Publications

Dominik Janzing

Co-authored book

[1] J. Peters, D. Janzing, and B. Schölkopf. *Elements of Causal Inference – Foundations and Learning Algorithms*. MIT Press, 2017.

Articles in peer-reviewed conference proceedings

[1] Elias Eulig, Atalanti A. Mastakouri, Patrick Blöbaum, Michaela Hardt, and Dominik Janzing. Toward falsifying causal graphs using a permutation-based test. In *Proceedings of the 39th Annual AAAI Conference on Artificial Intelligence*, 2025.

[2] Philipp M. Faller, Leena Chennuru Vankadara, Atalanti A. Mastakouri, Francesco Locatello, and Dominik Janzing. Self-compatibility: Evaluating causal discovery without ground truth. AISTATS, 2024.

[3] D. Janzing, P. Blöbaum, A. Mastakouri, P.M. Faller, L. Minorics, and K. Budhathoki. Quantifying indirect causal influence via structure preserving interventions. AISTATS, 2024.

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