

# Eric Nalisnick

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enalisnick.github.io

## Academic Employment

**Assistant Professor, Tenure Track**  
*Amsterdam Machine Learning Lab*

University of Amsterdam  
September 2020 - present

**Postdoctoral Research Associate**  
Supervisor: José Miguel Hernández-Lobato

University of Cambridge  
September 2018 to September 2020

**Graduate Student Researcher**  
Supervisor: Padhraic Smyth

University of California, Irvine  
October 2013 to June 2018

## Industrial Employment

**Research Scientist**  
Part time, one day per week (20% FTE)

Google DeepMind  
February 2019 to January 2020

**Research Scientist Intern**  
Supervisor: Balaji Lakshminarayanan

Google DeepMind  
Summer 2018

**Applied Scientist Intern**  
Supervisors: Vijai Mohan, Eiman Elnahrawy

Amazon  
Fall 2016

**Research Intern**  
Supervisor: Hugo Larochelle

Twitter  
Summer 2016

**Research Intern**  
Supervisors: Rich Caruana, Nick Craswell

Microsoft  
Summer 2015

**Research Scientist Intern**  
Supervisors: Vijai Mohan, Rahul Bhagat

Amazon  
Summer 2014

## Education

**Ph.D. Computer Science**  
**University of California, Irvine**

2013 - 2018

**M.S. Computer Science**  
**Lehigh University**

2012 - 2013

**B.S. Computer Science & English Literature**  
**Lehigh University**

2008 - 2012

## Academic Honors and Awards

Veni Laureate, Dutch Research Council (NWO)	2021
ELLIS Scholar, <i>European Lab for Learning and Intelligent Systems Society</i>	2021
Top / Best Reviewer	NeurIPS 2017, ICML 2019, ICML 2020

## Research Funding

### PRINCIPAL INVESTIGATOR

<i>Continual Learning under Human Guidance</i>	June 2022 - 2026
€280,000	
Veni, Talent Programme, Dutch Research Council (NWO): Science Domain	
Single Principal Investigator, Acceptance Rate: 16%.	

### CO-INVESTIGATOR

<i>UvA-Bosch Delta Lab</i>	November 2021 - 2026
Gift funding for 10 PhD students from the Bosch Group.	
Role: Supervisor for 4 PhD students.	
PIs: Theo Gevers, Jan-Willem van de Meent.	
<i>Hybrid Intelligence Centre</i>	January 2020 - 2030
Gravitation Programme, Dutch Research Council (NWO)	
Role: Co-supervisor for 2 PhD students.	
PIs: F. v. Harmelen, C. Jonker, M. d. Rijke, R. Verbrugge, P. Vossen, M. Welling.	

## Academic Supervision

### PHD CANDIDATES

Mona Schirmer	University of Amsterdam, 2022 -
Dharmesh Tailor	University of Amsterdam, 2021 -
Saba Amiri (with Adam Belloum and Sander Klous)	University of Amsterdam, 2021 -
Putra Manggala (with Holger Hoos)	University of Amsterdam, 2021 -
Urja Khurana (with Antske Fokkens)	Vrije University Amsterdam, 2020 -
Mrinank Sharma (with Tom Rainforth and Yee Whye Teh)	University of Oxford, 2020 -

### MASTERS STUDENTS

Nils Lehmann	University of Amsterdam, 2021 -
Shuai Wang	University of Amsterdam, 2021 -
Rajeev Verma	University of Amsterdam, 2021 -
Arsen Sheverdin	University of Amsterdam, 2021 -

Daniël Nobbe

University of Amsterdam, 2021

## Teaching

*Introduction to Machine Learning (Leren)*, University of Amsterdam 2020 - 2022  
 ~ 180 Undergraduate Students, ~ 12 Teaching Assistants

*Bayesian Deep Learning* Module, Deep Learning II, University of Amsterdam 2022  
 123 Graduate Students, 5 Teaching Assistants

## Professional Service

### ORGANIZATION

Anomaly Detection for Scientific Discovery 2021 - present  
 Tractable Probabilistic Modeling, UAI Workshop 2022  
 Bayesian Deep Learning, NeurIPS Workshop 2021  
 Bayesian Deep Learning, ELLIS Workshop / NeurIPS Meetup 2020  
 Bayesian Deep Learning, NeurIPS Workshop 2019

### AREA CHAIR / SENIOR PROGRAM COMMITTEE

Uncertainty in Artificial Intelligence (UAI) 2021 - present  
 Neural Information Processing Systems (NeurIPS) 2022  
 International Conference on Machine Learning (ICML) 2022  
 Artificial Intelligence and Statistics (AISTats) 2022  
 International Conference on Learning Representations (ICLR) 2021

