

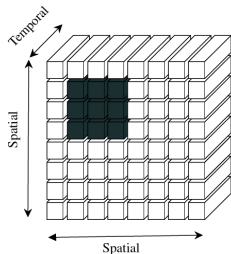
Sequential learning of SAR image time series for monitoring Earth deformation

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17 June, 2026

Mathematical modeling of multi-temporal SAR data



For a given pixel x_i with p temporal observations:

$$x_i = [x_i^1, x_i^2, \dots, x_i^p]^T$$

An homogeneous neighbourhood (similar backscattering and statistical properties): $\Omega = \{x_i\}_{i=1}^n$

Second order statistic: $E[x^q(x^l)^H] = \gamma_{q,l} \sigma_q \sigma_l e^{j(\theta_q - \theta_l)}$
in matrix form

$$\mathbb{E}[xx^H] \triangleq \Sigma = \text{eddiag}(w_\theta) \underbrace{((\sigma\sigma^T) \circ \Gamma)}_{\Psi} \text{eddiag}(w_\theta)^H$$

with $w_\theta = [e^{j\theta_1}, \dots, e^{j\theta_p}]$

$$\Sigma(\Psi, \theta) = \text{mod}(\Sigma) \odot \arg(\Sigma) \triangleq \Psi \odot (w_\theta w_\theta^H)$$

Σ complex covariance matrix
— interferogram

Ψ module of the covariance
matrix

Γ coherence matrix

w_θ vector of phase exponential

State of the art: Phase Linking

- Objective: Estimate $p - 1$ phase differences from p SAR images $\rightarrow w_\theta$
- Main idea: minimize the negative log-likelihood

$$\begin{aligned} \min_{\Psi, w_\theta} \quad & \mathcal{L}(\Sigma(\Psi, \theta)) \\ \text{subject to} \quad & \Psi \text{ real and symmetric} \\ & \theta_0 = 0 \end{aligned}$$

Assuming that Ψ is known, the method is equivalent to optimizing the following problem:

$$\begin{aligned} \min_{w_\theta} \quad & w_\theta^H (\Psi^{-1} \circ \mathbf{S}) w_\theta \\ \text{subject to} \quad & \theta_1 = 0 \end{aligned}$$

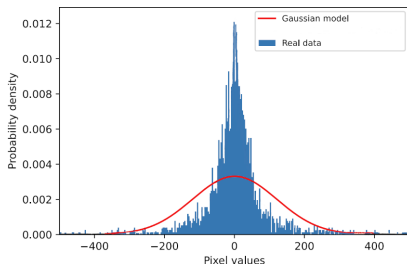
with $\mathbf{S}$ sample covariance matrix

In reality, Ψ is unknown

- [1] proposed to use a plug-in estimator: $\Psi = |\mathbf{S}| \rightarrow$ **not optimal**
- [2] proposed to estimate Ψ jointly with w_θ

- [1] A. Guarnieri et al. (2008) "On the exploitation of target statistics for SAR interferometry applications".
[2] P.V.H. Vu et al. (2022) "A new phase linking algorithm for multi-temporal InSAR based on the Maximum Likelihood Estimator".

State of the art: Phase Linking



Empirical distribution of a Sentinel-1 image (July 3 2019) over the Mexico City.

- Gaussian distribution

$\{\mathbf{x}^i\}_{i=1}^n$ an i.i.d set, $\rightarrow \mathbf{x} \sim \mathcal{CN}(0, \Sigma)$

$$\begin{aligned}\mathcal{L}(\mathbf{x}; \Sigma) &= -\log \left(\prod_{i=1}^n f(\mathbf{x}^i, \Sigma) \right) \\ &\propto n \log(|\Sigma|) + n \text{Tr}(\Sigma^{-1} \mathbf{S})\end{aligned}$$

