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## Education

Ph.D., University of Massachusetts Amherst, Computer Science, 2006  
Dissertation: *Statistical Models and Analysis Techniques for Learning in Relational Data*  
Advisor: David Jensen

M.S., University of Massachusetts Amherst, Computer Science, 2004

B.S. *summa cum laude*, University of Massachusetts Amherst, Computer Science, 2000

## Professional Experience

2021–present: *Senior Principal Research Manager*, Microsoft Research Redmond  
2021–2022: *Senior Principal Researcher*, Microsoft Research Redmond  
2020–present: *Samuel Conte Professor of Computer Science*, Purdue University  
Joint appointment: Computer Science (75%) and Statistics (25%)  
2020 *Visiting Researcher*, Microsoft Research/Microsoft MSAI  
2019–2020: *Professor*, Purdue University  
Joint appointment: Computer Science (75%) and Statistics (25%)  
2014–2019: *Miller Family Chair*, Purdue University  
2013–2019: *Associate Professor*, Purdue University  
Joint appointment: Computer Science (75%) and Statistics (25%)  
2013 *Visiting Scientist*, Simons Institute for Theoretical Computing, UC Berkeley  
2006–2013: *Assistant Professor*, Purdue University  
Joint appointment: Computer Science (75%) and Statistics (25%)  
2000–2006: *Research Assistant*, University of Massachusetts Amherst  
2000–2000: *Research Intern*, AT&T Shannon Laboratory

## Awards and Honors

Program Committee Chair, 37<sup>th</sup> AAAI Conference on Artificial Intelligence, 2023  
Distinguished Lecture, University of California San Diego, Computer Science Department, 2020  
Purdue University Faculty Scholar, 2020  
Keynote Address, 23<sup>rd</sup> Pacific-Asia Conference on Knowledge Discovery and Data Mining, 2019  
Keynote Address, 2<sup>nd</sup> IEEE Data Science Workshop, 2019.  
Keynote Address, 2<sup>nd</sup> Machine Learning in Science and Engineering Forum, Georgia Tech 2019.

Keynote Address, IUPUI Data Science Summit, 2019.

Invited Plenary Panelist: *Future of Artificial Intelligence*,  
33<sup>rd</sup> AAAI Conference on Artificial Intelligence, 2019

Program Committee Chair, 18<sup>th</sup> SIAM International Conference on Data Mining, 2019

Distinguished Lecture, Virginia Tech University, Computer Science Department, 2019

ACM Doctoral Dissertation Award Committee, 2019-2021

Keynote Address, 28<sup>th</sup> Keck Annual Research Conference, 2018

Invited Plenary Panelist: *Societal Impact of Data Science and Artificial Intelligence*,  
ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, 2018

AAAI Executive Council, Elected Councilor, 2015-2018

Keynote Address, 27<sup>th</sup> International Conference on Inductive Logic Programming, 2017

Keynote Address, 3<sup>rd</sup> ACM International Conference on the Theory of Information Retrieval, 2017

Keynote Address, 6<sup>th</sup> International Conference on Complex Networks and Their Applications, 2017

Invited Plenary Panelist *Is Deep Learning the New 42?*,  
ACM SIGKDD Conference on Knowledge Discovery and Data Mining, 2016

Distinguished Lecture, Max Planck Institute of Informatics, 2016

Program Committee Chair, 9<sup>th</sup> ACM International Conference on Web Search and Data Mining, 2016

Keynote Address, 2<sup>nd</sup> European Network Intelligence Conference, 2015

Purdue College of Science Research Award, 2014

Keynote Address, 27<sup>th</sup> International Conference of the Florida Artificial Intelligence Research Society, 2014

Purdue College of Science Graduate Mentoring Award, 2013

Outstanding Achievement by a Young Alum, School of CS, University of Massachusetts Amherst, 2013

NSF Career Award, 2012

Purdue College of Science Team Award, 2012

Purdue Seed for Success Award, 2011-2013

Purdue College of Science Interdisciplinary Award, 2009, 2010

ICDM Best Research Paper Award Runner-Up, 2009

IEEE Intelligent Systems Top Ten to Watch, 2008

Microsoft New Faculty Fellowship Finalist, 2007

DARPA Computer Science Study Panel Member, 2007

Nominated for ACM Doctoral Dissertation Award, University of Massachusetts, 2006

KDD Cup First Place Open Task, 9<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery  
and Data Mining, 2003

Young Investigator Award, DARPA IPTO Cognitive Systems Conference, 2003

AT&T Labs Graduate Fellowship, 2000-2006

National Science Foundation Graduate Research Fellowship, 2000-2003

National Physical Science Consortium Fellowship, 2000 (*declined*)

Bell Labs Graduate Research Fellowship, 2000 (*declined*)

## Publications<sup>1</sup>

### *Journal Articles*

1. Generating Post-Hoc Explanations for Skip-gram-based Node Embeddings by Identifying Important Nodes with Bridgeness  
H. Park, and J. Neville.  
*Neural Networks*, 2023.
2. Ensemble learning for relational data.  
H. Eldardiry, J. Neville, and R. Rossi.  
*Journal of Machine Learning Research*, 21 (1), 1780-1816.
3. Scalable and exact sampling method for probabilistic generative graph models.  
S. Moreno<sup>†</sup>, J. Pfeiffer III<sup>†</sup>, and J. Neville.  
*Data Mining and Knowledge Discovery*, Volume 32, Issue 6, pp 1561-1596, 2018.
4. Designing Size Consistent Statistics for Accurate Anomaly Detection in Dynamic Networks.  
T. La Fond<sup>†</sup>, J. Neville, and B. Gallagher.  
*ACM Transactions on Knowledge Discovery from Data*, 12:14, 2018.
5. Tied Kronecker Product Graph Models to Capture Variance in Network Populations.  
S. Moreno, J. Neville, and S. Kirshner  
*ACM Transactions on Knowledge Discovery from Data*, 12:3, 2018.
6. Graphlet Decomposition: Framework, Algorithms, and Applications.  
N. Ahmed<sup>†</sup>, J. Neville, R. Rossi<sup>†</sup>, N. Duffield, T. Willke  
*Knowledge and Information Systems*, Volume 50, Issue 3, pp 689-722, 2017.
7. Network Sampling: From Static to Streaming Graphs.  
N. Ahmed<sup>†</sup>, J. Neville, and R. Kompella  
*Transactions on Knowledge Discovery and Data Mining*, Vol. 8, Issue 2, 2014.
8. Transforming Graph Data for Statistical Relational Learning.  
R. Rossi<sup>†</sup>, L. McDowell, D. Aha and J. Neville  
*Journal of Artificial Intelligence Research*, Vol. 45, 363-441, 2012.
9. Correcting Evaluation Bias of Relational Classifiers with Network Cross Validation.  
J. Neville, B. Gallagher, T. Eliassi-Rad, and T. Wang  
*Knowledge and Information Systems*, 30-1, 31-55, 2012.
10. Gender demographics trends and changes in U.S. CS departments.  
D. Baumann<sup>†</sup>, S. Hambrusch, and J. Neville.  
*Communications of the ACM*, 54:11, 38-42, 2011.
11. Guided Data Repair  
M. Yakout<sup>†</sup>, A. Elmagarmid, J. Neville, M. Ouzzani, and I. Ilyas  
*Proceedings of the VLDB Endowment*, 2011.
12. Prediction models for long-term Internet prefix availability  
R. Khosla<sup>†</sup>, S. Fahmy, Y. C. Hu, and J. Neville  
*Computer Networks*, 2010.
13. A Bias-Variance Decomposition for Collective Inference Models.  
J. Neville and D. Jensen.  
*Machine Learning Journal*, 73:1, pages 87-106, 2008.

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<sup>1</sup>My authorship policy is to list students authors ahead of me, regardless of contribution. Graduate students are denoted with <sup>†</sup>; undergraduates are denoted with <sup>‡</sup>.

14. Dependency Networks for Relational Data.  
J. Neville and D. Jensen.  
*Journal of Machine Learning Research*, 8(Mar):653–692, 2007.
15. Exploiting Relational Structure to Understand Publication Patterns in High-Energy Physics.  
A. McGovern, L. Friedland, M. Hay, B. Gallagher, A. Fast, J. Neville and D. Jensen.  
*SIGKDD Explorations*, Volume 5, Issue 2, pages 165-172, 2003.

### **Conference Papers**

16. DYANE: DYnamic Attributed Node rolEs Generative Model.  
G. Zeno and J. Neville.  
*Proceedings of the 32<sup>nd</sup> ACM International Conference on Information and Knowledge Management (CIKM)*, 2023.
17. Hindsight Learning for MDPs with Exogenous Inputs.  
S. Sinclair, F. Frujeri, C. Cheng, L. Marshall, H. Barbalho, J. Li, J. Neville, I. Menache, A. Swaminathan.  
*Proceedings of the 40<sup>th</sup> International Conference on Machine Learning (ICML)*, 2023.
18. Stationary Algorithmic Balancing Over Dynamic Email Re-Ranking Problem  
J. Liu<sup>†</sup> and J. Neville.  
*Proceedings of the 29<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2023.
19. Workplace Recommendation with Temporal Network Objectives  
K. Tomlinson, J. Neville, L. Yang, M. Wang, C. Lu  
*Proceedings of the 29<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2023.
20. Expressive and Efficient Representation Learning for Ranking Links in Temporal Graphs  
S. Suresh<sup>†</sup> M. Shrivastava, A. Mukherjee, J. Neville, and P. Li.  
*Proceedings of the 32<sup>nd</sup> ACM Web Conference*, 2023.
21. Adversarial Graph Augmentation to Improve Graph Contrastive Learning  
S. Suresh<sup>†</sup> P. Li, C. Hao, and J. Neville.  
*Advances in Neural Information Processing Systems (NeurIPS)*, 2021.
22. A Collective Learning Framework to Boost GNN Expressiveness for Node Classification.  
M. Hang<sup>†</sup> J. Neville, and B. Ribeiro.  
*Proceedings of the 38<sup>th</sup> International Conference on Machine Learning (ICML)*, 2021.
23. Breaking the Limit of Graph Neural Networks by Improving the Assortativity of Graphs with Local Mixing Patterns  
S. Suresh<sup>†</sup> V. Budde, J. Neville, P. Li, J. Ma.  
*Proceedings of the 27<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2021.
24. Towards Decentralized Social Reinforcement Learning via Ego-Network Extrapolation  
M. Goindani<sup>†</sup> and J. Neville.  
*Proceedings of the 20<sup>th</sup> International Conference on Autonomous Agents and Multi-Agent Systems (AAMAS)*, 2021.
25. Dymond: Dynamic motif-nodes network generative model  
G. Zeno<sup>†</sup> T. La Fond, and J. Neville.  
*Proceedings of the 30<sup>th</sup> International Web Conference*, 2021.

