

# Jonathan H. Huggins

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CONTACT INFORMATION	655 Huntington Avenue Building 2 Boston, MA 02115 USA	✉ <a href="mailto:jhuggins-at-mit-dot-edu">jhuggins -at- mit -dot- edu</a> 🌐 <a href="http://jhhuggins.org">jhhuggins.org</a>	
EDUCATION	<b>Massachusetts Institute of Technology</b> , Cambridge, MA USA Ph.D., Computer Science. Advisor: Tamara Broderick S.M., Computer Science. Advisor: Joshua B. Tenenbaum		2014 - 2018 2012 - 2014
	<b>Columbia University, Columbia College</b> , New York, NY USA B.A., Mathematics. Advisors: Liam Paninski and Frank D. Wood		2008 - 2012
ACADEMIC EXPERIENCE	<b>Harvard University, Department of Biostatistics</b> , Boston, MA USA Postdoctoral Research Fellow. Advisor: Jeffrey Miller		2018 -
	<b>Microsoft Research New England</b> , Cambridge, MA USA Research Intern. Advisor: Lester Mackey		2017
HONORS AND AWARDS	ISBA@NIPS travel award (2016) DoD National Defense Science and Engineering Graduate Fellowship (2013-2015) NSF Graduate Research Fellowship (2013) ( <i>declined for DoD NDSEG</i> ) Hertz Fellowship Finalist (2013) Summa Cum Laude, Columbia University (2012) Phi Beta Kappa (2011) Rabi Scholar, Columbia College (2008-2012) Intel Science Talent Search Finalist (2008)		
PREPRINTS		<ul style="list-style-type: none"><li>• <b>J. H. Huggins</b>, M. Kasprzak, T. C. Campbell &amp; T. Broderick. Practical bounds on the error of Bayesian posterior approximations: A nonasymptotic approach. <i>arXiv:1809.09505 [stat.TH]</i>.</li><li>• <b>J. H. Huggins</b>, T. C. Campbell, M. Kasprzak &amp; T. Broderick. Scalable Gaussian process inference with finite-data mean and variance guarantees. <i>arXiv:1806.10234 [stat.ML]</i>.</li><li>• R. Agrawal, T. C. Campbell, <b>J. H. Huggins</b> &amp; T. Broderick. Data-dependent compression of random features for large-scale kernel approximation.</li></ul>	
PUBLICATIONS		<ol style="list-style-type: none"><li>14. <b>J. H. Huggins</b>* &amp; L. Mackey* (2018). Random feature Stein discrepancies. In <i>Proc. of the 32nd Annual Conference on Neural Information Processing Systems</i>.</li><li>13. T. C. Campbell*, <b>J. H. Huggins</b>*, J. P. How &amp; T. Broderick (To appear). Truncated Random Measures. <i>Bernoulli</i>.</li><li>12. <b>J. H. Huggins</b>* &amp; D. M. Roy* (To appear). Sequential Monte Carlo as approximate sampling: bounds, adaptive resampling via <math>\infty</math>-ESS, and an application to particle Gibbs. <i>Bernoulli</i>.</li><li>11. <b>J. H. Huggins</b>, R. P. Adams &amp; T. Broderick (2017). PASS-GLM: polynomial approximate sufficient statistics for scalable Bayesian GLM inference. In <i>Proc. of the 31st Annual Conference on Neural Information Processing Systems</i>.</li></ol> <p>▷ Selected for spotlight presentation (top 22% of accepted papers)</p>	

