

General Situation

- A sample of partial orders (e.g. compare the performance of ml algorithms on a fixed data set w.r.t multiple performance measures at once).
 - Each partial order is indeed a precise observation. Thus, items are allowed to be incomparable.
- ⇒ Partial orders as a special case of *non-standard data*.

Contribution

- Introducing the ufg depth function (gives a center-outward order) to obtain a description of the distribution of partial orders (\mathcal{P}) based on a sample.
- Explicitly addressing incomparability in the data description.

Idea: Adaptation of the Simplicial Depth

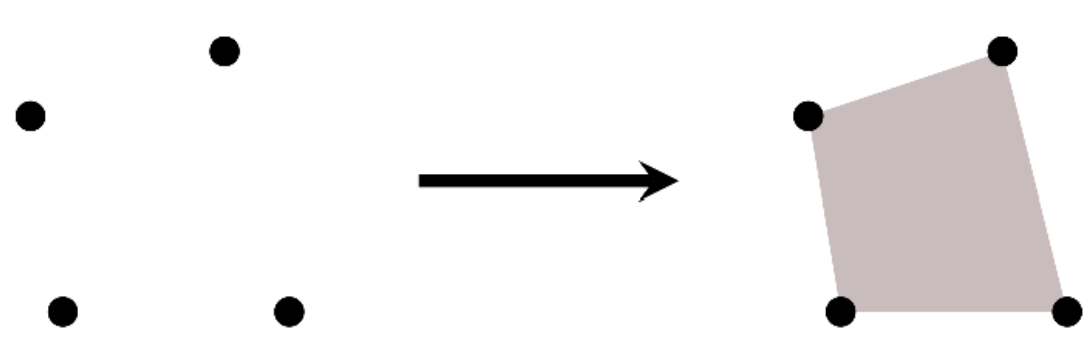
Depth functions measure the centrality and outlyingness of a data point with respect to a data cloud or an underlying distribution.

Simplicial Depth

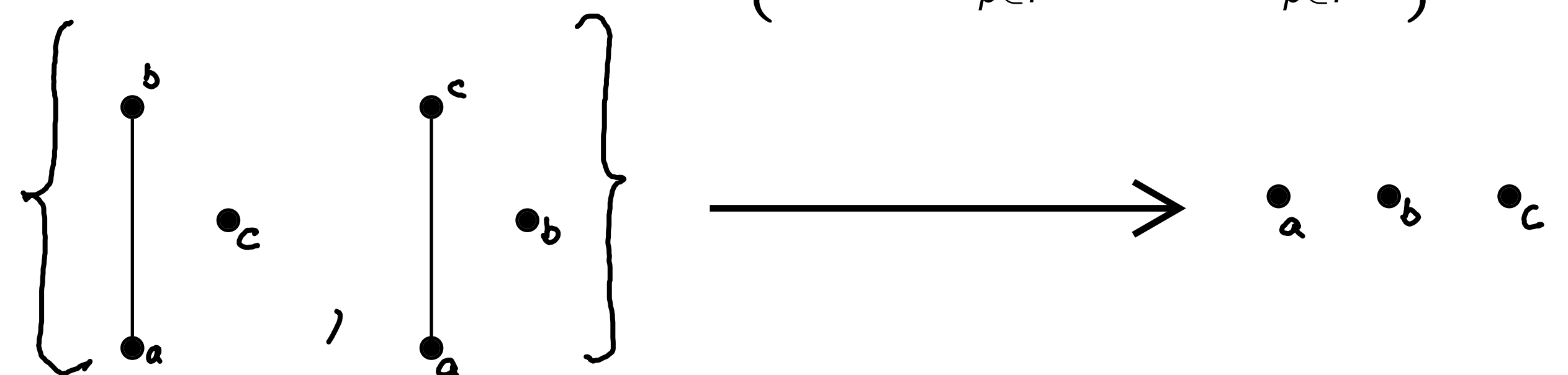
Union-Free Generic Depth

Define the Closure Operator/System

$$\gamma_{\mathbb{R}^d}: 2^{\mathbb{R}^d} \rightarrow 2^{\mathbb{R}^d}; A \mapsto \left\{ x \in \mathbb{R}^d \mid \begin{array}{l} x = \sum_{i=1}^k \lambda_i a_i \text{ with } a_i \in A, \\ \lambda_i \in [0, 1], \sum_{i=1}^k \lambda_i = 1, k \in \mathbb{N} \end{array} \right\}$$

