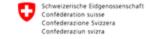
# DIGITAL FINANCE

This project has received funding from the Horizon Europe research and innovation programme under the Marie Skłodowska-Curie Grant Agreement No. 101119635









# explainable AI in Public Employment Services

08.10.2025 - MSCA Training Week on Explainable Al Julius Kooistra - BFH







#### Outline



Introduction to the use case



Modelling choices



System design under real-world constraints



Explainable AI layer







#### Introduction to the use case

- AVA Bern
- Social insurance of financial damages from unemployment
- Help unemployed with:
  - Finding a new job
  - Improving employability









#### Problem statement



Define clusters in their job seekers

Predict Long-Term Unemployment (LTU) for incoming job seekers







#### Data sources

- 1.5 years of data
- 80,000 observations
- Financial administration
- Personal information
- Labour market meassures
- Sanctions imposed







#### Data sources

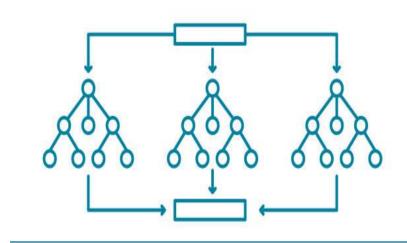
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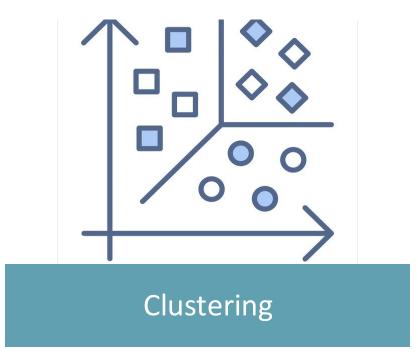




# Modelling choices



Predictive modelling







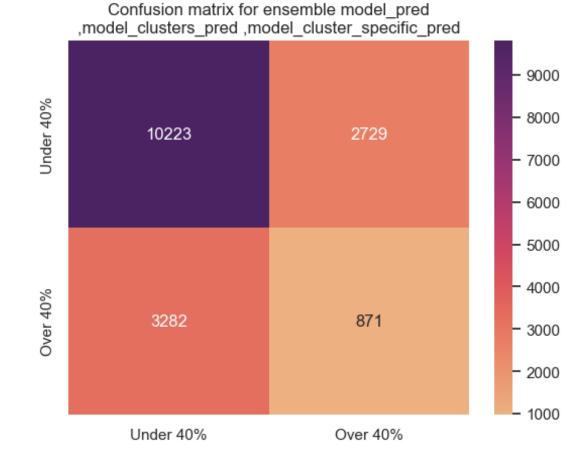


# Predictive modelling

- Binary classification
- Ensemble of different CatBoost configs
- Calibrated

• Acc: 0.749

• F1: 0.560





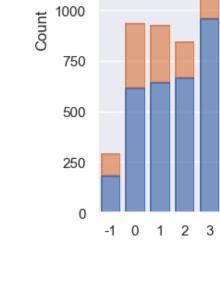




# Clustering

- HDBSCAN
- Clustering of subset of features
- Explain > 80% of variance
- Predicted value

• MSC: 0.580



1750

1500

1250







11 12 13 14 15 16 17 18

Cluster

Prediction

Under 40%

Over 40%

# System design under real-world constraints

- MicroStrategy environment
  - How to fit the ML pipe into the system?
  - Daily ETL pipeline
- New observations change the distance matrix:(
  - Train a supervised classifier based on the clustering output