- Scaled Gaussian distribution

$\{\mathbf{x}^i\}_{i=1}^n$, an i.i.d set, $\mathbf{x}^i \sim \mathcal{CN}(0, \tau_i \Sigma)$

$$\begin{aligned}\mathcal{L}(\mathbf{x}; \Sigma, \{\tau_i\}_{i=1}^n) &= n \log |\Sigma| + n \text{Tr}\{\Sigma^{-1} \mathbf{S}\} \\ &\quad + \sum_{i=1}^n p \log \tau_i + \text{const}\end{aligned}$$

$\rightarrow$ closed-form solutions in both cases

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- [1] A. Guarnieri et al. (2008) "On the exploitation of target statistics for SAR interferometry applications".
[2] P.V.H. Vu et al. (2022) "A new phase linking algorithm for multi-temporal InSAR based on the Maximum Likelihood Estimator".

State of the art: covariance fitting

Minimizing a dissimilarity measure between an empirical covariance matrix and the theoretical covariance matrix.

$$\begin{aligned} \min_{\mathbf{w}_\theta} \quad & f^d(\tilde{\Sigma}, \Psi \odot \mathbf{w}_\theta \mathbf{w}_\theta^H) \\ \text{subject to} \quad & \theta_1 = 0 \end{aligned}$$

Possible options

Matrix distance f^d :

- Kullback - Leibler divergence
- Frobenius norm (least square)

Plug-in (empirical) estimator $\tilde{\Sigma}$:

- Sample covariance
- M-estimator
- Phase-only matrix ($re^{j\phi} \rightarrow e^{j\phi}$)

Considering decorrelation mechanism

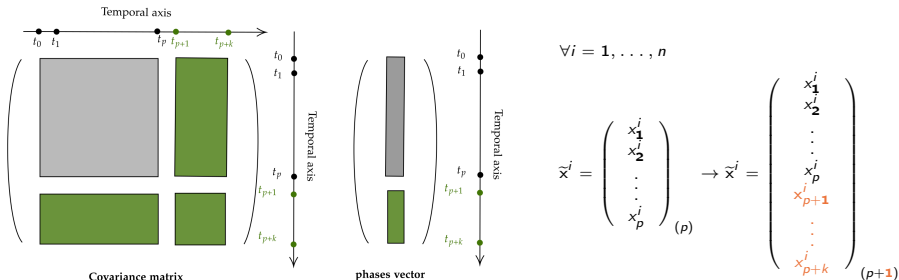
Regularization:

- Low rank
- Banding
- Shrinkage

$$\tilde{\Sigma}(\beta) = \beta \hat{\Sigma} + (1 - \beta) \frac{\text{tr}(\hat{\Sigma})}{p} \mathbf{I}$$

with $\beta \in [0, 1]$.

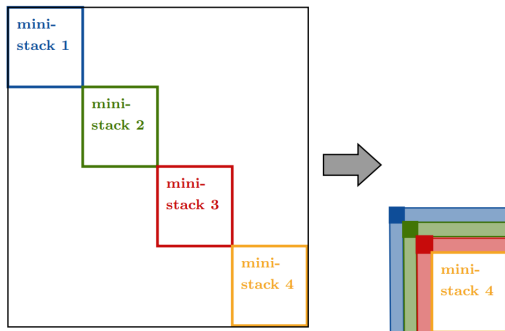
When more images are coming ...



Integrating **new** SAR acquisitions

- **Estimation** of the increasing covariance matrix, **Re-Estimation** of the past covariance matrix
- **Estimation** of the new phases, **Re-Estimation** of the past phases
 - **Huge computation time**
 - **Need of sequential approaches**

State-of-the-art: Sequential estimator



Processing steps:

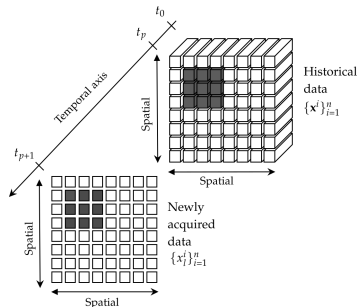
- Mini stack division: split the time series into small temporal blocks
- Phase linking: estimate phase values in each block
- Date compression: reduce the data to one virtual image for each block
- Datum connection: link all mini-stacks by a global phase linking