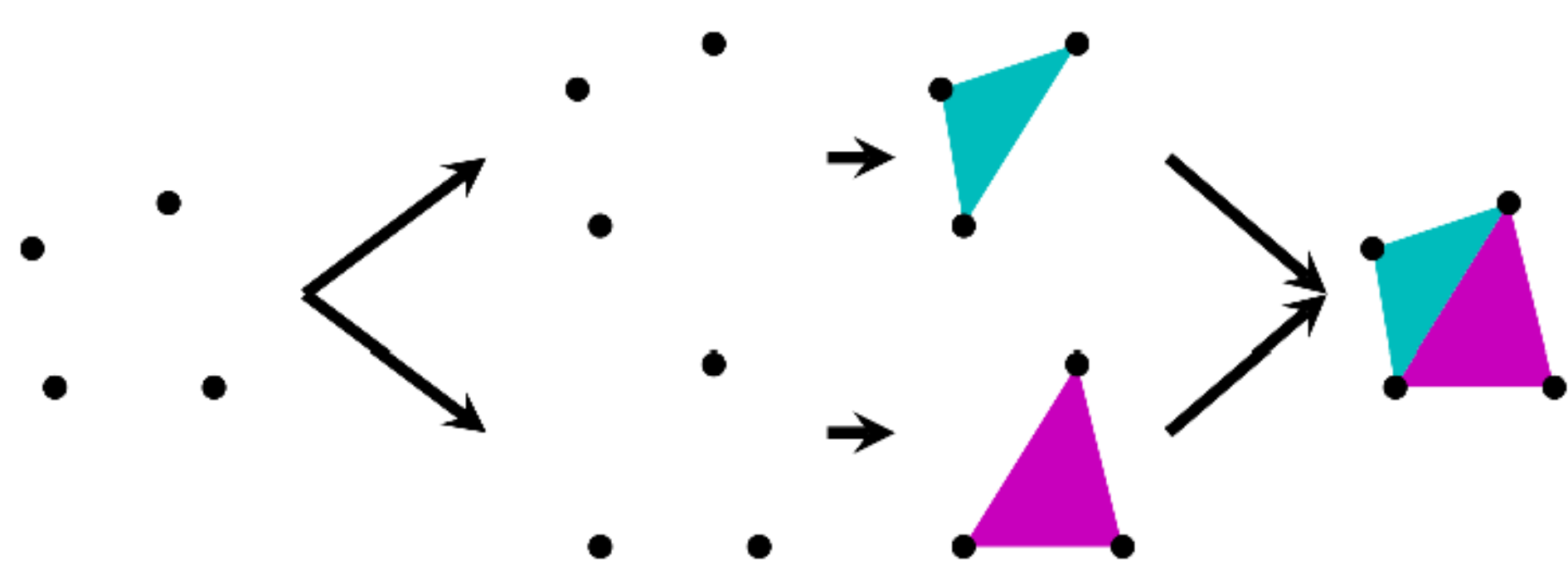


$$\gamma: 2^{\mathcal{P}} \rightarrow 2^{\mathcal{P}}; P \mapsto \left\{ p \in \mathcal{P} \mid \bigcap_{\tilde{p} \in P} \tilde{p} \subseteq p \subseteq \bigcup_{\tilde{p} \in P} \tilde{p} \right\}$$



Reduce the Input Set (It is Still Sufficient to Describe the Closure Operator)