26. A Hybrid Model for Learning Embeddings and Logical Rules Simultaneously from Knowledge Graphs  
S. Suresh<sup>†</sup> and J. Neville.  
*Proceedings of the 20<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 6 pages, 2020.  
(Acceptance rate: 20%)
27. Online Bayesian Sparse Learning with Spike and Slab Priors  
S. Fang, S. Zhe, K. Lee, K. Zhang, and J. Neville.  
*Proceedings of the 20<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 8 pages, 2020.  
(Acceptance rate: 10%)
28. MERL: Multi-View Edge Representation Learning in Social Networks  
Y. Lai<sup>†</sup>, and J. Neville.  
*Proceedings of the 29<sup>th</sup> ACM International Conference on Information and Knowledge Management (CIKM)*, 10 pages, 2020. (Acceptance rate: 21%)
29. Role Equivalence Attention for Label Propagation in Graph Neural Networks  
H. Park<sup>†</sup> and J. Neville  
*Proceedings of the 24<sup>th</sup> Pacific-Asia Conference on Knowledge Discovery and Data Mining (PAKDD)*, 12 pages, 2020. (Acceptance rate: 21%)
30. Cluster-Based Social Reinforcement Learning  
M. Goindani<sup>†</sup> and J. Neville.  
*Proceedings of the 19<sup>th</sup> International Conference on Autonomous Agents and Multi-Agent Systems (AAMAS)*, 3 pages, 2020. (Acceptance rate: 40%)
31. ReadNet: A Hierarchical Transformer Framework for Web Article Readability Analysis  
C. Meng<sup>†</sup>, M. Chen, J. Mao, and J. Neville  
*Proceedings of the 42<sup>nd</sup> European Conference on Information Retrieval (ECIR)*, 15 pages, 2020.  
(Acceptance rate: 23%)
32. Exploiting Interaction Links for Node Classification with Deep Graph Neural Networks  
H. Park<sup>†</sup> and J. Neville  
*Proceedings of the 28<sup>th</sup> International Joint Conference on Artificial Intelligence (IJCAI)*, 8 pages, 2019.  
(Acceptance rate: 18%)
33. Learning How to Intervene in True News Diffusion to Combat Fake News Spread.  
M. Goindani<sup>†</sup> and J. Neville.  
*Proceedings of the 35<sup>th</sup> Conference on Uncertainty in Artificial Intelligence (UAI)*, 10 pages, 2019.  
(Acceptance rate: 26%)
34. HATS: A Hierarchical Sequence-Attention Framework for Inductive Set-of-Sets Embeddings.  
C. Meng<sup>†</sup>, J. Yang<sup>†</sup>, B. Ribeiro, and J. Neville.  
*Proceedings of the 25<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 10 pages, 2019. (Acceptance rate: 9%)
35. A Stein-Papangelou Goodness-of-Fit Test for Point Processes.  
J. Yang<sup>†</sup>, V. Rao and J. Neville.  
*Proceedings of the 23<sup>rd</sup> International Conference on Artificial Intelligence and Statistics (AISTAT)*, 10 pages, 2019. (Acceptance rate: 32%)
36. TransConv: Relationship Embedding in Social Networks.  
Y. Lai<sup>†</sup>, J. Neville, and D. Goldwasser.  
*Proceedings of the 33<sup>rd</sup> Conference on Artificial Intelligence (AAAI)*, 9 pages, 2019.  
(Acceptance rate: 16%)
37. Multi-level hypothesis testing for populations of heterogeneous networks  
G. Gomes<sup>†</sup>, J. Neville, and V. Rao.  
*Proceedings of the 18<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 6 pages, 2018.  
(Acceptance rate: 20%)

38. The Indian Buffet Hawkes Process to Model Evolving Latent Influences.  
X. Tan<sup>†</sup>, V. Rao and J. Neville.  
*Proceedings of the 34<sup>th</sup> Conference on Uncertainty in Artificial Intelligence (UAI)*, 10 pages, 2018.  
(Acceptance rate: 31%)
39. Exploring Student Check-In Behavior for Improved Point-of-Interest Prediction.  
M. Hang<sup>†</sup>, I. Pytlarz, and J. Neville.  
*Proceedings of the 24<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 10 pages, 2018. (Acceptance rate: 22%)
40. Goodness-of-fit Testing for Discrete Distributions via Stein Discrepancy.  
J. Yang<sup>†</sup>, Q. Liu, V. Rao and J. Neville.  
*Proceedings of the 35<sup>th</sup> International Conference on Machine Learning (ICML)*, 9 pages, 2018.  
(Acceptance rate: 25%)
41. Nested CRP with Hawkes-Gaussian Processes.  
X. Tan<sup>†</sup>, V. Rao and J. Neville.  
*Proceedings of the 22<sup>nd</sup> International Conference on Artificial Intelligence and Statistics (AISTAT)*, 10 pages, 2018. (Acceptance rate: 33%)
42. Subgraph Pattern Neural Networks for High-Order Graph Evolution Prediction.  
C. Meng<sup>†</sup>, C. Mouli<sup>†</sup>, B. Ribeiro, and J. Neville.  
*Proceedings of the 32<sup>nd</sup> Conference on Artificial Intelligence (AAAI)*, 9 pages, 2018.  
(Acceptance rate: 24%)
43. Decoupling Homophily and Reciprocity with Latent Space Network Models  
J. Yang<sup>†</sup>, V. Rao and J. Neville  
*Proceedings of the 33<sup>rd</sup> Conference on Uncertainty in Artificial Intelligence (UAI)*, 10 pages, 2017.  
(Acceptance rate: 31%)
44. Unified Representation and Lifted Sampling for Generative Models of Social Networks  
P. Robles<sup>†</sup>, S. Moreno, and J. Neville  
*Proceedings of the 26<sup>th</sup> International Joint Conference on Artificial Intelligence (IJCAI)*, 8 pages, 2017.  
(Acceptance rate: 25%)
45. Should We Be Confident in Peer Effects Estimated From Partial Crawls of Social Networks?  
J. Yang<sup>†</sup>, B. Ribeiro and J. Neville  
*Proceedings of the 11<sup>th</sup> International AAAI Conference on Weblogs and Social Media (ICWSM)*, 4 pages, 2017. Acceptance rate: 19%)
46. Deep Collective Inference  
J. Moore<sup>†</sup> and J. Neville  
*Proceedings of the 31<sup>st</sup> Conference on Artificial Intelligence (AAAI)*, 9 pages, 2017.  
(Acceptance rate: 21%)
47. Sampling of Attributed Networks from Hierarchical Generative Models  
P. Robles<sup>†</sup>, S. Moreno, and J. Neville  
*Proceedings of the 22<sup>nd</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 10 pages, 2016. (Acceptance rate: 9%)
48. Efficient Graphlet Counting for Large Networks  
N. Ahmed<sup>†</sup>, J. Neville, R. Rossi<sup>†</sup>, and N. Duffield.  
*Proceedings of the 15<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 10 pages, 2015.  
(Acceptance rate: 8%)
49. Overcoming Relational Learning Biases to Accurately Predict Preferences in Large Scale Networks  
J. Pfeiffer III<sup>†</sup>, J. Neville, and P. Bennett  
*Proceedings of the 24<sup>th</sup> International World Wide Web Conference (WWW)*, 11 pages, 2015.  
(Acceptance rate: 14%)

50. Incorporating Assortativity and Degree Dependence into Scalable Network Models  
S. Mussmann<sup>‡</sup>, J. Moore<sup>‡</sup>, J. Pfeiffer III<sup>†</sup>, and J. Neville  
*Proceedings of the 29<sup>th</sup> Conference on Artificial Intelligence (AAAI)*, 9 pages, 2015.  
(Acceptance rate: 12%)
51. Composite Likelihood Data Augmentation for Within-Network Statistical Relational Learning  
J. Pfeiffer III<sup>†</sup>, J. Neville, and P. Bennett  
*Proceedings of the 14<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 10 pages, 2014.  
(Acceptance rate: 10%)
52. A Scalable Method for Accurate Sampling from Kronecker Models  
S. Moreno, J. Pfeiffer III<sup>†</sup>, and J. Neville  
*Proceedings of the 14<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 10 pages, 2014.  
(Acceptance rate: 10%)
53. Active Exploration in Networks: Using Probabilistic Relationships for Learning and Inference  
J. Pfeiffer III<sup>†</sup>, J. Neville, and P. Bennett  
*Proceedings of the 23<sup>rd</sup> ACM International Conference on Information and Knowledge Management (CIKM)*, 10 pages, 2014. (Acceptance rate: 21%)
54. Graph Sample and Hold: A Framework for Big-Graph Analytics  
N. Ahmed<sup>†</sup>, N. Duffield, J. Neville, and R. Kompella  
*Proceedings of the 20<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 10 pages, 2014. (Acceptance rate: 15%)
55. Attributed Graph Models: Modeling network structure with correlated attributes  
J. Pfeiffer III<sup>†</sup>, S. Moreno<sup>†</sup>, T. La Fond<sup>†</sup>, J. Neville, and B. Gallagher  
*Proceedings of the 23<sup>rd</sup> International World Wide Web Conference (WWW)*, 11 pages, 2014.  
(Acceptance rate: 13%)
56. Network Hypothesis Testing Using Mixed Kronecker Product Graph Models  
S. Moreno<sup>†</sup> and J. Neville  
*Proceedings of the 13<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 6 pages, 2013.  
(Acceptance rate: 19%)
57. Learning Mixed Kronecker Product Graph Models with Simulated Method of Moments  
S. Moreno<sup>†</sup>, J. Neville and S. Kirshner  
*Proceedings of the 19<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 9 pages, 2013. (Acceptance rate: 17%)
58. Collective Inference for Network Data with Copula Latent Markov Networks  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the 6<sup>th</sup> ACM International Conference on Web Search and Data Mining (WSDM)*, 10 pages, 2013. (Acceptance rate: 19%)
59. Modeling Dynamic Behavior in Large Evolving Graphs  
R. Rossi<sup>†</sup>, B. Gallagher, J. Neville and K. Henderson  
*Proceedings of the 6<sup>th</sup> ACM International Conference on Web Search and Data Mining (WSDM)*, 10 pages, 2013. (Acceptance rate: 19%)
60. An Analysis of How Ensembles of Collective Classifiers Improve Predictions in Graphs  
H. Eldardiry<sup>†</sup> and J. Neville  
*Proceedings of the 21<sup>st</sup> ACM International Conference on Information and Knowledge Management (CIKM)*, 10 pages, 2012. (Acceptance rate: 13%)
61. Fast Generation of Large Scale Social Networks While Incorporating Transitive Closures  
J. Pfeiffer III<sup>†</sup>, T. La Fond<sup>†</sup>, S. Moreno<sup>†</sup>, and J. Neville  
*Proceedings of the 4<sup>th</sup> ASE/IEEE International Conference on Social Computing (SocialCom)*, 12 pages, 2012. (Acceptance rate: 10%)

