

Monitoring and Interpreting Deformation along Linear Infrastructure Using Deep Clustering of MT-InSAR Analyses

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FRINGE 2023

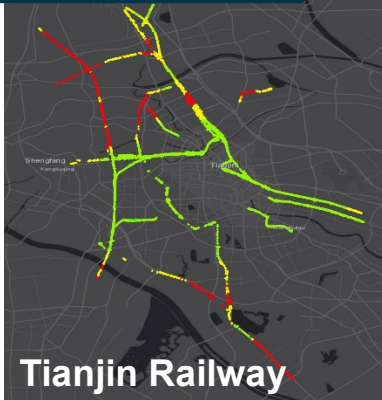
University of Leeds, UK | 11 - 15 September 2023.



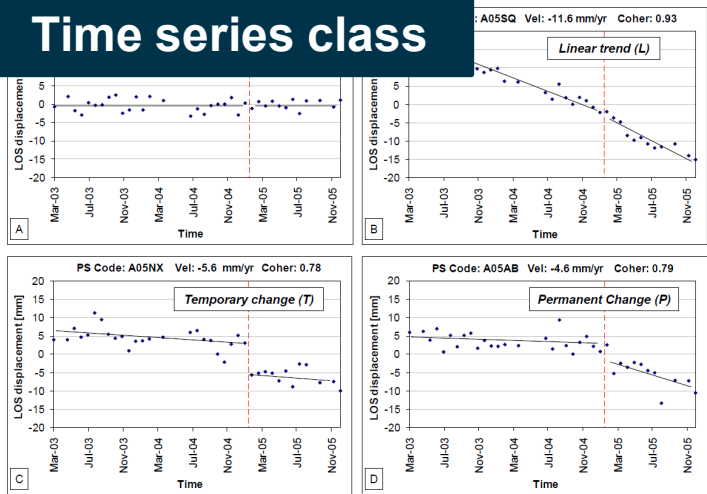
Motivation



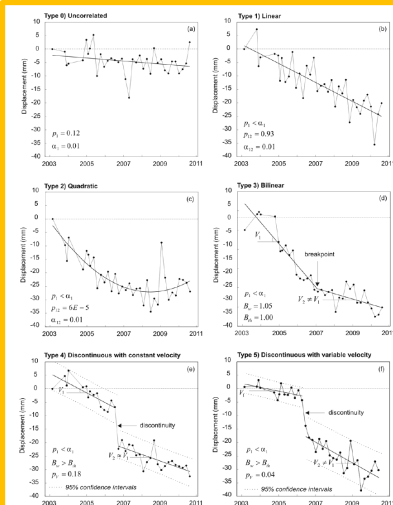
Deformation velocity



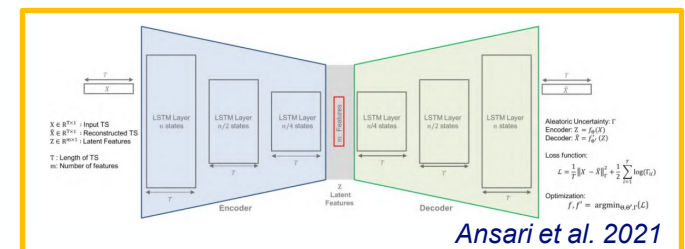
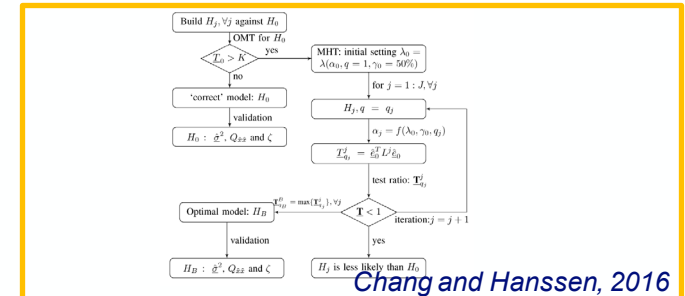
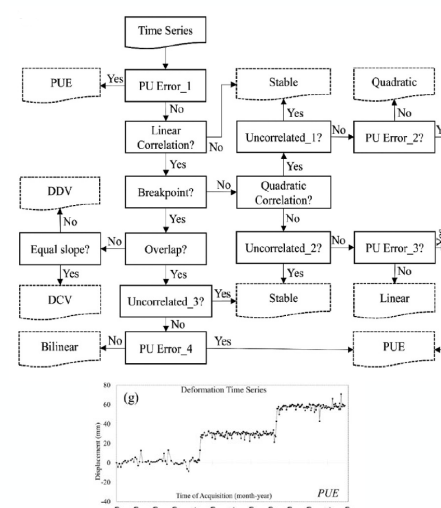
Time series class



Cigna et al., 2011

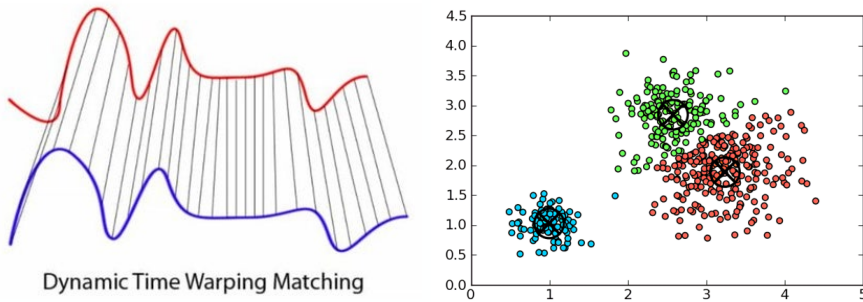


Berti et al., 2013 S. Mohammad Mirmazloumi et al., 2022

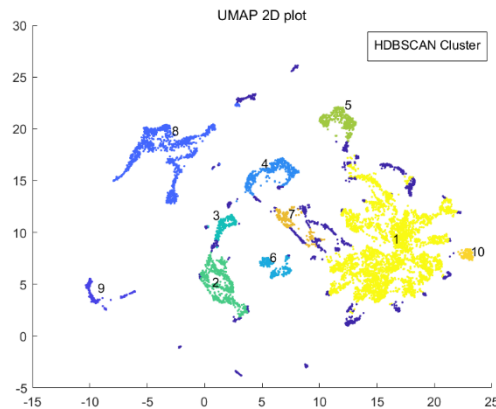


Baseline method:

1. K-mean (DTW distance)

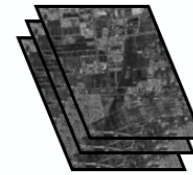


2. UMAP+HDBSCAN

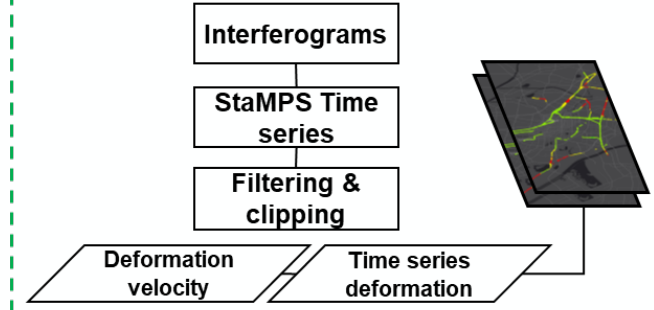


Proposed method:

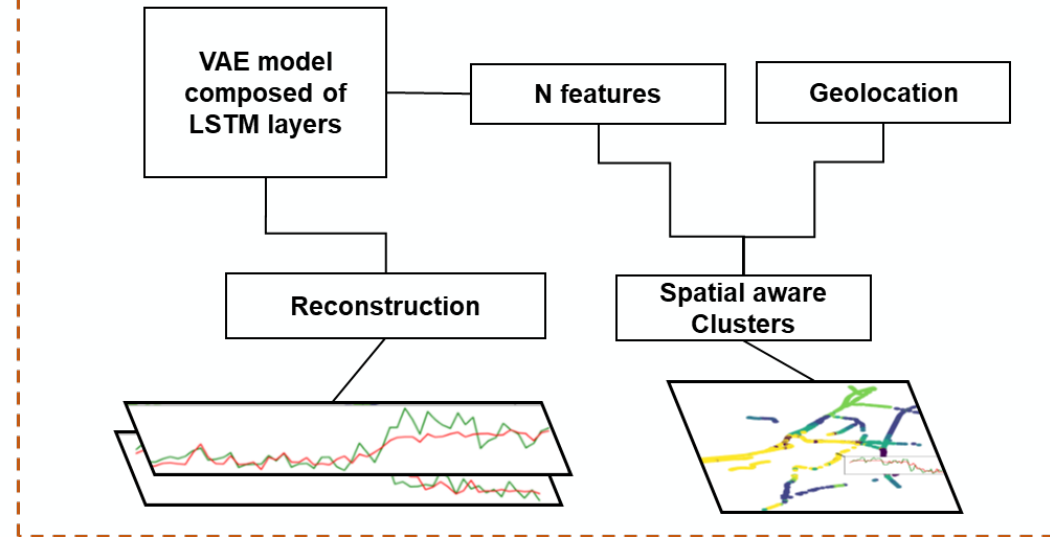
TerraSAR-X dataset



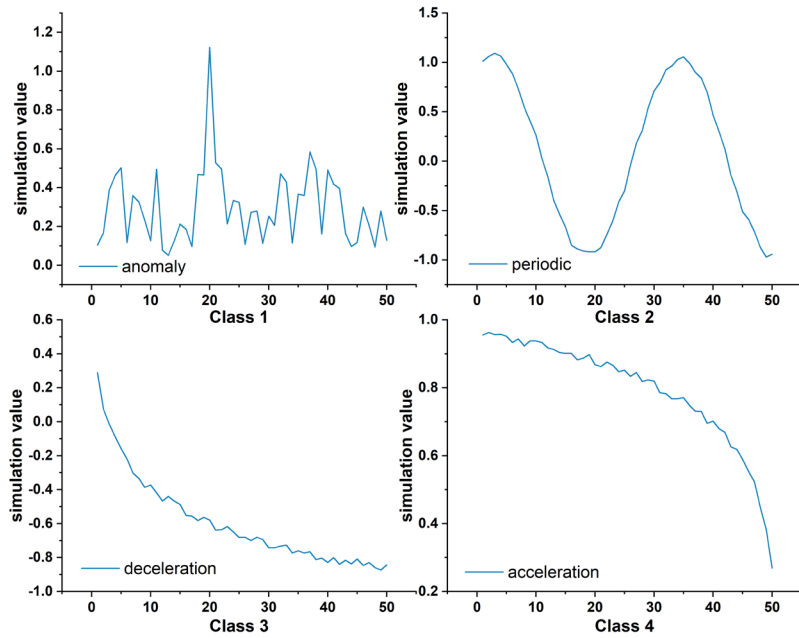
MT-InSAR processing



Spatial aware VAE-Clustering



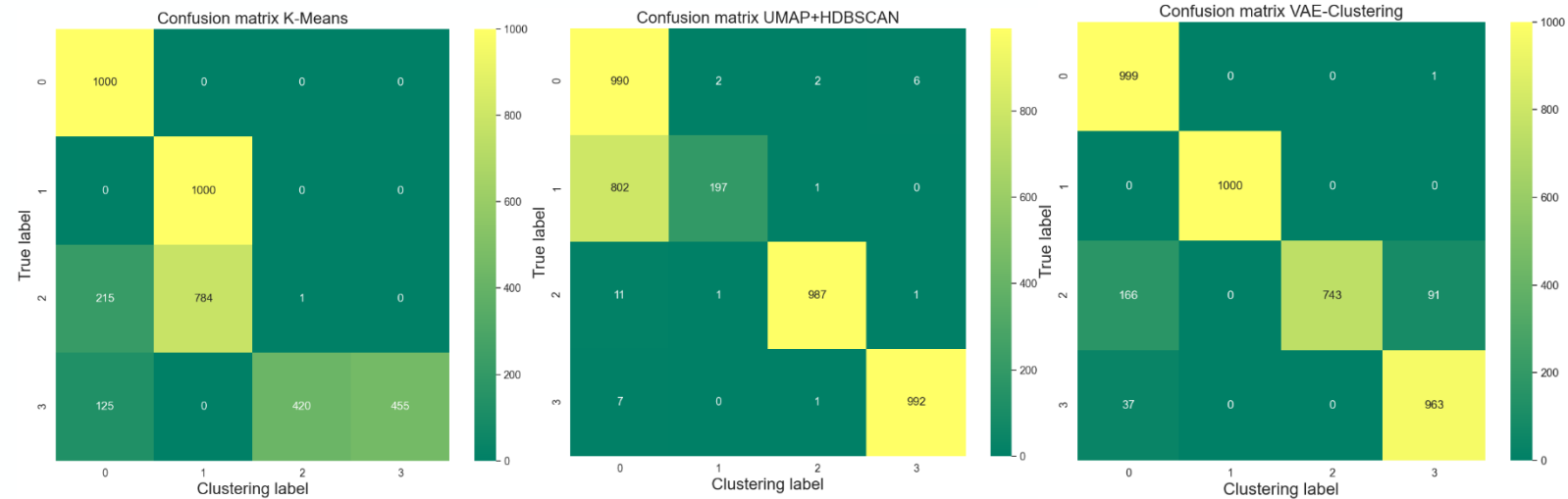
Test on Synthetic data



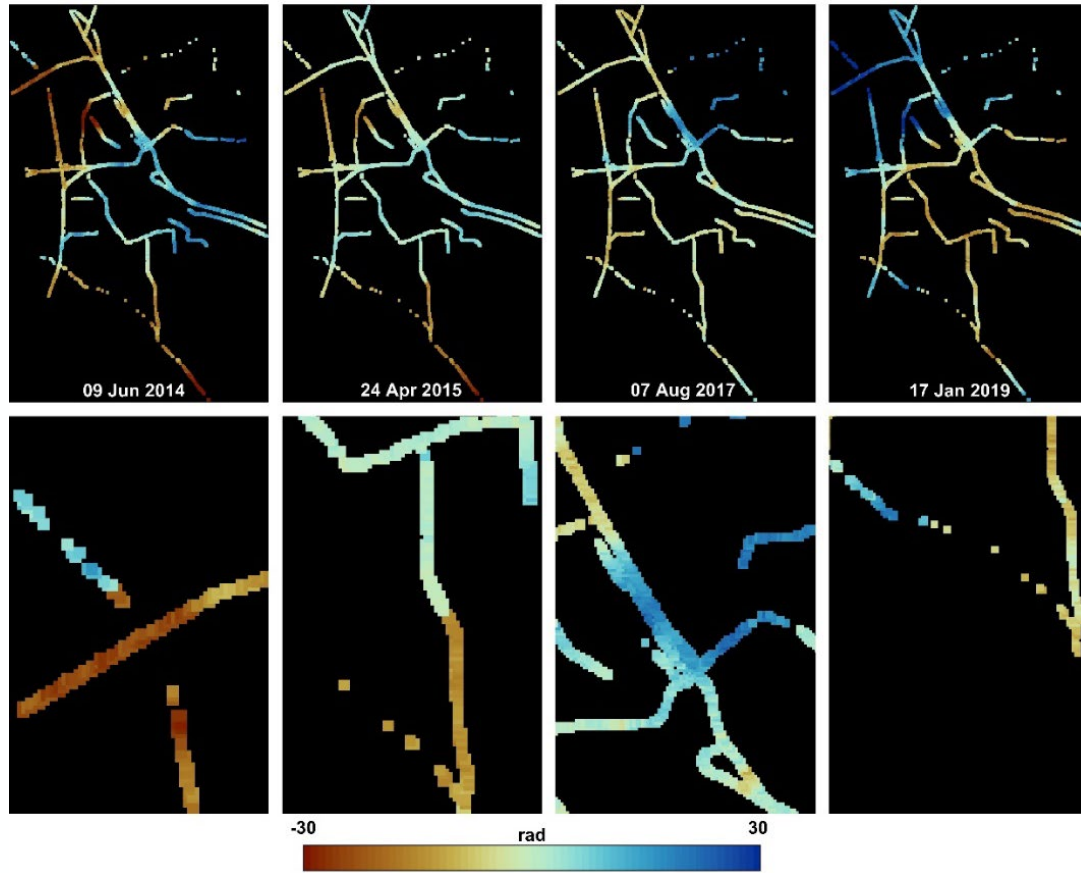
Simulating data

	Accuracy	ARI	NMI
K-means	0.614	0.514	0.642
UMAP+HDBSC AN	0.791	0.691	0.794
VAE-clustering	0.926	0.821	0.831

