

Predictive Analytics for Emergency Department Visits Based on Local Short-Term Pollution and Weather Exposure

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The Healthcare Partner





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She received the master's degree in Mathematical Engineering from Politecnico di Milano, Italy in 2023. She is a researcher at GPI's Research & Development department.

Her research interests are machine learning, mathematical modelling and computational techniques.

In her free time, she enjoys travelling and hiking.

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R&D Department: a few current projects

Population Health Management

- Algorithmic Solutions and Virtual Care Technology applied on clinical big data

One Health

- Using AI to predict outbreaks of diseases that can affect humans, animals, and ecosystems, crucial for early warning systems and for preventing or mitigating epidemics and pandemics

Vocal biomarkers analysis

- For diagnosis and pathology monitoring purposes, focusing on fields like neurodegenerative diseases and emotional health

Virtual Care

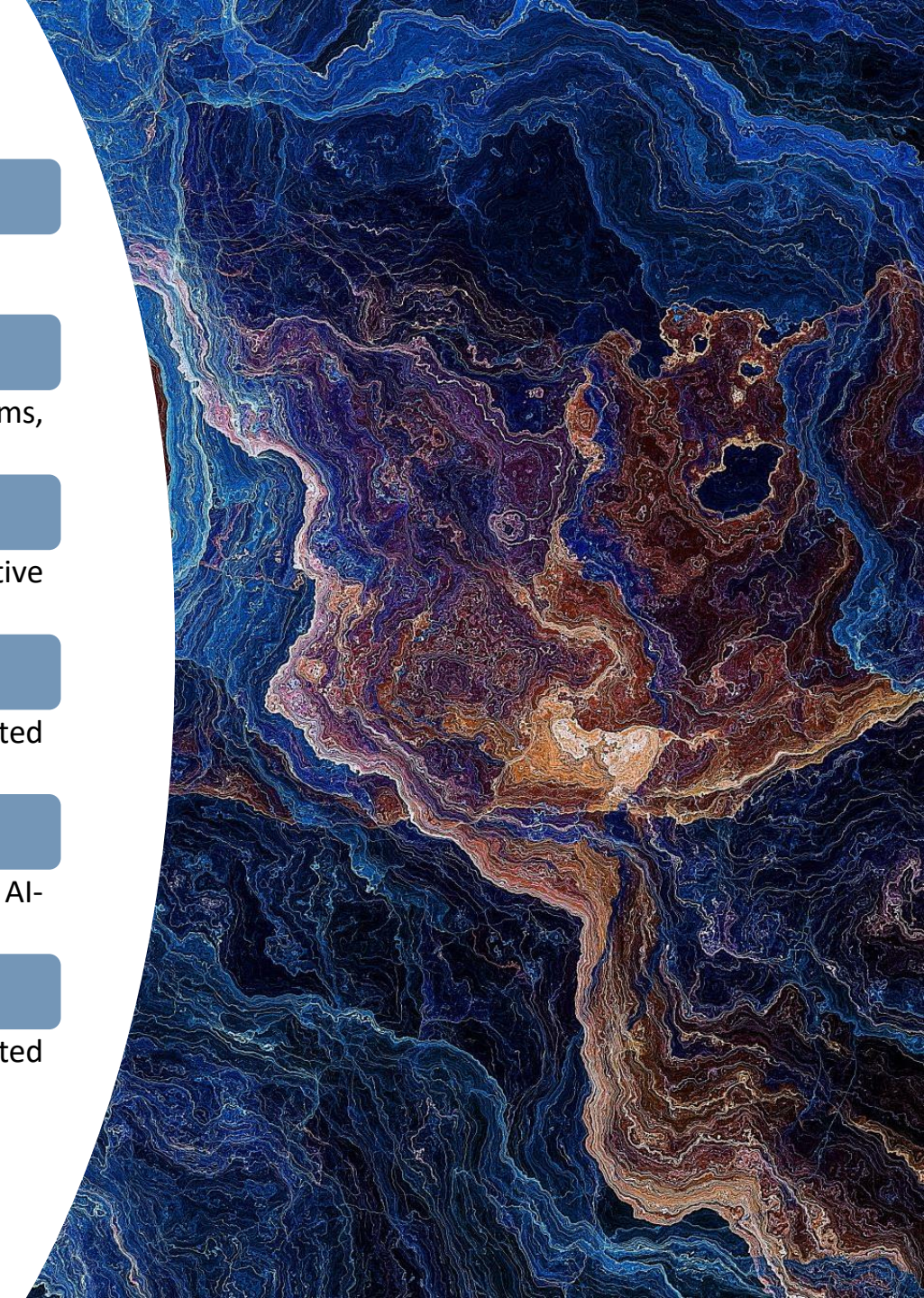
- Telehealth, Teleconsultation, Telemonitoring, Telerehabilitation, Patient Portal, Augmented Telemedicine, Digital Therapeutics, ...

