

PADHRAIC SMYTH

Department of Computer Science, Bren Hall 4216
School of Information and Computer Sciences
University of California, Irvine
CA 92697-3435
telephone: (949) 824 2558
email: smyth@ics.uci.edu

Professional Positions

April 1996–present: Distinguished Professor, Department of Computer Science, University of California, Irvine

- Distinguished Professor: 2023 to present
- Chancellor’s Professor: 2018 to 2023
- Full Professor: July 2003 to 2018
- Associate Professor: July 1998 to June 2003
- Assistant Professor: April 1996 to June 1998

October 1988–March 1996: Member of Technical Staff and Technical Group Leader, Jet Propulsion Laboratory, California Institute of Technology, Pasadena.

Education

PhD, 1988: California Institute of Technology, Department of Electrical Engineering.

MSEE, 1985: California Institute of Technology, Department of Electrical Engineering.

BE, 1984: National University of Ireland, University College Galway. Bachelor of Engineering (Electronic) with First-Class Honors.

Additional Professional Roles and Affiliations

Joint (Courtesy) Faculty Appointments:

- Department of Statistics, UC Irvine, July 2008–present.
- Department of Education, UC Irvine, July 2017–present.

Founding Director, UCI Data Science Initiative, University of California, Irvine, July 2014–June 2018.

Founding Director, Center for Machine Learning and Intelligent Systems, University of California, Irvine, January 2007–June 2014.

Co-Director, HPI Research Center in Machine Learning and Data Science at UCI, 2020–present.

Center Affiliations (Faculty Member)

- Steckler Center for Responsible, Ethical, and Accessible Technology, UC Irvine, 2021–present.
- Connected Learning Lab, UC Irvine, 2024–present.
- Institute for Genomics and Bioinformatics (IGB), UC Irvine, Member 2001–present.
- Institute for Mathematical Behavioral Sciences (IMBS), UC Irvine, 1999–2022.
- Center for Digital Transformation, UC Irvine, 2012–present.
- Program for Mathematical, Computational, and Systems Biology (MCB), UC Irvine, 2007–present.

- Center for Research on Information Technology and Organizations (CRITO), UC Irvine, 2008–2012.

Founding Director and Executive Committee Member of the ACM Special Interest Group on Knowledge Discovery and Data Mining (SIGKDD), 1998.

Visiting Principal Researcher, Jet Propulsion Laboratory, California Institute of Technology, Pasadena, 1996–2001.

Honors and Awards

Best paper awards:

- International Conference on AI and Statistics: outstanding paper with lead student author, 2024.
- Educational Data Mining Conference (EDM): best paper, 2018
- ACM/IEEE Joint Conference on Digital Libraries (JCDL): shortlist for best paper, 2007
- ACM SIGKDD Conference: best paper (1997, 2002), runner-up best paper (1998, 2000)

Fellow, Institute for Electrical and Electronic Engineers (IEEE), elected 2023

Fellow, American Association for the Advancement of Science (AAAS), elected 2022

Fellow, Association for Computing Machinery (ACM), elected 2013

Fellow, Association for the Advancement of Artificial Intelligence (AAAI), elected 2010

ACM SIGKDD Innovation Award, 2009

Qualcomm Faculty Awards, 2019/2020/2021

Google Faculty Research Awards, 2008 and 2014

IBM Faculty Partnership Award, 2001

National Science Foundation CAREER award, 1997

ACM Teaching Award, UC Irvine, 1997

NASA Group Achievement award, Jet Propulsion Laboratory, 1997

Lew Allen Award for Excellence in Research, Jet Propulsion Laboratory, 1993

17 NASA Certificates for Technical Innovation (1991–1996)

Advisory and Consulting Activities

Smith Baluch LLP (2022-present); Cove Fund (2021-present); Candor Technologies (2021-present); Fenwick and West LLP (2019-present); Fox Rothschild LLP (2021); Fish and Richardson (2021); AdvanceOC Advisory Board (2020-present); Wilson, Sonsoni, Goodrich and Rosati (2019-2021); QuinnEmanuel LLP (2019-2020); Morgan Lewis and Bockius LLP (2019); Erise IP (2017-2018; Toshiba (2018-2019), First American (2018-2019); ProLung, Inc (2017-2019); Unified Patents (2016-2019); University of Washington (2016-2019); Klarquist LLP (2015-2016); Frost Data Capital (2014-2015); AST Inc (2013-2015); Samsung (2012-2015); SOCCCD (2012-present); DigitalRisk (2010-2012); CoreLogic (2011-2014); IdentityMetrics (2010-2012); Microsoft (2010-2011); ImageCat (2010); eBay (2009-2011); DataAnalytics LLC (2009-2011); QuinnEmanuel LLP (2011); Latham and Watkins (2008-2009, 2011); Netflix (2006-2009); Topicseek LLC (2005-2008); Yahoo! (2005-2008); Strativa (2005); IET (2004-2005); JWDirect (2001-2004); Credit Sciences (2000-2004); Nokia Research (2000); First Quadrant Financial Services (1998-1999); Smith-Kline Beecham (1998); AT&T (1996-1998).

Postdoctoral Advisees and Current Positions

Ralf Krestel, 2011-2013; Professor, Kiel University/ZBW, Kiel, Germany.
Tracy Holsclaw, 2011-2014; Consultant, San Jose, CA.
Romain Thibaux, 2008-2009; Software Engineer, Waymo, Mountain View, CA.
Alex Ihler, 2005-2006; Professor, Department of Computer Science, UC Irvine.
Michael Duff, 2005-2006; Researcher, Fred Hutchinson Cancer Research Center, Seattle, WA.
Michal Rosen-Zvi, 2003-2004; Director, AI for Healthcare and Lifescience, IBM Research, Israel; Adjunct Professor, Hebrew University.

PhD Students

PhD Advisees and Current Positions

Gavin Kerrigan, PhD 2024; Postdoctoral Fellow, Department of Statistics, Oxford University
Alex Boyd (co-advised with Stephan Mandt), PhD 2024; GE Healthcare, Seattle
Robert Logan IV (co-advised with Sameer Singh), PhD 2022; Dataminr, New York
Disi Ji, PhD 2020; Instagram, Menlo Park, CA
Chris Galbraith, PhD 2020; Google, Philadelphia, PA
Jihyun Park, PhD 2019; Apple, Cupertino, CA
Dimitris Kotzias, PhD 2018; Google, Zurich
Eric Nalisnick, PhD 2018; Assistant Professor, Department of Computer Science, Johns Hopkins University
Moshe Lichman, PhD 2017; Google, Irvine, CA
Nick Navaroli, PhD 2014; Google, Irvine, CA
Jimmy Foulds, PhD 2014: Associate Professor, Department of Information Systems, UMBC
Chris DuBois, PhD 2013: Apple, Seattle
America Chambers, PhD 2013: Associate Professor and Chair, Department of Mathematics and Computer Science, University of Puget Sound
Drew Frank (co-advised with Alex Ihler), PhD 2013: Apple, Seattle
Arthur Asuncion, PhD 2011: Google, Seattle, WA
Jon Hutchins (co-advised with Alex Ihler), PhD 2010: Google, Pittsburgh, PA
Chaitanya Chemudugunta, PhD 2009: Director, Data Science/Research, Pandora, CA
Seyoung Kim, PhD 2007: Associate Professor, Department of Epidemiology, University of Pittsburgh
Darya Chudova, PhD 2007: Chief Technical Officer, Guardant Health, Redwood City, CA
Sergey Kirshner, PhD 2005: Amazon, Palo Alto, CA
Scott Gaffney, PhD 2004: VP of Search Engineering, eBay, San Jose, CA
Xianping Ge, Google (retired), PhD 2002
Igor V. Cadez, Consultant, PhD 2002
Dimitry Pavlov, Consultant, PhD 2001

Current PhD Students

Advanced to Candidacy: Rachel Longjohn, Markelle Kelly, Yuxin Chang
Pre-Candidacy: Catarina Belem (co-advised with Sameer Singh), Giosue Migliorini, Shang Yu, Coen Adler, Saatvik Kher, Noah Benjamin (co-advised with Stephan Mandt)

Professional Activities

Journals: Associate/Action Editor

ACM Transactions on Knowledge Discovery and Data, guest editor of special issue on best papers from *ACM SIGKDD 2011 Conference*, TKDD 6(4), 2012.

