TimeSAT - Ground Motion Pattern Detection and Classification in massive Satellite Image Time Series



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 Terradue srl., Roma, Italy;
 Terranum, Bussigny, Switzerland





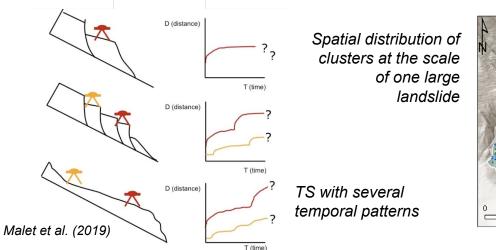
Context (1)

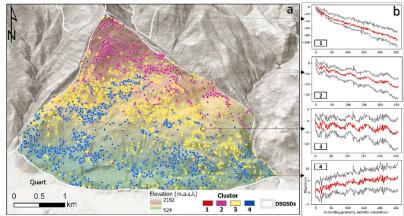
- Availability of massive ground motion products/datasets (large scales, long observation periods)
 - from MT-InSAR (e.g. EGMS, P-SBAS, LiCSAR, FLATSIM, SNAPPING, ...)
 - from MT-Optical Image Correlation (GDM-OPT, ITS_LIVE, ...)
- But challenges to exploit them and identify relevant signals
 - noisy observations
 - temporal patterns
 - spatial clusters
- Cloud/HPC processing for "rapid" analysis
 - signal processing challenges: discrete vs. non discrete information, outlier detection, time series transformation, …
 - technical processing challenges: scaling up, datacube formatting, ...

Context (2)

- Several identification and classification approaches:
 - The user knows which signal(s) is(are) of interest: supervised learning
 - The user wants to discover unknown signals: unsupervised approaches
- Service developed initially for landslide signals detection
 - e.g. ESA eo4alps-landslides project <u>eo4alps-landslides.eu</u>
- But usable for any ground deformation analysis





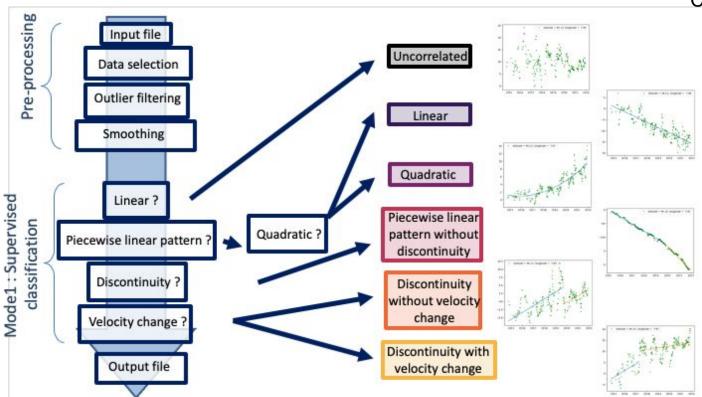


Festa et al. (2023)

TimeSAT: a satellite TS (time series) processing service on-line

- Goal
 - Identify temporal and spatial clusters of homogeneous styles of ground motion
- Input dataset
 - *discrete:* points clouds (e.g. ground motion time series from PS/DS InSAR)
 - *continuous:* datacube of optical offset-tracking time series
 - Input data in .csv or asset/stac compliant formats
- The service includes several components
 - *Pre-processing module:* outlier detection, smoothing, data selection
 - *Temporal time series module:* two modes of clustering / classification (based only on the temporal information)
 - Spatio-temporal time series module: clustering / classification (integrates spatial information from the features) ongoing work

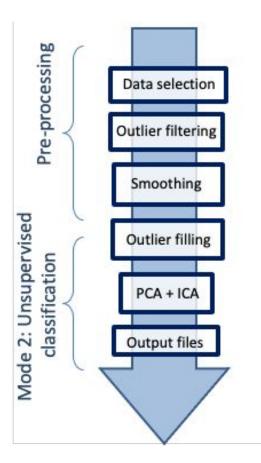
TimeSAT: Mode 1 - Supervised classification



