




MicroPyzzotMet: A Lightweight Python Package for Climate Downscaling

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Summary

Modern reanalysis products provide continuous global climate information extending back decades and into future projections, yet their spatial resolution remains too coarse to represent the meteorological variability imposed by mountain terrain (Fan et al., 2019). For researchers working on snow, glaciers, permafrost, and alpine hydrology, this mismatch remains a persistent limitation: surface energy-balance and mass-balance models depend on local meteorological fields that capture how terrain modulates atmospheric conditions. Generating such fields does not always require complex dynamical downscaling systems, which—despite their accuracy—often demand substantial computational resources and sophisticated model setups.

MicroPyzzotMet addresses this gap with a lightweight downscaling framework focused on practicality and broad usability. Rather than implementing complex physical parameterizations or spatial clustering techniques, it applies a streamlined set of MicroMet-inspired corrections to temperature, radiation, humidity, wind, and precipitation (Liston & Elder, 2006). Because it requires only a minimal set of essential climate variables, the tool can operate with virtually any reanalysis or climate dataset. A further strength is its integration with EarthDataHub, enabling rapid access to ERA5-Land and digital terrain models through the Zarr format, which significantly reduces I/O overhead and speeds up preprocessing.

In contrast to more advanced packages such as TopoPyScale, which is designed for detailed terrain-driven heterogeneity and fine-scale modelling (Filhol et al., 2023), MicroPyzzotMet prioritizes computational efficiency and conceptual clarity. This makes it ideal for large-domain experiments, operational workflows, or rapid prototyping, while still remaining compatible with higher-resolution approaches when more elaborate topographic corrections are required.

Statement of need

MicroPyzzotMet is an open-source Python package for downscaling meteorological variables from reanalysis climate datasets. It is inspired by the MicroMet methodology (Liston & Elder, 2006) but reimplemented in a modern Python framework, making the workflow more accessible, flexible, and easier to integrate in contemporary data-processing pipelines.

The increasing availability of atmospheric reanalyses—such as ERA5 and ERA5-Land at 25 km and 9 km spatial resolution—has enabled a wide range of cryospheric and hydrological studies. However, these coarse spatial grids remain inadequate for mountain regions, where elevation, slope, and aspect strongly modulate near-surface climate. Tools such as TopoPyScale (Filhol et al., 2023), based on the TopoSCALE and TopoSUB approaches (Fiddes & Gruber, 2014; ?), address this limitation by applying sophisticated 3-D interpolation schemes, multi-level

atmospheric corrections, and DEM segmentation into terrain clusters. These methods allow detailed reconstruction of fine-scale meteorological patterns, especially over complex alpine terrain.

Yet not all applications require this level of complexity. For coarser target resolutions (e.g., 250–500 m), for large domains processed at high temporal frequency, or for long climatological time series, the computational and data requirements of advanced downscaling frameworks may become limiting. In these cases, the original MicroMet methodology offers an attractive balance between physical robustness and computational simplicity.

MicroPyzzotMet builds upon this philosophy. It applies lapse-rate corrections, radiative geometry adjustments, vapor-pressure formulations, and precipitation-elevation relationships following MicroMet, using only the set of climate variables typically available in major reanalysis datasets. This makes the tool broadly applicable, lightweight, and extremely fast, while still delivering spatially coherent meteorological fields suitable for surface energy and mass-balance modelling.

A key feature of MicroPyzzotMet is its integration with EarthDataHub ([Earth Data Hub \(DestinE\), 2025](#)), which distributes global reanalysis datasets—including ERA5-Land—preconverted into Zarr format. This enables efficient cloud-native data access and processing through Xarray/Dask, greatly accelerating workflows and reducing storage overhead.

Figure 1 illustrates an example for the Maipo region in Chile, comparing native ERA5-Land fields to MicroPyzzotMet downscaled products for daily air temperature and incoming shortwave radiation.

