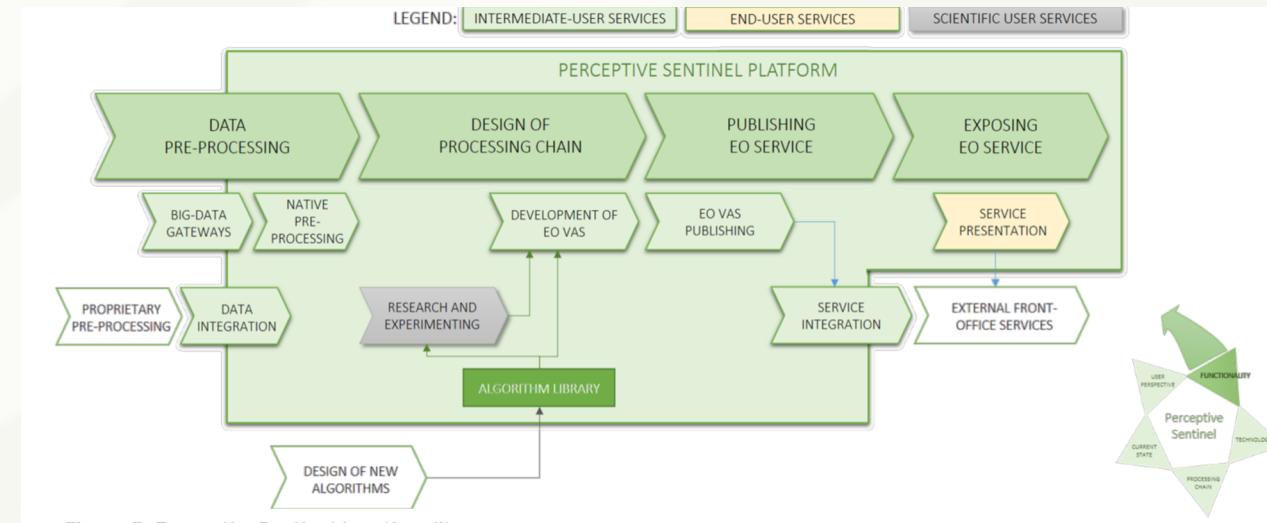




# Perceptive Sentinel eo-learn Python package

Dr. Anže Zupanc for the EO Research Group and Perceptive Sentinel

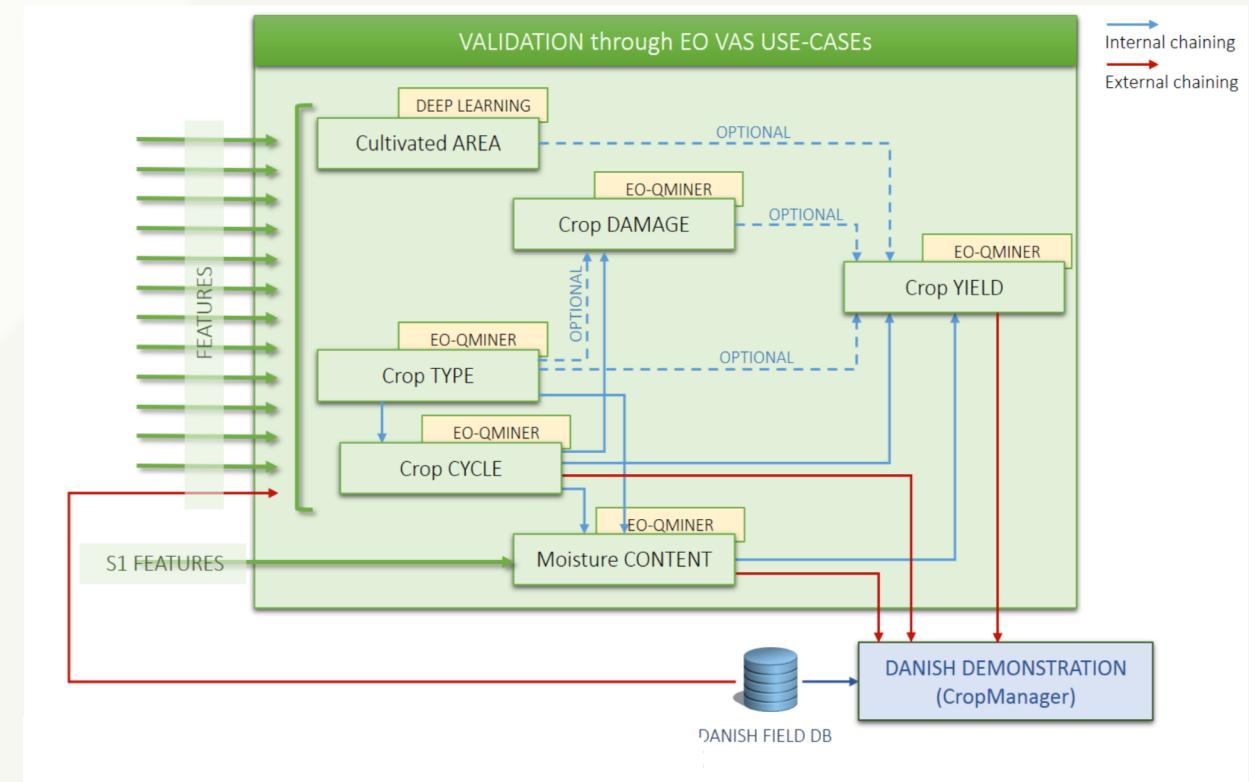
# Perceptive Sentinel in a nutshell



- Intermediate Earth observation (EO) service providing
  - *modelling* and *publishing* capabilities for
  - *design, exposure* and *exploitation* of EO-processing chains for
  - *forecasting, monitoring* and *historical analysis* based on
  - *multi-temporal* and *multi-spectral* EO and non-EO data *modelling*

# Perceptive Sentinel in a nutshell

- Partners
  - [Sinergise](#), Slovenia
  - [GeoVille](#), Austria
  - [Magellum](#), France
  - [Jožef Stefan Institute](#), Slovenia