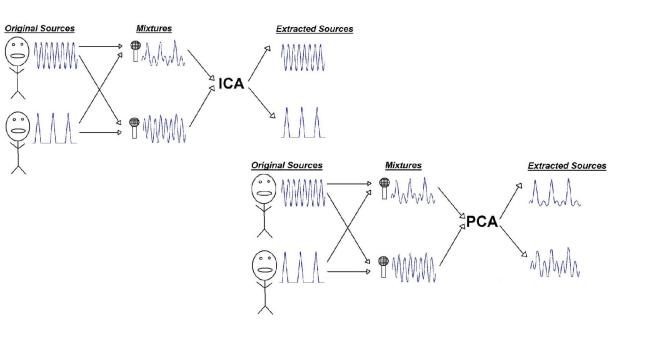
Classification:

- in six predefined distinctive ground motion trends
- based on a sequence of advanced statistical tests

TimeSAT: Mode 2 - Unsupervised classification (ICA/PCA)



Classification:using a combination of principal component analysis (PCA) and independent component analysis (ICA) to detect and classify specific patterns.



TimeSAT: technological developments for rapid/massive processing

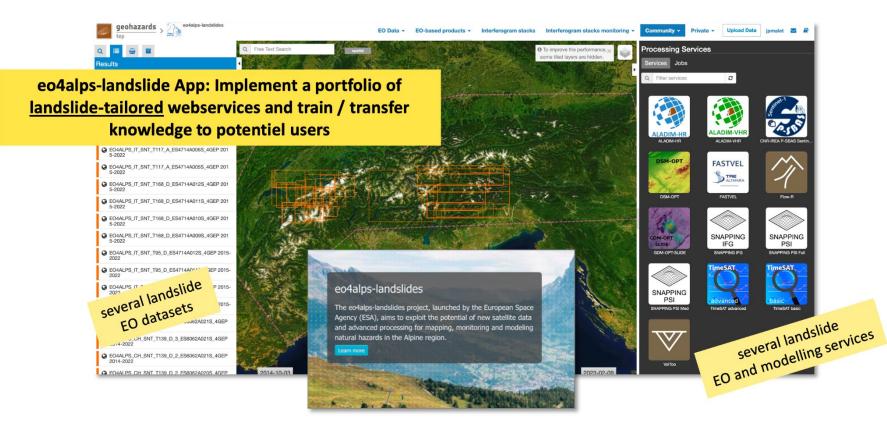
- All datasets are re-organized, prior to the TimeSat processing, in optimized datacubes (Zarr format) for easy transformation of the dimensions
 - spatial coordinates time value
 - time spatial coordinates value
 - value time spatial coordinates
- The time serie of value need to be contiguous in the computer memory (stride optimization using chunking/blocking strategies)





TimeSAT: Service on GEP

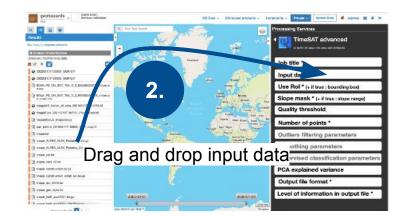
Developed with the support of ESA, and part of the eo4landslides App on GEP

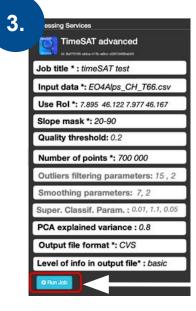


TimeSAT: service on GEP



Select the processing service TimeSAT





Tune the parameters:

4

- Job title: Defines the title of the job
- Input data: Input data in csv or kml format + nb of points
- **Time series classification approach:** Type of classification approach (e.g. supervised / mode 1 or unsupervised /mode 2 pattern identification or both
- Data selection: quality mask, AOI, slope mask
- **Pre-processing parameters:** filtering, smoothing, ...
- **Output**: format, level of information

Run the service

TimeSAT: tutorial on GEP

(docs.terradue.com/geohazards-tep/tutorials/timesat.html)



