



EvoLand
LAND MONITORING EVOLUTION

T2.21 Source Code

Work Package 2 - Method development

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Start date	01.01.2023	Duration	36 months
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Project Coordinator	Vlaamse Instelling voor Technologisch Onderzoek N.V. (VITO)		
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Version	Issue date	Stage	Changes	Contributor
0.1	10.02.2025	Draft	Review of D2.20	ALL
1.0	31.03.2025	Final	Review and finalization	VITO

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1 Weakly supervised learning

1.1 DLR code

Example of code

Module Imports

```
from lightning.pytorch.trainer import Trainer
from evoland.data import DataDownloader, patchify, TreeCoverDataModule
from evoland.models import SemanticSegmentationModule
import wget, os
```

Model Inference Demo

```
import lightning.pytorch
```

```
model = SemanticSegmentationModule(
    num_classes=2,
    model="fcn",
    backbone="custom_resnet18",
    weights=None,
    in_channels=12,
    ignore_index=255,
    loss='ce',
)
```

Setting up SimpleSegmentationModel

```
type(model), isinstance(model, lightning.pytorch.LightningModule)
```

```
(evoland.models.forest_segmentation.SemanticSegmentationModule, True)
```

```
datamodule = TreeCoverDataModule(
    root = ".",
    img_dir_name = IMAGE_DIR,
    batch_size = 1,
    num_workers = 0,
    patch_size = PATCH_SIZE,
    use_masks = False,
)
```

```
type(datamodule), isinstance(datamodule, lightning.pytorch.LightningDataModule)
```

```
(evoland.data.treecover.TreeCoverDataModule, True)
```

```
MODEL_PATH = 'custom_resnet18_l2a.ckpt'
trainer = Trainer(
    accelerator='cpu',
    devices=1,
)
output = trainer.predict(
    model=model,
    datamodule=datamodule,
    ckpt_path=MODEL_PATH,
)
```

Description	
Name	EvoLand - software
Description	Python package with modules for AI methods evoland.models and evoland.data for loading corresponding data inputs including helper classes and functions in evoland.utils. An overview for installation (mamba/conda), documentation (Pdoc), and usage (Jupyter notebook) provides https://github.com/Evoland-Land-Monitoring-Evolution/evoland-software/blob/main/README.md
Version	0.1 (interim version)
Access	EvoLand GitHub: https://github.com/Evoland-Land-Monitoring-Evolution/evoland-software
License	Apache-2.0

Citation	Braham, N., Jancauscas, V., Wang, Y., Albrecht, C. M., “evoland – a Python package for self-supervised AI methodologies for remote sensing employed and developed for the EU Horizon project EvoLand” (2024)
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1.2 CESBIO code

1.2.1 MMDC pre-training

Description	
Name	MMDC-SingleDate
Description	Models and training code for the Multi-Modal Data Cube generic embeddings using Sentinel-1 and Sentinel-2 in single-date configuration.
Version	0.1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/MMDC-SingleDate
License	Apache-2.0 license
DOI	https://doi.org/10.5281/zenodo.10876755

1.2.2 MALICE and MALICE Aux pre-training

Description	
Name	malice
Description	Models and training code for Multimodal ALLse Cross-Encoder (MALICE) and MALICE with auxiliary data embeddings using Sentinel-1 and Sentinel-2 in multi-date configuration.
Version	0.1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/malice
License	Apache-2.0 license

1.2.3 ProsailVAE embeddings extraction with openEO

Description	
Name	openeo_pvae
Description	Sentinel-2 single-date embeddings extraction with prosailVAE model using openEO client
Version	0.1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/openeo_pvae
License	Apache-2.0 license

1.2.4 MALICE embeddings extraction with openEO

Description	
Name	openeo_malice
Description	Sentinel-2 and Sentinel-1 ASC multi-temporal embeddings extraction with MALICE model using openEO client
Version	0.1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/openeo_malice
License	Apache-2.0 license

1.2.5 ProsailVAE pre-training

Description	
Name	prosailvae
Description	Models and training code for ProsailVAE embeddings using Sentinel-2 in single-date configuration.
Version	0.1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/prosailvae
License	Apache-2.0 license

2 Improved resolution

2.1 UT3 code

Example of code

```

Code Blame 438 lines (369 loc) · 12.8 KB

1  """
2  Inference script
3  """
4
5  import argparse
6  import logging
7  import os
8  import sys
9  from dataclasses import dataclass
10
11  import numpy as np
12  import onnxruntime as ort # type: ignore[import-untyped]
13  import rasterio as rio # type: ignore[import-untyped]
14  from affine import Affine # type: ignore[import-untyped]
15  from sensorsio.sentinel2 import Sentinel2
16  from sensorsio.sentinel2_llc import Sentinel2LLC
17  from tqdm import tqdm
18
19
20  @dataclass(frozen=True)
21  class Chunk:
22      source_area: rio.coords.BoundingBox
23      target_area: rio.coords.BoundingBox
24
25
26  def generate_chunks(
27      roi: rio.coords.BoundingBox,
28      tile_size_in_meters: float,
29      margin_in_meters: float,
30  ) -> list[Chunk]:
31      """