Artificial Intelligence

- Topics include Predictivity in Augmented Epidemiology, NLP, Automatic Pattern Recognition, AI-supported Proactive Medical Centers

Diagnostics, Laboratory and Blood Bank Information Systems

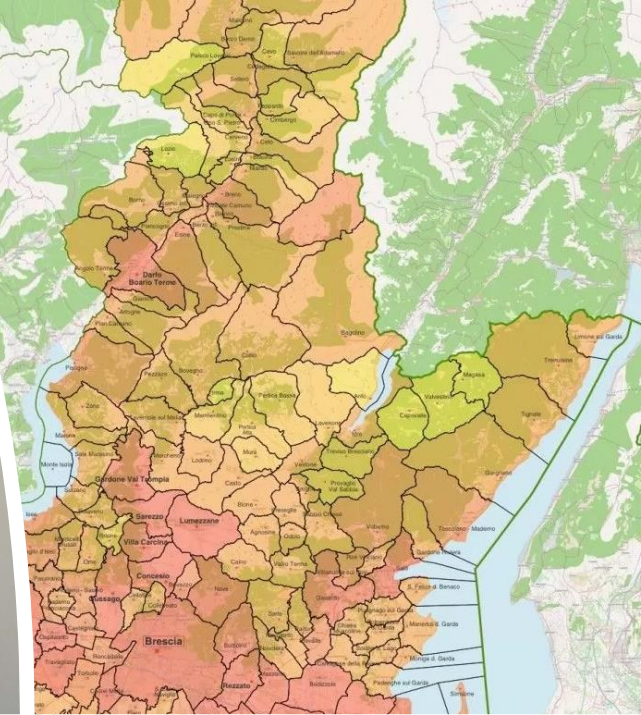
- Application of AI to increase accuracy and efficiency in diagnostics through automated processing of laboratory tests or analysis of blood samples





“Care shouldn’t start in the emergency room”

James Douglas



Our aim

ER historical data



Integration of pollution and weather data



Predictive analysis of visitors' volume based on local environment





Data

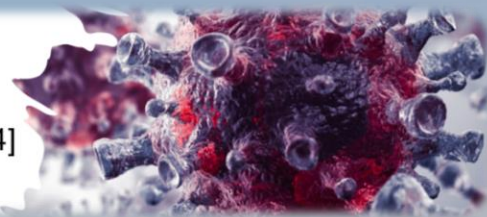
Brescia (2018-2022)

ER accesses

Environment



COVID-19 pandemic
(2020 and 2021):
decrease in ER accesses [23, 24]



Data

Brescia (2018-2022)