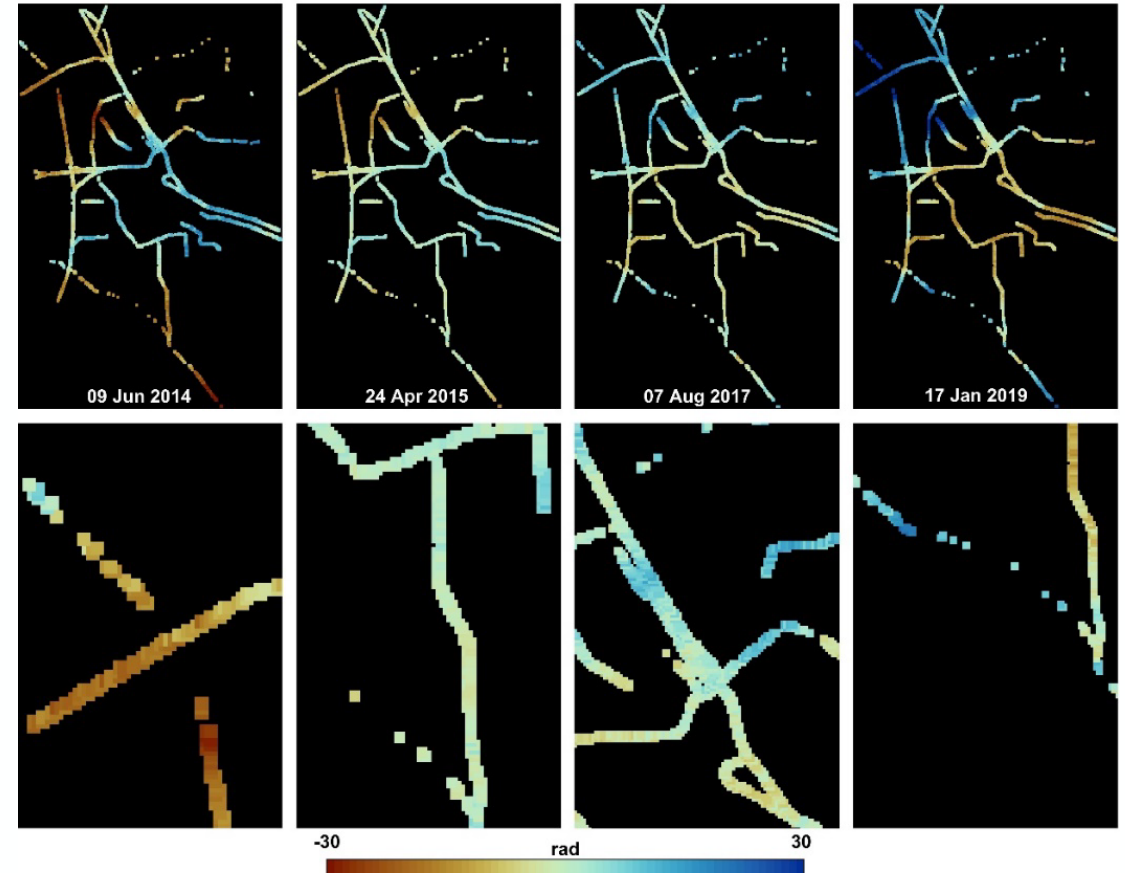
Performance of different methods



Case study #1 – Tianjin Railway

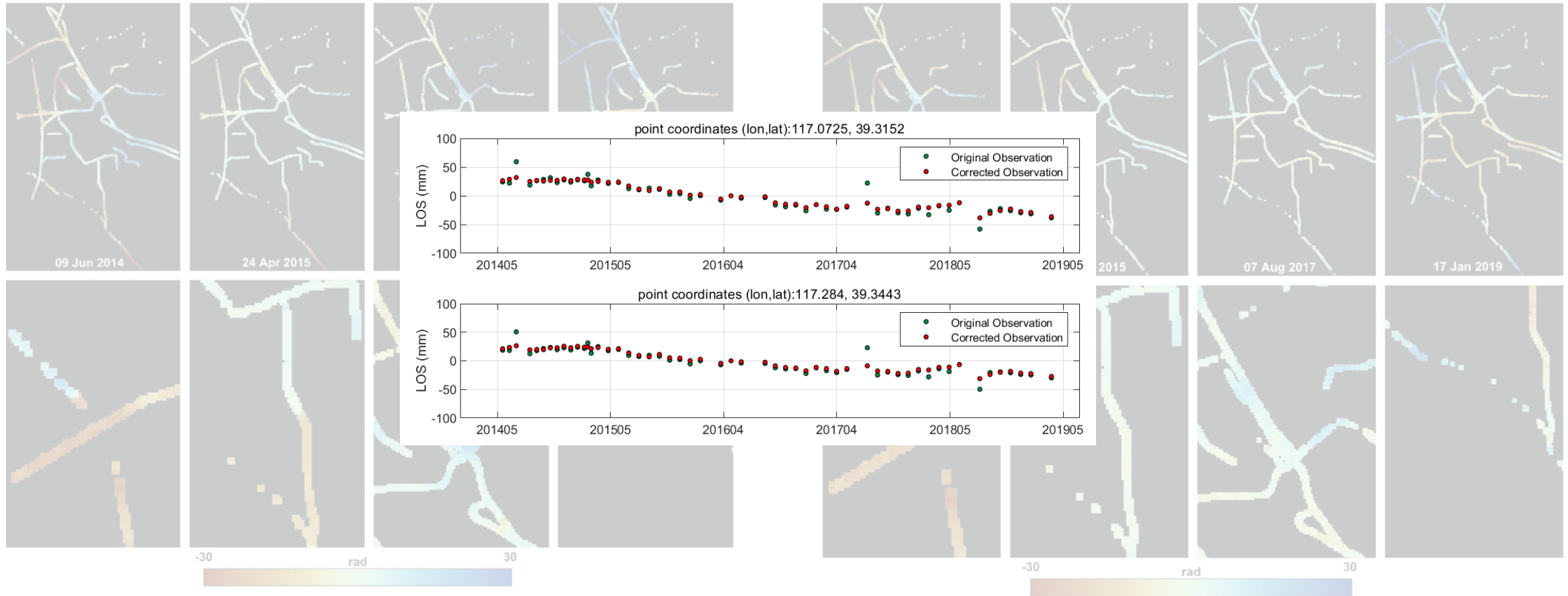


Phase unwrapping error



Corrected time series

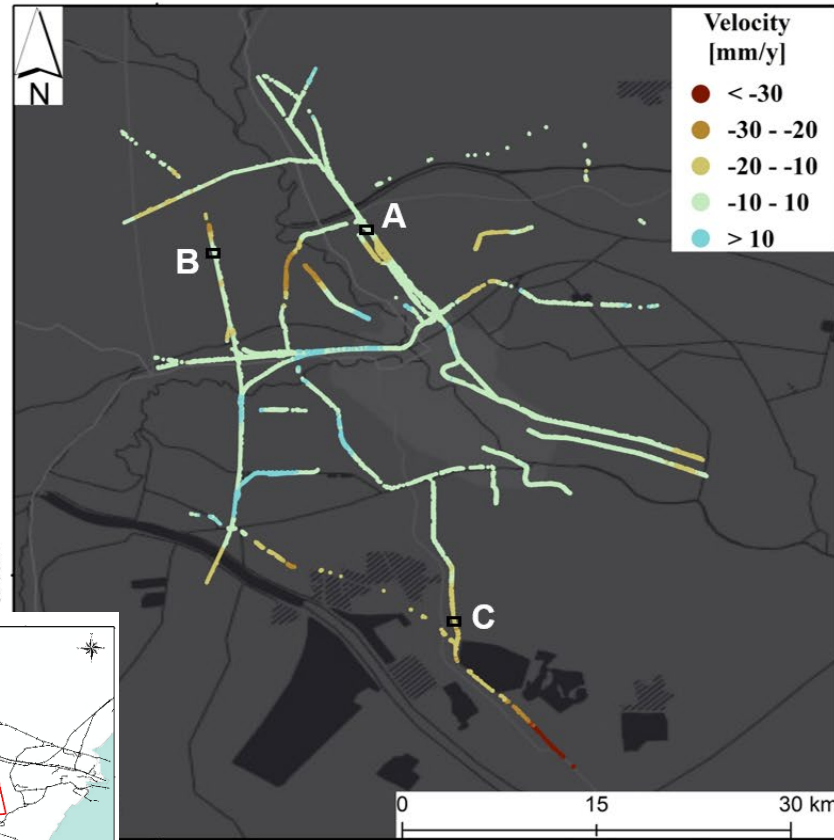
Case study #1 – Tianjin Railway



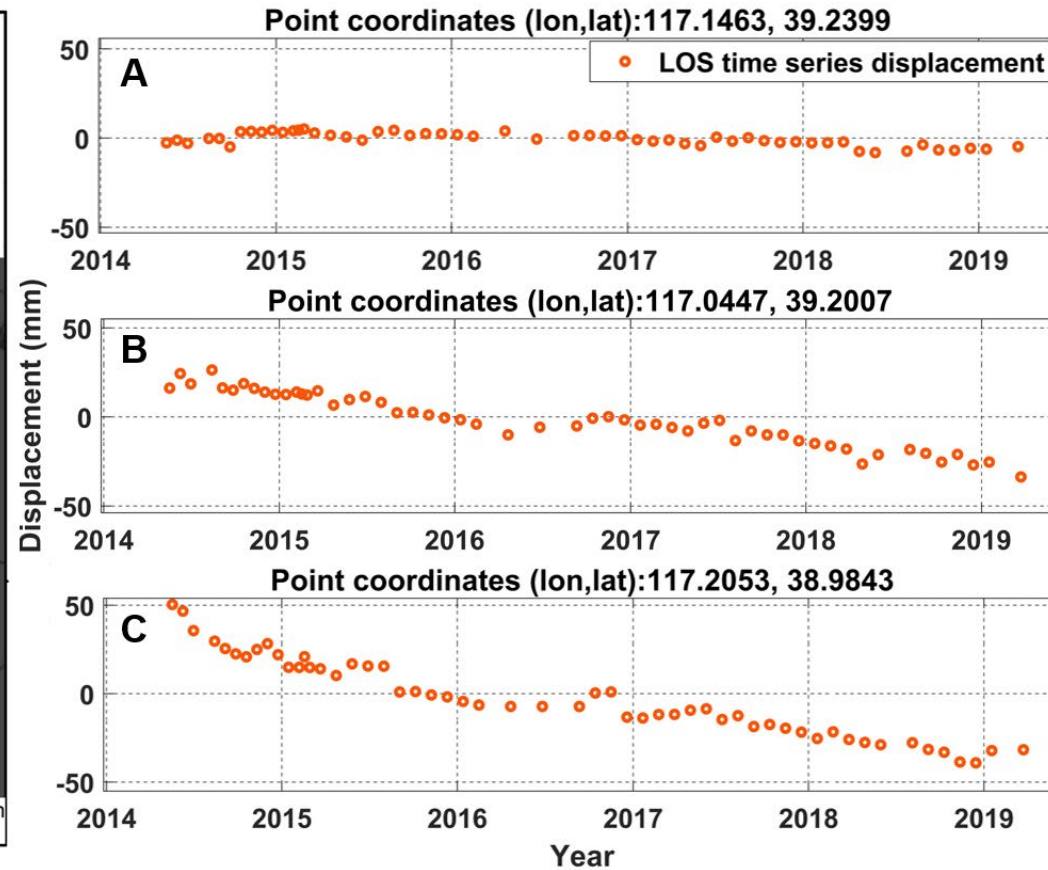
Phase unwrapping error

Corrected time series

Case study #1 – Tianjin Railway

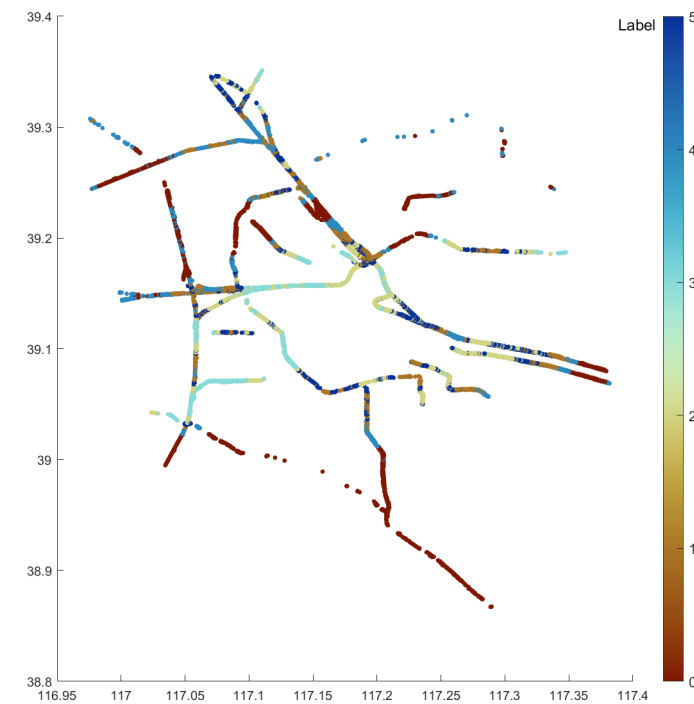
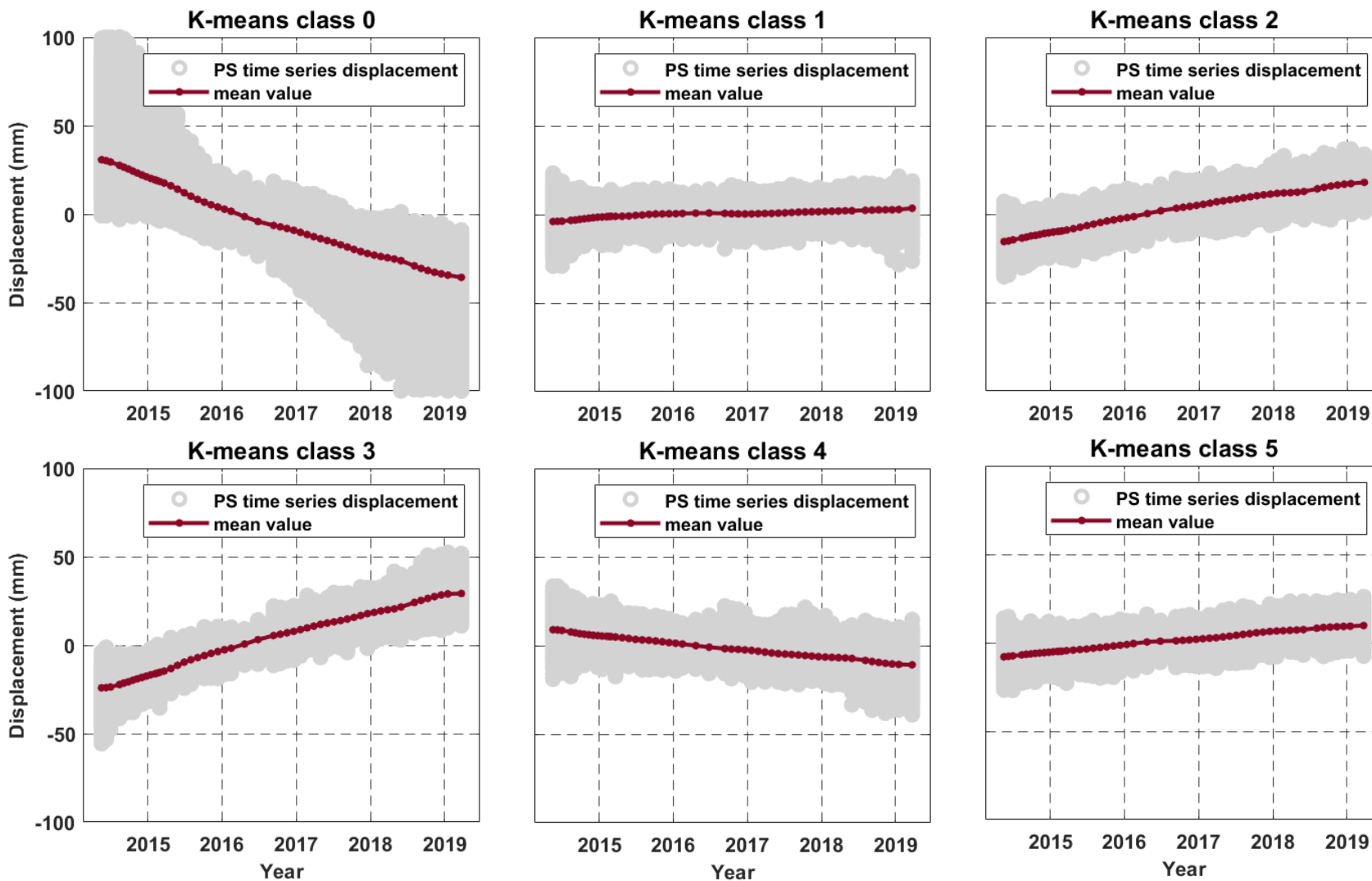


Deformation velocity map

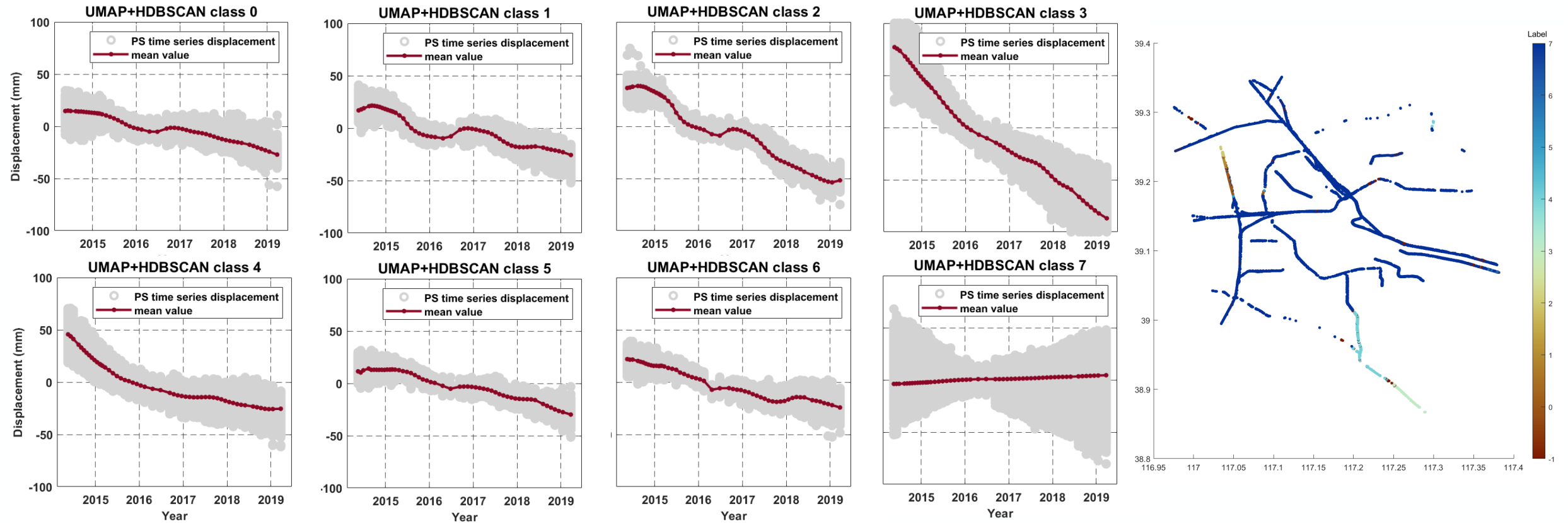


Deformation time series

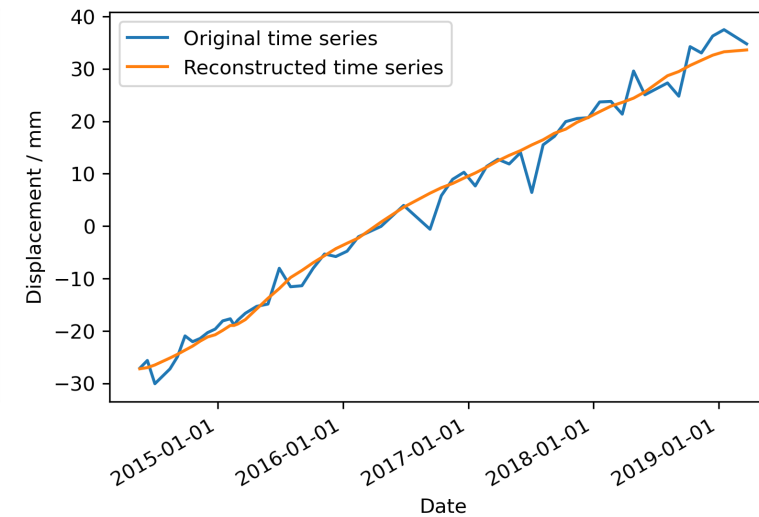
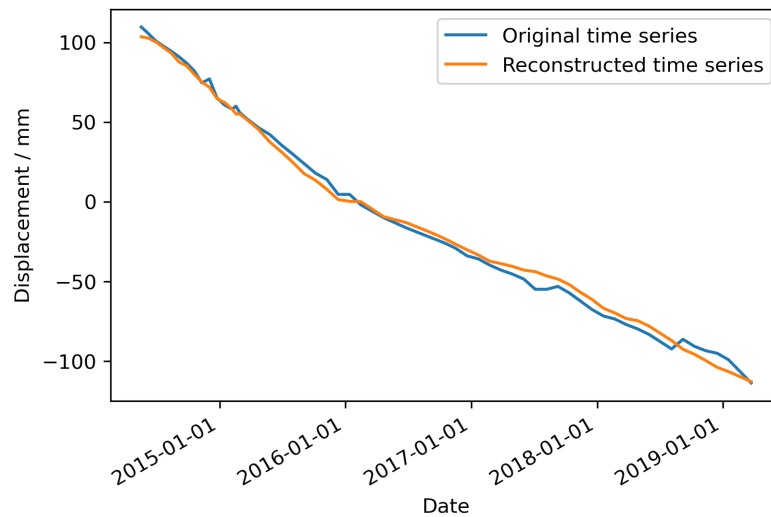
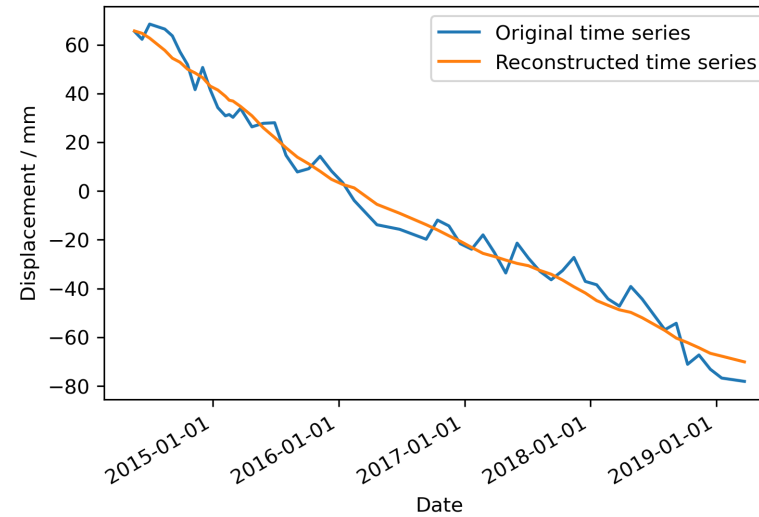
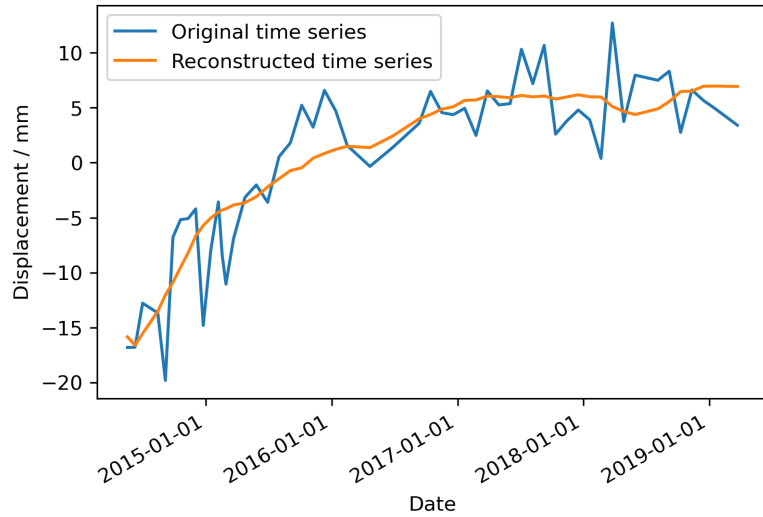
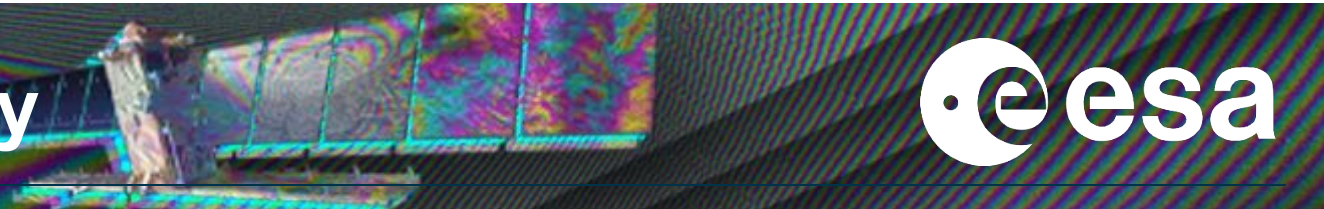
Case study #1 – Tianjin Railway