Site - Page - « STEMP - Surfa... VolToo: VOLum... » Source

TERRA)UE

Site +

Overview

Community Portal User Guide

Community Advanced User Guide

Community Portal Administrator Guide

Cloud Operations Administrator Guide

Thematic Applications

Processing tutorials

Active fire detection with Sentinel-3 SI STR

ALADIM-HR and ALADIM-VHR: Preparation of input datasets

ALADIM-HR: Automatic LAndslide Detection and Inventory Mapping from multispectral S2 & L8 data

ALADIM-VHR: Automatic LAndslide Detection and Inventory Mapping from multispectral Very-High Resolution data

Avalanche Mapping with Sentinel-1 [norce]

Burned Area Analysis with Sentinel-2

Change Detection Analysis [VAL] COIN - Coherence and Intensity change for Sentinel-1

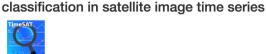
COLOC - Multi-mission data colocation [VAL, cppp]

COMBI - Band Combination

DIAPASON InSAR - StripMap(SM) **DIAPASON InSAR Sentinel-1**

TOPSAR(IW.EW) **DLR Sentinel-1 InSAR Browse** [VAL]

DSM-OPT: Digital surface models from optical stereo satellite images FASTVEL for displacement velocity



TimeSAT

Satellite image time series and derived products are increasingly available thanks to the launch of Earth Observation missions which aim at providing a coverage of the Earth every few days with high spatial resolution. The high revisit time of Copernicus (Sentinel-1. Sentinel-2) and Landsat satellites allow for the setup of systematic calculation of ground motion products, opening the way to science and operational monitoring capacities of geohazards. Many services are deployed in order to offer to users systematic or on-demand calculation of optical and InSAR time series products representing ground deformation. Satellite-derived products and services (e.g. EGMS; EPOS satellite products; GEP, Comet and ARIA services, etc) for the processing of SAR and optical imagery allow accessing displacement/velocity time series over large areas and time periods. Exploiting these datasets (stacks of interferograms, PSInSAR time series, optical derived ground motion, possibly organized in datacubes) necessitates the development of post-processing tools in order to combine the datasets and investigate the spatial and temporal behavior of the studied variables.

TimeSAT - Ground motion pattern detection and

TimeSAT is a service developed by CNRS-EOST (Strasbourg, France) for classifying ground motion displacement time series in specific behaviors/patterns, detect changes in the time series (increase, decrease, periodicity) and identify spatial clusters of homogeneous styles of ground motion. The service currently allows ingesting PSInSAR and SBAS InSAR time series and optical offset-tracking time series. It consists of:

A. A module for data pre-processing (advanced Savitzky-Golay filtering, data subset masking); B. A module for time series classification, for which several processing workflows are possible:

- Mode 1 / supervised: The classification in pre-defined distinctive patterns (uncorrelated trend, linear trend, guadratic trend, bilinear trend) based on a sequence of conditional statistical tests,
- · Mode 2 / unsupervised: The classification using a combination of principal component analysis (IPA) and independent component analysis (ICA) to detect and classify specific patterns.

TimeSAT allows the processing of time series non structured and unevenly distributed in time and in space. The workflow is computationally optimized and parallelised and is implemented on the Mésocentre/HPC infrastructure of the University of Strasbourg. Thanks to the parallelization and scaling of the code, the processing of about 1 million time series (pixel, PS/DS points) of 5 a years period lasts less than 2 hours.

TimeSAT Pre-processing is using Python librairies based on [1] and [2],

TimeSAT Mode 1 implements part of the supervised classification approach described in [3] using advanced statistical tests [4], [5], [6],

TimeSAT Mode 2 implements unsupervised classification approaches such as those used by [7], [8] and [9].