Journal of the American Statistical Association, 2002 to 2005.

IEEE Transactions on Knowledge and Data Engineering, 2002 to 2004.

Machine Learning Journal, July 1998 to December 2001.

Machine Learning Journal, guest editor of special issue on probabilistic learning, 1997.

Journals, Book Series, Centers: Advisory/Editorial Board Member

Journal of Machine Learning Research, 2000-2020.

Journal of Data Mining and Knowledge Discovery, 1997-present.

Chapman and Hall: Series in Computer Science and Data Analysis, 2002-2008.

Bayesian Analysis, 2004-2007.

Insight Center for Data Analytics, University College Dublin, Scientific Advisory Member, 2015-2020.

Conference Program and General Chair Positions

Associate Program Chair, International Joint Conference on Artificial Intelligence (IJCAI), 2022

Program Chair for the Uncertainty in Artificial Intelligence (UAI) Conference, 2013.

Program Chair for 17th ACM SIGKDD Conference, San Diego, 2011.

Program Chair for the Symposium on the Interface between Statistics and Computing, Costa Mesa, CA, June 2001.

General Chair for the Sixth International Conference on Artificial Intelligence and Statistics, January 1997.

Other Conference and Workshop Organization Roles

Conference Organization Roles: Senior Area Chair/Area Chair, NeurIPS 2017, 2018, 2019, 2020,2021,2024; Senior Area Chair/Area Chair, ICML 2018, 2019, 2020, 2021,2022,2023,2024,2025; Senior Area Chair, AAAI 2020; Panels Chair for ACM SIGKDD Fifth International Conference on Knowledge Discovery and Data Mining, 1999; Tutorials co-Chair for National Conference on Artificial Intelligence, 1998; Tutorials Chair for the ACM SIGKDD Conferences on Knowledge Discovery and Data Mining, 1997 and 1998; Publicity Chair for the ACM SIGKDD Conferences on Knowledge Discovery and Data Mining, 1995 and 1996.

Workshop Co-Chair/Organizer for: Dagstuhl Seminar, Automating Data Science, 2018; Workshop on Algorithmic and Statistical Approaches for Large Social Network Data Sets, NIPS Conference, Lake Tahoe, 2012; Workshop on User-Centered Modeling, Institute for Mathematics and its Applications (IMA), University of Minnesota, 2012.; Workshop on Scientific Data Mining, Institute for Pure and Applied Mathematics (IPAM), UCLA, 2002; Workshop on Temporal and Spatial Machine Learning, International Conference on Machine Learning (ICML), 2001; Massive Datasets workshop at the 1998 Neural Information Processing Conference (NIPS).

Publications List

Books and Conference Proceedings

- B5 A. Nicholson and P. Smyth (eds.), *Uncertainty in Artificial Intelligence: Proceedings of the 29th Conference*, ISBN 978-0-9749039-9-6, AUA Press, Corvallis, OR, 2013.
- B4 C. Apte, J. Ghosh, P. Smyth (eds.), *Proceedings of the 17th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining*, ISBN 978-1-4503-0813-7, ACM Press, New York, NY, 2011.
- B3 *Modeling the Internet and the Web: Probabilistic Methods and Algorithms*, P. Baldi, P. Frasconi, and P. Smyth, John Wiley, June 2003.
- B2 *Principles of Data Mining*, D. Hand, H. Mannila, and P. Smyth, Cambridge, MA: MIT Press, 2001.
- B1 *Advances in Knowledge Discovery and Data Mining*, U. Fayyad, G. Piatetsky-Shapiro, P. Smyth, and R. Uthurasamy (eds.), Palo Alto, CA: AAAI/MIT Press, 1996.

Journal Papers

- J95 M. Steyvers, H. Tejada, A. Kumar, C. Belem, S. Karny, X. Hu, L. Mayer, P. Smyth, 'Bridging the gap between what large language models know and what people think they know,' *Nature Machine Intelligence*, to appear.
- J94 R. Longjohn and P. Smyth, 'Likelihood ratios for changepoints in categorical event data with applications in digital forensics,' *Journal of Forensic Sciences*, doi:10.1111/1556-4029.15512, 2024.
- J93 T. Liu, James T. Randerson, Y. Chen, D. C. Morton, E. B. Wiggins, P. Smyth, E. Foufoula-Georgiou, R. Nadler, and O. Nevo, 'Systematically tracking the hourly progression of large wildfires using GOES satellite observations', *Earth System Science Data*, 16(3), 1395-1424, 2024.
- J92 J. J. Guilinger, E. Foufoula-Georgiou, A. B. Gray, J. Randerson, P. Smyth, N. C. Barth, M. L. Goulden, 'Predicting postfire sediment yields of small steep catchments using airborne lidar differencing,' *Geophysical Research Letters*, 50(16), e2023GL104626, doi: 10.1029/2023GL104626, 2023.
- J91 A. Kumar, P. Smyth, M. Steyvers, 'Differentiating mental models of self and others: a hierarchical framework for knowledge assessment,' *Psychological Review*, 130(6): 1566-1591, doi: 10.1037/rev0000443, 2023.
- J90 E. E. Robles, Y. Jin, P. Smyth, R. H. Scheuermann, J. D. Bui, H-Y. Wang, J. Oak, Y. Qian, 'A cell-level discriminative neural network model for diagnosis of blood cancers,' *Bioinformatics*, 39(10), doi: 10.1093/bioinformatics/btad585, October 2023.
- J89 P. Le, J. T. Randerson, R. Willett, S. Wright, P. Smyth, C. Guilloteau, A. Mamalakis, E. Foufoula-Georgiou, 'Climate-driven changes in the predictability of seasonal precipitation,' *Nature Communications*, 14(1):3822, 2023.
- J88 E. Nalisnick, D. Tran, P. Smyth, 'A brief tour of deep learning from a statistical perspective,' *Annual Review of Statistics and its Application*, 10:219-246, <https://doi.org/10.1146/annurev-statistics-032921-013738>, April 2023.
- J87 R. Longjohn, P. Smyth, and H. Stern, 'Likelihood ratios for categorical count data with applications in digital forensics,' *Law, Probability, and Risk*, 21(2):91-122, <https://doi.org/10.1093/lpr/mgac016>, December 2022.
- J86 H. Tejada, A. Kumar, P. Smyth, M. Steyvers, 'AI-assisted decision-making: a cognitive modeling approach to infer latent reliance strategies,' *Computational Brain and Behavior*, 5:491-508, <https://doi.org/10.1007/s42113-022-00157-y>, 2022.

