

DAOUDA SOW

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My research interests are mainly about **optimization methods for modern Machine Learning (ML), robust and safe ML**. On the theoretical side, I am interested in building new frameworks, especially through the lens of bilevel optimization, that capture modern ML applications such as Meta Learning, Adversarial Training, Hyperparameter Optimization, etc. On the practical side, I aim to design new algorithms with provable guarantees to tackle these problems as well as develop their practical and efficient implementations.

SKILLS

<ul style="list-style-type: none">• Large-scale Optimization Algorithms and Theory for training ML models• Implementation of Deep Learning methods and Optimization algorithms• Deep Learning for Computer Vision• Classical Computer Vision and Image Processing	<ul style="list-style-type: none">• Classical Machine Learning• Programming languages I use regularly: Python, MATLAB, C++.• Deep Learning frameworks: Pytorch, TensorFlow, Keras, Scikit Learn• Languages: English, French, Pulaar, and Wolof
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EDUCATION

<ul style="list-style-type: none">• The Ohio State University, Columbus, OHIO, USA Ph.D. student in Department of Electrical and Computer Engineering – since 2019• Beijing University Of Aeronautics And Astronautics, Beijing, CHINA M.Sc. in Pattern Recognition and Intelligent Systems – 2019• Ecole Polytechnic de Thies, Thies, SENEGAL Undergraduate studies in Electromechanical Engineering – 2017

SELECTED RESEARCH PROJECTS

<ul style="list-style-type: none">• Robust Learning:<ul style="list-style-type: none">○ Currently developing methods for online adversarial training that are robust to distributional shifts and/or imbalanced datasets. We formulate the problem into a novel compositional bilevel optimization problem and propose a principled way of optimizing the so-called instance weights associated with individual data losses. Our formulation yields +10% on average accuracy compared to the original framework by Madry et al.

In particular, our framework mitigates the issue of non-uniform per-class performances in adversarial training.

- We are **targeting NeurIPS 2023** for this project and will make our code public soon.
- **Online Meta Learning Under Distribution Shifts:**
 - Propose a novel online meta-learning algorithm in non-stationary environments without knowing the task boundaries.
 - Built 3 dynamic online few-shot benchmarks from a mixture of several vision datasets (MNIST, FashionMNIST, Omniglot, Tiered-ImageNet, Symbols) to simulate the realistic dynamic setting. Extensive experiments demonstrate the advantage of our method over existing baselines in all benchmarks.
 - Provide a regret analysis of the proposed algorithm. Our analysis captures a trade-off between the impact of task similarity on the performance of standard online meta-learning with known task boundaries and the performance under task boundary detection uncertainty.
 - Paper is **submitted for publication** and all codes will be publicly available soon.
- **Hessian-free Algorithms for Efficient Bilevel Optimization:**
 - Propose a simple but effective Hessian-free method which uses a zeroth-order-like approach to approximate the response Jacobian in bilevel optimization based on the difference between two gradient-based optimization paths. On the Few-shot Meta-learning problem over a ResNet-12 network, our approach outperforms the state-of-the-art optimization-based meta learners. Experiments on Hyperparameter optimization further demonstrate the competitiveness of our approach. Paper is **accepted by NeurIPS 2022**.
 - Propose a simple and easy-to-implement primal-dual bilevel optimization algorithm in the practical setting where the inner problem admits multiple minimizers. Paper is submitted for publication.
 - Provide the convergence rate analysis of all proposed algorithms.

PUBLICATIONS

- **Daouda Sow**, Sen Lin, Zhangyang Wang, Yingbin Liang. “Doubly Robust Instance-Reweighted Adversarial Training”, submitted for publication, 2023. [Arxiv](#)
- **Daouda Sow**, Kaiyi Ji, Yingbin Liang. “On the convergence theory for Hessian-free bilevel algorithms”, Advances in Neural Information Processing Systems (NeurIPS), 2022. [Arxiv](#)
- **Daouda Sow**, Sen Lin, Yingbin Liang, Junshan Zhang. “Algorithm Design for Online Meta-Learning with Task Boundary Detection”, submitted for publication, 2023. [Arxiv](#)
- Sen Lin, **Daouda Sow**, Yingbin Liang, Ness Shroff. “Online Bilevel Optimization: A Single-Loop Method with Window Averaging”, submitted for publication, 2023.
- **Daouda Sow**, Kaiyi Ji, Ziwei Guan, Yingbin Liang. “A primal-dual approach to bilevel optimization with multiple inner minima”, submitted for publication, 2022. [Arxiv](#)

- **Daouda Sow**, Zengchang Qin, Mouhamed Niasse, Tao Wan. “A Sequential Guiding Network With Attention for Image Captioning”, IEEE ICASSP 2019. [Arxiv](#)