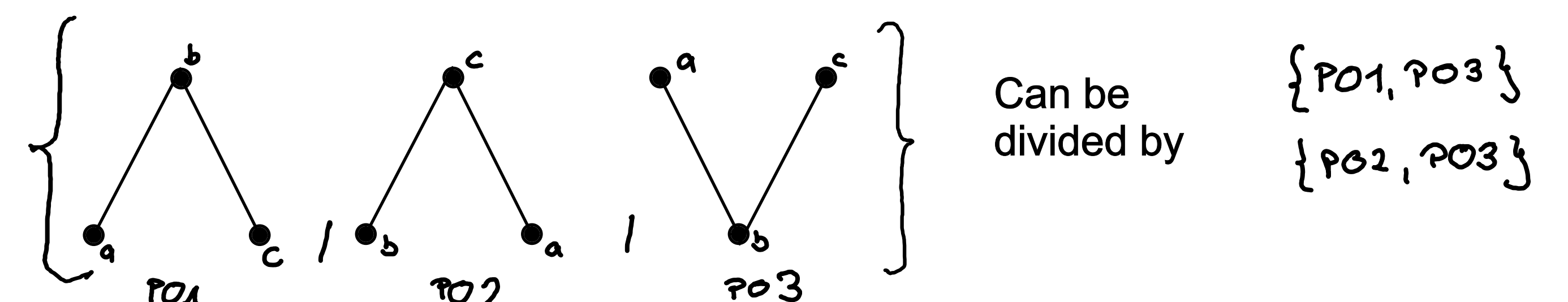
$$\mathcal{S} = \{\{a, b, c\} \subseteq \mathbb{R}\}$$



$$\mathcal{S} = \{P \subseteq \mathcal{P} \mid \text{Condition (C1) and (C2) hold}\}$$

with

- (C1) $P \subsetneq \gamma(P)$,
- (C2) There does not exist a family $(A_i)_{i \in \{1, \dots, \ell\}}$ such that for all $i \in \{1, \dots, \ell\}$, $A_i \subsetneq P$ and $\bigcup_{i \in \{1, \dots, \ell\}} \gamma(A_i) = \gamma(P)$.



Define the Depth Function (\mathcal{M} is a Set of Probability Measures)

$$D: \mathbb{R}^d \times \mathcal{M} \rightarrow [0, 1], \\ (x, \nu) \mapsto \nu(x \in \gamma_{\mathbb{R}^d}\{X_1, \dots, X_{d+1}\}),$$

$$\mathcal{P} \times \mathcal{M} \rightarrow [0, 1] \\ D: (p, \nu) \mapsto \begin{cases} 0, & \text{if for all } S \in \mathcal{S}: \prod_{\tilde{p} \in S} \nu(\tilde{p}) = 0 \\ c \sum_{S \in \mathcal{S}: p \in \gamma(S)} \prod_{\tilde{p} \in S} \nu(\tilde{p}), & \text{else,} \end{cases}$$

Comparison of Machine Learning Algorithms

For each data set we compare a set of ml algorithms based on multi-dimensional performance measures. This leads to a partial order for every data set.

- Data Sets: 80 classification problems from OpenML.
- ML Algorithms: Random Forests (RF), Decision Tree (CART), Logistic regression (LR), L1-penalized logistic regression (Lasso) and k-nearest neighbours(KNN).
- Performance Measures: area under the curve, F-score, predictive accuracy and Brier score.

