

MetObs-Toolkit - a Python toolkit for using non-traditional meteorological observations

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Summary

In-situ weather station observations are highly important for meteorological and climatological research. The evolution towards more affordable sensor technology and data communication has resulted in the emergence of novel meteorological networks alongside the traditional high-quality measurement networks of meteorological institutions. Examples include urban measurement networks intended to study the impact of cities (Caluwaerts et al., 2020) and networks consisting of devices of weather enthusiasts e.g. (C. L. Muller et al., 2015). However, exploiting the data of such non-traditional networks comes with significant challenges (Meier et al., 2017). Firstly, sensors and data communication protocols are usually low-cost and this in general results in an increase of measurement errors, biases and data gaps. Secondly, data storage formats and temporal measurement frequencies are not consistent or compatible across different networks. Finally, metadata, such as station location, elevation, and instrument specifications, are not easily accessible or documented. The MetObs-toolkit is a Python package developed to address these issues and facilitate the use of non-traditional observations. The package provides automated quality control (QC) techniques to identify and flag erroneous observations and includes methods to fill data gaps. Additionally, the package also offers a set of tools for analyzing the data e.g. linkage with popular land use datasets (Demuzere et al., 2022; Zanaga et al., 2022) is included such that microclimate effects can be investigated by the MetObs toolkit.

Statement of need

The primary objective of the MetObs toolkit is to enable scientists to process meteorological observations into datasets ready for analysis. The data cleaning process involves three steps:

1. resampling the time resolution if necessary,
2. identifying erroneous or missing records, and
3. filling the missing records.

Sophisticated software such as TITAN/CrowdQC+ (Båserud et al., 2020; Fenner et al., 2021) exists for identifying erroneous observations (QC), which is one aspect of cleaning a dataset. These packages are designed specifically for this purpose. Moreover, researchers often face the challenge of coding scripts that can generate analyses, particularly when using geographical datasets such as landcover datasets. Traditionally, this requires the installation of numerous packages, storage of geographical datasets, and GIS manipulations (often manually done with specific GIS software). The toolkit implements one user-friendly framework for creating various plots, generating analysis statistics, and incorporating GIS data through the use of the Google

Earth engine. By using the toolkit, scientists can set up a pipeline to process raw data into analysis in an easy-to-use (and install) manner. Additionally, the developed pipeline can be directly applied to other datasets without any formatting issues.

