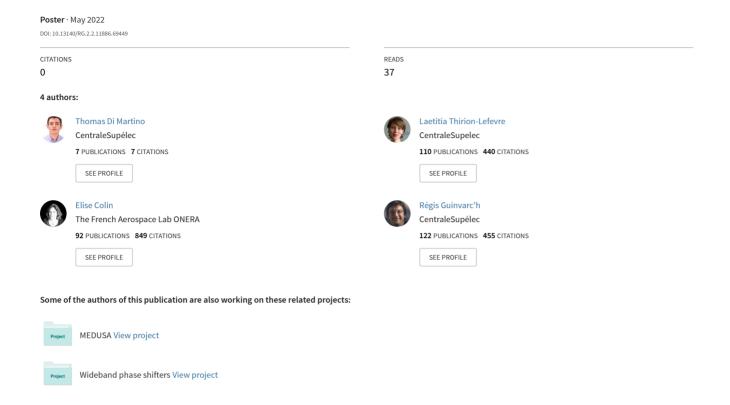
Convolutional Autoencoder for the unsupervised extraction of fire footprints from Sentinel-1 time-series





Convolutional Autoencoder for the unsupervised extraction of fire footprints from Sentinel-1 time-series Thomas Di Martino^{1,2}, Laetitia Thirion-Lefevre², Elise Colin¹, Régis Guinvarc'h²



CentraleSupélec

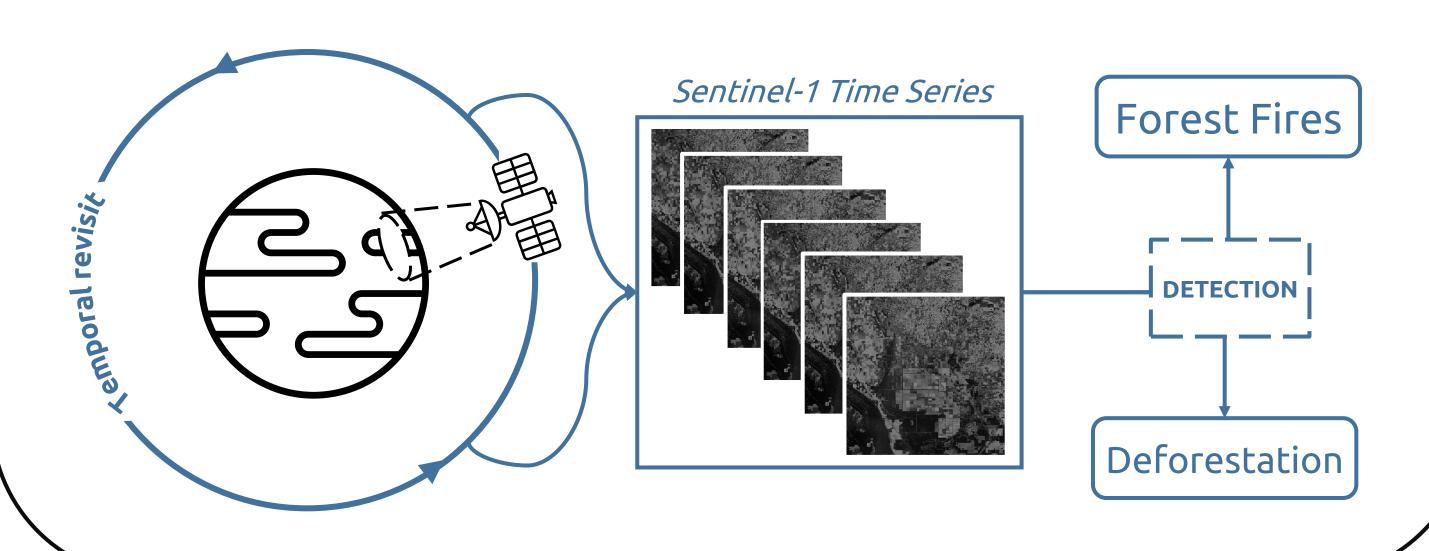
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> Session C1.09 Representation learning in remote sensing: from unsupervised, to self-and meta-learning

