



MultiMiner
Earth Observation for Smart Mining



Exploring Innovative Methods for Air Dust Pollution Monitoring using Satellite Data and Products: A Case Study of Chalkidiki (Greece)

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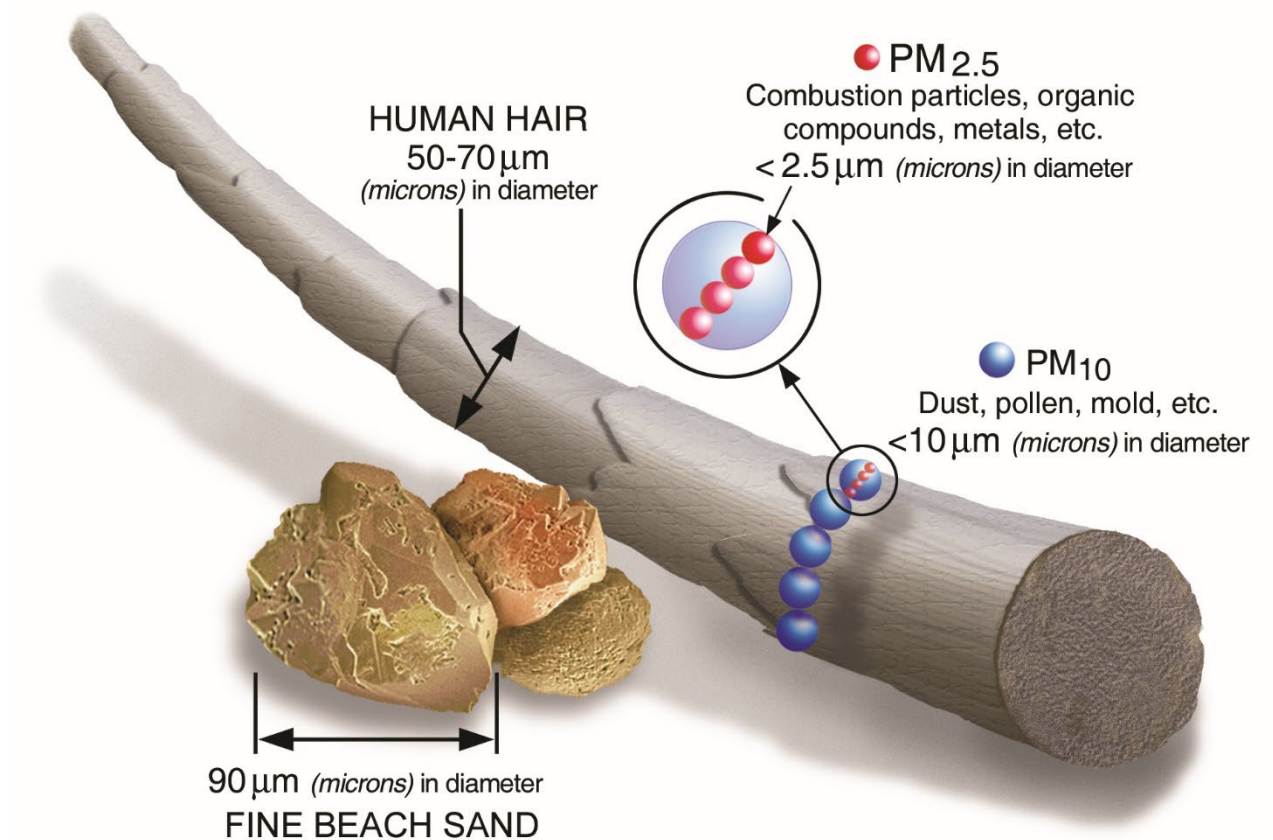


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Introduction: atmospheric and surface dust monitoring



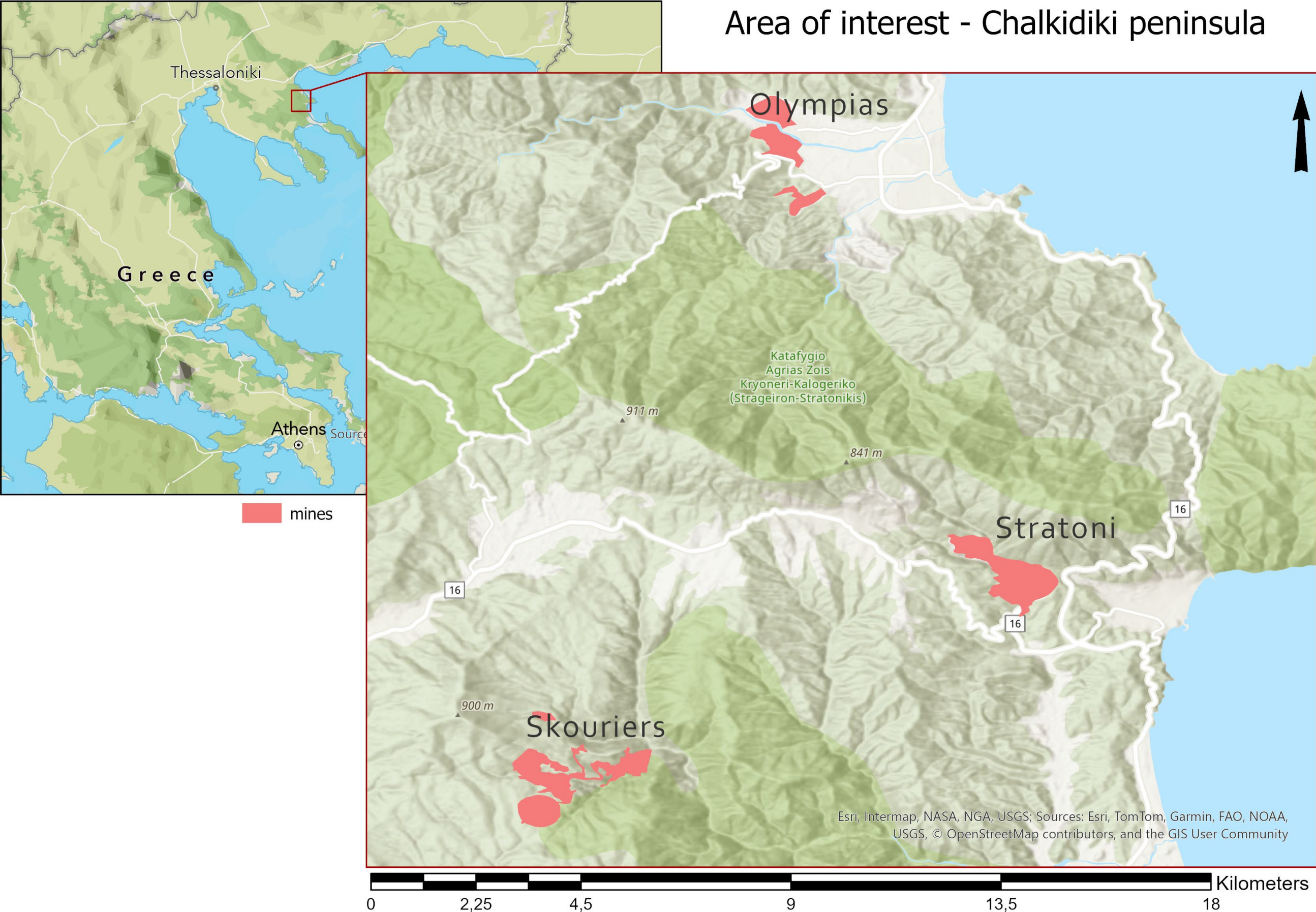
- Surface mining activities significantly contribute to environmental degradation and pose risk to human health
 - One of the main type of pollution is particulate matter that is dispersed into the air through mining processes
 - Particulate matter (PM10) or air dust pollutants are particles with an aerodynamic diameter $\leq 10 \mu\text{m}$, suspended in the air
 - PM10 particles are inhalable, penetrating into the respiratory tract and can cause respiratory and cardiovascular diseases, long term exposure is linked to reducing lung function
 - Monitoring of this type of pollution can be challenging due its spatial distribution in three dimensional space
 - Satellite data can offer stable spatial and temporal coverage over large and inaccessible areas
- Focus on overall dust pollution evaluation and the usability of remote sensing data in the measurement of particular pollution**



Test site: Chalkidiki, Greece



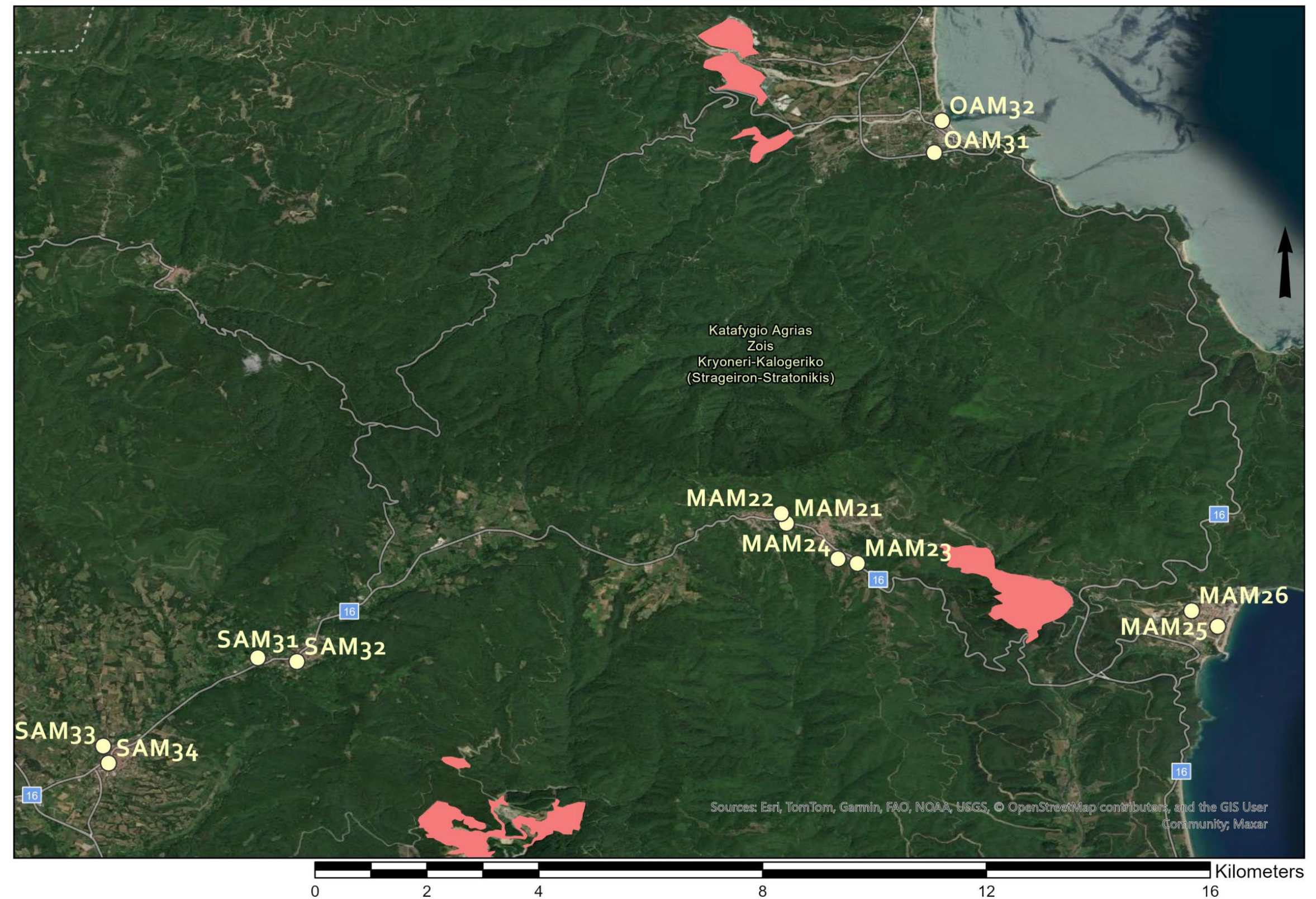
- Open pit mining sites operated by Hellas Gold, a subsidiary of Eldorado Gold Corporation
- Copper, gold, silver, lead, zinc