ER accesses

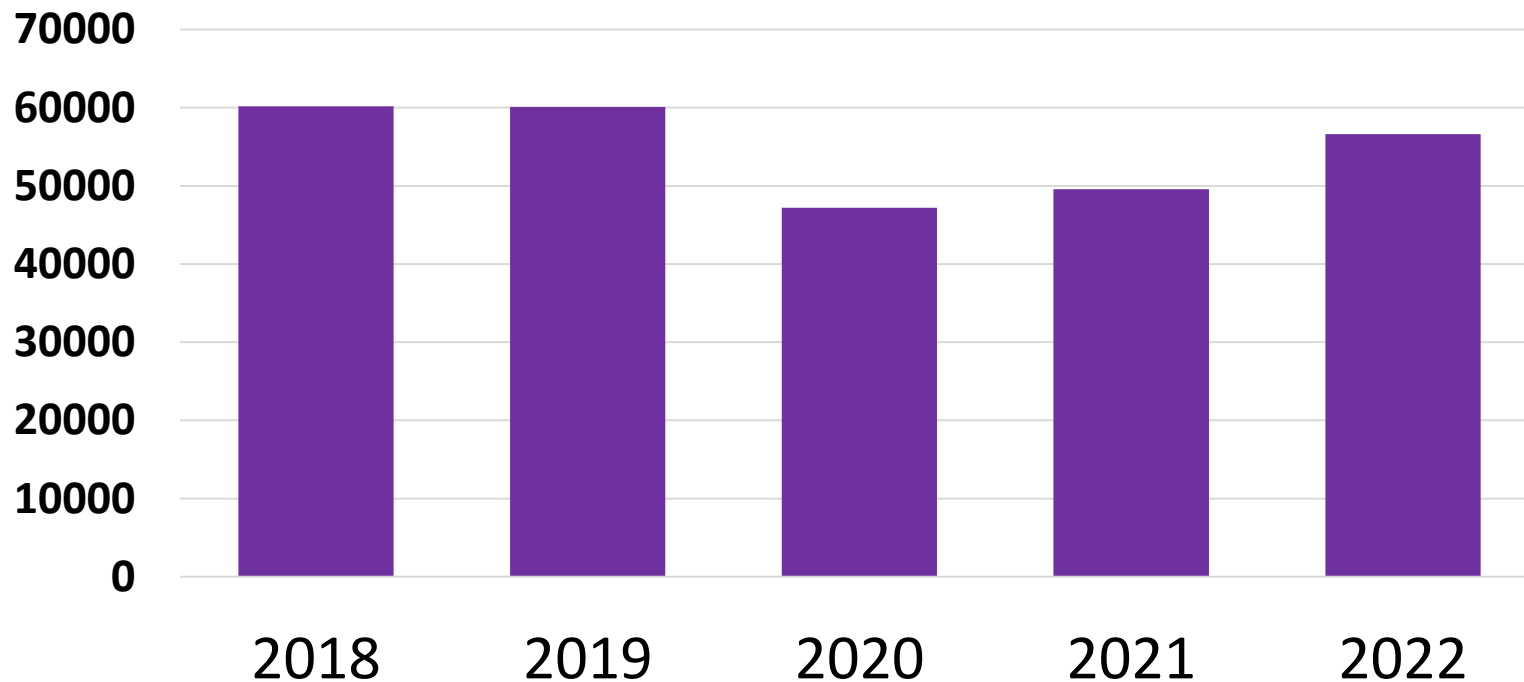
Environment

One of the most polluted
areas in Europe [28]

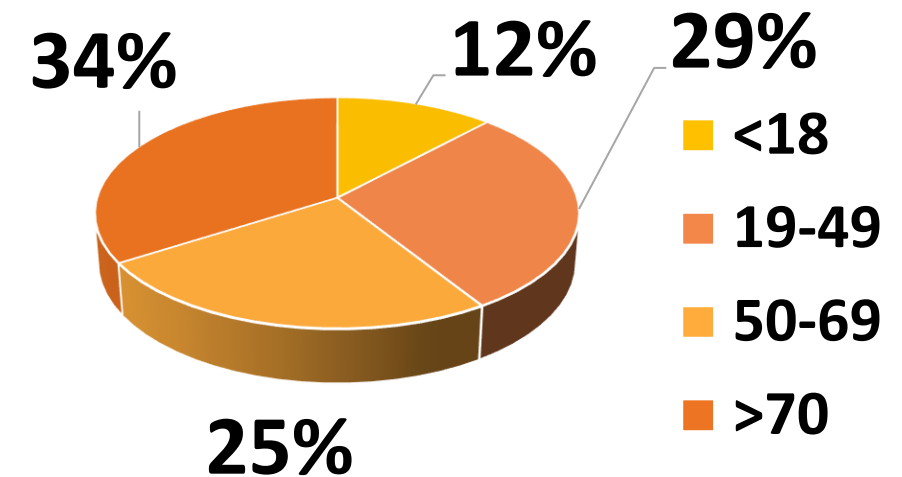


Clinical data

Number of accesses to ER



Patients' age - 2022



After pre-processing:

daily number of accesses or hospitalisations to the ER, limited to those patients coming **only from the city of Brescia** + rolling mean of the daily accesses computed on a seven-day window

Environmental data



Temperature
(T_{\min} and T_{\max} [°C])



Humidity
(RH_{\min} and RH_{\max} [%])