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Set the processing parameters

Some processing parameters can be adjusted. When hovering over the parameter fields, you will see a short explanation for each of the parameters

General

- · Job title: Defines the title of the job
- Input data: Input data in csv or kml format
- . Time series classification approach: Type of classification approach (e.g. supervised / mode 1 or unsupervised /mode 2 pattern identification or both)

Data selection

- · Use Rol (Region Of Interest) If set to True the spatial data selection is activated
- · Region Of Interest bounding box Define the bounding box of the Region of Interest (Rol)
- · Slope mask if set to True the data selection based on the slope in activated (Slope computed using Copernicus DEM)
- · Slope mask: range minimum / maximum Defines the slope range for which the points are keeping
- · Minimum quality threshold (coherence or correlation) Defines a minimum InSAR interferogram coherence OR coefficient of correlation for image correlation for which the points are keeping

Pre-processing

- · Pre-processing: Length of filtering window for outlier removal Defines the length of the sliding window for outlier. using Hampel identifier
- Pre-processing: Standard deviation value for outlier removal Defines the number of standard deviations to dete using Hampel Identifier
- Pre-processing: Length of filtering window for time series smoothing Defines the length of filtering window (Sat) filter): the value must be less than or equal to the number of dates in the time series
- · Pre-processing: Polynomial order for time series smoothing Defines the order of the polynomial used for sample (Savitzky-Golay filter): the value must be less than the length of the filtering window.

Mode 1: Supervised classification

- Supervised classification: Linear term level of significance Defines the linear ANOVA test threshold (a1); the time classified as uncorrelated pattern if q1 < p-value.
- Supervised classification: Quadratic term level of significance Defines the quadratic ANOVA test threshold (α2): series is classified as linear pattern if q2 < p-value.
- · Supervised classification: Piecewise linear pattern level of significance Defines the BIC threshold (Bth): the tim classified as piecewise linear pattern if the minimum BIC of segmented regression models is less than the BIC of qu linear models and if the evidence ratio is > Bth.



TimeSAT: processing report

TimeSAT Report Job : TimeSAT_Advanced_XXXX 2023-08-08 Date: Author: (c) CNRS-EOST (ForM@Ter) Type of input data: InSAR

Location:



Country: Switzerland AOI: center = (lon: 7.99, lat: 46.069), surface = 753 km²

Processing parameters:

slope mask	[0.0,90.0]
Quality threshold	0.5
% of explained variance	0.981

Dataset:

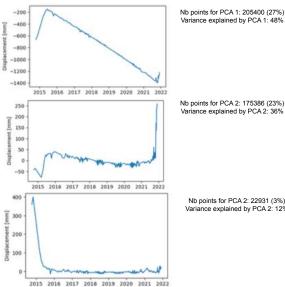
Input file	EO4ALPS_CH_SNT_T66_D_3_ES6062A021S_reformate.csv
Nb of measurement points:	761493
Nb of dates:	257 (2015-06-12 to 2022-05-23)

Results Mode #1 : Supervised classification:

0: uncorrelated	93584	12%
1: linear	0	0%
2: quadratic	1324	0.1%
3: piecewise linear with break point	362376	47%
4: piecewise linear with discontinuity	59946	8%
5: piecewise linear with discontinuity and velocity change	244263	32%

Results Mode #2 : Unsupervised classification:

Nb of PCA/ICA components: 25



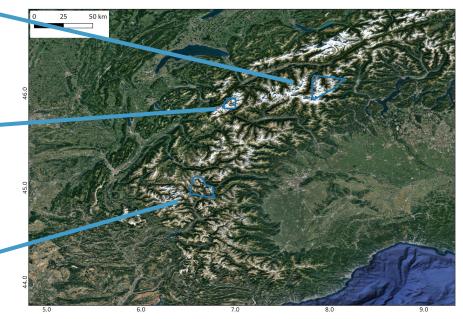
Nb points for PCA 2: 175386 (23%) Variance explained by PCA 2: 36%