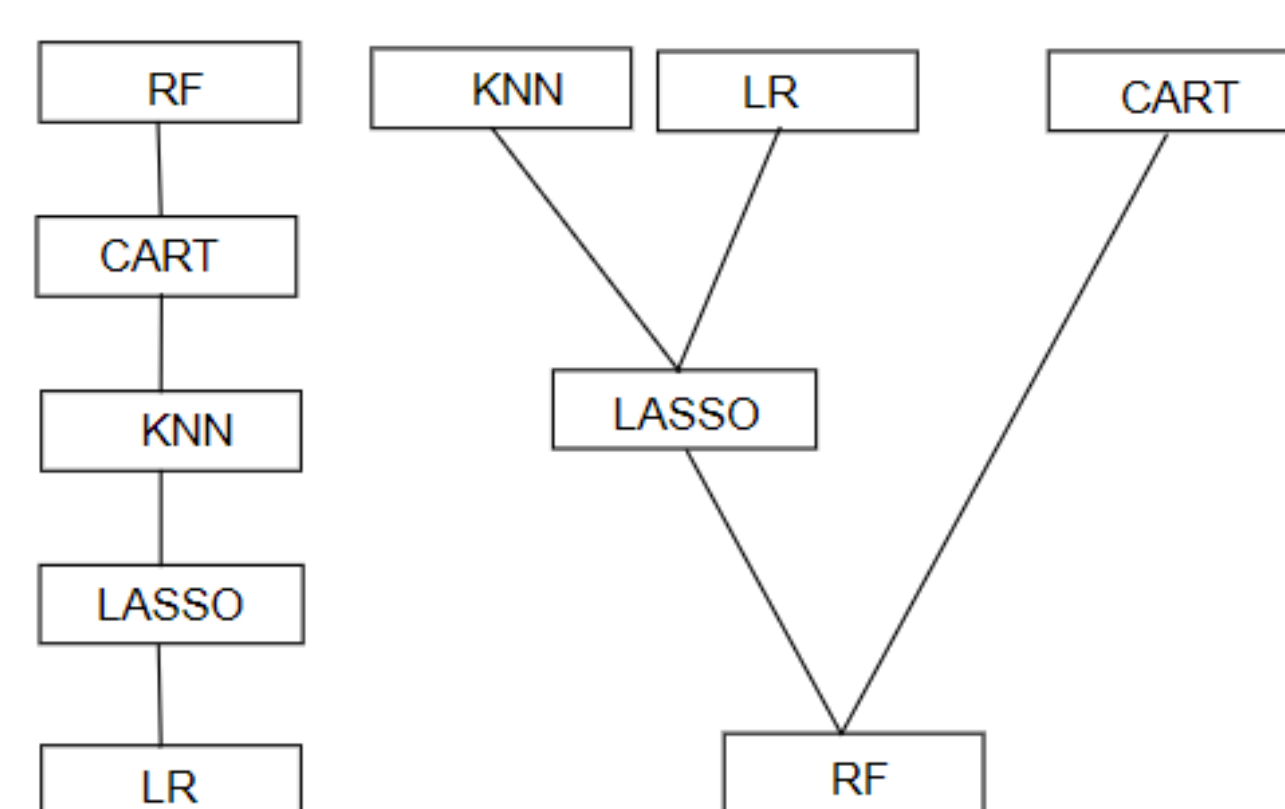


Figure 1: Observed poset with maximal (left) and minimal (right) ufg depth.

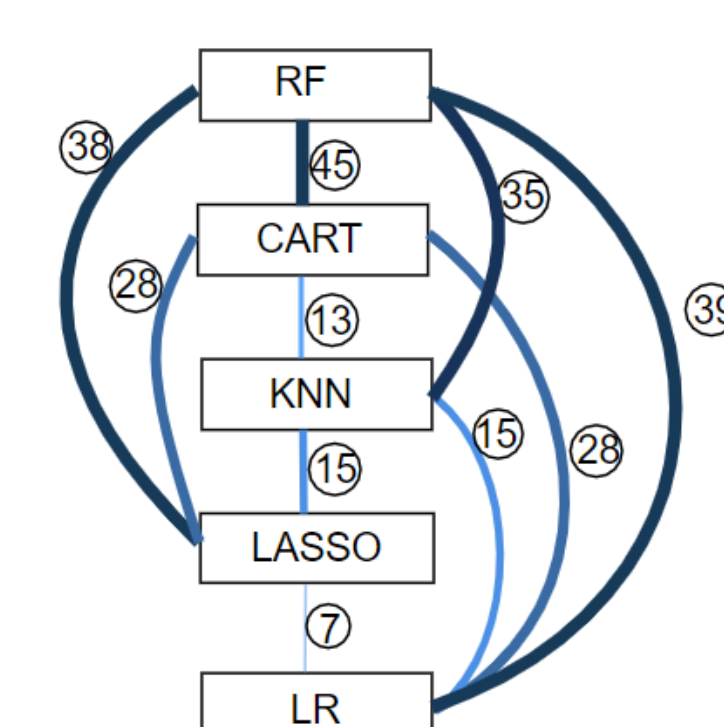


Figure 2: Represents what the observed posets with the k highest depth values have in common.

Outlook

- Other types of non-standard data
- Statistical Inference
- Other ML problems and criteria
- ...

References

- Blocher, Schollmeyer, Jansen, Nalenz (2023): Depth Functions for Partial Orders with a Descriptive Analysis of Machine Learning Algorithms. *Forthcoming in: ISIPTA '23*.
- Blocher, Schollmeyer, Jansen (2022): Statistical models for partial orders based on data depth and formal concept analysis. In: Ciucci, D.; Couso, I.; Medina, J.; Slezak, D.; Petturiti, D.; Bouchon-Meunier, B.; Yager, R.R. (eds): *IPMU Communications in Computer and Information Science*, vol 1602, Springer.

