

# Dustin Tran

Senior Staff Research Scientist  
Google DeepMind  
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## Education

Ph.D. Computer Science, Columbia University Advisors: David M. Blei, Andrew Gelman	2016–2020
Ph.D. Statistics, Harvard University (transferred) M.S. Computational Science & Engineering, Harvard University Advisor: Edoardo M. Airolidi	2014–2015
B.A. (Hon.) Mathematics, Statistics, University of California, Berkeley	2010–2014

## Employment

Senior Staff Research Scientist	2023–
Staff Research Scientist	2021–2023
Senior Research Scientist Google DeepMind	2018–2021
Research Intern Google Brain	Oct 2017 – Jan 2018
Research Intern OpenAI	May 2017 – Oct 2017

## Awards

John M. Chambers Statistical Software Award (for Edward)	2018
Google Ph.D. Fellowship in Machine Learning (\$34,000 + tuition/fees)	2017–2020
Columbia SEAS Fellowship (Full funding)	2016–2020
Adobe Research Fellowship (\$10,000)	2016
LinkedIn Economic Graph Challenge	2015
Harvard GSAS Fellowship (Full funding)	2015
Regents' and Chancellor's Scholarship (Full funding)	2010–2014
Cal Alumni Leadership Scholarship (\$2,500)	2010

## Publications

## PREPRINTS

1. Gemini Team. Gemini 2.5: Pushing the frontier with advanced reasoning, multimodality, long context, and next generation agentic capabilities. 2025.
2. Gemma Team. Gemma 3 technical report. 2025.
3. Gemini Team. Gemini 1.5: Unlocking multimodal understanding across millions of tokens of context. 2024.
4. Gemini Team. Gemini: A family of highly capable multimodal models. 2023.
5. J. Wei, J. Wei, Y. Tay, **D. Tran**, A. Webson, Y. Lu, X. Chen, H. Liu, D. Huang, D. Zhou, and others. Larger language models do in-context learning differently. 2023.
6. **D. Tran**, J. Liu, M. W. Dusenberry, D. Phan, M. Collier, J. Ren, K. Han, Z. Wang, Z. Mariet, H. Hu, N. Band, T. G. J. Rudner, K. Singhal, Z. Nado, J. van Amersfoort, A. Kirsch, R. Jenatton, N. Thain, H. Yuan, K. Buchanan, K. Murphy, D. Sculley, Y. Gal, Z. Ghahramani, J. Snoek, and B. Lakshminarayanan. Plex: Towards reliability using pretrained large model extensions. 2022.
7. Z. Nado, N. Band, M. Collier, J. Djolonga, M. W. Dusenberry, S. Farquhar, Q. Feng, A. Filos, M. Havasi, R. Jenatton, G. Jerfel, J. Liu, Z. Mariet, J. Nixon, S. Padhy, J. Ren, T. G. J. Rudner, F. Sbahi, Y. Wen, F. Wenzel, K. Murphy, D. Sculley, B. Lakshminarayanan, J. Snoek, Y. Gal, and **D. Tran**. Uncertainty Baselines: Benchmarks for uncertainty & robustness in deep learning. 2021.
8. J. Lee, **D. Tran**, O. Firat, and K. Cho. On the discrepancy between density estimation and sequence generation. 2020.
9. J. Nixon, M. W. Dusenberry, L. Zhang, G. Jerfel, and **D. Tran**. Measuring calibration in deep learning. 2018.
10. M. Hoffman, P. Sountsov, J. V. Dillon, I. Langmore, **D. Tran**, and S. Vasudevan. Neutra-lizing bad geometry in Hamiltonian Monte Carlo using neural transport. 2019.
11. **D. Tran**, Y. Burda, and I. Sutskever. Feature-matching auto-encoders. 2017.
12. J. Dillon, I. Langmore, **D. Tran**, E. Brevdo, S. Vasudevan, D. Moore, B. Patton, A. Alemi, M. Hoffman, and R. Saurous. TensorFlow Distributions. 2017.
13. **D. Tran**, A. Kucukelbir, A. B. Dieng, M. Rudolph, D. Liang, and D. M. Blei. Edward: A library for probabilistic modeling, inference, and criticism. 2016.
14. **D. Tran**, F. J. R. Ruiz, S. Athey, and D. M. Blei. Model criticism for Bayesian causal inference. 2016.
15. **D. Tran**, P. Toulis, and E. M. Airoldi. Stochastic gradient descent methods for estimation with large data sets. 2016.

