

National Research University Higher School of Economics Faculty of Computer Science School of Data Analysis and Artificial Intelligence

Neural Networks in Formal Concept Analysis

Subject: Ordered Sets In Data Analysis

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Outline

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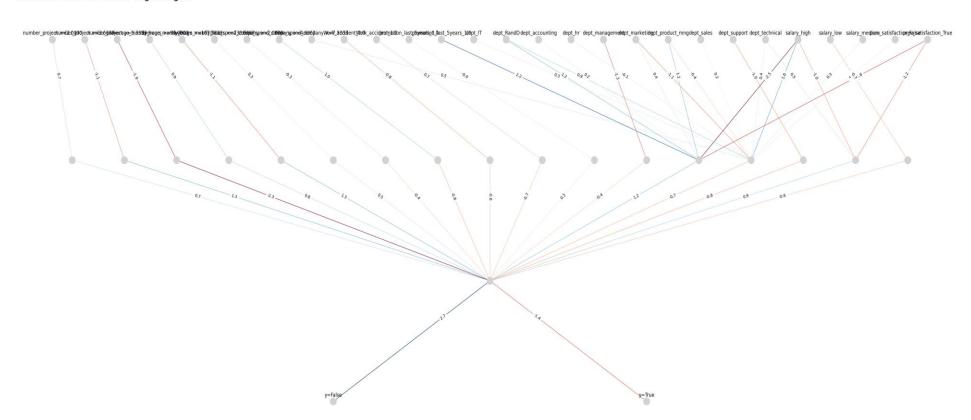
Problem Statement

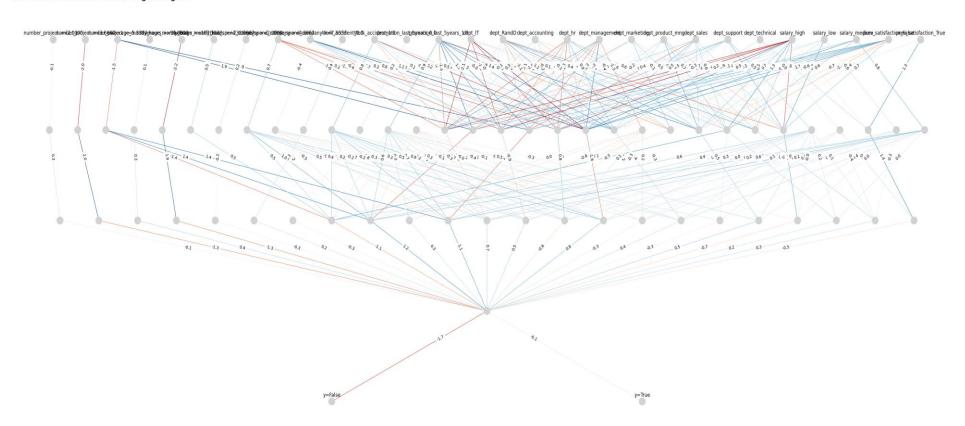
The overarching task of this homework is to implement a merger of machine learning and formal concept analysis. For this, a dataset has to be chosen, its data binarized using scaling (binarization) strategy of choice, and finally the target attribute defined. Then, a comparison between several standard classification methods and NN should be made by calculating performance metrics best suited for the dataset.

First dataset chosen was the **Employee Attrition** dataset. It describes employees' satisfaction with their workplace in a certain company.

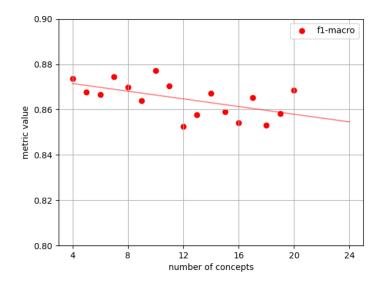
Emp ID	satisfaction_level	last_evaluation	number_project	average_montly _hours	time_spend_co mpany	Work_accident	promotion_last _5years	dept	salary
1	0.38	0.53	2.0	157.0	3.0	0.0	0.0	sales	low
2	0.80	0.86	5.0	262.0	6.0	0.0	0.0	sales	medium
3	0.11	0.88	7.0	272.0	4.0	0.0	0.0	sales	medium
4	0.72	0.87	5.0	223.0	5.0	0.0	0.0	sales	low
5	0.37	0.52	2.0	159.0	3.0	0.0	0.0	sales	low

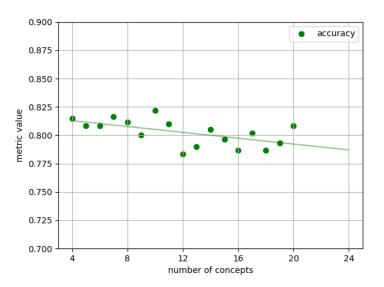
Let's see how the number of concepts used impacts the interpretability of a neural network