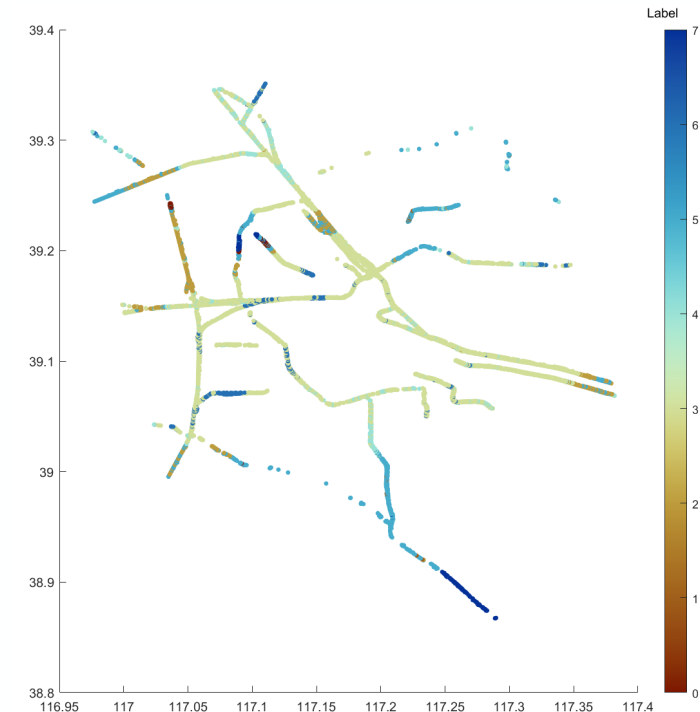
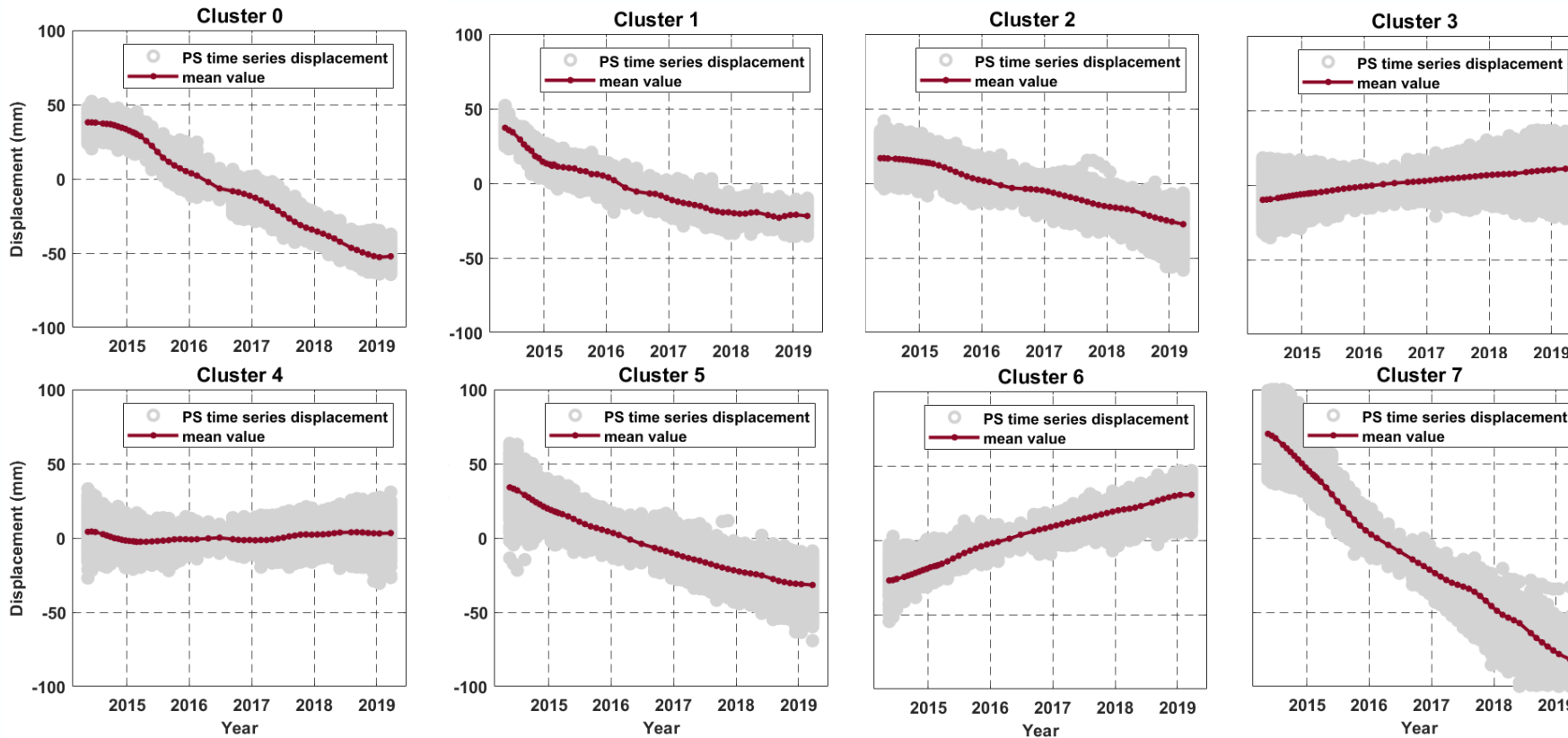
Case study #1 – Tianjin Railway



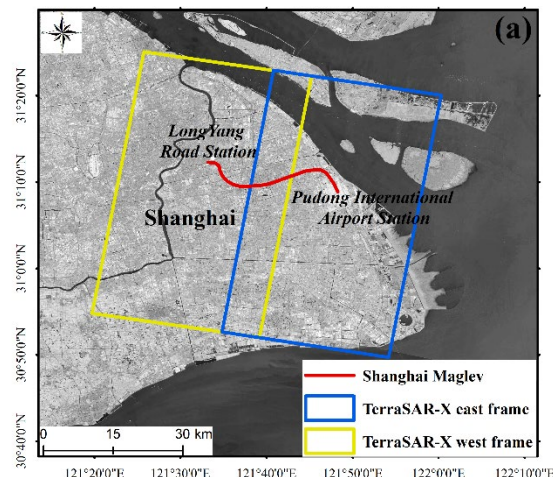
Case study #1 – Tianjin Railway



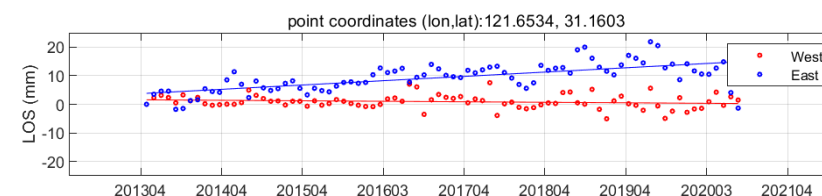
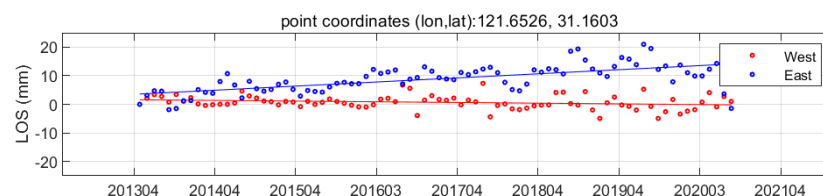
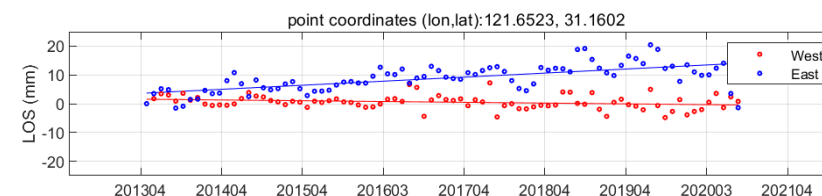
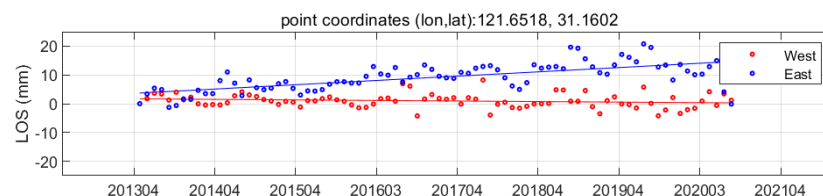
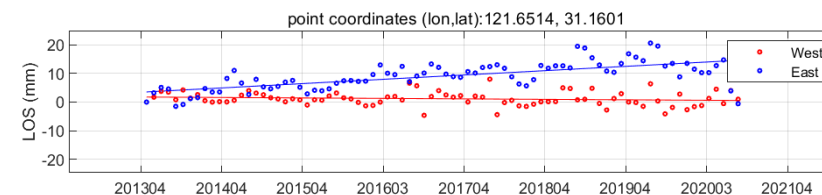
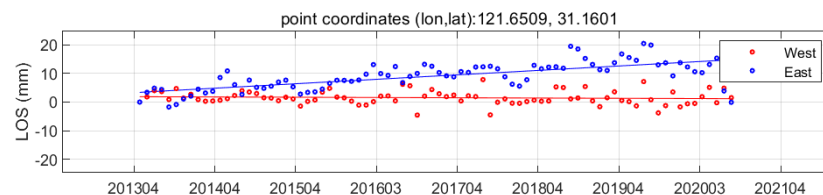
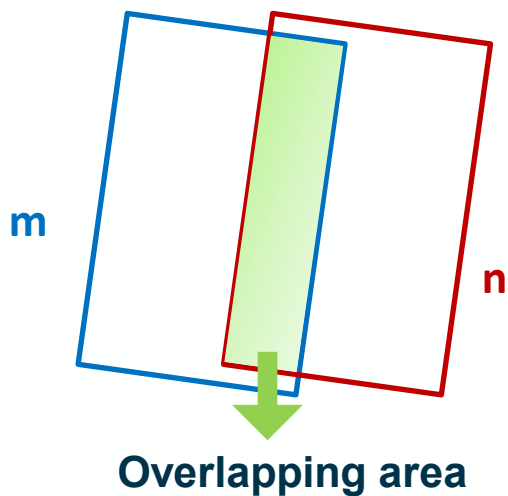
Case study #1 – Tianjin Railway