  - [Agricultural Institute Slovenia](#), Slovenia
  - [Seges](#), Denmark
- Demonstration Use-cases
  - Cultivated Area
  - Crop Cycle
  - Crop Type
  - Crop Damage
  - Moisture Content
  - Crop Yield



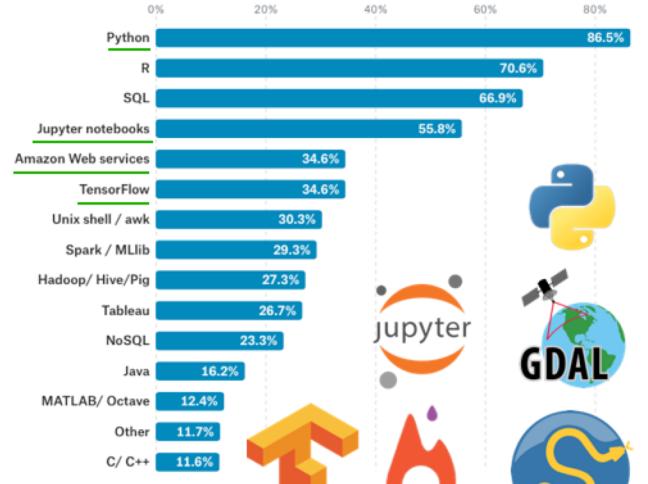
# eo-learn library



- collection of modular Python sub-packages that allow processing of spatio-temporal data to prototype, build, and automate large scale EO workflows
- operates on Area of Interest (AOI) of any size
- acts as a bridge between EO and Python ecosystem for data science and machine learning
- it's open source (MIT License)

*eo-learn enables experts and non-experts alike to explore, discover, and share value from vast amounts of satellite imagery*

