bound of fields for rescaling
- **kwargs** (*dict*) – parameters for opencv optical flow algorithm if not given, default is taken from `_farnebaeck_default_parameters`

**Returns flow** – displacement vector as index shift 1st component is related to u, i.e. displacement along row, 2nd to v, i.e. along column

**Return type** *numpy array, 3dim*

```
trophy.analysis_tools.optical_flow.displacement_vector(tracer_field, vmin=0,
                                                         vmax=1.0)
```

Calculates displacement vector for several times.

#### Parameters

- **tracer\_field** (*numpy array*) – 3d fields (time on 1st axis)
- **vmin** (*float, optional, default = 0.*) – lower bound of fields for rescaling needed for flow calculation
- **vmax** (*float, optional, default = 1.*) – upper bound of fields for rescaling needed for flow calculation

**Returns flow** – displacement vector

**Return type** *numpy array*

`tropy.analysis_tools.optical_flow.flow_velocity(lon, lat, f3d, dt=5.0, flow_input=None, vmin=0, vmax=1.0)`

Calculates flow field for Lagrangian displacements of tracers in the field.

#### Parameters

- **lon** (*numpy array*) – longitude
- **lat** (*numpy array*) – latitude
- **f3d** (*numpy array*) – 3d fields (time on 1st axis)
- **dt** (*float, optional, default = 5.*) – time interval in minutes
- **flow\_input** (*numpy array, optional, default = None*) – predefined flow field for u/v calculation not use if None
- **vmin** (*float, optional, default = 0.*) – lower bound of fields for rescaling needed for flow calculation
- **vmax** (*float, optional, default = 1.*) – upper bound of fields for rescaling needed for flow calculation

#### Returns

- **u** (*numpy array*) – zonal velocity,
- **v** (*numpy array*) – meridional velocity

#### Notes

Velocities are derived from displacements between subsequent time slots and hence, time dimension is ntimes - 1.

`tropy.analysis_tools.optical_flow.morph_trans_opt_flow(f, flow, method='forward')`

Applies morphological transformation of field f given a displacement field.

#### Parameters

- **f** (*numpy array*) – 2d field that is transformed (source)
- **flow** (*numpy array*) – displacement vector between source and target stage

**Returns** **ftrans** – transformed field

**Return type** `numpy array`

## 2.1.4 tropy.analysis\_tools.parallax

## 2.1.5 tropy.analysis\_tools.segmentation

Package to perform segmentation of fields.

The special emphasis is on threshold-based segmentation of 2-dim atmospheric fields for which object-based analysis techniques should be applied.

#### Notes

- the package has variants of connected component analysis & watershedding
- some split-and-merge rules exists
- most of the method might be also applied to 3-dim fields

`trophy.analysis_tools.segmentation.clustering` (*f*, *thresh*, *cluster\_method*='watershed', *min\_size*=20, *\*\*kwargs*)

General interface to apply either connectivity or watershed clustering.

**Parameters**

- **f** (*np.array*) – field to be segmented
- **thresh** (*float*) – threshold for masking, set foreground to  $f > \text{thresh}$
- **cluster\_method** (*{'watershed', 'watershed\_merge', 'connect'}*, *optional*) – switch between watershed or connectivity clustering
- **min\_size** (*int*, *optional*, *default* = 20) – minimum cluster size (px) left in the cluster field
- **\*\*kwargs** (*dict*, *optional*) – keyword argument for selected cluster algorithm

**Returns** **c** – categorical cluster field

**Return type** *np.array*

`trophy.analysis_tools.segmentation.combine_object_stack` (*c3d*)

Sequentially combines object label fields by searching for new objects that have no predecessor in the higher hierarchy level.

**Parameters** **c3d** (*numpy array*, *3dim* or *4dim*) – labeled object data stack

**Returns** **c** – combined label field

**Return type** *numpy array*, *2dim* or *3dim*

`trophy.analysis_tools.segmentation.connected_sequences2pairlists` (*connected\_pairs\_list*)

The routine gets a set of paired numbers which show connection between numbers and collects all connections in a list of pairs.

**Parameters** **connected\_pairs\_list** (*list*) – list of number pairs

**Returns** **pair\_list** – dictionary that contains pair lists

**Return type** *dict*

`trophy.analysis_tools.segmentation.connectivity_clustering` (*f*, *thresh*, *filter\_method*='curve', *numberOfIterations*=5, *ctype*=4, *cluster\_masking*=True, *sig\_gauss*=1.0, *\*\*kwargs*)

Applies connectivity clustering to a field.

**Parameters**

- **f** (*np.array*) – field to be segmented
- **thresh** (*float*) – threshold for masking, set foreground to  $f > \text{thresh}$
- **filter\_method** (*{'gauss', 'curve'}*, *optional*) – filter method used to smooth the input field
  - 'gauss' for 2d-Gaussian
  - 'curve' for curvature flow filter
- **numberOfIterations** (*int*, *optional*) – number of repeated application of curvature flow filter the larger then smoother if selected
- **ctype** (*{4, 8}*, *optional*) – set either 4- or 8-connectivity (edge vs. edges+corners)

- **cluster\_masking** (*{True, False}, optional*) – if original (True) or smoothed (False) threshold mask is used.
- **siggauss** (*float, optional*) – sigma for Gaussian filter if selected
- **\*\*kwargs** (*dict, optional*) – set of additional, but not used keywords

**Returns** **c** – categorial cluster field

**Return type** np.array

```
trophy.analysis_tools.segmentation.markers_from_iterative_shrinking(mask,
                                                                    marker_shrinkage_Niter=3,
                                                                    marker_shrinkage_dmin=0.1,
                                                                    **kwargs)
```

The routine tries to find core parts of objects via iterative shrinking.

Each object is normalized by its maximum distance-to-background which makes the algorithm scale-invariant.

**Parameters** **mask** (*np.array*) – binary mask

**marker\_shrinkage\_Niter** [int, optional] option for marker creation, how often shrinking is applied

**marker\_shrinkage\_dmin** [float, optional] option for marker creation

How much of the edge is taken away. Distance is measured relative to the maximum distance, thus  $0 < dmin < 1$ .

**Returns** **markers** – markers field with increasing index per marker region

**Return type** np.array

```
trophy.analysis_tools.segmentation.multithreshold_clustering(f, thresh_min,
                                                            thresh_max,
                                                            nthresh=10,
                                                            use_percentile_threshold=False,
                                                            **kws)
```

Performs sequential segmentation and returns the combine result only keeping the predecessor object alive.

**Parameters**

- **f** (*numpy array, 2dim or 3dim*) – input field
- **thresh\_min** (*float*) – minimum threshold value
- **thresh\_max** (*float*) – maximum threshold value
- **nthresh** (*int, optional*) – number of threshold values
- **use\_percentile\_threshold** (*{True, False}, optional*) – if threshold list is derived from field percentiles
- **\*\*kws** (*dict*) – keywords passed to segmentation routine

**Returns** **c** – combined label field

**Return type** numpy array, same shape as f

```
trophy.analysis_tools.segmentation.percentiles_from_cluster(f, c, p=[25, 50, 75], index=None)
```

Calculates percentiles of cells in an segmented (labeled) field.

Functionality missing in `scipy.ndimage.measurements`.

**Parameters**



- **f** (*np.array*) – the field as basis for percentile calculation
- **c** (*np.array*) – categorial cluster field
- **p** (*list or np.array, optional*) – percentiles array
- **index** (*list or None, optional*) – list of object labels for which percentiles are calculated

**Returns** **pc** – percentiles per cell (including background (set to zero))

**Return type** *np.array*

```
trophy.analysis_tools.segmentation.remove_clustersize_outside(c, min_size=6,
                                                             max_size=inf)
```

Counts numbers of pixels per cluster cell and removes clusters outside a size range defined by a minimum and maximum size.

In addition, it sorts clusters by size in decreasing order.

