# Wei Fu

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## **EDUCATION**

North Carolina State University, Raleigh, NC

May. 2013 - May. 2018

Ph.D. Computer Science | Adviser: Dr. Tim Menzies

Beijing University of Posts and Telecommunications, Beijing, China

Sep. 2009 - Mar. 2012

M.S. Electrical Engineering

Nanjing University of Technology, Nanjing, China

Sep. 2005 - Jun. 2009

B.S. Electrical Engineering

#### **SKILLS AND INTERESTS**

- Deliver computer vision based solutions with deep learning and traditional machine learning methods.
- Machine learning solution development: data analysis/ organization, ML algorithm exploration, error analysis.
- Tools and Frameworks: Python/C++, Scikit-learn, TensorFlow, PyTorch, AWS, docker.
- Interests: problems can be solved with machine learning.

#### **INDUSTRY EXPERIENCE**

# Machine Learning Engineer | Landing Al, Palo Alto, CA

Data Pipeline, ML Algorithm Exploration, Model Iteration/Optimization, Error Analysis

May. 2018 - Now

- Designed, implement and deployed a computer vision-based software solution to a multinational electronics contract manufacturer to inspect visual defects on IC chips. Owned data pipeline, model iteration and error analysis.
- Key contributor to design and implement Al-harvesters to enable automatic harvesting. My contributions include data labeling pipeline, building regression set, algorithm research, key computer vision model iterations and error analysis.
- Drive to design, build and implement an end to end solution for a multinational biopharmaceutical company to do syringe quality control. The system greatly helps the customer to reduce the false positive of the existing AVI system.

## Research Intern | ABB, Raleigh, NC

Software Project Data Visualization and Exploration

May. 2016 - Aug. 2016

- Design mapping rules and building a web interface to visualize and manage data with asp.net and database.
- Based on ABB industrial data, building predictive models to improve quality of software development process.
- Compare and analyze the different characteristics of open source software data and proprietary data for building analytics models.

# **RESEARCH PROJECTS**

#### **NSF Funded: Search-based Software Engineering Research**

Sep, 2014 - May, 2018

Research Assistant Under Dr. Tim Menzies, North Carolina State University, USA

- Hyper-parameter Tuning for Software Analytics: Software researchers and practitioners routinely use machine learning to explore software project data. However, they rarely tune hyper-parameter of their learners. By applying *Differential Evolution* on defect predictors, we find it (DE) can quickly find tunings that alter detection precision from 0% to 60%; DE can dramatically reduce clustering instability for LDA.
- **Differential Evolution v.s. Grid Search:** Grid search has become a de-facto hyper-parameter tuning technique for machine learning algorithms over years. However, for software analytics, we show that Differential Evolution as a tuner has better performance than grid search and also 210X faster.
- Supervised v.s. Unsupervised Learning: There's a debate about choices of defect predictors. In this project, we show that, in practice, unsupervised learning is not stable for deploying defect prediction, some supervised data is required to prune weaker models when building effort-aware just-in-time defect predictors.
- Simple Techniques for Software Analytics: Deep Learning has become a buzzword in both academia and industry. It seems that every single task should be solved by deep learning. In this project, we revisited a SE task recently solved by deep learning. However, after applying differential evolution-based parameter tuning on SVM, our results outperform the deep learning method in terms of performance metrics and also 84X faster.

## **NSF Funded: Transfer Knowledge between Software Projects**

Aug, 2015 - Dec, 2017

Research Assistant Under Dr. Tim Menzies, North Carolina State University, USA

- **Heterogeneous Defect Prediction:** For new software projects, historical data is missing. To build defect predictors, we proposed that historical data with different metrics from different projects can be used to build software quality models to predict quality of the target project. By using the mathematical models, we identify categories of data sets as few as 50 instances are enough to build a defect prediction model.
- Bellwether Effect in Software Analytics We find a "bellwether" effect in software analytics. Given N data sets, we find there always one data set produces the best predictions on all the others. This "bellwether" data set then can be used for all subsequent predictions.

## **SELECTED PUBLICATIONS**

- A. Agrawal, **W. Fu**, D. Chen, X. Shen, and T. Menzies. *How to" DODGE" Complex Software Analytics?*. **TSE**, 2019. △ http://tiny.cc/Dodge.
- D. Chen, **W. Fu**, R. Krishna, and T. Menzies. *Applications of psychological science for actionable analytics.* **FSE'18**. △→tiny.cc/wfufft.
- W. Fu and T. Menzies. Easy over Hard: A Case Study on Deep Learning. FSE'17. △ tiny.cc/wfuDL.
- W. Fu and T. Menzies. Revisiting Unsupervised Learning for Defect Prediction. FSE'17. △ → tiny.cc/wfuOneWay.
- JC Nam, **W. Fu**, S. Kim, T. Menzies, and L. Tan. *Heterogeneous Defect Prediction*. **TSE**, 2017. △ tiny.cc/wfuHDP.
- W. Fu, T. Menzies, X. Shen, *Tuning for Software Analytics: is it Really Necessary.* IST, 2016. △ tiny.cc/wfuTuning.
- R. Krishna, T. Menzies, and W. Fu. Too Much Automation? the Bellwether Effect and Its Implications for Transfer Learning. ASE'16. △→tiny.cc/wfuBellwether.