### EDITORIAL BOARD

Editor, Probabilistic Methods for Deep Learning, Special Issue of *Entropy* 2021

### JOURNAL REVIEWING

American Statistical Association (JASA) 2020 - present  
 Machine Learning Research (JMLR) 2018 - present  
 Advances in Statistical Analysis 2020 - present

### CONFERENCE REVIEWING

Neural Information Processing Systems (NeurIPS) 2016 - 2021  
 International Conference on Learning Representations (ICLR) 2018 - present  
 International Conference on Machine Learning (ICML) 2018 - 2021

Artificial Intelligence and Statistics (AISTats)	2019 - 2021
Uncertainty in Artificial Intelligence (UAI)	2019 - 2020
Hybrid Human-Artificial Intelligence (HHAI)	2022
Association for the Advancement of Artificial Intelligence (AAAI)	2020 - 2021
International Joint Conference on Artificial Intelligence (IJCAI)	2019

## Departmental / Institute Service

Education program committee for Bachelors and Masters of AI	2021 - present
Hiring committee for one AMLab / ELLIS unit faculty position	2021
Hiring committee for two AMLab faculty positions	2021

## Publications

\* Denotes equal contribution

### JOURNAL ARTICLES

1. G. Papamakarios\*, **E. Nalisnick**\*, D. J. Rezende, S. Mohamed, and B. Lakshminarayanan. Normalizing Flows for Probabilistic Modeling and Inference. *Journal of Machine Learning Research* (JMLR), 2021.

### CONFERENCE ARTICLES

2. R. Verma and **E. Nalisnick**. Calibrated Learning to Defer with One-vs-All Classifiers. In *Proceedings of the 39th International Conference on Machine Learning* (ICML), 2022.
3. J. Antoran, D. Janz, J. U. Allingham, E. Daxberger, R. Barbano, **E. Nalisnick**, and J. M. Hernández-Lobato. Adapting the Linearised Laplace Model Evidence for Modern Deep Learning. In *Proceedings of the 39th International Conference on Machine Learning* (ICML), 2022.
4. E. Daxberger, **E. Nalisnick**\*, J. U. Allingham\*, J. Antoran\*, and J. M. Hernández-Lobato. Expressive yet Tractable Bayesian Deep Learning via Subnetwork Inference. In *Proceedings of the 38th International Conference on Machine Learning* (ICML), 2021.
5. **E. Nalisnick**, J. Gordon, and J. M. Hernández-Lobato. Predictive Complexity Priors. In *Proceedings of the 24th International Conference on Artificial Intelligence and Statistics* (AISTATS), 2021.
6. R. Pinsler, J. Gordon, **E. Nalisnick**, and J. M. Hernández-Lobato. Bayesian Batch Active Learning as Sparse Subset Approximation. In *Advances in Neural Information Processing Systems* (NeurIPS), 2019.
7. **E. Nalisnick**, J. M. Hernández-Lobato, and P. Smyth. Dropout as a Structured Shrinkage Prior. In *Proceedings of the 36th International Conference on Machine Learning* (ICML), 2019.

8. **E. Nalisnick\***, A. Matsukawa\*, Y. W. Teh, D. Gorur, and B. Lakshminarayanan. Hybrid Models with Deep and Invertible Features. In *Proceedings of the 36th International Conference on Machine Learning (ICML)*, 2019.
9. **E. Nalisnick**, A. Matsukawa, Y. W. Teh, D. Gorur, and B. Lakshminarayanan. Do Deep Generative Models Know What They Don’t Know? In *Proceedings of the 7th International Conference on Learning Representations (ICLR)*, 2019.
10. D. Ji, **E. Nalisnick**, Y. Qian, R. Scheuermann, and P. Smyth. Bayesian Trees for Automated Cytometry Data Analysis. In *Proceedings of Machine Learning for Healthcare (MLHC)*, 2018.
11. **E. Nalisnick** and P. Smyth. Learning Priors for Invariance. In *Proceedings of the 21st International Conference on Artificial Intelligence and Statistics (AISTATS)*, 2018.
12. **E. Nalisnick** and P. Smyth. Learning Approximately Objective Priors. In *Proceedings of the 33rd Conference on Uncertainty in Artificial Intelligence (UAI)*, 2017.
13. **E. Nalisnick** and P. Smyth. Stick-Breaking Variational Autoencoders. In *Proceedings of the 5th International Conference on Learning Representations (ICLR)*, 2017.
14. **E. Nalisnick**, B. Mitra, N. Craswell, and R. Caruana. Improving Document Ranking with Dual Word Embeddings. In *Proceedings of the 25th World Wide Web Conference (WWW)*, 2016.
15. **E. Nalisnick** and H. Baird. Character-to-Character Sentiment Analysis in Shakespeare’s Plays. In *Proceedings of the 51st Annual Meeting of the Association for Computational Linguistics (ACL)*, 2013.
16. **E. Nalisnick** and H. Baird. Extracting Sentiment Networks from Shakespeare’s Plays. In *Proceedings of the 12th International Conference on Document Analysis and Recognition (ICDAR)*, 2013.