Datasets: IN-SITU measurements of dust in residential areas



- PM10 data from Hellas Gold platform
 - Hourly values of PM10 aggregated in daily medians at 12 stations
 - From 6/2015 to 8/2024 (based on location)
- Data available at <https://environmental.hellas-gold.com/>



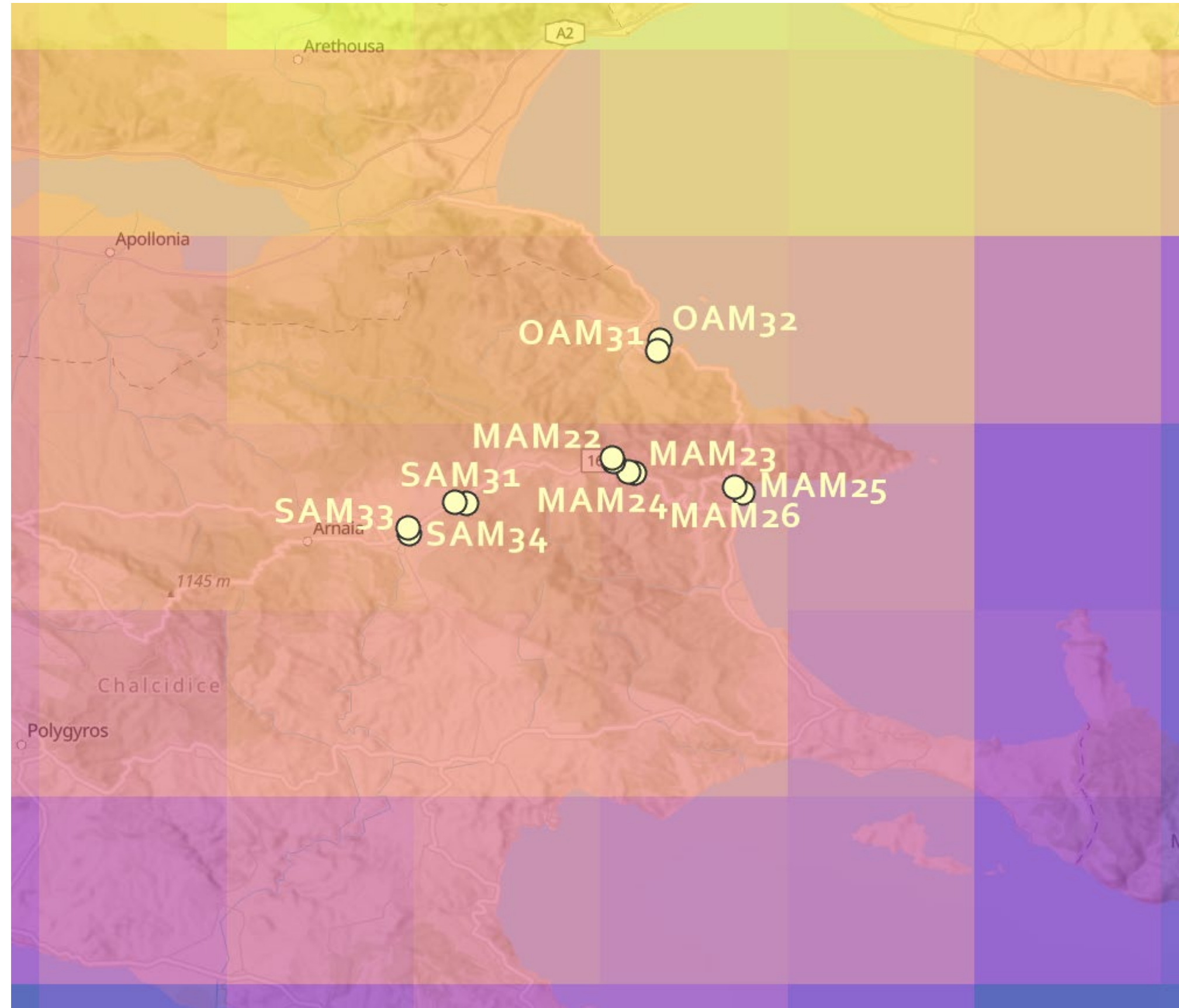
Datasets: Satellite data product - CAMS



Atmosphere
Monitoring Service

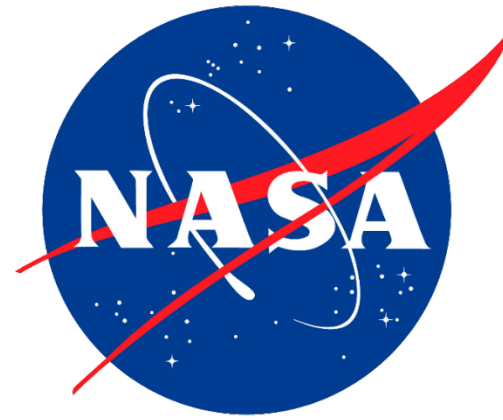
atmosphere.copernicus.eu

- Copernicus Atmosphere Monitoring Service
- Numerical models from satellite and ground based observations
- CAMS European air quality reanalyses and analysis data
 - Particulate matter < 10 μm (PM10)
 - Level 0
 - Hourly values aggregated in daily medians
 - From 1/2015 to 8/2024
 - Spatial resolution ~10 km



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Datasets: Satellite data - MODIS



- NASA's Terra and Aqua satellites
- Remote sensing measurements of aerosol optical depth
- Optical depth of atmosphere
 - MOD04_L2 – Optical_Depth_Land_And_Ocean
 - 10 km
 - MOD04_3K – Optical_Depth_Land_And_Ocean
 - 3 km
 - MAIAC MCD19A2.061 – Optical_Depth_055
 - 1 km
- Data aggregated into daily medians (max. four overpass per day)

MOD04_3K Land and ocean AOD from 29.8. 2024

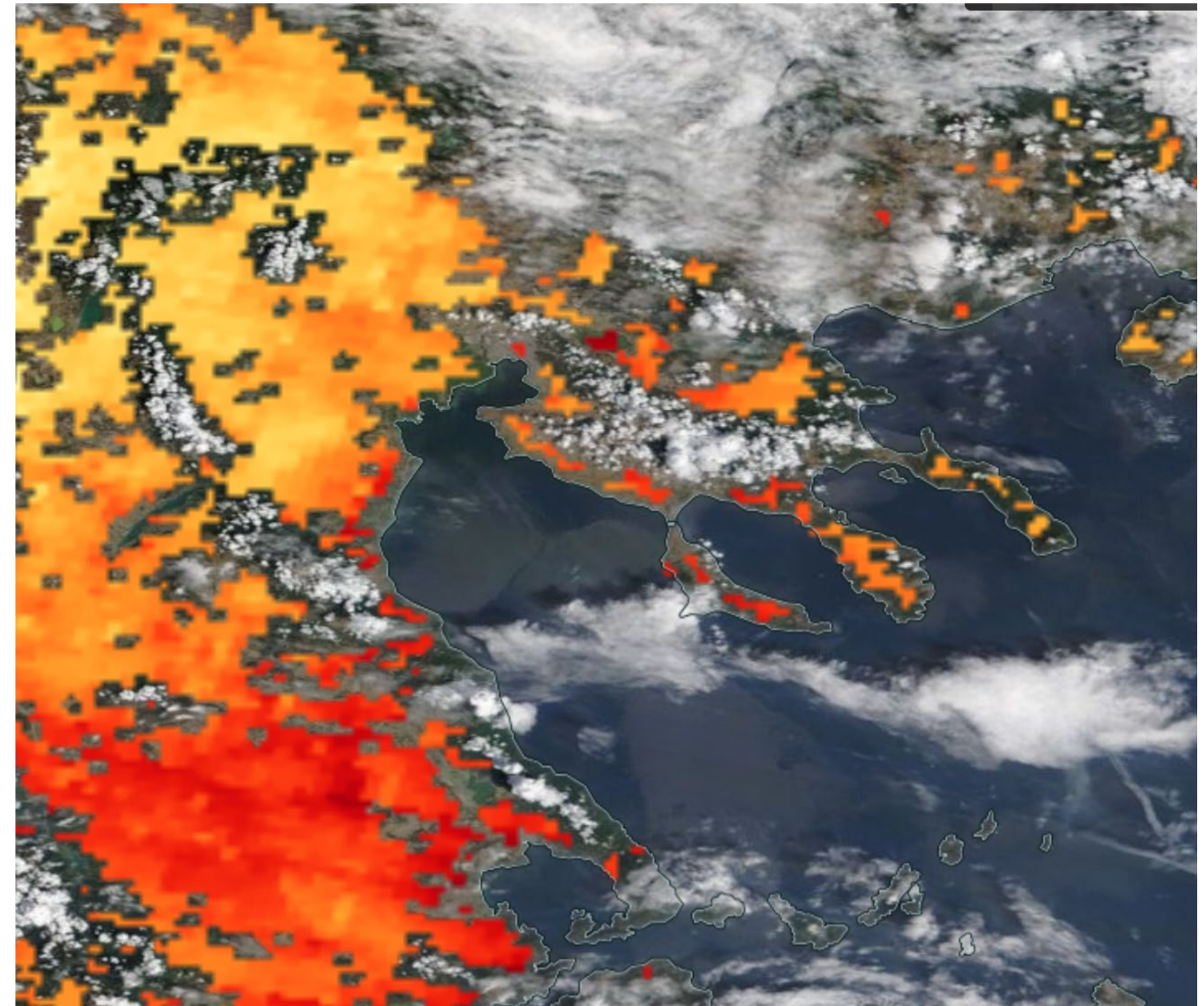


Fig.2 - NASA Worldview Snapshots, accessed Dec 2, 2024



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Methodology



Preprocessing	Data Acquisition: data downloaded via API python scripts for 1/2015 - 8/2024 related to station positions and aggregated in to daily median values Spatial data aggregation: based on CAMS resolution – means from „pixel containing“ stations measurements
Correlation Analysis	Spearman correlation for observing relationships in: <ul style="list-style-type: none">• Three AOD MODIS datasets and IN-SITU data for year 2015• CAMS, MODIS and IN-SITU datasets for individual station measurements and satellite-derived values for 2015 - 2024
Station Specificity Analysis	PCA analysis of in-situ time series data for observing similarity in stations measurements and time series visualization – annual and monthly means
Trends and Seasonality Analysis	Seasonal decomposition of time series into trend, seasonal and residual variables, observing differences between CAMS and IN-SITU datasets
Prophet Forecasting	Used Meta prophet forecasting procedure for modelling and predicting seasonal trends for both CAMS and IN-SITU clustered data Useful for predicting time series with strong seasonality on longer time series <ul style="list-style-type: none">• Trained on 2015-2023 monthly values• Validation with 2024 data• Prediction extended to the end of 2025
Spatial Trends and Seasonality	PCA analysis of CAMS rasters divided in wet season (Nov-Apr) and dry season (May-Oct) for main wind directions and dust hotspots observing