10. **J. H. Huggins\*** & J. Zou\* (2017). Quantifying the Accuracy of Approximate Diffusions and Markov Chains. In *Proc. of the 19th International Conference on Artificial Intelligence and Statistics*.
9. **J. H. Huggins**, T. C. Campbell & T. Broderick (2016). Coresets for Scalable Bayesian Logistic Regression. In *Proc. of Advances in Neural Information Processing Systems*.
8. **J. H. Huggins** & J. B. Tenenbaum (2015). Risk and Regret of Hierarchical Bayesian Learners. In *Proc. of the 32nd International Conference on Machine Learning*.
7. **J. H. Huggins\***, A. Saeedi\*, K. Narasimhan\* & V. K. Mansinghka (2015). JUMP-Means: Small-Variance Asymptotics for Markov Jump Processes. In *Proc. of the 32nd International Conference on Machine Learning*.
6. **J. H. Huggins** & C. Rudin (2014). A statistical learning theory framework for supervised pattern discovery. In *Proc. of SIAM International Conference on Data Mining*.
5. A. Pakman, **J. H. Huggins**, C. Smith & L. Paninski (2014). Fast state-space methods for inferring dendritic synaptic connectivity. *Journal of Computational Neuroscience* 36(3), 415-443.
4. E. Pnevmatikakis, K. Rahnama Rad, **J. H. Huggins** & L. Paninski (2014). Fast low-SNR Kalman filtering and forward-backward smoothing via a low-rank perturbative approach. *Journal of Computational and Graphical Statistics* 23(2), 316-339.
3. **J. H. Huggins** & L. Paninski (2012). Optimal experimental design for sampling voltage on dendritic trees in the low-SNR regime. *Journal of Computational Neuroscience* 32(2), 347-66.
2. M. Vilain, **J. H. Huggins** & B. Wellner (2009). Sources of performance in CRF transfer training: a business name-tagging case study. In *Proc. of Recent Advances in Natural Language Processing 2009*.
1. M. Vilain, **J. H. Huggins** & B. Wellner (2009). A simple feature-copying approach to long-distance dependencies. In *Proc. of the 13th Conference on Computational Natural Language Learning 2009*.

\* = contributed equally

#### MISCELLANEA

2. **J. H. Huggins**, A. Saeedi & M. J. Johnson (2014). Detailed Derivations of Small-variance Asymptotics for some Hierarchical Bayesian Nonparametric Models. *arXiv:1501.00052 [stat.ML]*.
1. **J. H. Huggins** & F. Wood (2014). Infinite structured hidden semi-Markov models. *arXiv:1407.0044 [stat.ME]*.

#### INVITED TALKS

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|---|------------|
| ISBA World Meeting (contributed session)<br><i>Scaling Bayesian inference using exponential family approximations</i> | June 2018  |
| SPA 2018<br><i>Finite-dimensional Approximations of Completely Random Measures</i>                                    | June 2018  |
| Boston Bayesian Meetup<br><i>Scaling Bayesian Inference by Constructing Approximating Exponential Families</i>        | April 2018 |
| Schlumberger Doll Research  | April 2018 |

*Scaling Bayesian Inference by Constructing Approximating Exponential Families*

Raytheon BBN Technologies

February 2018

*Scaling Bayesian Inference: Theoretical Foundations and Practical Methods*

CONTRIBUTED  
TALKS

BNP 2017 (contributed talk)  
*Truncated Random Measures*

June 2017

PROFESSIONAL  
SERVICE

*Journal Reviewer:* PLoS One, Journal of Machine Learning Research

*Conference Reviewer:* Advances in Neural Information Processing Systems, International Conference on Machine Learning, Artificial Intelligence and Statistics

TEACHING

*Massachusetts Institute of Technology*

- Teaching Assistant, 6.862 Applied Machine Learning (Graduate-level) 2017
- Guest Lecturer, 6.438 Fundamentals of Probability 2016
- Teaching Assistant, 6.867 Machine Learning (Graduate-level) 2016

*Columbia University*

- Teaching Assistant, Data Structures 2011
- Guest Lecturer, Statistical Analysis of Neural Data (Graduate-level) 2011

PROFESSIONAL  
EXPERIENCE

**Google Inc.**, New York, NY USA  
Summer Engineering Intern

2012

**MITRE Corp.**, Bedford, MA USA  
Technical Co-op

2007 - 2009