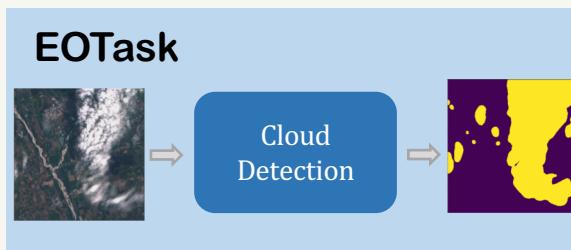
Tools used by data scientists in Industry (The State of Data Science '17)



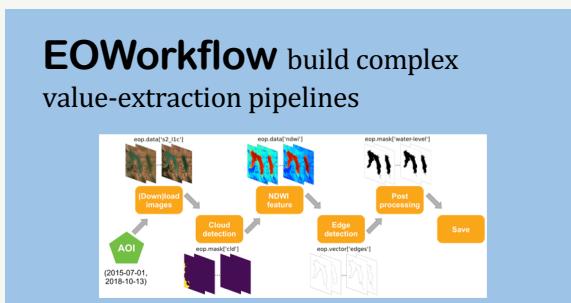
# eo-learn building blocks



- a common data-object for *spatio-temporal* EO and non-EO data, and their derivatives (numpy arrays, shapely polygons)
- sensor agnostic (optical, radar, ...)



- a single, well-defined action being performed on existing **EOPatch(es)**
- each **EOTask** takes an **EOPatch** as an input and returns a modified **EOPatch**
- little to no overhead in order to implement a new task