#### Parameters

- **c** (*np.array*) – categorial cluster field
- **min\_size** (*int, optional*) – minimum cluster size (px) left in the cluster field
- **max\_size** (*int, optional*) – maximum cluster size (px) left in the cluster field

#### Returns

**c\_re** – categorial cluster field with sorted objects (from large to small)

- objects smaller than or equal *min\_size* are removed
- objects larger than or equal *max\_size* are removed

**Return type** *np.array*

```
trophy.analysis_tools.segmentation.remove_clustersize_outside_slow(c,
                                                                    min_size=6,
                                                                    max_size=inf,
                                                                    sort=True)
```

OLD VERSION -> much slower ...

Counts numbers of pixels per cluster cell and removes clusters outside a size range defined by a minimum and maximum size.

In addition, it sorts clusters by size in decreasing order.

#### Parameters

- **c** (*np.array*) – categorial cluster field
- **min\_size** (*int, optional*) – minimum cluster size (px) left in the cluster field
- **max\_size** (*int, optional*) – maximum cluster size (px) left in the cluster field
- **sort** (*{True, False}, optional*) – if objects are sorted by size

#### Returns

**c\_re** – categorial cluster field with sorted objects (from large to small)

- objects smaller than or equal *min\_size* are removed
- objects larger than or equal *max\_size* are removed

**Return type** *np.array*

`trophy.analysis_tools.segmentation.remove_small_clusters(c, min_size=6)`

Counts numbers of pixels per cluster cell and removes clusters smaller than minimum size.

In addition, it sorts clusters by size in decreasing order.

**Parameters**

- **c** (*np.array*) – categorial cluster field
- **min\_size** (*int, optional*) – minimum cluster size (px) left in the cluster field

**Returns**

**c\_re** – categorial cluster field with sorted objects (from large to small)  
objects smaller than or equal *min\_size* are removed

**Return type** *np.array*

`trophy.analysis_tools.segmentation.sequential_segmentation(f, thresh_min, thresh_max, nthresh=10, use_percentile_threshold=False, **kws)`

Performs sequential segmentation.

**Parameters**

- **f** (*numpy array, 2dim or 3dim*) – input field
- **thresh\_min** (*float*) – minimum threshold value
- **thresh\_max** (*float*) – maximum threshold value
- **nthresh** (*int*) – number of threshold values
- **use\_percentile\_threshold** (*{True, False}, optional*) – if threshold list is derived from field percentiles
- **\*\*kws** (*dict*) – keywords passed to segmentation routine

**Returns** **c3d** – stack of labeled field, from max. thresh to min. thresh

**Return type** *numpy array*, one dimension added

`trophy.analysis_tools.segmentation.set_connectivity_footprint(ctype, ndim)`

Return connectivity footprint for either 4-connectivity or 8-connectivity.

4-connectivity only includes neighboring sides, 8-connectivity also includes neighboring diagonal values.

**Parameters**

- **ctype** (*{4, 8}*) – set either 4- or 8-connectivity (edge vs. edges+corners)
- **ndim** (*{2, 3}*) – dimension of the field (either 2-dim or 3-dim)

**Returns** **footprint** – *ndim*-dimensional connectivity footprint

**Return type** *np.array*

`trophy.analysis_tools.segmentation.sort_clusters(c)`

It sorts clusters by size (number of pixels) in decreasing order.

**Parameters** **c** (*np.array*) – categorial cluster field

**Returns** **c\_re** – categorial cluster field with sorted objects (from large to small)

**Return type** *np.array*

```
trophy.analysis_tools.segmentation.watershed_clustering(f, thresh, dradius=3, ctype=4,
    marker_field='dist', filter_method='curve',
    numberOfIterations=5,
    cluster_masking=True,
    exclude_border=False,
    marker_method='mahotas_regmax',
    siggauss=1.0,
    marker_shrinkage_Niter=3,
    marker_shrinkage_dmin=0.1,
    **kwargs)
```

Applies watershed clustering to a field.

### Parameters

- ***f*** (*np.array*) – field to be segmented
- ***thresh*** (*float*) – threshold for masking, set foreground to  $f > \text{thresh}$
- ***dradius*** (*int*, *optional*) – minimal distance for neighboring maxima
- ***ctype*** (*{4, 8}*, *optional*) – set either 4- or 8-connectivity (edge vs. edges+corners)
- ***marker\_field*** (*str*, *optional*) – switch if input field *f* or a distance transform as used to set the watershed markers
  - if ‘dist’, then max. in Euclidean distance to background is used to set initial markers, else the max. in field *f* are taken
- ***filter\_method*** (*{'gauss', 'curve'}*, *optional*) – filter method used to smooth the input field
  - ‘gauss’ for 2d-Gaussian
  - ‘curve’ for curvature flow filter
- ***numberOfIterations*** (*int*, *optional*) – number of repeated application of curvature flow filter the larger then smoother if selected
- ***cluster\_masking*** (*{True, False}*, *optional*) – if original (True) or smoothed (False) threshold mask is used.
- ***exclude\_border*** (*{True, False}*, *optional*) – if clusters that touch domain borders are excluded (set to zero)
- ***marker\_method*** (*str*, *optional*) – defines a method for marker assignment
  - ‘brute\_force\_local\_max’ : just uses `scipy.ndimage.maximum_filter`
  - ‘skimage\_peak\_local\_max’ : uses `skimage` local maximum routine
  - ‘mahotas\_regmax’ : uses `mahotas` regional maximum detection
  - ‘iterative\_shrinking’ : uses iterative object shrinking depending on relative distance-to-background
- ***siggauss*** (*float*, *optional*) – sigma for Gaussian filter if selected
- ***marker\_shrinkage\_Niter*** (*int*, *optional*) – option for marker creation, how often shrinking is applied
- ***marker\_shrinkage\_dmin*** (*float*, *optional*) – option for marker creation

How much of the edge is taken away. Distance is measured relative to the maximum distance, thus  $0 < d_{min} < 1$ .

**Returns** **c** – categorial cluster field

**Return type** np.array

`tropy.analysis_tools.segmentation.watershed_merge_clustering` (*f*, *thresh*,  
*min\_size*=20,  
*merge\_ratio*=0.5,  
*\*\*kwargs*)

Applies watershed clustering to a field and then tries to merge them again.

The merging is based on the ratio between maximum distances-to-background within the subclusters and on the cluster borders.

#### Parameters

- **f** (*np.array*) – field to be segmented
- **thresh** (*float*) – threshold for masking, set foreground to  $f > \text{thresh}$
- **min\_size** (*int, optional, default = 20*) – optional, minimum cluster size (px) left in the cluster field
- **merge\_ratio** (*float, optional*) – essentially the ratio between interface length and cluster size  
 if the interface is longer than  $\text{merge\_ratio} * \text{cluster\_size}$  than to two connected objects are merged.
- **\*\*kwargs** (*dict, optional*) – keywords passed to *watershed\_clustering* routine

**Returns** **c\_update** – categorial cluster field (updated for merges)

**Return type** np.array

## 2.1.6 tropy.analysis\_tools.statistics

Package for some statistical utility functions.

**Special Highlights** are:

- conditioned percentiles
- rank transformation / distribution mapping
- drawing from an empirical 1d / 2d distribution

`tropy.analysis_tools.statistics.KStest` (*x*, *y*, *Nx\_eff*=None, *Ny\_eff*=None)

Calculates Kolmogorov Smirnov test for the distributions of two samples in a standard way, but allows for input of effective degrees of freedom.

Nullhypothesis that the two samples share the same distribution is rejected if p-value is smaller than a threshold (typically 0.05 or 0.01).

#### Parameters

- **x** (*numpy array*) – sample of 1st variable
- **y** (*numpy array*) – sample of 2nd variable
- **Nx\_eff** (*int, optional, default = None*) – number of effective degrees of freedom of variable x set to size of x if  $Nx\_eff == \text{None}$

- **Ny\_eff** (*int, optional, default = None*) – number of effective degrees of freedom of variable y set to size of y if Ny\_eff == None

#### Returns

- **D** (*float*) – KS test statistic
- **pval** (*float*) – p-value

`tropy.analysis_tools.statistics.autocorr(x, Nmax=None)`

Calculates autocorrelation function of masked array.