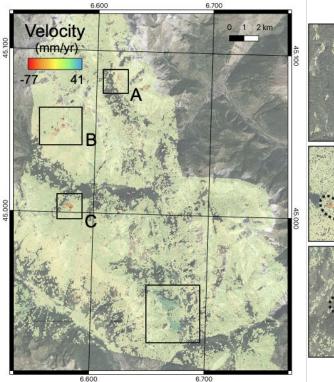
Nb points for PCA 2: 22931 (3%) Variance explained by PCA 2: 12%

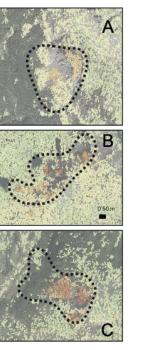


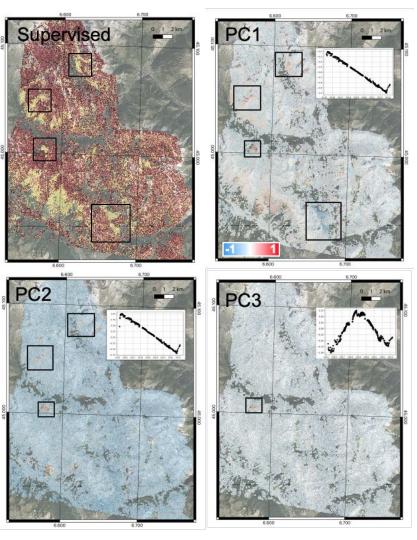
TimeSAT: use cases presented in this work

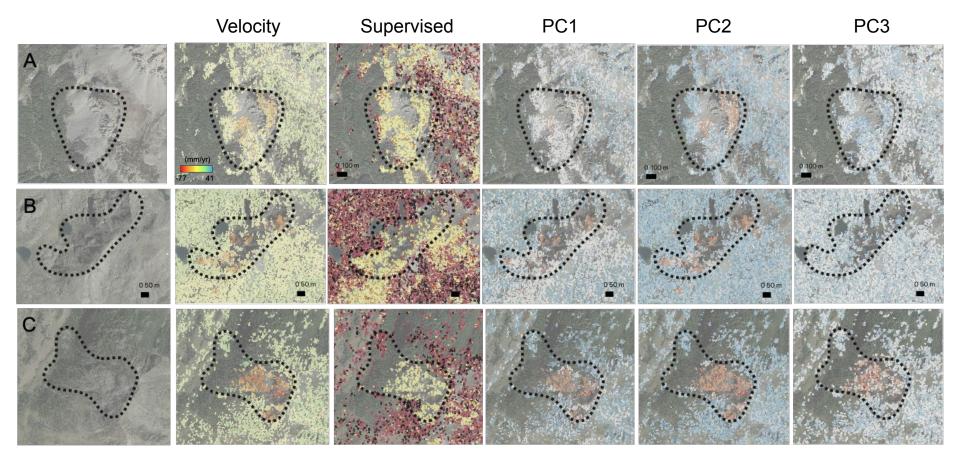
- Wallis Cantoon (Alps, Switzerland)
 - Ground motion time series from InSAR processing (*PS/DS*)
 - > 700.000 points, 257 dates (2014-10-10 to 2022-07-12)
- Mont-Blanc massif (Mer de Glace, France)
 - Optical offset-tracking time series (GDM-OPT-ICE)
 - > 300.000 points 158 dates (2015-08-29 to 2023-07-03)
- Briançonnais Massif (Alps, France)
 - Ground motion time series from InSAR processing (*PS/DS*)
 - >600.000 points, 298 dates (2015-04-08 to 2023-04-02)







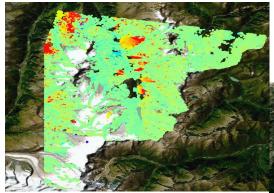




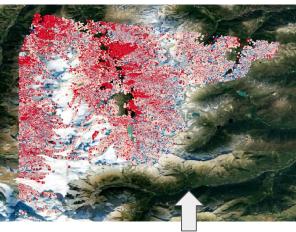
Velocity change (mm/yr)