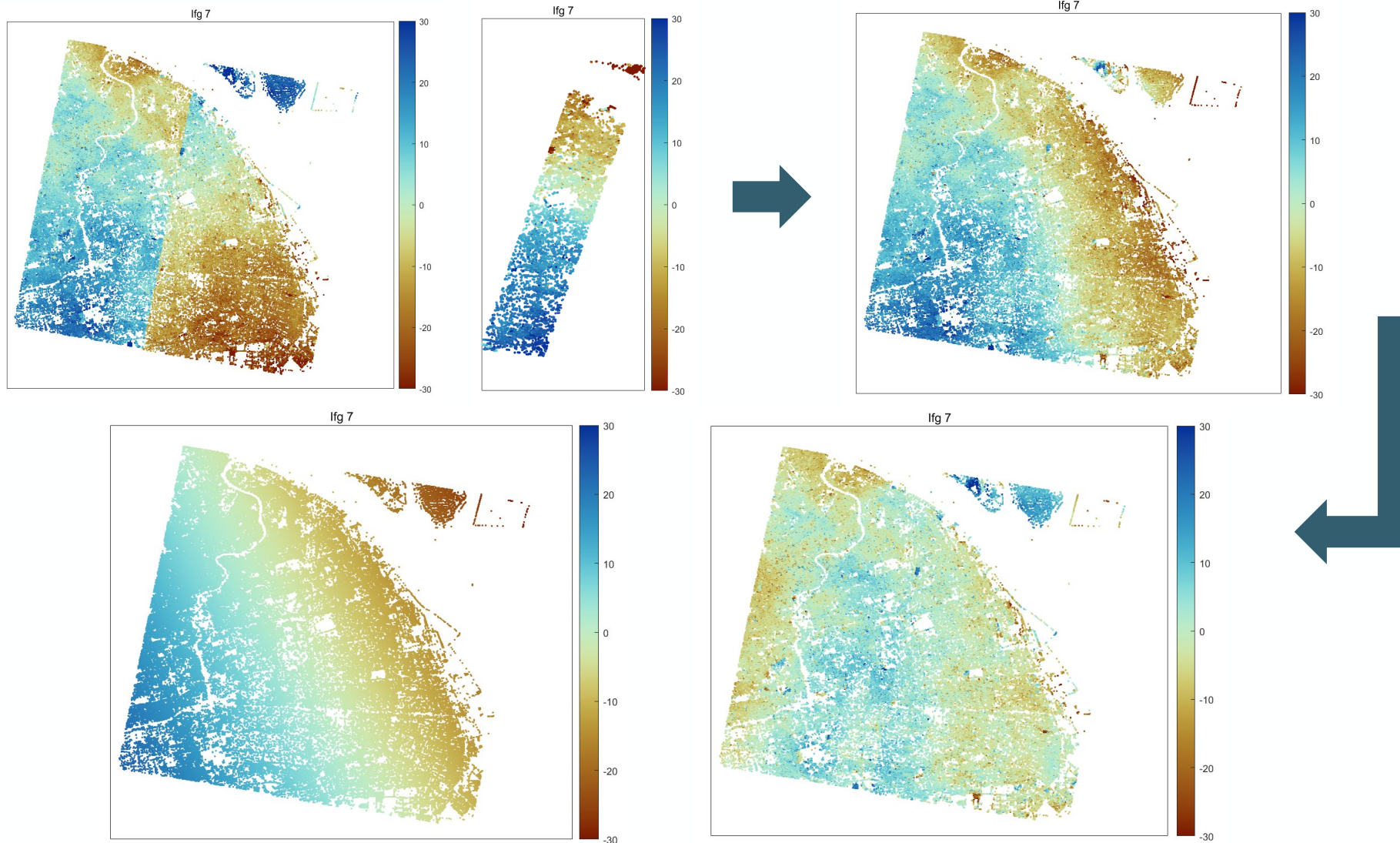
Case study #2 – Shanghai maglev



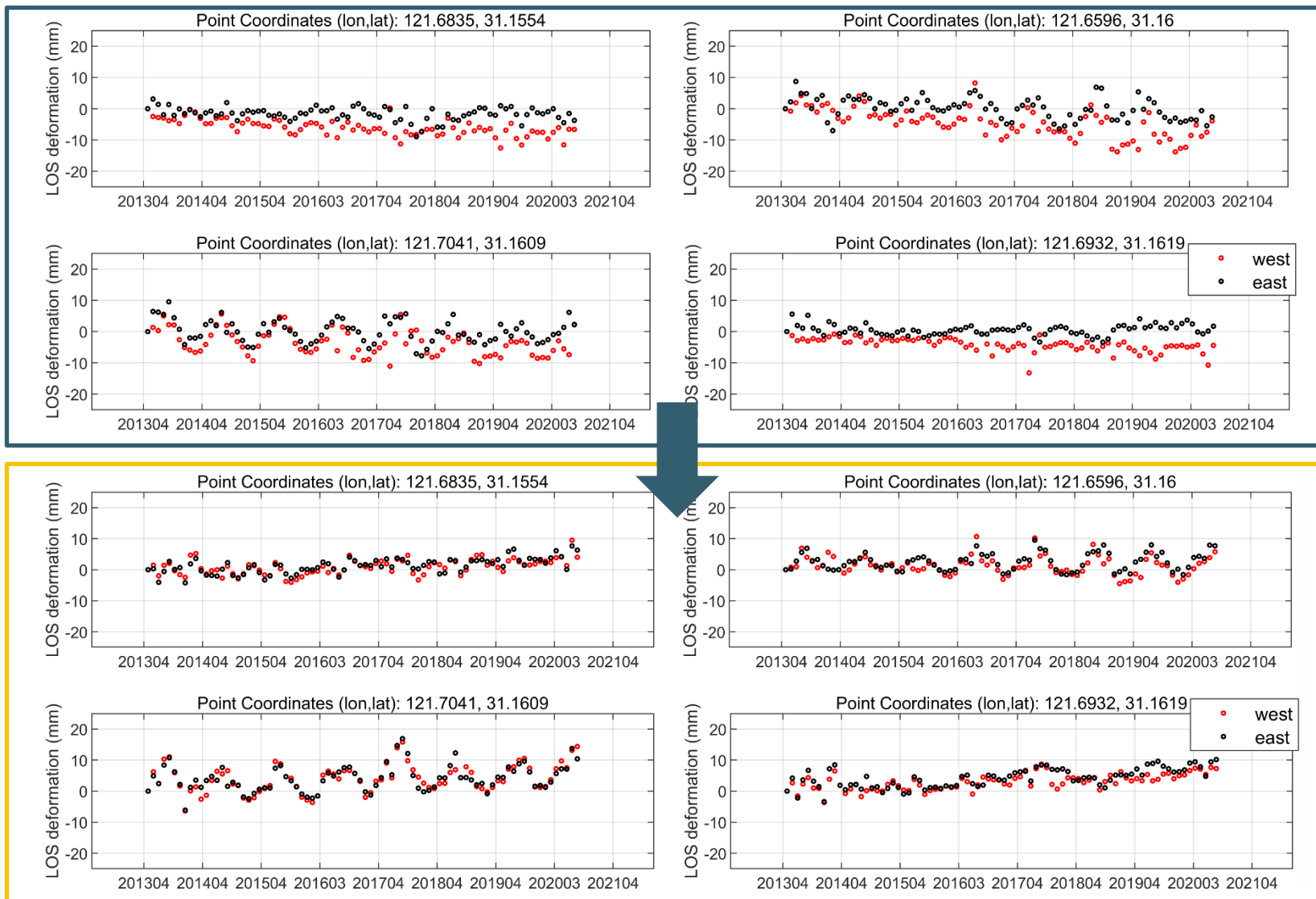
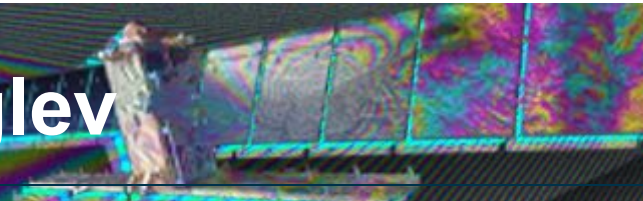
There is difference on the time series.



Case study #2 – Shanghai maglev

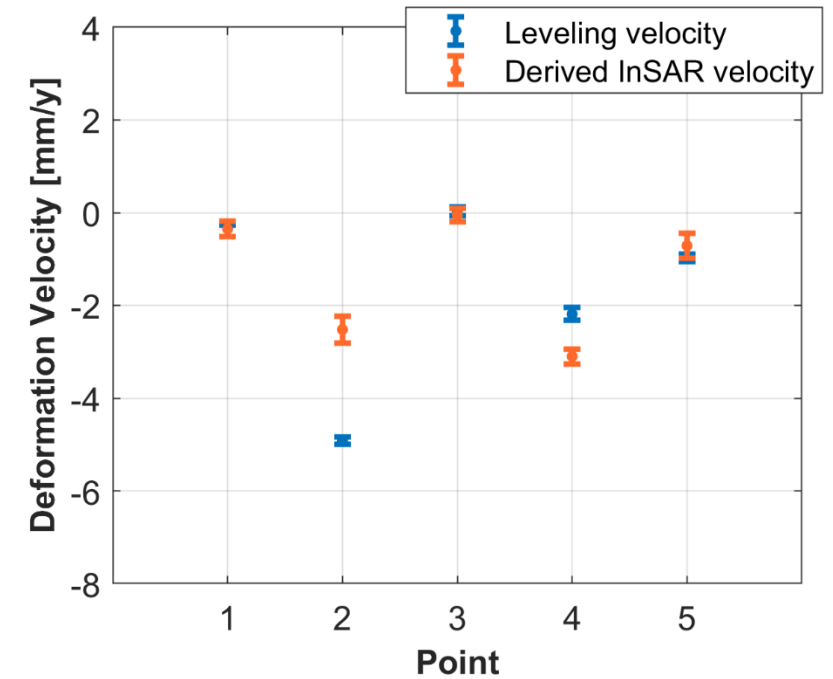
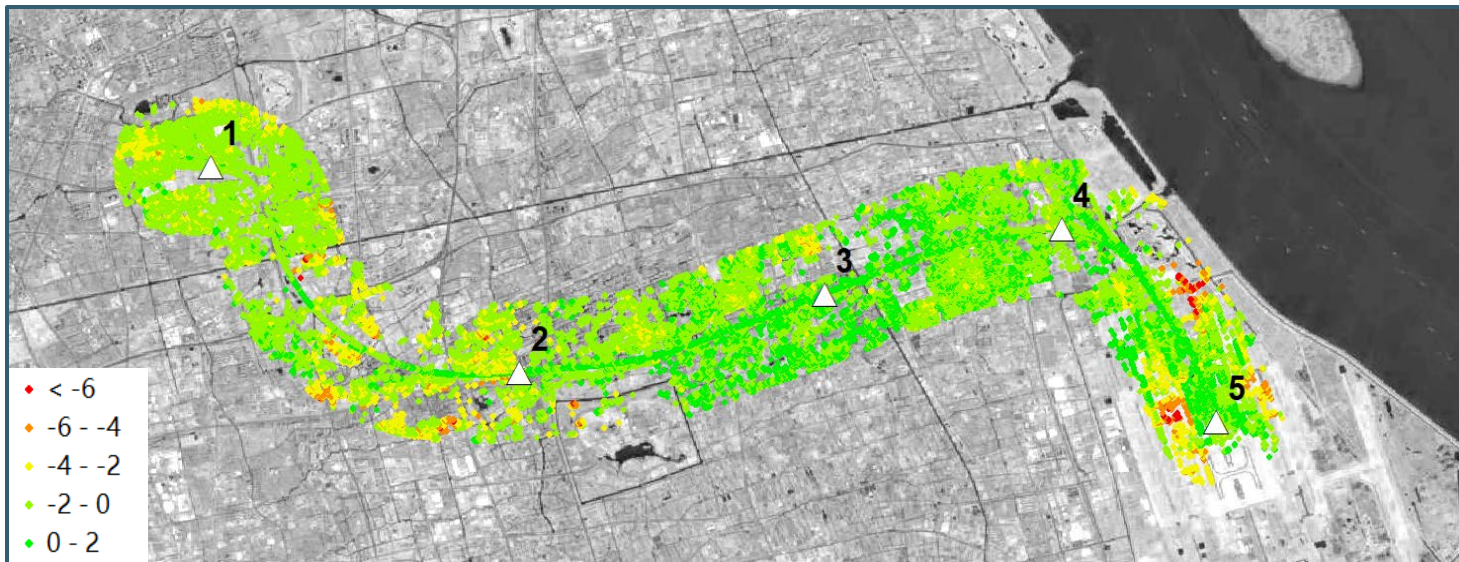