#### Parameters

- **x** (*numpy array*) – sample of 1st variable
- **Nmax** (*int, optional, default = None*) – maximum lag Nmax is set to len(x) - 1 if Nmax == None

**Returns** **c** – autocorrelation function for lag 0 ... Nmax,

**Return type** numpy array

`tropy.analysis_tools.statistics.cond_perc_simple(v1, v2, x1, p=[10, 25, 50, 75, 90], sig=0, medsize=0)`

This is a simple routine to emulate the Rosenfeld T-Re plots.

Input are v1 (T) and v2 (Re) and percentiles of v2 conditioned on intervals of v1 given by x1 are calculated.

#### Parameters

- **v1** (*numpy array*) – sample of 1st variable of bi-variate distribution
- **v2** (*numpy array*) – sample of 2nd variable of bi-variate distribution
- **x1** (*numpy array*) – intervals at which 1st variable is gathered
- **p** (*list, optional, default = [10, 25, 50, 75, 90]*) – percentiles of 2nd variables which will be calculated for each interval [ x1[i], x1[i + 1] )
- **sig** (*float, optional, default = 0.*) – sigma for Gaussian filter to smooth percentile curves (not applied for sig == 0)
- **medsize** (*int, optional, default = 0*) – footprint size of median filter to smooth percentile curves (not applied for sig != 0 OR medsize == 0)

**Returns** **v2p** – percentile statistics of variable v2 conditioned on different x1 intervals

**Return type** numpy array

`tropy.analysis_tools.statistics.correlation_bootstrap(x, y, perc=[2.5, 50, 97.5], xerr=0, yerr=0, Nsample=1000, corr=<function pearsonr>)`

Calculates percentile values of possible correlation using a bootstrap approach.

(x,y) are interpreted as mean values and error bars as standard deviations of 2d Gaussian distribution per point and the fit is repeated for random draws out of these. After Nsample fits, the range of possible fits is analyzed for their percentiles.

#### Parameters

- **x** (*numpy array, 1-dim*) – independent values
- **y** (*numpy array, 1-dim*) – dependent values

- **perc** (*list or numpy array (1-dim)*, optional, default = [2.5, 50, 97.5]) – list of percentiles for which correlation is estimated
- **xerr** (*float or numpy array (1-dim)*, optional, default = 0.) – standard deviation of x
- **yerr** (*float or numpy array (1-dim)*, optional, default = 0.) – standard deviation of y
- **Nsample** (*int*, optional, default = 1000) – size of bootstrap sample
- **corr** (func, optional, default = `scipy.stats.pearsonr`) – function object used for correlation calculation

**Returns** **corr** – percentiles of possible correlation obtained from bootstrapping

**Return type** numpy array, 1-dim

`trophy.analysis_tools.statistics.crosscorr(x, y, Nmax=None)`

Calculates time-lagged crosscorrelation function of masked array.

**Parameters**

- **x** (*numpy array*) – 1st masked input array, evaluated at  $t + \tau$
- **y** (*numpy array*) – 2nd masked input array
- **Nmax** (*int*, optional, default = None) – maximum lag Nmax is set to `len(x) - 1` if `Nmax == None`

**Returns** **c** – autocorrelation function for lag 0 ... Nmax,

**Return type** numpy array

`trophy.analysis_tools.statistics.cumsum_data_fraction(h, divide_by_total_sum=True)`

The function uses iso-lines of equal density and maps the fraction of data enclosed by these lines onto it.

**Parameters**

- **h** (*numpy array (n-dim)*) – histogram / density or other variable for which cumsum makes sense
- **divide\_by\_total\_sum** (*bool*, optional, default = True) – switch if cumsum field should be divided by total sum this generates a relative field with range (0, 1)

**Returns** **f** – data fraction field

**Return type** numpy array (n-dim)

`trophy.analysis_tools.statistics.draw_from_empirical_1ddist(bins, hist, Nsamp=100)`

Draw random number from an empirical distribution. Implemented for 1d histograms.

**Parameters**

- **bins** (*numpy array*) – bins edges (1dim)
- **hist** (*numpy array (1-dim)*) – histogram values (either absolute frequencies or relative)
- **Nsamp** (*int*, optional, default = 100) – number of random samples that are drawn

**Returns** **rvalues** – random values that are distributed like hist

**Return type** numpy array

`trophy.analysis_tools.statistics.draw_from_empirical_dist(bins, hist, Nsamp=100, discrete_version=False)`

Draw random number from an empirical distribution. Implemented for 2d histograms.

**Parameters**

- **bins** (*list of two numpy arrays*) – bins edges (2dim)
- **hist** (*numpy array (2-dim)*) – histogram values (either absolute frequencies or relative)
- **Nsamp** (*int, optional, default = 100*) – number of random samples that are drawn
- **discrete\_version** (*bool, optional, default = False*) – if True: samples can only be placed at bin midpoints if False: samples are uniformly distributed within each bin

**Returns** **rvalues** – random values (2d) that are distributed like hist

**Return type** list of two numpy arrays

`trophy.analysis_tools.statistics.fdistrib_mapping(fin, fmap, keep_range=False)`

Performs a transformation of field fin to have the same distribution as fmap.

**Parameters**

- **fin** (*numpy array (n-dim)*) – input field
- **fmap** (*numpy array (n-dim)*) – field from which distribution is taken
- **keep\_range** (*bool, optional, default = False*) – switch if output field is scaled to input range (with min-max transformation)

**Returns** **fout** – transformed field fin

**Return type** numpy array (n-dim)

`trophy.analysis_tools.statistics.get_outlier_from_residuals(e, thresh=3.0)`

Calculate outliers from residuals using percentile-based standardization.

**Parameters**

- **e** (*numpy array*) – input residuals
- **thresh** (*float, optional, default = 3.*) – value of residual (standardized) at which outliers are identified

**Returns** **m** – outlier mask

**Return type** numpy array

`trophy.analysis_tools.statistics.normalize_field(f, vmin=None, vmax=None)`

Normalization of a field is applied.

**Parameters**

- **f** (*numpy array, 2dim*) – 2dim field
- **vmin** (*float, optional, default = None*) – lower bound of fields for rescaling
- **vmax** (*float, optional, default = None*) – upper bound of fields for rescaling

**Returns** **f\_norm** – 2dim field

**Return type** numpy array, 2dim

`tropy.analysis_tools.statistics.odrfit_with_outlier_removal` (*func*, *x*, *y*, *xerr*=0, *yerr*=0, *n\_iter*=10, *outlier\_thresh*=3.0, *p0*=None)

Calculates fitting parameter object using a orthogonal distance regression.

(x,y) are interpreted as mean values and error bars as standard deviations of 2d Gaussian distribution per point. The range of possible fits is analyzed for their percentiles.

#### Parameters

- **func** (*function*) – fitting function `func(args, x)`, args: parameters of the function ! ARGVS HAVE BEEN TURNED !
- **x** (*numpy array, 1-dim*) – independent values
- **y** (*numpy array, 1-dim*) – dependent values
- **xerr** (*float or numpy array (1-dim), optional, default = 0.*) – standard deviation of x
- **yerr** (*float or numpy array (1-dim), optional, default = 0.*) – standard deviation of y
- **n\_iter** (*int, optional, default = 10*) – numbers of iteration used for outlier removal
- **outlier\_thresh** (*float, optional, default = 3.*) – threshold for outlier removal (standardized fit residuals)
- **p0** (*list or tuple, optional, default = None*) – first guess of fitted parameters (for increased convergence)

**Returns** `fit_result` – result of the ODR fit

**Return type** “`scipy.odr.ODR`”

`tropy.analysis_tools.statistics.rank_transformation` (*f*, *normalize*=True, *gamma*=1.0)

The routine performs rank transformation of a field, meaning that the ranks of the individual field values are returned.