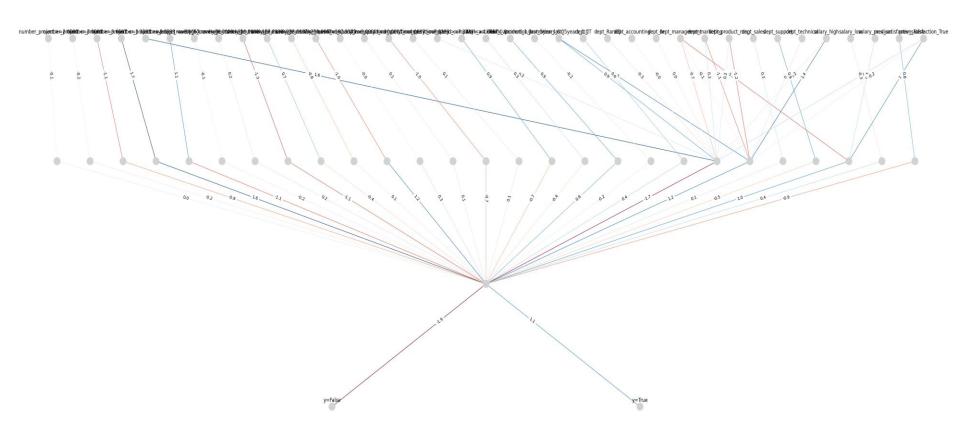


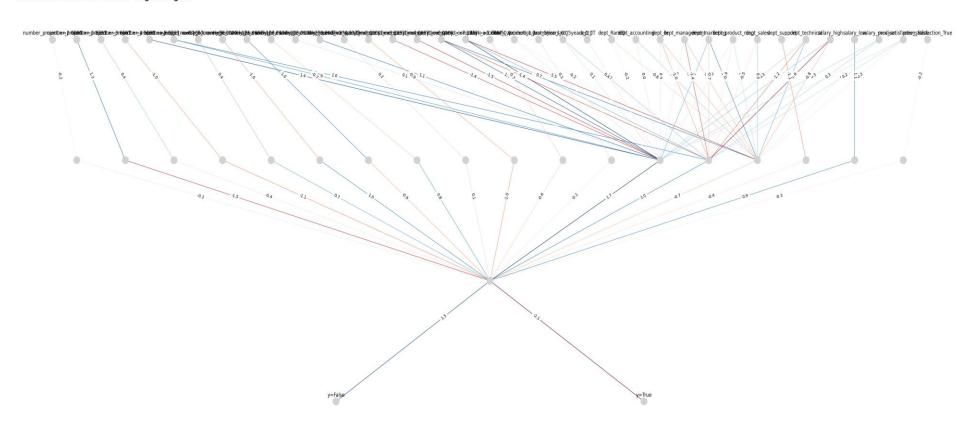


Let's consider the impact of an increase in the number of concepts used on performance metrics too:







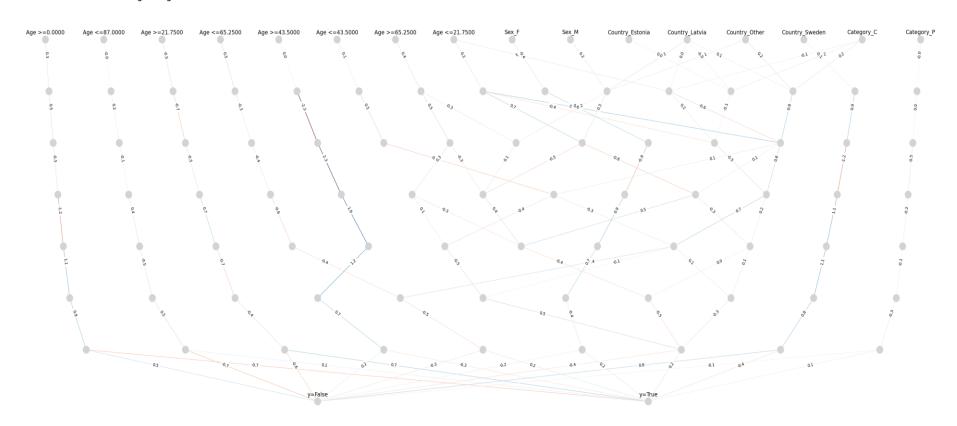


Let's look at the performance metrics of the neural network and several other classifiers on the Employee Attrition dataset and compare them.

Classifier	f1-macro	accuracy
GaussianNB	0.8318	0.7567
RandomForest	0.8889	0.8383
HistGradientBoosting	0.8791	0.8300
NeuralFCA(base)	0.8643	0.8017
NeuralFCA(interordinal)	0.8707	0.8100
NeuralFCA(coarse)	0.9001	0.8550

We should also check how the neural network performs on skewed data. For this task the **Estonia Disaster** dataset was used.

PassengerId	Country	Firstname	Lastname	Sex	Age	Category	Survived
1	Sweden	ARVID KALLE	AADLI	М	62	Р	0
2	Estonia	LEA	AALISTE	F	22	С	0
3	Estonia	AIRI	AAVASTE	F	21	С	0
4	Sweden	JURI	AAVIK	М	53	С	0
5	Sweden	BRITTA ELISABET	AHLSTROM	F	55	Р	0



Results

Coarser binarization produces **better results** in terms of model metrics than both the baseline and binarization using interordinal encoding with similar number of nodes, however this comes **at the cost of interpretability**. Increasing the number of formal concepts used to construct the concept lattice does not improve the model's performance; on the contrary, it seems to correlate with decrease in classification performance. Furthermore, the NN architecture used in this work struggles with overfitting on skewed data as exemplified by the Estonia Disaster dataset.

Conclusion

Neural Networks based on concept lattices perform **on par with several ensemble models** and show promise in some machine learning applications. As exemplified by the Estonia Disaster dataset, these NNs can show connections between various features and their impact on its final decision, hence **improving human interpretability** of their inner workings.

References

1. Kuznetsov, S.O., Makhazhanov, N., Ushakov, M. On Neural Network Architecture Based on Concept Lattices // Foundations of Intelligent Systems. ISMIS 2017. Lecture Notes in Computer Science(), vol 10352, pp. 653-663. Springer, Cham. https://doi.org/10.1007/978-3-319-60438-1_64