JUNTANG ZHUANG

Email: j.zhuang@yale.edu ⋄ Website: https://juntang-zhuang.github.io ⋄ Phone: +1 475-224-8282 ⋄ [Google Scholar]

RESEARCH STATEMENT

My research aims to develop new machine learning techniques for biomedical applications. I approach this from 3 aspects: 1) develop better optimizers for neural networks, 2) develop accurate numerical methods for prior-informed continuous deep neural networks, 3) develop interpretable invertible networks and feature selection methods. My research lies in **Optimization**, **Deep Learning** (specifically **Continuous** and **Invertible** models) and **Medical Image Analysis**.

EDUCATION

Yale University
Ph.D. in Biomedical Engineering (Advisor: James S. Duncan)   Sep 2016 - May 2022 (Expected)

Yale University
M.A. in Statistics, M.Phil in Biomedical Engineering (GPA:4.0/4.0)   Sep 2017 - May 2018

Tsinghua University
B.E. in Engineering Physics   Sep 2012 - May 2016

PUBLICATIONS

1. **J. Zhuang**, Yifan Ding, et al, Momentum centering and asynchronous update for adaptive gradient methods *Conference on Neural Information Processing Systems* (NeurIPS 2021)[project page]

2. **J. Zhuang**, N. Dvornek, et al. MALI: a memory efficient and reverse accurate integrator for Neural ODEs, *International Conference on Learning Representations* (ICLR 2021)[project page]

3. **J. Zhuang**, N. Dvornek, et al. Multiple-shooting adjoint method for whole-brain dynamic causal modeling , *Information Processing in Medical Imaging* (IPMI 2021, oral presentation) [project page]

4. **J. Zhuang**, T. Tang, et al. AdaBelief Optimizer: adapting stepsizes by the belief in observed gradients, *Conference on Neural Information Processing Systems* (NeurIPS 2020, Spotlight) [project page]

5. **J. Zhuang**, N. C. Dvornek, et al. Adaptive Checkpoint Adjoint Method for Gradient Estimation in Neural ODE, *International Conference on Machine Learning* (ICML 2020) [project page]

6. **J. Zhuang**, J. Yang, et al., ShelfNet for fast semantic segmentation, *Workshop on Computer Vision for Road Scene Understanding and Autonomous Driving* (CVRSUAD 2019)

7. **J. Zhuang**, N. C. Dvornek, et al., Decision Explanation and Feature Importance for Invertible Networks, *Workshop on Interpretating and Explaining Visual Artificial Intelligence Models* (XAIC 2019) [oral]

8. **J. Zhuang**, N. C. Dvornek, et al., Invertible Network for Classification and Biomarker Selection for ASD , *International Conference on Medical Image Computing & Computer Assisted Intervention* (MICCAI 2019)

9. **J. Zhuang**, N. C. Dvornek, et al., Prediction of Pivotal response treatment outcome with task fMRI using random forest and variable selection, *International Symposium on Biomedical Imaging* (ISBI 2018)

10. **J. Zhuang**, N. C. Dvornek, et al., Prediction of severity and treatment outcome for ASD from fMRI, *International Workshop on PRedictive Intelligence In MEdicine* (PRIME 2018)

11. N. C. Dvornek, X. Li, **J. Zhuang**, et al., Demographic-Guided Attention in Recurrent Neural Networks for Modeling Neuropathophysiological Heterogeneity (MLMI 2020)

12. X. Li, N. C. Dvornek, M. Zhang, **J. Zhuang**, et al., Pooling Regularized Graph Neural Network for fMRI Biomarker Analysis, (MICCAI 2020)

13. **J. Zhuang** LadderNet: Multi-path networks based on U-Net for medical image segmentation, ArXiv 2018

14. **J. Zhuang**, et al, Towards a flat minima with an ascent step in the gradient descent update,, in submission, 2021

WORK EXPERIENCE

Student researcher at Google   Los Angeles USA, June-Oct 2021
Intern at Tencent AI   Bellevue, W.A. USA, June-Sep 2019

OPEN-SOURCE PROJECTS

AdaBelief-optimizer (*941* stars on github, added to official repositories such as tensorflow-addons, Google Flax and Deepmind optax);   ShelfNet (246 stars);   LadderNet (102 stars);   TorchDiffEqPack

INVITED TALKS

- Computational Neuroscience Laboratory at Stanford University   2020
- SyncedTech (name in Chinese pinyin: JiQi ZhiXin) [video]   2020

AWARDS & SCHOLARSHIPS

- Best paper award, Machine Learning in Medical Imaging (MLMI)   2019
- Top-1 winner for CNI Transfer Learning Challenge, MICCAI   2019
- Graduate fellowship, Yale University   2016
- Award for excellent learning performance, Tsinghua University   2015
- Meritorious award for Mathematical Contest in Modeling (top 10% teams worldwide)   2015
- National encouragement award (for excellent learning performance), Tsinghua University   2014
- Sparks Program (Undergraduate High-tech Club) membership, Tsinghua University   2014

SKILLS

C/C++, R, Python, MATLAB, PyTorch, Keras, Jax, Tensorflow

RESEARCH EXPERIENCE

Towards a flat minima with an ascent step in the gradient descent update,   Jun - Oct 2021
Mentor: Ting Liu, Boqing Gong et al.   Google Research

- Proposed a generic method to improve the generalization of neural networks. Provided extensive theoretical analysis on both the convergence and provably better generalization performance of the new method.