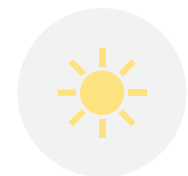
Precipitations
(Prec [mm])



PM₁₀ and PM_{2.5} [$\mu\text{g}/\text{m}^3$]



NO_x, SO₂ and O₃
[$\mu\text{g}/\text{m}^3$]

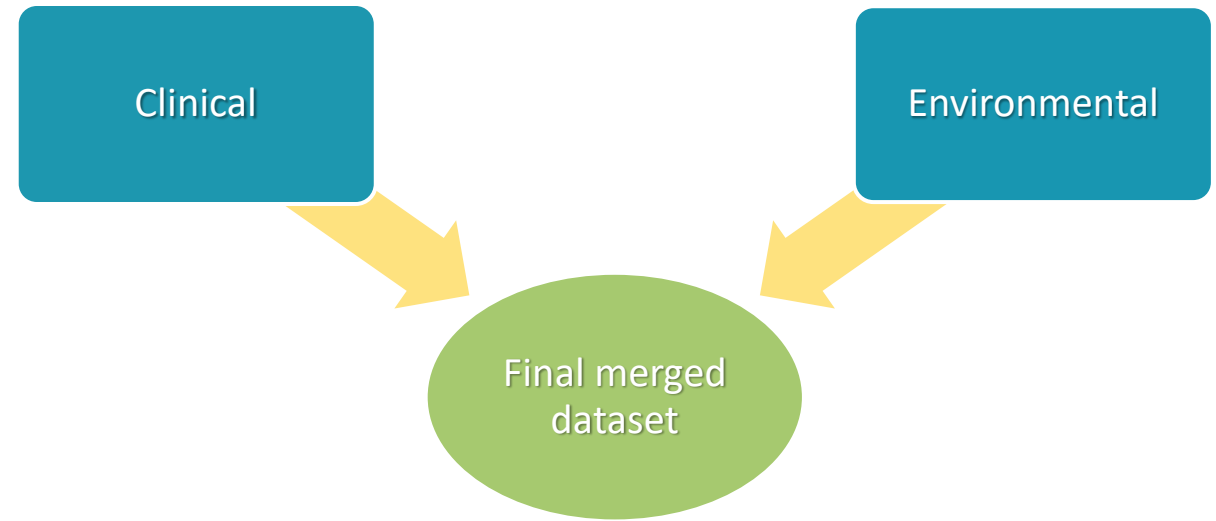


Total solar irradiance
(SSW_{tot} [Wh/m²])

After pre-processing [30]:

each variable has been labelled with the zip code it refers to, so we have all same date data on one row

Case studies - predictive algorithms



A

1) Final pre-processed accesses dataset 2) 1-day and 5-day lagged data 3) Only on the 2 most important features, as computed by the model