#### Parameters

- **f** (*numpy array (n-dim)*) – field to be transformed
- **normalize** (*bool, optional, default = True*) – option if the rank field should be normalized to one
- **gamma** (*float, optional, default = 1.*) – gamma correction factor

**Returns** `f` – field ranks

**Return type** `numpy array (n-dim)`

## 2.1.7 tropy.analysis\_tools.thermodynamic\_variables

Package for the calculation of several thermodynamical properties of moist air (including condensate).

#### ToDo

- Some final tests of the routines should be done.
- There was an issue with mixing ratio vs. specific humidity, but might be resolved ...



`trophy.analysis_tools.thermodynamic_variables.H2r(hrel, p, T)`

Converts relative humidity H into mixing ratio r.

**Parameters**

- **hrel** (*numpy array*) – relative humidity in [%]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]

**Returns** **es** – water vapor pressure in [Pa]

**Return type** *numpy array*

`trophy.analysis_tools.thermodynamic_variables.absolute_humidity(r, p, T)`

Calculates absolute humidity.

**Parameters**

- **r** (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]

**Returns** **rho\_v** – absolute humidity / m\*\*3]

**Return type** *numpy array*

`trophy.analysis_tools.thermodynamic_variables.both_mass_fractions2mixing_ratios(qv, qw)`

Converts water and the vapor mass fraction to condensed water and vapor mixing ratio.

**Parameters**

- **qv** (*numpy array*) – water vapor mass fraction [kg / kg]
- **qw** (*numpy array*) – condensed water mass fraction [kg / kg]

**Returns**

- **r** (*numpy array*) – water vapor mixing ratio [kg / kg]
- **rw** (*numpy array*) – condensed water mixing ratio [kg / kg]

`trophy.analysis_tools.thermodynamic_variables.dew_point(r, p, T)`

Calculates dew point temperature after Markowski book p 13 eq (2.25)

**Parameters**

- **r** (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]

**Returns** **Td** – dew point temperature in [K]

**Return type** *numpy array*

`trophy.analysis_tools.thermodynamic_variables.dry_air_density(r, p, T)`

Calculates dry air density.

**Parameters**

- **r** (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]

- **T** (*numpy array*) – temperature [K]

**Returns rho\_d** – dry air density in [kg / m\*\*3]

**Return type** numpy array

`trophy.analysis_tools.thermodynamic_variables.dry_air_potential_temperature` (*r*,  
*p*,  
*T*)

Calculates dry air potential temperature.

**Parameters**

- **r** (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]

**Returns theta\_d** – dry air potential temperature (K)

**Return type** numpy array

`trophy.analysis_tools.thermodynamic_variables.equivalent_potential_temperature` (*r*,  
*p*,  
*T*,  
*eq=15*)

Calculates equivalent potential temperature. after Bolton (1980) MWR 108, p.1046, eq. (43)

Davies-Jones (2009) MWR137, eq. (6.3)

also using either one of (15), (21) or (22)

**Parameters**

- **r** (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]
- **eq** (*int, optional, default = 15*) – which eq. of Bolton (1980) to be chosen for lifting condensation level temperature eq in [15, 21, 22]

**Returns theta\_E** – equivalent potential temperature

**Return type** numpy array

`trophy.analysis_tools.thermodynamic_variables.lifting_condensation_level_temperature` (*r*,  
*p*,  
*T*,  
*eq=15*)

Calculates lifting condensation level temperature. after Bolton (1980) MWR 108, p.1046, eq. (15), (21) or (22)

**Parameters**

- **r** (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]
- **eq** (*int, optional, default = 15*) – which eq. of Bolton (1980) to be chosen eq in [15, 21, 22]

**Returns TL** – lifting condensation level temperature in [K]

**Return type** numpy array

`trophy.analysis_tools.thermodynamic_variables.moist_gas_constant(r)`

Calculates gas constant of moist air.

**Parameters** `r` (*float* or *numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]

**Returns** `R_m` – gas constant of moist air [J / kg / K]

**Return type** *float* or *numpy array*

`trophy.analysis_tools.thermodynamic_variables.moist_potential_temperature(r, p, T)`

Calculates moist air potential temperature.

**Parameters**

- `r` (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- `p` (*numpy array*) – total gas pressure [Pa]
- `T` (*numpy array*) – temperature [K]

**Returns** `theta_m` – moist air potential temperature (K)

**Return type** *numpy array*

`trophy.analysis_tools.thermodynamic_variables.moist_specific_heat(r)`

Calculates specific heat at constant pressure of moist air.

**Parameters** `r` (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]

**Returns** `c_pm` – specific heat at constant pressure of moist air

**Return type** *numpy array*

`trophy.analysis_tools.thermodynamic_variables.relative_humidity(r, p, T)`

Calculates relative humidity.

**Parameters**

- `r` (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- `p` (*numpy array*) – total gas pressure [Pa]
- `T` (*numpy array*) – temperature [K]

**Returns** `H` – relative humidity in [%]

**Return type** *numpy array*

`trophy.analysis_tools.thermodynamic_variables.saturation_over_ice(T, method='PK')`

Calculates saturation water vapor pressure over ice.

**Parameters**

- `T` (*numpy array*) – temperature [K]
- `method` (*str*, optional, default = 'PK') – use an equation which approximates the vapor pressure over ice method in ['PK', 'Murphy']

**Returns** `es` – water vapor pressure in [Pa]

**Return type** *numpy array*

## Notes

‘PK’ refers to Pruppacher and Klett (1997) ‘Murphy’ refers to Murphy and Koop (2005) eq. (7)

## References

Murphy, D. M., and T. Koop (2005), Review of the vapour pressures of ice and supercooled water for atmospheric applications, Quart. J. Roy. Meteor. Soc., 131(608), 1539-1565, doi:10.1256/qj.04.94.

`trophy.analysis_tools.thermodynamic_variables.saturation_pressure(T)`

Calculates saturation water vapor pressure after Bolton (1980) MWR 108, p.1046, eq. (10)

**Parameters** `T` (*numpy array*) – temperature [K]

**Returns** `es` – water vapor pressure in [Pa]

**Return type** `numpy array`

`trophy.analysis_tools.thermodynamic_variables.specific_humidity(r)`

Calculates specific humidity.

**Parameters** `r` (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]

**Returns** `q` – specific humidity i.e. vapor mass per moist air mass [kg / kg]

**Return type** `numpy array`

`trophy.analysis_tools.thermodynamic_variables.thermodynamic_constants()`

Sets some thermodynamic constants.

**Parameters** `None` –

**Returns** `const` – set of thermodynamic constants saved in dictionary

**Return type** `dict`

`trophy.analysis_tools.thermodynamic_variables.total_density(r, r_w, p, T)`

Calculates total density of moist air with hydrometeors.

**Parameters**

- `r` (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- `r_w` (*numpy array*) – mixing ratio of condensed water i.e. condensed water mass per dry air mass in [kg / kg]
- `p` (*numpy array*) – total gas pressure [Pa]
- `T` (*numpy array*) – temperature [K]

**Returns** `rho_d` – dry air density in [kg / m\*\*3]

**Return type** `numpy array`

`trophy.analysis_tools.thermodynamic_variables.water_vapor_pressure(r, p, T)`

Calculates water vapor pressure.

**Parameters**

- `r` (*numpy array*) – mixing ratio i.e. vapor mass per dry air mass in [kg / kg]
- `p` (*numpy array*) – total gas pressure [Pa]
- `T` (*numpy array*) – temperature [K]

**Returns** `e` – water vapor pressure in [Pa]

**Return type** numpy array

`trophy.analysis_tools.thermodynamic_variables.watermass_fraction2rw(qw, r)`  
 Converts water mass fraction to condensed water mixing ratio.

**Parameters**

- **qw** (*numpy array*) – condensed water mass fraction [kg / kg]
- **r** (*numpy array*) – mixing ration of water vapor [kg / kg]
- **p** (*numpy array*) – total gas pressure [Pa]
- **T** (*numpy array*) – temperature [K]

**Returns** **rw** – condensed water mixing ratio [kg / kg]

**Return type** numpy array

## 2.2 trophy.io\_tools package

### 2.2.1 Submodules

### 2.2.2 trophy.io\_tools.HRIT module

`trophy.io_tools.HRIT.SEVIRI_channel_list()`

`trophy.io_tools.HRIT.bit_conversion(input_set, inbit=8, outbit=10)`

An integer array which is based on a n-bit representation is converted to an integer array based on a m-bit representation.