- J85 Y. Chen, S. Hantson, N. Andela, S. Coffield, C. Graff, D. Morton, L. Ott, E. Foufoula-Georgiou, P. Smyth, M. Goulden, J. Randerson, ‘California wildfire spread derived using VIIRS satellite observations and an object-based tracking system,’ *Scientific Data*, 9:249, <https://doi.org/10.1038/s41597-022-01343-0>, 2022.
- J84 M. Steyvers, H. Tejada, G. Kerrigan, P. Smyth, ‘Bayesian modeling of human-AI complementarity,’ *Proceedings of the National Academy of Sciences*, 119(11):1-7, <https://doi.org/10.1073/pnas.2111547119>, March 2022
- J83 H. Do, S. Nandi, P. Putzel, P. Smyth, J. Zhong, ‘A joint fairness model with applications to risk predictions for under-represented populations,’ *Biometrics*, 79(2), 826–840, doi.org/10.1111/biom.13632, published online February 2022 (in print, June 2023).
- J82 A. Mamalakis, J. T. Randerson, J-Y Yu, M. Pritchard, G. Magnúsdóttir, P. Smyth, P. A. Levine, S. Yu, E. Foufoula-Georgiou, ‘Zonally contrasting shifts of the tropical rainbelt in response to climate change,’ *Nature Climate Change*, <https://doi.org/10.1038/s41558-020-00963-x>, 11: 143151, January 2021.
- J81 Park, J., Jindal, A., Kuo, P., Tanana, M., Elston Lafata, J., Tai-Seale, M., Atkins, D. C., Imel, Z. E., Smyth, P, ‘Automated rating of patient and physician emotion in primary care visits,’ *Patient Education and Counseling*, <https://doi.org/10.1016/j.pec.2021.01.004>, 2021.
- J80 A. Stevens, R. Willett, A. Mamalakis, E. Foufoula-Georgiou, A. Tejedor, J. Randerson; P. Smyth, S. Wright., ‘Graph-guided regularized regression of Pacific Ocean climate variables to increase predictive skill of southwestern US winter precipitation,’ *Journal of Climate*, 34(2):737–754, <https://doi.org/10.1175/JCLI-D-20-0079.1>, 2021.
- J79 Y. Chen, J. T. Randerson, S. R. Coffield, E. Foufoula-Georgiou, P. Smyth, C. A. Graff, D. C. Morton, N. Andela, G. R. van der Werf, L. Giglio, L. E. Ott, ‘Forecasting global fire emissions on sub-seasonal to seasonal (S2S) timescales,’ *Journal of Advances in Modeling Earth Systems*, 12(9), e2019MS001955, [doi:/10.1029/2019MS001955](https://doi.org/10.1029/2019MS001955), 2020.
- J78 C. Galbraith, P. Smyth, H. S. Stern, ‘Statistical methods for the forensic analysis of geolocated event data,’ *Forensic Science International*, <https://doi.org/10.1016/j.fsidi.2020.301009>, 33:1–12, July 2020.
- J77 C. Galbraith, P. Smyth, H. Stern, ‘Quantifying the association between discrete event time series with applications to digital forensics,’ *Journal of the Royal Statistical Society A*, 183(3):1005–1027, 2020.
- J76 C. A. Graff, S. R. Coffield, Y. Chen, E. Foufoula-Georgiou, J. T. Randerson, P. Smyth, ‘Forecasting daily wildfire activity using Poisson regression,’ *IEEE Transactions on Geoscience and Remote Sensing*, 58(7):4837–4851, 2020.
- J75 R. Baker, D. Xu, J. Park, R. Yu, Q. Li, B. Cung, C. Fischer, F. Rodriguez, M. Warschauer, P. Smyth, ‘The benefits and caveats of clickstream data to understand student self-regulatory behaviors: opening the black box of learning processes,’ *International Journal of Educational Technology in Higher Education*, 17(13):1–24, 2020.
- J74 D. Ji, P. Putzel, Y. Qian, I. Chang, A. Mandava, R. H. Scheuermann, J. D. Bui, H-Y Wang, P. Smyth, ‘Machine learning of discriminative gate locations for clinical diagnosis,’ *Cytometry A: Special Issue: Machine Learning for Single Cell Data*, 97(3):296–307, 2020.
- J73 C. Fischer, Z. Pardos, R. Baker, J. J. Williams, P. Smyth, R. Yu, S. Slater, R. Baker, M. Warschauer, Mining big data in education: Affordances and challenges, *Review of Research in Education*, 44(1):130-160, 2020.
- J72 S. Coffield, C. Graff, Y. Chen, P. Smyth, E. Foufoula-Georgiou, J. Randerson, ‘Machine learning to predict final fire size at the time of ignition,’ *International Journal of Wildland Fire*, 28(11):861–873, 2019.

- J71 J. Park, D. Kotzias, P. Kuo, R. L. Logan, K. Merced, S. Singh, M. Tanana, E. Karra-Taniskidou, J. Elston Lafata, D. C. Atkins, M. Tai-Seale, Z. E. Imel, and P. Smyth, ‘Detecting conversation topics in primary care office visits from transcripts of patient-provider interactions,’ *Journal of the American Medical Informatics Association (JAMIA)*, 26(12):1493–1504, 2019.
- J70 D. Kotzias, M. Lichman, and P. Smyth, ‘Predicting consumption patterns with repeated and novel events,’ *IEEE Transactions on Knowledge and Data Engineering*, 31(2), 371-384, 2018.
- J69 J. R. Hipp, C. Bates, M. Lichman, and P. Smyth, ‘Using social media to measure temporal ambient population: does it help explain local crime rates?’ *Justice Quarterly*, 36(4), 714-748, March 2018.
- J68 C. Galbraith and P. Smyth, ‘Analyzing user-event data using score-based likelihood ratios with marked point processes,’ *Journal of Digital Investigation*, 22, 106-114, 2017.
- J67 T. Holsclaw, A. M. Greene, A. W. Robertson, P. Smyth, ‘Bayesian non-homogeneous Markov models via Polya-Gamma data augmentation with applications to rainfall modeling’, *Annals of Applied Statistics*, 11(1):393–426, 2017.
- J66 G. Gaut, M. Steyvers, Z. E. Imel, D. C. Atkins, P. Smyth, ‘Content coding of psychotherapy transcripts using labeled topic models,’ *IEEE Journal of Biomedical and Health Informatics*, 21(2):476–487, 2017.
- J65 C. Haffke, G. Magnusdottir, D. Henke, P. Smyth, Y. Peings, ‘Daily states of the March-April east Pacific ITCZ in three decades of high-resolution satellite data,’ *Journal of Climate*, doi:10.1175/JCLI-D-15-0224.1, 29(8):2981-2995, 2016.
- J64 P. Arnesen, T. Holsclaw, P. Smyth, ‘Bayesian detection of changepoints in finite-state Markov chains for multiple sequences,’ *Technometrics*, doi:10.1080/00401706.2015.1044118, 58(2), 205-213, 2016.
- J63 T. Hoslclaw, A. Greene, A. R. Robertson, P. Smyth, ‘A Bayesian hidden Markov model of daily precipitation over South and East Asia,’ *Journal of Hydrometeorology*, doi:10.1175/JHM-D-14-0142.1, 17(1):3–25, 2016.
- J62 T. Hoslclaw, K. A. Hallgren, M. Steyvers, P. Smyth, D. C. Atkins, ‘Measurement error and outcome distributions: Methodological issues in regression analyses of behavioral coding data,’ *Psychology of Addictive Behaviors*, doi:10.1037/adb0000091, 29(4):1031-1040, 2015
- J61 M. L. Salmans, Z. Yu, K. Watanabe, E. Cam, P. Sun, P. Smyth, X. Dai, B. Andersen, ‘The co-factor of LIM domains (CLIM/LDB/NLI) maintains basal mammary epithelial stem cells and promotes breast tumorigenesis,’ *PLOS Genetics*, July 2014, doi: 10.1371/journal.pgen.100452.
- J60 A. J. Frank, P. Smyth, A. T. Ihler, ‘Beyond MAP estimation with the track-oriented multiple hypothesis tracker,’ *IEEE Transactions on Signal Processing*, 62(9):2413–2423, 2014.
- J59 D. C. Atkins, M. Steyvers, Z. E. Imel, P. Smyth, ‘Scaling up the evaluation of psychotherapy: evaluating motivational interviewing fidelity via statistical text classification,’ *Implementation Science*, 9:49:1–11, 2014.
- J58 C. DuBois, C. T. Butts, D. McFarland, P. Smyth, ‘Hierarchical models for relational event sequences,’ *Journal of Mathematical Psychology*, 57(6):297–309, 2013.
- J57 N. Navaroli, C. DuBois, P. Smyth, ‘Modeling individual email patterns over time with latent variable models,’ *Machine Learning*, 92(2–3):431-455, May 2013.
- J56 M. Geyfman, V. Kumar, Q. Liu, R. Ruiz, W. Gordon, F. Espitia, E. Cam, S. E. Millar, P. Smyth, A. Ihler, J. Takahashi, B. Andersen, ‘Bmal1 controls circadian cell proliferation and susceptibility to UVB-induced DNA damage in the epidermis,’ *Proceedings of the National Academies of Science*, 109(29):11758-63, doi:10.1073/pnas.1209592109, July 2012.