Case study #2 – Shanghai maglev

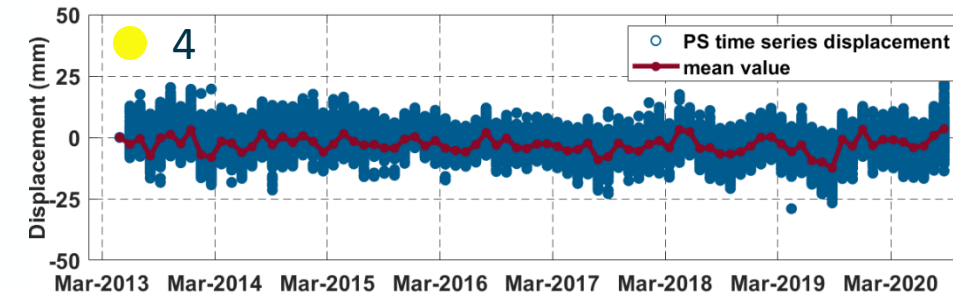
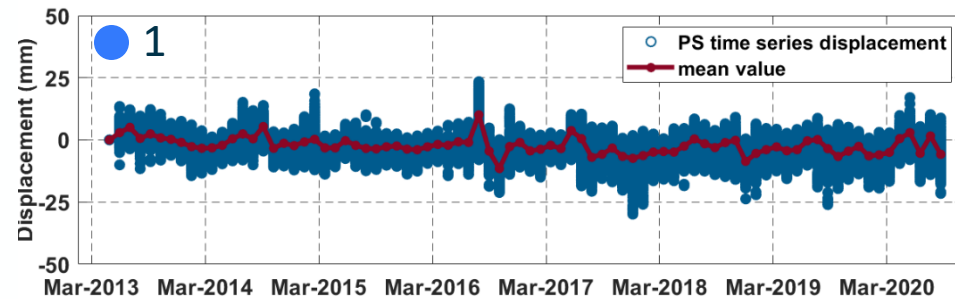
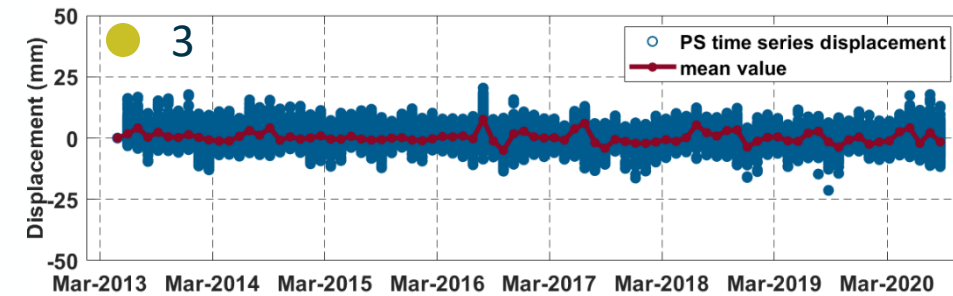
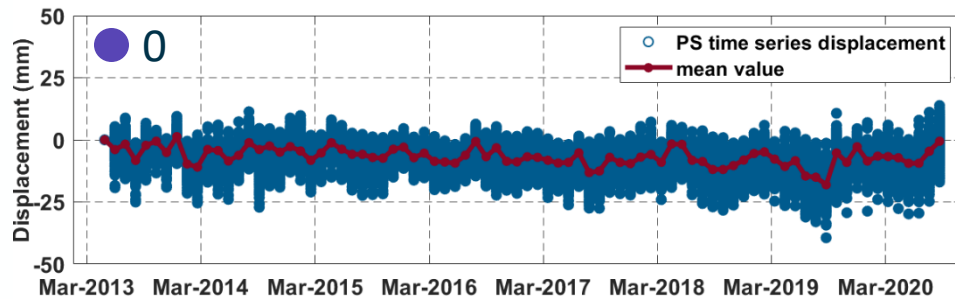
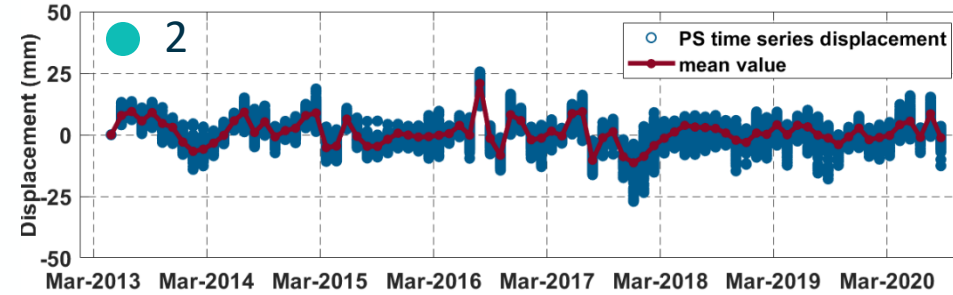
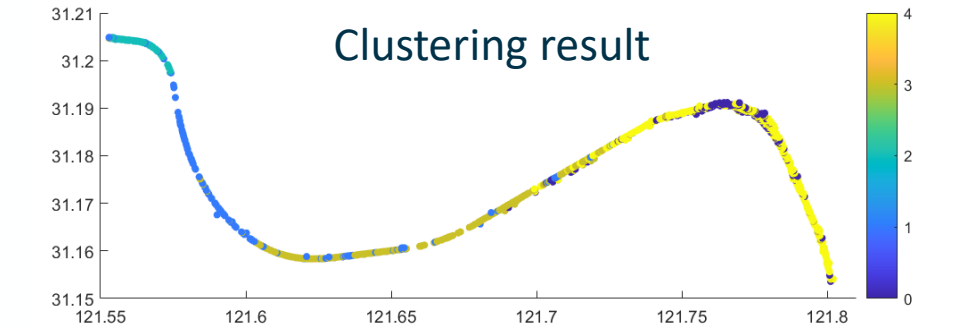
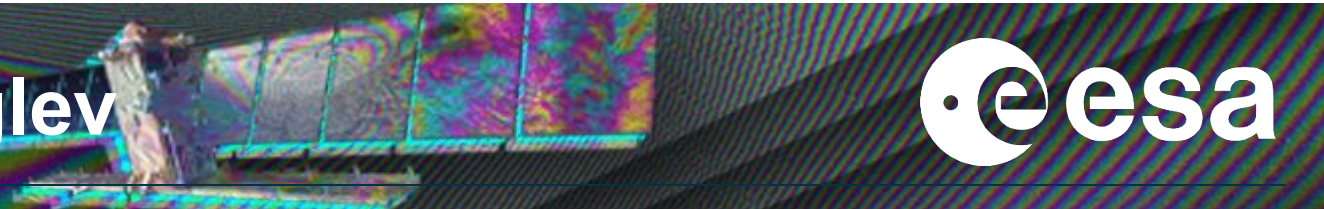


Case study #2 – Shanghai maglev

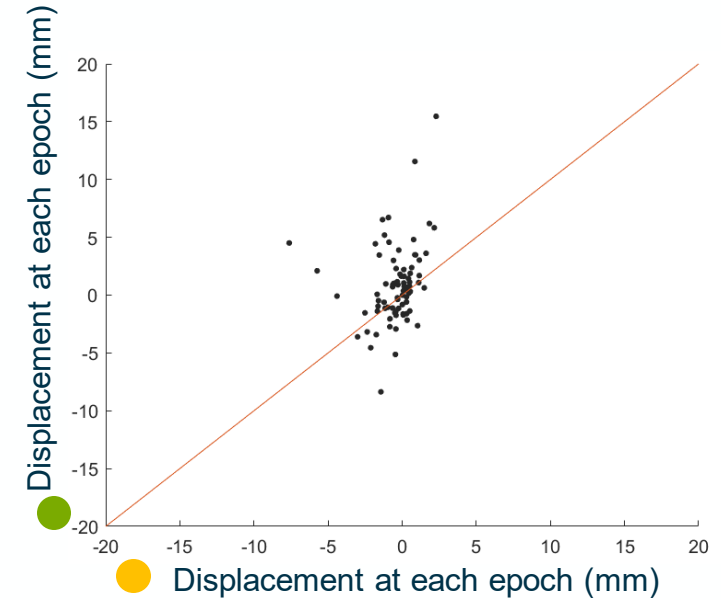
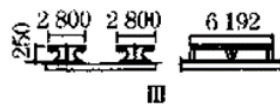
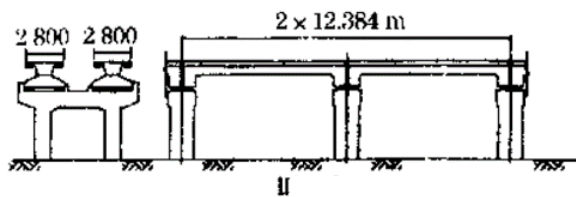
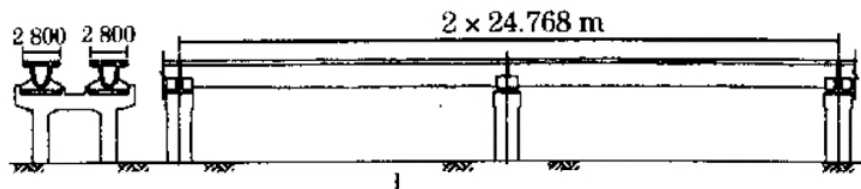
Deformation Velocity [mm/y]



Case study #2 – Shanghai maglev



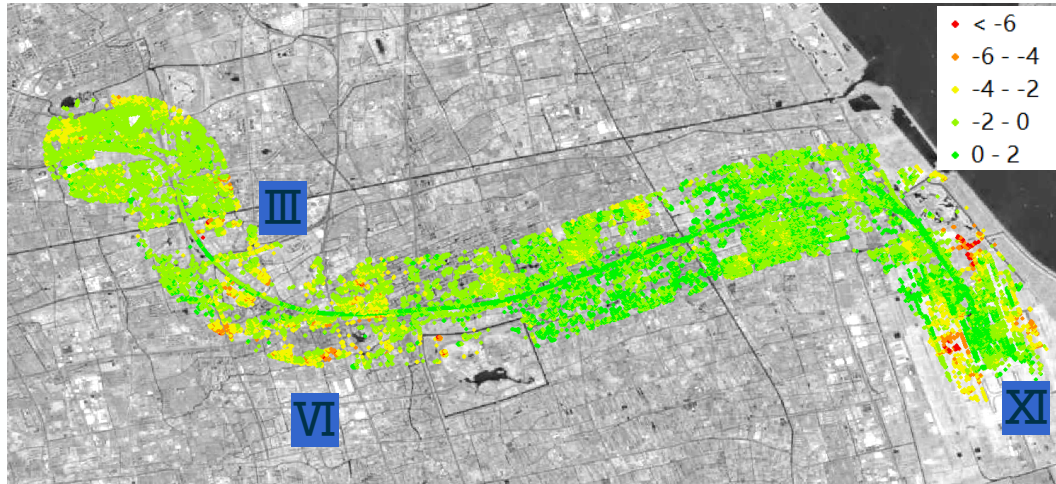
Case study #2 – Shanghai maglev



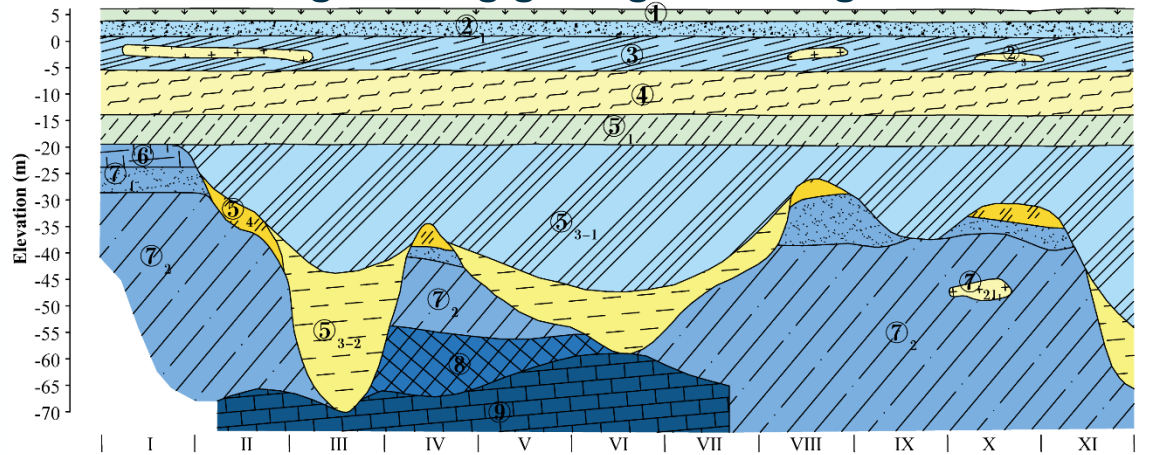
Beam end pier: $\Delta y = L / 6000$
Intermediate piers of multi-span beams: $\Delta y = L / 4500$

Case study #2 – Shanghai maglev

Deformation velocity [mm/y]



Engineering geological background

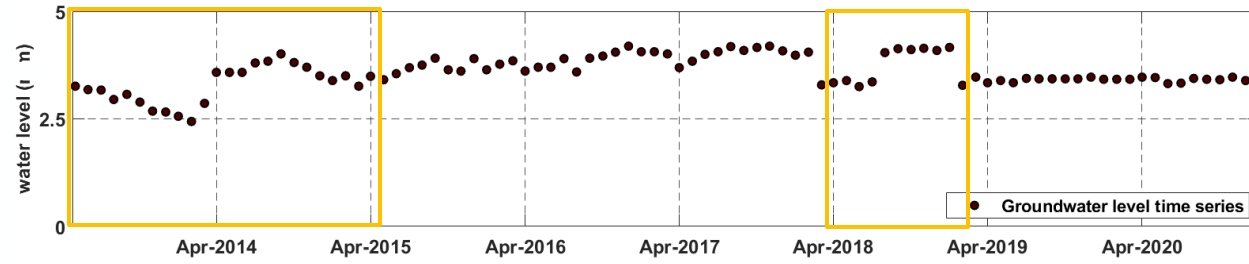


- ① filling
- ②₁ Brown-yellow clay
- ③ Grey mucky silty clay
- ④ Grey mucky clay
- ⑤₃₋₂ Grey sandy silt with silty clay
- ⑤₁ Grey-green clay
- ⑥₂ Grey fine sand
- ⑦₂₁ Weak interlayer
- ②₂ Grey-yellow clay
- ⑤₁ Brown-grey cohesive soil
- ⑥ Dark green-grass yellow clay
- ⑧ Gray silty sticky sand
- ②₃ Grey silty soil, silt
- ⑤₃₋₁ Grey silty sticky silt sand
- ⑦₁ Grass yellow-grey silty soil, silt
- ⑨ Grey fine sand