62. The Impact of Communication Structure and Interpersonal Dependencies on Distributed Teams  
T. La Fond<sup>†</sup>, D. Roberts<sup>‡</sup>, J. Neville, J. Tyler, and S. Connaughton  
*Proceedings of the 4<sup>th</sup> ASE/IEEE International Conference on Social Computing (SocialCom)*, 8 pages, 2012. (Acceptance rate: 17%)
63. Network Sampling Designs for Relational Classification  
N. Ahmed<sup>†</sup>, J. Neville, and R. Kompella  
*Proceedings of the 6<sup>th</sup> International AAAI Conference on Weblogs and Social Media (ICWSM)*, 4 pages, 2012. (Acceptance rate: 26%)
64. Time-Evolving Relational Classification and Ensemble Methods  
R. Rossi<sup>†</sup> and J. Neville  
*Proceedings of the 16<sup>th</sup> Pacific-Asia Conference on Knowledge Discovery and Data Mining (PAKDD)*, 12 pages, 2012. (Acceptance rate: 36%)
65. Structured Comparative Analysis of Systems Logs to Diagnose Performance Problems  
K. Nagaraj<sup>†</sup>, C. Killian, and J. Neville.  
*Proceedings of the 9<sup>th</sup> USENIX Symposium on Networked Systems Design and Implementation (NSDI)*, 2012. (Acceptance rate: 18%)
66. Understanding Propagation Error and Its Effect on Collective Classification  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the 11<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 10 pages, 2011. (Acceptance rate, full paper: 12%)
67. Correcting Bias in Statistical Tests for Network Classifier Evaluation  
T. Wang, J. Neville, B. Gallagher, and T. Eliassi-Rad  
*Proceedings of the 21<sup>st</sup> European Conference on Machine Learning (ECML)*, 16 pages, 2011. (Acceptance rate: 20%)
68. Relational Active Learning for Joint Collective Classification Models  
A. Kuwadekar<sup>‡</sup> and J. Neville  
*Proceedings of the 28<sup>th</sup> International Conference on Machine Learning (ICML)*, 8 pages, 2011. (Acceptance rate: 25%)
69. Across-Model Collective Ensemble Classification  
H. Eldardiry<sup>†</sup> and J. Neville  
*Proceedings of the 25<sup>th</sup> Conference on Artificial Intelligence (AAAI)*, 7 pages, 2011. (Acceptance rate: 25%)
70. Methods to Determine Node Centrality and Clustering in Graphs with Uncertain Structure  
J. Pfeiffer III<sup>†</sup> and J. Neville  
*Proceedings of the 5<sup>th</sup> International AAAI Conference on Weblogs and Social Media (ICWSM)*, 4 pages, 2011.
71. Relational Learning with One Network: An Asymptotic Analysis  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the 14<sup>th</sup> International Conference on Artificial Intelligence and Statistics (AISTAT)*, 11 pages, 2011. (Acceptance rate, oral presentation: 8%)
72. ERACER: A Database Approach for Statistical Inference and Data Cleaning  
C. Mayfield<sup>†</sup>, J. Neville, and S. Prabhakar  
*Proceedings of the 2010 ACM SIGMOD Conference (SIGMOD)*, 12 pages, 2010. (Acceptance rate: 19%)
73. Predicting Prefix Availability in the Internet  
R. Khosla<sup>†</sup>, S. Fahmy, C. Hu, and J. Neville  
*Proceedings of the 29<sup>th</sup> IEEE Conference on Computer Communications (INFOCOM) Mini-Conference*, 5 pages, 2010. (Acceptance rate: 24%)



74. Randomization tests for distinguishing social influence and homophily effects  
T. LaFond<sup>†</sup> and J. Neville  
*Proceedings of the 19<sup>th</sup> International World Wide Web Conference (WWW)*, 10 pages, 2010.  
(Acceptance rate: 14%)
75. Modeling Relationship Strength in Online Social Networks  
R. Xiang<sup>†</sup>, J. Neville, and M. Rogati  
*Proceedings of the 19<sup>th</sup> International World Wide Web Conference (WWW)*, 10 pages, 2010.  
(Acceptance rate: 14%)
76. Tied Kronecker Product Graph Models to Capture Variance in Network Populations  
S. Moreno<sup>†</sup>, S. Kirshner, J. Neville and S.V.N. Vishwanathan  
*Proceedings of the 48<sup>th</sup> Annual Allerton Conference on Communications, Control and Computing*, 8 pages, 2010.
77. Using Transactional Information to Predict Link Strength in Online Social Networks  
I. Kahanda<sup>†</sup> and J. Neville  
*Proceedings of the 3<sup>rd</sup> International AAAI Conference on Weblogs and Social Media (ICWSM)*, 8 pages, 2009. (Acceptance rate: 17%)
78. Evaluating Statistical Tests for Within-Network Classifiers of Relational Data.  
J. Neville, B. Gallagher, and T. Eliassi-Rad. **Best Paper Award Runner-Up**  
*Proceedings of the 9<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 10 pages, 2009.  
(Acceptance rate, full paper: 9%)
79. Temporal-Relational Classifiers for Prediction in Evolving Domains.  
U. Sharan<sup>†</sup> and J. Neville  
*Proceedings of the 8<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 10 pages, 2008.  
(Acceptance rate, full paper: 10%)
80. A Shrinkage Approach for Modeling Non-Stationary Relational Autocorrelation.  
P. Angin<sup>†</sup> and J. Neville  
*Proceedings of the 8<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 6 pages, 2008.  
(Acceptance rate, short paper: 20%)
81. Pseudolikelihood EM for Within-Network Relational Learning.  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the 8<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, 6 pages, 2008.  
(Acceptance rate, short paper: 20%)
82. Database support for probabilistic attributes and tuples.  
S. Singh<sup>†</sup>, C. Mayfield<sup>†</sup>, R. Shah, S. Prabhakar, S. Hambrusch, J. Neville, R. Cheng.  
*The 24<sup>th</sup> International Conference on Data Engineering (ICDE)*, 9 pages, 2008.  
(Acceptance rate: 19%)
83. Bias/Variance Analysis for Relational Domains.  
J. Neville and D. Jensen.  
*The 17<sup>th</sup> International Conference on Inductive Logic Programming, Lecture Notes in Artificial Intelligence 4894 (ILP)*, pages 27-28, 2007. (Acceptance rate: 34%)
84. Leveraging Relational Autocorrelation with Latent Group Models.  
J. Neville and D. Jensen.  
*Proceedings of the 5<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, pages 322-329, 2005.  
(Acceptance rate: 11%)
85. Using Relational Knowledge Discovery to Prevent Securities Fraud.  
J. Neville, O. Simsek, D. Jensen, J. Komoroske, K. Palmer and H. Goldberg.  
*Proceedings of the 11<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD)*, pages 449-458, 2005. (Acceptance rate: 19%)

86. Dependency Networks for Relational Data.  
J. Neville and D. Jensen.  
*Proceedings of the 4<sup>th</sup> IEEE International Conference on Data Mining (ICDM)*, pages 170-177, 2004.  
(Acceptance rate: 9%)
87. Why Collective Inference Improves Relational Classification.  
D. Jensen, J. Neville and B. Gallagher.  
*Proceedings of the 10<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD)*, pages 593-598, 2004. (Acceptance rate: 25%)
88. Simple Estimators for Relational Bayesian Classifiers.  
J. Neville, D. Jensen and B. Gallagher.  
*Proceedings of the 3<sup>rd</sup> IEEE International Conference on Data Mining (ICDM)*, pages 609-612, 2003.  
(Acceptance rate: 23%)
89. Learning Relational Probability Trees.  
J. Neville, D. Jensen, L. Friedland and M. Hay.  
*Proceedings of the 9<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD)*, pages 625-630, 2003. (Acceptance rate: 27%)
90. Avoiding Bias When Aggregating Relational Data with Degree Disparity.  
D. Jensen, J. Neville and M. Hay.  
*Proceedings of the 20<sup>th</sup> International Conference on Machine Learning (ICML)*, pages 274-281, 2003.  
(Acceptance rate: 32%)
91. Autocorrelation and Linkage Cause Bias in Evaluation of Relational Learners.  
D. Jensen and J. Neville.  
*Proceedings of the 12<sup>th</sup> International Conference on Inductive Logic Programming (ILP)*, pages 101-116, 2002. (Acceptance rate: 54%)
92. Linkage and Autocorrelation Cause Feature Selection Bias in Relational Learning.  
D. Jensen and J. Neville.  
*Proceedings of the 19<sup>th</sup> International Conference on Machine Learning (ICML)*, pages 259-266, 2002.  
(Acceptance rate: 33%)

### **Referred Workshop Papers<sup>2</sup>**

93. Creating Generalizable Downstream Graph Models with Random Projections.  
A. Amirov, C. Quirk, J. Neville.  
*Proceedings of the Workshop on Machine Learning on Graphs, (WSDM)*, 2023.
94. Lightweight Compositional Embeddings for Incremental Streaming Recommendation  
M. Hang<sup>†</sup>, T. Schnabel, L. Yang, and J. Neville.  
*Proceedings of the Workshop on Machine Learning on Graphs, (WSDM)*, 2022.
95. Social Reinforcement Learning  
M. Goindani<sup>†</sup> and J. Neville  
*Proceedings of the AAAI Symposium on Challenges and Opportunities for Multi-Agent Reinforcement Learning, (COMARL)*, 4 pages, 2020.
96. Dynamic Network Modeling from Motif-Activity  
G. Zeno<sup>†</sup>, T. La Fond, and J. Neville  
*Proceedings of the 4<sup>th</sup> International Workshop on Mining Actionable Insights from Social Networks, WWW*, 8 pages, 2020.

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<sup>2</sup>Note: In data mining and machine learning, workshops are a key venue to present and discuss preliminary work at the top conferences. Workshop papers are peer-reviewed and acceptance rates are typically in the 25-75% range.

