

Novel use of EO satellite data and AI in railways: SPATRA Project



EUSPA AI week 2026

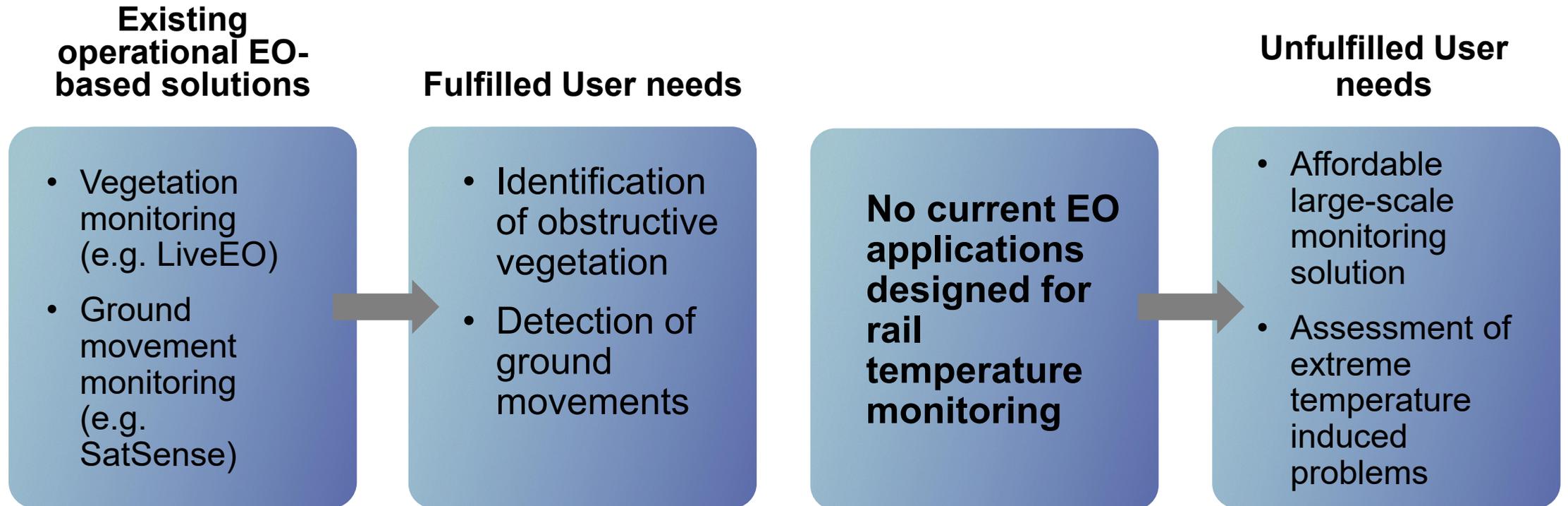
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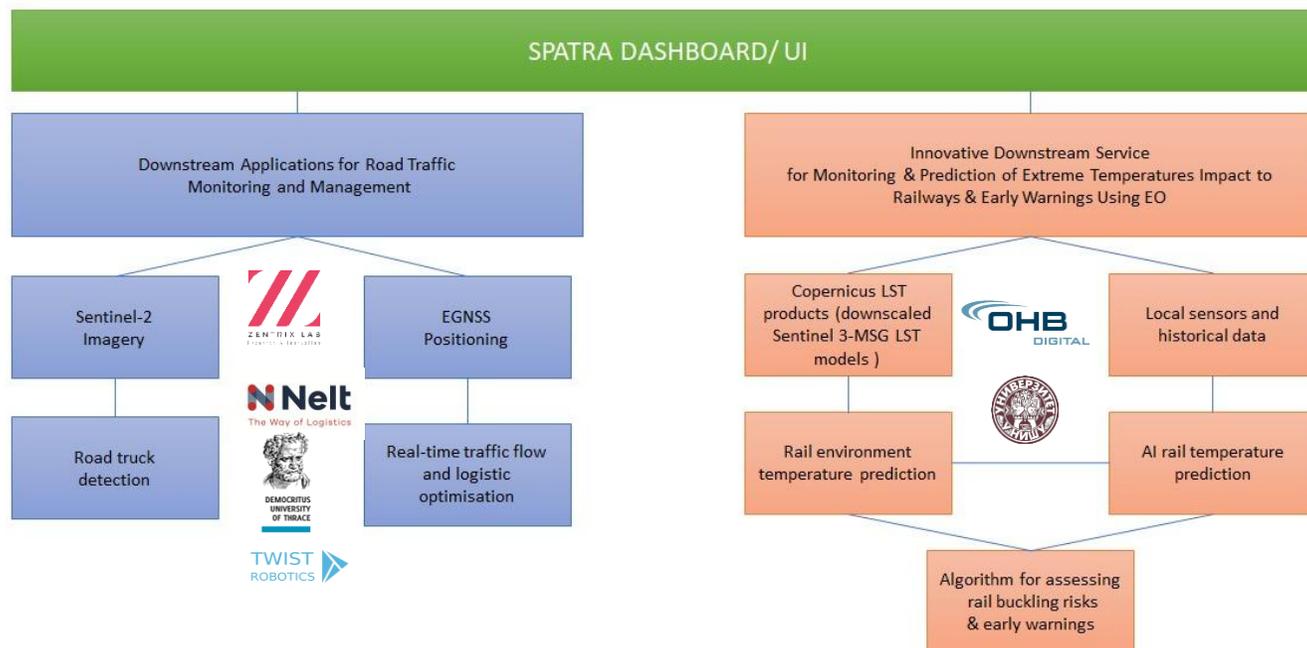
Usage of satellite data in railways