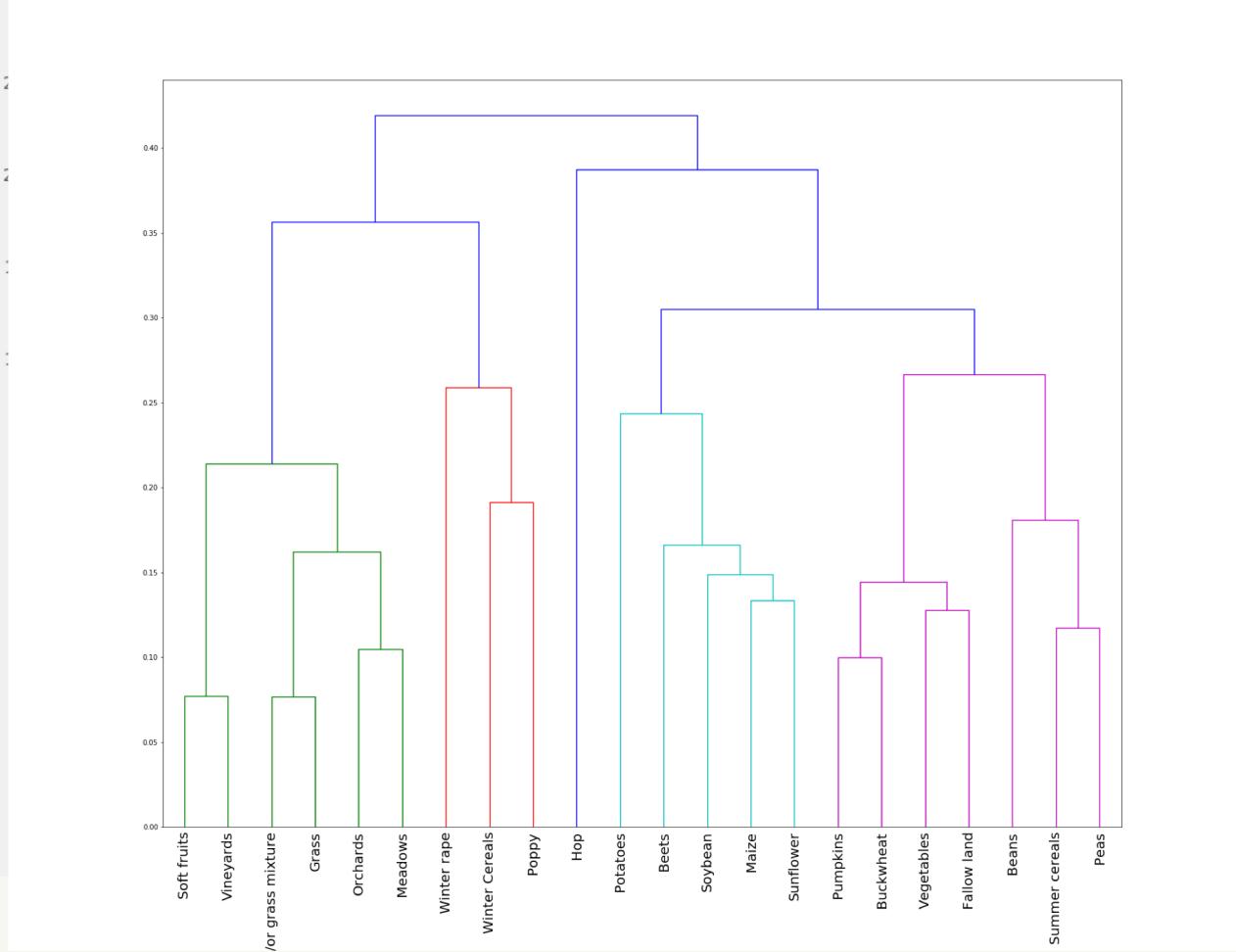


- a collection of **EOTasks** that together represent an *EO-value-adding-processing* chain

# Crop-type mapping

## Processing pipeline in a nutshell:

- Divide AOI into patches
- Download EO data
  - L2A
  - sen2cor's scene classification
  - s2cloudless
- Add non-EO data
  - GSAA/LPIS
- Rasterize
- Sample
- Develop crop-type classification algorithm
- Predict for AOI



# Hands on exercises

- Go to <https://github.com/sentinel-hub/eo-learn-workshop>
- Launch Binder instance

The screenshot shows a GitHub repository page for 'eo-learn-workshop'. At the top, there are buttons for 'Branch: master', 'New pull request', 'Create new file', 'Upload files', 'Find file', and 'Clone or download'. Below this is a list of commits by 'Anze Zupanc':

- Merge pull request #5 from sentinel-hub/check\_land\_cover
- Updated readme and image according to the new repo name.
- Updates the instructions to allow running the workshop with your own ...
- Last touch-ups of the sentinel1 notebook.
- Merge branch 'master' into feat/water-level
- Updated readme and image according to the new repo name.
- Create apt.txt
- first commit of data exploration notebook
- Presentation given as intro at the workshop.

Below the commit list is a preview of the 'README.md' file content:

Bridging Earth Observation data and Machine Learning in Python

[launch eo-learn tutorial](#)

Quick note to the reader

This material has been prepared for the Nordic Remote Sensing 2019 conference with everything set up beforehand. We encourage you to run the tutorial by yourself, but please have a look at the "Running the tutorial with your Sentinel Hub account" section below, which should help you with the setup. We will be happy to hear back from you; feedback is welcome.

At the bottom of the page are logos for Synchrone, ECOS, and the Agricultural Institute of Slovenia (Institut Štefan).

