billweasley20092@gmail.com

Github: https://github.com/billweasley

Linkedin: https://www.linkedin.com/in/horace-haoxuan-wang

Personal Website: http://shellcottage.me

Work Experience

- 2022.03 Now Machine Learning Engineer, ASR and Language Tech @ Zoom
 - Designed and implemented a multilingual ASR service supporting 36 languages and 5+ workflows. The architecture is based on AsyncMQ/Kafka, deployed via Kubernetes (K8s) and Istio, supporting dynamic CPU-based scaling to serve global Zoom customers.
 - Conducted ASR model performance optimization, fine-tuning inference parameters, memory usage, and CPU allocation for efficient large-scale deployment.
 - Developed LLM-based ASR error correction workflows, achieving a Rare Word WER reduction from 37.8% to 17.5% on medical datasets
 using closed-source LLMs and bias word lists.
 - Independently implemented and optimized Whisper inference, leveraging in-house VAD (Voice Activity Detection) and WhisperX, achieving higher throughput and lower WER than OpenAl's reference implementation.
 - Built Multi-Head Attention (MHA)-based time alignment on LAS/Seq2Seq models, providing accurate word-level timestamps for multilingual transcription. * [Filed a US patent for this] *
 - Independently experimented with LLM for ASR in multimodality setting, aiming to improve the consistency of ASR decoding results. The orthographic (fully-format) WER (word error rate), and the rare word WER achieved a better result compared with the production model.
 - Created LLM-based data augmentation pipelines using Mistral MoE 8x7B, generating diverse textual datasets for model training.
- - Initially a Java Software Engineer supporting Barclaycard, supporting backend cache for bouns.
 - 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-turn BERT
 models in an iterative way. Finish the inference on 6 million internal pair-wised 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 from 89% where we started. The
 model does inference offline on our Spark cluster in a distributed way. For 6 million pair-wised samples, the running time is under 1 hour (on a cluster
 with 80 CPUs).
 - Built time-series forecasting models using the Informer architecture, predicting aggregated user transaction volume and value. Constructed counterfactual analyses to assess financial losses during system downtimes.

Education

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

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

Personal Project

2024.06 - Fine-tuning and evaluation of medical record data on Large Language Models (LLMs)

Fine-tuned various LLMs (Llama3-instruct, Llama3 Chinese-chat, Qwen2) on Chinese medical records datasets, focusing on tasks such as department classification, record summarization, and discharge certification.

- Consultation Summary/Discharge Summary: BLEU improved from 0%-30% to 49%-55%, ROUGE-L from 20%-30% to 60%-64%.
- Department Classification: Accuracy improved from 0%-36% to 69%-71%.

Future plans include open-sourcing the dataset.

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

Early Stage Project

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

Built data crawlers for **Gab.com** and pre-processed datasets using **Regular Expressions**. Fine-tuned **BERT and XLNet** for classification tasks, achieving an **86% F1 score** on hashtag-balanced datasets. Visualized results using **t-SNE**, performed **cross-validation**, and evaluated metrics including **Accuracy**, **F1 score**, **and Matthews Correlation Coefficient**.

• 2019.02 - 2019.03 Integrated BERT and Embeddings in CommonsenseQA Challenge

Fine-tuned Google BERT for CommonsenseQA Challenge 1.0, integrating ConceptNet Numberbatch and ELMo embeddings. Achieved 68.79% accuracy on validation datasets (BERT only: 67.47%).