97. Joint Embedding Models for Textual and Social Analysis  
C. Li<sup>†</sup>, Y. Lai<sup>†</sup>, D. Goldwasser, and J. Neville  
*Proceedings of the 1<sup>st</sup> Workshop on Deep Structured Prediction*, (ICML), 5 pages, 2017.
98. Stochastic Gradient Descent for Relational Logistic Regression via Partial Network Crawls  
J. Yang<sup>†</sup>, B. Ribeiro and J. Neville  
*Proceedings of the 7<sup>th</sup> International Workshop on Statistical Relational AI*, UAI, 7 pages, 2017.
99. Deep Dynamic Relational Classifiers: Exploiting Dynamic Neighborhoods in Complex Networks  
H. Park<sup>†</sup>, J. Moore<sup>†</sup>, and J. Neville  
*Proceedings of the Mining Actionable Insights from Social Networks Workshop*, WSDM, 7 pages, 2017.
100. Online Spike-and-slab Inference with Stochastic Expectation Propagation  
S. Zhe<sup>†</sup>, K. Lee, K. Zhang, and J. Neville  
*Proceedings of the 2016 Workshop on Advances in Approximate Bayesian Inference*, NIPS, 2016.
101. Generating Local Explanations of Network Anomalies via Score Decomposition  
T. La Fond<sup>†</sup>, J. Neville, and B. Gallagher  
*Proceedings of the ODD 4.0: Outlier Definition, Detection, and Description on Demand*, KDD, 6 pages, 2016.
102. Investigating the Impact of Graph Structure and Attribute Correlation on Collective Classification Performance  
G. Zeno<sup>†</sup> and J. Neville  
*Proceedings of the 13<sup>th</sup> Workshop on Mining and Learning with Graphs*, KDD, 8 pages, 2016.
103. Combining Gradient Boosting Machines with Collective Inference to Predict Continuous Values  
I. Alodah<sup>†</sup> and J. Neville  
*Proceedings of the 6<sup>th</sup> International Workshop on Statistical Relational AI*, IJCAI, 7 pages, 2016.
104. Better Together: Combining Language and Social Interactions into a Shared Representation  
Y. Lai<sup>†</sup>, C. Li<sup>†</sup>, D. Goldwasser, and J. Neville  
*Proceedings of the TextGraphs 2016*, (NAACL), 5 pages, 2016.
105. Analyzing the Transferability of Collective Inference Models Across Networks  
R. Niu<sup>‡</sup>, S. Moreno and J. Neville  
*Proceedings of the International Workshop on Information Analysis and Data Mining Over Social Network*, (ICDM), 9 pages, 2015.
106. Using Bayesian Network Representations for Effective Sampling from Generative Network Models  
P. Robles<sup>†</sup>, S. Moreno, and J. Neville  
*Proceedings of the 5<sup>th</sup> International Workshop on Statistical Relational AI*, UAI, 6 pages, 2015.
107. Assortativity in Chung Lu Random Graph Models  
S. Mussmann<sup>‡</sup>, J. Moore<sup>‡</sup>, J. Pfeiffer III<sup>†</sup>, and J. Neville  
*Proceedings of the 8th SNA-KDD Workshop*, KDD, KDD, 8 pages, 2014.
108. Anomaly Detection in Networks with Changing Trends  
T. La Fond<sup>†</sup>, J. Neville, and B. Gallagher  
*Proceedings of the Outlier Detection & Description under Data Diversity Workshop*, KDD, 10 pages, 2014.
109. Block Kronecker Product Graph Models  
S. Moreno<sup>†</sup>, P. Robles<sup>†</sup>, and J. Neville  
*Proceedings of the 11<sup>th</sup> Workshop on Mining and Learning with Graphs*, KDD, 6 pages, 2013.
110. Combining Active Sampling with Parameter Estimation and Prediction in Single Networks  
J. Pfeiffer III<sup>†</sup>, J. Neville, and P. Bennett  
*Proceedings of the Structured Learning: Inferring Graphs from Structured and Unstructured Inputs Workshop*, ICML, 6 pages, 2013.

111. Space-Efficient Sampling from Social Activity Streams  
N. Ahmed<sup>†</sup>, J. Neville, and R. Kompella  
Space-Efficient Sampling from Social Activity Streams. *Proceedings of the 1<sup>st</sup> International Workshop on Big Data, Streams and Heterogeneous Source Mining, KDD*, 8 pages, 2012.
112. Using Latent Communication Styles to Predict Individual Characteristics  
J. Bates<sup>‡</sup>, J. Neville, and J. Tyler  
*Proceedings of the 3<sup>rd</sup> Workshop on Social Media Analytics, KDD*, 8 pages, 2012.
113. Active Sampling of Networks  
J. Pfeiffer III<sup>†</sup>, J. Neville, and P. Bennett  
*Proceedings of the 10<sup>th</sup> Workshop on Mining and Learning with Graphs, ICML*, 8 pages, 2012.
114. On the Mismatch Between Learning and Inference for Single Network Domains  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the Workshop on Inferning: Interactions between Inference and Learning, ICML*, 6 pages, 2012.
115. Role-Dynamics: Fast Mining of Large Dynamic Networks  
R. Rossi<sup>†</sup>, Brian Gallagher, J. Neville, and Keith Henderson  
*Proceedings of the 1<sup>st</sup> International Workshop on Large Scale Network Analysis, WWW*, 9 pages, 2012.
116. Understanding Propagation Error and Its Effect on Collective Classification  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the 9th Workshop on Mining and Learning with Graphs, KDD*, 8 pages, 2011.
117. Modeling the Variance of Network Populations with Mixed Kronecker Product Graph Models  
S. Moreno<sup>†</sup>, J. Neville, S. Kirshner and S.V.N. Vishwanathan. **Most Promising Paper Award**  
*Proceedings of the Workshop on Analyzing Networks and Learning with Graphs, 24<sup>th</sup> Annual Conference on Neural Information Processing Systems*, 8 pages, 2010.
118. Reconsidering the Foundations of Network Sampling  
N. Ahmed<sup>†</sup>, J. Neville, and R. Kompella  
*Proceedings of the 2nd Workshop on Information in Networks*, 5 pages, 2010.
119. Time-Based Sampling of Social Network Activity Graphs  
N. Ahmed<sup>†</sup>, F. Berchmans<sup>†</sup>, J. Neville, and R. Kompella  
*Proceedings of the 8th Workshop on Mining and Learning with Graphs, KDD*, 8 pages, 2010.
120. Multi-Network Fusion for Collective Inference  
H. Eldardiry<sup>†</sup> and J. Neville  
*Proceedings of the 8th Workshop on Mining and Learning with Graphs, KDD*, 8 pages, 2010.
121. Probabilistic Paths and Centrality in Time  
J. Pfeiffer III<sup>†</sup> and J. Neville  
*Proceedings of the 4th SNA-KDD Workshop, KDD*, 8 pages, 2010.
122. Combining Semi-supervised Learning and Relational Resampling for Active Learning in Network Domains  
A. Kuwadekar<sup>‡</sup> and J. Neville. **Best Paper Award**  
*Proceedings of the Budgeted Learning Workshop, ICML*, 8 pages, 2010.
123. Modeling the Evolution of Discussion Topics and Communication to Improve Relational Classification  
R. Rossi<sup>†</sup> and J. Neville  
*Proceedings of the 1st Workshop on Social Media Analytics, KDD*, 8 pages, 2010.
124. Ranking for Data Repairs  
M. Yakout<sup>†</sup>, A. Elmagarmid, and J. Neville  
*Proceedings of the 4<sup>th</sup> International Workshop on Ranking in Databases, ICDE 2010*, 6 pages, 2010.

125. Modeling Relationship Strength in Online Social Networks.  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the Workshop on Analyzing Networks and Learning with Graphs, 23<sup>rd</sup> Annual Conference on Neural Information Processing Systems*, 8 pages, 2009.
126. An Investigation of the Distributional Characteristics of Generative Graph Models.  
S. Moreno<sup>†</sup> and J. Neville.  
*Proceedings of the 1<sup>st</sup> Workshop on Information in Networks*, 5 pages, 2009.
127. A Shrinkage Approach for Modeling Non-Stationary Relational Autocorrelation.  
P. Angin<sup>†</sup> and J. Neville.  
*Proceedings of the 2<sup>nd</sup> Social Network Analysis Workshop, 14<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining*, 6 pages, 2008.
128. A Resampling Technique for Relational Data Graphs.  
H. Eldardiry<sup>†</sup> and J. Neville.  
*Proceedings of the 2<sup>nd</sup> Social Network Analysis Workshop, 14<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining*, 6 pages, 2008.
129. Pseudolikelihood EM for Within-Network Relational Learning.  
R. Xiang<sup>†</sup> and J. Neville  
*Proceedings of the 2<sup>nd</sup> Social Network Analysis Workshop, 14<sup>th</sup> ACM SIGKDD Conference on Knowledge Discovery and Data Mining*, 8 pages, 2008.
130. Exploiting Time-Varying Relationships in Statistical Relational Models.  
U. Sharan<sup>†</sup> and J. Neville.  
*Proceedings of the 1<sup>st</sup> Social Network Analysis KDD Workshop, 13<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery and Data Mining*, 7 pages, 2007.
131. Bias/Variance Analysis for Network Data.  
J. Neville and D. Jensen.  
*Proceedings of the Workshop on Statistical Relational Learning, 23<sup>rd</sup> International Conference on Machine Learning*, 8 pages, 2006.
132. Structure Learning for Statistical Relational Models.  
J. Neville.  
*Proceedings of the 20<sup>th</sup> National Conference on Artificial Intelligence (Doctoral Consortium)*, pages 1656-1657, 2005.
133. Autocorrelation and Relational Learning: Challenges and Opportunities.  
J. Neville, O. Simsek and D. Jensen.  
*Proceedings of the Workshop on Statistical Relational Learning, 21<sup>st</sup> International Conference on Machine Learning*, 8 pages, 2004.
134. Collective Classification with Relational Dependency Networks. J. Neville and D. Jensen.  
*Proceedings of the 2<sup>nd</sup> Multi-Relational Data Mining Workshop, 9<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery and Data Mining*, pages 77-91, 2003.
135. Statistical Relational Learning: Four Claims and a Survey.  
J. Neville, M. Rattigan and D. Jensen.  
*Proceedings of the Workshop on Learning Statistical Models from Relational Data, 18<sup>th</sup> International Joint Conference on Artificial Intelligence*, 5 pages, 2003.
136. Clustering Relational Data Using Attribute and Link Information.  
J. Neville, M. Adler and D. Jensen.  
*Proceedings of the Text Mining and Link Analysis Workshop, 18<sup>th</sup> International Joint Conference on Artificial Intelligence*, 6 pages, 2003.

137. Schemas and Models.  
D. Jensen and J. Neville.  
*Proceedings of the Multi-Relational Data Mining Workshop, 8<sup>th</sup> ACM SIGKDD International Conference on Knowledge Discovery and Data Mining*, 15 pages, 2002.
138. Supporting Relational Knowledge Discovery: Lessons in Architecture and Algorithm Design.  
J. Neville and D. Jensen.  
*Proceedings of the Data Mining Lessons Learned Workshop, 19<sup>th</sup> International Conference on Machine Learning*, pages 57-64, 2002.
139. Correlation and Sampling in Relational Data Mining.  
D. Jensen and J. Neville.  
*Proceedings of the 33<sup>rd</sup> Symposium on the Interface of Computing Science and Statistics*, 14 pages, 2001.
140. Iterative Classification in Relational Data.  
J. Neville and D. Jensen.  
*Proceedings of the Workshop on Learning Statistical Models from Relational Data, 17<sup>th</sup> National Conference on Artificial Intelligence*, pages 42-49, 2000.

### **Book Chapters**

141. Relational Dependency Networks.  
J. Neville and D. Jensen.  
*Introduction to Statistical Relational Learning*, L. Getoor and B. Taskar, editors, pages 239-268, 2007.

### **Invited Papers**

142. Data Mining in Social Networks.  
D. Jensen and J. Neville.  
*National Academy of Sciences Symposium on Dynamic Social Network Analysis*, 13 pages, 2002.

### **Technical Reports**

143. Identifying User Survival Types via Clustering of Censored Social Network Data.  
C. Mouli<sup>†</sup>, A. Naik<sup>‡</sup>, B. Ribeiro, and J. Neville.  
ArXiv e-prints, 1703.03401, 2017.
144. Learning the Latent State Space of Time-Varying Graphs.  
N. Ahmed<sup>†</sup>, C. Cole<sup>‡</sup>, and J. Neville.  
ArXiv e-prints, 1403.3707, 2014.
145. Network Sampling via Edge-based Node Selection with Graph Induction.  
N. Ahmed<sup>†</sup>, J. Neville, R. Kompella.  
Purdue University, CSD TR 11-0016, 2011.
146. Correcting Bias in Statistical Tests for Network Classifier Evaluation.  
J. Neville, T. Wang, B. Gallagher, and T. Eliasi-Rad.  
Purdue University, CSD TR 10-012, 2010.
147. Spectral Clustering with Links and Attributes.  
J. Neville, M. Adler and D. Jensen.  
University of Massachusetts Amherst, Technical Report 04-42, 2004.
148. Randomization Tests for Relational Learning.  
D. Jensen, J. Neville and M. Rattigan.  
University of Massachusetts Amherst, Technical Report 03-05, 2003.