Results: Correlations

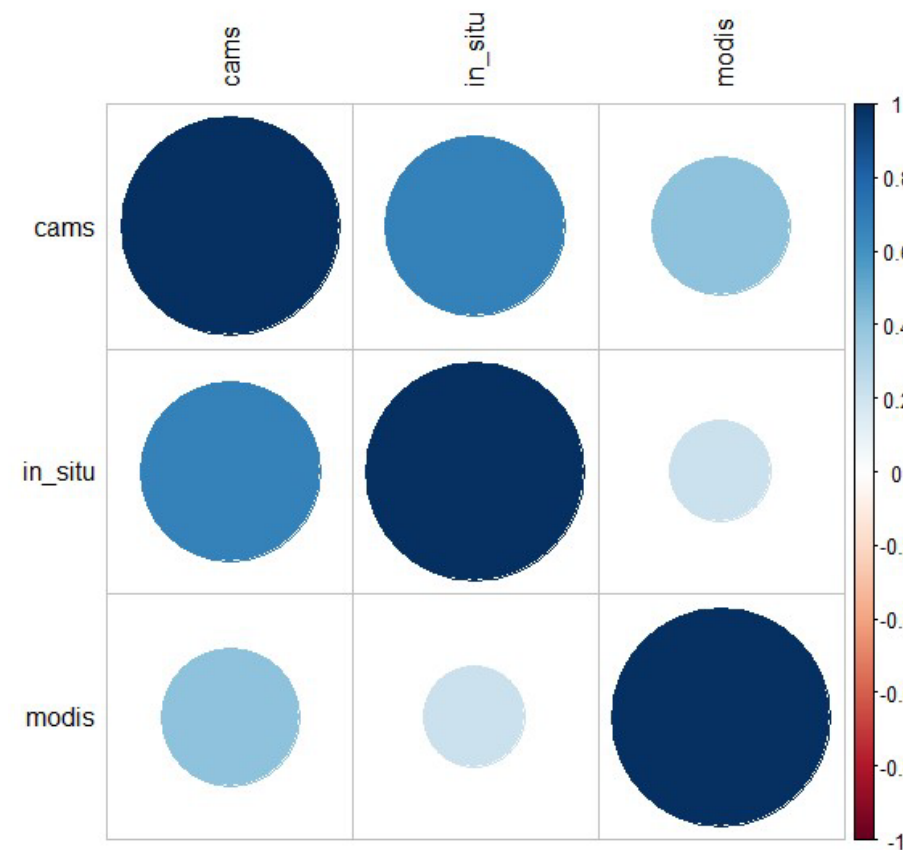


Correlations of MODIS datasets

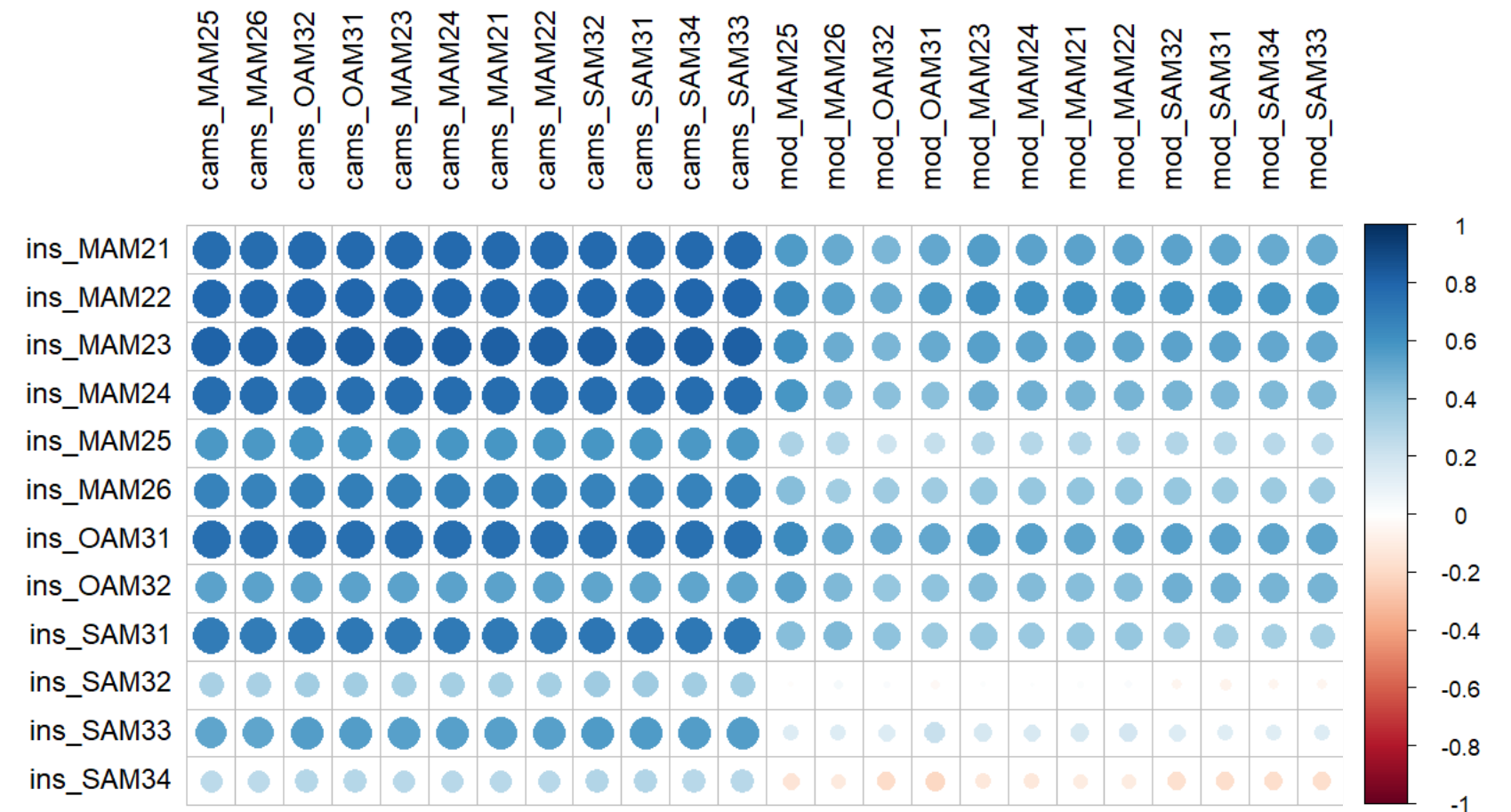
- Examination of three AOD datasets – MOD04_L2, MOD04_3K, MCD19a2.061
- Challenging due cloud coverage and missing values
- MOD04_3k best results based on highest available measurements and highest correlation with in-situ
- **$r_s \approx 0,35$ (based on location), 61,9% missing values -> insufficient, will not be used**

CAMS correlations

- Examination of pm10 variable
- Challenging due low spatial resolution, lack of identifying local extremes
- **$r_s \approx 0,72$ (based on location) -> sufficient, will be examined deeper**