## JOURNAL ARTICLES

16. E. Nalisnick, P. Smyth, and **D. Tran**. A brief tour of deep learning from a statistical perspective. *Annual Review of Statistics and Its Application*, 10(1):219–246, 2023.

17. J. Z. Liu, S. Padhy, J. Ren, Z. Lin, Y. Wen, G. Jerfel, Z. Nado, J. Snoek, **D. Tran**, and B. Lakshminarayanan. A simple approach to improve single-model deep uncertainty via distance-awareness. *Journal of Machine Learning Research*, 24(42):1–63, 2022.
18. M. Havasi, J. Snoek, **D. Tran**, J. Gordon, and J. M. Hernández-Lobato. Sampling the variational posterior with local refinement. *Entropy*, 23(11):1475, 2021.
19. A. Vehtari, A. Gelman, T. Sivula, P. Jylanki, **D. Tran**, S. Sahai, P. Blomstedt, J. P. Cunningham, D. Schiminovich, and C. P. Robert. Expectation propagation as a way of life: A framework for Bayesian inference on partitioned data. *Journal of Machine Learning Research*, 21(17):1–53, 2020.
20. **D. Tran** and D. M. Blei. Comment, “Fast approximate inference for arbitrarily large semiparametric regression models via message passing”. *Journal of the American Statistical Association*, 112(517):156–158, 2017.
21. A. Kucukelbir, **D. Tran**, R. Ranganath, A. Gelman, and D. M. Blei. Automatic differentiation variational inference. *Journal of Machine Learning Research*, 18(14):1–45, 2017.

#### CONFERENCE ARTICLES

22. J. Wei, C. Yang, X. Song, Y. Lu, N. Hu, **D. Tran**, D. Peng, R. Liu, D. Huang, C. Du, and others. Long-form factuality in large language models. In *Neural Information Processing Systems*, 2024.
23. M. Dehghani, J. Djolonga, B. Mustafa, P. Padlewski, J. Heek, J. Gilmer, A. Steiner, M. Caron, R. Geirhos, I. Alabdulmohsin, R. Jenatton, L. Beyer, M. Tschannen, A. Arnab, X. Wang, C. Riquelme, M. Minderer, J. Puigcerver, U. Evci, M. Kumar, S. van Steenkiste, G. F. Elsayed, A. Mahendran, F. Yu, A. Oliver, F. Huot, J. Bastings, M. P. Collier, A. Gritsenko, V. Birodkar, C. Vasconcelos, Y. Tay, T. Mensink, A. Kolesnikov, F. Pavetić, **D. Tran**, T. Kipf, M. Lučić, X. Zhai, D. Keysers, J. Harmsen, and N. Houlsby. Scaling vision transformers to 22 billion parameters. In *International Conference on Machine Learning*, 2023.
24. J. U. Allingham, J. Ren, M. W. Dusenberry, J. Z. Liu, X. Gu, Y. Cui, **D. Tran**, and B. Lakshminarayanan. A simple zero-shot prompt weighting technique to improve prompt ensembling in text-image models. In *International Conference on Machine Learning*, 2023.
25. J. U. Allingham, F. Wenzel, Z. E. Mariet, B. Mustafa, J. Puigcerver, N. Houlsby, G. Jerfel, V. Fortuin, B. Lakshminarayanan, J. Snoek, and others. Sparse MoEs meet efficient ensembles. *Transactions on Machine Learning Research*, 2022.
26. V. Fortuin, M. Collier, F. Wenzel, J. Allingham, J. Liu, **D. Tran**, B. Lakshminarayanan, J. Berent, R. Jenatton, and E. Kokiopoulou. Deep classifiers with label noise modeling and distance awareness. *Transactions on Machine Learning Research*, 2022.
27. M. Minderer, J. Djolonga, R. Romijnders, F. Hubis, X. Zhai, N. Houlsby, **D. Tran**, and M. Lucic. Revisiting the calibration of modern neural networks. In *Neural Information Processing Systems*, 2021.
28. N. Band, T. G. J. Rudner, Q. Feng, A. Filos, Z. Nado, M. W. Dusenberry, G. Jerfel, **D. Tran**, and Y. Gal. Benchmarking Bayesian deep learning on diabetic retinopathy detection tasks. In *Neural Information Processing Systems*, 2021.