Potential drawbacks

- Need time to get meaningful mini stack
- Based on Gaussian assumptions on SAR data
- Lack of robustness against various decorrelation mechanism

- Sequential Phase Linking based on Maximum Likelihood Estimation (S-MLE-PL)
- Sequential Covariance Fitting Interferometric Phase Linking (S-COFI-PL)
- SLiding interferometric Phase Linking (SL-IPL)

Sequential MLE Phase Linking



- Updated data

$$\tilde{\mathbf{x}}^i = [\underbrace{x_1^i, \dots, x_p^i}_{\mathbf{x}^i}, x_{p+1}^i]^T \in \mathbb{C}^{l=p+1}$$

- Incremental update of the covariance

$$\tilde{\Sigma} = \begin{pmatrix} \Sigma & w_{\theta_l}^* \text{diag}(\mathbf{w}_{\theta}) \gamma^T \\ \gamma \text{diag}(\mathbf{w}_{\theta})^H w_{\theta_l} & \gamma_l \end{pmatrix}$$

- Incremental update of the log-likelihood (scaled Gaussian distribution) - closed form solution

$$\begin{aligned} \min_{\gamma, \gamma_l, w_{\theta_l}, \{\tau_i\}_{i=1}^n} \quad & \mathcal{L}(\tilde{\mathbf{x}}^i; \gamma, \gamma_l, w_{\theta_l}, \{\tau_i\}_{i=1}^n) \\ \text{subject to} \quad & \theta_1 = 0, |w_{\theta_l}| = 1, \gamma, \gamma_l \text{ real} \end{aligned}$$

Summary: w_{θ} is estimated with the past $\hat{w}_{\theta}$ and $\hat{\Sigma}$ as well as the new data

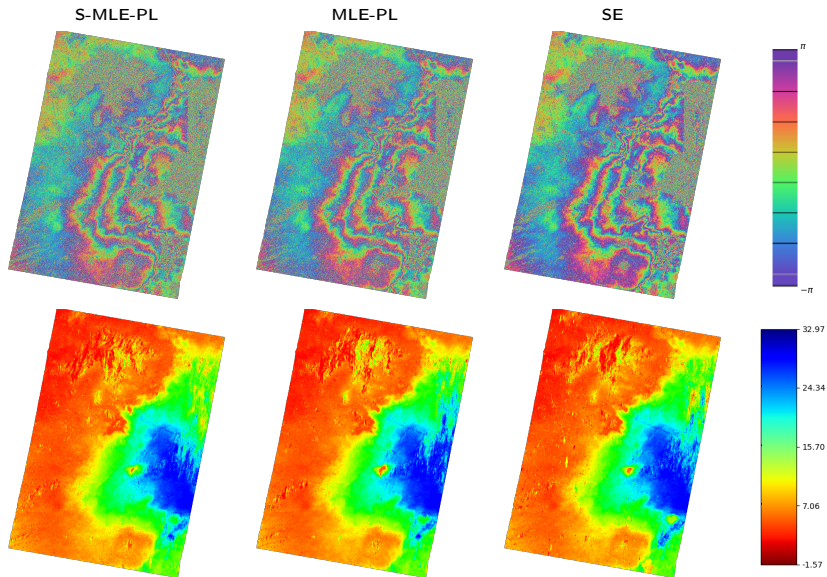
→ conditional statistics of the new phase on the past.

D. El Hajjar et al. (2024) "Sequential Phase Linking : Progressive Integration of SAR Images for Operational Phase Estimation."

D. El Hajjar et al. (2025) "Robust Sequential Phase Estimation Using Multi-Temporal SAR Image Series"



Sequential MLE Phase Linking



Mexico City, Sentinel-1, 14/04/2019 - 10/04/2020, time series of 20 images, SE with blocks of 5 images.