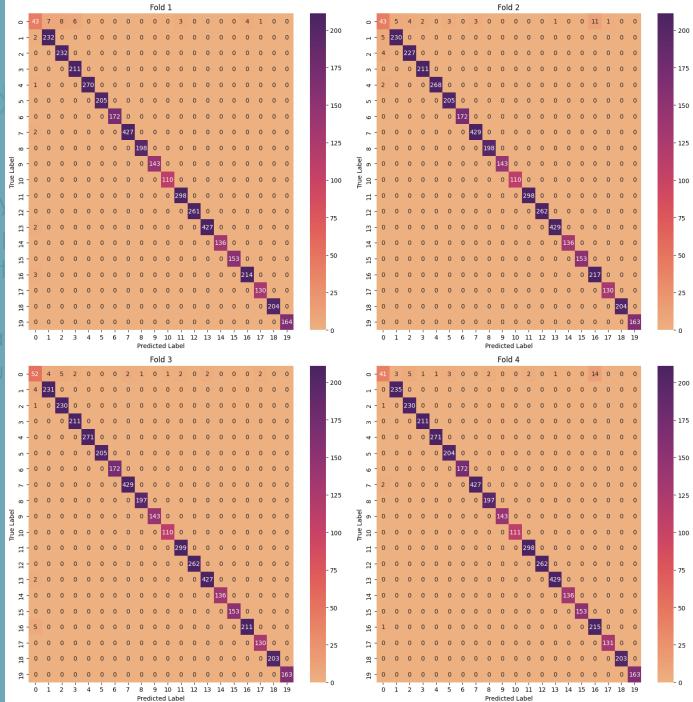






# System c

- MicroStrate
  - Daily ETL
  - How to f
- New observo
  - Train a su







# eXplainable Al layer

- Predictive model
  - Problem: we are using an ensemble
  - Solution: Using the additive property of SHAP, we average the SHAP values of all the models in the ensemble.

$$f(x) = \frac{1}{M} \sum_{m=1}^{M} \phi_{m,0} + \sum_{j=1}^{p} \left( \frac{1}{M} \sum_{m=1}^{M} \phi_{m,j} \right)$$

where *M* is the number of models in the ensemble, and *p* is the number of features

- Clustering:
  - Unexpected benefit of clustering solution: explainable clustering







### XAIs potential

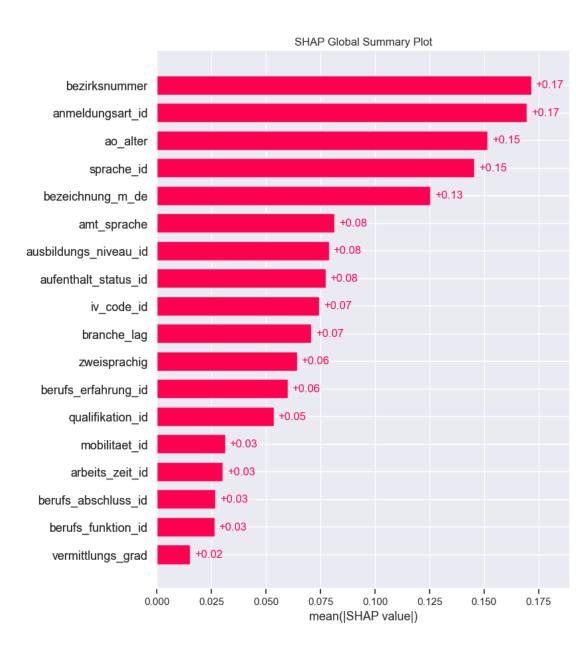
- Graphs
- Generating natural language explanations
- •

Die Modellvorhersage wurde beeinflusst, weil ausbildungs\_niveau\_id (Wert: Sek. I) hat die Vorhersage stark erhöht; anmeldungsart\_id (Wert: Wiederanmeldung laenger 6 Monate) hat die Vorhersage stark erhöht; ...









#### XAI in action @AVA

- Actual usage:
  - Inspecting the model after (re-)training
  - At counsellor level ......





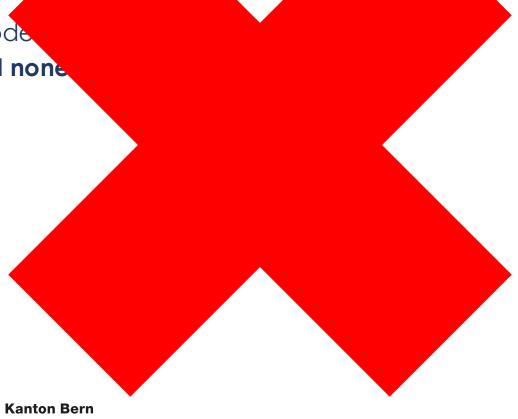


## XAI in action

• Actual usage:

Inspecting the mode

At counsellor level none









# "The end user is **not** a data wizard"

How to present explanations to different end users in a PES context?











Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or Horizon Europe: Marie Skłodowska-Curie Actions. Neither the European Union nor the granting authority can be held responsible for them.