- J55 D. Henke, P. Smyth, C. Haffke, G. Magnusdottir, ‘Automated analysis of the temporal behavior of the double Intertropical Convergence Zone over the east Pacific,’ *Remote Sensing of Environment*, 123:418–433, August 2012.
- J54 T. Rubin, A. Chambers, P. Smyth, and M. Steyvers, ‘Statistical topic models for multi-label document classification,’ *Machine Learning*, doi: 10.1007/s10994-011-5272-5, 88(1-2):157–208, July 2012.
- J53 B. Gretarsson, J. O’ Donovan, S. Bostandjiev, T. Hollerer, A. Asuncion, D. Newman, and P. Smyth, ‘TopicNets: Visual analysis of large text corpora with topic modeling,’ *ACM Transactions on Intelligent Systems and Technology*, 3(2):1–26, February 2012.
- J52 M. Steyvers, P. Smyth, and C. Chemudugunta, ‘Combining background knowledge and learned topics,’ *Topics in Cognitive Science*, 3(1):18–47, January 2011.
- J51 A. M. Greene, A. W. Robertson, P. Smyth, and S. Triglia, ‘Downscaling projections of Indian monsoon rainfall using a nonhomogeneous hidden Markov model,’ *Quarterly Journal of the Royal Meteorological Society*, 137(655):347–359, January 2011.
- J50 T. T. Van Leeuwen, A. J. Frank, Y. Jin, P. Smyth, M. L. Goulden, G. R. van der Werf, J. T. Randerson, ‘Optimal use of land surface temperature data to detect changes in tropical forest cover,’ *Journal of Geophysical Research—Biogeosciences*, 116, G02002, doi:10.1029/2010JG00148, 2011.
- J49 A. Asuncion, P. Smyth, and M. Welling, ‘Asynchronous distributed estimation of topic models for document analysis,’ *Statistical Methodology*, 8(1):3–17, January 2011.
- J48 C. Bain, G. Magnusdottir, P. Smyth, H. Stern, ‘The diurnal cycle of the intertropical convergence zone in the east Pacific,’ *Journal of Geophysical Research*, 115, D23116, doi:10.1029/2010JD014835, 2010.
- J47 C. Bain, J. DePaz, J. Kramer, G. Magnusdottir, P. Smyth, H. Stern, C-C. Wang, ‘Detecting the ITCZ in instantaneous satellite data using spatial-temporal statistical modeling: ITCZ climatology in the east Pacific,’ *Journal of Climate*, 138(6):2132–2148, 2010.
- J46 S. Kim, P. Smyth, and H. Stern, ‘A Bayesian mixture approach to modeling spatial activation patterns in multi-site fMRI data,’ *IEEE Transactions on Medical Imaging*, 29(6):1260–1274, June 2010.
- J45 L. Scharenbroich, G. Magnusdottir, P. Smyth, H. Stern and C. Wang, ‘A Bayesian framework for storm tracking using a hidden-state representation,’ *Monthly Weather Review*, 138(6):2132–2148, June 2010.
- J44 Q. Liu, K. K. Lin, B. Andersen, P. Smyth, and A. Ihler, ‘Estimating replicate time-shifts using Gaussian process regression,’ *Bioinformatics*, 26(6):770–776, 2010.
- J43 M. Rosen-Zvi, C. Chemudugunta, T. Griffiths, P. Smyth, and M. Steyvers, ‘Learning author-topic models from text corpora,’ *ACM Transactions on Information Systems*, 28(1):1–38, 2010.
- J42 D. Chudova, A. T. Ihler, K. K. Lin, B. Andersen, P. Smyth, ‘Bayesian detection of non-sinusoidal periodic patterns in circadian expression data,’ *Bioinformatics*, 25(23):3114–3120, 2009.
- J41 D. Newman, A. Asuncion, P. Smyth, and M. Welling, ‘Distributed algorithms for topic models,’ *Journal of Machine Learning Research*, 10:1801–1828, 2009.
- J40 K. K. Lin, V. Kumar, M. Geyfman, D. Chudova, A. T. Ihler, P. Smyth, R. Paus, J. S. Takahashi, B. Andersen, ‘Circadian clock genes contribute to the regulation of hair follicle cycling,’ *PLOS Genetics*, 5(7): e1000573. doi:10.1371/journal.pgen.1000573, 2009.
- J39 A. Ihler, J. Hutchins, and P. Smyth, ‘Learning to detect events with Markov-modulated Poisson processes,’ *ACM Transactions on Knowledge Discovery from Data*, 1(3):1–23, 2007.
- J38 S. J. Gaffney, A. W. Robertson, P. Smyth, S. J. Camargo and M. Ghil, ‘Probabilistic clustering of extratropical cyclones using regression mixture models,’ *Climate Dynamics*, 29(4):423–440, 2007

- J37 S. J. Camargo, A. W. Robertson, S. J. Gaffney, P. Smyth, and M. Ghil, ‘Cluster analysis of typhoon tracks. Part I: general properties,’ *Journal of Climate*, 20:3635–3653, 2007.
- J36 S. J. Camargo, A. W. Robertson, S. J. Gaffney, P. Smyth, and M. Ghil, ‘Cluster analysis of typhoon tracks. Part II: large-scale circulation and ENSO,’ *Journal of Climate*, 20:3654–3676, 2007.
- J35 L. Friedman, Stern, Brown, Mathalon, Turner, Glover, Gollub, Lauriello, Lim, Cannon, Greve, Bockholt, Belger, Mueller, Doty, He, Wells, Smyth, Pieper, Kim, Kubicki, Vangel, and Potkin, Test-retest and between-site reliability in a multicenter fMRI study, *Human Brain Mapping*, 29(8):958–972, 2008.
- J34 A. Ihler, S. Kirshner, M. Ghil, A. Robertson, P. Smyth, ‘Graphical models for statistical inference and data assimilation,’ *Physica D*, 230(1–2):72–87, 2007.
- J33 S. Kim and P. Smyth, ‘Segmental hidden Markov models with random effects for waveform modeling,’ *Journal of Machine Learning Research*, 7(Jun):945–969, 2006.
- J32 A. W. Robertson, S. Kirshner, P. Smyth, S. P. Charles, B. Bates, ‘Subseasonal-to-interdecadal variability of the Australian monsoon over North Queensland,’ *Quarterly Journal of the Royal Meteorological Society*, 132:519–542, 2006.
- J31 Turner, J.A., Smyth, P., Macciardi, F., Fallon, J.H., Kennedy, J.L., Potkin, S.G., ‘Imaging phenotypes and genotypes in schizophrenia,’ *Neuroinformatics*, 4(1):21–50, March 2006.
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- BC1 P. Smyth, J. Statman, G. Oliver and R. Goodman, ‘Combining knowledge-based techniques and simulation with applications to communications network management,’ in *Integrated Network Management II*, I. Krishnan and W. H. Zimmer (eds.), Elsevier Science Publishers, April 1991.