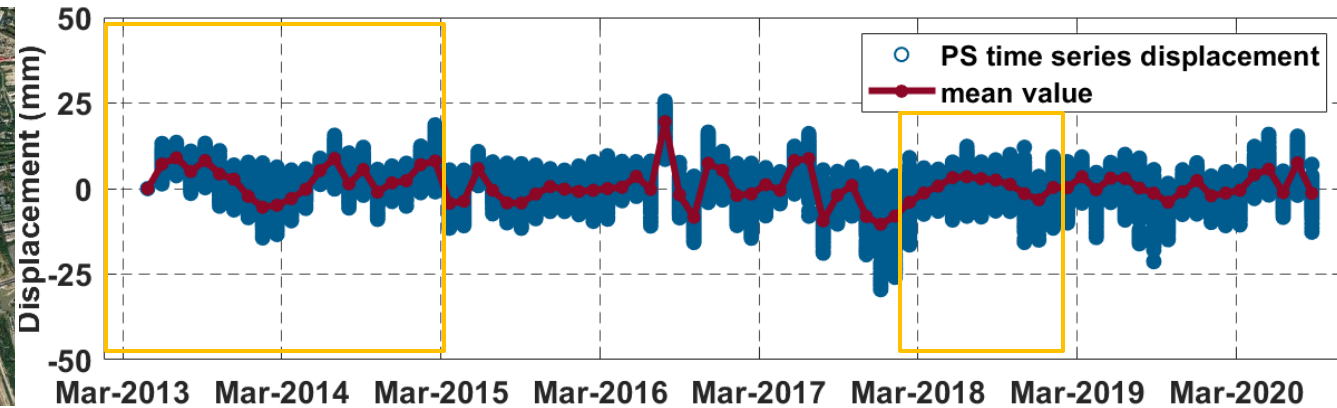
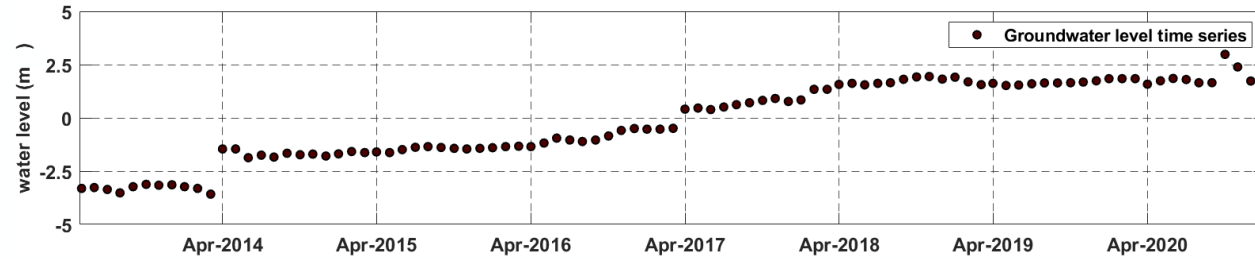
Section	Stake number
I	K0+000~K2+906
II	K2+906~K4+661
III	K4+661~K5+406
IV	K5+406~K7+247
V	K7+247~K8+544
VI	K8+544~K11+052
VII	K11+052~K13+529
VIII	K13+529~K16+997
IX	K16+997~K17+938
X	K17+938~K29+232
XI	K29+232~K29+842

Case study #2 – Shanghai maglev

FS9-00W (aquifer 0)

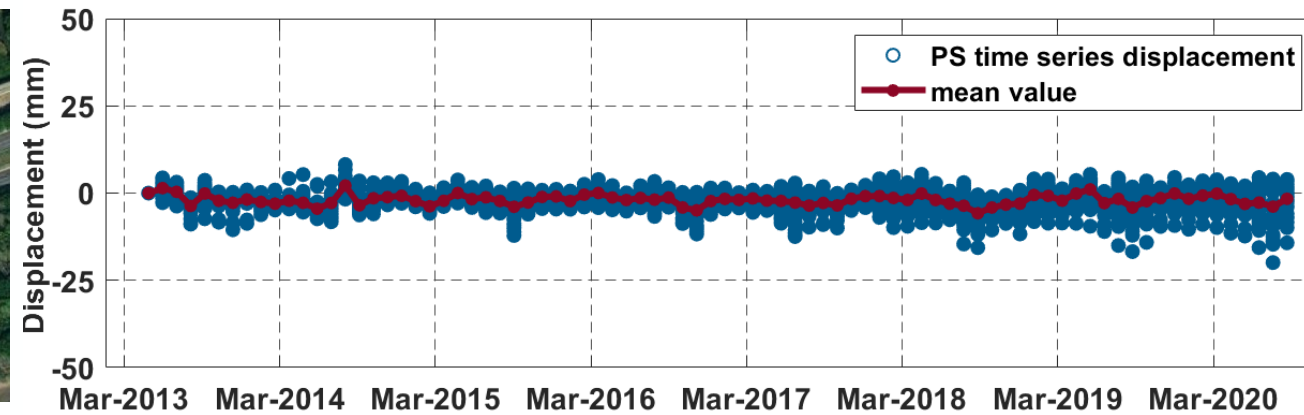
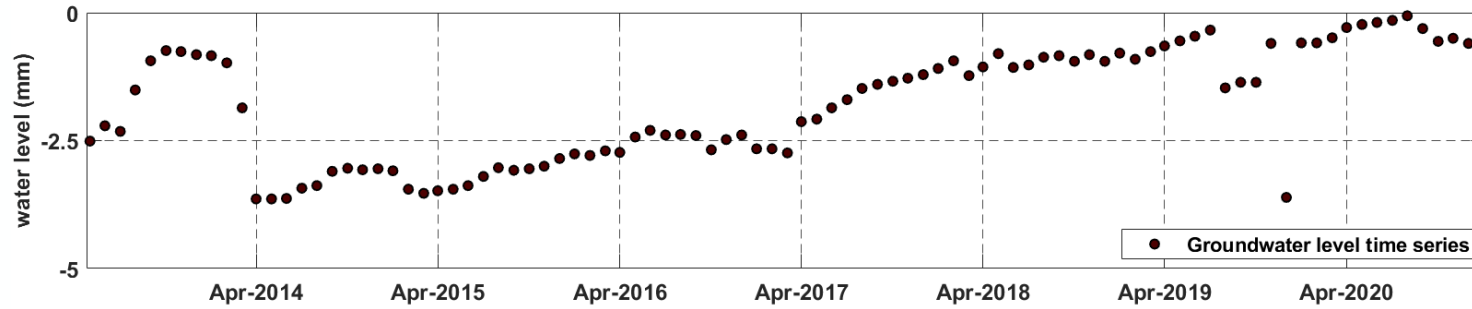


FS9-01W (aquifer 1)



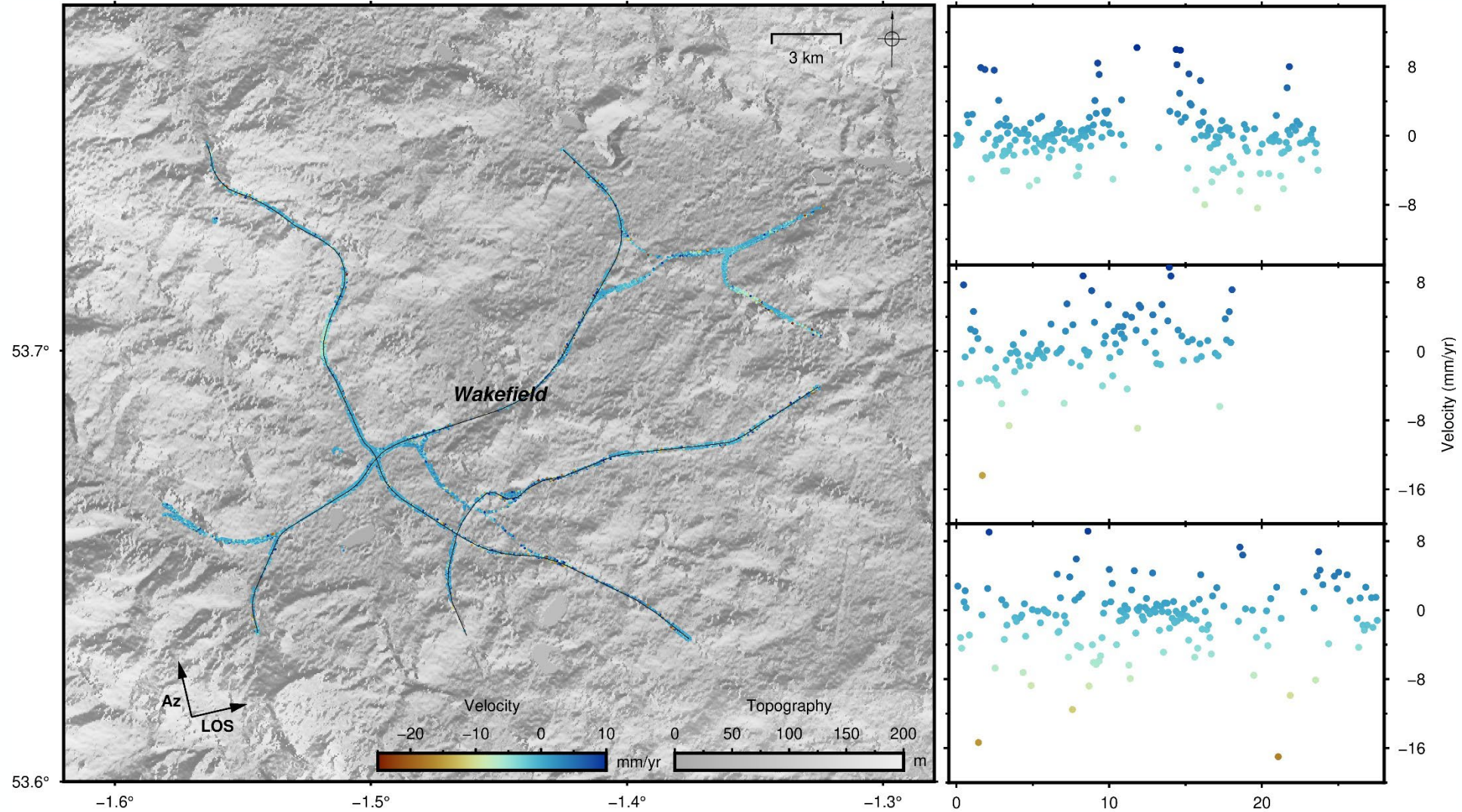
Case study #2 – Shanghai maglev

FS11-01W (aquifer 1)

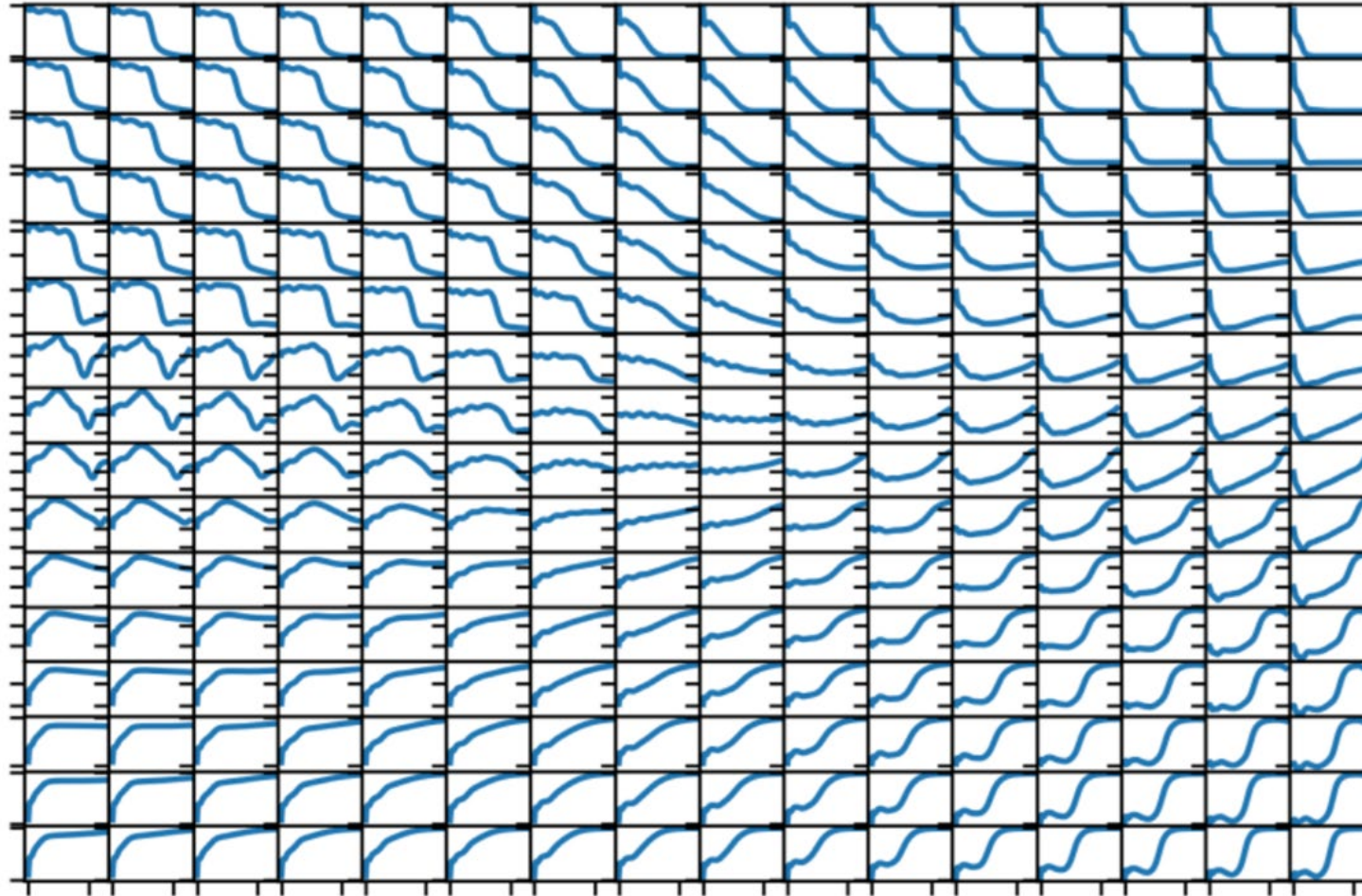


Groundwater level change of aquifer 0 has more correlations with surface displacement.