TimeSat results, supervised mode

Significant velocity change before and after the breakpoint **Cluster of points of class 5** (TS with jump(s) and velocity change) Date of the Breaking Point (All points of the cluster highlight a breakpoint around a same date: spring 2016)



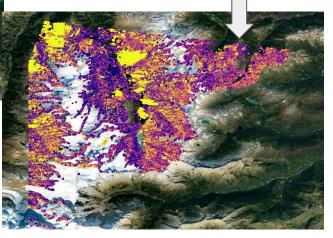
Linear trend of Input data (SqueeSAR for EO4alps on, ©TreMap)

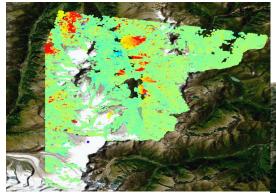


TimeSat results, supervised mode Class 5 in red highlights the instabilities (TS with jump(s) and velocity change)

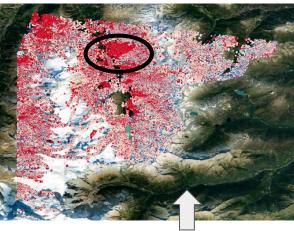
TimeSat results, unsupervised mode

For each point : weight of the first principal component. Max (yellow) and min (dark blue) values allow to identify the instability of the region.





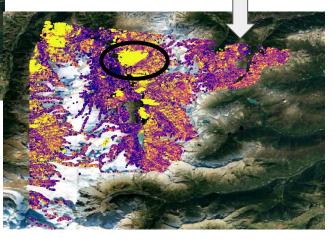
Linear trend of Input data (SqueeSAR for eo4alps-landslides, ©TRE-A)



TimeSat results, supervised mode Class 5 in red highlights the slope instabilities (TS with jump(s) and velocity change)

TimeSat results, unsupervised mode

For each point : weight of the first principal component. Max (yellow) and min (dark blue) values allow to identify the instability of the region.



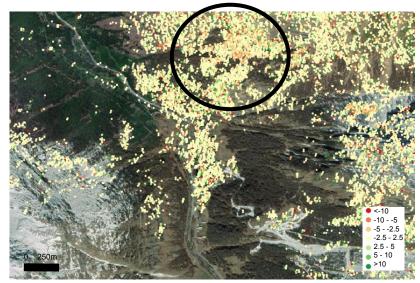
TimeSAT results, supervised mode



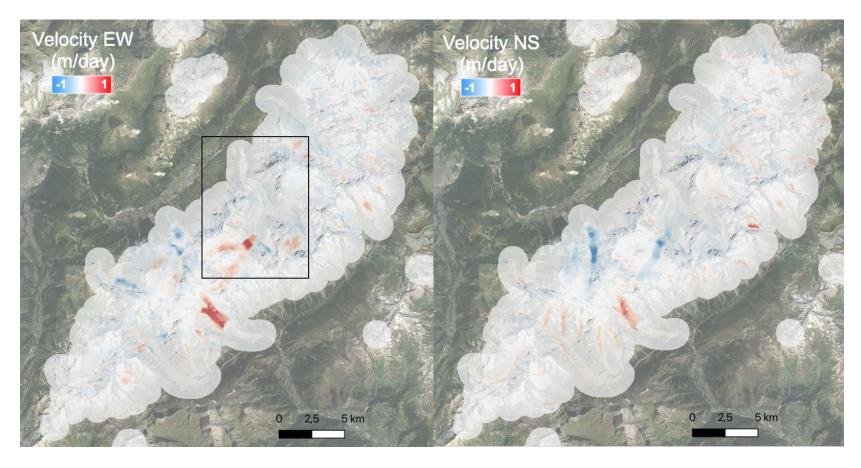
Date of the breaking point (All points of the cluster highlight a breakpoint around a same date: spring 2018)

Velocity change (mm.yr-1)