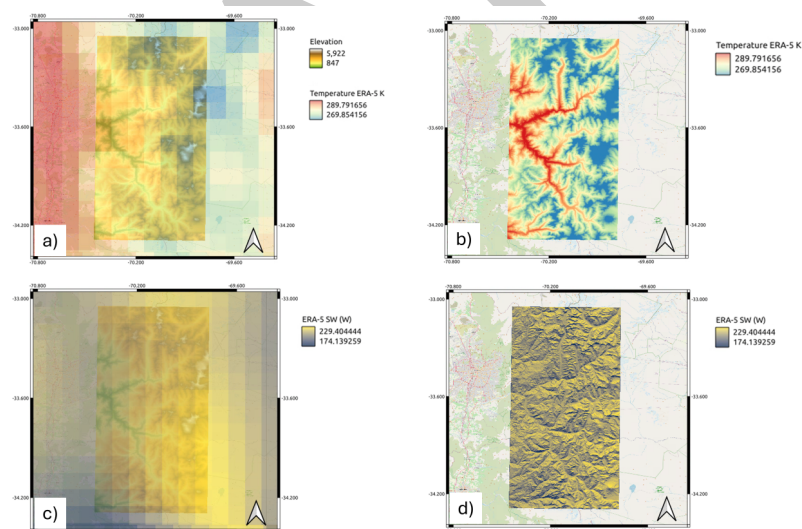


Figure 1: In the figure are presented two examples of downscaled variables—daily air temperature and shortwave incoming radiation—over the Maipo region in Chile on January 1, 2017. Panels a) and c) show the native ERA5-Land fields at 9 km resolution, overlaid with the DEM used for downscaling at 50 m resolution. Panels b) and d) show the corresponding fields produced by MicroPyzzotMet.

Toolbox methods and structure

MicroPyzzotMet is implemented entirely in Python and builds on widely used scientific and geospatial libraries, including NumPy, pandas, xarray, and its Zarr engine for cloud-native data access. It uses rasterio and rioxarray for raster handling, and pyproj for coordinate transformations. Terrain derivatives such as slope, aspect, and curvature are generated with

67 rasterio, gdaldem, and custom convolution kernels, while **pvlb** is employed to compute solar
68 geometry required for shortwave radiation corrections. Parallel processing is handled via **joblib**
69 to distribute downscaling tasks across CPU cores.

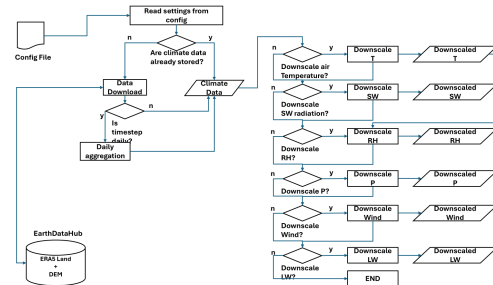


Figure 2: Workflow of the MicroPyzzotMet processing pipeline.

70 **Figure 2:** Schematic overview of the *MicroPyzzotMet* downscaling workflow. Coarse-resolution
71 reanalysis data (e.g., ERA5-Land) and a Digital Elevation Model (DEM) constitute the primary
72 inputs. Terrain derivatives (slope, aspect, curvature) are computed from the DEM and
73 combined with solar geometry to drive MicroMet-inspired corrections. Each meteorological
74 variable is processed independently through reprojection to the DEM grid, vertical (lapse-rate)
75 adjustment, and terrain-based corrections, producing high-resolution meteorological fields
76 written as NetCDF outputs suitable for cryospheric and hydrological modelling.

77 The workflow of *MicroPyzzotMet* is controlled by a single JSON configuration file and
78 orchestrated by the main execution function. The pipeline begins by creating a standard
79 folder structure (inputs/climate, inputs/dem, outputs) and by loading or downloading a
80 Digital Elevation Model (DEM). When no DEM is provided, the tool retrieves a Copernicus
81 GLO-30 subset from EarthDataHub as a Zarr dataset, reprojects and resamples it to the
82 user-defined grid, and writes it to GeoTIFF. Slope, aspect, and curvature metrics are then
83 computed and stored for use in the downscaling routines.