#### PEER-REVIEWED WORKSHOP ARTICLES

17. S. Wang and **E. Nalisnick**. Active Learning for Multilingual Fingerspelling Corpora. *Adaptive Experimental Design and Active Learning in the Real World*, ICML 2022.
18. R. Verma, D. Barrejón, and **E. Nalisnick**. On the Calibration of Learning to Defer to Multiple Experts. *Human-Machine Collaboration and Teaming*, ICML 2022.
19. J. U. Allingham and **E. Nalisnick**. A Product of Experts Approach to Early-Exit Ensembles. *Dynamic Neural Networks*, ICML 2022.
20. P. Manggala, H. Hoos, and **E. Nalisnick**. Bayesian Weak Supervision via an Optimal Transport Approach. *Human-Machine Collaboration and Teaming*, ICML 2022.
21. U. Khurana, I. Vermeulen, **E. Nalisnick**, M. van Noorloos, and A. Fokkens. Hate Speech Criteria: A Modular Approach to Task-Specific Hate Speech Definitions. *Online Abuse and Harms*, NAACL 2022.
22. S. Amiri, A. Belloum, **E. Nalisnick**, S. Klous, and L. Gommans. On the Impact of Non-IID Data on the Performance and Fairness of Differentially Private Federated Learning. *Dependable and Secure Machine Learning*, DSN 2022.

23. J. Antoran, J. U. Allingham, D. Janz, E. Daxberger, **E. Nalisnick**, and J. M. Hernández-Lobato. Linearised Laplace Inference in Networks with Normalisation Layers and the Neural g-Prior. *Symposium on Advances in Approximate Bayesian Inference*, 2022.
24. U. Khurana, **E. Nalisnick**, and A. Fokkens. How Emotionally Stable is ALBERT? Testing Robustness with Stochastic Weight Averaging on a Sentiment Analysis Task. *Evaluation and Comparison of NLP Systems*, EMNLP 2021.
25. P. Manggala, H. Hoos, and **E. Nalisnick**. Bayesian Regression from Multiple Sources of Weak Supervision. *Machine Learning for Data*, ICML 2021.
26. Y. Zhang and **E. Nalisnick**. On the Inconsistency of Bayesian Inference for Misspecified Neural Networks. *Symposium on Advances in Approximate Bayesian Inference*, 2021.
27. E. Daxberger, **E. Nalisnick\***, J. U. Allingham\*, J. Antoran\*, and J. M. Hernández-Lobato. Expressive yet Tractable Bayesian Deep Learning via Subnetwork Inference. *Symposium on Advances in Approximate Bayesian Inference*, 2021.
28. **E. Nalisnick**, J. Gordon, and J. M. Hernández-Lobato. Predictive Complexity Priors. *Uncertainty & Robustness in Deep Learning*, ICML 2020.
29. **E. Nalisnick**, A. Matsukawa, Y. W. Teh, and B. Lakshminarayanan. Detecting Out-of-Distribution Inputs to Deep Generative Models Using Typicality. *Bayesian Deep Learning*, NeurIPS 2019.
30. **E. Nalisnick** and J. M. Hernández-Lobato. Automatic Depth Determination for Bayesian ResNets. *Bayesian Deep Learning*, NeurIPS 2018.
31. **E. Nalisnick**, A. Matsukawa, Y.W. Teh, D. Gorur, and B. Lakshminarayanan. Do Deep Generative Models Know What They Don’t Know? *Bayesian Deep Learning*, NeurIPS 2018.
32. **E. Nalisnick\***, A. Matsukawa\*, Y.W. Teh, D. Gorur, and B. Lakshminarayanan. Hybrid Models with Deep and Invertible Features. *Bayesian Deep Learning*, NeurIPS 2018.
33. O. Rybakov, V. Mohan, A. Misra, S. LeGrand, R. Joseph, K. Chung, S. Singh, Q. You, **E. Nalisnick**, L. Dirac, and R. Luo. The Effectiveness of a Two-Layer Neural Network for Recommendations. Workshop Track, ICLR 2018.
34. D. Ji, **E. Nalisnick**, and P. Smyth. Mondrian Processes for Flow Cytometry Analysis. *Machine Learning for Health*, NeurIPS 2017.
35. **E. Nalisnick** and P. Smyth. Variational Inference with Stein Mixtures. *Advances in Approximate Bayesian Inference*, NIPS 2017.
36. **E. Nalisnick** and P. Smyth. The Amortized Bootstrap. *Implicit Models*, ICML 2017.
37. **E. Nalisnick** and P. Smyth. Variational Reference Priors. Workshop Track, ICLR 2017.
38. **E. Nalisnick**, L. Hertel, and P. Smyth. Approximate Inference for Deep Latent Gaussian Mixtures. *Bayesian Deep Learning*, NeurIPS 2016.
39. **E. Nalisnick** and P. Smyth. Nonparametric Deep Generative Models with Stick-Breaking Priors. *Data-Efficient Machine Learning*, ICML 2016.