Sequential MLE Phase Linking

- **S-MLE-PL**: uses $p = 19$ historical images and one new image.
- **MLE-PL** [6]: uses all $I = 20$ available images.
- **SE** [4]: uses a temporal sliding window of size $s = 5$ images.

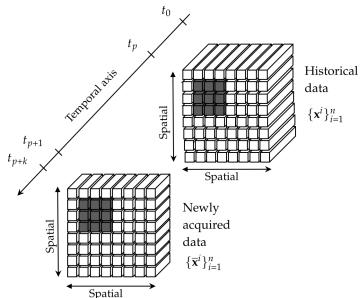
	RMSE (↓)	SSIM (↑)	Colinearity (↑) [3]	Computation time
MLE-PL [6]	Ref	Ref	0.61	38.4 min
S-MLE-PL	2.87	0.70	0.57	4.72 min
SE [4]	2.96	0.67	0.76	16.97 min

with a 95-core CPU running at 2.2 GHz and 125 G of RAM with calculations executed in parallel across the CPUs.

SSIM: Structural SIMilarity

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- [4] H. Ansari et al. (2017) "Sequential Estimator: Toward efficient InSAR time series analysis".
[5] B. Pinel-Puysségur et al. (2012) "Multi-link InSAR timeseries: Enhancement of a wrapped interferometric database".
[6] P.V.H. Vu et al. (2023) "Robust Phase Linking in InSAR".

Sequential Covariance Fitting Interferometric Phase Linking



- Updated data vector

$$\tilde{\mathbf{x}}^i = [\underbrace{x_1^i, \dots, x_p^i}_{\mathbf{x}^i}, \underbrace{x_{p+1}^i, \dots, x_{p+k}^i}_{\tilde{\mathbf{x}}^i}]^T \in \mathbb{C}^{p+k}$$

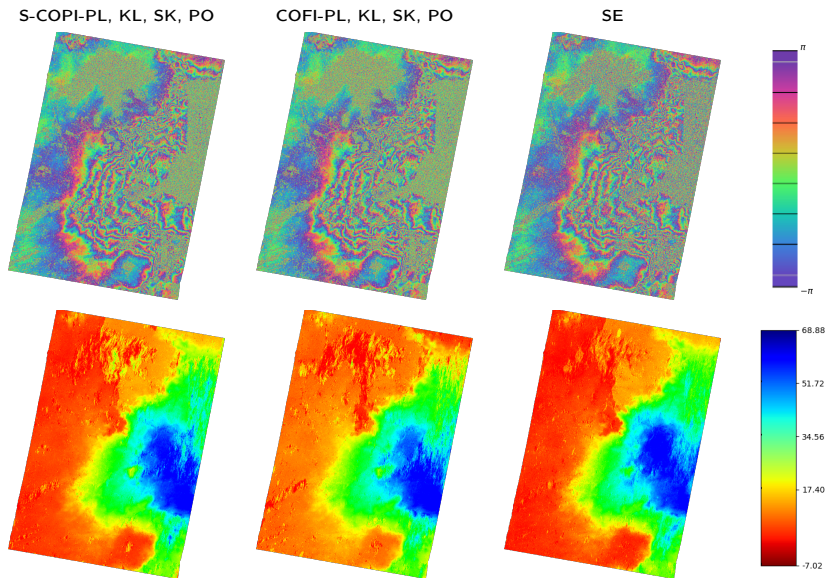
- Incremental update of the covariance

$$\tilde{\Sigma} = \begin{pmatrix} \Sigma_p & (\Sigma_{pn})^H \\ \Sigma_{pn} & \Sigma_n \end{pmatrix}$$

- Optimization (Minimization - Majorization algorithm)

$$\begin{aligned} \min_{\tilde{\mathbf{w}}_{\theta}} \quad & f_{\tilde{\Sigma}}^d(\tilde{\Sigma}, \tilde{\Psi} \circ \tilde{\mathbf{w}}_{\theta} \tilde{\mathbf{w}}_{\theta}^H) \\ \text{subject to} \quad & \theta_1 = 0 \\ & \tilde{\mathbf{w}}_{\theta} \in \mathbb{T}_k \end{aligned}$$