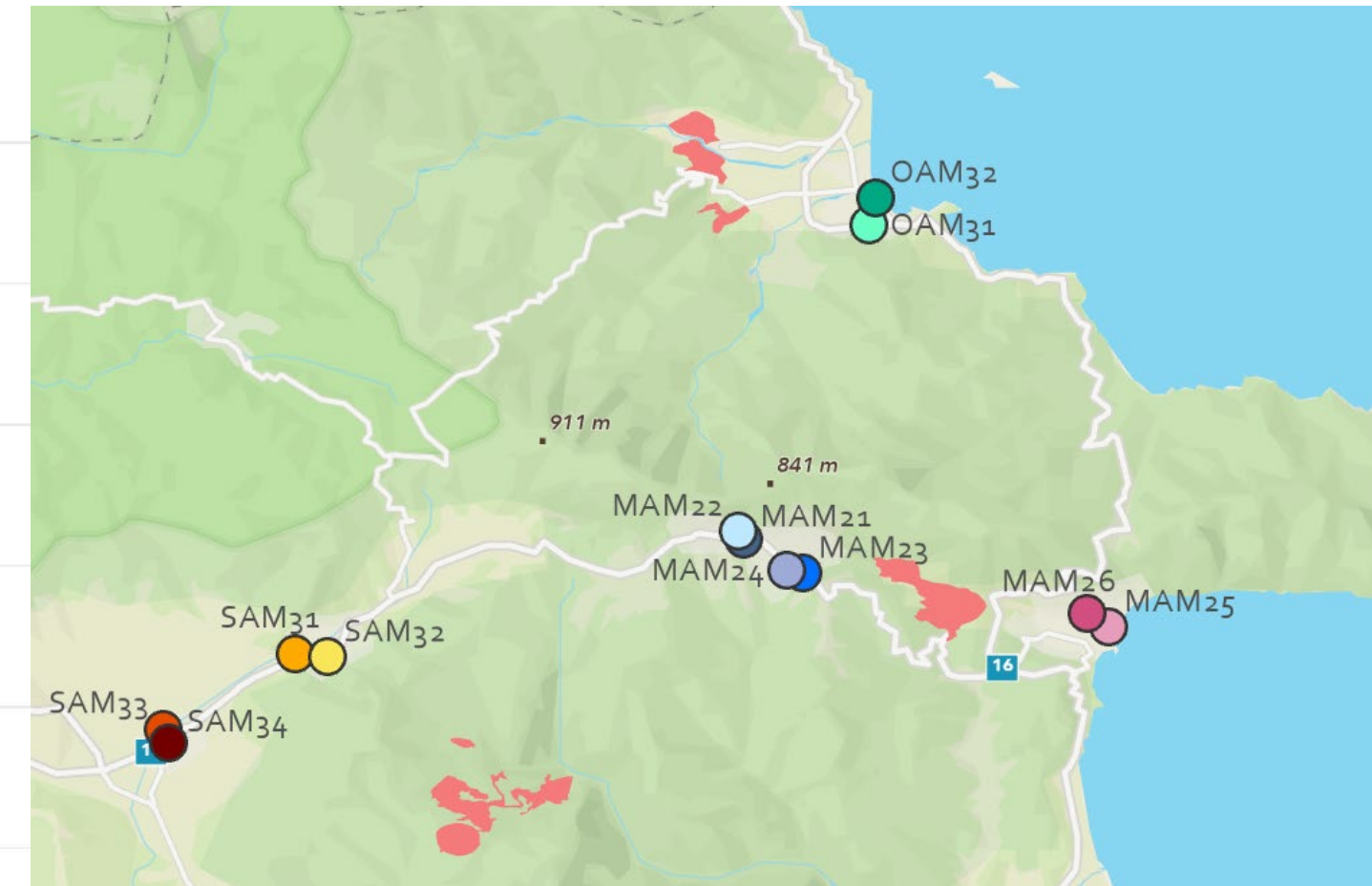
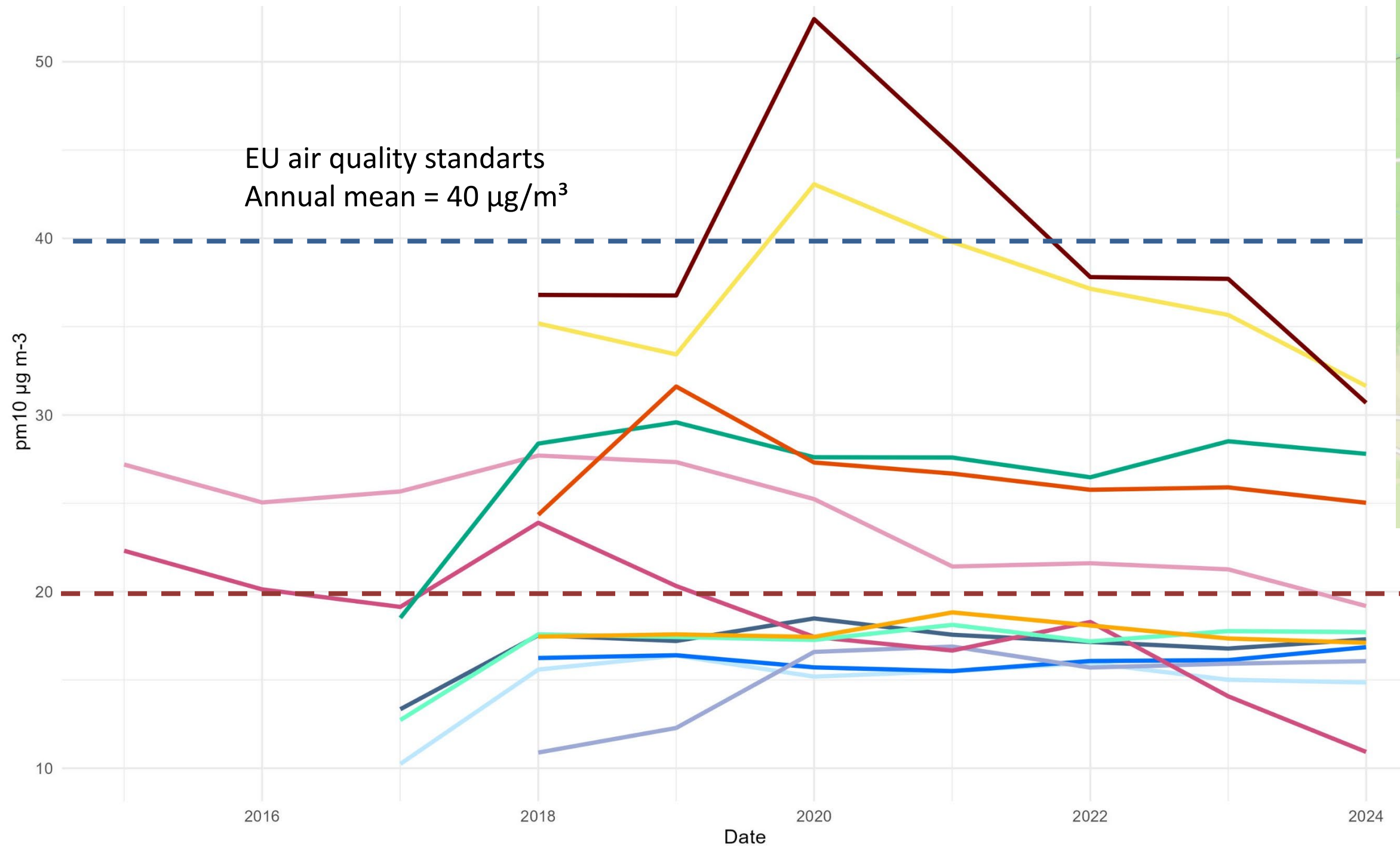


Correlation between datasets



Correlation between sites datasets measurements

Analysis of stations: annual IN-SITU means

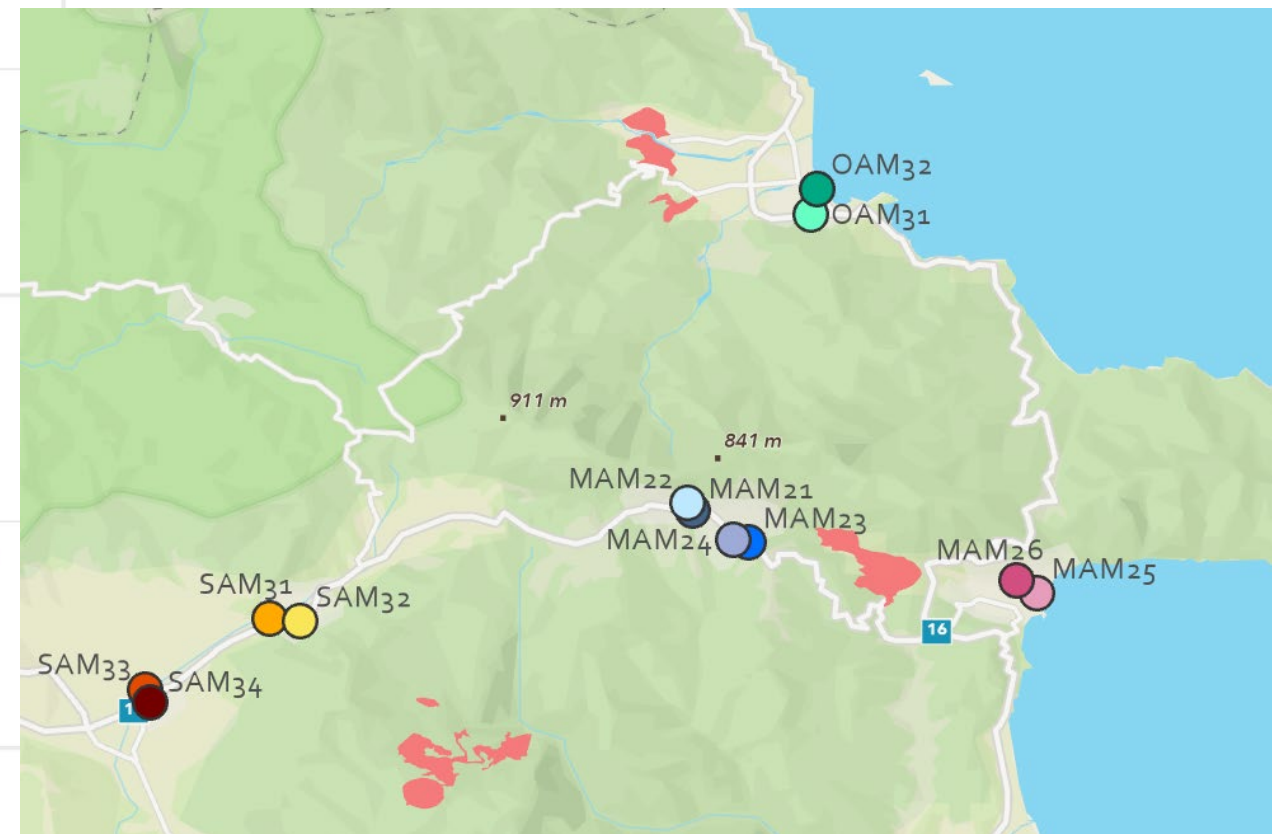
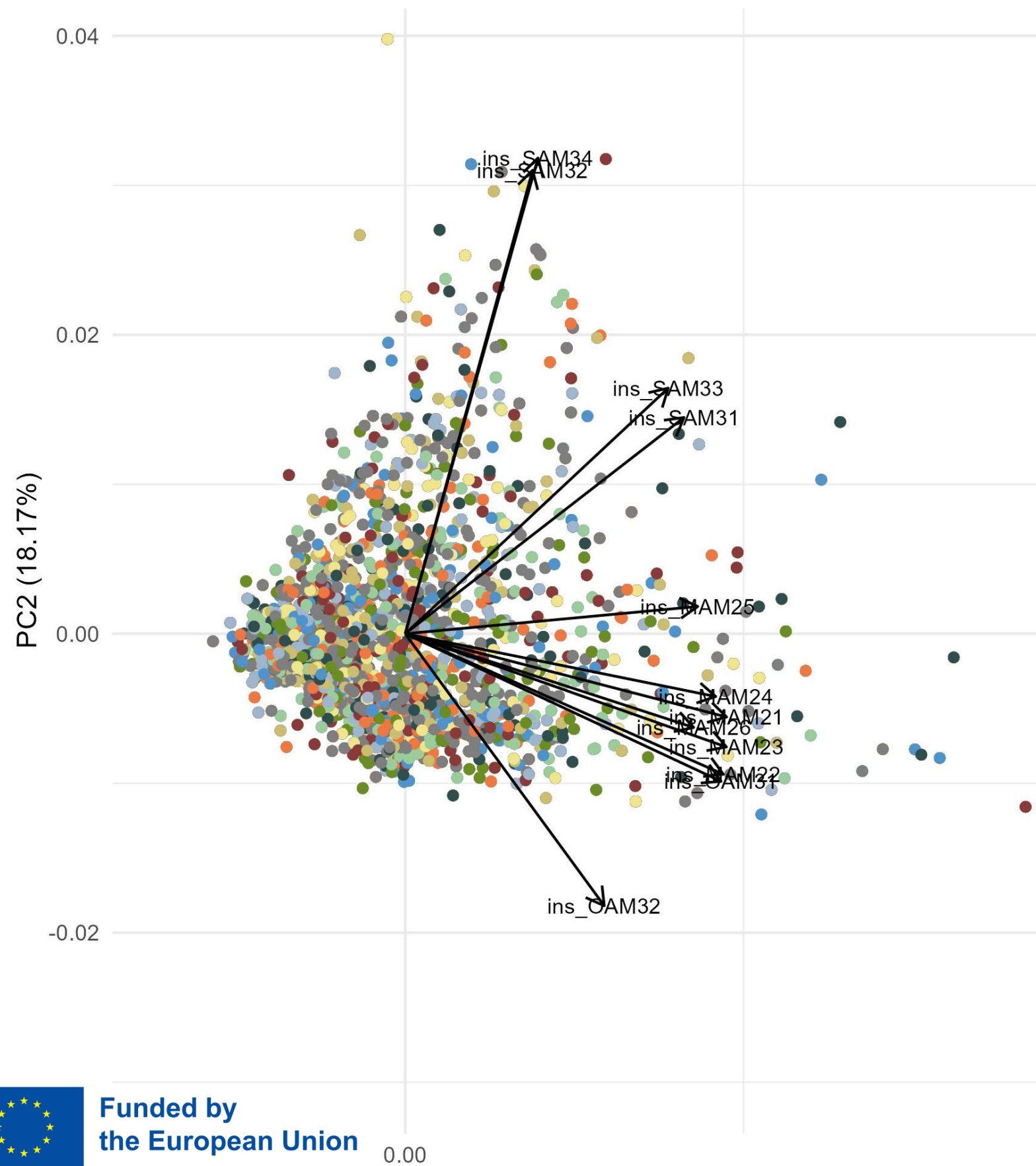