40. J. Park, M. Blume-Kohout, R. Krestel, **E. Nalisnick**, and P. Smyth. Analyzing NIH Funding Patterns over Time with Statistical Text Analysis. *Scholarly Big Data*, AAAI 2016.

#### THESES

1. **E. Nalisnick**. On Priors for Bayesian Neural Networks. *Doctoral Dissertation*, University of California, Irvine, 2018.
2. **E. Nalisnick**. Automatic Methods for Tracking Sentiment Dynamics in Plays. *Master's Thesis*, Lehigh University, 2013.
3. **E. Nalisnick**. A Combinatorial Explanation for a Conjecture of Fomin and Zelevinsky. *Honors Thesis*, Lehigh University, 2012.

#### PATENTS

1. E. M. H. Elnahrawy, V. Mohan, and **E. Nalisnick**. Generation and Use of Model Parameters in Cold-Start Scenarios. U.S. Patent Number 10,726,334. 28 July 2020.

### Invited Talks

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|---|------|
| 1. Towards Informative Priors for Bayesian Deep Learning, DAGSTUHL SEMINAR                  | 2022 |
| 2. On the Calibration of Learning-to-Defer Systems, UNIV. OF MANCHESTER, STAT. SEMINAR      | 2022 |
| 3. Predictive Complexity Priors, UNIVERSITY OF EDINBURGH, ANC SEMINAR                       | 2021 |
| 4. Predictive Complexity Priors, GENERATIVE MODELS AND UNCERTAINTY QUANTIFICATION           | 2021 |
| 5. Predictive Complexity Priors, IMPERIAL COLLEGE STATISTICS SEMINAR                        | 2021 |
| 6. Predictive Complexity Priors, ALAN TURING INSTITUTE                                      | 2020 |
| 7. Detecting Distribution Shift with Deep Generative Models, SYDNEY ML MEETUP               | 2020 |
| 8. Detecting Distribution Shift with Deep Generative Models, INN+ , ICML WORKSHOP           | 2020 |
| 9. Building and Critiquing Models for Probabilistic Deep Learning, GATSBY UNIT, UCL         | 2020 |
| 10. Building and Critiquing Models for Probabilistic Deep Learning, CARNEGIE MELLON UNIV.   | 2020 |
| 11. Building and Critiquing Models for Probabilistic Deep Learning, UNIV. OF NORTH CAROLINA | 2020 |
| 12. Deep Learning & Statistics: Bridging the Gap with Prob. Structure, UNIV. OF AMSTERDAM   | 2020 |
| 13. Deep Learning & Statistics: Bridging the Gap with Prob. Structure, UC SANTA BARBARA     | 2020 |
| 14. Deep Learning Under Covariate Shift, UCI AI/ML SEMINAR                                  | 2019 |
| 15. Normalizing Flows for Tractable Probabilistic Modeling and Inference, T-PRIME, NEURIPS  | 2019 |
| 16. Deep Learning: A Synthesis from Probabilistic Foundations, RAND CORP STATS. SEMINAR     | 2019 |
| 17. Evaluating Deep Generative Models on Out-of-Distribution Inputs, OXFORD STATS. SEMINAR  | 2019 |
| 18. Do Deep Generative Models Know What They Don't Know?, CAMAIML (MSR CAMBRIDGE)           | 2019 |

19. Do Deep Generative Models Know What They Don't Know?, CAMBRIDGE LTL SEMINAR 2019
20. Structured Shrinkage Priors for Neural Networks, IMPERIAL COLLEGE STATISTICS SEMINAR 2018
21. Deep Learning: A Synthesis from Probabilistic Foundations, UCI STATISTICS SEMINAR 2018
22. Approximate Inference for Frequentist Uncertainty Estimation, SoCAL ML SYMPOSIUM 2017
23. Deep Generative Models with Stick-Breaking Priors, UCI AI/ML SEMINAR 2017
24. Alternative Priors for Deep Generative Models, OPENAI 2017