Sequential Covariance Fitting Interferometric Phase Linking



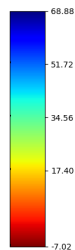
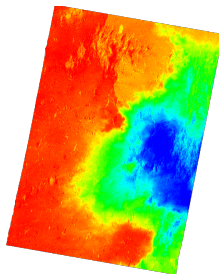
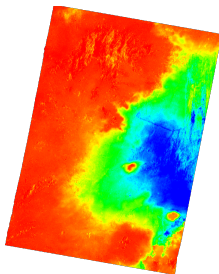
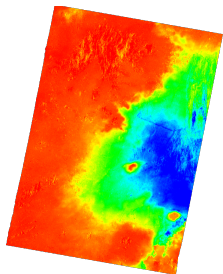
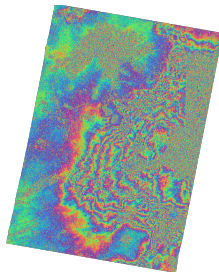
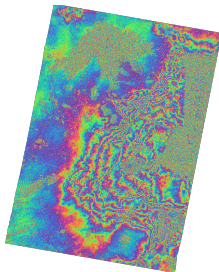
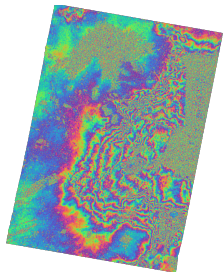
Mexico City, Sentinel-1, 14/04/2019 - 06/12/2020. time series of 40 images, block of 5 images.

Sequential Covariance Fitting Interferometric Phase Linking

S-COPI-PL, FN, BW, PO

COFI-PL, FN, BW, PO

SE



Mexico City, Sentinel-1, 14/04/2019 - 06/12/2020. time series of 40 images, block of 5 images.

Sequential Covariance Fitting Interferometric Phase Linking

- **S-COFI-PL**: uses $p = 35$ historical images and $k = 5$ new images.
- **COFI-PL** [7]: uses all $I = 40$ available images.
- **SE** [4]: uses a temporal sliding window of size $s = 5$ images.

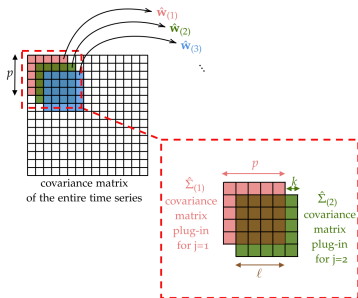
		RMSE (↓)	SSIM (↑)	Colinearity [5] (↑)	Computation time
KL	S-COFI-PL	7.54	0.62	0.64	0.44h
SK-PO	COFI-PL	7.73	0.55	0.60	0.52h
FN	S-COFI-PL	2.07	0.89	0.89	0.37h
BW-PO	COFI-PL	Ref	Ref	0.89	0.43h
SE [4]		7.47	0.24	0.72	0.5h

with a 95-core CPU running at 2.2 GHz and 125 G of RAM with calculations executed in parallel across the CPUs.

KL : Kullback-Leibler divergence; FN : Frobenius Norm; BW : banding; SK : Shrinkage-to-identity;
PO : Phase-Only; SSIM : Structural SIMilarity

Sliding interferometric Phase Linking

Temporal sliding window:



p : size of the temporal window

k : size of the stride or the new added images

$\ell = p - k$: size of the overlap

No need of storing all the historical data — memory and computational efficiency

- Covariance fitting formulation

$$\begin{aligned} \min_{\mathbf{w}_{(j)}} \quad & f_{\hat{\Sigma}_{(j)}}^{FN}(\Sigma_{(j)}, \Psi_{(j)} \circ \mathbf{w}_{(j)} \mathbf{w}_{(j)}^H) \\ & + \lambda h(\mathbf{w}_{(j)}) \\ \text{subject to} \quad & \mathbf{w}_{(j)} \in \mathbb{T}_p \end{aligned}$$

with a penalty term h that guarantees the phase estimates to stay close to the previous ones in the window overlap

$$h(\mathbf{w}_{(j)}) = \left\| \begin{pmatrix} \mathbf{w}_{\ell(j-1)} \\ 0 \end{pmatrix} - \mathbf{w}_{(j)} \right\|^2$$