## Invited Presentations

- Towards Relational AI: The good, the bad, and the ugly of learning over networks (**distinguished lecture**), University of California San Diego, Computer Science Department, 2020.
- Learning Over Sets, Subgraphs, and Streams (**invited speaker**), Microsoft Research, 2020.
- Learning Over Sets, Subgraphs, and Streams (**invited speaker**), From Local to Global Information Workshop, NSF Center for Science of Information, 2020.
- Natural Intelligence vs. Artificial Intelligence (**invited panelist**), Purdue Engineering Distinguished Lecture Panel Discussion w/James DiCarlo, 2019.
- Analyzing Behavioral Traces to Improve Student Outcomes (**invited speaker**), Microsoft Faculty Summit, Microsoft Research, Bellevue, Washington, 2019.
- Towards Relational AI: The good, the bad, and the ugly of learning over networks (**invited keynote**), IEEE Data Science Workshop, Minneapolis, Minnesota, 2019.
- Towards Relational AI: The good, the bad, and the ugly of learning over networks (**invited keynote**), MLSE Data Science Forum, Georgia Tech Institute for Data Engineering and Science, 2019.
- Deep Learning for Relational Networks (**invited keynote**), IUPUI Data Science Summit, IUPUI, Indianapolis, Indiana, 2019.
- Towards Relational AI: The good, the bad, and the ugly of learning over networks (**invited keynote**), Pacific-Asia Conference on Knowledge Discovery and Data Mining, Macau, China, 2019.
- Statistical methods for prediction and anomaly detection in dynamic networks (**invited speaker**), Foundations of Data Science Workshop, Purdue University, 2019.
- Can we trust AI? (**invited panelist**), Purdue 150<sup>th</sup> Celebration Ideas Festival Panel Discussion w/Vint Cerf, 2019.
- Towards Relational AI: The good, the bad, and the ugly of learning over networks (**distinguished lecture**), Virginia Tech University, Computer Science Department, 2019.
- Essential machine learning hacks that everyone should know about (**invited keynote**), *Women in Data Science*, Purdue, West Lafayette, Indiana, 2019.
- The role of machine learning in AI (**invited panelist**), Plenary Panel, 33<sup>rd</sup> AAAI Conference on Artificial Intelligence, 2019.
- Statistical methods for prediction and anomaly detection in dynamic networks, *Microsoft Research*, Redmond, WA, 2018.
- Deep Learning for Relational Networks (**invited keynote**), 5<sup>th</sup> *International Workshop on High Performance Big Graph Data Management, Analysis, and Mining*, IEEE BigData 2018.
- Deep Learning for Relational Networks (**invited speaker**), Purdue Workshop on Deep Learning, West Lafayette, Indiana, 2018.
- Machine learning for network data (**invited panelist**), 4<sup>th</sup> *Annual Financial Stability Conference*, Department of the Treasury, Washington DC, 2018.
- Using machine learning to exploit complex data in biomedical domains (**invited keynote**), Keck Annual Conference, Houston, TX, 2018.
- Societal Impact of Data Science and Artificial Intelligence (**invited panelist**), Plenary Panel, ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, 2018.

Networks, Crowds & Confusion (**invited panelist**), Purdue Engineering Distinguished Lecture Panel Discussion w/Eva Tardos, 2018.

Mining social network interactions to understand and predict user behavior (**invited speaker**), 6<sup>th</sup> *Purdue Symposium on Psychological Sciences*–Big Data for Psychological Sciences, West Lafayette, Indiana, 2018.

Network Machine Learning (**invited speaker**), *Women in Data Science*, Purdue, West Lafayette, Indiana, 2018.

Learning from single networks—the impact of network structure on relational learning (**invited keynote**), 27<sup>th</sup> *International Conference on Inductive Logic Programming*, France, 2017.

Learning in networks: How to exploit relationships to improve predictions (**invited keynote**), 3<sup>rd</sup> *ACM International Conference on the Theory of Information Retrieval*, Netherlands, 2017.

The impact of network structure on relational machine learning (**invited keynote**), 6<sup>th</sup> *International Conference on Complex Networks and Their Applications*, France, 2017.

Learning in networks: How to exploit relationships to improve predictions, *Ecole Polytechnique Federale de Lausanne (EPFL)*, Computer Science Department, 2017.

Exploiting User Relationships to Accurately Predict Preferences in Large Scale Networks (**invited speaker**), *First Data Institute Conference*, San Francisco, CA, 2017.

Semi-supervised learning for node classification in networks (**invited speaker**), *Workshop on Machine Learning in Network Science*, NetSci, Indianapolis, IN, 2017.

Lifted and Constrained Sampling of Attributed Graphs with Generative Network Models (**invited speaker**), *Workshop on Statistical Inference of Network Models*, NetSci, Indianapolis, IN, 2017.

Learning in networks: How to exploit relationships to improve predictions (**invited speaker**), *Workshop on Diversity in AI*, AAAI, San Francisco, CA, 2017.

AI-Easy vs AI-Hard (**invited speaker**), Dawn or Doom Symposium, Purdue University, West Lafayette, IN, 2016.

Is Deep Learning the New 42? (**invited panelist**), Plenary Panel, ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, 2016.

Statistical methods for modeling network distributions (**invited keynote talk**), *Workshop on Mining and Learning from Graphs*, KDD, San Francisco, CA, 2016.

Statistical methods for modeling network distributions, *MIT Lincoln Laboratory*, Boston, MA, 2016.

AI-Easy vs AI-Hard: Machine learning and its impact on the development of AI systems (**invited speaker**), Purdue Student Pugwash Midwest Regional Conference, West Lafayette, IN, 2016.

Network machine learning: How to exploit relationships to improve node-level predictions (**invited speaker**), Institute of Science and Technology Austria, Young Scientist Symposium, Klosterneuburg, Austria, 2016.

Statistical methods for modeling network distributions (**invited speaker**), Mathematical Biosciences Institute Workshop on Generalized Network Structures and Dynamics, Columbus, Ohio, 2016.

Learning in networks: How to exploit relationships to improve predictions (**distinguished lecture**), Max Planck Institute for Informatics, Saarbrücken, Germany, 2016.

Exploiting User Relationships to Accurately Predict Preferences in Large Scale Networks (**invited speaker**), Netflix Workshop on Personalization, Recommendations, and Search, Los Gatos, CA, 2016.



Learning How to Transfer Collective Classification Models Across Networks (**invited speaker**), Santa Fe Institute Workshop: Inference on Networks, Santa Fe, New Mexico, 2015.

Collective Classification in Large-Scale Networks (**invited tutorial**), *Data Science and Advanced Analytics* (DSAA-15), Paris, France, 2015.

Sampling Attributed Networks From Generative Graph Models (**invited speaker**) Workshop on Information in Networks (WIN), New York, NY, 2015.

Network machine learning: How to use friends to your advantage algorithmically, 2<sup>nd</sup> European Network Intelligence Conference (**invited keynote**), Karlskrona, Sweden, 2016.

Machine learning methods for accurate estimation and prediction in partially-labeled complex networks, Skytree, San Jose, CA, 2015.

Network Sampling: Methods and Applications (**tutorial**), with Mohammad Al Hasan and Nesreen Ahmed, *SIAM International Conference on Data Mining* (SDM-15), Vancouver, Canada, 2015.

An introduction to big data: Opportunities for CS/AG collaborations, Purdue Big Data in Agriculture Seminar Series, West Lafayette, IN, 2015.

Predicting user behavior in networks: The impact of structure on machine learning methods, Indiana University, Bloomington, IN, 2014.

Predicting user behavior in networks: The impact of structure on machine learning methods. Data, Society, and Inference Seminar, *Stanford University*, Palo Alto, CA, 2014.

Predictive modeling for online social networks: Machine learning methods for networks (**invited speaker**), Westwood Colloquium, Purdue University, , West Lafayette, IN, 2014.

Are we too smart for our own good? How large-scale machine learning systems can vastly exceed human level decision-making abilities (**invited speaker**), Dawn or Doom: The New Technology Explosion, Purdue University, West Lafayette, IN, 2014.

How to Exploit Network Properties to Improve Learning in Relational Domains, IBM Thomas J Watson Research Center, Yorktown Heights, NY, 2014.

Prediction in complex networks: The impact of structure on learning and inference (**invited keynote**), 27th International Conference of the Florida Artificial Intelligence Research Society, Pensacola, FL, 2014.

How to learn from a single network to support classification and hypothesis testing in graphs, iCeNSA Seminar, *Notre Dame*, Notre Dame, IN, 2014.

Strengthening Computer Science, with Sunil Prabhakar, *Purdue President's Council*, Mollenkopf/Keyes Weekend, Naples, FL, 2014.

Network Sampling: Methods and Applications (**tutorial**), with Mohammad Al Hasan and Nesreen Ahmed, *International Conference on Data Mining* (ICDM-13), Dallas, TX, 2013.

Prediction in complex networks: The impact of structure on learning and inference (**invited keynote**), Women in Machine Learning Workshop, NIPS 2013, Lake Tahoe, CA, 2013.

Supporting hypothesis testing over graphs (**invited speaker**), Workshop on the Frontiers of network analysis: Methods, models, and applications, NIPS 2013, Lake Tahoe, CA, 2013.

Prediction in complex networks: The impact of structure on learning and inference, RAIN Seminar, *Stanford University*, Palo Alto, CA, 2013.

The impact of network structure on relational learning and inference,  
 Simons Institute, *University California Berkeley*, Berkeley, CA, 2013.

Purdue Moves: Growth in Computer Science, with Sunil Prabhakar,  
*President's Forum*, Purdue University, November 2013.

How to learn from a single network to support classification and hypothesis testing in graphs,  
*SRI International*, Menlo Park, CA, 2013.

How to learn from a single network to support classification and hypothesis testing in graphs,  
*Facebook*, Menlo Park, CA, 2013.

Prediction in complex networks: The impact of structure on learning and inference,  
*PARC*, Palo Alto, CA, 2013.

Prediction in complex networks: The impact of structure on learning and inference,  
 Neyman Seminar, *University of California Berkeley*, Berkeley, CA, 2013.

Machine learning methods for diagnosis, maintenance, and repair of data,  
 AMP Lab, *University California Berkeley*, Berkeley, CA, 2013.

Prediction in complex networks: The impact of structure on learning and inference,  
*Google*, Mountain View, CA, 2013.

How to learn from a single network to support classification and hypothesis testing in graphs,  
*Sandia National Laboratory*, Livermore, CA, 2013.

Network Sampling: Methods and Applications (**tutorial**), with Mohammad Al Hasan and Nesreen Ahmed<sup>†</sup>,  
*International Conference on Knowledge Discovery and Data Mining (KDD-13)*, Chicago, IL, 2013.