- Empirically validated that the proposed method consistently improves the test performance of neural networks. Specifically, for the ImageNet top-1 accuracy of Vision Transformers, the proposed method achieved as much as +11.3% improvement over AdamW, and +3.2% improvement over the state of the art, using the same architecture, data augmentation and training budget.

- Paper submitted to ICLR 2022.

Momentum centering and asynchronous update for adaptive gradient methods   Jan - May 2021
Mentor: James S. Duncan, Sekhar Tatikonda   Yale University

- Proposed ACprop, which is an adaptive optimizer combining momentum centering and asynchronous update. ACProp has a convergence rate of $O(\frac{1}{\sqrt{T}})$ for the stochastic non-convex case, which matches the oracle rate. Furthermore, ACProp has a weak condition for convergence.

- Validated ACProp in extensive empirical studies: ACProp outperforms both SGD and other adaptive optimizers in image classification with CNN, and outperforms well-tuned adaptive optimizers in the training of various GAN models, reinforcement learning and transformers.

- Paper accepted to NeurIPS 2021. [project page]

MALI: a memory efficient and reverse accurate integrator for Neural ODEs   Sep - Nov 2020
Mentor: James S. Duncan, Sekhar Tatikonda   Yale University

- Proposed MALI, which is a new solver for Neural ODEs. Every numerical step in MALI is accurately invertible, so MALI can achieve accuracy in reverse-time trajectory, hence accuracy in gradient estimation for Neural ODEs. Due to the numerical invertibility, intermediate states can be deleted and later on reconstructed, hence MALI achieves a constant memory cost w.r.t integration time.

- Due to the accuracy and memory efficiency, MALI is suitable for large-scale Neural ODEs, such as FFJORD, and large ODE-CNNs on ImageNet. MALI achieves new state-of-the-art for image generation with continuous models.

- Paper accepted by International Conference on Learning Representations (ICLR 2021) [project page]

Evolutionary causal modeling of brain states from task-fMRI data   Mar - Sep 2020
Mentor: James S. Duncan   Yale University

- Modeled the effective connectome of the brain, which is the directional influence between different regions of the brain. Developed a differential equation model to simulate the dynamical evolution of brain states.

- Developed a method to efficiently learn the dynamical model from high-dimensional fMRI data, based on previous machine-learning projects such as AdaBelief Optimizer, Adaptive Checkpoint Adjoint and Invertible Networks (see details below).

- Paper accepted by 27th International Conference on Information Processing in Medical Imaging (IPMI 2021).

AdaBelief optimizer: a fast, accurate and stable optimizer for deep learning   Jan - June 2020
Mentor: James S. Duncan, Sekhar Tatikonda   Yale University

- Developed an optimizer for deep learning models. To our knowledge, it's the first to achieve three goals simultaneously: fast training speed, good generalization performance, and stability of training. Performed extensive validation in computer vision tasks, language modeling and generative adversarial networks (GAN) and reinforcement learning.
- Paper is accepted as Spotlight Presentation by NeurIPS 2020. [project page]

Adaptive checkpoint adjoint method for gradient estimation in neural ODE   Jun - Dec 2019
Mentor: James S. Duncan, Sekhar Tatikonda   Yale University

- Proposed and implemented a family of adaptive ODE solvers for accurate gradient estimation, which is a core factor of most deep learning models. Achieved both accuracy and computation efficiency.
- To our knowledge, our method is the first to enable neural-ODE to achieve comparable results to state-of-the-art discrete-layer models on benchmark classification tasks.
- Our method enables the real-world application of deep learning models combined with physical and biological knowledge in the form of differential equations. For example, our method is applied to estimate the mass of stars in a three-body problem, and learn the dynamical evolution of brain networks in fMRI data.
- Paper accepted by International Conference on Machine Learning (ICML 2020). [project page]

Invertible networks for model decision interpretation   Jan - June 2019
Mentor: James S. Duncan, Nicha C. Dvornek   Yale University

- Proposed a two-stage model for classification tasks, an invertible transform from input domain to feature domain, and a linear classifier in the feature domain. With invertible networks, we explicitly determine the decision boundary in the input domain, and calculate the projection of a point onto the decision boundary. The difference between a point and its projection onto the decision boundary can be viewed as the explanation for model decision.
- Applied the proposed method on fMRI data, selected biomarkers for ASD, and validated biomarkers in prediction of phenotype scores.
- Papers accepted by ICCV Workshop on Interpreting and Explaining Visual Artificial Intelligence Models (ICCV 2019, XAIC) and International Conference on Medical Image Computing & Computer Assisted Intervention (MICCAI 2019).

ShelfNet for real-time semantic segmentation   Aug - Nov 2018
Mentor: Nicha C. Dvornek   Yale University

- Proposed ShelfNet, a multi-path network with a shelf-like structure for real-time semantic segmentation. Achieved both faster running speed and higher mIoU than state-of-the-art real-time semantic segmentation models such as BiSeNet.
- Paper accepted by ICCV Workshop on Computer Vision for Road Scene Understanding and Autonomous Driving (ICCV 2019, CVRSUAD).

Treatment outcome prediction and biomarker selection for Autism from fMRI   2017-2018

- Designed a two-level feature selection approach for the whole-brain image to deal with the high dimensionality: first select predicitve ROIs, then select predictive voxels within surviving ROIs. Developed interpretable and accurate predictive models.
- Tested on both task fMRI and resting-state fMRI datasets and achieved state-of-the art accuracy on different clinical scores.
- Selected biomarkers for autism and validated the results with Neurosynth decoder.
- Paper accepted in PRIME workshop, MICCAI 2018 and ISBI 2018.