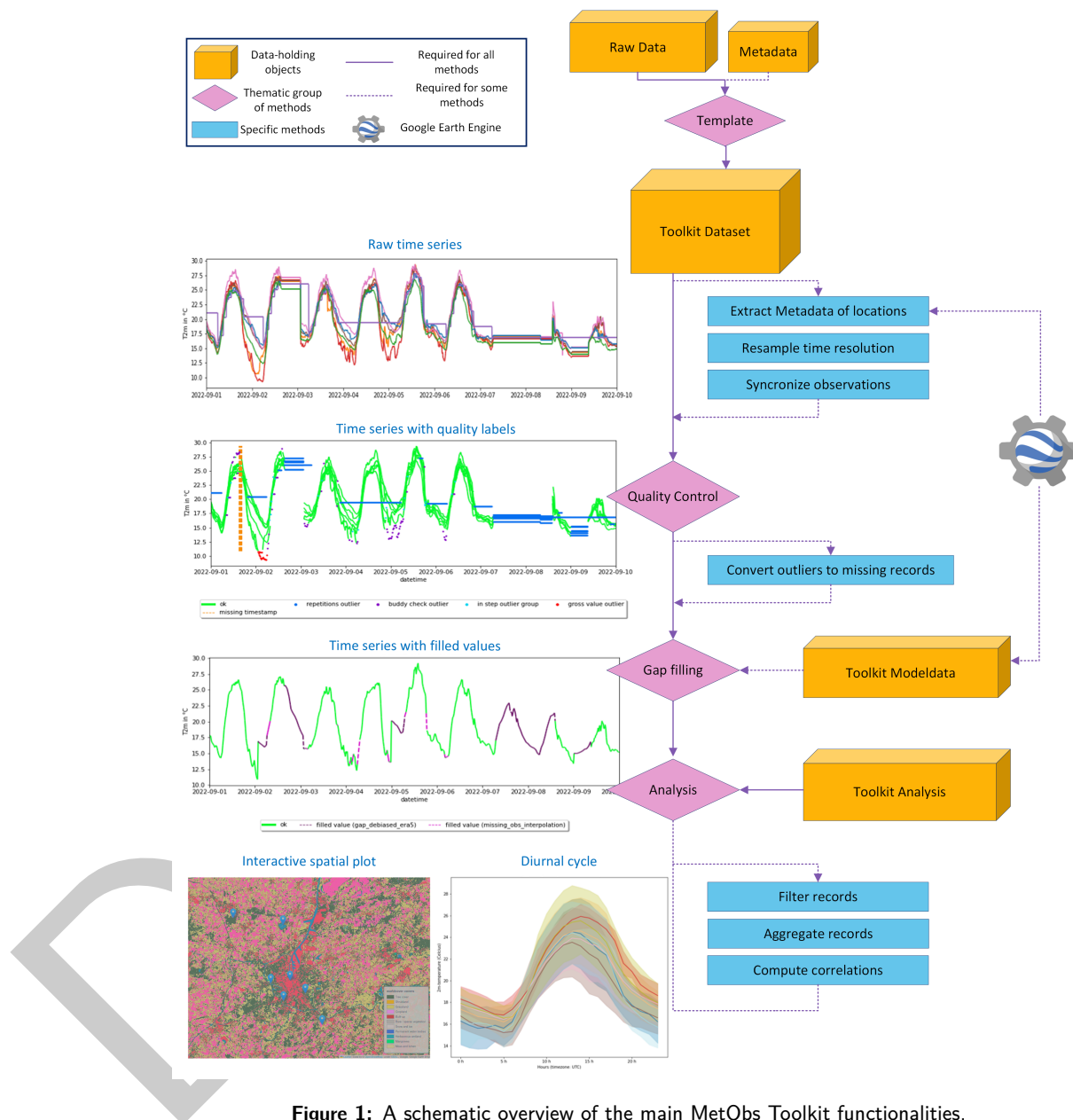


Figure 1: A schematic overview of the main MetObs Toolkit functionalities.

Technical implementation

The MetObs-Toolkit provides a comprehensive framework for scientists to process raw meteorological data for analysis. The process consists of the following steps, visualized in the figure below:

Mapping the raw data to the toolkit standards by creating a template. Once the raw data is imported into the Toolkit Dataset, missing observations are identified and methods to resample and synchronize observations can be used. Quality control is performed in the form of a series of checks. Advanced quality control methods are available through the implementation of TITAN

into the toolkit. The user can choose to keep the outliers or convert them to missing records (which can be filled). Gap filling is applied by using interpolation methods and/or importing ERA5 reanalysis time series to fill the gaps. The latter is stored as a Toolkit Modeldata, which has a set of methods to directly import the required time series through the use of the Google Earth engine API. The user obtains a cleaned-up dataset ready for analysis. A set of typical analysis techniques such as filters, aggregation schemes, and landcover correlation estimates are implemented in the Toolkit-Analysis class.

Figure 1 gives an overview of the main framework of the MetObs-toolkit, but it is an evolving project that responds to the community's needs and input. As an example, the development of a graphical user interface (GUI) for the toolkit is planned. A GUI would increase the ease of use by enabling to create templates, adjust QC settings and plot data interactively.

It is to be expected that some novel functionalities will be added in the coming years and the documentation will be updated accordingly.

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References

- Båserud, L., Lussana, C., Nipen, T. N., Seierstad, I. A., Oram, L., & Aspelien, T. (2020). TITAN automatic spatial quality control of meteorological in-situ observations. *Advances in Science and Research*, 17, 153–163. <https://doi.org/10.5194/asr-17-153-2020>
- Caluwaerts, S., Hamdi, R., Top, S., Lauwaet, D., Berckmans, J., Degrauwe, D., Dejonghe, H., De Ridder, K., De Troch, R., Duchêne, F., Maiheu, B., Van Ginderachter, M., Verdonck, M.-L., Vergauwen, T., Wauters, G., & Termonia, P. (2020). The urban climate of ghent, belgium : A case study combining a high-accuracy monitoring network with numerical simulations. *URBAN CLIMATE*, 31, 19. <http://dx.doi.org/10.1016/j.uclim.2019.100565>
- Caluwaerts, S., Top, S., Vergauwen, T., Wauters, G., Ridder, K. D., Hamdi, R., Mesuere, B., Schaeybroeck, B. V., Wouters, H., & Termonia, P. (2021). Engaging schools to explore meteorological observational gaps. *Bulletin of the American Meteorological Society*, 102(6), E1126–E1132. <https://doi.org/https://doi.org/10.1175/BAMS-D-20-0051.1>
- Demuzere, M., Kittner, J., Martilli, A., Mills, G., Moede, C., Stewart, I. D., Vliet, J. van, & Bechtel, B. (2022). *Global map of local climate zones* (Version 1.0.0) [Data set]. Zenodo. <https://doi.org/10.5281/zenodo.6364594>
- Fenner, D., Bechtel, B., Demuzere, M., Kittner, J., & Meier, F. (2021). CrowdQC+—a quality-control for crowdsourced air-temperature observations enabling world-wide urban climate applications. *Frontiers in Environmental Science*, 9. <https://doi.org/10.3389/fenvs.2021.720747>

