Jun Yu

Redmond, WA 98052 | (541) 678-8666 | yujunnokia@gmail.com | sites.google.com/site/yujunnokia

A proven track record of developing machine learning algorithms on large-scale data and QUALIFICATION leading a team to build end-to-end machine learning solution to deliver business value. □ 10+ years experience in developing machine learning algorithms and applying machine learning techniques to solve large scale real world problems. Computer Science PhD focusing on machine learning and probabilistic graphical models with 15+ publications in top-tier artificial intelligence journals and conferences. **EXPERIENCE** Amazon - Senior Applied Scientist and Tech Lead October 2017 – Now ☐ [Promotion] Lead a team of applied scientists to forecast sale velocity for various Amazon deals, generate promotion recommendations for sellers and vendors and optimize the deal scheduling to maximize customer satisfaction. We build large-scale demand forecasting models on Spark and a ML inference platform to surface model predictions in real time. eBay - Applied Science Tech Lead and Engineer December 2013 – October 2017 ☐ [Marketing] Led a team of 8 applied researchers to power eBay's paid internet marketing improving the bidding strategies on Google and Facebook to maximize ROI. Our models improved the ROI by 27% in 2016, leading to millions more GMB annually. ☐ [Trust] Built a large-scale machine learning model to predict seller risk and reduce the number of defective transactions on eBay via search ranking demotion, which translates to 100+ million GMB lift annually. [4] ☐ [Search] Implemented a topic model based approach to retrieve diverse items based on user buying intents and improved the user satisfaction by 6+%. [8] ☐ Award: eBay Critical Talent Award & eBay Seattle Technical Achievement Award. Oregon State University - PhD Researcher September 2006 – December 2013 Developed probabilistic graphical models to predict species distribution with citizen science data, which significantly improved the accuracy of species distribution. [3, 5, 10, 16, 6, 14]. Developed a generative mixture model in the eBird human/computer learning network to quantify the skill level of citizen scientists in the eBird project. [1, 2, 7, 9, 11, 12] **EDUCATION** Ph.D. in Computer Science September 2006 – December 2013 Oregon State University, OR GPA: 3.98 Thesis: Machine Learning For Improving The Quality of Citizen Science Data. **B.S. in Computer Science** September 2002 – June 2006 Wuhan University, China SKILL SETS **Python and R** for training machine learning models, data analytics and rapid prototyping.

Python and R for training machine learning models, data analytics and rapid prototyping. **Scala and Java** for building and deploying end-to-end application in production. **Spark and Scalding** for building ML pipelines and training larges-scale ML models. **SQL** for ad-hoc queries on RDBMS

PUBLICATION

- [1] Can observation skills of citizen scientists be estimated using species accumulation curves? PLOS-ONE, 2015.
- [2] The eBird enterprise: an integrated approach to the development and application of citizen science. Biological Conservation, 2014.
- [3] eBird: a human/computer learning network for biodiversity conservation and research. Al magazine, 2013.
- [4] Bayesian and empirical Bayesian forests. ICML. 2015.
- [5] Modeling misidentification of bird species by citizen scientists. AAAI. 2014.
- [6] HC-Search for Multi-Label Prediction: An Empirical Study. AAAI. 2014.
- [7] Clustering species accumulation curves to identify groups of citizen scientists with similar skill levels. IAAI. 2014.
- [8] Latent Dirichlet Allocation based diversified retrieval for e-commerce search. WSDM. 2014.
- [9] Clustering species accumulation curves to identify groups of citizen scientists with similar skill levels. NIPS workshop. 2013.
- [10] Modeling misidentification of bird species by citizen scientists. NIPS workshop, 2013.
- [11] Automated data verification in a large-scale citizen science project: a case study. eScience, 2013.
- [12] eBird: a human/computer learning network for biodiversity conservation and research. IAAI. 2012.
- [13] Crowdsourcing citizen science data quality with a human-computer learning network. NIPS workshop. 2013.
- [14] Multi-label classification for species distribution modeling. ICML workshop. 2011.
- [15] The implementation of automated data verification processes in a large-scale citizen science project. eScience workshop. 2012.
- [16] Modeling experts and novices in citizen science data for species distribution modeling. ICDM. 2010.
- [17] Learning algorithms for link prediction based on chance-constraints. ECML. 2010.
- [18] Chance-constrained programs for link prediction. NIPS workshop. 2009.