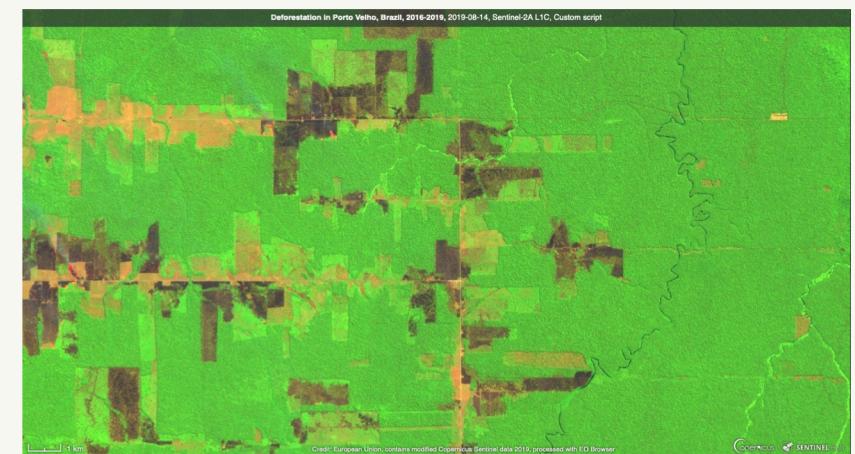
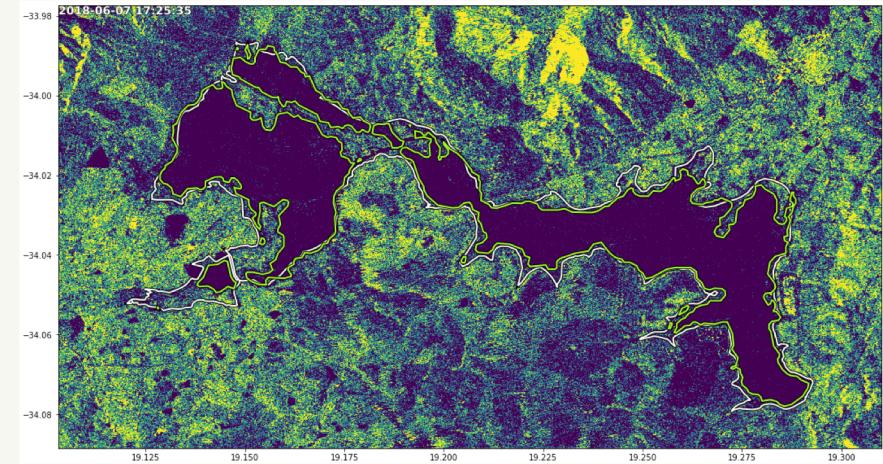
Click!

Remember (you're in control):

- eo-learn can be installed anywhere
  - Your laptop
  - Your cluster
  - DIASes
  - Cloud
  - AWS

# H2020 EO Big Data Hackathon Challenge ideas

- Participants of the Hackathon are challenged to
  - Run **eo-learn** on their own platforms.
  - Adapt **eo-learn** surface water level extraction workflow to work with Sentinel-1 data as an input.
  - Adapt **eo-learn** surface water level extraction workflow to detect deforested areas in Amazonia (i.e. Porto Velho, Brazil).
  - Run **s2cloudless** cloud detection algorithm as User Defined Function within openEO.
  - Run **eo-learn** workflow as User Defined Function within openEO.
  - Implement **eo-learn** task that uses openEO's API to get satellite imagery from any of the openEO's backends.



# H2020 Hackathon Outcomes

- openEO and eo-learn were successfully merged
  - a joint processing chain was executed on Vito's back-end
  - Code available on github

<https://github.com/iovsn/eohack>

iovsn / eohack

No description, website, or topics provided.

6 commits 1 branch 0 packages 0 releases 2 contributors

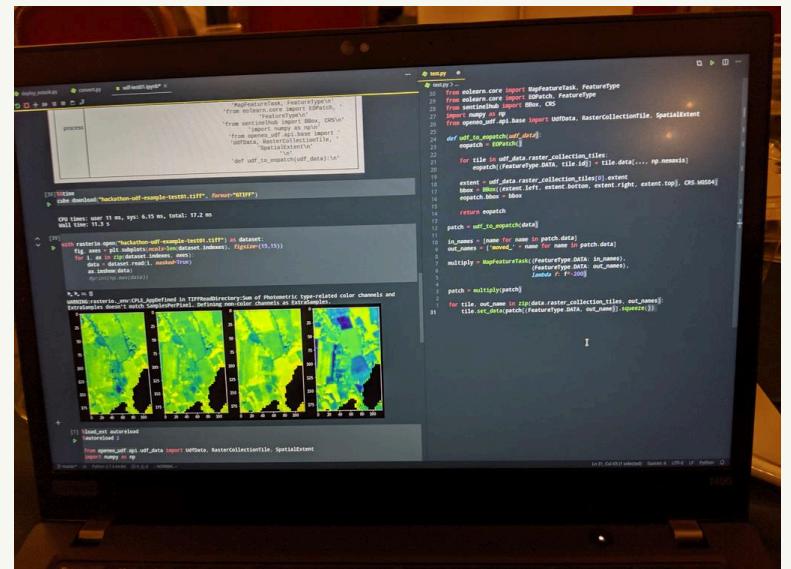
Branch: master New pull request Create new file Upload files Find file Clone or download