Supporting Statistical Hypothesis Testing over Graphs, (**invited session**)  
 Statistical Inference for Networks, *2013 Joint Statistical Meetings (JSM)*, Montreal, QC, Canada, 2013.

Supporting Statistical Hypothesis Testing over Graphs, (**invited speaker**)  
*International Conference on Network Science (NetSci)*, Copenhagen, Denmark, 2013.

Mining Social Network Activity to Understand and Predict User Behavior,  
*Smith College*, Northampton, MA, 2013.

Prediction in complex networks: The impact of structure on learning and inference, (**invited speaker**)  
*Graph Exploitation Symposium*, MIT Lincoln Laboratory, MA, 2013.

Prediction in complex networks: The impact of structure on learning and inference,  
 Statistics Department Colloquia, *Carnegie Mellon University*, Pittsburgh, PA, 2013.

Active Exploration in Networks,  
*CSoI Big Data Workshop*, Honolulu, HI, 2013.

Prediction in complex networks: The impact of structure on learning and inference,  
 Computer Science Colloquia, *University of Alberta*, Edmonton, AB, Canada, 2013.

Prediction in complex networks: The impact of structure on learning and inference,  
*Microsoft Research*, Redmond, WA, 2012.

How to learn from a single network: Statistical relational learning for social network domains,  
 Computer Science Colloquia, *University of Maryland College Park*, College Park, MD, 2012.

Measurement and methodology for mining mobile, cloud-based, social systems, (**invited speaker**)  
 NSF Workshop on Social Networks and Mobility in the Cloud, Arlington, VA, 2012.

Prediction in complex networks: The impact of structure on learning and inference,  
 Computer Science Colloquia, *Ohio State University*, Columbus, OH, 2012.

- How to learn from a single network: Statistical relational learning for social network domains,  
Statistics Colloquia, *Columbia University*, New York, NY, 2012.
- Prediction in complex networks: The impact of structure on learning and inference, (**invited speaker**)  
Workshop on Information in Networks (WIN), New York, NY, 2012.
- Mining Social Network Activity to Understand and Predict User Behavior (**invited speaker**),  
*Data Sciences Summer Institute*, University of Illinois Urbana-Champaign, Urbana, IL, 2012.
- How to learn from a single network: Relational learning for social networks, (**invited speaker**)  
Workshop on Machine Learning: Theory and Computation, *Institute for Mathematics and Its Applications* (IMA), Minnesota, MN, 2012.
- Supporting Statistical Hypothesis Testing Over Graphs, (**invited speaker**)  
Workshop on Network Links: Connecting Social, Communication & Biological Network Analysis, *Institute for Mathematics and Its Applications* (IMA), Minnesota, MN, 2012.
- How to learn from a single network: Statistical relational learning for social network domains,  
Computer Science Colloquia, *Duke University*, Durham, NC, 2011.
- How to learn from one sample? Statistical relational learning for single network domains,  
AI Seminar, *University of Texas Austin*, Austin, TX, 2011.
- Modeling online social networks to understand and predict user behavior,  
IROM Seminar, *McCombs School of Business*, Austin, TX, 2011.
- Understanding the Effects of Collective Classification on Learning and Inference (**invited keynote talk**),  
*Workshop on Collective Learning and Inference for Structured Data*, ECML, Athens, Greece, 2011.
- Mining Social Network Activity to Understand and Predict User Behavior (**invited keynote talk**),  
*Workshop on Enriching Information Retrieval*, SIGIR, Beijing, China, 2011.
- Modeling Complex Social Networks: Challenges and Opportunities for Statistical Learning and Inference,  
(**invited speaker**), *Machine Learning Summer School*, Purdue, West Lafayette, IN, 2011.
- Modeling Complex Social Networks: Challenges and Opportunities for Statistical Learning and Inference,  
(**invited speaker**), *Science of Information Summer School*, Purdue, West Lafayette, IN, 2011.
- Statistical Relational Learning in Single Network Domains,  
PRiML Seminar, *University of Pennsylvania*, Philadelphia, PA, 2011.
- Modeling and Mining Social Networks,  
Fantastic Lectures in Computer Science Series, *Bryn Mawr College*, Bryn Mawr, PA, 2011.
- Hypothesis testing methods for social network mining,  
AI Seminar, *Information Sciences Institute*, Marina Del Ray, CA, 2010.
- Hypothesis testing methods for social network mining,  
Neyman Seminar, *University of California Berkeley*, Berkeley, CA, 2010.
- Hypothesis testing methods for social network mining,  
IS Research Seminar, *New York University Stern School of Business*, New York, NY, 2010.
- Capturing the Natural Variability of Real Networks with Kronecker Product Graph Models,  
*Sandia National Laboratory*, Livermore, CA, 2010.
- Evaluation Strategies for Network Classification Models (**invited keynote talk**),  
*Workshop on Mining and Learning from Graphs*, KDD, Washington, DC, 2010.
- Evaluation Strategies for Network Classification Models,  
*University of Maryland College Park*, College Park, MD, 2010.

- Modeling Relationship Strength in Online Social Networks,  
*IUPUI*, Indianapolis, IN, 2010.
- Modeling Relationship Strength in Online Social Networks,  
DAIS Seminar, *University of Illinois Urbana-Champaign*, Urbana, IL, 2010.
- Predictive Modeling with Social Networks (**invited tutorial**), with Foster Provost,  
*International Conference on Weblogs and Social Media (ICWSM-09)*, San Jose, CA, 2009.
- Social Network Mining (**tutorial**), with Foster Provost,  
*International Conference on Knowledge Discovery and Data Mining (KDD-08)*, Henderson, NV, 2008.
- Social Network Mining (**invited tutorial**), with Foster Provost,  
*National Conference on Artificial Intelligence (AAAI-08)*, Chicago, IL, 2008.
- Exploiting Temporal Variations in Relational Domains,  
*Lawrence Livermore National Laboratory*, Livermore, CA, 2008.
- Exploiting Temporal Variations in Relational Domains,  
*University of Maryland College Park*, College Park, MD, 2008.
- Statistical Models for Learning and Inference in Complex Relational Domains.  
*National Security Agency*, Fort Meade, MD, 2007.
- Leveraging Autocorrelation with Latent Group Models.  
Auton Lab, School of Computer Science, *Carnegie Mellon University*, Pittsburgh, PA, 2005.
- Leveraging Autocorrelation with Latent Group Models.  
Dagstuhl Seminar on Probabilistic, Logical and Relational Learning: Towards a Synthesis. *Schloss Dagstuhl*, Wadern, Germany, 2005.
- Knowledge Discovery with Relational Dependency Networks.  
Weekly Computer Science Colloquium, *Williams College*, Williamstown, MA, 2004.
- Dependency Networks for Relational Data.  
*The Boeing Company*, Phantom Works, Mathematics & Computing Technology Unit, Seattle, WA, 2004.
- Probability Estimation Trees for Relational Data.  
Computer Science Department, *Williams College*, Williamstown, MA, 2004.
- Collective Classification with Relational Dependency Networks (*poster*).  
DARPA IPTO Cognitive Systems Conference, Washington, DC, 2003.
- Knowledge Discovery in Networks.  
Talent Advancement Program Seminar, Computer Science Department, *University of Massachusetts*, Amherst, MA, 2003.
- Clustering Relational Data (*poster*).  
Grace Hopper Celebration of Women in Computing, Vancouver, BC, 2002.
- Data Mining in Networks.  
International Sunbelt Social Network Conference XXII, New Orleans, LA, 2002.

## Service

### *Outreach*

Machine learning and its impact on the development of AI systems, Cyber Technology Conference & Policy Summit, Indianapolis, IN, 2019.

Session organizer: Fairness and Bias in Machine Learning, SciFoo'18, Google, 2018.

A few things about machine learning, Anvil Software Stir lecture, West Lafayette, Indiana, 2017.

Machine Learning Over Social Networks, Iridescent Learning, Ask the Experts Lecture and Interview, 2017.

<https://www.youtube.com/watch?v=U0vtr6L0rzU&t=78s>

<https://www.youtube.com/watch?v=iKUXaagu0-I&t=45s>

AI-easy vs. AI-hard?, Purdue Dawn or Doom Symposium, 2016.

<https://www.c-span.org/video/?415869-2/jennifer-neville-discusses-artificial-intelligence>

Machine learning and its impact on the development of AI systems, Purdue Student Pugwash Midwest Regional Conference, 2016.

Panelist: Techpoint Indianapolis, Dawn or Doom Panel, 2015.

An introduction to big data: Opportunities for CS/AG collaborations, Purdue Big Data in Agriculture Seminar, 2015.

Are we too smart for our own good?, Purdue Dawn or Doom Symposium, 2014.

<https://www.youtube.com/watch?v=GpltbAljmbg>

Strengthening Computer Science, with Sunil Prabhakar, *Purdue President's Council*, Naples, FL, 2014.

### *University*

150<sup>th</sup> Ideas Festival Working Group, 2018-2019

Faculty Search Committee OIE Facilitator, 2014-2018

Diversity and Inclusion Symposium, Panel Moderator, 2014

Conference for Pre-Tenure Women, Mentor, 2014

Presentations:

CoE EFC on Security and Privacy, 2019

CoE Engineering Advisory Panel, 2017

Center for The Environment Big Data Panel, 2017

Westwood Colloquium, 2014

President's Council Weekend, 2014

President's Forum, 2013

### *College*

Integrative Data Science Initiative Steering Committee, 2018-2019

Integrated Imaging Cluster Hiring Committee, 2012-2014

CS Realignment Task Force, 2009-2010

Presentations:

College of Science Woman for Purdue event, 2010, 2014  
College of Science Alumni Board Meeting, 2010, 2104  
Purdue President and Provost, Tour of the College of Science, 2010

Other:

Purdue Day of Giving Interview, 2014  
Purdue Insights Article, 2011, 2014

*Department*

Computer Science

AI Search Committee Chair, 2019-2020

Head Search Committee, 2018-2019

Hiring Committee, 2007-2008, 2012-2016, 2017-2019

New Data Science Degree Committee, 2016-present

Hiring Committee, 2007-2008, 2012-2016

Big Data Cluster Hiring Committee, 2012-2016

Advisory Committee, 2014-present

Colloquium Chair, 2008-2010

Course Development

Development of introductory data science course for Data Science major (2017-2018)  
Development of undergrad machine learning course for Machine Intelligence track (2011)  
Development of joint MS program for CS/Stat (2009-10)

Graduate Curriculum Committee, 2011-2013

Strategic Planning Committee, 2009-2010

Undergraduate Curriculum Committee, 2006-2007, 2010-2011

Qual 2 Committees<sup>3</sup> (31 students):

2006: Rimma Nehme  
2007: Jing Li, Yinian Qi  
2008: Suleyman Cetintas  
2009: Mohamed Yakout  
2010: Nan Ding, Yi Fang, Yao Zhu, Chandima Hewanadungodage, Youhan Fang  
2011: Pei He, Bin Shen, Fang Yu Rao, Feng Yan, Ahmet Bugdayci, Jiaqi Ge, Xi Tan, Syed Abaas  
2012: Ruby Tahboub, Romilla Pradhan, Jihwan Lee  
2014: Ning Zhang, Xuejiao Kang, Hao Peng  
2015: Shandian Zhe, Shawn Merrill, Zhiwei Zhang  
2016: Cosmo Xiang, Ashish Ghoshal, Kristen Johnson, Jin Di

Prelim Committees (that do not include PhD committee):

Feng Yan  
Jiaqi Qi  
Qifan Wang  
Lei Cen

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<sup>3</sup>Note that Purdue CS eliminated Qual 2 exams in 2016.