- The use of satellite data in railways primarily revolves around navigation and the application of Global Navigation Satellite Systems (GNSS).
- Using of Earth Observation (EO) data is a cutting-edge approach that leverages satellite technology to enhance efficiency and safety of rail networks.
- Advanced satellite sensors delivering detailed EO data, combined with AI, enable rail network management to become more efficient, effective, and safer.

Using of EO data for railway infrastructure monitoring



SPATRA: Space-based applications for transport monitoring and management



Use-case 1 - Road

- Combining the positioning, navigation, and timing with Earth Observation services through the use of EGNSS and Copernicus for the real-time monitoring of road traffic flow and prediction of logistics networks

Use-case 2 - Rail

- Space-based service predicting the risk of rail buckling based on downstream application for estimation of the rail temperature that can support rail management plan to prevent track buckling induced train derailments



Cross border road congestion

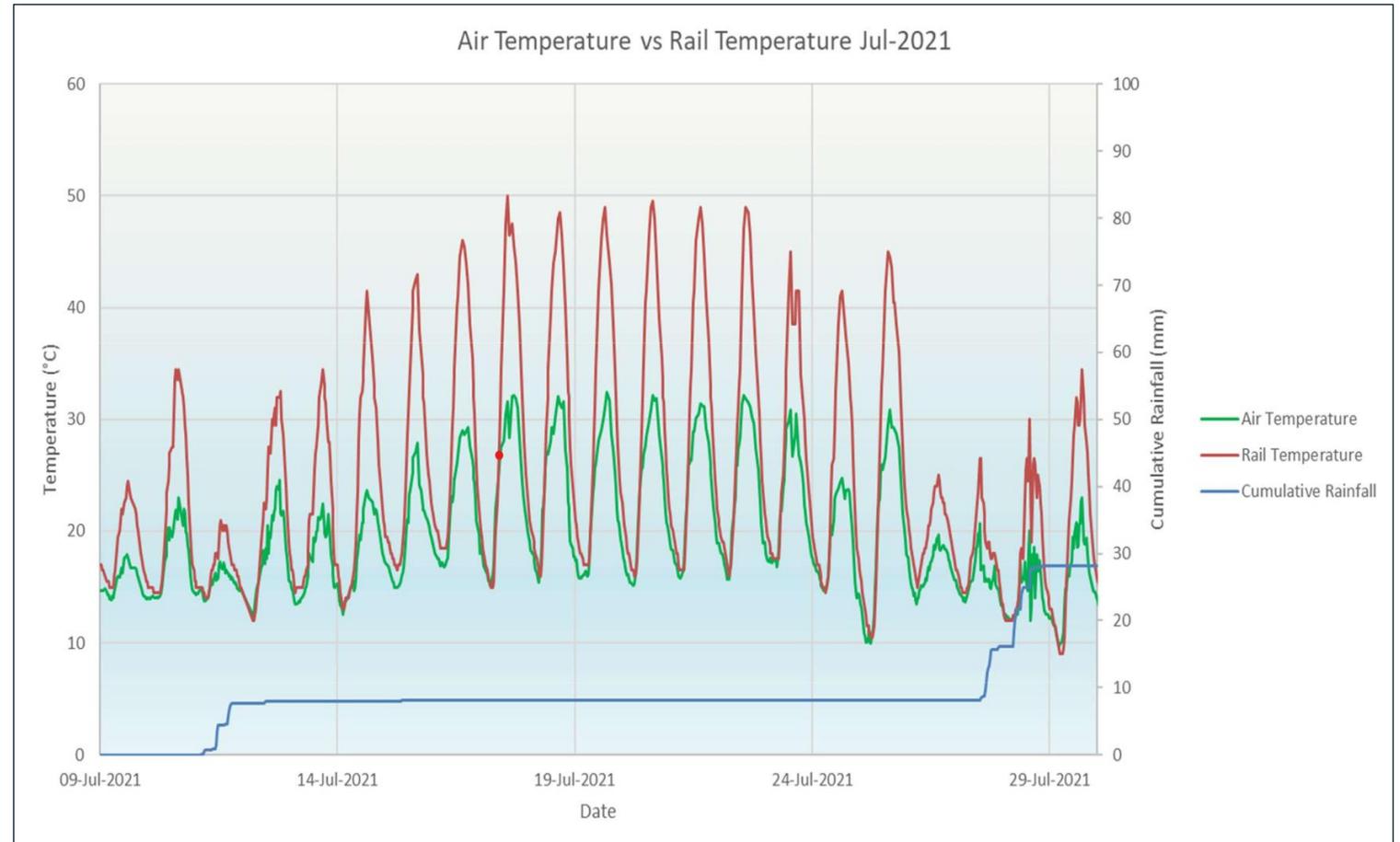


Lateral buckling of the track due to high temperatures

Impact of extreme temperatures to rail infrastructure

! The rail temperature can significantly exceed the air temperature.