Patents

- U. S. Patent no. 4807280, *Cross-Connect Switch*, assigned to Pacific Bell, inventors are E. C. Posner and P. Smyth, issued February 21 1989.
- U. S. Patent no. 4845736, *Cross-Connect Switch and Method for Providing Test Access Thereto*, assigned to Pacific Bell, inventors are E. C. Posner and P. Smyth, issued July 4 1989.
- U. S. Patent no. 5465321, *Hidden Markov Models for Fault Detection in Dynamic Systems*, assigned to NASA, inventor is P. Smyth, issued November 7 1995.

Research and Training Grants, Contracts and Gifts

- R84 *Time Series Prediction with Deep Learning*, Google gift, \$50,000, September 2024.
- R83 *CAIG: Advancing Wildfire Science, Prediction, and Management with Machine Learning*, NSF 2425932, \$889,209, Oct 1 2024 to Sept 30 2027, Co-principal investigator (PI: Jim Randerson, Earth System Sciences, UCI).
- R82 *Individualized Learning and Prediction for Heterogeneous Multimodal Data from Wearable Devices*, NIH R01 CA297869-01, \$1,200,000, July 1 2024 to June 30th 2028, Multi-Principal Investigator (MPI) (PI: Annie Qu, Statistics, UCI).
- R81 *AI/ML and Data Science Training Datasets*, subaward to NIH 3OT2OD032581, \$489,452, Jan 1 2023 to March 16 2024, Principal Investigator.
- R80 *Improving Prediction of Fire Extremes in the GEOS Forecasting System on Daily and Seasonal Timescales*, NASA, Sept 1 2021 to June 30 2025, \$1,040,166, Co-principal investigator (PI: Jim Randerson, Earth System Sciences, UCI).
- R79 *Fair Risk Predictions for Underrepresented Populations using Electronic Health Records*, NIH R01AG065330-02S1, Sept 1 2021 to April 30 2022, \$167,792, co-investigator, (PI: Judy Zhong, Biostatistics, NYU).
- R78 *Data Science Training and Practices: Preparing a Diverse Workforce via Academic and Industrial Partnership*, NSF IIS-2123366, Sept 1 2021 to Aug 31 2024, \$751,921, Co-principal investigator (PI: Babak Shahbaba, Statistics, UCI).
- R77 *Personalized Risk Predictions with Deep Learning Methods in the Presence of Missing and Biased Electronic Health Record Data*, NIH R01-LM013344, Aug 6 2021 to May 31 2025, \$498,957 (UCI portion), Principal Investigator (MPI with Judy Zhong, Biostatistics, NYU).

- R76 *Center of Excellence in Forensic Statistics (CSAFE2)*, National Institute of Standards and Technology (NIST), award number 70NANB20H019 , \$20,000,000 (\$4,000,000 for UC Irvine), June 2020 to May 2025; co-Investigator (UCI PI: Hal Stern).
- R75 *HPI Research Center in Machine Learning and Data Science at UC Irvine*, Hasso Plattner Institute (gift), April 2020 to Dec 2024, \$3,592,500, Co-principal investigator (PI: Erik Sudderth, UCI).
- R74 *Addressing the Critical Role of Innate/Adaptive Immunity by Integrating Novel Informatics, Translation Technologies and Ongoing Clinical Trial Research*, NIH 3UL1TR001414-06S1, Sept 2020 to June 2021, \$ 1,088,735, co-investigator (PI: Dan Cooper, School of Medicine, UCI).
- R73 *Analyzing Information Exchange in Human-Human Dialog using Machine Learning*, SAP Innovation Center, \$124,000, April 1 2020 to March 31 2021, Principal Investigator.
- R72 *Generative Expectation-based Response and Novelty Identification*, DARPA/SRI-HR001120C0021, \$1,087,251, Oct 1 2019 to May 30 2022, Co-investigator (PI: Stephan Mandt, Computer Science, UCI).
- R71 *Machine Learning Democratization via a Linked, Annotated, Repository of Datasets*, National Science Foundation (CCRI: ENS), award number NSF-1925741, \$1,792,952, Oct 1 2019 to Sept 30 2022. Co-principal investigator (PI: Sameer Singh, Computer Science, UCI).
- R70 *Hybrid Human Algorithm Predictions: Balancing Effort, Accuracy, and Perceived Autonomy*, National Science Foundation (EAGER: AI-DCL), award number NSF-1927245, \$293,923, Aug 15 2019 to Aug 14 2021. Co-principal investigator (PI: Mark Steyvers, Cognitive Sciences, UCI).
- R69 *Assessment of Machine Learning Algorithms in the Wild*, National Science Foundation, award number NSF-1900644, \$1,199,898, Oct 1 2019 to Sept 30 2023, Principal Investigator.
- R68 *Qualcomm Faculty Award*, \$225,000 (gift), May 2019/March 2022, Principal Investigator.
- R67 *Innovation Center for Advancing Ecosystem Climate Solutions*, California Strategic Growth Council, award number CCR20021, \$4,604,140, 4/01/2019 to 3/31/2022, co-investigator (PI: Mike Goulden, Earth Systems Sciences, UCI).
- R66 *Hands-free Documentation in Clinical Practice*, SAP, \$172,000 (gift/sponsored project), October 2018, co-Principal Investigator (with Kai Zheng, Department of Informatics, UCI).
- R65 *TRIPODS-X: Data Science Frontiers in Climate Science*, National Science Foundation, award number NSF-1839336, \$300,000, Oct 1 2018 to Sept 30 2021, co-PI (PI: Efi Foufoula-Georgiou, Civil and Environmental Engineering, UCI).
- R64 *Large-Scale Classification Algorithms*, eBay Labs, \$30,000 (gift), Dec 1 2017, Principal Investigator.
- R63 *Center for Machine Learning and Intelligent Systems*, Cylance, \$50,000 (gift), Dec 1 2017, Principal Investigator.
- R62 *Development of Computational Methods for Evaluating Patient-Doctor Communication*, PCORI, \$270,000 (UCI portion), award number ME-1602-34167, July 1 2017 to June 30th 2019, co-Investigator (PI: Zac Imel, U Utah).
- R61 *NRT-DESE: Team Science for Integrative Graduate Training in Data Science and Physical Science*, NSF, award number NSF-1633631, Sep 15 2016 to Aug 31 2021, \$2,967,150, Principal Investigator.
- R60 *Learning Individual Predictive Choice Models*, Adobe Research Award, \$50,000, October 2016, Principal Investigator.
- R59 *Transformative Computational Infrastructures for Cell-Based Biomarker Diagnostics*, NIH, award number U01TR001801-01, 09/01/16 08/31/21, \$766,000 (UCI portion), co-Investigator (PI: Richard Scheuermann, Venter Institute/UCSD).

