




# 1 SenSARP: A pipeline to pre-process Sentinel-1 SLC data 2 by using ESA SNAP Sentinel-1 Toolbox

3 **Thomas Weiß<sup>1</sup> and Tonio Fincke<sup>2</sup>**

4 **1** Department of Geography, Ludwig-Maximilians-Universität München, 80333 Munich, Germany **2**  
5 Brockmann Consult GmbH, 21029 Hamburg, Germany;

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## Software

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## 6 Summary

7 The Sentinel-1 mission consists of two polar-orbiting satellites acquiring Synthetic Aper-  
8 ture Radar data (SAR) at C-band (frequency of 5.405 GHz) with a revisit time of 6 days.  
9 The SAR data is distributed free of charge via the Copernicus Open Access Hub (<https://scihub.copernicus.eu/>) by European Space Agency (ESA) and the European Commission.  
10 Large archives are also provided by Data and Information Access Services (DIAS) which serve  
11 the purpose to facilitate the access and use of Sentinel Data. Due to the specific imaging  
12 geometry of the radar system, the acquired radar data contains different radiometric and geo-  
13 metric distortions. The radiometric quality is affected by spreading loss effect, the non-uniform  
14 antenna pattern, possible gain changes, saturation, and speckle noise. Geometric distortions  
15 such as foreshortening, layover or shadowing effects are based on the side looking radar ac-  
16 quisition system. To account for these radiometric and geometric distortions, the Sentinel-1  
17 Level 1 data has to be corrected radiometrically and geometrically before the data can be  
18 used for further analysis or within third party applications. Therefore, either an automatic or  
19 manual pre-processing of Sentinel-1 images is needed.  
20

## 21 Statement of need

22 Sentinel-1 satellites will provide continuous free available microwave remote sensing data of  
23 the entire globe at least until the end of 2030. Furthermore, ESA is not only providing  
24 Sentinel satellite images (e.g. Sentinel-1, Sentinel-2, Sentinel-3) but they also developed free  
25 open source toolboxes (Sentinel-1, 2, 3 toolboxes) for scientific exploitation. The toolboxes  
26 can be accessed and used via the Sentinel Application Platform (SNAP). SNAP offers a  
27 graphical interface where expert users can develop different processing schemes and apply  
28 them on the satellite images. Although, Sentinel-1 satellite data and a processing software  
29 are freely available, the usage of the data is mainly limited to expert users in the field of  
30 microwave remote sensing as different pre-processing steps need to be applied before using  
31 Sentinel-1 images.

32 SenSARP was developed to provide a push-button option to easily apply a rigid pre-processing  
33 pipeline with sensible defaults to a Sentinel-1 Level 1 SLC time series data as well as single  
34 Sentinel-1 Level 1 SLC images. Thus, non-expert users in the field of pre-processing microwave  
35 data are able to use radiometric and geometric corrected sigma nought backscatter data for  
36 their specific applications. Beside a rigid pre-processing pipeline, SenSARP provides filter  
37 options to retrieve only images of a specific year or images that contain a specific area of  
38 interest from a stack of downloaded Sentinel-1 data. Furthermore, the default processing  
39 scheme of SenSARP can handle if an area of interest is contained in two tiles of the same  
40 swath (due to storage reasons data of one Sentinel-1 satellite swath is provided by ESA within

different tiles). Additionally, SenSARP checks if within a stack of Sentinel-1 images, one specific image was multiple processed by ESA and uses the newest.

For expert users, SenSARP provides the possibility to automate their pre-processing on a large scale by either modifying the default pre-processing scheme (modification of xml graph `pre_processing_step1.xml`) or create their own pre-processing scheme (create a new xml graph) with the graph builder of the SNAP software. They can benefit from the filter options, the default pre-processing step 2 (co-registration of images) and the SenSARP functions to stack all processed and co-registered images within a netCDF file with additional image information e.g. satellite name, relative orbit and orbit direction.

## Method

This Python package generates a file list of to be processed Sentinel-1 images (already downloaded and stored in a specific folder) based on different user defined criteria (specific year, area of interest). Additionally, specific cases of repeatedly processed data are handled, as sometimes Sentinel-1 data were initially processed multiple times and stored under similar names on the Copernicus Open Access Hub. Also, cases where Sentinel-1 data within the user-defined area of interest might be stored in consecutive tiles are considered.