! Especially in high temperature ranges, differences of over 20°C occur



Source: Tren Trace (2021*)

Impact of extreme temperatures to rail infrastructure

! Temperatures above approx. 60°C are classified as critical

Quelle: DZSF(2020)

! Possible consequence: track distortions / „buckling“



Source: ABproTWE
(2013)

Impact of extreme temperatures to rail infrastructure



Safety risks

- Danger of derailments
- Failure of signalling and control technology
- Increased risk of fire
- Health risks for employees and passengers



Costs

- Delays/cancellations
- Maintenance and repairs
- Energy costs

! In 2024, NetworkRail reported that buckled rails caused by hot weather led to more than 350,000 minutes – or 240 days of delays.

Satellite-based solutions vs current practice



Weather forecast

- + Cost-effective
- + Scalable
- + Predictive
- Imprecise
- Delayed in time
- Not localised



In-situ sensors

- + Precise
- + Real-time information
- + Localised
- Not scalable
- Maintenance intensive

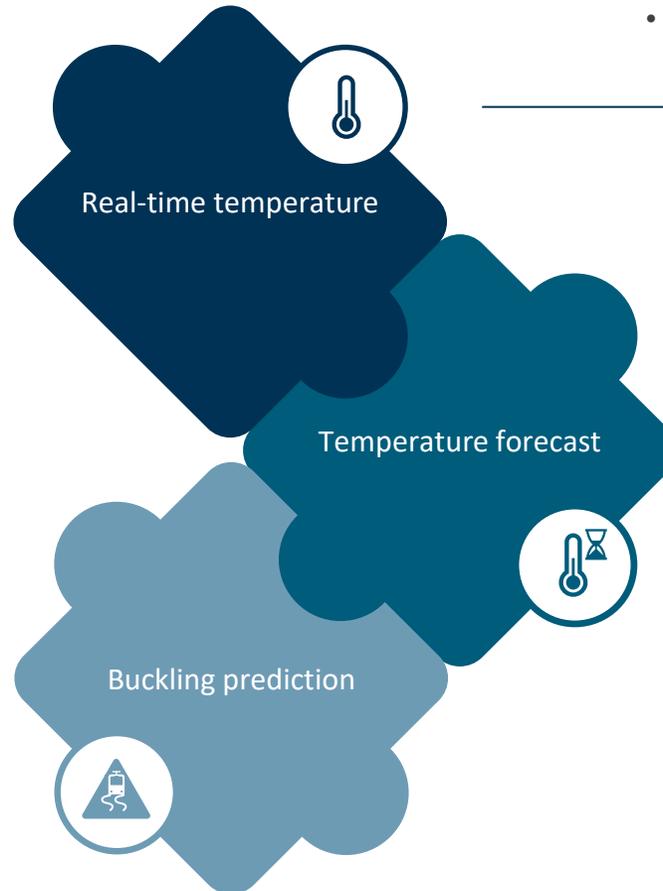


Satellite-based Rail Temp estimation

- + Cost-effective
- + Scalable
- + Predictive
- + Precise
- + Real-time information
- + Localised

SPATRA solution for rail buckling risk estimation - Novel use of EO satellite data and AI in railways