84 Meteorological forcing is obtained either from user-supplied NetCDF files or directly from
85 ERA5-Land via EarthDataHub. When downloaded through EarthDataHub, the Zarr dataset is
86 spatially subsetting to match the DEM extent and written to monthly NetCDF files containing
87 variables such as 2 m air temperature and dewpoint, surface pressure, 10 m wind components,
88 precipitation, and shortwave and longwave radiation. Cumulative fluxes are optionally converted
89 to hourly or daily rates.

90 Once the DEM and climate inputs are prepared, *MicroPyzzotMet* applies a set of variable-
91 specific downscaling functions. These functions implement MicroMet-style parameterizations:
92 - **Temperature** is adjusted using monthly lapse rates or dynamically calibrated rates.
93 - **Shortwave radiation** is corrected using topographic metrics and solar geometry.
94 - **Relative humidity** is derived from temperature and dewpoint using vapor-pressure relationships.
95 - **Precipitation** is scaled with elevation using empirical gradients.
96 - **Wind fields** are modified based on terrain metrics.
97 - **Longwave radiation** is adjusted using cloudiness estimates derived from humidity and
98 temperature.

99 Each routine reads a single monthly climate file, reprojects the coarse fields to the DEM grid,
100 applies vertical and topographic corrections, and writes a NetCDF output file.

101 The selection of variables to downscale is fully configurable, allowing modular development
102 and efficient processing of large datasets.

103 **Table 1:** Default downscaled output variables of *MicroPyzzotMet* (based on ERA5-Land
104 inputs).

| Name | Variable | Unit | Downscaling type |
|-------------------------------|----------|-------------------|--|
| 2 m Air temperature | t2m | K | Vertical lapse-rate adjustment; reprojection to DEM grid |
| Relative humidity | RH | % | Lapse-rate corrections; vapor-pressure formulation |
| Surface pressure | sp | Pa | Reprojection to DEM grid (optional elevation adjustment) |
| 10 m Wind speed and direction | u10, v10 | m s ⁻¹ | Reprojection and terrain-based adjustments |
| Precipitation | P | mm | Elevation-dependent scaling using empirical gradients |
| Incoming longwave radiation | LW | W m ⁻² | Atmospheric and cloudiness corrections |
| Incoming shortwave radiation | SW | W m ⁻² | Topographic and solar-geometry corrections |

Working examples

A complete working example of MicroPyzzotMet is available in the public repository:
<https://github.com/bare92/micropyzzotmet>.

The included demonstration applies the downscaling workflow to the Maipo basin in central Chile, a region characterized by steep elevation gradients and strong spatial variability in meteorological forcing.

The example is configured through the file `micro_config_DEMO_MAIP0.json` and executed with a simple shell script. In this workflow:

- A DEM covering the Maipo catchment is downloaded from EarthDataHub as a Zarr dataset, reprojected to EPSG:32719 (UTM 19S), and resampled to 50 m resolution.
- ERA5-Land meteorological inputs for **1 April to 31 May 2017** are fetched via EarthDataHub, enabling fast cloud-native access to reanalysis data.
- All major variables—air temperature, shortwave and longwave radiation, relative humidity, precipitation, and wind—are downscaled using MicroMet-based parameterizations.
- Outputs are written as monthly NetCDF files and can be converted into S3M-compatible forcing files.

This demonstration illustrates the typical usage of MicroPyzzotMet: a lightweight, configuration-driven workflow capable of producing high-resolution atmospheric forcing fields with minimal user intervention. The Maipo setup can be adapted to other regions by modifying the spatial extent, DEM specifications, and processing period.

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