

## Work Experience

- 2022.03 - Now **Machine Learning Engineer, ASR and Language Tech** @ Zoom

Tech stack: Torch/Pytorch, ONNX, Tensorrt, Kubernetes, Pynn/Kaldi

- Evaluate and turn the ASR (automatic speech recognition) system parameters to minimize its inference latency (i.e. real time factor, RTF) and maximize its throughput.
- Implement, test, maintain and review the critical features in inference side for offline ASR. This mainly includes the asr decoding codes, attention based word level timestamp generation and post-processing/punctuation codes.
- Train and experiment on the transcription model.
- Watch closely on online bugs reported from customers, and fix them either from code, or from data quality side.
- Working closely with Intel and Nvidia, explore the ways of transforming current system using ONNX, Intel extension for Pytorch and TensorRT for speeding up the inference and minimizing the cost of the existing inference system.
- Work with web team, infra team and downstream services, design internal changes for offline ASR system for better fulfilling the transcription requirements from other teams.

- 2020.09 - 2021.12 **Data Scientist** @ Barclays

Tech stack: Spark / PySpark (on Elastic Data Platform), Amazon Deep Java Library (DJL), Tensorflow / Keras, BitBucket, Pandas, Jupyter, Pretrained Transformers / Transformers / Likelihood Ratio

- Company address matching and entity matching without internal GPU and labeled data available. Solve using an active learning method. Start from constructing some small datasets only with external data and training an XGBoost tree, then label samples in the boundary and fine-tune BERT models in an iterative way. Finish the inference on 6 million internal pair-wise samples with this model on a CPU cluster, using a DJL based pipeline built from scratch on my own. It achieved a very satisfying result of 94% F1 score on a noisy testing dataset. The model does inference offline on our Spark cluster in a distributed way. For 6 million pair-wise samples, the running time is under 1 hour (on a cluster with 80 CPUs).
- Predict the aggregated user's transaction activity (volume and value) using the historical mean and Informer model, a variant of Transformer for time-series modeling. Following that, a counterfactual was constructed to provide an evaluation of how much finance loss that the bank suffers from system downtime and to find out the critical period for the system reliability.
- Maintain the Spark cluster for the team, and build up pipelines for distributed inference by combining DJL / PySpark UDF with models. Collaborated with one of my colleagues, we created a team-wise package to start a Spark session within 4 lines of codes, which significantly reduces the overhead of using Spark for colleagues who are not with a distributed computing background.

- 2019.08 - 2020.09 **Java Backend Developer** @ Barclays

Tech stack: Jenkins, Jira, Confluence, BitBucket, OpenShift (Kubernetes), Docker, GridGain, Maven, Gradle, Wiremock, Mockito, Spring Boot, SonarQube, Karate, AppDynamics

- End-to-end function development, testing (unit, functional, performance), deployment (CD)
- Add cache layer to the existing APIs to reduce the latency for repetitive data access
- Migrate legacy codes to internal Spring Boot templates, with refactors to enhance code readability and performance
- Build up handy internal tools (e.g. git hooks) and scripts (python / bash) from scratch to automate software development processes

## Education

2018 - 2019	MSc Web Science and Big Data Analytics @ <b>University College London</b> , Distinction
2016 - 2018	BSc Internet Computing @ <b>University of Liverpool</b> , First class
2014 - 2016	BSc Information and Computing Science @ <b>Xi'an Jiaotong-Liverpool University</b>

2+2 pathway routine (first 2 years in Suzhou, China and final 2 years in Liverpool, UK), dual degree.

## Selected Projects

- 2021.09 - 2021.10 **Wechat chat history analysis**

It was a gift for one of my important friends for a friendship anniversary. I collected all of our chat history (in Chinese). Take an analysis of the following aspects: the emotion appeared in our single chat sentences (80% accuracy for fine-tuning a Chinese version Roberta model for a 6-classes dataset including happy, natural, angry, fearful, anxious, exciting), word cloud generation, Wechat emoji counting, and hourly chat statistics. The final delivery is a mobile html5 page constructed using a library wechat-h5-boilerplate. This was an end-to-end project from data collection (WeChat does not provide any public way of exporting chat history).

- 2021.08 - 2021.09 **Implementation of Conflict-Free Replicated Data Type (CRDT) - set and graph**

A project from one of my previous take-home interview exercises finished in a week. It is a self-contained, fully-functional and properly-tested implementation of Last Write Win (LWW) Graph and Set in Python, which is one type of CRDT. CRDTs can be replicated across systems, they can be updated independently and concurrently without coordination between the replicas, and it is always mathematically possible to resolve inconsistencies that might result. Github: <https://github.com/billweasley/Last-write-win-CRDT-graph>

- 2019.06 - 2019.09 **Project Internship (Master Degree Thesis) @ Astroscreen**

Social media posting language source identification (tweets and gabs) project. Finished a crawler for collecting language (posts) data from Gab.com, pre-processed data using Regular Expression, built models for classifying the source of these data by fine-tuning BERT and XLNet, visualized results using t-SNE, did "leave-one-hashtag-out" cross-validation, and evaluated models using some common metrics (Accuracy, F1 score, Confusion Matrix, Matthews Correlation Coefficient).

- 2019.02 - 2019.03 **Integrating BERT and Embeddings into CommonsenseQA Challenge**

We fine-tuned Google BERT to CommonsenseQA challenge 1.0 (with 3 options for each question) and then integrated Conceptnet Numberbatch and ELMo embeddings attempting to improve the model performance. The challenge involves a set of MCQ questions requiring human commonsense knowledge. We achieved 68.79% of accuracy on validation set using BERT + ELMo (soly BERT : 67.47%; BERT + Numberbatch: 67.68%).

## Technical Article

- "Accelerating Deep Learning on the JVM with Apache Spark and NVIDIA GPUs"**

Author: Haoxuan Wang, Qin Lan [AWS], Carol McDonald [Nvidia]; Link: [https://www.infoq.com/articles/deep-learning-apache-spark-nvidia-gpu/?itm\\_source=articles\\_about\\_ai-ml-data-eng&itm\\_medium=link&itm\\_campaign=ai-ml-data-eng](https://www.infoq.com/articles/deep-learning-apache-spark-nvidia-gpu/?itm_source=articles_about_ai-ml-data-eng&itm_medium=link&itm_campaign=ai-ml-data-eng)