

# <sup>1</sup> MicroPyzzotMet: A Lightweight Python Package for Climate Downscaling

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DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

## Software

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Editor: ↗

Submitted: 19 January 2026

Published: unpublished

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## <sup>7</sup> Summary

<sup>8</sup> Modern reanalysis products provide continuous global climate information extending back  
<sup>9</sup> decades and into future projections, yet their spatial resolution remains too coarse to represent  
<sup>10</sup> the meteorological variability imposed by mountain terrain ([Fan et al., 2019](#)). For researchers  
<sup>11</sup> working on snow, glaciers, permafrost, and alpine hydrology, this mismatch remains a  
<sup>12</sup> persistent limitation: surface energy-balance and mass-balance models depend on local  
<sup>13</sup> meteorological fields that capture how terrain modulates atmospheric conditions. Generating  
<sup>14</sup> such fields does not always require complex dynamical downscaling systems, which—despite  
<sup>15</sup> their accuracy—often demand substantial computational resources and sophisticated model  
<sup>16</sup> setups.

<sup>17</sup> MicroPyzzotMet addresses this gap with a lightweight downscaling framework focused on  
<sup>18</sup> practicality and broad usability. Rather than implementing complex physical parameterizations  
<sup>19</sup> or spatial clustering techniques, it applies a streamlined set of MicroMet-inspired corrections  
<sup>20</sup> to temperature, radiation, humidity, wind, and precipitation ([Liston & Elder, 2006](#)). Because  
<sup>21</sup> it requires only a minimal set of essential climate variables, the tool can operate with virtually  
<sup>22</sup> any reanalysis or climate dataset. A further strength is its integration with EarthDataHub,  
<sup>23</sup> enabling rapid access to ERA5-Land and digital terrain models through the Zarr format, which  
<sup>24</sup> significantly reduces I/O overhead and speeds up preprocessing.

<sup>25</sup> In contrast to more advanced packages such as TopoPyScale, which is designed for detailed  
<sup>26</sup> terrain-driven heterogeneity and fine-scale modelling ([Filhol et al., 2023](#)), MicroPyzzotMet  
<sup>27</sup> prioritizes computational efficiency and conceptual clarity. This makes it ideal for large-domain  
<sup>28</sup> experiments, operational workflows, or rapid prototyping, while still remaining compatible with  
<sup>29</sup> higher-resolution approaches when more elaborate topographic corrections are required.  
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## Statement of need

<sup>31</sup> MicroPyzzotMet is an open-source Python package for downscaling meteorological variables  
<sup>32</sup> from reanalysis climate datasets. It is inspired by the MicroMet methodology ([Liston & Elder,](#)  
<sup>33</sup> [2006](#)) but reimplemented in a modern Python framework, making the workflow more accessible,  
<sup>34</sup> flexible, and easier to integrate in contemporary data-processing pipelines.

<sup>35</sup> The increasing availability of atmospheric reanalyses—such as ERA5 and ERA5-Land at 25 km  
<sup>36</sup> and 9 km spatial resolution—has enabled a wide range of cryospheric and hydrological studies.  
<sup>37</sup> However, these coarse spatial grids remain inadequate for mountain regions, where elevation,  
<sup>38</sup> slope, and aspect strongly modulate near-surface climate. Tools such as TopoPyScale ([Filhol](#)  
<sup>39</sup> [et al., 2023](#)), based on the TopoSCALE and TopoSUB approaches ([Fiddes & Gruber, 2014;](#)  
<sup>40</sup> ?), address this limitation by applying sophisticated 3-D interpolation schemes, multi-level

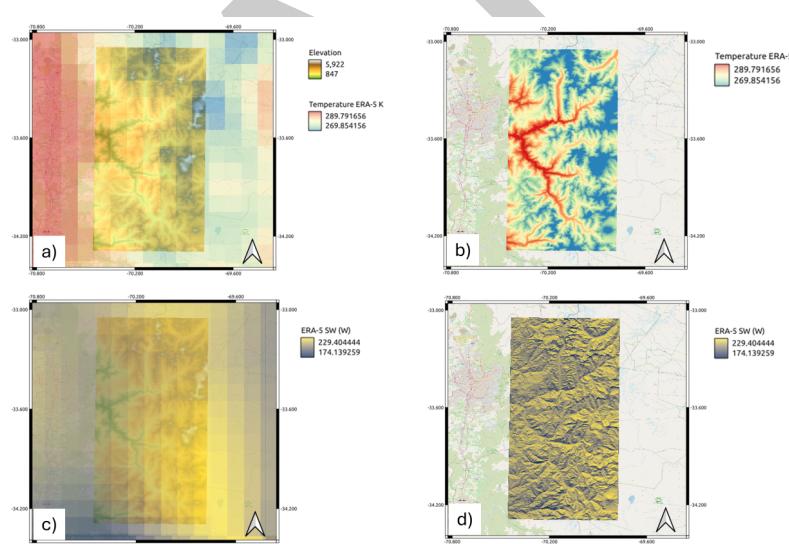
41 atmospheric corrections, and DEM segmentation into terrain clusters. These methods allow  
 42 detailed reconstruction of fine-scale meteorological patterns, especially over complex alpine  
 43 terrain.

44 Yet not all applications require this level of complexity. For coarser target resolutions (e.g.,  
 45 250–500 m), for large domains processed at high temporal frequency, or for long climatological  
 46 time series, the computational and data requirements of advanced downscaling frameworks  
 47 may become limiting. In these cases, the original MicroMet methodology offers an attractive  
 48 balance between physical robustness and computational simplicity.

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

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

59 [Figure 1](#) illustrates an example for the Maipo region in Chile, comparing native ERA5-Land fields  
 60 to MicroPyzzotMet downscaled products for daily air temperature and incoming shortwave  
 61 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.

## 62 Toolbox methods and structure

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

67 rasterio, gdaldem, and custom convolution kernels, while `pvlib` 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.

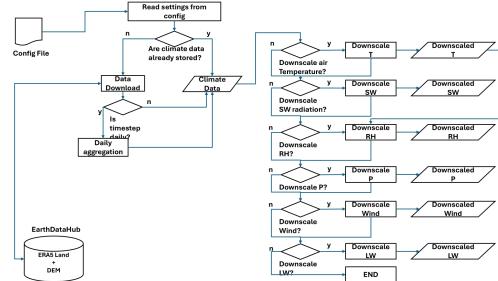


Figure 2: Workflow of the MicroPyzzotMet processing pipeline.

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 subsetted 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 <sup>-1</sup>	Reprojection and terrain-based adjustments
Precipitation	P	mm	Elevation-dependent scaling using empirical gradients
Incoming longwave radiation	LW	W m <sup>-2</sup>	Atmospheric and cloudiness corrections
Incoming shortwave radiation	SW	W m <sup>-2</sup>	Topographic and solar-geometry corrections

## 105 Working examples

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

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

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

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

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

## 128 Acknowledgements

129 This project has received funding from the European Union's Horizon Research and Innovation  
 130 Actions programme under Grant Agreement 101180133, and from the Swiss State Secretariat  
 131 for Education, Research and Innovation (SERI).

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