`output_set = bit_conversion(input_set, inbit=8, outbit=10)`

`input_set` : numpy integer array in a n-bit representation  
`inbit` : number of bits for `input_set` (n)  
`outbit` : number of bits for `output_set` (m)

`output_set` : numpy integer array in a m-bit representation

The size of `input_set` should be a multiple of `inbit` / `outbit`!

`trophy.io_tools.HRIT.channel_segment_sets(set_name)`

Outputs a predefined set of channels and segments'

`chan_seg = channel_segment_sets(set_name)`

`set_name`: name of a predefined set

`chan_seg`: Dictionary of channels with a list of segments.

`trophy.io_tools.HRIT.combine_segments(seg_dict, missing=-1)`

Combines individual segments of SEVIRI scans. Segments will be sorted by key.

`combined = combine_two_segments(seg_list)`

`seg_list`: dictionary of segment, sorted by key from bottom to top

`combined`: combination of both segments

`trophy.io_tools.HRIT.combine_two_segments(seg1, seg2)`

Combines two segments where `seg1` is the upper and `seg2` the lower segment.

`seg = combine_two_segments(seg1, seg2)`

`seg1`: upper segment `seg2`: lower segment

seg: combination of both segments

`tropy.io_tools.HRIT.divide_two_segments(seg)`

Divides one segment into two with half the line size.

`seg1, seg2 = divide_two_segments(seg)`

seg: combination of two segments

seg1: upper segment seg2: lower segment

`tropy.io_tools.HRIT.get_HRIT_from_arch(day, chan_seg={'IR_016': [1, 2, 3, 4, 5, 6, 7, 8],  
'VIS006': [1, 2, 3, 4, 5, 6, 7, 8], 'VIS008': [1, 2,  
3, 4, 5, 6, 7, 8]}, scan_type='pzs', tarname=None,  
arch_dir=None, out_path=None)`

Gets and decompresses the original HRIT files from archive. A typical set of channels and segments can be chosen.

`file_list = get_HRIT_from_arch(day, chan_set = 'nc-full', scan_type = 'pzs' , arch_dir = None, out_path = None)`

day: datetime object which include day and time of MSG time slot  
chan\_set = name for a channel-segment set  
scan\_type: which type of msg is used,

- ) allowed keys are: 'pzs' and 'rss'

arch\_dir = archive directory out\_path = directory where the HRIT files are saved.

file\_list : list of extracted HRIT files

`tropy.io_tools.HRIT.read_HRIT_data(day, chan_seg, calibrate=True, scan_type='pzs',  
arch_dir=None, standard_seg=(464, 3712),  
add_meta=False, **kwargs)`

Read HRIT data given a time slot, satellite type and channel-segment set

`combined = read_HRIT_data(day, chan_seg, calibrate=True, scan_type = 'pzs', arch_dir = None, add_meta = False, **kwargs)`

day = datetime object which include day and time of MSG time slot  
chan\_seg = a channel-segment set, dictionary of channel names with lists of segments  
calibrate = flag, if slope / offset calibration should be done  
scan\_type = which type of msg is used,

- ) allowed keys are: 'pzs' and 'rss'

arch\_dir = archive directory add\_meta = flag, if meta data should be added  
out\_path = directory where the HRIT files are saved.

combined = dictionary of retrieved channels with combined segments

`tropy.io_tools.HRIT.read_HRIT_seg_data(hrit_file)`

Reads directly the counts data for an HRIT segment.

`seg_data = read_HRIT_seg_data(hrit_file)`

hrit\_file : HRIT file of the considered segment

seg\_data: integer segment data field of radiance counts

`tropy.io_tools.HRIT.read_slope_offset_from_prolog(pro_file, NJUMP=387065)`

Reads directly slope and offset from HRIT PRO file.

`slope, offset = read_slope_offset_from_prolog(pro_file)`

pro\_file : PROLOG file of the considered time slot

slope: slope of the data offset: offset of the data

Following transformation is applied:

$DATA = SLOPE * COUNTS + OFFSET$

if COUNTS are valid data points!

!! ATTENTION !! Jump in binary file is hard-coded!

```
trophy.io_tools.HRIT.segments_for_region(region, Nrow=3712, Nseg=8, seg_size=464)
```

```
trophy.io_tools.HRIT.write_HRIT_seg_data(seg_data, hrit_file)
```

Reads directly the counts data for an HRIT segment.

```
write_HRIT_seg_data(seg_data, hrit_file)
```

hrit\_file : HRIT file of the considered segment seg\_data: integer segment data field of radiance counts

### 2.2.3 trophy.io\_tools.bit\_conversion module

### 2.2.4 trophy.io\_tools.data\_collection module

```
class trophy.io_tools.data_collection.DataCollection
```

Bases: dict

```
add(vname, array, setting='bt', subpath=None)
```

```
list()
```

```
save(fname)
```

```
class trophy.io_tools.data_collection.Dataset(name, data=None, setting={'_FillValue':
0, 'add_offset': 0, 'dtype': 'lu2',
'longname': 'Brightness Temperature',
'scale_factor': 0.01, 'unit': 'K'})
```

Bases: object

The Dataset class is a building block of the DataCollection class which will be used to easy input and output dataset sets.

```
add(array)
```

```
array()
```

```
attrs()
```

```
load(fname, subpath=None)
```

```
save(fname, subpath=None, mode='overwrite', time_out=300.0)
```

### 2.2.5 trophy.io\_tools.data\_settings module

```
trophy.io_tools.data_settings.Dge_settings()
```

```
trophy.io_tools.data_settings.bt_settings()
```

```
trophy.io_tools.data_settings.histogram_settings()
```

```
trophy.io_tools.data_settings.refl_settings()
```

```
trophy.io_tools.data_settings.settings(vartype='bt')
```

```
trophy.io_tools.data_settings.standard_int_setting(longname='Unknown',
unit='Unknown')
```

`trophy.io_tools.data_settings.standard_real_setting` (*longname='Unknown', unit='Unknown'*)

## 2.2.6 trophy.io\_tools.file\_status module

`trophy.io_tools.file_status.test` (*fname*)

Returns status of a file:

True: if open False: if not open

## 2.2.7 trophy.io\_tools.find\_latest\_slot module

## 2.2.8 trophy.io\_tools.hdf module

`trophy.io_tools.hdf.dict_merge` (*a, b*)

Recursively merges dict's. not just simple `a['key'] = b['key']`, if both a and b have a key whose value is a dict then `dict_merge` is called on both values and the result stored in the returned dictionary.

from <https://www.xormedia.com/recursively-merge-dictionaries-in-python/>

`trophy.io_tools.hdf.get_seviri_chan` (*chan\_list, day, scan\_type='rss', fname=None, calibrate=True, add\_meta=False, add\_geo=False, arch\_dir=None*)

Reads MSG - SEVIRI radiance of a given channel for a given date.

`var = get_seviri_chan(chan_list, day, scan_type = 'rss', calibrate=True, add_meta = False)`

**chan\_list:** name of channel, e.g. **ir\_108** or **IR\_108** (case does not matter) or list of channel names

**day:** datetime object which include day and time of MSG time slot

**scan\_type:** sets of scanning modus, i.e. 'rss' or 'pzs' (DEFAULT: 'rss')

**calibrate :** optional, decides if output is radiance (True) or counts (False)

**add\_meta:** meta data of the MSG-SEVIRI hdf file

**var:** dictionary including radiance/counts of channel **<ch\_name>** at date **<day>** and if chosen meta data

`trophy.io_tools.hdf.list_hdf_groups` (*fname*)

Makes a list of hdf groups.

Only the 1st level is implemented, TDB: recursive listing.