- 97 Harris, C. R., Millman, K. J., Walt, S. J. van der, Gommers, R., Virtanen, P., Cournapeau, D.,
 98 Wieser, E., Taylor, J., Berg, S., Smith, N. J., Kern, R., Picus, M., Hoyer, S., Kerkwijk,
 99 M. H. van, Brett, M., Haldane, A., Río, J. F. del, Wiebe, M., Peterson, P., ... Oliphant,
 100 T. E. (2020). Array programming with NumPy. *Nature*, 585(7825), 357–362. <https://doi.org/10.1038/s41586-020-2649-2>
 101
- 102 Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, A., Muñoz-Sabater, J., Nicolas,
 103 J., Peubey, C., Radu, R., Schepers, D., Simmons, A., Soci, C., Abdalla, S., Abellan,
 104 X., Balsamo, G., Bechtold, P., Biavati, G., Bidlot, J., Bonavita, M., ... Thépaut, J.-N.
 105 (2020). The ERA5 global reanalysis. *Quarterly Journal of the Royal Meteorological Society*,
 106 146(730), 1999–2049. <https://doi.org/https://doi.org/10.1002/qj.3803>
- 107 Jordahl, K., Bossche, J. V. den, Fleischmann, M., Wasserman, J., McBride, J., Gerard,
 108 J., Tratner, J., Perry, M., Badaracco, A. G., Farmer, C., Hjelle, G. A., Snow, A. D.,
 109 Cochran, M., Gillies, S., Culbertson, L., Bartos, M., Eubank, N., maxalbert, Bilogur,
 110 A., ... Leblanc, F. (2020). *Geopandas/geopandas: v0.8.1* (Version v0.8.1). Zenodo.
 111 <https://doi.org/10.5281/zenodo.3946761>
- 112 McKinney, Wes. (2010). Data Structures for Statistical Computing in Python. In Stéfan van
 113 der Walt & Jarrod Millman (Eds.), *Proceedings of the 9th Python in Science Conference*
 114 (pp. 56–61). <https://doi.org/10.25080/Majora-92bf1922-00a>
- 115 Meier, F., Fenner, D., Grassmann, T., Otto, M., & Scherer, D. (2017). Crowdsourcing air
 116 temperature from citizen weather stations for urban climate research. *Urban Climate*, 19,
 117 170–191. <https://doi.org/https://doi.org/10.1016/j.uclim.2017.01.006>
- 118 Muller, Catherine L., Chapman, L., Grimmond, C. S. B., Young, D. T., & Cai, X. (2013).
 119 Sensors and the city: A review of urban meteorological networks. *International Journal of*
 120 *Climatology*, 33(7), 1585–1600. <https://doi.org/https://doi.org/10.1002/joc.3678>
- 121 Muller, C. L., Chapman, L., Johnston, S., Kidd, C., Illingworth, S., Foody, G., Overeem, A.,
 122 & Leigh, R. R. (2015). Crowdsourcing for climate and atmospheric sciences: Current
 123 status and future potential. *International Journal of Climatology*, 35(11), 3185–3203.
 124 <https://doi.org/https://doi.org/10.1002/joc.4210>
- 125 Stewart, I. D., & Oke, T. R. (2012). Local climate zones for urban temperature studies.
 126 *Bulletin of the American Meteorological Society*, 93(12), 1879–1900. <https://doi.org/https://doi.org/10.1175/BAMS-D-11-00019.1>
 127
- 128 team, T. pandas development. (2023). *Pandas-dev/pandas: pandas* (Version v2.0.0). Zenodo.
 129 <https://doi.org/10.5281/zenodo.7794821>
- 130 Zanaga, D., Van De Kerchove, R., Daems, D., De Keersmaecker, W., Brockmann, C., Kirches,
 131 G., Wevers, J., Cartus, O., Santoro, M., Fritz, S., Lesiv, M., Herold, M., Tsendbazar, N.-E.,
 132 Xu, P., Ramoino, F., & Arino, O. (2022). *ESA WorldCover 10 m 2021 v200* (Version v200)
 133 [Data set]. Zenodo. <https://doi.org/10.5281/zenodo.7254221>