- R58 *The Big DIPA: Data Image Processing and Analysis*, NIH BD2K Program, award number 1R25EB022366-01, \$486,000, Sept 30 2015 to June 30th 2018, co-Investigator (UCI PI: Charless Fowlkes).
- R57 *Investigating Virtual Learning Environments*, National Science Foundation, award number NSF-1535300, \$2,500,000, Oct 1 2015 to Sept 30th 2020, co-Investigator (UCI PI: Mark Warschauer).
- R56 *Center of Excellence in Forensic Statistics (CSAFE2)*, National Institute of Standards and Technology (NIST), award number 70NANB15H176, \$20,000,000 (\$3,700,000 for UC Irvine), June 2015 to May 2020; co-Investigator (UCI PI: Hal Stern).
- R55 *Data-Intensive Research and Education Center in Science, Technology, Engineering, and Mathematics (DIRECT-STEM)*, NASA MIRO program, award number NNX15AQ06A, \$5,000,000 (\$1,250,000 for UC Irvine), Sept 1 2015 to Aug 31st 2020, Principal Investigator.
- R54 *Analyzing Individual Event Data over Time*, Google Faculty Research Award, \$60,000, March 2014, Principal Investigator.
- R53 *Peer Assessment and Academic Achievement in a Gateway MOOC*, Bill and Melinda Gates Foundation, Oct 1 2013, \$25,000, Co-Investigator (PI: Mark Warschauer, UC Irvine).
- R52 *Statistical Learning Algorithms for Micro-Event Time Series Data*, National Science Foundation, award number IIS-1320527, Oct 1 2013 to Sept 30th 2018, \$499,880, Principal Investigator.
- R51 *Balancing the Portfolio: Efficiency and Productivity of Federal Biomedical R&D Funding*, National Science Foundation, award number 1158699, Aug 15 2012 to July 31 2015, \$297,331, Principal Investigator (original PI, David Newman).
- R50 *Location-based Social Media for Context-based Analysis of Transportation Data*, Xerox UAC Research Award, Jan 1st 2013 to Dec 31st 2015, \$90,000 gift, Principal Investigator.
- R49 *Collaborative Research, Type 1: Decadal Prediction and Stochastic Simulation of Hydroclimate over Monsoonal Asia*, US Department of Energy, award number DOE SC0006619, Sept 1st 2011 to August 31st 2014, \$180,000, Co-Investigator (PI: Andrew Robertson, Columbia University).
- R48 *Copernicus: System for Foresight and Understanding from Scientific Exposition*, IARPA, contract number D11PC20155, September 2011 to August 2016, \$1,097,420, Principal Investigator.
- R47 *Probabilistic Alignment and Distributed Analytics*, IARPA/AFRL FA8650-10-C-7060, Oct 1 2010 to Dec 31 2011, \$334,537, Principal Investigator.
- R46 *Biomedical Informatics Training Program (supplement)*, award number NIH LM07443-10S1, 7/1/10-6/30/11, \$153,485, Senior Personnel (PI: Pierre Baldi, UC Irvine).
- R45 *Automating Behavioral Coding via Text-Mining and Speech Signal Processing*, National Institutes of Health, award number R01AA018673, \$3.1 million, (UC Irvine portion is \$953,952), Sept 1 2010 to August 31 2015, Co-Investigator (PI: David Atkins, University of Washington).
- R44 *UC Irvine Clinical Translational Science Center*, National Institutes of Health, award number UL1RR031985, \$7,075,320 awarded to date, July 1 2010 to March 31st 2015, Senior Personnel (PI: Dan Cooper, UC Irvine).
- R43 *Scaling Statistical Topic Modeling Algorithms to Massive Data Sets*, Yahoo! Faculty Research (FREP) award, \$10,000 gift, May 2010, Principal Investigator.
- R42 *Scalable Methods for the Analysis of Network-based Data*, Office of Naval Research: Multidisciplinary University Research Initiative (MURI) Award), award number N00014-08-1-1015, \$5,381,300, May 1 2008 to April 30 2013, Principal Investigator.

- R41 *Scaling Statistical Topic Modeling Algorithms to Massive Data Sets*, Google Research Award, \$60,000, April 2008, Principal Investigator.
- R40 *Research in Cyber-Fraud Detection and Prevention*, gift from Experian, Inc., \$200,000, February 2008, Co-Principal Investigator with Michael Goodrich.
- R39 *Collaborative Research: Regional Climate-Change Projections Through Next-Generation Empirical and Dynamical Models*, Department of Energy, Scientific Discovery through Advanced Computing: Climate Change Prediction, award number DE-FG02-07ER64429, \$360,000, Oct 1 2007 to Sept 30 2010, Principal Investigator.
- R38 *CRI: Collaborative Research: Improving Experimental Computer Science with a Searchable Web Portal for Datasets*, National Science Foundation, award number CNS-0551510, \$400,000, March 15, 2006 to February 28, 2009, Co-Principal Investigator with Andrew McCallum (University of Massachusetts).
- R37 *Functional Biomedical Informatics Research Network (FBIRN)*, National Institutes of Health, U24RR021992, \$23,992,092, from February 8th 2006 to November 30th 2010, Senior Personnel (PI: Steven Potkin, UC Irvine).
- R36 *Characterizing ITCZ Dynamics and Breakdown using Statistical Learning Methods and Satellite Data*, National Science Foundation, award number ATM-0530926, \$618,000, 10/1/2005 to 9/30/2008, Co-Investigator (PI: Gudrun Magnusdottir, UC Irvine).
- R35 *UC Irvine Knowledge Discovery Evaluation Challenge Project*, Entity Analytics Division, International Business Machines (IBM), \$73,430, 7/15/05 to 12/31/05, Principal Investigator.
- R34 *Bringing Probabilistic Text Mining Techniques to Historical Document Collections: An Early American Case Study*, UCI CORCLR Award MI-05-06-14, \$18,080, 7/1/2005 - 6/30/2006, Co-Investigator (PI: Sharon Block, UC Irvine).
- R33 *Transdisciplinary Imaging Genetics Center*, NIH Grant No. 1-P20-RR020837-01, total award is \$1,724,026, 9/28/04 to 7/31/07, Co-Investigator (PI: Steven Potkin, UC Irvine).
- R32 *National Alliance for Medical Image Computing (NAMIC)*, National Institutes of Health, award number NIH U54 EB005149, total UCI award is \$609,253 from 9/17/04 to 8/31/06, Co-Investigator (PI: Ron Kikinis, Brigham and Women's Hospital).
- R31 *Morphometry Biomedical Informatics Research Network (MBIRN)*, National Institutes of Health, U24-RR021382, total UCI award is \$579,880 from 9/30/04 to 5/31/06, Co-Investigator (PI: Bruce Rosen, Massachusetts General Hospital).
- R30 *Studies of regional-scale climate variability and change: Hidden Markov models and coupled ocean-atmosphere modes*, funded by the Climate Change Prediction Program, US Department of Energy, October 1st 2004 to September 30th 2007, Principal Investigator.
- R29 *Statistical Data Mining of Time-Dependent Data with Applications in Geoscience and Biology*, NSF-IIS-0431085, National Science Foundation, \$566,644, October 1st 2004 to September 30th 2007, Principal Investigator.
- R28 *NSF-ITR: Responding to the Unexpected*, Information Technology Research (ITR) program, National Science Foundation, \$9,480,928, award number NSF-ITR-0331707, October 1st 2003 to September 30th 2008, Co-Investigator (PI: Sharad Mehrotra, UC Irvine).
- R27 *NSF-ITR: The OptIPuter*, Information Technology Research (ITR) program, National Science Foundation, award number , \$13,500,000, October 1st 2002 to September 30th 2007, Co-Investigator (PI: Larry Smarr, UCSD).
- R26 *Biomedical Informatics Training Program*, National Institutes of Health and National Library of Medicine, award number T15-LM-07443, \$8,840,297, July 1st 2002 to June 30th 2012, Senior Personnel (PI: Pierre Baldi, UC Irvine).