B

Analogous to A, but the rolling mean feature was discarded

C

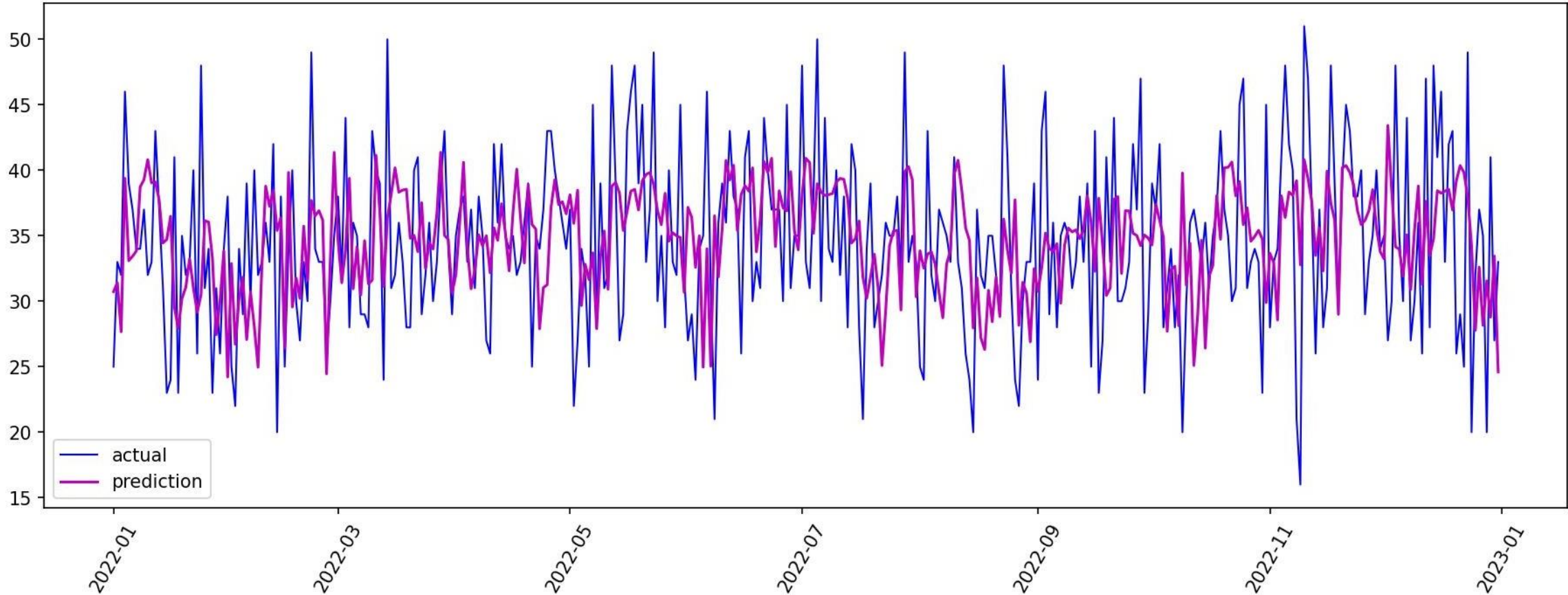
1) 1-day lagged hospitalisation data without rolling mean – cardiovascular patients 2) Only on the 2 most important features

D

Analogous to C, but on data belonging to patients affected by respiratory diseases

Random Forest [29]