- Rail lateral displacement prediction
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- High-resolution Land Surface Temperature (LST) data in rail environment
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- Rail track temperature prediction (RTTP)
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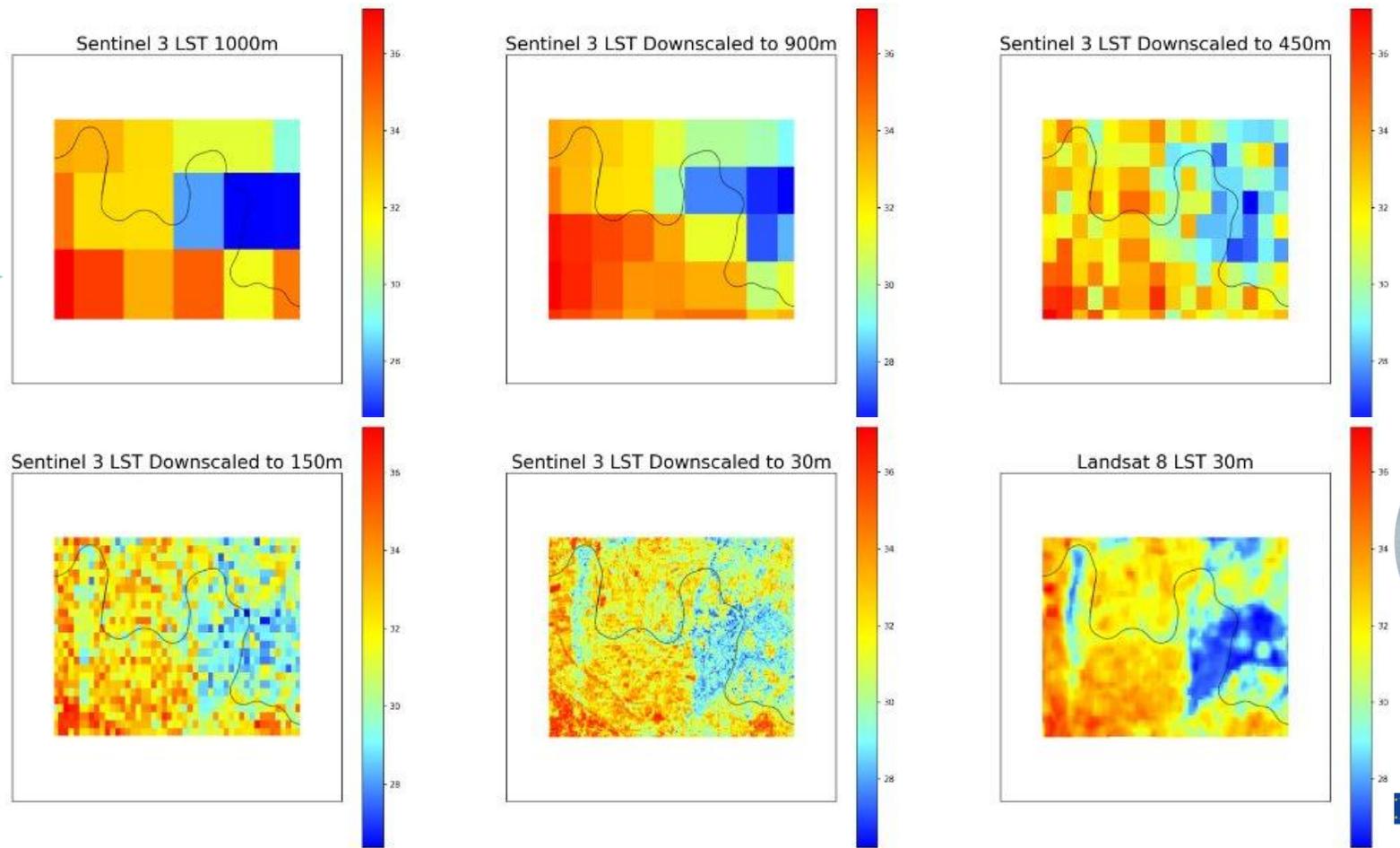
High-resolution LST data in rail environment

Thermal Downscaling

Data: Land Surface Temperature (LST) from

- Sentinel-3 (1km)
 - Copernicus Land Monitoring Service (CLMS) (5km)
 - Aqua/Terra MODIS (1km)
- combined with optical data from Sentinel-2

Method: stepwise random forest algorithm for improvement from 1km/5km spatial resolution to 500m or finer¹



1 | Li, X.; Zhang, G.; Zhu, S.; Xu, Y. Step-By-Step Downscaling of Land Surface Temperature Considering Urban Spatial Morphological Parameters. Remote Sens. 2022, 14, 3038. <https://doi.org/10.3390/rs14133038>