`glist = list_hdf_groups(fname)`

**fname:** filename of hdf file

**glist:** list of group names

`trophy.io_tools.hdf.read_dict_cont` (*f, d*)

Recursively read nested dictionaries.

`trophy.io_tools.hdf.read_dict_from_hdf` (*fname*)

The content of an hdf file with arbitrary depth of subgroups is saved in nested dictionaries.

`d = read_dict_from_hdf(fname)`

**fname:** filename of output hdf file

**d:** nested dictionary which contains the data



`trophy.io_tools.hdf.read_var_from_hdf(fname, vname, subpath=None)`  
 A specific variable is read from an hdf file. Group and subgroups can be specified.  
`v = read_var_from_hdf(fname, vname, subpath = None)`  
 fname: filename of input hdf file vname: variable name as given in hdf file  
 subpath: group or subgroup path with /  
 v: variable as numpy array

`trophy.io_tools.hdf.save_dict2hdf(fname, d, mode='w')`  
 The content of nested dictionaries of arbitrary depth is saved into an hdf file.  
 The key, subkeys, etc. are mapped into the hdf-groups directory structure.  
`save_dict2hdf(fname, d)`  
 fname: filename of output hdf file d: dictionary which contains the data  
 None

`trophy.io_tools.hdf.save_dict_cont(f, d)`  
 Recursively saves nested dictionaries.

`trophy.io_tools.hdf.update_dict_in_hdf(fname, din)`  
 The content of nested dictionaries of arbitrary depth is updated in an hdf file.  
 The key, subkeys, etc. are mapped into the hdf-groups directory structure.  
`update_dict_in_hdf(fname, d)`  
 fname: filename of output hdf file d: dictionary which contains the data  
 None

## 2.2.9 trophy.io\_tools.netcdf module

`trophy.io_tools.netcdf.read_icon_2d_data(fname, var_list, itime=0)`  
`trophy.io_tools.netcdf.read_icon_4d_data(fname, var_list, itime=0, itime2=None)`  
 Read netcdf data.  
`dset = read_icon_4d_data(fname, var_list, itime = 0)`  
 fname: netcdf filename var\_list: variable name or list of variables itime (OPTIONAL): index of 1st dimension (assumed to be time) to be read only  
 if itime = None, the full field is read  
 dset: dictionary containing the fields  
`trophy.io_tools.netcdf.read_icon_dimension(fname, dim_name)`  
`trophy.io_tools.netcdf.read_icon_georef(fname)`  
`trophy.io_tools.netcdf.read_icon_time(fname, itime=0)`  
`trophy.io_tools.netcdf.roundTime(dt=None, roundTo=60)`  
**Round a datetime object to any time laps in seconds** dt : datetime.datetime object, default now.  
 roundTo : Closest number of seconds to round to, default 1 minute. Author: Thierry Husson 2012 - Use it as you want but don't blame me.  
`trophy.io_tools.netcdf.save_icon_georef(fname, geopath=None)`

```
trophy.io_tools.netcdf.save_icon_time_reference(fname, outfile=None)
```

## 2.2.10 trophy.io\_tools.radolan module

### 2.2.11 Module contents

## 2.3 trophy.l15\_msevi package

### 2.3.1 Submodules

#### 2.3.2 trophy.l15\_msevi.msevi module

#### 2.3.3 trophy.l15\_msevi.msevi\_config module

#### 2.3.4 trophy.l15\_msevi.msevi\_rgb module

### 2.3.5 Module contents

## 2.4 trophy.plotting\_tools package

### 2.4.1 Submodules

#### 2.4.2 trophy.plotting\_tools.bmaps module

#### 2.4.3 trophy.plotting\_tools.colormaps module

```
trophy.plotting_tools.colormaps.colormap_based_cmap(colname, start_col=None,
                                                    final_col=None, reverse=False)
```

A matplotlib colormap is constructed based on one color'

**cmap = colormap\_based\_cmap(colname, start\_col = None, final\_col = None, reverse = False)**

colname : supported colname as basis for colormap start\_col : color at one end of the colormap final\_col : color at the other end of the colormap reverse: option if order is reversed

cmap: resulting colormap

```
trophy.plotting_tools.colormaps.dwd_sfmap()
```

```
trophy.plotting_tools.colormaps.enhanced_colormap(vmin=200.0, vmed=240.0,
                                                    vmax=300.0)
```

```
trophy.plotting_tools.colormaps.enhanced_wv62_cmap(vmin=200.0, vmed1=230.0,
                                                    vmed2=240.0, vmax=260.0)
```

```
trophy.plotting_tools.colormaps.nice_cmaps(cmap_name)
```

#### 2.4.4 trophy.plotting\_tools.compare module

```
trophy.plotting_tools.compare.compare(*args, **kwargs)
```

The function compare can be used to fastly compare several 2d fields.

The coordinates of the x and y axis are needed. An arbitrary number of fields is plotted using the function shaded. Corresponding option keywords can be used.

`trophy.plotting_tools.compare.compare_shaded(x, y, *args, **kwargs)`

The function compare can be used to fastly compare several 2d fields.

The coordinates of the x and y axis are needed. An arbitrary number of fields is plotted using the function shaded. Corresponding option keywords can be used.

## 2.4.5 trophy.plotting\_tools.histograms module

Module for histogram calculations and plotting.

Histogram calculations include \* partial normalization to get marginal distributions \* conditional averages, standard deviations and percentiles

Histogram plotting is for \* conditioned histograms with averages or percentiles

`trophy.plotting_tools.histograms.ave_sig_from_hist(xe, ye, h)`

Use output from the numpy 2d histogram routine to calculate y-mean and standard deviations along y direction.

### Parameters

- **xe** (*np.array*) – x-part of histogram grid (edge values)
- **ye** (*np.array*) – y-part of histogram grid (edge values)
- **h** (*np.array*) – frequency of occurrence

### Returns

- **yave** (*np.array*) – mean of y weighed by h along y-direction
- **ysig** (*np.array*) – standard deviation of y weighed by h along y-direction

`trophy.plotting_tools.histograms.axis_average_from_hist3d(bins3d, h, axis=0)`

Calculates the average of a 3-dim histogram along a selected axis.

### Parameters

- **bins3d** (*list of 3dim np.array*) – mesh of bin edges
- **h** (*np.array*) – absolute histogram counts
- **axis** (*{0, 1}, optional*) – axis along which the calculations are performed

**Returns** **ave** – conditional average field

**Return type** *np.array*

`trophy.plotting_tools.histograms.conditioned_hist(h, axis=0)`

Make conditioned histogram.

Prob per bin (not PDF!!!).

**Parameters** **h** (*np.array*) – absolute frequency of occurrence

**Returns** “normalized” frequency

**Return type** *np.array*

`trophy.plotting_tools.histograms.hist2d_scatter(x, y, bins=200, **kwargs)`

A 2d histogram is constructed and displayed as scatter plot.

### Parameters

- **x** (*np.array*) – 1st data vector

- **y** (*np.array*) – 2nd data vector

#### Returns

- **hxy** (*np.array*) – frequency of occurrence
- **xs** (*np.array*) – x-part of histogram grid (edge values)
- **ys** (*np.array*) – y-part of histogram grid (edge values)

`trophy.plotting_tools.histograms.hist3d(v1, v2, v3, bins)`

A wrapper for 3d histogram binning.

It additionally generates the average bin values with the same 3d shape as histogram itself.

#### Parameters

- **v1** (*np.array*) – 1st data vector
- **v2** (*np.array*) – 2nd data vector
- **v3** (*np.array*) – 3rd data vector
- **bins** (*list of 3 np.arrays*) – bins argument passed to the `np.histogramdd` function

#### Returns

- **bins3d** (*list of 3dim np.array*) – mesh of bin edges
- **h** (*np.array*) – absolute histogram counts

`trophy.plotting_tools.histograms.max_from_hist(xe, ye, h)`

Use output from the numpy 2d histogram routine to calculate y-positions where maximum occurrences are located.