29. A. Karandikar, N. Cain, **D. Tran**, B. Lakshminarayanan, J. Shlens, M. C. Mozer, and B. Roelofs. Soft calibration objectives for neural networks. In *Neural Information Processing Systems*, 2021.
30. Y. Wen, G. Jerfel, R. Muller, M. W. Dusenberry, J. Snoek, B. Lakshminarayanan, and **D. Tran**. Combining ensembles and data augmentation can harm your calibration. In *International Conference on Learning Representations*, 2021.
31. M. Havasi, R. Jenatton, S. Fort, J. Z. Liu, J. Snoek, B. Lakshminarayanan, A. M. Dai, and **D. Tran**. Training independent subnetworks for robust prediction. In *International Conference on Learning Representations*, 2020.
32. F. Wenzel, J. Snoek, **D. Tran**, and R. Jenatton. Hyperparameter ensembles for robustness and uncertainty quantification. In *Neural Information Processing Systems*, 2020.
33. J. Z. Liu, Z. Lin, S. Padhy, **D. Tran**, T. Bedrax-Weiss, and B. Lakshminarayanan. Simple and principled uncertainty estimation with deterministic deep learning via distance awareness. In *Neural Information Processing Systems*, 2020.
34. M. Mladenov, C.-w. Hsu, V. Jain, E. Ie, C. Colby, N. Mayoraz, H. Pham, **D. Tran**, I. Vendrov, and C. Boutilier. Demonstrating principled uncertainty modeling for recommender ecosystems with RecSim NG. In *ACM Conference on Recommender Systems*, 2020.
35. M. W. Dusenberry, G. Jerfel, Y. Wen, Y. Ma, J. Snoek, K. Heller, B. Lakshminarayanan, and **D. Tran**. Efficient and scalable Bayesian neural nets with rank-1 factors. In *International Conference on Machine Learning*, 2020.
36. Y. Wen, **D. Tran**, and J. Ba. Batchensemble: An alternative approach to efficient ensemble and lifelong learning. In *International Conference on Learning Representations*, 2020.
37. M. W. Dusenberry, **D. Tran**, E. Choi, J. Kemp, J. Nixon, G. Jerfel, K. Heller, and A. Dai. Analyzing the role of model uncertainty in electronic health records. In *ACM Conference on Health, Inference, and Learning*, 2020.
38. **D. Tran**, K. Vafa, K. K. Agrawal, L. Dinh, and D. Poole. Discrete flows: Invertible generative models for discrete data. In *Neural Information Processing Systems*, 2019.
39. **D. Tran**, M. W. Dusenberry, D. Hafner, and M. van der Wilk. Bayesian layers: A module for neural network uncertainty. 2019.
40. D. Hafner, **D. Tran**, A. Irpan, T. Lillicrap, and J. Davidson. Noise contrastive priors for functional uncertainty. In *Uncertainty in Artificial Intelligence*, 2019.
41. **D. Tran**, M. D. Hoffman, D. Moore, C. Suter, S. Vasudevan, A. Radul, M. Johnson, and R. A. Saurous. Simple, distributed, and accelerated probabilistic programming. In *Neural Information Processing Systems*, 2018.
42. N. Shazeer, Y. Cheng, N. Parmar, **D. Tran**, A. Vaswani, P. Koanantakool, P. Hawkins, H. Lee, M. Hong, C. Young, R. Sepassi, and B. Hechtman. Mesh-TensorFlow: Deep learning for supercomputers. In *Neural Information Processing Systems*, 2018.
43. M. D. Hoffman, M. Johnson, and **D. Tran**. Autoconj: Recognizing and exploiting conjugacy without a domain-specific language. In *Neural Information Processing Systems*, 2018.