MAE = 5.1
Acc = 84.42 %



Case study A2

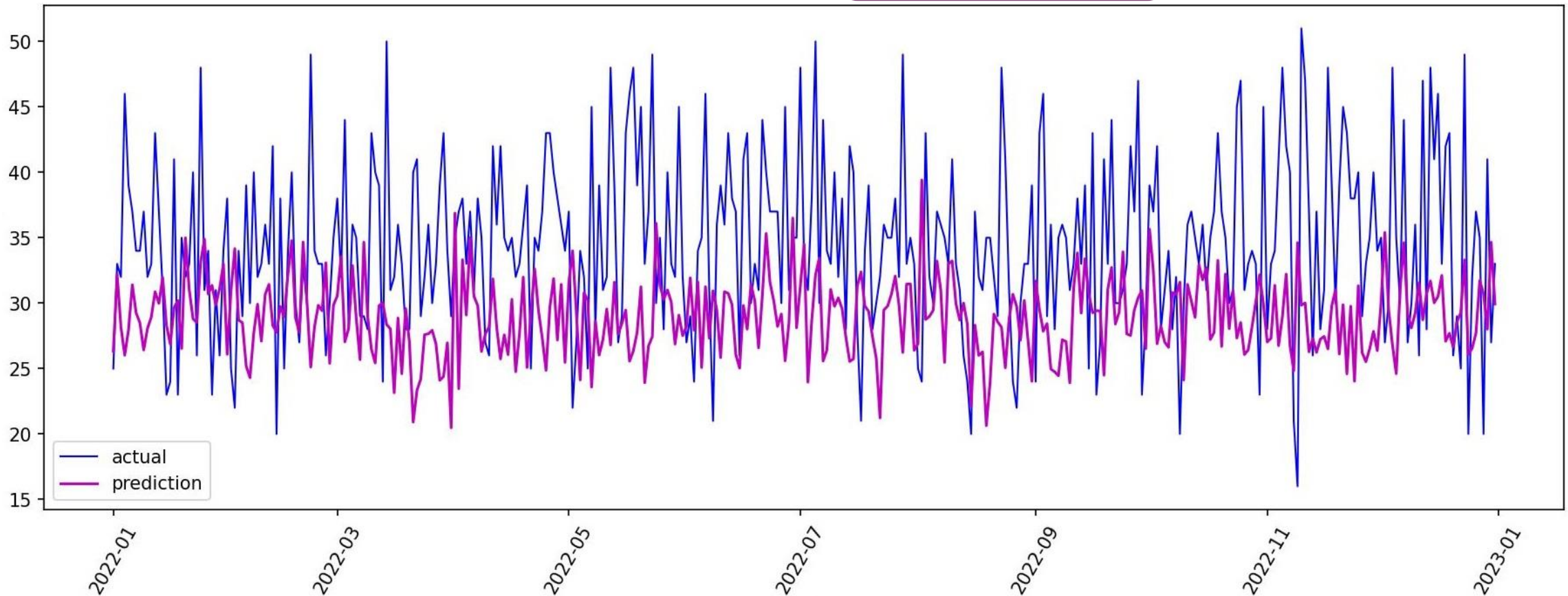
1-day lag



Random Forest

MAE = 6.33

Acc = 82.44 %