Rail track temperature prediction

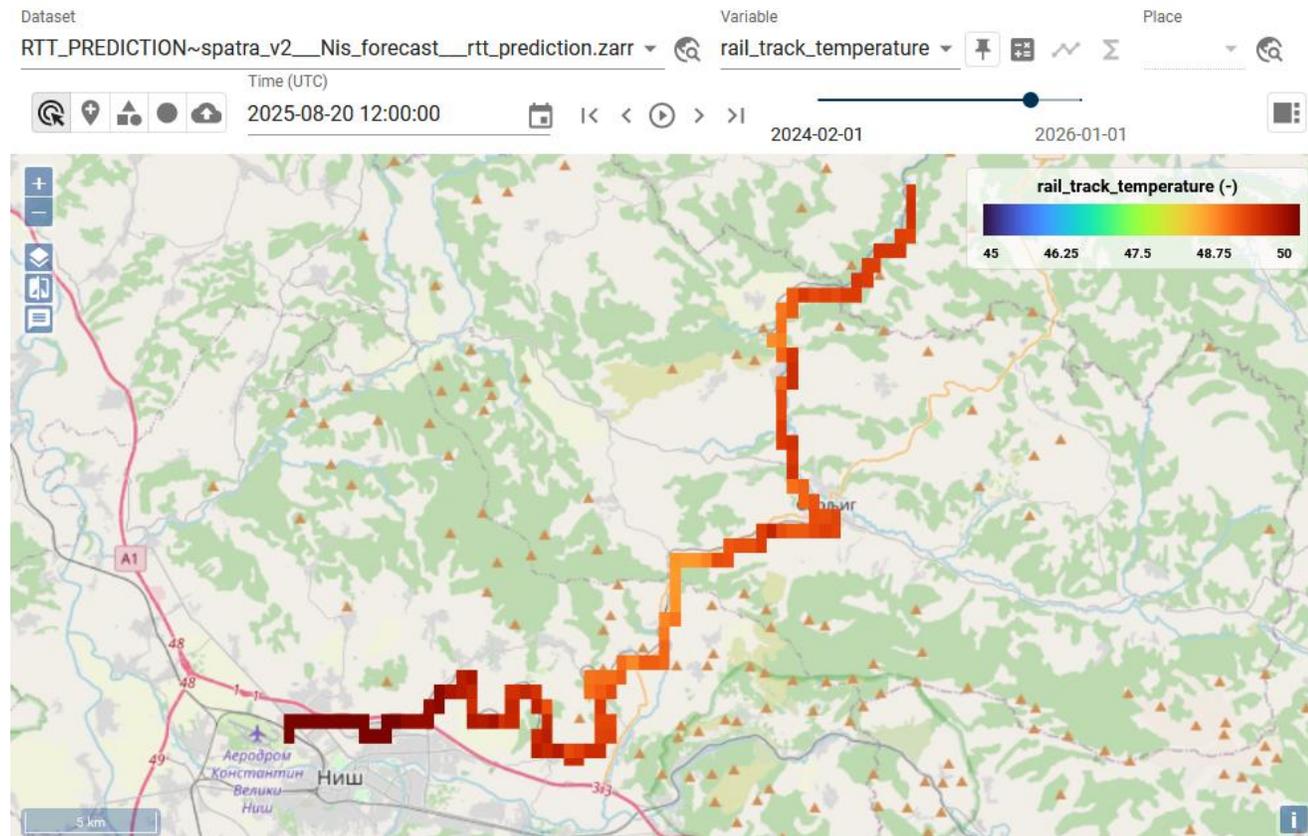
Rail Track Temperature (RTT) Prediction

Data:

- in situ RTT measurements
- downscaled high-res. LST
- weather forecasts
- solar effects

Method: : eXtrem Gradient Boosting (XGBoost) algorithm to predict hourly rail track temperature²

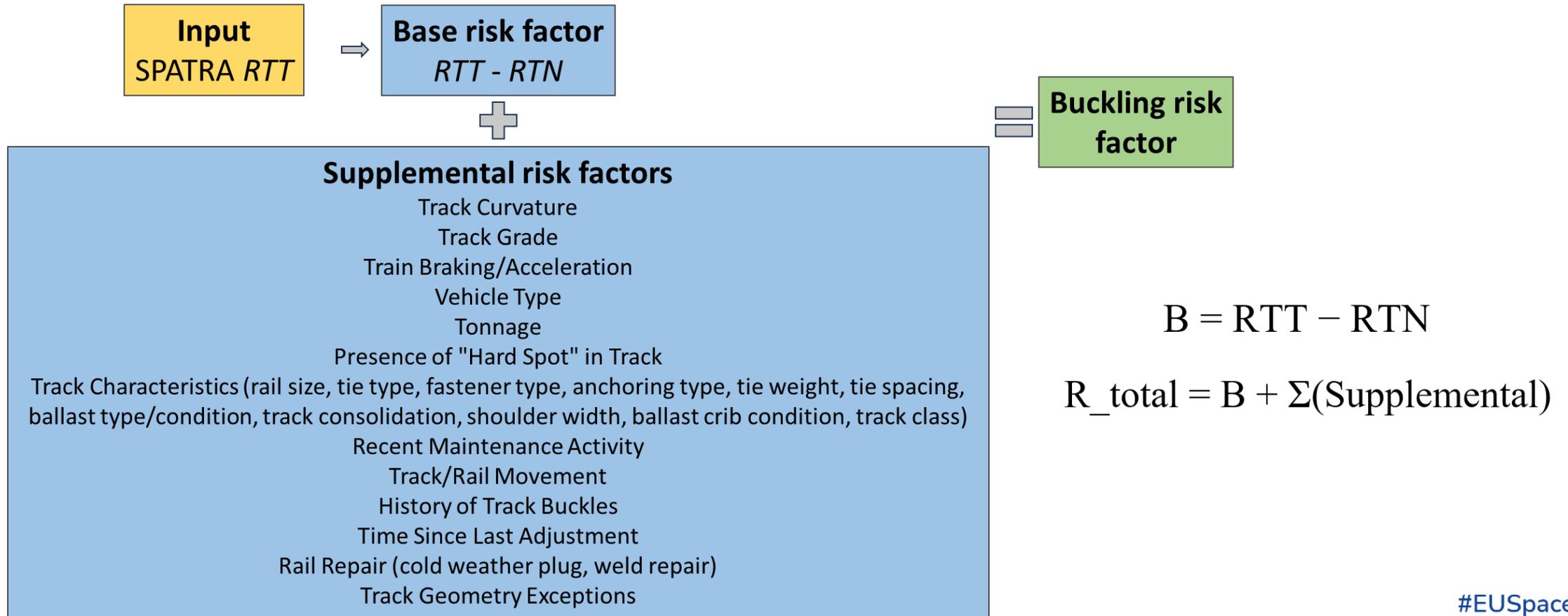
Validation: using the internal - in situ data with a training-/test-data split and thermal drone measurements for validation at larger scale