- R25 *Predicting Coupled Ocean-Atmosphere Modes With A Climate Modeling Hierarchy*, US Department of Energy: Climate Change Prediction Program, \$396,000, February 1st 2002 to January 31st 2005, Co-Investigator (with Andrew Robertson and Michael Ghil, UCLA).
- R24 *Intelligent Time-Series Pattern Matching*, Jet Propulsion Laboratory, June 15th to September 30th 2002, \$80,920, Principal Investigator.
- R23 *Preclinical Detection and Disease Measurement of Alzheimer's Disease and Related Disorders Using EEG, Psychophysical and Data Mining Methods*, Alzheimer's Association of America, September 1st 2001 to August 30th 2003, \$250,000, Co-Investigator (PI: Rod Shankle, UC Irvine).
- R22 *Spatial Data Mining for Massive Scientific Data Sets*, Lawrence Livermore National Laboratory, May 1st 2001 to August 31st 2002, \$100,000, Principal Investigator.
- R21 *IBM Faculty Partnership Award*, gift from IBM Watson Research Center, May 18th 2001, \$40,000, Principal Investigator.
- R20 *Data Mining of Digital Behavior*, NSF-IIS-0083489, Principal Investigator:
- Original award: September 15th 2001 to August 30th 2004, \$425,000.
 - Supplemental award: September 1st 2003 to December 31st 2010, \$1,816,750.
- R19 *Predictive Models for Cancer Detection and Therapy*, November 1st 2000 to October 31st 2001, University of California, Irvine, Cancer Research Grants, \$14,301, Co-Investigator (PI: Christine McLaren, UC Irvine).
- R18 *Probabilistic Clustering of Dynamic Trajectories for Scientific Data Mining*, Institute for Scientific Computer Research, Lawrence Livermore National Laboratory, October 1 2000 to September 30 2001, \$39,178, Renewal: October 1 2001 to September 30 2002, \$28,448, Principal Investigator.
- R17 *Sequential Data Analysis for Biomedical Applications*, UCI CORCLR Program, July 1 2000 to June 30th 2001, \$12,000, Co-Investigator (PI: Christine McLaren, UC Irvine).
- R16 *Spatio-Temporal Data Mining of Scientific Trajectory Data*, Lawrence Livermore National Laboratory, March 1st to September 30th 2000, \$42,937, Principal Investigator.
- R15 *Research in Data Mining*, gift from Microsoft Research, October 1999, \$60,000, Principal Investigator.
- R14 *Data Mining of Multivariate Time-Series Sensor Data for Semiconductor Manufacturing*, NIST/National Semiconductor corporation, April 1 1999 through Dec 31 2001, \$162,000, Principal Investigator.
- R13 *Clustering of Sequences and Time Series*, HNC Software, Inc, \$40,913, January 1 1999 through Dec 31 1999, Principal Investigator.
- R12 *SGER: An Online Repository of Large Data Sets for Data Mining Research and Experimentation*, National Science Foundation, NSF IIS-9813584, Aug 15, 1998 to January 31, 2000, \$99,737, Principal Investigator.
- R11 *Data Mining of High-Dimensional Structure-Activity Data Sets*, from SmithKline Beecham Research, September 1st 1998 to April 1st 1999, \$22,730, Principal Investigator.
- R10 *Graduate Fellowships in Biomedical Computing*, US Department of Education, \$750,000. Sept 1, 1997 to August 31, 2001, Co-Investigator (PI: Lubomir Bic, UC Irvine).
- R9 *A Distributed Biomedical Computing Laboratory*, National Science Foundation (CISE Research Instrumentation), NSF-9617349, co-investigator with L. Bic et al. (University of California, Irvine), March 1 1997 to February 1 1998, \$69,986. Co-Investigator.

- R8 *Turbo-Decoding of High Performance Error-Correcting Codes via Belief Propagation*, AFOSR, grant F49620-97-1-0313, May 1 1997 to December 31 1998, \$300,000. Co-Investigator (PI: Robert McEliece, Caltech).
- R7 *Automated Cloud Screening for Remote Exploration and Experimentation (REE) Applications to the Earth Orbiting-1 (EO-1) Satellite and Similar Platforms*, the Jet Propulsion Laboratory, June 16th 1997 to November 15th 1997, \$34,601, Principal Investigator.
- R6 *Exploring QSAR Data using Probabilistic Data Mining*, SmithKline Beecham Research, July 1st to December 31st 1997, \$35,048, Principal Investigator.
- R5 *Probabilistic Knowledge Discovery and Data Mining: An Integrated Approach at the Interface of Computer Science and Statistics*, National Science Foundation (CAREER award), NSF-9703120, September 1st 1997 to August 31st 2001, \$304,379, Principal Investigator.
- R4 *Clustering and Mode Classification of Engineering Time Series Data*, Jet Propulsion Laboratory, June 15th 1996 to October 17th 1996, \$34,401, Principal Investigator.
- R3 *Automated Detection of Natural Features in SAR Images*, Jet Propulsion Laboratory Director's Discretionary Fund, January 1st 1994 to December 31st 1994, \$140,000, Co-Investigator with Usama Fayyad (JPL) and Pietro Perona (Caltech).
- R2 *Using Information Theory to Discover Patterns in Databases*, Lew Allen Award research grant, Jet Propulsion Laboratory. January 1st 1994 to December 31st 1995, \$25,000, Principal Investigator.
- R1 *An Information-Theoretic Approach to Distributed Inference and Learning*, AFOSR, and ONR. Original award AFOSR-90-0199, February 1st 1990 to May 30th 1992, \$338,161. Continuation award NOOO14-92-J-1860: July 1st 1992 to March 30th 1995, \$394,118. Co-Investigator (PI: Rodney Goodman, Caltech).

Reviewing: Funding Agencies, Journals, Conferences

Reviewer for *National Science Foundation, Science, Journal of Machine Learning Research, Communications of the ACM, Journal of the American Statistical Association, Bayesian Analysis, IEEE Transactions on Information Theory, IEEE Trans. on Neural Networks, IEEE Trans. on Signal Processing, IEEE Trans. on Circuits and Systems, IEEE Trans. on Pattern Analysis and Machine Intelligence, IEEE Trans. on Knowledge and Data Engineering, Statistics and Computing, Journal of Artificial Intelligence Research, Pattern Recognition Letters, Neural Networks, Machine Learning, NeurIPS, ICML, UCI, AI-Stats, ECML/PKDD, ACM SIGKDD, WWW, ICPR, IJCAI, AAAI.*

Conference Keynote Talks (Invited)

- Open Data Science Conference, *Overconfidence in machine learning: do our models know what they don't know?*, Boston, April 2022.
- Fudan International Conference on Data Science, *Deep learning and statistics: connections*, Fudan University, Shanghai, December 2018.
- Annual Irish Conference on Artificial Intelligence and Cognitive Science (AICS), *Promise and peril in machine learning*, September 2016
- British International Conference on Databases (BICOD), *Statistical thinking in machine learning*, Edinburgh, UK, July 2015.
- Climate Informatics 2014, *Climate data analysis with machine learning and statistics: a tale of two tribes*, National Center for Atmospheric Research (NCAR), Boulder, September 2014.