Case study B

1-day lag without rolling mean

Case study C1

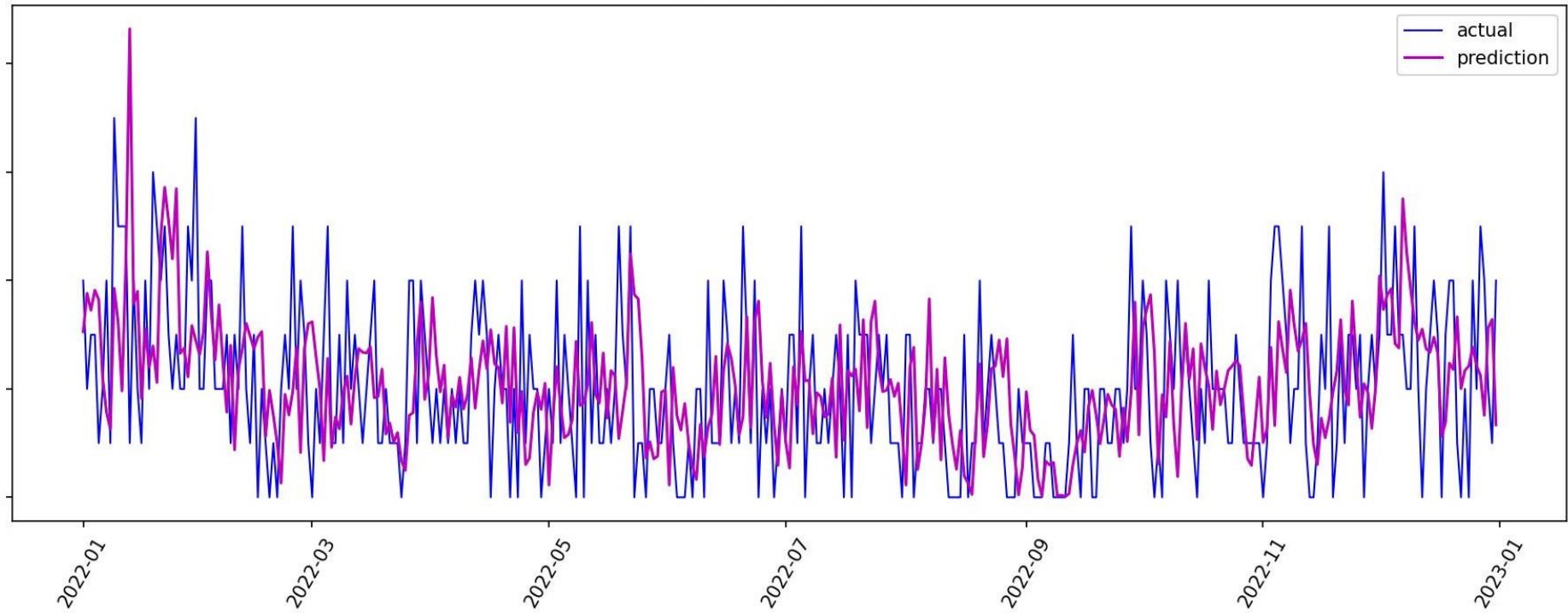
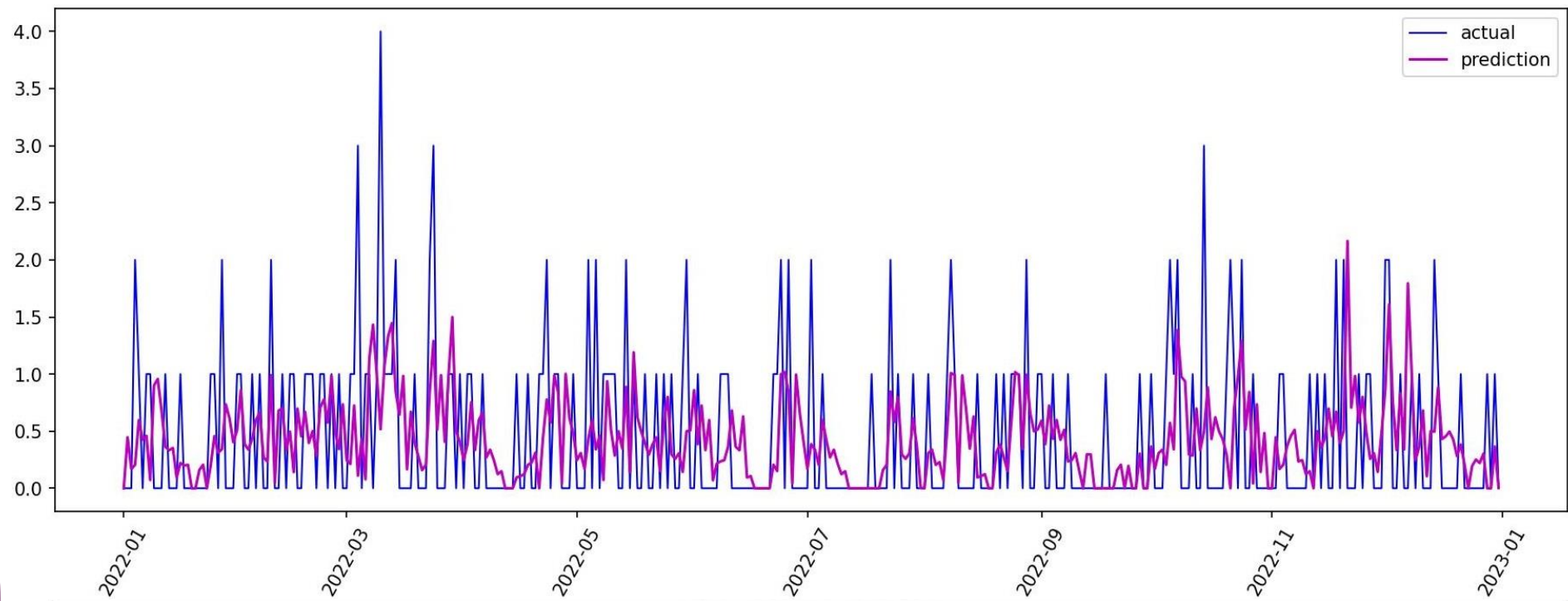