Bahar Saberidokht  
Amr Ebiad  
Tanay Saha

PhD Committees (44 students):

*Computer Science:*

Ferit Akova, Vasil Denchev, Nicole Eikmeier, Juan Esquivel, Jiaqi Ge, Ashish Ghoshal, Chandima Hewanadungodage, Andrew Hoblitzell, Kristen Johnson, Ishita Khan, Chang Li, I-Ta Lee, Jihwan Lee, Koray Manchuan, Mohamed Mohamed, Abbas Naqvi, Rimma Nehme, Mehmet Nergiz, Gen Nishida, Keewan Park Hao Peng, Rahul Potharaju, Yinian Qi, Mahmudur Rahman, Sarvjeet Singh, Christine Task, Leonardo Teixeira, Mohamed Yakout, Pinar Yardanag, Dan Zhang, Xiao Zhang

*Aerospace Engineering:* Alden Black

*Biology:* Biaobin Jiang, Muye Liu

*Electrical and Computer Engineering:*

Suhas Javagal, Subrata Mitra, Nawanol Theera-Ampornpant, Tsung-Tai Yeh

*Industrial Engineering:* Arora Viplove

*Sociology:* Min Yim

*Statistics:* Emery Goosens, Ryan Murphy, Hyokun Yun

*External:* Shuangfei Fan, Virginia Tech

Presentations:

CS197: *Topics in Computer Sciences (Honors)*, 2008, 2010, 2012

CS291: *Sophomore Development Seminar*, 2008

CS591C: *Research Seminar for Graduate Students*, 2006, 2007, 2008, 2010

Women in CS Career Day: *Purdue CS outreach event for high school girls*, 2008, 2009, 2010

Undergrad Research Bonanza, 2009

Undergrad Track Fair, 2014

Corporate Partners Program, 2009

Grad Orientation, 2009, 2012

Corporate presentations 2014-2016: John Deere, General Electric, Northrup Grumman, Lockheed Martin, Infosys, AFRL, Tsukuba

Outreach:

STAR lunches with incoming freshman, 2009

CSWN events with women faculty, 2008, 2010

Statistics

Hiring Committee, 2007-2008, 2017-2018

Presentations:

MA108: *Mathematics as a Profession and a Discipline*, 2006

STAT VIGRE Seminar: *Exploring Statistical Sciences Research*, 2006, 2008, 2010

**Professional**

Program committee chair

National Conference on Artificial Intelligence (AAAI): 2023

SIAM International Conference on Data Mining (SDM): 2019

ACM International Conference on Web Search and Data Mining (WSDM): 2016

Associate chair

Neural Information Processing Systems (NIPS), 2017-2019

International Joint Conference on Artificial Intelligence (IJCAI), Machine Learning Track: 2015, 2018  
National Conference on Artificial Intelligence (AAAI), Machine Learning Track: 2013, 2019

#### Area chair

ACM International Conference on Knowledge Discovery and Data Mining (KDD): 2016-2019  
ACM International Conference on on Web Search and Data Mining (WSDM): 2013, 2014, 2017  
International Conference on Machine Learning (ICML): 2011, 2015, 2018-2019  
International Conference on Uncertainty in Artificial Intelligence (UAI): 2019  
International Joint Conference on Artificial Intelligence (IJCAI), 2016, 2017  
National Conference on Artificial Intelligence (AAAI): 2014  
Neural Information Processing Systems (NIPS): 2014  
IEEE International Conference on Data Mining (ICDM): 2014, 2015  
SIAM conference on Data Mining (SDM): 2013, 2014

#### Program committees

ACM International Conference on Knowledge Discovery and Data Mining (KDD): 2008-2012, 2014  
European Conference on Machine Learning (ECML/PKDD): 2007, 2008, 2012  
IEEE International Conference on Data Mining (ICDM): 2007, 2009-2012  
International Conference on Artificial Intelligence and Statistics (AISTATS): 2009, 2011  
International Conference on Inductive Logic Programming (ILP): 2007  
International Conference on Machine Learning (ICML): 2006, 2008-2012  
International Joint Conference on Artificial Intelligence (IJCAI): 2009  
International World Wide Web Conference (WWW): 2011  
National Conference on Artificial Intelligence (AAAI): 2006-2008, 2012  
Neural Information Processing Systems (NIPS): 2012  
SIAM Conference on Data Mining (SDM), 2006

#### Journal editorial boards

Journal of Artificial Intelligence Research, 2010-2013  
Machine Learning Journal, 2011-present  
Data Mining and Knowledge Discovery, 2014-present

#### Journal reviewing

Journal of Machine Learning Research  
Machine Learning Journal  
Transactions on Knowledge Discovery  
Data Mining and Knowledge Discovery Journal

#### Proposal reviewing

NSF Expeditions Blue Ribbon Panel, 2015, 2018  
NSF Information & Intelligent Systems Panel, 2005, 2008, 2010-2015, 2017  
NSF Information & Intelligent Systems CAREER Panel, 2012, 2016  
NASA Earth Science Technology Office Proposals, 2005

#### Conference/workshop organization

Workshop Co-chair:  
Relational Representation Learning Workshop, NeurIPS: 2018



NII Shonan Meeting Seminar 113, Meta-Programming for Statistical Machine Learning: 2018  
13<sup>th</sup> Workshop on Mining and Learning with Graphs (MLG): 2017  
Unifying Theory and Experiment for Large-Scale Networks, Simons Institute, UC Berkeley: 2014  
4<sup>th</sup> ACM SIGSPATIAL International Workshop on Location-Based Social Networks (LBSN): 2012  
Statistical Issues with Modeling of Networks, 8th Int'l Purdue Symposium on Statistics: 2012  
9<sup>th</sup> Workshop on Mining and Learning with Graphs (MLG): 2011  
Machine Learning Summer School, Purdue: 2011

Award Committee:

*Doctoral Dissertation Award*: Association for Computing Machinery (ACM): 2019-2021  
*Best Papers*, ACM Int'l Conf on Knowledge Discovery and Data Mining (KDD): 2015, 2019  
*Test-of-Time Papers*, ACM Int'l Conf on Knowledge Discovery and Data Mining (KDD): 2018  
*Best Papers* (Chair), ACM Int'l Conf on Knowledge Discovery and Data Mining (KDD): 2017  
*Dissertation Award*: ACM Int'l Conf on Knowledge Discovery and Data Mining (KDD): 2011-2013

Diversity events:

Women in KDD Panel, ACM International Conference on Knowledge Discovery and Data Mining (KDD): 2017-2018  
Working Group on Gender Diversity, ACM International Conference on Web Search and Data Mining (WSDM): 2013  
Women in AI, International Conference on Artificial Intelligence (AAAI): 2016-2017, 2019

Tutorial Chair:

ACM International Conference on Knowledge Discovery and Data Mining (KDD): 2012, 2015  
International Conference on Machine Learning (ICML): 2009, 2015

Panelist:

White House Office of Science and Technology Policy, Workshop on AI and Social Good, 2016

Treasurer:

International Machine Learning Society (IMLS): 2009-2013

Professional societies

Association for the Advancement of Artificial Intelligence (AAAI)  
Association for Computing Machinery (ACM)  
ACM Special Interest Group on Knowledge Discovery and Data Mining (SIGKDD)  
Institute of Electrical and Electronics Engineers (IEEE)  
International Machine Learning Society (IMLS)  
Sigma Xi

## **Sponsored Research**<sup>4</sup>

*Learning from Theorem Proving Search via Graph Representations*

NSF/CISE/FMitF, co-Primary Investigator  
\$250,000 (33% of total), 08/01/19 - 07/31/22

*I<sup>2</sup>DS: Intelligent Interaction Defense System*

DARPA/I2O, Primary Investigator  
\$744,302 (33% of total), 09/01/18 - 08/31/22

*Robust Machine Learning*

Integrative Data Science Institute Purdue, Primary Investigator  
\$264,989 (100% of total), 06/01/18 - 05/31/20

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<sup>4</sup>Reported amounts reflect funds 100% under my responsibility.

*Transfer Learning Within & Across Networks for Collective Classification*

NSF/CISE/IIS, Primary Investigator  
\$495,308 (100% of total), 07/01/16 - 06/30/19

*Models, Algorithms, and Software for Spatial-Relational Networks*

NSF/CISE/IIS, co-Primary Investigator  
\$300,000 (33% of total), 09/01/15 - 08/31/19

*Frontiers of Science of Information*

NSF/Science & Technology Center, Senior Personnel  
*approx.* \$200,000 (0.8% of total), 08/01/15 - 07/31/20

*A Heterogeneous Inference Framework for 3D Modeling and Rendering of Sites*

NSF/CISE/CVG, co-Primary Investigator  
\$297,782 (50% of total), 07/01/13 - 06/30/16

*Parametric Statistical Models to Support Statistical Hypothesis Testing over Graphs*

NSF/CISE/IIS, co-Primary Investigator  
\$245,920 (50% of total), 09/01/12 - 08/31/15

*Career: Machine Learning Methods and Statistical Analysis Tools for Single Network Domains*

NSF/CISE/IIS, Primary Investigator  
Amount: \$496,638 (100% of total), 01/01/12 - 12/31/16

*Toward Intrusion Tolerant Clouds*

DARPA/I2O, co-Primary Investigator (subcontract from Johns Hopkins)  
\$367,028 (9% of total), 11/01/11 - 10/31/15

*Modeling Tools to Support Advanced Analysis of Multi-Source Network Data*

IARPA/KDD, co-Primary Investigator (subcontract from SAIC)  
\$563,950 (5% of total), 11/01/10 - 01/31/15

*Sampling and Modeling Dynamic Streaming Networks*

CISCO, co-Primary Investigator  
\$102,076 (50% of total), 02/07/11 - 02/07/99

*Towards Better Modeling of Communication Activity Dynamics in Large-Scale Online Social Networks*

NSF/CISE/NETSE, Primary Investigator  
\$248,226 (50% of total), 09/01/10 - 08/31/13

*Emerging Frontiers of Science of Information*

NSF/Science & Technology Center, Senior Personnel  
*approx.* \$200,000 (0.8% of total), 08/01/10 - 07/31/15

*Algorithms for Sampling Similar Graphs Using Subgraph Signatures*

NSF/CISE/IIS, co-Primary Investigator  
\$164,846 (33% of total), 09/01/09 - 08/31/11

*Machine Learning Techniques to Model the Impact of Relational Communication on Distributed Team Effectiveness*

NSF/SES/IOS, Primary Investigator  
\$205,311 (50% of total), 09/01/08 - 08/31/11

*MAASCOM: Modeling, Analysis, and Algorithms for Stochastic Control of Multi-Scale Networks*

ARO/MURI, co-Primary Investigator (subcontract from Ohio State)  
\$250,000 (5% of total), 5/29/08–10/28/08

