





Prof. Barbara Caputo

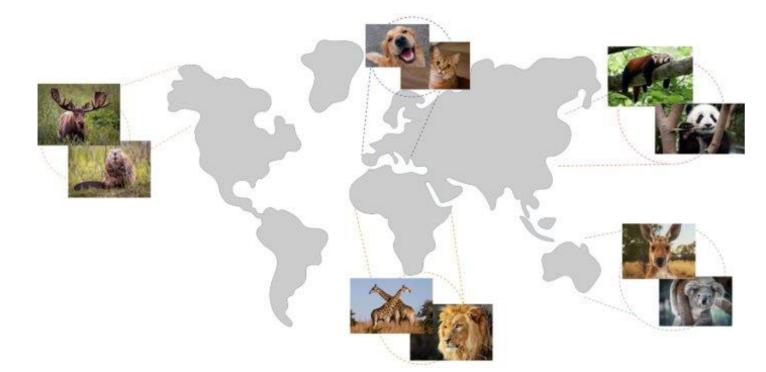
Federated Learning across Domains

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1. Introduction

Federated Learning (FL)¹ aims at learning model in a **distributed** way preserving the client's privacy. In realistic 5 scenarios, the clients' data is non-i.i.d. w.r.t. the global distribution and unbalanced.



2. Goal

The heterogeneous distribution of the clients' data leads to degraded convergence performance and unstable learning, posing a major challenge in realistic scenarios.

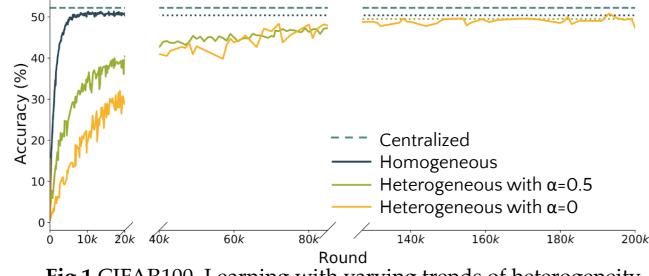
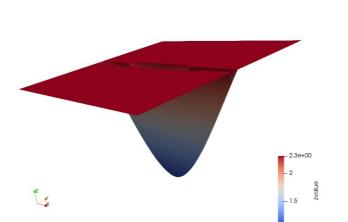


Fig.1 CIFAR100. Learning with varying trends of heterogeneity.

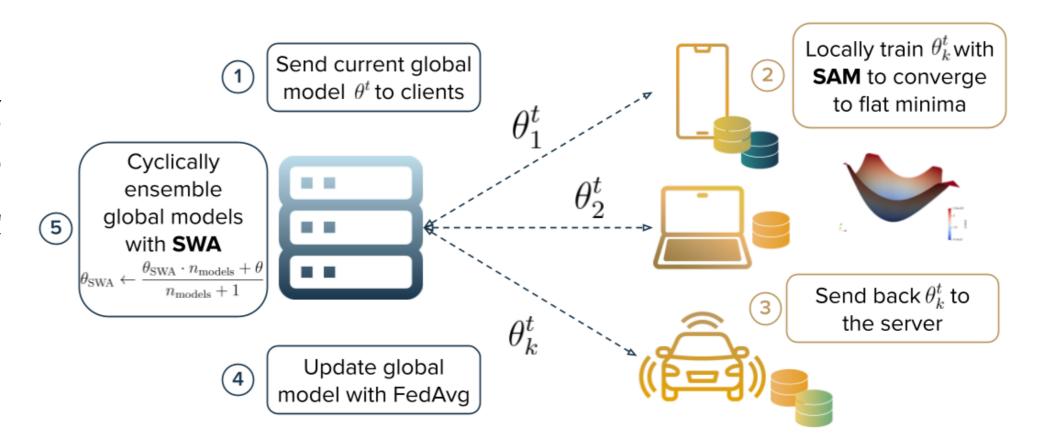
This thesis aims at explaining the poor **generalization** of the model through the lens the loss landscape and developing algorithms for faster and more stable learning. We extend our research to realistic applications, such as autonomous driving.

3. Improving generalization

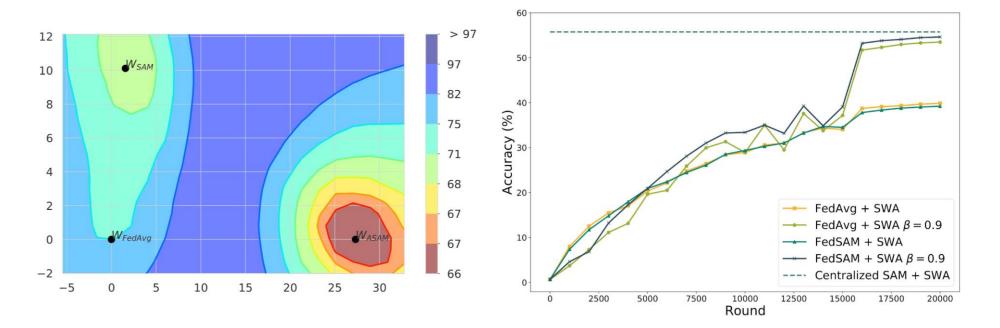
Federated models converge to sharp minima, linked with poor generalization.



Can we encourage convergence towards flat minima to improve generalization?

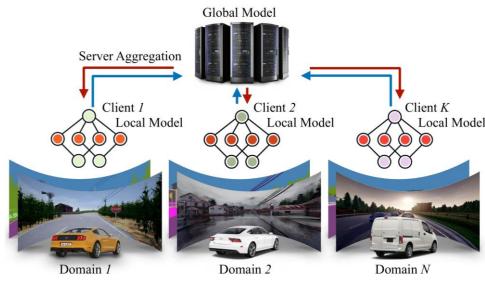


Results on heterogeneous CIFAR100 (1 class per client):



4. FL for autonomous driving

Task: Semantic Segmentation.



heterogeneity Challenges: class and domain shift across clients.

Main contributions:

- First benchmark for federated SS in autonomous driving⁵
- Self-supervised learning over clients' unlabelled data (more realistic)

5. References

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- Foret, P., et al. Sharpness-aware minimization for efficiently improving generalization. ICLR (2021)
- Izmailov, P., et al. Averaging weights leads to wider optima and better generalization. UAI (2018)
- Caldarola, Debora, Barbara Caputo, and Marco Ciccone. "Improving generalization in federated learning by seeking flat minima."
- Shenaj, Donald, et al. "Learning Across Domains and Devices: Style-Driven Source-Free Domain Adaptation in Clustered Federated Learning." WACV 2023