INTRODUCTION

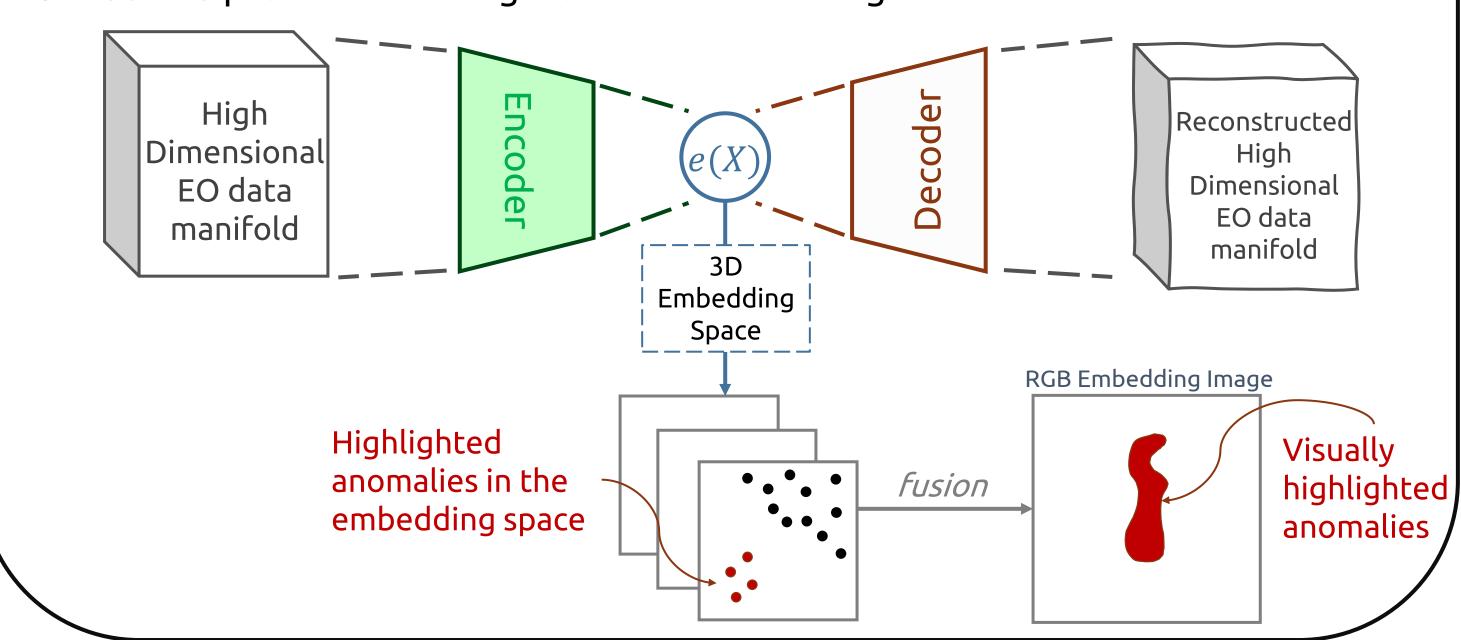
Detection of forest perturbations is possible with Sentinel-1 time series, through temporal processing:

- Detection of deforestation [1,2]
- Detection of wildfires [3]

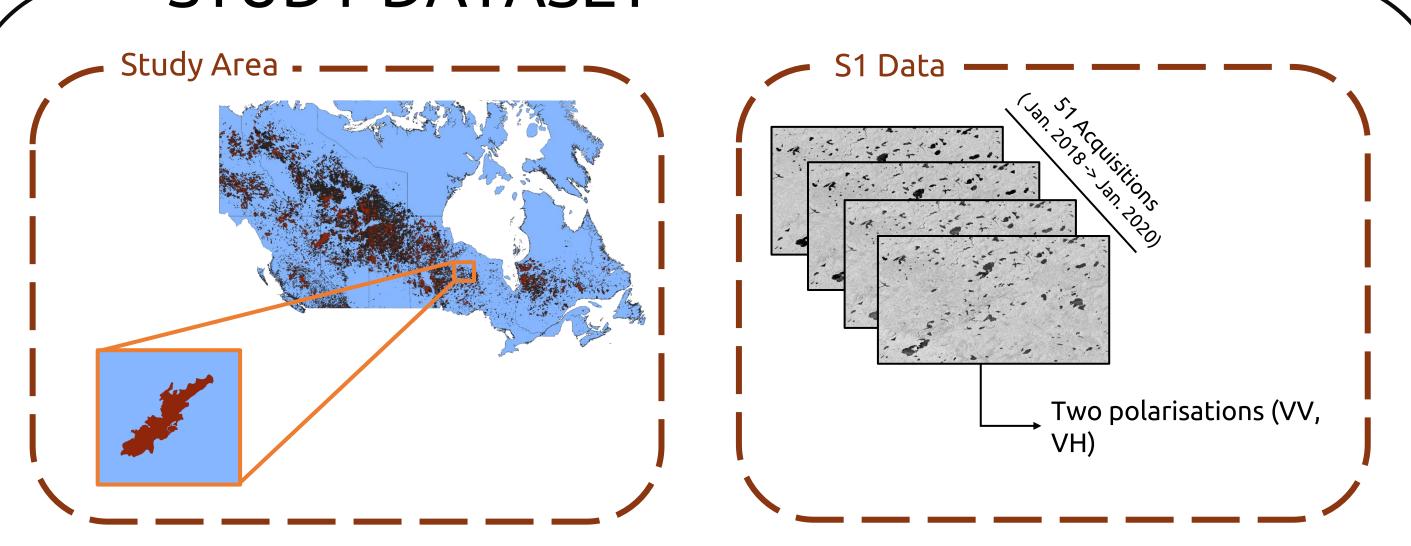


Unsupervised temporal anomaly detection method:

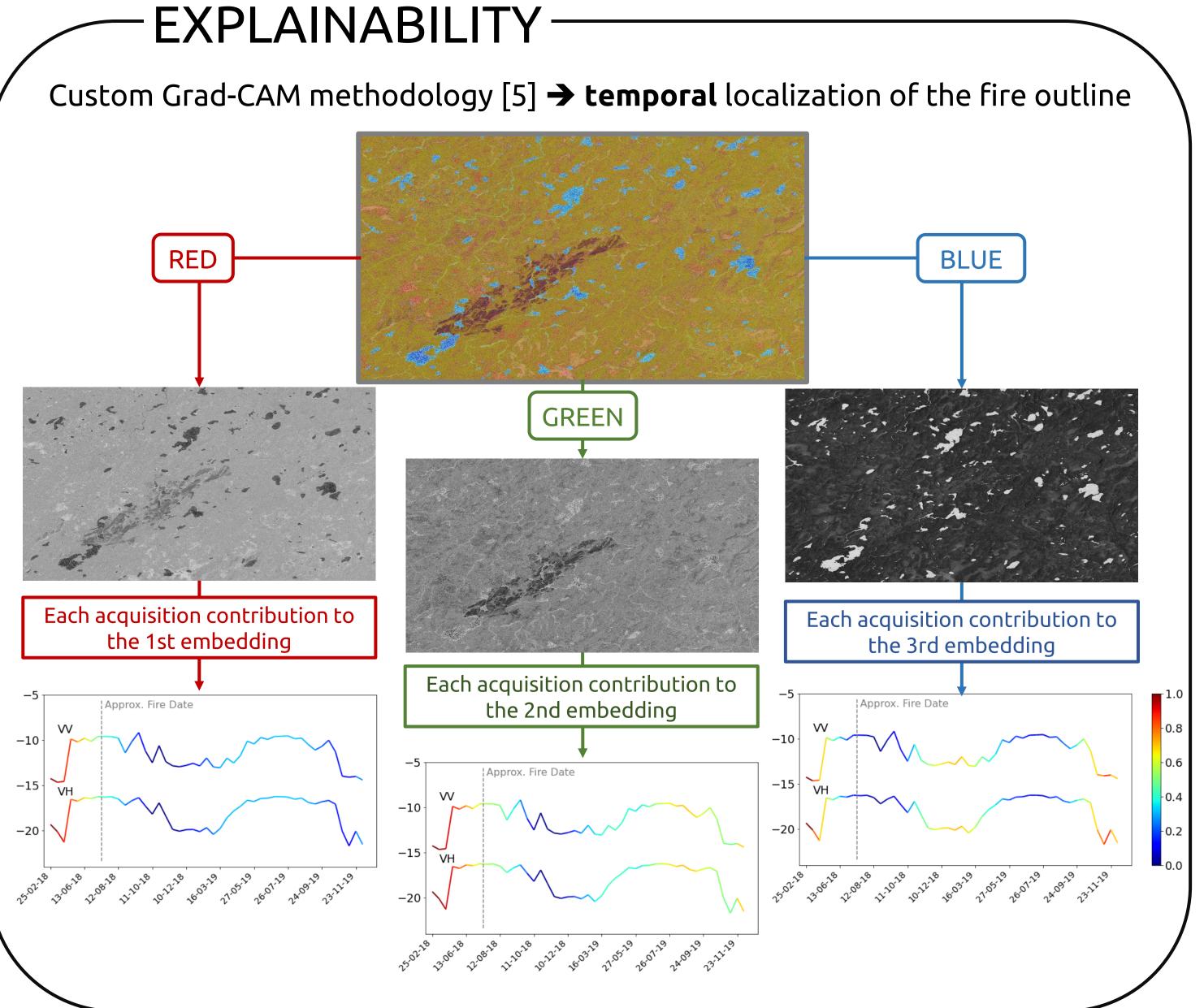
- 1. Model radiometric temporal profile of forests using Convolutional Autoencoders [4].
- 2. Visualize variations in the profiles of forest using the generated embedding space.
- 3. Isolate profiles deviating from a norm as being fire.



STUDY DATASET-



- Ontario Fire, starting date: 21st of June 2018.
- Superficies: ~760 Ha.
- Fire outline data source: 2021 Canadian National Fire Database product



RESULTS 1. Model Sentinel-1 Time-Recreated Sentinel Reconstruction loss: 1 Time-series $L(p, \tilde{p}) = \frac{1}{N} \sum_{i=1}^{N} \left(\frac{1}{T} \sum_{t=1}^{T} \left(p_i^{(t)} - \tilde{p_i^{(t)}} \right)^2 \right)$ series (VV, VH) $\{\widetilde{p_1},\widetilde{p_2},\cdots,\widetilde{p_N}\}$ $\{p_1, p_2, \cdots, p_N\}$ Official Fire outline 3D Embedding Image, mapped to RGB color space Visual interpretation of embeddings \rightarrow spatial localization of the fire outline Fire outline and embedding image superposition 3. Isolate

CONCLUSION

Thanks to the modeling of SAR time series of forested environments with Convolutional Autoencoders, we can:

- Extract and visualize the main temporal profiles within a forested scene.
- Adopt an anomaly detection viewpoint to model "normal" forest temporal signatures, and the **extraction** of "abnormal" forest temporal signatures.
- Leverage this viewpoint to retrieve fire outlines as "abnormal" profiles without supervision.

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