ZUOLIN CHENG

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SUMMARY

Highly skilled Ph.D. in Bioinformatics with 8+ years of experience in designing/implementing multi-omics data analysis pipelines & developing novel machine-learning/statistical models. Proficient in multiple programming languages, machine-learning/data-analysis libraries, and bioinformatics tools. Proven track record of successful model development & omics data analysis that has enabled significant biological discoveries in cross-functional, collaborative research.

CORE COMPETENCIES

• AI / Machine Learning:

- Model Development: Large Language Models (LLMs), Neural Networks, Supervised Learning (Random Forests, SVM), Unsupervised Learning, Statistical Modeling
- o Al Applications: Al for Biological Insight (Driving Transcriptional Factor Identification, etc.)

• Bioinformatics & Omics Analysis:

- Data Types: Omics Data Analysis (scRNA-seq, Spatial Transcriptomics, scATAC-seq, ChIP-seq, CUT&RUN, proteomics, metabolomics)
- Techniques: Proficient in Omics Technologies, End-to-End Omics Data Analysis Pipeline Development,
 Gene Set Enrichment Analysis, Regulatory Network Analysis, Cell Type Annotation

Programming & Tools:

- o Languages: Python, R, C++, MATLAB, Bash Scripting.
- o Tools & Environment: Git / GitHub, Linux Environment

WORK EXPERIENCE

Virginia Tech

Postdoctoral Research Associate

Cupertino, CA, USA

Feb. 2023 – Present

Lead Next Generation Sequencing (NGS) data analysis in collaborative biological research studies, effectively facilitating scientific discoveries. Develop state-of-the-art NGS data analysis models and methods, providing user friendly R packages to benefit a larger audience. Mentor & train junior graduate students in research practices.

Project 1: GO-BERT, Cell Type Annotation for scRNA-seq Data Using Large Language Model (LLM)

Data: scRNA-seq data, Gene Ontology (GO);

Language: Python, PyTorch

- o Innovatively **integrated domain-specific priors** (gene ontology) into neural network architecture during supervised fine tuning, serving as regularizations to guide model toward more plausible predictions
- o Significantly enhanced cell type annotation performance, increasing F1 score to 0.76 (previously 0.69)
- o Reduced 31% parameters in fine tuning without loss of performance compared to the state-of-the-art
- Propose a domain-insight-based pretraining strategy for large-scale single-cell data, overcoming limitations of existing models (e.g. scBERT, scGPT, Geneformer) in capturing awareness of global context

• Project 2: A Machine Learning Model Unifying Various NGS Analysis Tasks

o *Data*: various types of NGS data;

Language: C++, R, MATLAB

- Unified various NGS analysis tasks (e.g. single cell RNA-seq data normalization, cancer gene detection, GO term activity inference, etc.) by pinpointing the shared key mathematical problem behind them
- Developed a novel model of joint inference of multiple hidden factors by integrating sophisticated statistical/machine-learning techniques, enhancing confounding factor control in downstream analyses
- Proposed a highly efficient customized optimizer leveraging model structure & code optimization (C++),
 achieving scalability on large-scale real-world data (reduced running time: days to minutes)
- o Developed user-friendly R packages on different OS (Windows, Linux, iOS), and distributed via GitHub

Virginia Tech, Computational Bioinformatics and Imaging Lab Graduate Research Assistant

Arlington, VA, USA Aug. 2015 – Dec. 2022

Developed innovative AI/ML and statistical tools for NGS data analysis. Designed and implemented NGS analysis pipelines in cross-functional projects, collaborating with biologists and clinicians from leading institutions such as Stanford University, Cincinnati Children's Hospital Medical Center, and the Salk Institute.

Project 1: RNA-seq Analysis Pipeline to Uncover Microglial Heterogeneity During Development

o <u>Data</u>: bulk and single cell RNA-seq data;

Language: R

- o Designed and implemented **end-to-end analysis pipeline**, including sequencing alignment, quality control (QC), RNA-seq normalization, clustering, data visualization, DEG analysis, GSEA, etc.
- Customized some modules in the pipeline, achieving best solutions to project-specific requests
- o Enabled biological discoveries, collaborating with Stanford University: revealed a novel microglia subset

• Project 2: Machine Learning Model for Driving Transcription Factor (TF) Prediction

o Data: scRNA-seq, ChIP-seq Database, DNase-seq, Motif Database (JASPAR);

Language: R

- o Integrated ChIP-seq, motif databases, and epigenetic data to predict driving TF behind biological process
- Discovered a long-ignored key factor and integrated it into ML model, boosting the best record of disease driving TF detection based on large TF-gene binding database (38.8% to 49.0% on benchmark)
- o Validated predictions (Zfp36l1 TF) via wet-lab collaboration, published in Cell Stem Cell

• Project 3: Full-length RNA-seq Data Normalization

o Data: scRNA-seq;

Language: R, C, C++

- Identified a key confounding factor (effective cDNA-length during PRC) ignored by previous analyses in full-length RNA-seq raw counts, by combining empirical data mining (high-throughput intermedia data of sequencing alignment) and theoretical reasoning (deep understanding of sequencing technologies)
- Proposed a ML algorithm to learn and address the factor during RNA-seq data normalization, enhancing the accuracy of downstream differential expressed gene (DEG) detection (AUC: 0.886 to 0.938)

EDUCATION

Virginia TechArlington, VA, USAPhD, Electrical EngineeringAug. 2015 – Dec. 2022

• Research areas: Bioinformatics; Machine Learning

Peking University

Beijing, China

M.S., Electronic Science and Technology
• GPA: 3.87/4.0; **Top Ranked**: 1/30

PUBLICATIONS

TOTAL PUBLICATIONS: 13; TOTAL CITATIONS: 1600 +

https://scholar.google.com/citations?user=6pHncW4AAAAJ&hl=en&oi=ao

- Li Q*, Cheng Z* (co-first author), Zhou L, Darmanis S, ... Tony Wyss-Coray, Ben A Barres. Developmental heterogeneity of microglia and brain myeloid cells revealed by deep single-cell RNA sequencing. *Neuron*. 2019. (Best of Neuron 2018-2019).
- Barclay K, Nora A, **Cheng Z**, Kim M, Zhou L, Yang J, Rustenhoven J, et al. An inducible genetic tool to track and manipulate specific microglial states reveals their plasticity and roles in remyelination. *Immunity*. 2024.
- Cheng Z, Wei S, Wang Y, Wang Y, Lu R, Wang Y, Yu G. An Efficient and Principled Model to Jointly Learn the Agnostic and Multifactorial Effect in Large-Scale Biological Data. *bioRxiv* (2024): 2024-04.
- Cheng Z, Wei S, Yu G. A Single-Cell-Resolution Quantitative Metric of Similarity to a Target Cell Type for scRNA-seq Data. In 2022 IEEE International Conference on Bioinformatics and Biomedicine (BIBM) 2022.