WHO annual mean guideline
Annual mean = 20 $\mu\text{g}/\text{m}^3$

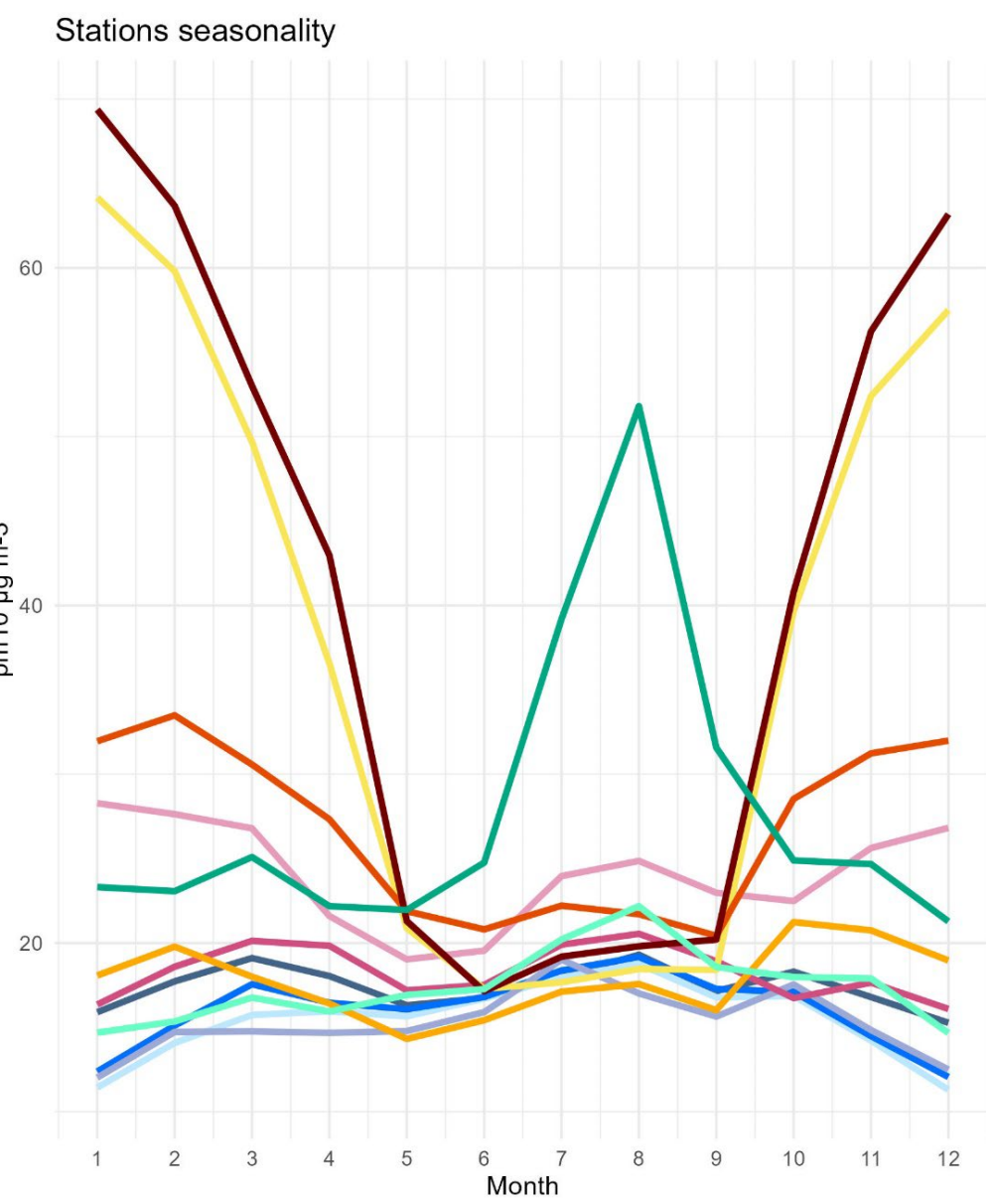
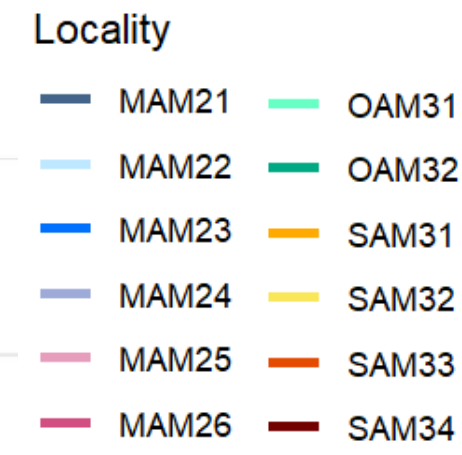
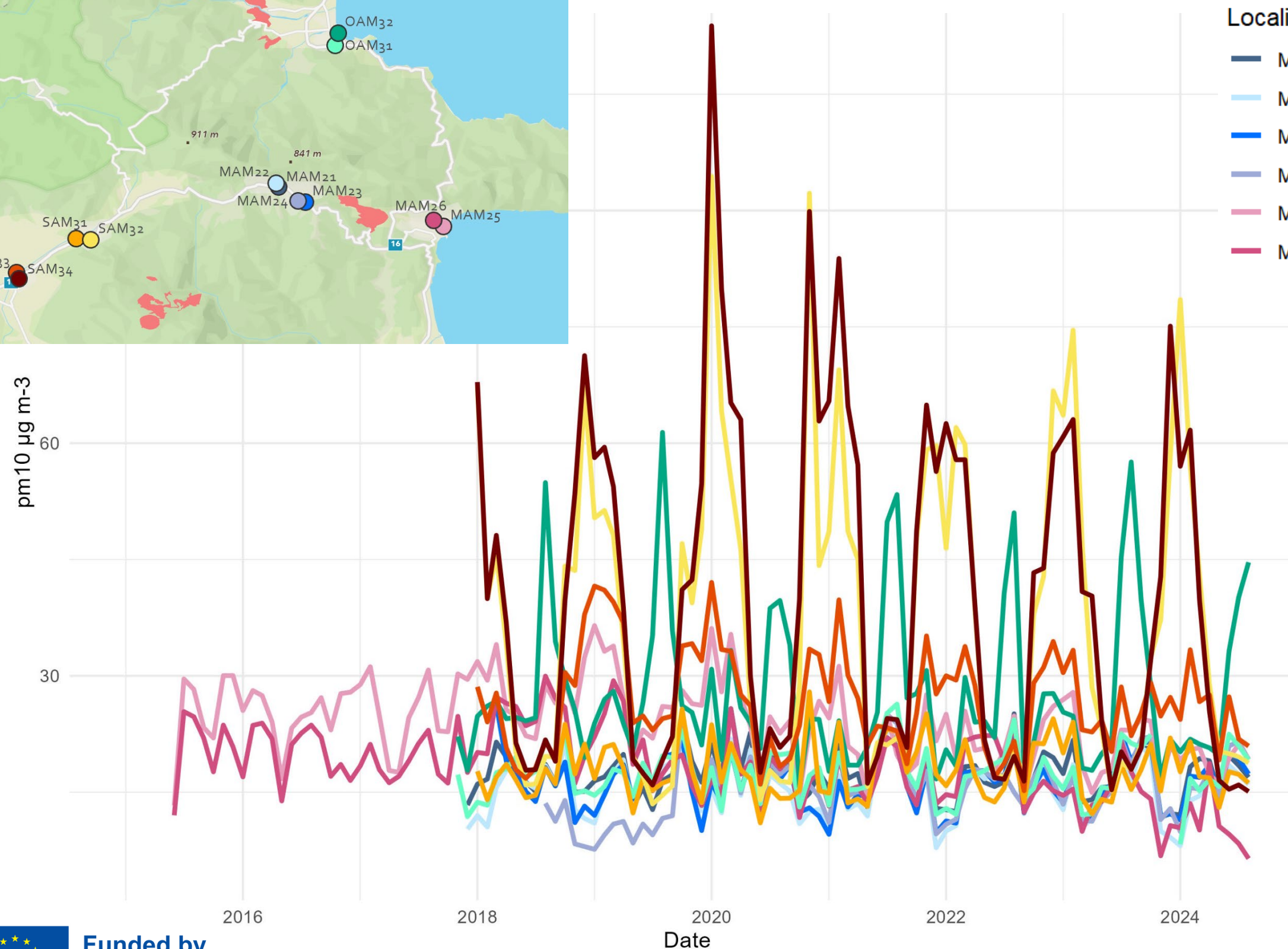
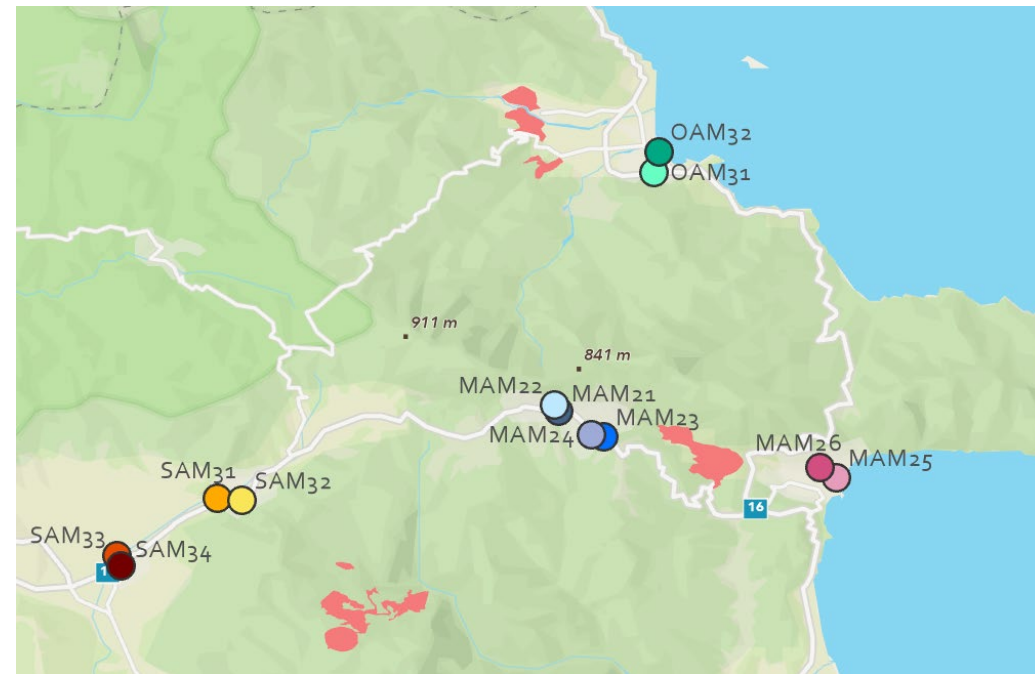
Locality

