

# JUNTANG ZHUANG

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## 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 Interpreting 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  
Intern at Tencent AI

Los Angeles USA, June-Oct 2021  
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

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

## AWARDS & SCHOLARSHIPS

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- 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

## PROFESSIONAL ACTIVITY

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Served as reviewer for MIDL 2019, ICML 2021, NeurIPS 2021, ICLR 2022, EMBC 2021 and MEDIA.

## TEACHING ACTIVITY

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Biomedical Image Processing and Analysis

## SKILLS

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

## RESEARCH EXPERIENCE

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**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**

Mentor: James S. Duncan

Mar - Sep 2020

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**

Mentor: James S. Duncan, Sekhar Tatikonda

Jan - June 2020

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**

Mentor: James S. Duncan, Sekhar Tatikonda

Jun - Dec 2019

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**

Mentor: James S. Duncan, Nicha C. Dvornek

Jan - June 2019

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**

Mentor: Nicha C. Dvornek

Aug - Nov 2018

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 predictive 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.