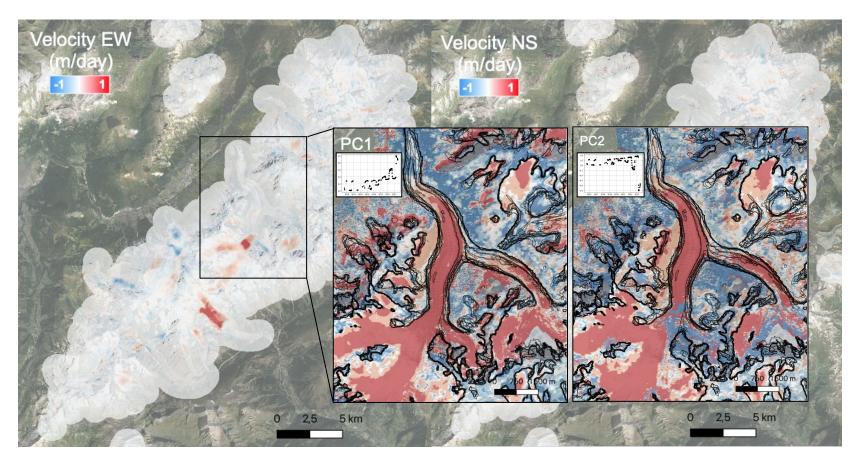
Velocity change before and after the breakpoint less significant than the previous example



TimeSAT: use case, over Mont-Blanc glaciers, preliminary results



TimeSAT: use case, over Mont-Blanc glaciers, preliminary results



Conclusion and perspectives

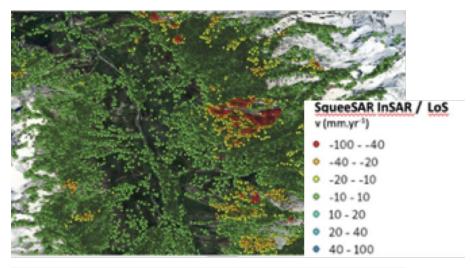
- TimeSAT is accessible on GEP since mid-2023
 - testers and feedbacks are welcome

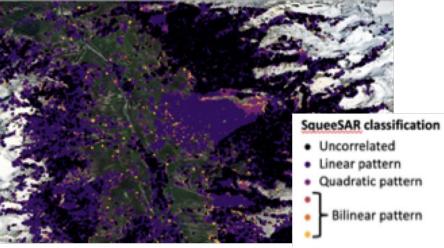


- many information are delivered to the user: class, breaking point date, velocity, etc.
- Current limitation: no clustering of the PCA/ICA
- Regular updates of the service
 - selection of one deformation pattern for the supervised classification
 - integrate other deformation pattern in the supervised classification
 - integrate methods for spatial and component clustering
 - at mid-term: towards a multi-source aggregator of motion measurements



Thanks for your attention!

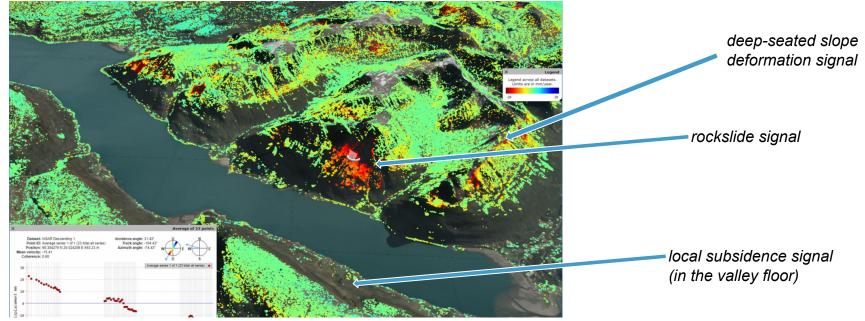




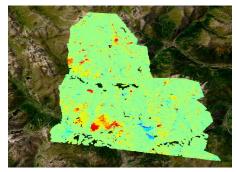
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Context (2)

- Several identification and classification approaches:
 - The user knows which signal(s) is(are) of interest: *supervised learning*
 - The user wants to discover unknown signals: unsupervised approaches