Predicted rail track temperature for test track from Niš to Svrljig, Serbia

Algorithm for assessment of rail track buckling risk

Classical approach



Algorithm for assessment of rail track buckling risk

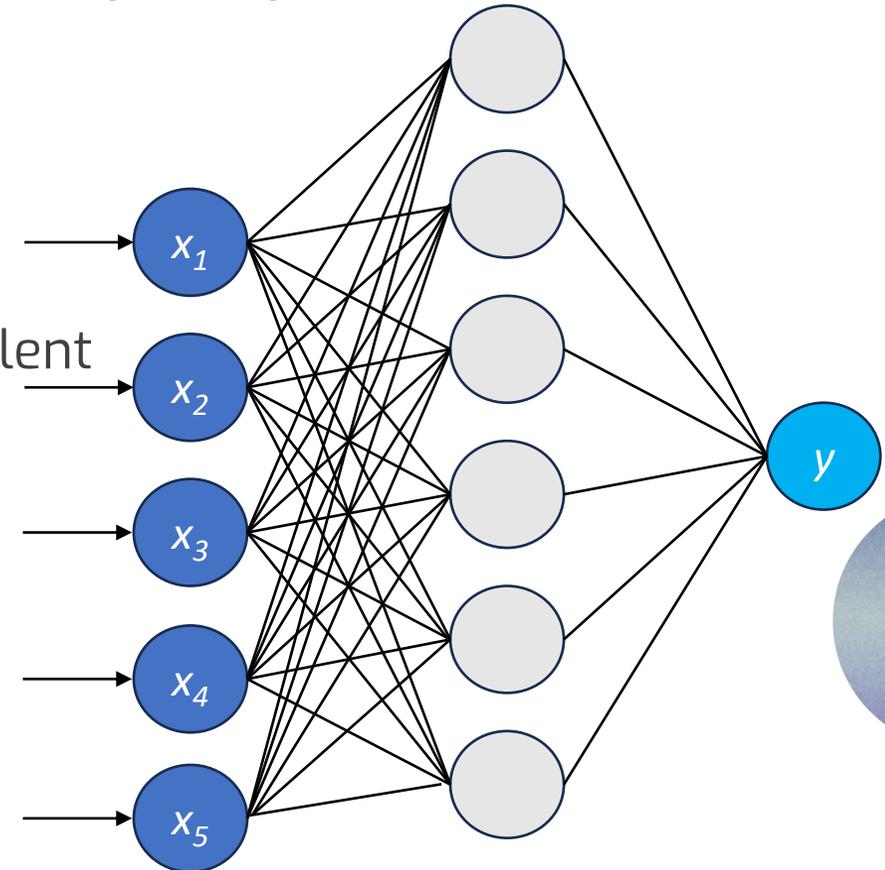
Machine learning - artificial neural network (ANN) model

Inputs:

- sleeper type (x_1) - binary; 1 = concrete, 0 = wood
- track curvature (x_2) - continuous
- rail track condition (x_3) - ordinal 1–10; 1 = very poor, 10 = excellent
- hard point (x_4) - binary; 1 = present, 0 = absent
- rail track temperature (RTT – x_5) - continuous

Output:

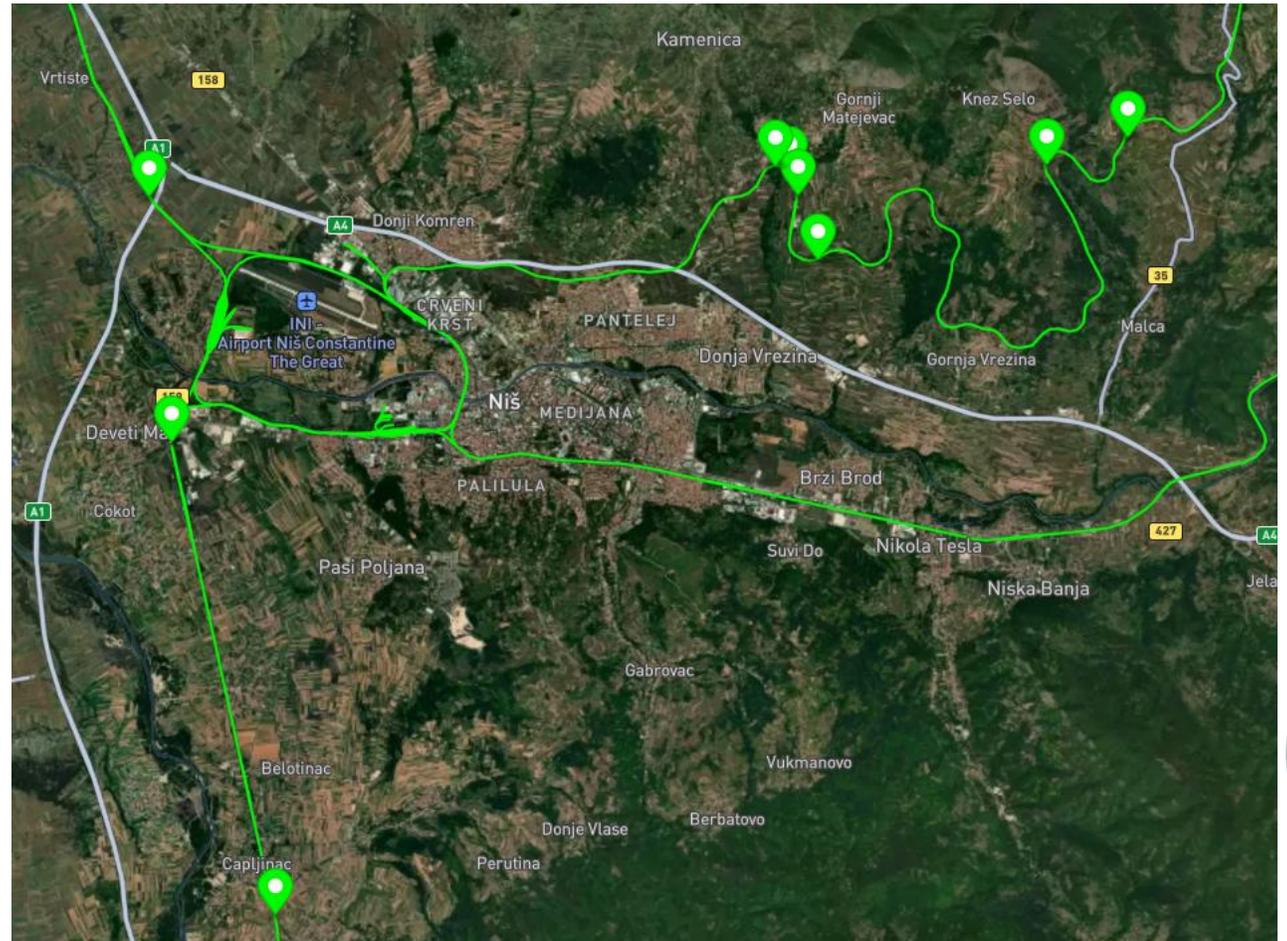
- rail lateral displacement (y)