- MAM21
- MAM22
- MAM23
- MAM24
- MAM25
- MAM26
- OAM31
- OAM32
- SAM31
- SAM32
- SAM33
- SAM34
- mines

Station-Specific Insights: PCA analysis of stations



Station-Specific Insights: IN-SITU monthly means

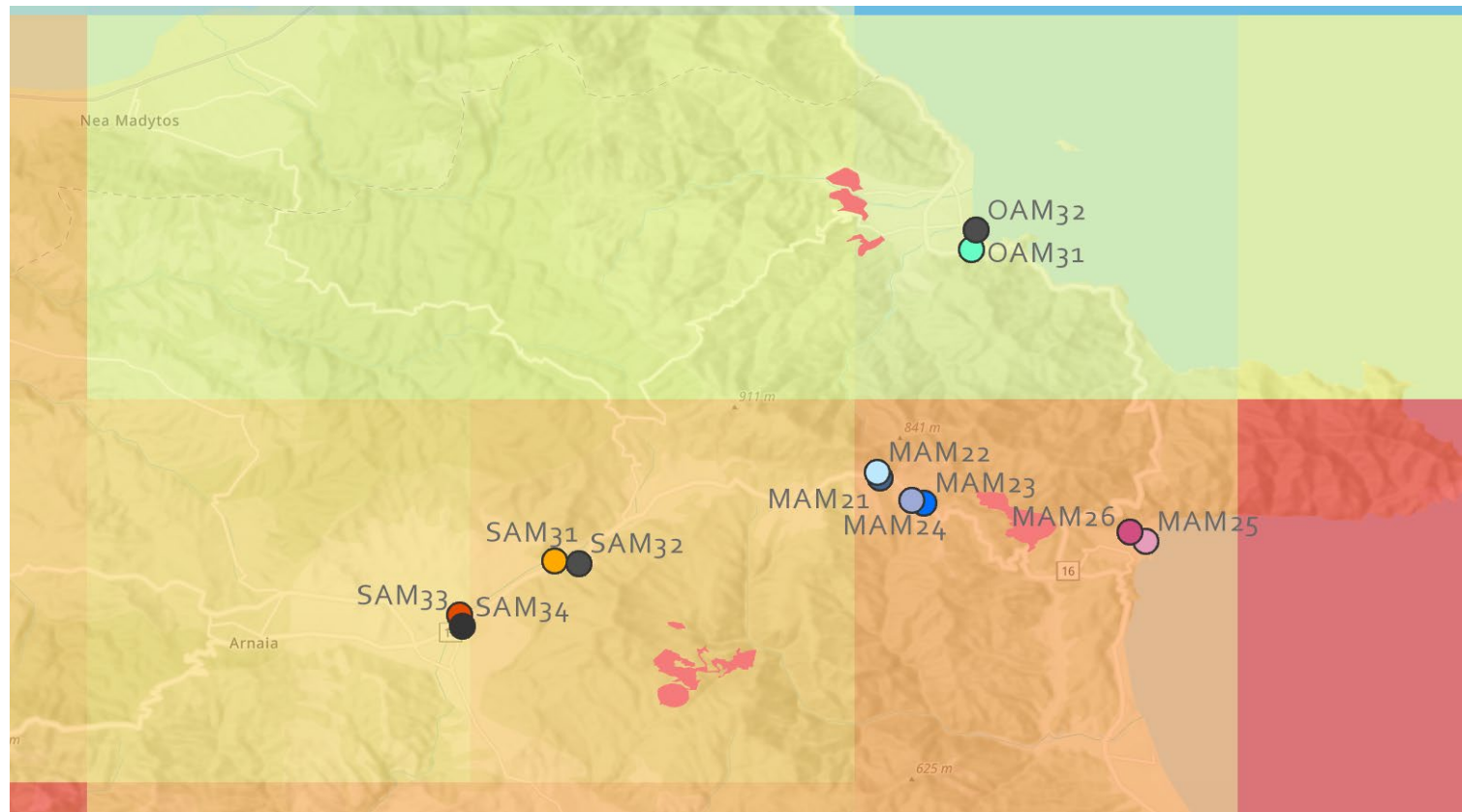


- Extremely high values repeatedly observed at three stations
 - SAM34, SAM32 – winter time extremes due to the stations locations situated in residential area where heating activity results in higher emissions
 - OAM32 – summer time extremes due to the sandy seaside position

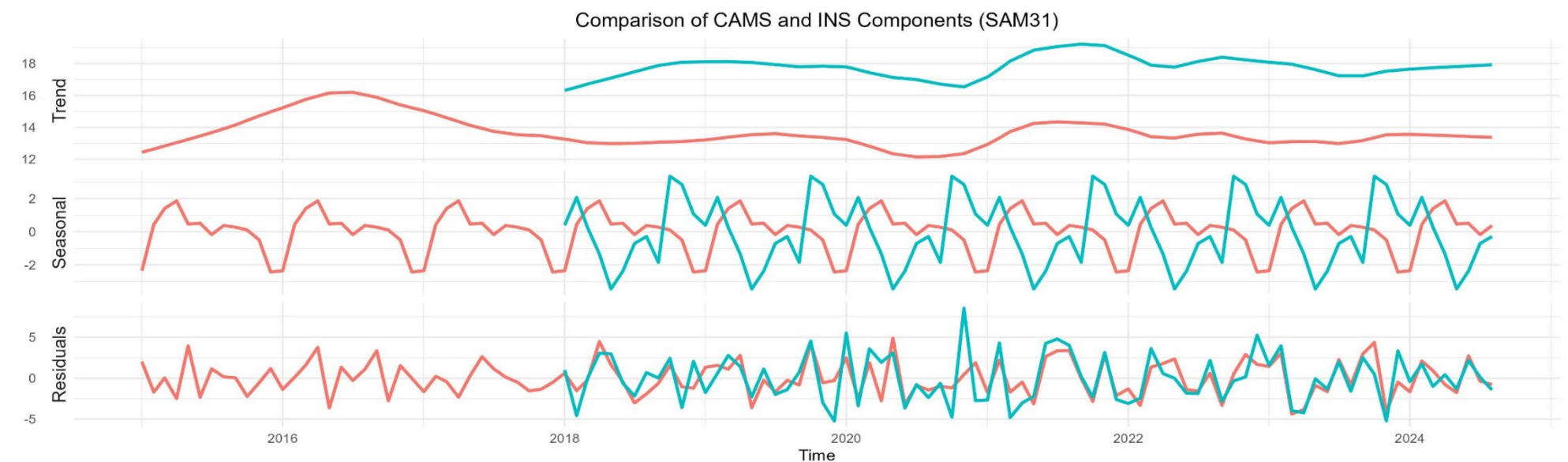
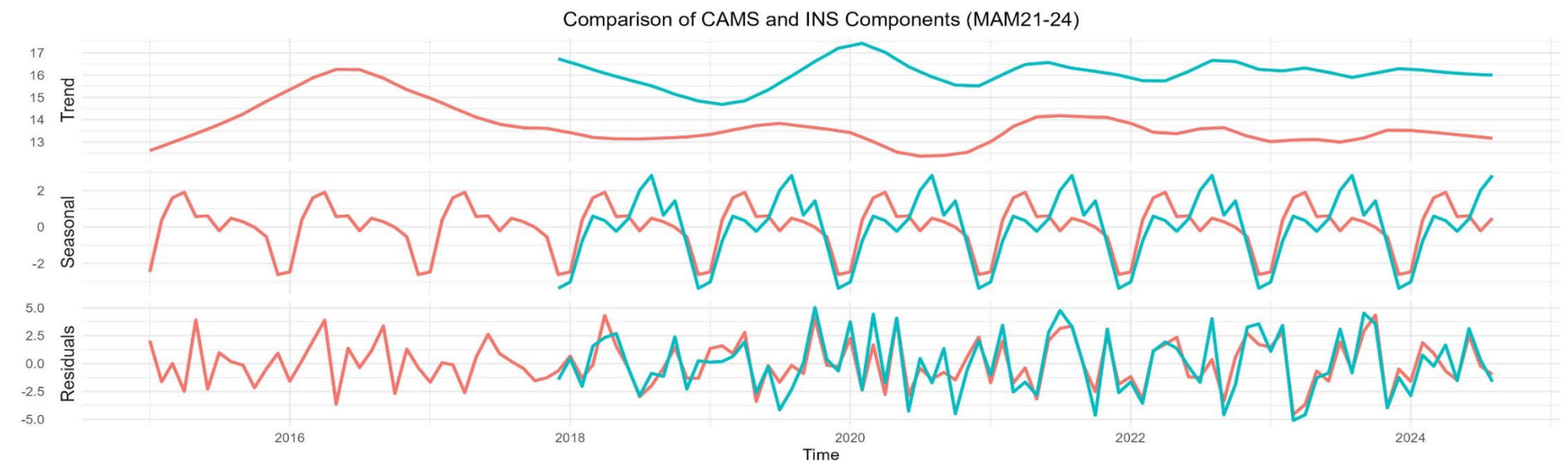
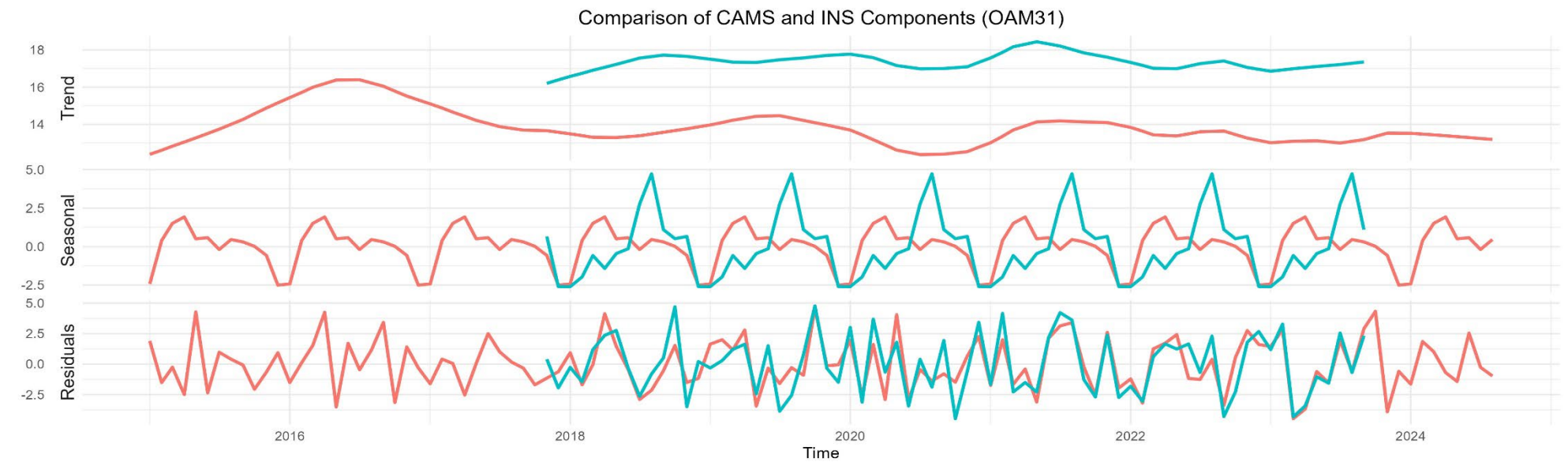
Seasonality – Trends: decompositions



- Seasonal decomposition of monthly CAMS and clustered IN-SITU data



- The correlation between both datasets can be observed in the trend component, with in situ always reaching higher values by 5 – 15 $\mu\text{g}/\text{m}^3$