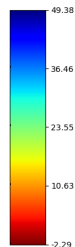
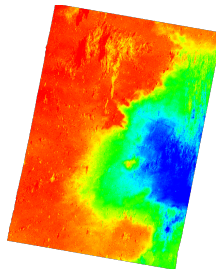
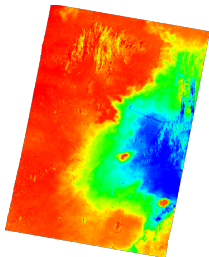
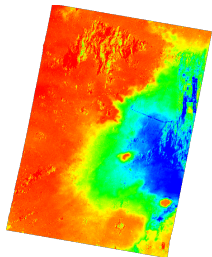
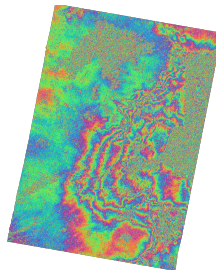
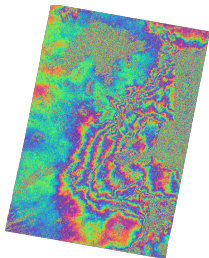
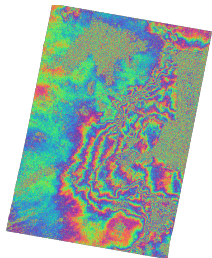
- Optimization: Minimization - Majorization algorithm

Sliding interferometric Phase Linking

SL-IPL

COFI-PL, FN

SE



Mexico City, Sentinel-1, 14/04/2019 - 06/12/2020. time series of 30 images, block of 5 images, stride = 1.

Sliding interferometric Phase Linking

- **SI-IPL**: uses $p = 5$, $\ell = 4$, and $k = 1$.
- **COFI-PL** [7]: uses all $I = 30$ available images.
- **SE** [4]: uses a temporal sliding window of size $s = 5$ images.

method	colinearity [5] ($\uparrow$)	SSIM ($\uparrow$)	Computation time
COFI-PL [7]	0.91	Ref	19.35 min
SI-IPL	0.90	0.94	3.9 min
SE [4]	0.84	0.85	17.1 min

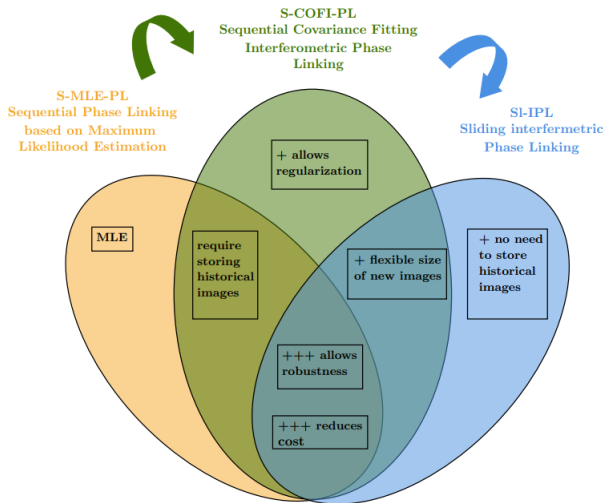
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SSIM: Structural SIMilarity

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[5] B. Pinel-Puysségur et al. (2012) "Multi-link InSAR timeseries: Enhancement of a wrapped interferometric database".

Conclusions



Codes availability:

<https://github.com/DanaElhajjar/S-G-MLE-PL>

<https://github.com/DanaElhajjar/S-COFI-PL>

<https://github.com/DanaElhajjar/S-SG-MLE-PL>

<https://github.com/DanaElhajjar/SI-IPL>