Algorithm for assessment of rail track buckling risk

Dataset generation

- 14 positions, 25 measurement points
- new (4 points), reconstructed (13 points) and old track (8 points)
- two rail sizes
- wooden and concrete sleepers with appropriate fastening system
- 5 different quality (accessed by ISR maintenance personal based on measurement car results)
- different rail curvature
- different sun exposure during the day



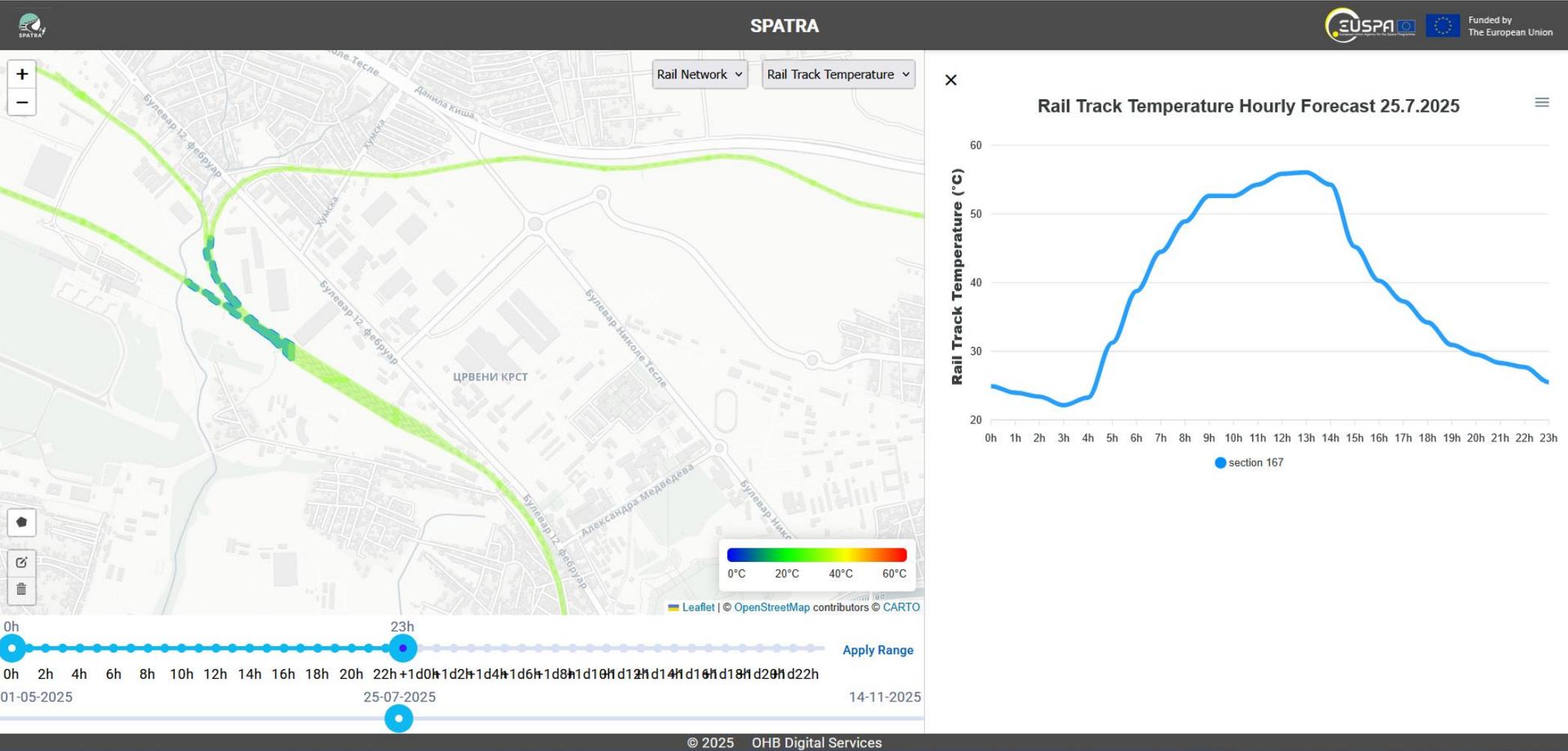
Algorithm for assessment of rail track buckling risk