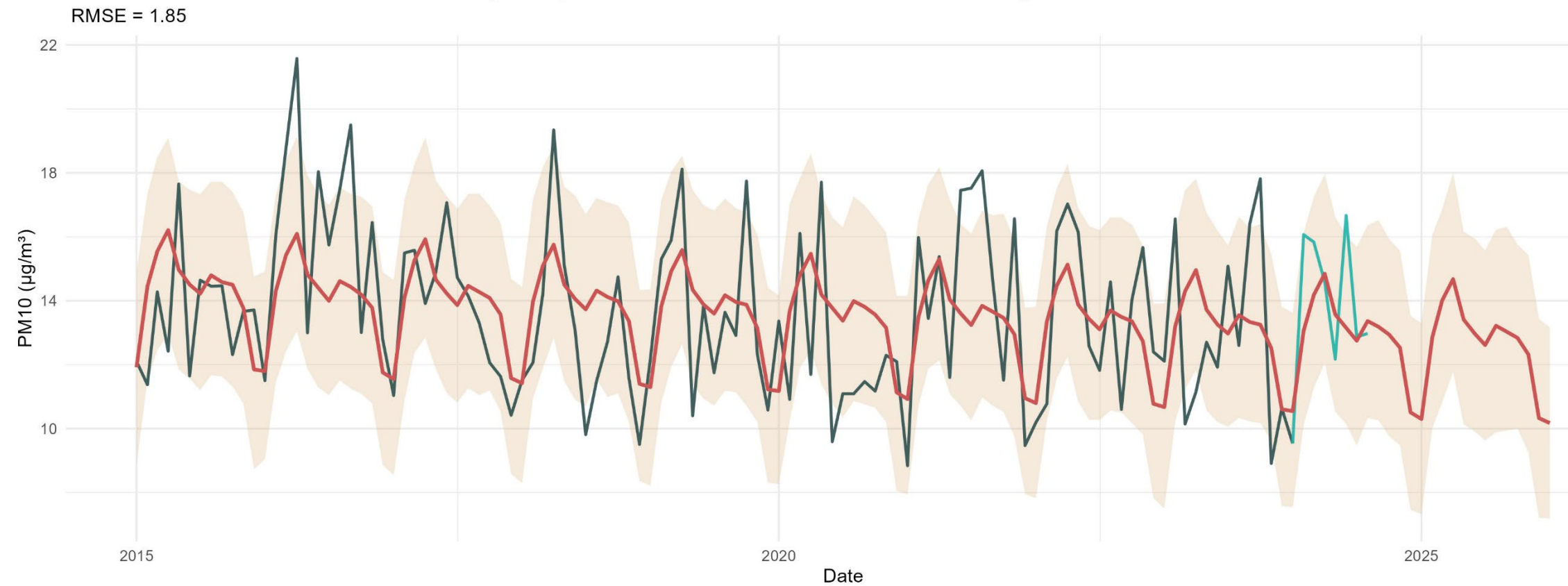


Seasonality – Trends: predictions

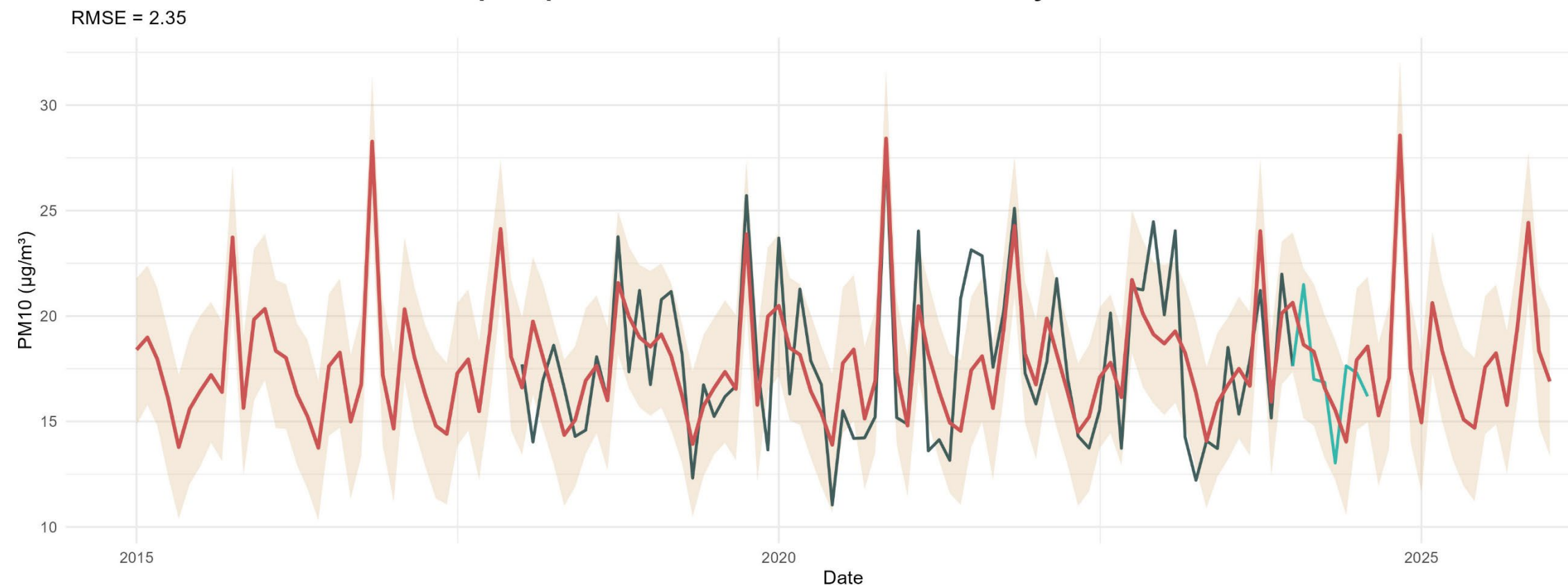


- Predictions of balanced, slightly decreasing trend
- Better results on CAMS dataset
 - $r \approx 0,7$ with $RMSE \approx 1,7 \mu\text{g}/\text{m}^3$
- IN-SITU results
 - $r \approx 0,35$ with $RMSE \approx 3,5 \mu\text{g}/\text{m}^3$

