Probability of Using Overrule to Evaluate Overlap in **Causal Inference**



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Background

Rule-based classification: Classification method based on IF-Then statements.

Research Question

What is the performance of Overrule on identifying overlap for different types of datasets?

- How does Overrule work with identifying overlap?
- How to find feasible hyperparameters for Overrule?
- Is Overrule sensitive to outliers?
- How does Overrule scale as the number of features and samples increases in the dataset?

Methodology

Overrule[1] package is used as the start of the project.



Figure 1. Overlap of two 1d normal-distributed classes

- 1. One-Class Support Vector Machine [2] to define the classes distribution.
- 2. Boolean Rule to extract the rules and evaluate the overlap.

Experimental Setup

Simulated data is used.

- scipy.stats is used to calculate the data distribution in order to find true overlap.
- Confusion matrix and Intersection over Union (IoU) is used to test the performance.
- Grid search for choosing hyperparameters.
- Performance of Overrule with different numbers of samples.
- Uniform distributed data is used to simulate outliers.
- Test with Iris dataset.

0.3



(a) True overlap estimator in 2d



- Overrule - One Class SVN 0.5 , ^{0.4} ₽ 0.5 0 3 0.2 0.1 0.0 One Class SVM 0.2 0.4 0.6 0.6 0.8 alpha (a) α (b) β







(a) IoU of Overrule with a different fraction of ourliers 0.9)

(b) IoU with different numbers of samples and features ($\alpha = 0.1, \beta =$

(c) Overrule for Iris

Conclusion and future work

- $\alpha = 0.1$, $\beta = 0.9$ works for most cases.
- Overrule can handle outliers effectively.
- More samples are needed for more features.
- Test with datasets with different distributions.
- Test with more than two classes and different variable types
- Test with real datasets.

References

- [1] Michael Oberst, Fredrik Johansson, Dennis Wei, Tian Gao, Gabriel Brat, David Sontag, and Kush Varshnev. Characterization of overlap in observational studies In International Conference on Artificial Intelligence and Statistics, pages 788–798. PMLR, 2020.
- [2] Bernhard Schölkopf, John C Platt, John Shawe-Taylor, Alex J Smola, and Robert C Williamson. Estimating the support of a high-dimensional distribution. Neural computation, 13(7):1443-1471, 2001.



(b) Outliers in 2d dataset

outliers for x1