*Fusion and Analysis of Multi-Source Relational Data*

DARPA/ISO, Primary Investigator  
\$499,877 (100% of total), 06/23/08–06/22/10

*Learning Compositional Simulation Models*

IARPA/Proactive Intelligence, co-Primary Investigator (subcontract from UMass)  
\$122,217 (33% of total), 04/01/07–02/09/09

*Mining Transaction Streams to Infer Semantic Relations*

Microsoft, Primary Investigator  
\$50,000 (100% of total), 06/01/07–06/01/99

*Statistical Models and Algorithms to Improve Decision-Making in Relational Domains*

DARPA/ISO, Primary Investigator  
\$100,000 (100% of total), 04/01/07–12/31/07

## **Teaching**

Spring 2020–Fall 2023: Research leave

CS471: Introduction to Artificial Intelligence, Fall 2019  
Requirement for Machine Intelligence Track  
115 students

CS 59000DEI: Data Engineering I and CS 59000FDM: Foundations of Decision Making, Spring 2019  
Developed material and delivered lectures for online modules

CS242: Introduction to Data Science, Fall 2018  
Requirement for new Data Science Major  
25 students

CS242: Introduction to Data Science, Spring 2018  
Requirement for new Data Science Major  
10 students

CS573: Data Mining, Spring 2017  
Graduate table course, EPE distance education course  
107 students

CS497: Honors Research, Spring 2017  
Undergraduate honors project  
Student: Benjamin Staiger  
Project: Classifying Relationships Based on Time Series Data

CS473: Web Information Search and Management, Fall 2016  
Requirement for Machine Intelligence Track  
75 students

ST598T: Statistical Network Analysis (cross-listed as CS590), Spring 2016  
Graduate seminar  
12 students

CS39000-DM0: Data Mining & Machine Learning, Fall 2015  
Requirement for Machine Intelligence Track  
55 students

CS573: Data Mining, Spring 2015  
Graduate table course, EPE distance education course  
101 students

CS39000-DM0: Data Mining & Machine Learning, Fall 2014  
Requirement for Machine Intelligence Track  
30 students

CS39000-DM0: Data Mining & Machine Learning, Spring 2014  
Requirement for Machine Intelligence Track  
40 students

CS497: Honors Research, Spring 2014  
Undergraduate honors project  
Student: Seong Gyeong Lee  
Project: Sentiment Analysis of Facebook Posts

Fall 2013: Sabbatical

CS39000-DM0: Data Mining & Machine Learning, Spring 2013  
Requirement for Machine Intelligence Track  
40 students

CS497: Honors Research, Spring 2013  
Undergraduate honors project  
Student: Dan Coroian  
Project: Playing Solitaire with Reinforcement Learning

CS573: Data Mining, Fall 2012  
Graduate table course  
66 students

CS497: Honors Research, Fall 2012  
Undergraduate honors project  
Student: Brian Donovan  
Project: Textual Analysis of Facebook Messages to Predict Social Characteristics

CS39000-DM0: Data Mining & Machine Learning, Spring 2012  
Requirement for Machine Intelligence Track  
32 students: 27 CS, 5 Other

CS251: Data Structures and Algorithms, Spring 2011  
Undergraduate requirement  
61 CS students

CS573: Data Mining, Fall 2010  
Graduate table course  
32 students, 32 CS

ST598T: Statistical Network Analysis (cross-listed as CS590), Spring 2010  
Graduate seminar  
19 students: 9 CS, 4 STAT, 5 Eng, 1 IPSC

Fall 2009: Buyout

CS497: Honors Research, Spring 2009

Undergraduate honors project

Student: Ankit Kuwadeker

Project: Active Learning for Relational Classification

CS573: Data Mining (cross-listed as ST598M), Spring 2009

Graduate table course

34 students, 21 CS, 3 STAT, 9 Eng, 1 EAS

CS471: Artificial Intelligence, Fall 2008

Undergraduate elective

30 CS students

ST598T: Statistical Network Analysis, Spring 2008

Graduate seminar

6 students: 4 CS, 2 STAT

CS590D: Data Mining (now CS573, cross-listed as ST598M), Fall 2007

Graduate table course

23 students, 16 CS, 2 STAT, 5 Eng

CS590N: Statistical Relational Learning, Spring 2007

Graduate seminar

10 students: 6 CS, 1 STAT, 3 Eng

## **Students**

### **Current PhD Students**

1. Ying-Chun (Jean) Lin  
Advising since: Fall 2017  
Current status: Post-prelim  
Funding: Research assistant
2. Jiayi Liu  
Advising since: Fall 2020  
Current status: Passed Qual 2  
Funding: Teaching assistant

### **Graduated Students**

1. Giselle Zeno  
Degree: PhD, July 2023  
Thesis: Dynamic Network Modeling from Temporal Motifs and Attributed Node Activity  
Current location: MIT Lincoln Labs
2. Susheel Suresh  
Degree: PhD, Dec 2022  
Thesis: Models and Representation Learning Mechanisms for Graph Data  
Current location: Microsoft
3. Mengyue Hang  
Degree: PhD, Dec 2021

Thesis: Graph Representation learning for unsupervised and semi-supervised learning tasks  
Current location: Facebook

4. Yi-yu Lai  
Degree: PhD, May 2021  
Thesis: Relational Representation Learning incorporating Textual Communication for Social Networks  
Current location: Intel
5. Mahak Goindani  
Degree: PhD, Dec 2020  
Thesis: Social Reinforcement Learning  
Current location: Apple
6. Hogun Park  
Degree: PhD, July 2020  
Thesis: Neural Representation Learning for Semi-Supervised Node Classification and Explainability  
Current location: Assistant Professor, College of Computing, Sungkyunkwan University, South Korea
7. Changping Meng  
Degree: PhD, April 2020  
Thesis: Scalable Representation Learning with Invariances  
Current location: ML Software Engineer, Google
8. Guilherme Gomes  
Degree: PhD, Dec 2019  
Thesis: Hypothesis Testing and Community Detection on Networks with Missingness & Block Structure  
Current location: Research Scientist, Bayer Pharma
9. Jiasen Yang  
Degree: PhD, June 2019  
Thesis: Statistical Learning and Model Criticism for Networks and Point Processes  
Current location: Quantitative Associate, Two Sigma
10. Xi Tan  
Degree: PhD, May 2018  
Thesis: Bayesian Nonparametrics to Model Content, User, and Latent Structure in Hawkes Processes  
Current location: Quantitative Associate, Goldman Sachs
11. Elizabeth Tigner  
Degree: BSc, May 2018  
Research project: Analysis of network discussion trends in Twitter using hashtag clusters  
Current location: Engineer, SEAKR Engineering
12. Caleb Belth  
Degree: BSc, May 2018  
Research project: Embedding influence graphs  
Current location: PhD Program, CS, University of Michigan
13. Shandian Zhe  
Degree: PhD, Nov 2017  
Thesis: Scalable Bayesian Nonparametrics and Sparse Learning for Hidden Relationships Discovery  
Current location: Assistant Professor, University of Utah, Computer Science Dept
14. Pablo Robles Granda  
Degree: PhD, Oct 2017  
Thesis: Generating Attribute Networks: Modeling, Learning, and Sampling  
Current location: Research Assistant Professor, Notre Dame University, Computer Science Dept

15. Gouthami Kamalnath  
Degree: BSc, May 2017  
Current location: Software Engineer, Microsoft
16. Shubhika Barjatya  
Degree: BSc (Statistics), May 2017  
Current location: Data Scientist, Wayfair
17. Benjamin Staiger  
Degree: BSc, May 2017  
Research project: Classifying Relationships Based on Time Series Data  
Current location: Software Engineer, Citadel
18. Sait Celebi  
Degree: MSc, May 2017  
Current location: Software Engineer, Google
19. Iman Alodah  
Degree: MSc, Dec 2016  
Current location: TBD
20. John Moore  
Degree: MSc, Dec 2016  
Thesis: Deep Collective Inference  
Current location: Software Engineer, Microsoft
21. Timothy La Fond  
Degree: PhD, July 2016  
Thesis: Controlling for Confounding Network Properties in Hypothesis Testing and Anomaly Detection  
Current location: Postdoc, Lawrence Livermore National Labs
22. Ransen Niu  
Degree: BSc, May 2016  
Research project: Transfer learning in networks  
Current location: PhD Program, CS, Cornell University
23. Nesreen Ahmed  
Degree: PhD, July 2015  
Thesis: Scaling Up Network Analysis and Mining: Statistical Sampling, Estimation, and Pattern Discovery  
Current location: Senior Staff Research Scientist, Intel Research
24. Stephen Mussmann  
Degree: BSc, May 2015  
Research project: Assortativity in statistical models of graphs  
Current location: PhD Program, CS, Stanford University
25. Joseph Pfeiffer III  
Degree: PhD, May 2015  
Thesis: Overcoming Uncertainty for Within-Network Relational Machine Learning  
Current location: Applied Researcher, Microsoft
26. Sebastian Moreno  
Degree: PhD, August 2014  
Thesis: Network Hypothesis Testing for Relational Data  
Current location: Assistant Professor, Universidad Adolfo Ibanez, Chile

27. Suvidha Kancharla  
Degree: MSc, May 2014  
Current location: Software Engineer, Microsoft
28. Seong Gyeong Lee  
Degree: BSc, May 2014  
Research project: Sentiment Analysis of Facebook Messages
29. Karthik Nagaraj  
Degree: PhD, October 2013  
Thesis: Enabling Richer Insight Into Runtime Execution of Systems  
Current location: Software Engineer, Google
30. Dan Coroian  
Degree: BSc, May 2013  
Research project: An Application of SARSA Learning to Klondike Solitaire  
Current location: PhD Program, Computer Science, Duke University
31. Brian Donovan  
Degree: BSc, May 2013  
Research project: Using Facebook Text to Predict Social Characteristics  
Current location: PhD Program, Civil Engineering, UIUC
32. Christopher Cole  
Degree: BSc, May 2013  
Research project: Analysis of Dynamic Email Graphs  
Current location: Software Engineer, Amazon
33. Daniel Roberts  
Degree: BSc, May 2013  
Research project: The Impact of Interpersonal Dependencies on Distributed Teams  
Current location: Software Engineer, EMC Isilon
34. Rongjing Xiang  
Degree: PhD, August 2012  
Thesis: Statistical Relational Learning for Single Network Domains  
Current location: Software Engineer, Google
35. Hoda Eldardiry  
Degree: PhD, February 2012  
Thesis: Ensemble Classification Techniques for Relational Domains  
Current location: Associate Professor, Virginia Tech University, CS
36. Chris Mayfield (co-advised with Sunil Prabhakar)  
Degree: PhD, Aug 2011  
Thesis: Statistical Inference and Data Cleaning in Relational Database Systems  
Current location: Associate Professor, James Madison University, CS
37. Ankit Kuwadekar  
Degree: BSc, May 2010  
Research project: Active Learning for Relational Domains  
Current location: Software Engineer, Amazon
38. Umang Sharan  
Degree: MSc, May 2008  
Thesis: Temporal-Relational Classifiers for Prediction in Evolving Domains  
Current location: Software Engineer, YouTube (Google)