

Measuring earth surface deformation, glacier dynamics and geomorphic changes from times series of optical satellite images with COSI-Corr

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Motivation



- 1. The surface of Earth and other planet changing as a result of both internal and external dynamic processes.
- 2. The volume of remote sensing data available is increasing exponentially, driven by commercial and defense applications.
- 3. Science opportunities have emerged also because of development of specific techniques to exploit those data.

Our goal: to provide tools (open source) to exploit optical sensing data for the Earth and Planetary sciences

Overview



- COSI-Corr presentation
- New COSI-Corr release
- Application to glacier flow monitoring using optical time series images

COSI-Corr: Co-registration of Optically Sensed Images and Correlation

- NNSTITUTE OF THCHNOLO
- Alignment of satellite and aerial images with sub-pixel accuracy: sub-pixel accuracy ten times smaller than the size of pixel
- Tracking and monitoring of ground motion, displacement, or changes
- Measure and track 3D surface motion (not distributed)
- ENVI toolbox for download since 2007 (last release 2014)
- Developed with IDL (+ C++, FORTRAN)

Implements (Tools):



The software package is available at : http://www.tectonics.caltech.edu/slip_history/spot_coseis/index.html

MDIS 2019

COSI-Corr: supported platforms

- Sensor-agnostic, can use any imagery, even old archives to document long-term landscape evolution
- Currently available for the following platforms (Pushbroom and aerial frames cameras):
 - Aerial photographs (films and digital)
 - All spot satellites
 - ASTER instruments
 - Quickbird satellite
 - ➤ Wolrdview
 - HiRISE and CTX instruments (Mars), and LROC-Na instruments (MOON)

COSI-Corr application

Common applications: Tectonics , Geomorphology and Glaciology











Cosi-Corr workflow 1/2 (Pre-processing)



Cosi-Corr workflow 2/2 (Correlation and Post-processing)



Correlation

Given two images i1 and i2 such that: $i_2(x, y) = i_1(x - \Delta_x, y - \Delta_y)$ \rightarrow How to retrieve (Δ_x, Δ_y) ?

Multi scale Statistical correlation (NCC)

$$NCC_{i_1,i_2}(\Delta_x, \Delta_y) = \frac{\sum_{N_x} \sum_{N_y} (i_1(x, y) - \bar{i}_1) (i_2(x + \Delta_x, y + \Delta_y) - \bar{i}_2)}{\sqrt{\sum_{N_x} \sum_{N_y} (i_1(x, y) - \bar{i}_1)^2 \sum_{N_x} \sum_{N_y} (i_2(x + \Delta_x, y + \Delta_y) - \bar{i}_2)^2}}$$

Multi-scale phase correlation

Fourier Shift Theorem

$$\begin{split} i_2(x,y) &= i_1(x - \Delta_x, y - \Delta_y) \\ I_2(\omega_x, \omega_y) &= I_1(\omega_x, \omega_y) e^{-j(\omega_x \Delta_x + \omega_y \Delta_y)} \end{split}$$

Normalized Cross-spectrum

$$C_{i_1i_2}(\omega_x,\omega_y) = \frac{I_1(\omega_x,\omega_y)I_2^*(\omega_x,\omega_y)}{|I_1(\omega_x,\omega_y)I_2^*(\omega_x,\omega_y)|} = e^{j(\omega_x\Delta_x+\omega_y\Delta_y)}$$

Finding the relative displacement

$$\phi(\Delta_x, \Delta_y) = \sum_{\omega_x = -\pi}^{\pi} \sum_{\omega_y = -\pi}^{\pi} \frac{W(\omega_x, \omega_y)}{|C_{i_1 i_2}(\omega_x, \omega_y) - e^{j(\omega_x \Delta_x + \omega_y \Delta_y)}|^2}$$

W weighting matrix. (Δ_x, Δ_y) such that ϕ minimum.

- At high noise level, statistical methods performs better than phase correlation method,
- Using iterative re-weighted method on W mitigates the strong noise-free requirement of the image in the phase correlation formulation and adds more robustness to the solution

Overview



COSI-Corr presentation

> New COSI-Corr release

Application to glacier flow monitoring using optical time series images

New COSI-Corr release 1/2



Open source command line tool

- Python , C++
- GDAL
- Could be used as standalone package or added to other free software such as ASP, MICMAC ...