Mobile in situ measuring station

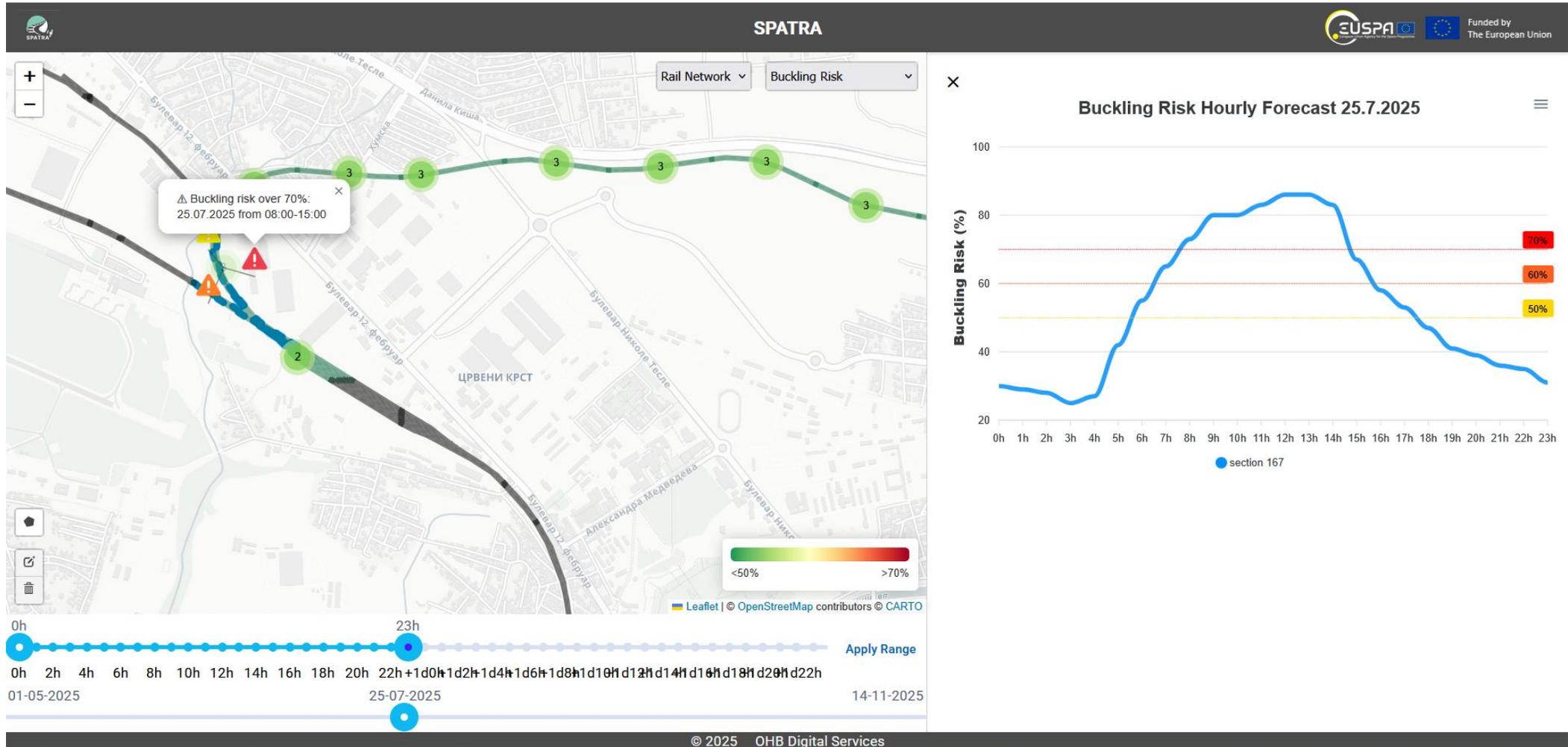
- Weather station (wind speed and direction, air temperature, pressure and humidity, solar radiation, rain fall, PM particles)
- Rail station, rail lateral deflection and temperature
- Data gather locally and transmitted to SPATRA cloud



SPATRA GUI- rail track temperature prediction



SPATRA GUI- rail track buckling risk



THANK YOU!

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