#### Parameters

- **xe** (*np.array*) – x-part of histogram grid (edge values)
- **ye** (*np.array*) – y-part of histogram grid (edge values)
- **h** (*np.array*) – frequency of occurrence

**Returns** **ymax** – y-position where h is maximal along y-direction

**Return type** `np.array`

`trophy.plotting_tools.histograms.percentiles_from_hist(xe, ye, h, p=[25, 50, 75], axis=0, sig=0)`

Use output from the numpy 2d histogram routine to calculate percentiles of the y variable based on relative occurrence rates.

#### Parameters

- **xe** (*np.array*) – x-part of histogram grid (edge values)
- **ye** (*np.array*) – y-part of histogram grid (edge values)
- **h** (*np.array*) – frequency of occurrence
- **p** (*list, optional, default = [25, 50, 75]*) – list of percentile values to be calculated (in 100th)
- **axis** (*{0, 1}, optional*) – axis along which the calculations are performed
- **sig** (*float, optional, default = 0*) – sigma of a Gaussian filter applied to the histogram in advance

Should be non-negative.

**Returns** `yperc` – list of arrays containing the percentiles

**Return type** `np.array`

`trophy.plotting_tools.histograms.plot_cond_hist` (*xe, ye, h, ax, axis=0, logscale=True, \*\*kwargs*)

Plot conditioned histogram (marginal distribution).

#### Parameters

- **xe** (*np.array*) – x-part of histogram grid (edge values)
- **ye** (*np.array*) – y-part of histogram grid (edge values)
- **h** (*np.array*) – frequency of occurrence
- **ax** (*plt.axes instance*) – a current axes where the plot is placed in
- **axis** (*{0, 1}, optional*) – axis along which the calculations are performed
- **logscale** (*{True, False}, optional*) – if histogram is plotted with logarithmic colorscale
- **\*\*kwargs** (*dict*) – further optional keywords passed to `plt.pcolormesh`

**Returns** `pcm`

**Return type** `plt.pcolormesh instance`

`trophy.plotting_tools.histograms.plot_hist_median_and_intervals` (*xe, ye, h, ax*)

Plot histogram with median and IQR. (axis = 1, fixed).

#### Parameters

- **xe** (*np.array*) – x-part of histogram grid (edge values)
- **ye** (*np.array*) – y-part of histogram grid (edge values)
- **h** (*np.array*) – frequency of occurrence
- **ax** (*plt.axes instance*) – a current axes where the plot is placed in

**Returns** `ax` – a current axes where the plot is placed in

**Return type** `plt.axes instance`

## 2.4.6 trophy.plotting\_tools.meta2png module

Module designed to write an Author Name and the Source Filename into the Meta-Data of an PNG file saved with matplotlib.

`trophy.plotting_tools.meta2png.meta2png` (*fname, meta*)

Saves meta data into image file.

#### Parameters

- **fname** (*str*) – name of png file
- **meta** (*dict*) – collection of extra meta data to be stored in image file

**class** `trophy.plotting_tools.meta2png.pngsave` (*\*args, \*\*kwargs*)

Bases: `object`

That class is designed to save pylab figures in png and add meta data.

#### Parameters

- **\*args** (*list*) – other positional arguments passed to *plt.savefig*
  - **\*\*kwargs** (*dict*) – other optional arguments passed to *plt.savefig*
- 'author' : Place your name into author keyword
- 'source' [Specify your source filename if needed,] if not set, it tries to automatically find the filename

**meta()**

Set the meta data, esp. the Author name and Source file information.

## 2.4.7 tropy.plotting\_tools.shaded module

A module to make shaded plots.

It contains a function for non-linear colormaps.

**class** tropy.plotting\_tools.shaded.**n1cmap**(*cmap, levels*)  
Bases: matplotlib.colors.LinearSegmentedColormap

A nonlinear colormap class.

### Parameters

- **cmap** (*matplotlib.cmap*) – a matplotlib colormap, e.g. matplotlib.cm.jet
- **levels** (*list or np.array*) – list of values where different colors should appear

### Notes

Derived from the matplotlib.colors.LinearSegmentedColormap

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**name** = 'n1cmap'

tropy.plotting\_tools.shaded.**set\_levs**(*nmax, depth, largest=5, sym=True, sign=True, zmax='None'*)

Makes a non-linear level set based on pre-defined numbers.

### Parameters

- **nmax** (*int*) – exponent of maximum number
- **depth** (*int*) – number of iterations used down to scales of  $10^{**}(\text{nmax} - \text{depth})$
- **largest** (*{5, 8}, optional*) – set the largest number in the base array either to 5 or 8
- **sym** (*{True, False}, optional*) – switch if levels are symmetric around origin
- **sign** (*{True, False}, optional*) – switches sign if negative
- **zmax** (*float, optional, default = 'none'*) – limiter for levels, **llevsl** > zmax are not allowed

**Returns** **levs** – set of non-linear levels

**Return type** np.array

tropy.plotting\_tools.shaded.**shaded**(*x, y, z, \*args, \*\*kwargs*)

The function shaded is a wrapper for the pylab function *contourf*.

In addition to *contourf*, shaded can plot filled contours for which a non-linear colormap is used.

**Parameters**

- **x** (*np.array*) – x-values passed to *plt.contourf*
- **y** (*np.array*) – y-values passed to *plt.contourf*
- **z** (*np.array*) – z-values passed to *plt.contourf* (color value)
- **\*args** (*list*) – other positional arguments passed to *plt.contourf*
- **\*\*kwargs** (*dict*) – other optional arguments passed to *plt.contourf*

special keywords:

- 'levels' : numpy array of color / contour levels
- 'lev\_depth' : gives the depth of an automatically generated level set i.e. an initial base (e.g. [1,2,3,5] ) that contains the maximum value of the field (e.g. 4.3) is downscaled by  $10^{*-1}, \dots, 10^{*-lev\_depth}$ .

Example: Given `lev_depth = 2`, for a positive field with maximum 4.3 a level set [0.01, 0.02, 0.03, 0.05, 0.1, 0.2, 0.3, 0.5, 1, 2, 3, 5] is generated.

**Returns**

**Return type** `pl.contourf` instance





## 3.1 tropy package

### 3.1.1 Submodules

`tropy.MSGtools.channel_segment_sets(set_name)`

Outputs a predefined set of channels and segments'

**Parameters** `set_name` (*str*) – name of a predefined set

**Returns** `chan_seg` – Dictionary of channels with a list of segments.

**Return type** *dict*

`tropy.MSGtools.get_HRIT_from_arch(day, chan_set='nc-full', scan_type='pzs', arch_dir=None, out_path=None)`

Gets and decompresses the original HRIT files from archive. A typical set of channels and segments can be chosen.

**Parameters**

- **day** (*datetime object*) – day and time of MSG time slot
- **chan\_set** (*str, optional, default = 'nc-full'*) – name of a predefined channels set
- **scan\_type** (*str, optional, default = 'pzs'*) – one of the two Meteosat scan types
- **arch\_dir** (*str, optional, default = None*) – archive directory where NWC-SAF files are saved
- **out\_path** (*str, optional, default = None*) – directory where the HRIT files are saved.

**Returns** `file_list` – list of extracted HRIT files

**Return type** *list*

`trophy.MSGtools.get_berendes_prod(prod, t, region='de', scan_type='rss', arch='none')`

Reads product of Berendes Cloud type product for a given date.

#### Parameters

- **prod** (*str*) – name of Berendes product, either ccm or texture
- **t** (*datetime object*) – day and time of MSG time slot
- **region** (*str, optional, default = 'de'*) – string that defines regional cutout
- **scan\_type** (*str, optional, default = 'rss'*) –
- **arch** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved (OPTIONAL)

**Returns** *var* – Berendes product as field

**Return type** *numpy array*

`trophy.MSGtools.get_cmsaf_prod(prod, day, scan_type='rss', region='eu', arch='none', calibrate=False, switch2new=True)`

Reads product of CM-SAF (KNMI retrieval) for a given date.