Pre-processing tools

- Optimization with RPC
- Automatic tie points determination and GCP generation

Correlation

• Regularization

Post-Processing

- Time series analysis tools
- Upgrade NLMF
- PCA (reconstruction and denoising)
- Local multiscale median filter
- Offset correction (de-ramping)
- Update the de-stripping tool
- > 3D deformation measurement

Similar to Coregis (Stumpf et al., 2018)



NLMF

Average of pixels with similar configuration in whole Gaussian neighborhood:



$$NL_h[u](\mathbf{x}) = \frac{1}{C(\mathbf{x})} \int_{\Omega} e^{-\frac{1}{h^2} \int_{\mathbb{R}^2} G_a(t) |u(\mathbf{x}+t) - u(\mathbf{y}+t)|^2 dt} u(\mathbf{y}) d\mathbf{y}$$

- Traditional NLMF uses a Gaussian weighting template with fixed weight coefficients when measuring the similarity of neighborhood distance
- When the noise level is large , the Gaussian weighting template with fixed weight coefficient will be subject to noise interference

To resolve the problem

- Adjusting the Gaussian weight coefficient with Laplace operator.
- An improved Non-Local Means Algorithms for Image Denoising (Leng,K. et al., 2017)

$$weight_j = \left|rac{G_{x_0,y_0}
abla^2 f(x_0,y_0)}{|\Sigma_{x,y} G_{x,y}
abla^2 f(x,y)|}
ight|$$

Local multiscale median filter

- Remove outliers of erroneous matches due to: clouds, snow cover or surfaces waters ...
- Local spatial filtering
- Initialize the PCA (in the spatial domain)









PCA (time series analysis)

> Noise removal (improve data quality)

- Local temporal filtering (temporal filtering)
- ➢ Re-weighting matrix

Visualization (modeling)





PCA (time series analysis)

- Noise removal (improve data quality)
- Local temporal filtering (temporal filtering)
- ➢ Re-weighting matrix
- ➢ Visualization (modeling)







PCA (resampling)





$$\widetilde{X}_{m \times \widetilde{n}} \approx \bigcup_{m \times k} \underbrace{S}_{k \times k} \widetilde{V^{T}}_{k \times \widetilde{n}}$$

3D deformation measurement

1. Optimize Viewing Parameters

- Pairwise image matching between all images,
- Only keep tie-points on stable surfaces (e.g., bedrock),
- Optimize external viewing parameters of all images jointly using regularized bundle adjustment.

2. Produce Disparity Maps

- Project all images on reference surface (e.g. DTM, GTOPO or smoothed GDEM),
- Cross-correlate image pairs using multi-scale, regularized image correlation.

3. Produce Point and Vector clouds (3D)

- Triangulate disparity maps,
 - (x₁, y₁, z₁)
 - (x_2, y_2, z_2)
 - $(x_1, y_1, z_1, D_x, D_y, D_z)$
- Output surface models at all times.

4. Grid Point Clouds and Vector Clouds

- Use standard gridding libraries on each components (only external processing).



3D deformation measurement

- Solving for 3-D displacements and the topography with 4 images.
- > Triangulate multiple disparity maps to retrieve 3D topography and displacement fields





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Location of the Shishper glacier

Current world-wide glaciers (Randolph Glacier Inventory) & Mass balance 2012-2016 (ice sheets: Bamber et al., 2018)





- Negative mass balance = melting worldwide
- > Several glacier surges were recorded during the last decades in the Karakoram region
- Hispar glacier : Paul et al., 2017
- Kyagar glacier : Round et al., 2017
- Shishper glacier : Begum et al., 2019

Why studying this glacier ?





- ✓ Surge reported in July 2018 in local press
- $\checkmark\,$ No velocity pattern studies of this surge
- ✓ Risk of GLOF (Glacial lake outburst floods)
- ✓ Risk : 40,000 people living downstream in the Hunza valley



Dataset and workflow





10/18/2019

Results



2016-05-21#2016-07-20 (SL) diff=60 EW velocity map

Before post-processing



After offset correction



After multi-scale Filtering



Results

2017-07-25#2017-11-02 (SL)



2018-04-01#2018-04-06 (SL)



2018-11-17#2018-12-02 (SL)







2018-05-01#2018-05-11 (SL)



2018-12-02#2018-12-07 (SL)



2018-01-26#2018-02-05 (SL)



2018-07-10#2018-07-15 (SL)



2019-08-04#2019-08-09 (SL)



- Apparition of the ice-dammed lake between 2018 and 2019
- Stronger and longer signal in 2019 at top and tongue
- 2019 shows a stronger dynamic than previous years, even 2018

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Results