iovsn split notebooks

Latest commit 6cd38c9 10 days ago

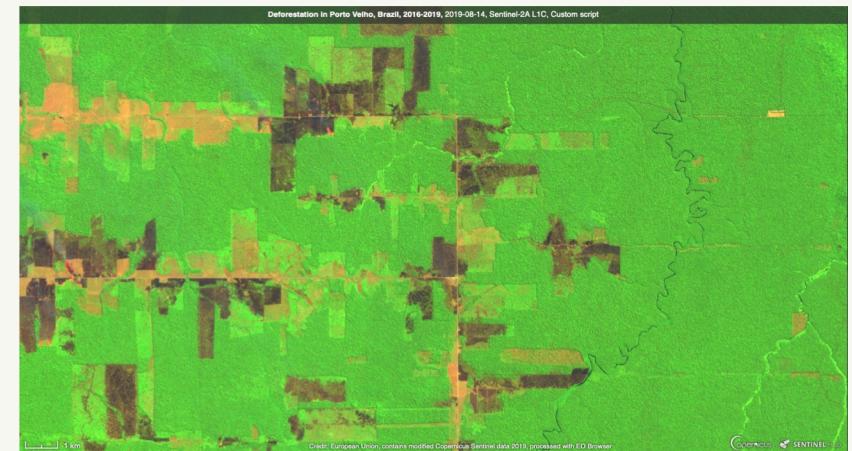
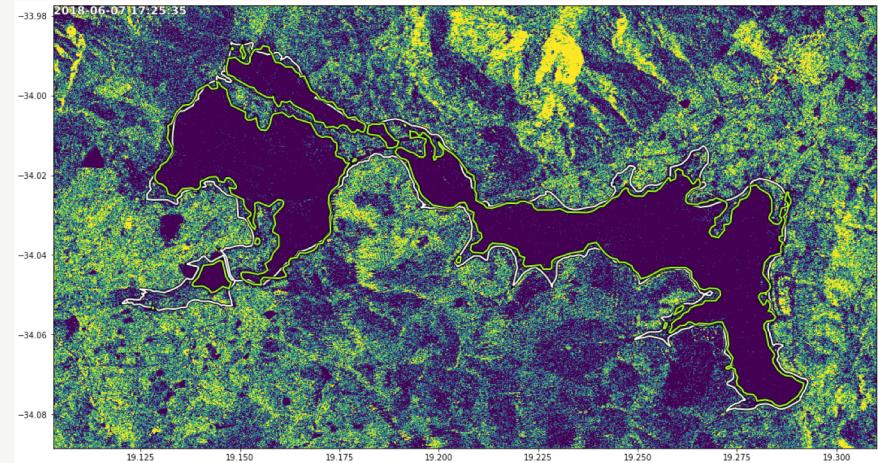
- README.md
- convert.py
- run\_udf.ipynb
- udf\_eopatch\_conversion.ipynb

A joint session of openEO and PerceptiveSentinel projects at the H2020 EO Big Data Hackaton.



# H2020 Hackathon Outcomes

- A lot of participants attempted to develop a new algorithm of surface water level extraction using Sentinel-1 and detection of deforestation in Amazonia
  - many first time users
  - due to WIFI issues participants working on the challenge were scattered
    - difficult to help all of them at the right time
  - But feedback that we got at the end was positive
    - Participants appreciated *getting hands dirty and let's code approach*



# More info

- <https://eo-learn.readthedocs.io/>
- <https://medium.com/sentinel-hub/>
- <https://github.com/sentinel-hub>
- <http://www.perceptivesentinel.eu>



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