Case study #3 – Wakefield Railway



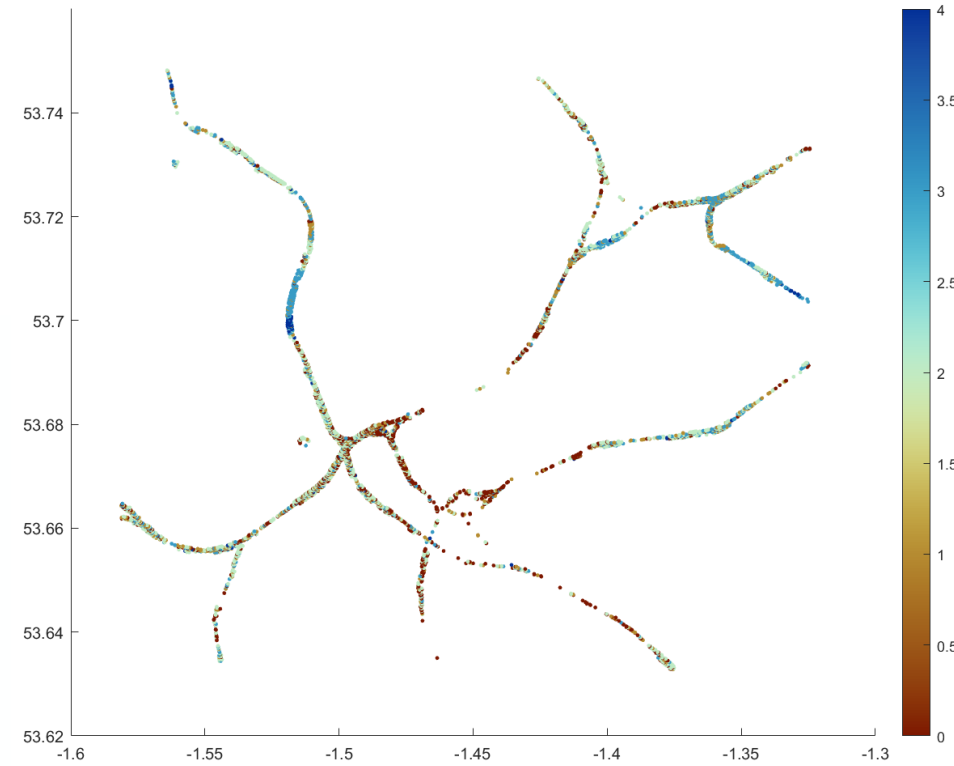
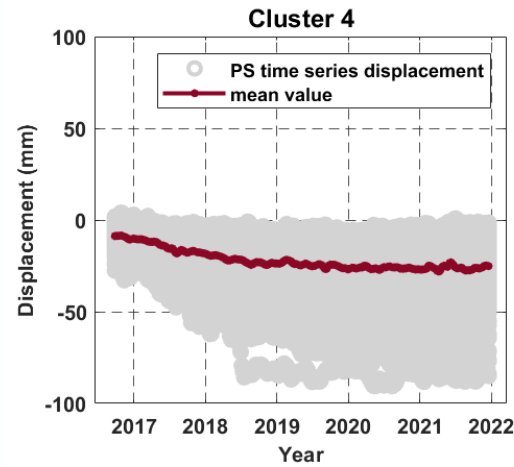
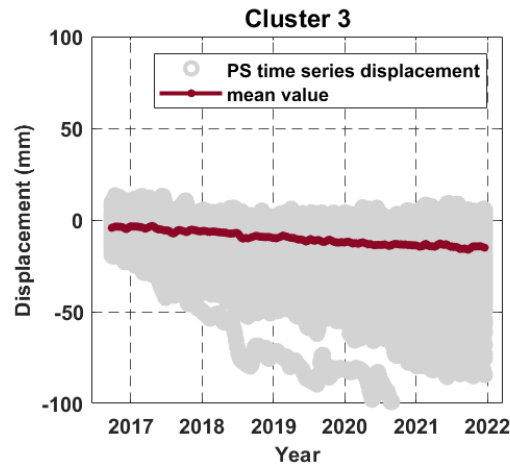
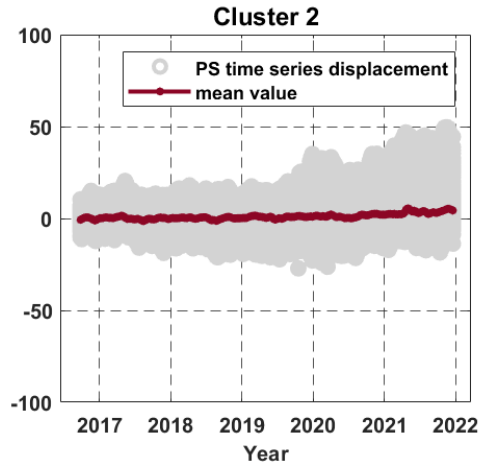
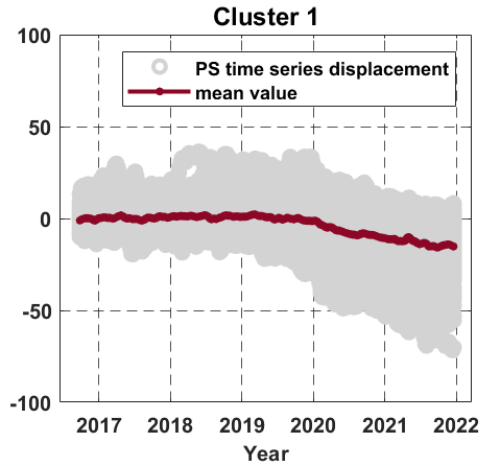
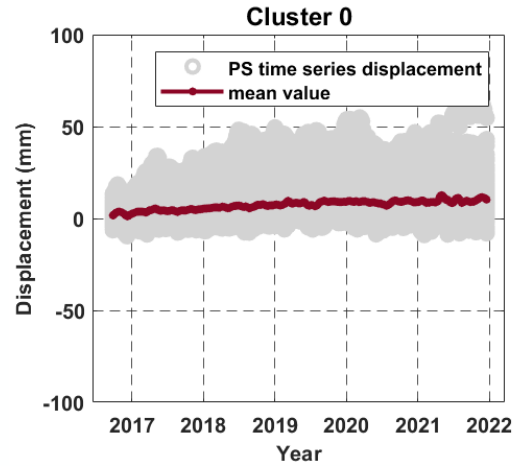
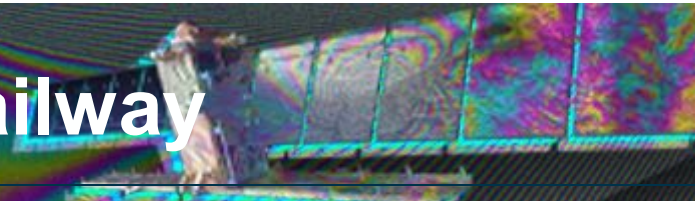
Case study #3 – Wakefield Railway



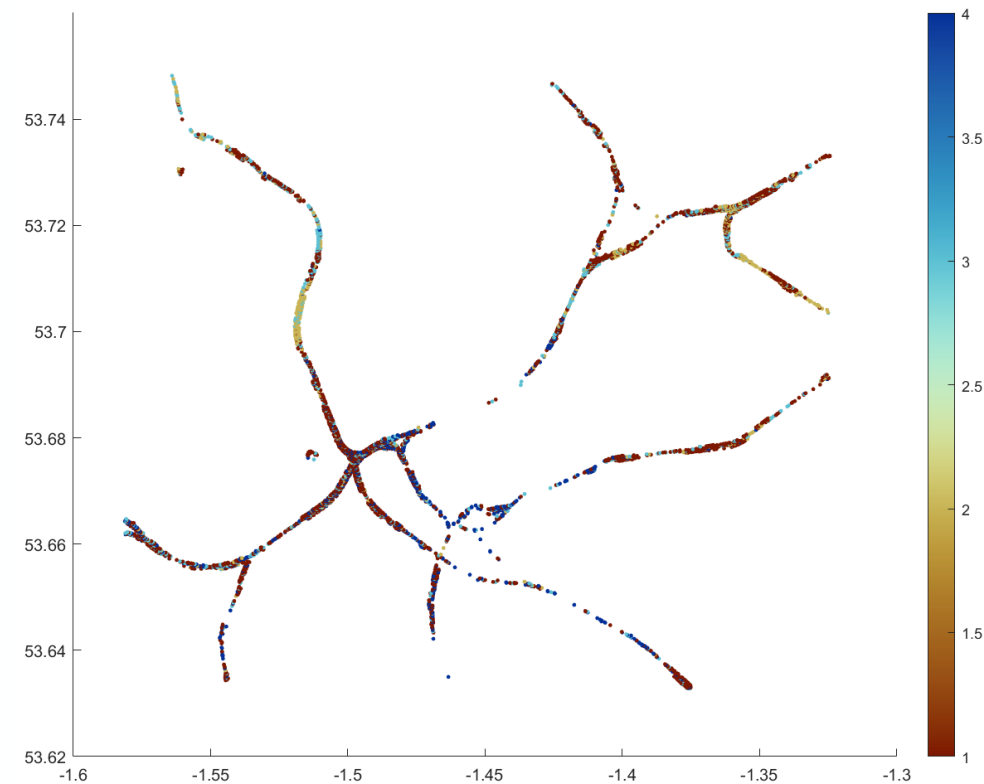
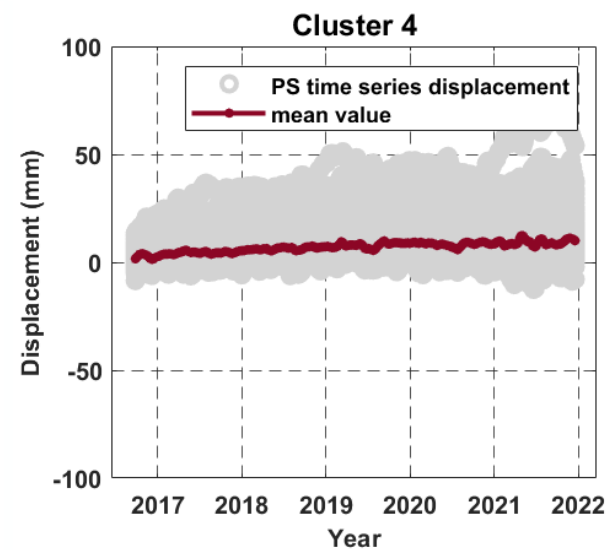
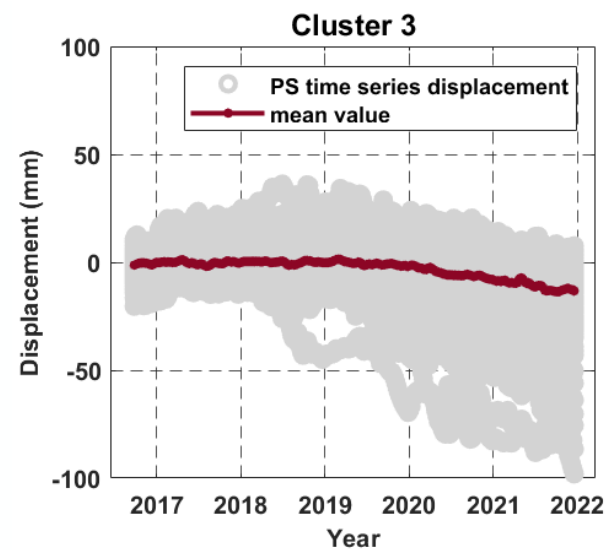
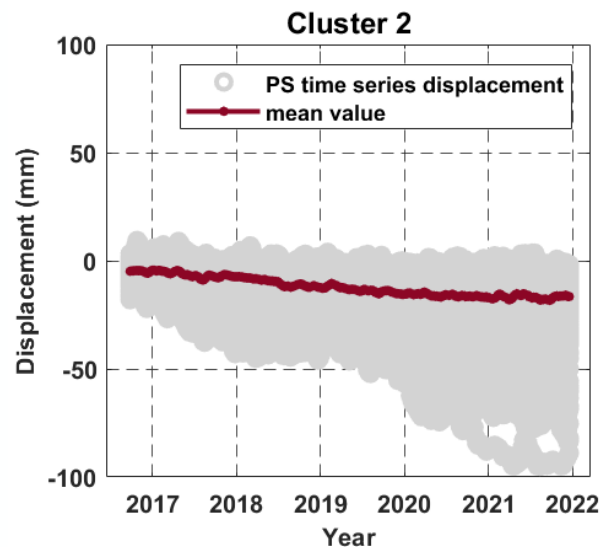
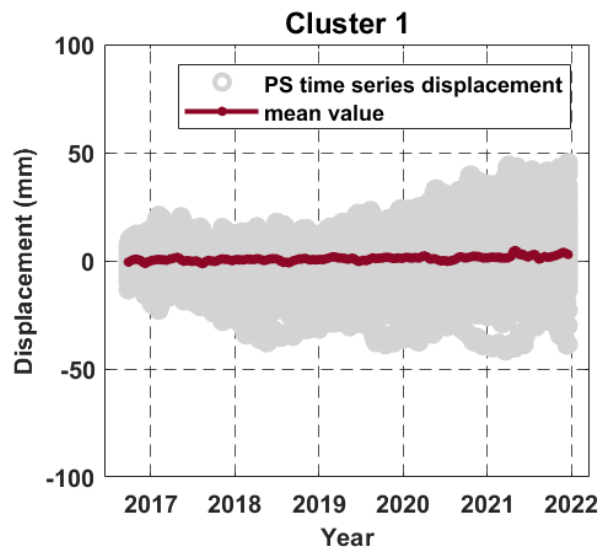
Time series in Latent space



Case study #3 – Wakefield Railway



Case study #3 – Wakefield Railway



1. **Ensure MT-InSAR processing to get accurate time series dataset is pre-requisite.**
2. **Spatial-aware deep clustering on temporal evolutions helps to interpret and analyze time series data.**
3. **Further analysis is needed to examine the various factors contributing to the deformation.**

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