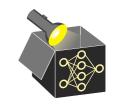
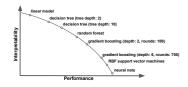
Interpretable Machine Learning Intro to IML Introduction, Motivation, and History



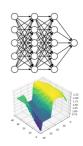


Learning goals

- Why interpretability?
- Developments until now?
- Use cases for interpretability

WHY INTERPRETABILITY?

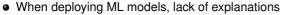
- ML: huge potential to aid decision-making process due to its predictive performance
- ML models are black boxes, e.g., XGBoost, RBF SVM or DNNs
 - → too complex to be understood by humans
- Some applications are "learn to understand"



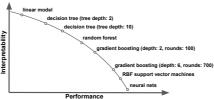


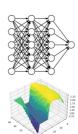
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- hurts trust
- creates barriers
- Many disciplines with required trust rely on traditional models, e.g., linear models, with less predictive performance







INTERPRETABILITY IN HIGH-STAKES DECISIONS

Examples of critical areas where decisions based on ML models can affect human life

- Credit scoring and insurance applications
 "Society of Actuaries" 2021
 - Reasons for not granting a loan
 - Fraud detection in insurance claims



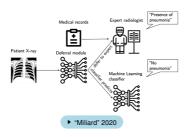


INTERPRETABILITY IN HIGH-STAKES DECISIONS

Examples of critical areas where decisions based on ML models can affect human life

- Credit scoring and insurance applications
 "Society of Actuaries" 2021
 - Reasons for not granting a loan
 - Fraud detection in insurance claims
- Medical applications
 - Identification of diseases
 - Recommendations of treatments
- •







NEED FOR INTERPRETABILITY

Need for interpretability becoming increasingly important from a legal perspective

- General Data Protection Regulation (GDPR) requires for some applications that models have to be explainable
 → "Goodman and Flaxman" 2017
 → EU Regulations on Algorithmic Decision-Making and a "Right to Explanation"
- Ethics guidelines for trustworthy AI → "European Commission" 2019





BRIEF HISTORY OF INTERPRETABILITY

- 18th and 19th century:
 Lin. regression models (Gauss, Legendre, Quetelet)
- 1940s: Emergence of sensitivity analysis (SA)
- Middle of 20th century:
 Rule-based ML, incl. decision rules and trees
- 2001: Built-in feature imp. measure of random forests
- >2010: Explainable AI (XAI) for deep learning
- >2015: IML as an independent field of research