Based on the generated file list the default processing pipeline of the Python package applies a pre-processing chain to Sentinel-1 Single Look Complex (SLC) time series or single images to generate radiometrically and geometrically corrected sigma nought backscatter values. Furthermore, if a time series is processed the images are co-registered and additional output files of multi-temporal speckle filtered data are generated. In addition, a single speckle filter instead of a multi-temporal one is applied as well and the output will be stored as a separate layer. To pre-process the images, the Python package uses the GPT (Graph Processing Tool) of SNAP to execute different operators provided by the Sentinel-1 Toolbox. The Sentinel Toolbox is available for download at <http://step.esa.int/>, its source code is available in the senbox-org organization on GitHub. Each of these operators performs a pre-processing step. The operators can be chained together to form a graph, which is used by the Python package to run on the Sentinel-1 data using the Graph Processing Framework (GPF). The graphs are stored in xml-files. Users may change the graphs by modifying the files directly or via the Sentinel Toolbox. User Guides to show how the GPF can be used are provided here: <https://senbox.atlassian.net/wiki/spaces/SNAP/pages/70503053/Processing>.

After the pre-processing the resulting radiometrically and geometrically corrected images are stored for further usage within a NetCDF4 stack file. The processing workflow was developed and optimized to use a Sentinel-1 time series of pre-processed sigma nought backscatter values to retrieve biophysical land surface parameters by the use of radiative transfer models. The sigma nought backscatter values provided by the default workflow of SenSARP might be used in other applications like flood risk analysis, monitoring land cover changes or monitoring global food security but it has to be mentioned that different applications have different demands and therefore, slight adjustments of the default workflow might be required. In the future, many more new products and operational third party services based on consistent Sentinel-1 time series might be developed.

## Applications

This Python package was developed within the Horizon 2020 project called MULTIscale SENTINEL land surface information retrieval Platform (MULTIPLY) (<http://www.multiply-h2020.eu/>, <https://cordis.europa.eu/project/id/687320>, <https://multiply.obs-website.eu-de.etc>.

t-systems.com). Furthermore, data processed by this package is used within Sentinel-Synergy-Study S3 project (<https://www.researchgate.net/project/Sentinel-Synergy-Study-S3>). In addition, the Python code was used to process Sentinel-1 time series images for the detection and analysis of temporary flooded vegetation (Tsyganskaya et al., 2018, 2019) and for the evaluation of different radiative transfer models for microwave backscatter estimation of wheat fields (Weiß et al., 2020).

## Other available Python software packages using ESA's SNAP software to pre-process SAR data

The ESA's SNAP toolbox has been written in Java. For Python users the developers provide a Python interface called Snappy. However, the Snappy interface is lacking in terms of installation, processing performance and usability. Hence, the remote sensing community developed different wrappers (e.g. SenSARP, snapista or pyroSAR) to use SNAP processing functionalities by utilizing the SNAP Graph Processing Tool (GPT).

### snapista

Snapista (<https://snap-contrib.github.io/snapista/index.html>) targets mainly experts remote sensing users with Python programming skills. It provides access to the processing operators of all toolboxes (e.g. Sentinel-1, Sentinel-2 or Sentinel-3) within SNAP. Expert users can generate processing graphs and execute their generated graphs in a pure Pythonic way. Guidelines about which processing steps are needed for different applications, or about which processing steps can or have to be combined, are not provided yet. Establishing guidelines about how to process different satellite data for different applications is not an easy task to do and would exceed the goal of snapista as a Python wrapper for the SNAP software. Summarizing, snapista provides access to all SNAP toolboxes (not just to Sentinel-1 Toolbox) via Python. But as it provides no default processing chains, snapista will be primarily usable by expert remote sensing users. The advantage of snapista is the accessibility of processing operators for SAR and optical data.

### pyroSAR

PyroSAR (<https://pyrosar.readthedocs.io/en/latest/index.html>) is a Python library which provides a Python wrapper to SAR pre-processing software SNAP and GAMMA (Wegnüller et al., 2016; Werner et al., 2000). The library provides utilities to read and store metadata information of downloaded satellite data within a database. Furthermore, pyroSAR provides access to processing operators of SNAP and GAMMA. A default workflow with different user options is provided to process single or time-series Sentinel-1 images. After executing the default processing workflow radiometric and geometric corrected gamma nought backscatter data are provided in Geotiff format (Truckenbrodt et al., 2019). The processed images can also be stored within an Open Data Cube. For expert users which might want to use a different processing workflow pyroSAR provides an option to create SNAP xml-workflows and execute them with the GPT. Summarizing, pyroSAR provides a similar push-button option to process Sentinel-1 data with a slightly different default workflow (pyroSAR: no temporal speckle filter, gamma nought backscatter output in Geotiff format) than SenSARP (SenSARP: temporal speckle filter, sigma nought backscatter output in netCDF format). PyroSAR, as a more complex library than SenSARP, provides on the one hand more changeable parameters within the processing workflow but on the other hand the usability for non-expert users might be narrowed compared to SenSARP. An advantage of SenSARP, especially for non-expert

130 users, might be the provision of background information (theory/purpose) of the different  
131 pre-processing steps within the documentation.

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