Proceedings of the 2nd International Workshop on Explainable and Interpretable Machine Learning (XI-ML) – PREFACE

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1. Introduction

With the current scientific discourse on explainable AI (XAI), algorithmic transparency, interpretability, accountability and finally, explainability of algorithmic models and decisions, the XI-ML workshop on explainable and interpretable machine learning tackles these themes from the modeling and learning perspective. In particular, it targets interpretable methods and models being able to explain themselves and their output, respectively. The workshop aims to provide an interdisciplinary forum to investigate fundamental issues in explainable and interpretable machine learning as well as to discuss recent advances, trends, and challenges.

With this edition of the workshop, we aimed to provide a discussion platform for topics related to XAI. The main emphasis of the call was on approaches such as rule learning and pattern mining for explainable and interpretable machine learning (cf., e.g., [1, 2, 3]). An equally important emphasis was put on cognitive approaches [4, 5] and human concept learning and contrastive explanation [6]. The workshop did not intend to cover only supervised (classification approaches) but also research on XAI in the unsupervised domain such as clustering [7], or subgroup discovery [8]. The third focus area was explanations of black box models as well as applications of all of the above (cf., [9, 10]).

The first edition of the XI-ML (Explainable and Interpretable Machine Learning) workshop was held on September 21, 2020, at the 43rd German Conference on Artificial Intelligence, Bamberg, Germany. This second edition of the workshop was held on September 20, 2022, in co-location with the 45th German Conference on Artificial Intelligence, Trier, Germany (held virtually).

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For the workshop, there were six accepted papers out of eight submissions in total. Below, we structure these according to the general topics of *explanation & methods* as well as *explanation & applications* with their specific methodological foci in the context of explainable and interpretable machine learning, as well as individual applications, respectively.

- Explanation & Methods:
 - Evolutionary Counterfactual Visual Explanation by Jacqueline Höllig, Steffen Thoma and Cedric Kulbach. The paper proposes a genetic algorithm for the important problem of explaining image classification models using counterfactuals.
 - Self-explaining variational Gaussian Processes for transparency and modelling of prior knowledge by Sarem Seitz. This paper presents an approach for combining Gaussian Processes with recent algorithms proposed in XAI to study feature importance.
 - Study on Criteria for Explainable AI for Laypeople by Thorsten Zylowski. This paper presented a survey aimed at identifying criteria important for the development of XAI solutions for the public. The results of this survey show that explanations for AI decisions are widely sought for. The authors suggest that global explanations are instrumental in building initial user trust, while local explanations probably contribute to maintaining trust.
- Explanation & Application:
 - Imitation Learning of Logical Program Policies for Multi-Agent Reinforcement Learning by Manuel Eberhardinger, Johannes Maucher and Setareh Maghsudi. This paper deals with making decisions in Multi-agent systems interpretable by using using the Logical Program Policies, a recently proposed approach originally developed for few-shot Bayesian imitation learning [11].
 - Mining Interesting Outlier Subgraphs In Attributed Graphs by Ahmad Mel and Tijl De Bie. The paper presents an approach for identifying interesting outlier subgraphs in attributed graphs. The algorithm searches for k interesting anomalous subgraphs by scanning the space of candidates using a branch-and-bound beam search procedure. The topic is well-motivated, the approach described algorithmically and also exemplified.
 - Multi-Perspective Anomaly Detection on Bipartite Multi-Layer Social Interaction Networks by Asep Maulana and Martin Atzmueller. This paper focuses on anomaly detection in complex networks, specifically bipartite multi-layer networks for modeling multiple relations. It presents an interactive approach for characterizing anomalies from different explanatory perspectives.

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