#### Parameters

- **prod** (*str*) – name of CMSAF product
- **day** (*datetime object*) – day and time of MSG time slot
- **region** (*str, optional, default = 'eu'*) – string that defines regional cutout
- **scan\_type** (*str, optional, default = 'rss'*) – one of the two Meteosat scan types
- **arch** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved
- **calibrate** (*bool, optional, default = False*) – switch for slope/offset calibration
- **switch2new** (*bool, optional, default = True*) – if the “new” set of CMSAF products should be selected “new” means collection two: “c2”

#### Returns

- **product** (*numpy array, optional*) – calibrated CMSAF product data (if calibrate == True)
- **counts** (*numpy array, optional*) – CMSAF product count data (if calibrate == False)
- **scal** (*float, optional*) – scale factor for CMSAF products (if calibrate == False)
- **offset** (*float, optional*) – offset for CMSAF products (if calibrate == False)

#### Notes

Data are saved as integer via:  $DATA = scal * count + offset$

The following structures can be returned:

`counts, scal, offset = get_cmsaf_prod(prod, day, calibrate = False)`

OR

`product = get_cmsaf_prod(prod, day, calibrate = True)`

```
trophy.MSGtools.get_cos_zen(day, scan_type='rss', arch='none')
```

Reads sun zenith angle from the partical satellite hdf file and calculates the cosine of the sun zenith.

#### Parameters

- **day** (*datetime object*) – day and time of MSG time slot
- **scan\_type** (*str, optional, default = 'rss'*) – one of the two Meteosat scan types

**Returns** **coszen** – cosine of sun zenith angle

**Return type** numpy array, optional

```
trophy.MSGtools.get_highres_cma(t, region='de', scan_type='rss', arch='none')
```

Reads product of High-Res Cloud Mask product (Bley et al., 2013) for a given date.

#### Parameters

- **t** (*datetime object*) – day and time of MSG time slot
- **region** (*str, optional, default = 'de'*) – string that defines regional cutout
- **scan\_type** (*str, optional, default = 'rss'*) –
- **arch** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved (OPTIONAL)

**Returns** **var** – Cma product as field

**Return type** numpy array

```
trophy.MSGtools.get_msg_lon_lat(region, scan_type='rss', arch_dir='none')
```

Reads longitudes and latitudes of Meteosat8 pixels saved in an auxiliary file in the MSG archive.

#### Parameters

- **region** (*str*) – predefined cutout of MSG fulldisk, e.g. 'eu'
- **scan\_type** (*str, optional, default = 'rss'*) – one of the two Meteosat scan types
- **arch\_dir** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved

#### Returns

- **lon** (*numpy array*) – longitudes of pixels
- **lat** (*numpy array*) – latitudes of pixels

```
trophy.MSGtools.get_msg_lsm(region, scan_type='rss', arch_dir='none')
```

Reads land sea mask of Meteosat8 pixels saved in an auxiliary file in the MSG archive.

#### Parameters

- **region** (*str*) – predefined cutout of MSG fulldisk, e.g. 'eu'
- **scan\_type** (*str*) – which type of msg is used, \* ) allowed keys are: 'pzs', 'rss'
- **arch\_dir** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved

**Returns** **lsm** – land sea mask

**Return type** numpy array

`trophy.MSGtools.get_msg_sat_zen(day, sat_type='rss')`

Reads satellite zenith angle from the partical satellite hdf file.

**Parameters**

- **day** (*datetime object*) – day and time of MSG time slot
- **sat\_type** (*str, optional, default = 'rss'*) – one of the two Meteosat scan types

**Returns** `satzen` – satellite zenith angle

**Return type** `numpy array`

`trophy.MSGtools.get_nwcsaf_prod(prod, day, scan_type='rss', arch='none', region='eu', calibrate=False, is_region_shifted=False)`

Reads product of NWC-SAF for a given date.

**Parameters** **prod** (*str*) – name of NWCSAF product `prod` in ['CMA', 'CMA\_DUST', 'CT', 'CTTH\_EFFECT', 'CTTH\_HEIGHT', 'CTTH\_PRESS', 'CTTH\_QUALITY', 'CTTH\_TEMPER', 'CT\_PHASE']

**day** [*datetime object*] day and time of MSG time slot

**region** [*str, optional, default = 'eu'*] string that defines regional cutout

**scan\_type** [*str, optional, default = 'rss'*] one of the two Meteosat scan types

**arch** [*str, optional, default = "none"*] archive directory where NWC-SAF files are saved

**calibrate** [*bool, optional, default = False*] switch for slope/offset calibration

**is\_region\_shifted** [*bool, optional, default = False*] switch if bugfix for (mistakingly) shifted regions has to be applied

**Returns**

- **product** (*numpy array, optional*) – calibrated NWCSAF product data (if `calibrate == True`)
- **counts** (*numpy array, optional*) – NWCSAF product count data (if `calibrate == False`)
- **scal** (*float, optional*) – scale factor for NWCSAF products (if `calibrate == False`)
- **offset** (*float, optional*) – offset for NWCSAF products (if `calibrate == False`)

**Notes**

Data are saved as integer via:  $DATA = scal * count + offset$

The following structures can be returned:

`counts, scal, offset = get_nwcsaf_prod(prod, day, calibrate = False)`

OR

`product = get_nwcsaf_prod(prod, day, calibrate = True)`

`trophy.MSGtools.get_seviri_chan(ch_name, day, scan_type='rss', arch='none', calibrate=True)`

Reads MSG - SEVIRI radiance of a given channel for a given date.

**Parameters**

- **ch\_name** (*str*) – name of MSG SEVIRI channel, e.g. `ir_108` or `IR_108` (case does not matter)

- **day** (*datetime object*) – day and time of MSG time slot
- **scan\_type** (*str, optional, default = 'rss'*) – one of the two Meteosat scan types
- **arch** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved
- **calibrate** (*bool, optional, default = False*) – switch for slope/offset calibration if True, radiances are output

**Returns** **out\_field** – radiance/counts (depending on calibrate option) of channel <ch\_name> at date <day>

**Return type** numpy array

**info** [dict] meta info list which includes calibration coefficients

`trophy.MSGtools.list_cmsaf_prod(day, arch='none')`

Lists products contained in a specific CM-SAF (KNMI retrieval) file at a given date.

**Parameters**

- **day** (*datetime object*) – day and time of MSG time slot
- **arch** (*str, optional, default = "none"*) – archive directory where NWC-SAF files are saved

**Returns** **L** – list of cmsaf products

**Return type** list

`trophy.MSGtools.prod_reference(prod)`

Given a NWC-SAF product, the routine returns the corresponding substring to address the product file.

**Parameters** **prod** (*str*) – name of NWC-SAF product, e.g. CMa, CT, CRR

**Returns** **name** – substring in the NWC-SAF product file

**Return type** str

## Notes

Further products with its corresponding substrings should be included in future.

`trophy.MSGtools.rad2bt(rad, info)`

Calculates brightness temperature of infrared channels given the radiance and info list containing calibration coefficients.

**Parameters**

- **rad** (*numpy array*) – radiance of an infrared channel
- **info** (*dict*) – dictionary which contains information about calibration coefficients.

**Returns** **bt** – brightness temperature

**Return type** numpy array

`trophy.MSGtools.rad2refl(rad, info, day, coszen)`

Conversion of satellite radiances in reflectances.

**Parameters**

- **rad** (*numpy array*) – radiance of a visible channel

- **info** (*dict*) – dictionary which contains information about calibration coefficients
- **day** (*datetime object*) – day and time of MSG time slot
- **coszen** (*numpy array*) – cosine of sun zenith angle

**Returns** refl – reflectance

**Return type** numpy array

`tropy.MSGtools.read_rad_from_hdf(hfile)`

Reads radiances from synthetic satellite calculation which are intermediately saved in an hdf file.

**Parameters** **hfile** (*str*) – file name of the hdf file where synthetic radiances are stored.

**Returns**

**RAD** – dictionary of radiances

the channel name, e.g. **IR\_108**, is used as key to access the data array (e.g. saved on COSMO grid)

**Return type** dict

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