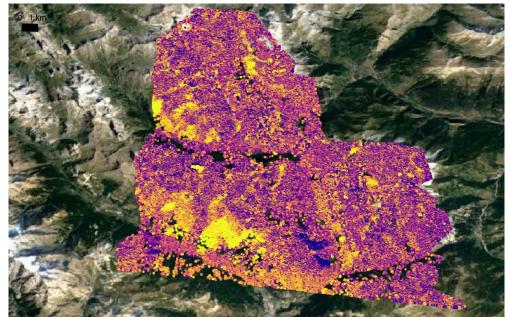
© InSAR Norway product

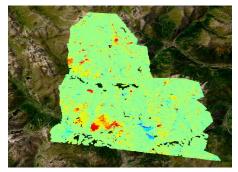


Linear trend of Input data (SqueeSAR for EO4alps on, ©TreMap)

TimeSat results, unsupervised mode

For each point : weight of the first principal component. Max (yellow) and min (dark blue) values allow to identify the instability of the region.

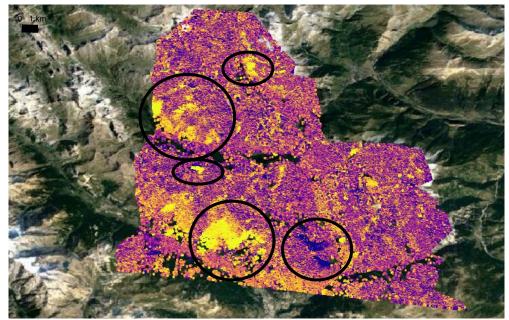


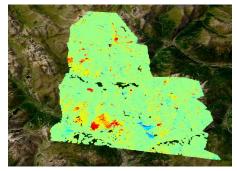


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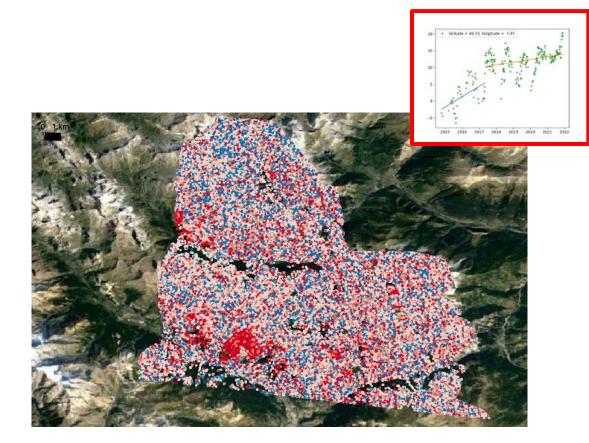


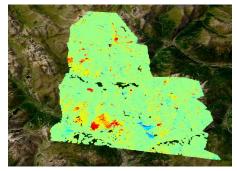
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TimeSat results, supervised mode

Class 5 in red highlights the instabilities

(TS with jump(s) and velocity change)



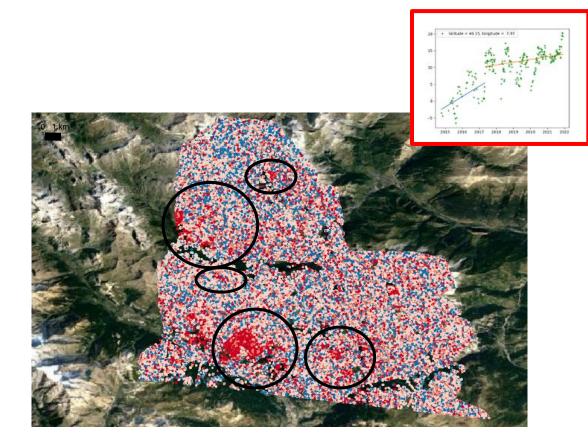


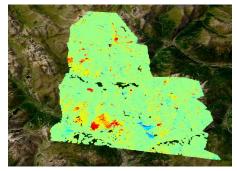
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TimeSat results, supervised mode

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