```

	Description
Name	sentinel2_superresolution
Description	Generate 5m super-resolved images from Sentinel-2 L2A (Theia) products (bands B02, B03, B04, B05, B06, B06, B08, B8A, B11 and B12) using Single Image Super-Resolution model trained in the frame of the EVOLAND Horizon Europe.
Version	1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/sentinel2_superresolution
License	Apache-2.0
DOI	https://doi.org/10.3390/data7070096
Reference	Michel, J.; Vinasco-Salinas, J.; Inglada, J.; Hagolle, O. SEN2VEN μ S, a Dataset for the Training of Sentinel-2 Super-Resolution Algorithms. Data 2022, 7, 96. https://doi.org/10.3390/data7070096

3 Continuous Monitoring

3.1 JR code

Example of code

```
run_nrt.py
1 import argparse
2 from datetime import datetime, date, timedelta
3 import os
4 import shutil
5 import sys
6
7 from nrt.monitor.ewma import EWMA
8 from nrt.monitor.cusum import CuSum
9 from nrt.monitor.mosum import MoSum
10
11 import numpy as np
12
13
14 def recode_timestamp(ga, output_file, writer):
15     """
16     Recode the detection timestamp used by the nrt module.
17
18     By default the nrt module uses a timestamp format based on the number of
19     days since 1970-01-01, which is not human-readable and difficult to interpret.
20     This function recodes the values to the format
21     1000*two-digit-year + day-of-year. Two examples: 2004-01-05 -> 4005,
22     2020-02-01 -> 20032
23
24     Args:
25     ga (rasterioutils.GeoArray): Array representation of the original nrt output. The date
26     info is expected to be in a layer called 'timestamp'.
27     output_file (str): Path of the recoded output to be created.
28     writer (rasterioutils.ArrayWriter): Writer instance
29
30     """
31     output_arr = np.zeros(ga.array.shape[2:], np.uint16)
32     ref_date = date(1970, 1, 1)
33     for delta_days in np.unique(ga.timestamp):
34         if delta_days > 0:
35             alarm_date = ref_date + timedelta(delta_days.item())
36             output_arr[ga.array[0, 0, :, :] == delta_days] = int(alarm_date.strftime('%y%j'))
```

	Description
Name	ts-modeling-benchmarking
Description	This repository represents the implementation of a benchmarking suite able to run several open-source 3 rd party algorithms for time series modelling and unsupervised change detection.
Version	1.0
Access	GitHub https://github.com/Evoland-Land-Monitoring-Evolution/ts-modeling-benchmarking
License	Apache-2.0

3.2 VITO Code

Example of code

```

65     else:
66         LSC_PATH = _DATASET_PATH_HPC
67         HIST_PATH = _HIST_PATH_HPC
68
69
70     def compute_pixel_coordinates(bounds, shape):
71         """
72         Compute the y and x coordinates for every pixel in an image.
73
74         Args:
75         bounds (tuple): A tuple containing (xmin, ymin, xmax, ymax).
76         shape (tuple): A tuple containing the image shape (rows, columns).
77
78         Returns:
79         tuple: Two arrays containing y and x coordinates for every pixel.
80         """
81         xmin, ymin, xmax, ymax = bounds
82         rows, cols = shape
83
84         x_res = (xmax - xmin) / cols
85         y_res = (ymax - ymin) / rows
86
87         if x_res != y_res:
88             raise ValueError("Different resolution for y and x axis are not "
89                               "supported. Bounds and shape are not consistent "
90                               "with the same resolution on both axis.")
91
92         res_half = x_res / 2
93
94         xx = np.linspace(xmin + res_half,
95                           xmax - res_half,
96                           cols)
97
98         yy = np.linspace(ymax - res_half,
99                           ymin + res_half,
100                          rows)
101
102         return yy, xx

```

	Description
Name	Evoland-lsc
Description	The first version of the python code for training LSC models (both the U-Net and CatBoost model) and inference is provided
Version	1.0
Access	https://github.com/Evoland-Land-Monitoring-Evolution/evoland-lsc
License	Creative Commons Legal Code CC0 1.0 Universal