44. N. Parmar, A. Vaswani, J. Uszkoreit, L. Kaiser, N. Shazeer, A. Ku, and **D. Tran**. Image Transformer. In *International Conference on Machine Learning*, 2018.
45. Y. Wen, P. Vicol, J. Ba, **D. Tran**, and R. Grosse. Flipout: Efficient pseudo-independent weight perturbations on mini-batches. In *International Conference on Learning Representations*, 2018.
46. **D. Tran** and D. M. Blei. Implicit causal models for genome-wide association studies. In *International Conference on Learning Representations*, 2018.
47. **D. Tran**, R. Ranganath, and D. M. Blei. Hierarchical implicit models and likelihood-free variational inference. In *Neural Information Processing Systems*, 2017.
48. A. B. Dieng, **D. Tran**, R. Ranganath, J. Paisley, and D. M. Blei. Variational inference via  $\chi$  upper bound minimization. In *Neural Information Processing Systems*, 2017.
49. **D. Tran**, M. D. Hoffman, R. A. Saurous, E. Brevdo, K. Murphy, and D. M. Blei. Deep probabilistic programming. In *International Conference on Learning Representations*, 2017.
50. R. Ranganath, J. Alotaibi, **D. Tran**, and D. M. Blei. Operator variational inference. In *Neural Information Processing Systems*, 2016.
51. R. Ranganath, **D. Tran**, and D. M. Blei. Hierarchical variational models. In *International Conference on Machine Learning*, 2016.
52. **D. Tran**, M. Kim, and F. Doshi-Velez. Spectral M-estimation with application to hidden Markov models. In *Artificial Intelligence and Statistics*, 2016.
53. P. Toulis, **D. Tran**, and E. M. Airolidi. Towards stability and optimality in stochastic gradient descent. In *Artificial Intelligence and Statistics*, 2016.
54. **D. Tran**, R. Ranganath, and D. M. Blei. The variational Gaussian process. In *International Conference on Learning Representations*, 2016.
55. **D. Tran**, D. M. Blei, and E. M. Airolidi. Copula variational inference. In *Neural Information Processing Systems*, 2015.

## Open-Source Software

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|---|------|
| 1. Uncertainty Baselines: High-quality implementations of standard and SOTA methods | 2020 |
| 2. Robustness Metrics: Evaluating out-of-distribution generalization                | 2020 |
| 3. Bayesian Layers: A module for neural network uncertainty                         | 2018 |
| 4. Mesh-TensorFlow: Deep learning for supercomputers                                | 2018 |
| 5. Edward2: Simple, distributed, and accelerated probabilistic programming          | 2018 |
| 6. Tensor2Tensor: Library of deep learning models and datasets                      | 2017 |
| 7. Edward: A library for probabilistic modeling, inference, and criticism           | 2016 |
| 8. sgd: An R package for large-scale estimation                                     | 2015 |

9. Stan: A platform for statistical modeling and high-performance statistical computation 2012

## Professional Service

### PROGRAM COMMITTEE

Area Chair: Neural Information Processing Systems	2019–2023
Area Chair: International Conference on Learning Representations	2020–2024
Area Chair: International Conference on Machine Learning	2019–2023
Senior Area Chair: Artificial Intelligence and Statistics	2023
Area Chair: Artificial Intelligence and Statistics	2019–2022
Area Chair: Association for the Advancement of Artificial Intelligence	2020–2023
Senior Program Committee: International Joint Conferences on Artificial Intelligence	2020–2023

### WORKSHOP ORGANIZATION

Symposium: Advances in Approximate Bayesian Inference	2019
Symposium: Advances in Approximate Bayesian Inference	2018
UAI Workshop: Uncertainty in Deep Learning	2018
NIPS Workshop: Advances in Approximate Bayesian Inference	2017
ICML Workshop: Implicit Generative Models	2017
NIPS Workshop: Advances in Approximate Bayesian Inference	2016
NIPS Workshop: Advances in Approximate Bayesian Inference	2015

### PH.D. THESIS COMMITTEE

Junjiao Tian (Georgia Tech)

### MENTORING

Eric Wallace (Google DeepMind Intern, 2023)  
 Kelly Buchanan (Google Student Researcher, 2022)  
 James Allingham (Google Brain Intern, 2021, 2022)  
 Frances Hubis (Google AI Resident, 2021)  
 Archit Karandikar (Google Software Engineer, 2021)  
 Jeremy Nixon (Google Software Engineer, 2020–2021)  
 Yeming Wen (Google Brain Intern, 2019–2020)  
 Aditya Grover (Google AI Resident, 2019)

Jason Lee (Google Brain Intern, 2019–2020)

Ghassen Jerfel (Google Student Researcher, 2019–2021)

Kumar Krishna Agrawal (Google AI Resident, 2019)

Michael W. Dusenberry (Google AI Resident, 2018–2020)

Keyon Vafa (Google Brain Intern, 2018)

Danijar Hafner (Google Brain Intern, 2018)

Akshay Khatri (M.S. Columbia University, 2017)