Prophet prediction for CAMS SAM31 locality measurements



Prophet prediction for IN-SITU SAM 31 locality measurements



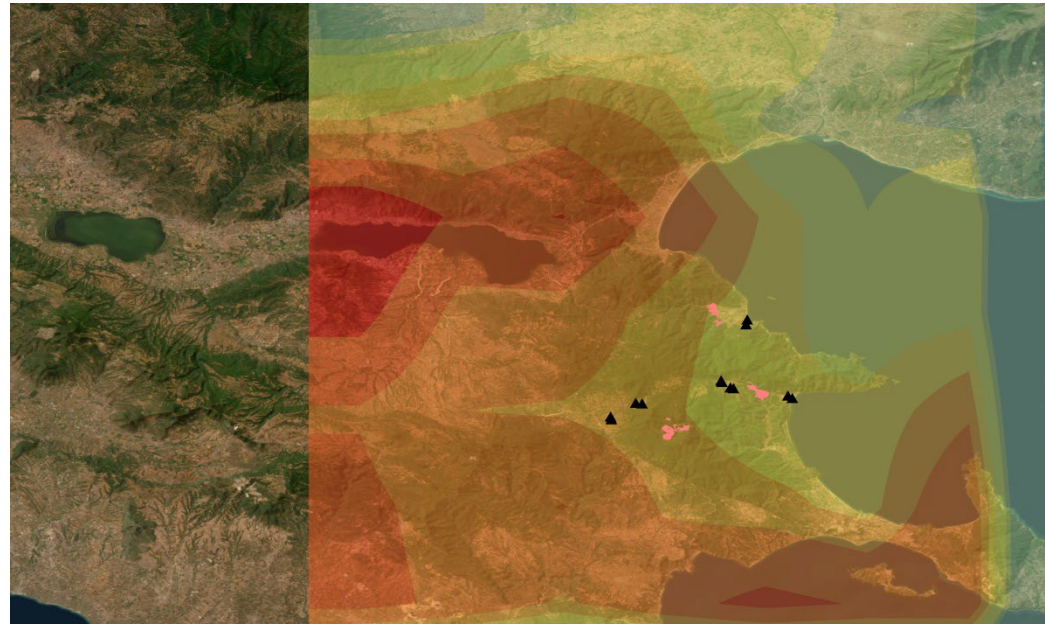
Legend

- Prediction
- Test data
- Train data

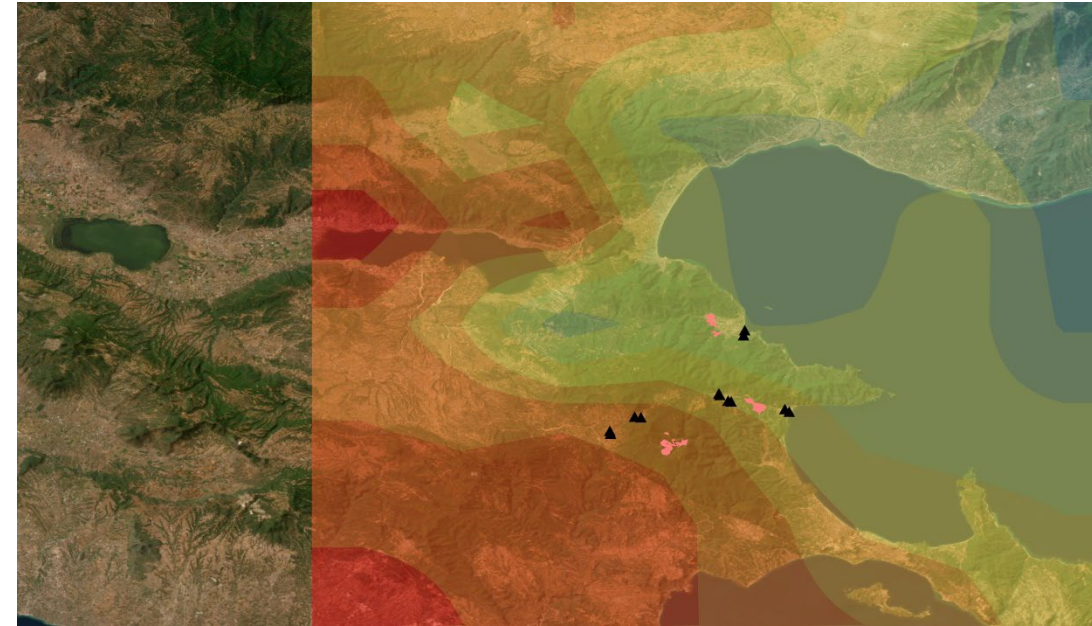
Results: Spatial trends and seasonality



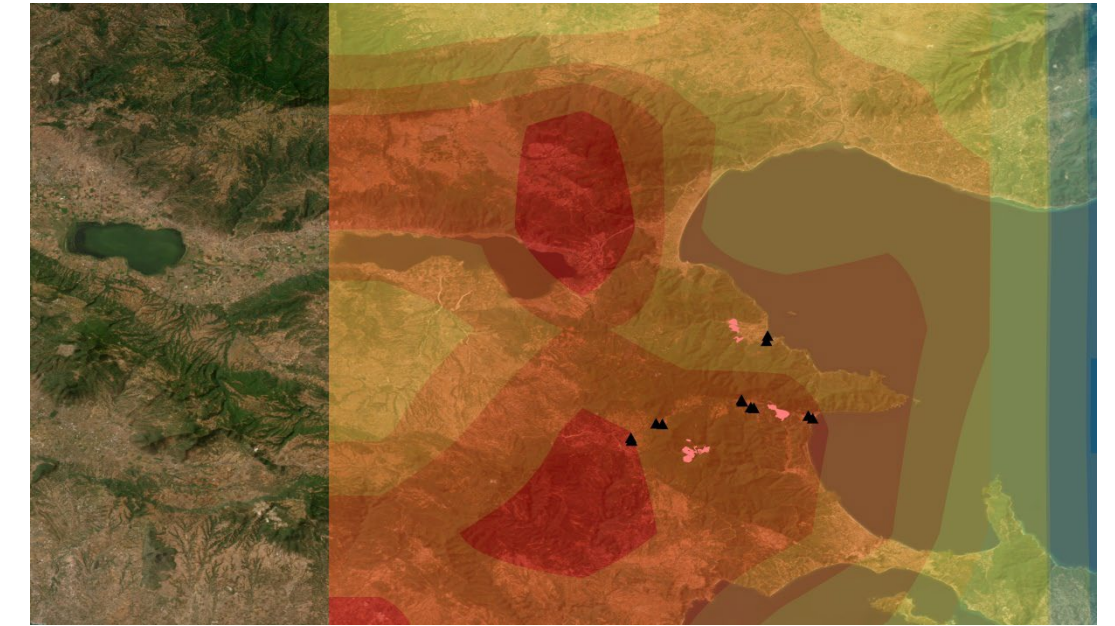
winter season (Nov-April)



2022-2023, 2019-2020

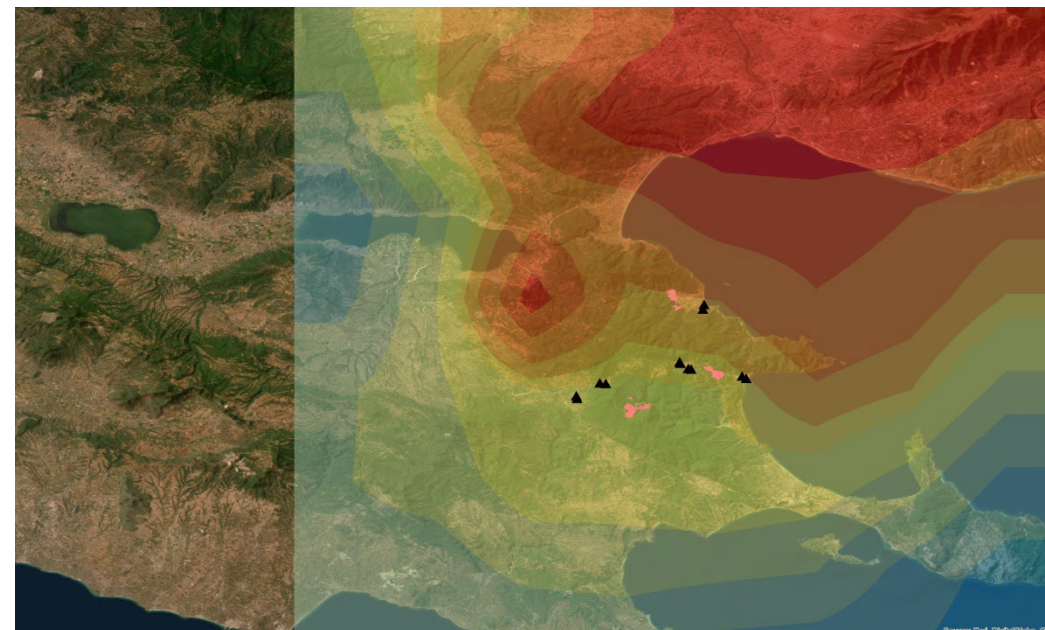


2021-2022, 2020-2021, 2018-2019, 2017-2018

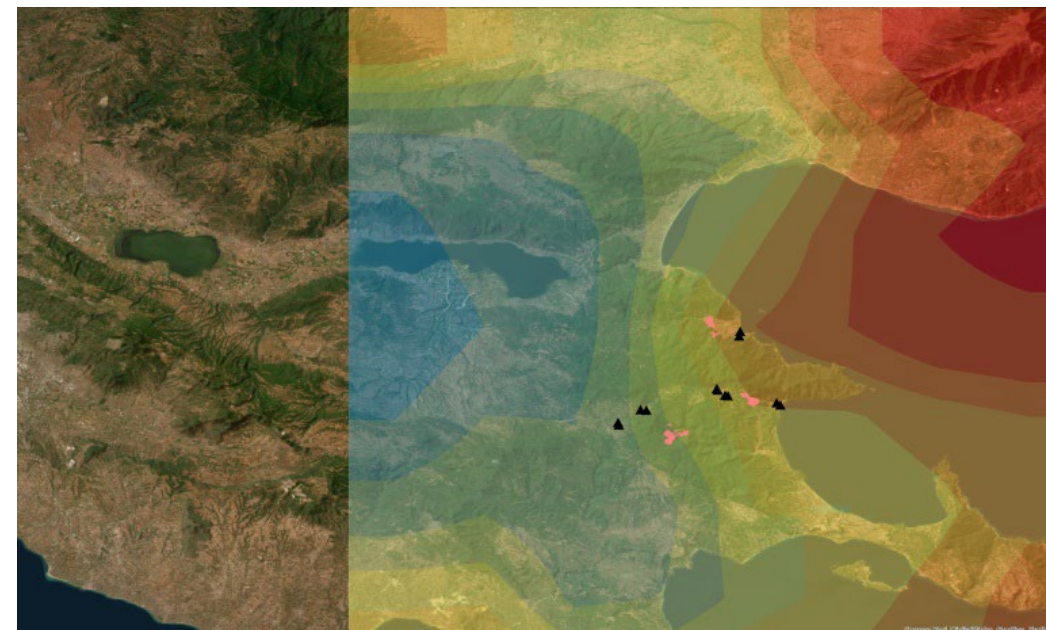


2020-2021, 2015-2016

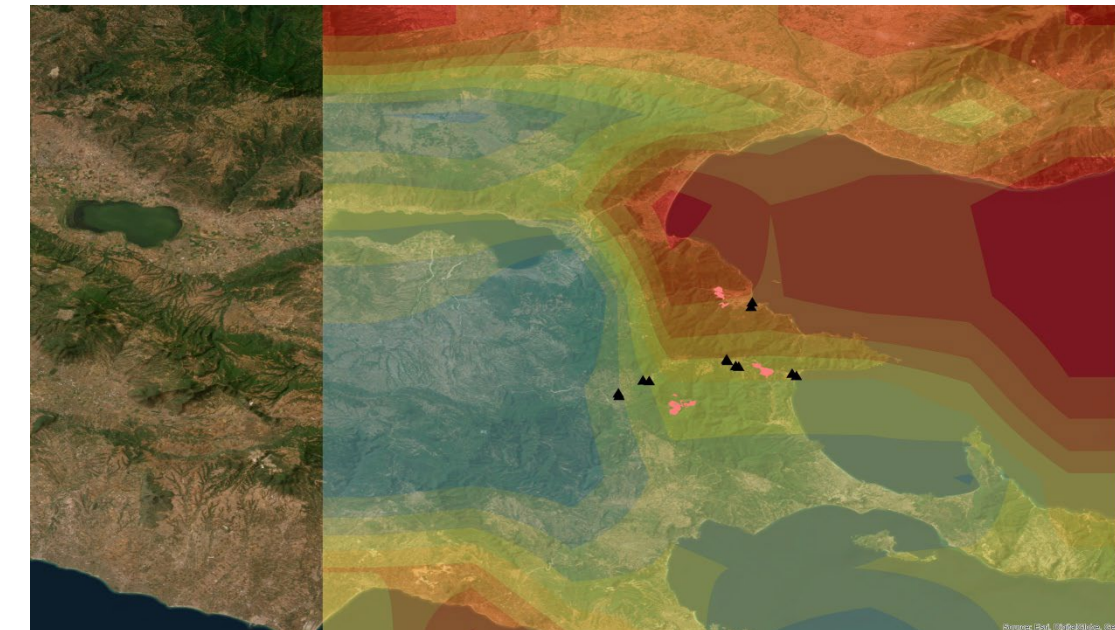
summer season (May-Oct)



2022-2023, 2016-2017



2021-2022, 2015-2016



2020-2021, 2018-2019, 2017-2018



Conclusion – future remarks



Conclusion

- A strong correlation was found between the CAMS dataset and in-situ measurements ($r_s \approx 0,72$), surpassing the correlation seen when using MODIS data ($r_s \approx 0,35$)
- Seasonal patterns and trends effectively identified using IN-SITU time series decomposition for years 2015-2024
- Prophet forecasting demonstrated reliable predictions of seasonal trends and confirmed decreasing trend in local pollution
- PCA revealed distinct station characteristics depending on surrounding environments
- PCA facilitated the visualization of distinct seasonal spatial patterns (winter/summer) in dust distribution

Future remarks

- Focusing on use of more advanced machine learning forecasting models
- Exploring new remote sensing data e.g. Dust RGB SEVIRI product
- Application of this workflow at second locality in Ihalainen, Finland





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Thank you.

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Sources



- Fig.1- Environmental Protection Agency (EPA). (n.d.). *Size comparisons for PM particles*. Retrieved December 3, 2024, from <https://www.epa.gov/pm-pollution/particulate-matter-pm-basics>.
- Fig.2- NASA Earth Observing System Data and Information System (EOSDIS). (2024). *NASA Worldview Snapshots*. Retrieved on December 2, 2024, from [https://worldview.earthdata.nasa.gov/?v=19.90832584072654,38.52225039100108,26.010412438077346,41.73236976090209&l=Reference_Labels_15m\(hidden\),Reference_Features_15m\(hidden\),Coastlines_15m,MODIS_Terra_Aerosol_Optical_Depth_3km,VIIRS_NOAA21_CorrectedReflectance_TrueColor\(hidden\),VIIRS_NOAA20_CorrectedReflectance_TrueColor\(hidden\),VIIRS_SNPP_CorrectedReflectance_TrueColor\(hidden\),MODIS_Aqua_CorrectedReflectance_TrueColor\(hidden\),MODIS_Terra_CorrectedReflectance_TrueColor&lg=true&t=2024-08-29-T07%3A39%3A52Z](https://worldview.earthdata.nasa.gov/?v=19.90832584072654,38.52225039100108,26.010412438077346,41.73236976090209&l=Reference_Labels_15m(hidden),Reference_Features_15m(hidden),Coastlines_15m,MODIS_Terra_Aerosol_Optical_Depth_3km,VIIRS_NOAA21_CorrectedReflectance_TrueColor(hidden),VIIRS_NOAA20_CorrectedReflectance_TrueColor(hidden),VIIRS_SNPP_CorrectedReflectance_TrueColor(hidden),MODIS_Aqua_CorrectedReflectance_TrueColor(hidden),MODIS_Terra_CorrectedReflectance_TrueColor&lg=true&t=2024-08-29-T07%3A39%3A52Z)
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