4 Biomass mapping

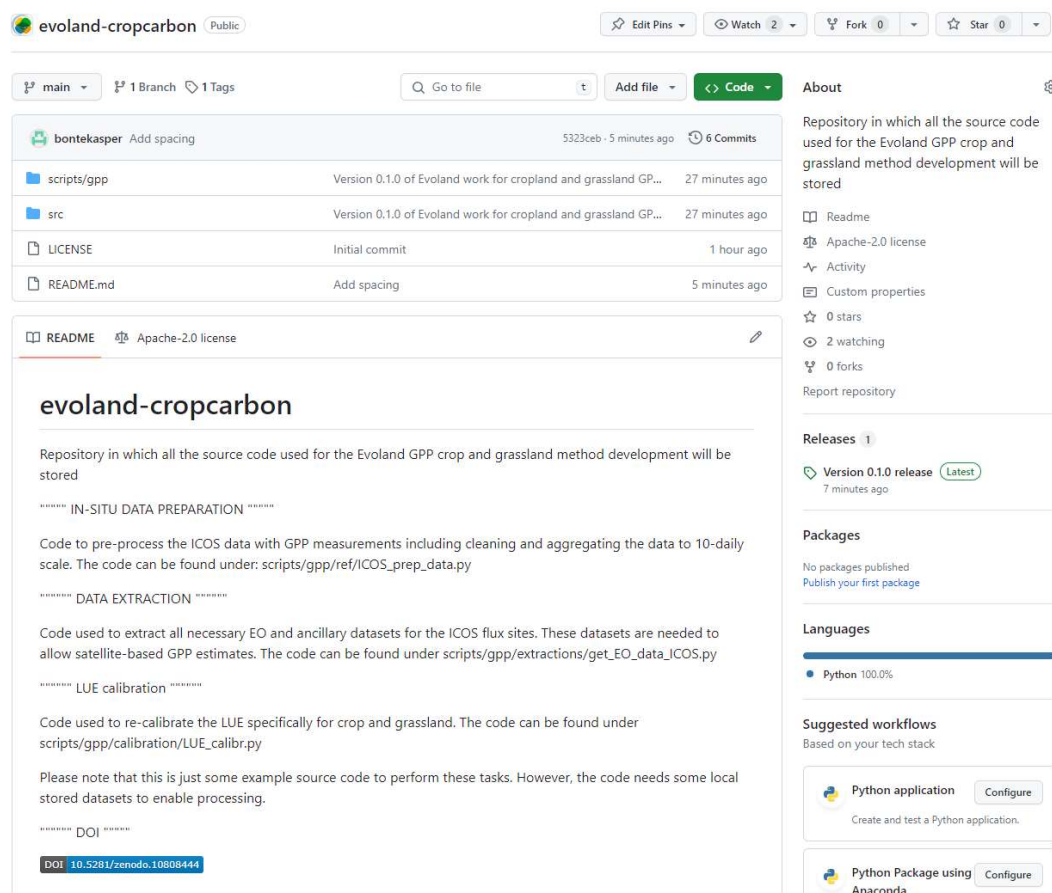
4.1 JR code

Description	
Name	RF1_Sweden_FCH_Predict.py
Description	The script uses a trained Random Forest Model for Forest Canopy Height (FCH) estimation and a multiband geotiff-File which includes various features. The script will output a GeoTIFF file with the predicted FCH. The script requires the following Python packages: - rasterio - numpy - pandas - joblib
Version	1.0
Access	https://github.com/Evoland-Land-Monitoring-Evolution/ForestBiomass
License	Creative commons attribution 4.0 international license

Description	
Name	RF1_Sweden_AGB_Predict.py
Description	The script uses a trained Random Forest Model for Above Ground Woody Biomass (AGB) estimation and a multiband geotiff-File which includes various features. The script will output a GeoTIFF file with the predicted AGB. The script requires the following Python packages: - rasterio - numpy - pandas - joblib
Version	1.0
Access	https://github.com/Evoland-Land-Monitoring-Evolution/ForestBiomass
License	Creative commons attribution 4.0 international license

5.1 VITO code

Example of code



evoland-crocarbon Public

Repository in which all the source code used for the Evoland GPP crop and grassland method development will be stored

Repository in which all the source code used for the Evoland GPP crop and grassland method development will be stored

***** IN-SITU DATA PREPARATION *****

Code to pre-process the ICOS data with GPP measurements including cleaning and aggregating the data to 10-daily scale. The code can be found under: scripts/gpp/ref/ICOS_prep_data.py

***** DATA EXTRACTION *****

Code used to extract all necessary EO and ancillary datasets for the ICOS flux sites. These datasets are needed to allow satellite-based GPP estimates. The code can be found under scripts/gpp/extractions/get_EO_data_ICOS.py

***** LUE calibration *****

Code used to re-calibrate the LUE specifically for crop and grassland. The code can be found under scripts/gpp/calibration/LUE_calibr.py

Please note that this is just some example source code to perform these tasks. However, the code needs some local stored datasets to enable processing.

***** DOI *****

DOI: [10.5281/zenodo.10808444](https://zenodo.org/doi/10.5281/zenodo.10808444)

Releases

Version 0.1.0 release (Latest)

Packages

No packages published

Languages

Python 100.0%

Suggested workflows

Based on your tech stack

Python application

Python Package using Anaconda

Description	
Name	evoland-crocarbon
Description	First version of the GPP crop and grassland monitoring workflow for EvoLand. In this repository all code used to prepare the in-situ data for calibration and to retrieve the needed EO and ancillary datasets is provided. In this release, the recalibration of the LUE for crop and grassland is included.
Version	0.1.0
Access	https://github.com/Evoland-Land-Monitoring-Evolution/evoland-crocarbon
License	Apache-2.0 license
DOI	https://zenodo.org/doi/10.5281/zenodo.10808443



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