

Stephan Rabanser

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EDUCATION

- **PhD in Computer Science** Toronto, Canada
University of Toronto, advised by Prof. Nicolas Papernot *September 2020 – Present*
 - **Supervisory Committee:** Prof. Nicolas Papernot, Prof. Rahul Krishnan, Prof. David Duvenaud, Prof. Roger Grosse, Prof. Zachary Lipton
 - **Research Interests:** Machine/Deep Learning, Robustness, Safety, Reliability, Uncertainty, Causality, Generative Modeling, Representation Learning, Anomaly Detection, Distribution Shifts, Interpretability, Out-of-Distribution Sample Detection, Healthcare Applications.
 - **TAing:** CSC2541: Neural Network Training Dynamics (Winter 2022), CSC2515: Introduction to Machine Learning (Fall 2022), ECE1784: Trustworthy Machine Learning (Fall 2022)
- **M.Sc. in Computer Science** Munich, Germany
Technical University of Munich (TUM), advised by Prof. Stephan Günnemann *October 2015 – July 2019*
- **Visiting Research Scholar** Pittsburgh, PA
Carnegie Mellon University (CMU), advised by Prof. Zachary Lipton *August 2018 – January 2019*
- **Honours Degree in Technology Management** Munich, Germany
Center for Digital Technology and Management (CDTM) *August 2015 – June 2017*
- **Visiting Research Student** Cambridge, MA
Massachusetts Institute of Technology (MIT), advised by Prof. Thomas Malone *February 2016 – June 2016*
- **B.Sc. in Computer Science, Minor in Economic Sciences** Munich, Germany
Technical University of Munich (TUM) *October 2012 – October 2015*

EXPERIENCE

- **Machine Learning Researcher** Toronto, CA
Vector Institute for Artificial Intelligence *September 2020 – Present*
- **Intern Applied Scientist** Munich, Germany
Amazon, AWS AI Labs *June 2021 – October 2021*
 - Designed context-invariant time series representations using contrastive and domain-adversarial learning.
- **Intern Applied Scientist** Munich, Germany
Amazon, AWS AI Labs *September 2019 – July 2020*
 - Systematically assessed the impact of I/O representations for deep-learning-based time-series forecasting.
- **Intern Applied Scientist** Munich, Germany
Amazon, AWS AI Labs *May 2018 – August 2018*
 - Evaluated existing and developed new ML-based algorithms for large-scale lossless data compression.
 - Implemented autoencoder-based probability distribution estimation for arithmetic coding on tabular data.
- **Intern Software Development Engineer** Berlin, Germany
Amazon, Core Machine Learning *August 2017 – October 2017*
 - Received an overview of standard time series analysis / forecasting techniques.
 - Implemented **Bayes by Backprop** (weight uncertainty quantification) for plain MLPs & RNNs in MXNet.
- **Intern Software Development Engineer** Berlin, Germany
Amazon, AWS OpsWorks *July 2016 – October 2016*
 - Developed internal business intelligence tool (business metrics reporting and automated dashboard generation) for new OpsWorks service offering (OpsWorks for Chef Automate).

PUBLICATIONS

- Adam Dziedzic, Stephan Rabanser, Mohammad Yaghini, and Nicolas Papernot. ***p*-DkNN: Out-of-Distribution Detection through Statistical Testing of Deep Representations**. *arXiv preprint arXiv:2207.12545*, 2022 [[paper](#)]
- Stephan Rabanser, Tim Januschowski, Kashif Rasul, Oliver Borchert, Richard Kurl, Jan Gasthaus, Michael Bohlke-Schneider, Nicolas Papernot, and Valentin Flunkert. **Intrinsic Anomaly Detection in Multi-Variate Time Series**. *arXiv preprint arXiv:2206.14342*, 2022 [[paper](#)]
- Stephan Rabanser, Anvith Thudi, Kimia Hamidieh, Adam Dziedzic, and Nicolas Papernot. **Selective Classification Via Neural Network Training Dynamics**. *arXiv preprint arXiv:2205.13532*, 2022 [[paper](#)]
- Stephan Rabanser, Tim Januschowski, Valentin Flunkert, David Salinas, and Jan Gasthaus. **The Effectiveness of Discretization in Forecasting: An Empirical Study on Neural Time Series Models**. In *7th KDD Workshop on Mining and Learning from Time Series (MiLeTS)*, 2020. **Oral presentation**. [[paper](#), [slides](#)]
- Stephan Rabanser, Stephan Günnemann, and Zachary Lipton. **Failing Loudly: An Empirical Study of Methods for Detecting Dataset Shift**. In *Advances in Neural Information Processing Systems*, pages 1394–1406, 2019 [[paper](#), [poster](#), [slides](#)]
- Stephan Rabanser, Oleksandr Shchur, and Stephan Günnemann. **Introduction to Tensor Decompositions and their Applications in Machine Learning**. *arXiv preprint arXiv:1711.10781*, 2017 [[paper](#)]

AWARDS & HONORS

- **Member of the Elite Network of Bavaria** *Since 2016*
- **Apple WWDC Student Scholarship** *June 2013*

COMMUNITY SERVICE

- **Reviewing:** NeurIPS (2022, 2021), ICML (2022, 2021), Distribution Shift Workshop @ ICML (2022), Distribution Shift Workshop @ NeurIPS (2021 – *outstanding reviewer*, 2022), Human Evaluation of Generative Models Workshop @ NeurIPS (2022), Time Series Workshop @ ICML (2021), Time Series Workshop @ KDD (2022), AAAI (2020)
- **Talks:** Microsoft Security Data Science Colloquium (2021), Private AI Institute Fall Workshop (2022), MIT MIMO Student Research Forum (2022)
- **Volunteering:** NeurIPS 2018, ICLR 2019

SELECTED COURSEWORK

- **De-noising Spectral Clustering Through Latent Data Decomposition** Munich, Germany
Guided Research Project @ TUM *October 2017 – March 2018*
 - Developed two new methods to make spectral clustering more robust (reduced sensitivity to noise).
 - Modeled problem as latent data decomposition instead of similarity graph decomposition.
 - Initial results outperform similar techniques on many datasets.
- **Data Science in Astrophysics and Industry** Munich, Germany
Interdisciplinary Project @ Max Planck Institute for Astrophysics (MPA) *March 2017 – July 2017*
 - Transformed an existing Gaussian mixture model (GMM) into Google TensorFlow.
 - Optimized the algorithmic implementation of the model (e.g. number of mixture components, hyper-parameters) and explored different training methods (stochastic vs. deterministic and expectation maximization (EM) vs. gradient descent vs. Newton).
 - Researched, implemented, and improved online learning techniques for GMMs and compared them to standard EM and tensor decomposition approaches.

PROGRAMMING SKILLS

- **Languages:** Python, Java, Swift, HTML/CSS/JS **ML Frameworks:** PyTorch, Tensorflow, JAX, MXNet