AAAI Conference on Artificial Intelligence (AAAI-14), *30 years of probability in AI and machine learning*, Quebec City, Canada, August 2014.

SIAM Data Mining Conference (SDM), *Modeling individual-level data in the 21st century*, Austin, Texas, 2013.

European Conference on Machine Learning and Principles of Knowledge Discovery in Data (ECML/PKDD), *Analyzing text and social network data with probabilistic models*, Bristol, UK, 2012.

41st Hawaii International Conference on System Sciences (HICSS), *From Gauss to Google: data analysis in the digital age*, Hawaii, 2008

Algorithmic Learning Theory (ALT) and Discovery Science (DS) 2006 Conferences, *Learning from data with probabilistic hidden variable models*, Barcelona, October 2006

Irish Conference on Artificial Intelligence, *Probabilistic learning and artificial intelligence: a review and update*, June 2003

European Conference on Machine Learning (ECML) and Principles of Knowledge Discovery in Data (PKDD), *Learning with mixture models: concepts and applications*, Helsinki, Finland, August 2002

SIAM Conference on Data Mining (SDM), *Mixture models for data exploration and prediction* Washington DC, April 2002

18th International Conference on Machine Learning (ICML), *A guided tour of finite mixture models: from Pearson to the Web* Williamstown, MA, June 2001

Joint session of the 1997 National Conference on Artificial Intelligence (AAAI) and IAAI-97, *Recent advances in knowledge discovery in databases*, Providence, RI, July 1997

Invited Research Seminars and Talks

- University of California Clinical Research Summit, Keynote talk, October 2024
- NIH Cancer Biomarkers and AI Workshop, Caltech, August 2024
- German National Library of Economics (ZBW/Leibniz), Kiel, Seminar, July 2024
- SAP Labs Developer Conference, SAP Newport Beach, Keynote Talk, June 2024
- New York University, Biostatistics Department Seminar, June 2024
- University College Dublin, SFI Center for Research Training in Machine Learning, June 2022
- Caltech, Electrical Engineering Department, December 2022
- Hasso-Plattner Research Symposium (virtual), Potsdam, April 2021
- Qualcomm ML Summit, San Diego, November 2019
- Monash University, Melbourne, Australia, Faculty of Information Technology, Dean's seminar series, July 2019
- University of Ghent, Belgium, UGent Data Science Seminar, May 2019
- Southern California AI and Biomedicine Symposium, UC Irvine, May 2019
- General Motors Research Symposium, Detroit, July 2018
- UC Irvine Department of Statistics Seminar, May 2019
- National University of Ireland (NUIG), Galway, School of Informatics, October 2017

- University of New South Wales, Electrical Engineering, August 2017
- Trinity College Dublin, School of Computer Science and Statistics, June 2017
- University of Amsterdam, Informatics Institute, June 2017
- Adobe Data Science Symposium, San Jose, June 2017
- California State University of Los Angeles, February 2017
- Adobe Research, San Jose, November 2016
- Amazon Machine Learning, Berlin, June 2016
- Jet Propulsion Laboratory, Pasadena, April 2016
- University of Edinburgh, Department of Informatics, July 2015
- Xerox Webster Research Labs, Rochester, NY, June 2015
- Cisco Research, San Jose, May 2015
- Insight Center for Data Analytics, University College Dublin, March 2015
- Hasso-Plattner Institute, University of Potsdam, March 2015
- Committee on National Statistics, Beckman National Academies Center, Irvine, February 2015
- Facebook, Palo Alto, September 2014
- Yahoo! Labs, Barcelona, Spain, July 2014
- Xerox Webster Research Labs, Rochester, April 2014 and June 2015
- eBay Research Labs, San Jose, April 2014
- IBM Research Center, Dublin, Ireland, September 2013
- Boeing Research & Technology Seminar, Bellevue, WA, July 2013
- L'Institut des Systèmes Complexes (ISC-PIF), Paris, April 2013
- École Normale Supérieure (ENS), Laboratoire de Météorologie Dynamique (LMD), Paris, April 2013
- Stanford Research Institute (SRI), Artificial Intelligence Center Seminar Series, September 2012
- Technische Universität Berlin (TU Berlin), Machine Learning Group Seminar, Berlin, June 2012
- Adobe Research, San Jose, February 2012
- UCLA, Statistics Seminar Series, October 2010
- National University of Ireland, Galway (NUIG), October 2010
- Fraunhofer Institute for Intelligent Analysis and Information Systems (IAIS) Seminar, May 2010
- Georgia Tech, College of Computing Seminar, April 2010
- Google Tech Talk, September 2009
- eBay Research Labs Seminar, September 2009
- Georgia Tech, College of Computing Seminar, April 2009
- National University of Ireland, Galway (NUIG), Information Technology Seminar, April 2009

- ETH Zurich, Computer Science Department Seminar, April 2008
- Caltech, IST Seminar Series, November 2006
- ISI/USC Seminar series, December 2006
- Yahoo! Research Labs Seminar, Sunnyvale, CA, February 2005
- UCSD, Computer Science and Engineering Seminar, November 2004
- Tufts University, Computer Science Department, October 2004
- Columbia University, International Research Institute for Climate Prediction, September 2004
- University of Washington, Computer Science Department Seminar, August 2004
- National University of Ireland, Galway (NUIG), Information Technology Seminar, July 2004
- Georgia Tech, College of Computing Seminar, April 2004
- Carnegie Mellon University, School of Computer Science, April 2004
- UC Davis, Department of Statistics Seminar, October 2003
- Amazon, Seattle, May 2003
- UCSD, Computer Science Department Seminar, November 2001
- Queen's University, Belfast, Department of Computer Science, August 2001
- IBM Watson Research Seminar, Yorktown Heights, NY, June 2001
- Helsinki Institute of Technology, Computer Science Department, November 2000
- Stanford University, Computer Science Department, November 2000
- RAND Corporation Seminar, October 2000
- Lawrence Livermore National Laboratory (LLNL), May 2000
- NEC/Teradata Research Center, San Diego, CA, April 2000
- Microsoft Research Seminar, August 1999
- IBM Almaden Research Center, July 1999
- Jet Propulsion Laboratory, March 1999
- UCLA, Department of Computer Science Seminar, February 1999
- UCLA, Department of Atmospheric Sciences Seminar, February 1999
- Smith-Kline Beecham Research Laboratories, Philadelphia, September 1996
- Microsoft Research Seminar, Redmond, WA, August, 1996
- Carnegie-Mellon University, Computer Science Department Seminar, July 1996
- Caltech, Electrical Engineering Systems Seminar, May 1996
- European Space Agency, Noordwijk, The Netherlands, June 1994
- Caltech, Electrical Engineering Systems Seminar, October 1992

Public Outreach Talks

- Newport Beach Public Library Science Seminar, “AI and Science”, November 2023
- ‘Big data: hope or hype?’
 - ACM Orange County Chapter, November 2015
 - Osher Lifelong Learning Institute and UC Irvine UClub Forum Talk Series, October
 - National Academy of Sciences Distinctive Voices Program, Beckman NAS Center, Irvine, June 2014