MAE = 0.51

Random Forest

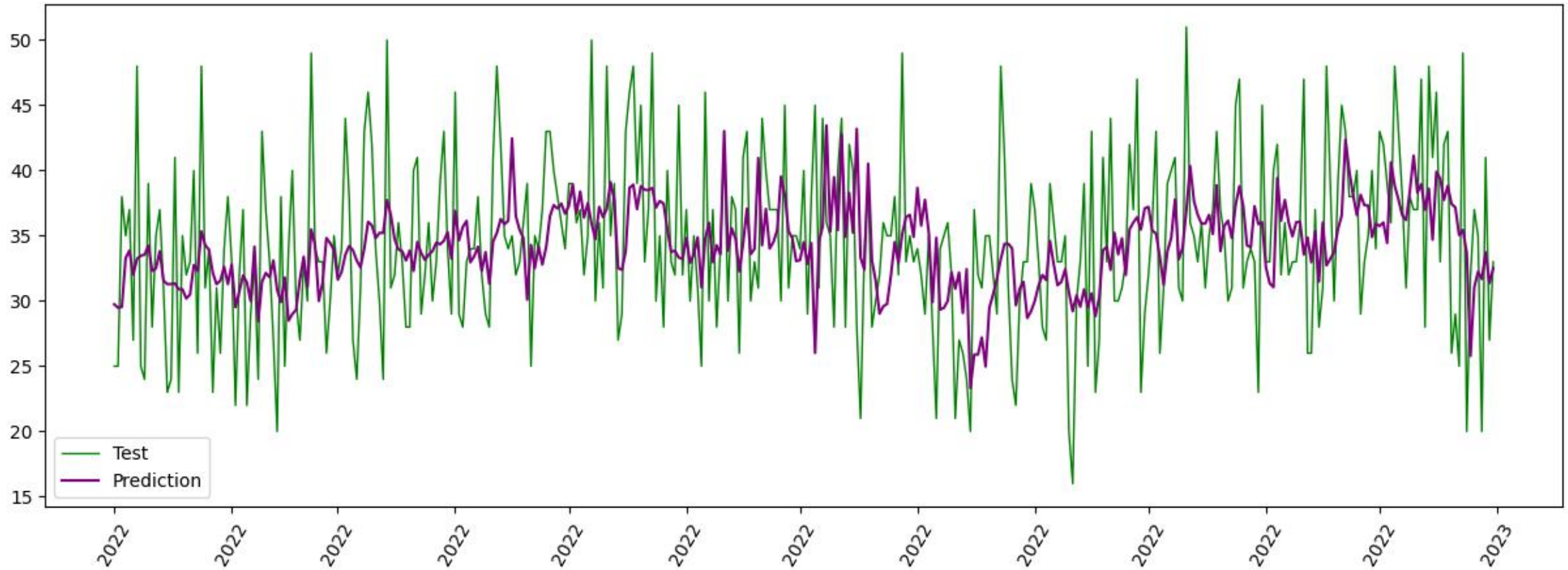
MAE = 1.09

Case study D1



ARIMA [32]

MAPE = 15% MSE = 39.5 RMSE = 6.3



Case study A1 (just the pre-processed final dataset)

Additional remarks

Same-day data and 5-day lagged data: worsening or no improvement in results.

When reducing the dataset to only the two features computed to be the most important ones, RF behaves even less precisely.

Decrease in ER accesses due to COVID-19 in 2020 and 2021: attempt at training only on 2018 and 2019 data, but predictions did not become more precise.

The number of hospitalisations for specific pathologies is limited to a few people every day and, sometimes, even none.

Conclusion and future work

This work is a starting point towards the time-series analysis of historical and environmental data for the prediction of ER accesses and hospitalisations in a specific geographical area.

Results are not optimal but generally promising, even though they cannot be generalised.

Future developments will include data belonging to the entire province of Brescia and test other algorithms.

This may be the offset of a new way of managing ER, monitoring entire populations and geographical areas, enabling a smart real-time predictive analysis able to improve the quality of healthcare and people's quality of life.



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Thank you for your attention,
let's keep in touch!

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Reduced bibliography

[...]

[16] R. D. Brook et al., “Particulate matter air pollution and cardiovascular disease: an update to the scientific statement from the American Heart Association”, *Circulation*, vol. 121, no. 21, pp. 2331–2378, 2010.

[17] F. Dominici et al., “Fine particulate air pollution and hospital admission for cardiovascular and respiratory diseases”, *Jama*, vol. 295, no. 10, pp. 1127–1134, 2006.

[...]

[23] F. Tartari, A. Guglielmo, F. Fuligni, and A. Pileri, “Changes in emergency service access after spread of COVID-19 across Italy”, *Journal of the European Academy of Dermatology and Venereology*, vol. 34, no. 8, p. e350, 2020.

[24] T. Ferrari, C. Zengarini, F. Bardazzi, and A. Pileri, “In-depth, single-centre, analysis of changes in emergency service access after the spread of COVID-19 across Italy”, *Clinical and Experimental Dermatology*, vol. 46, no. 8, pp. 1588–1589, 2021.

[...]

[28] S. Khomenko et al. “Premature mortality due to air pollution in European cities: a health impact assessment”, *The Lancet Planetary Health*, vol. 5, no. 3, pp. e121–e134, 2021

[29] <https://scikit-learn.org/stable/modules/generated/sklearn.ensemble.RandomForestRegressor.html> [Retrieved online: January 2024].

[30] <https://towardsdatascience.com/random-forest-in-python-24d0893d51c0> [Retrieved online: January 2024].

[...]

[32] <https://alkaline-ml.com/pmdarima/modules/generated/pmdarima.arima.AutoArima.html> [Retrieved online: January 2024].

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