

Studying local and global loss landscapes in Federated Learning

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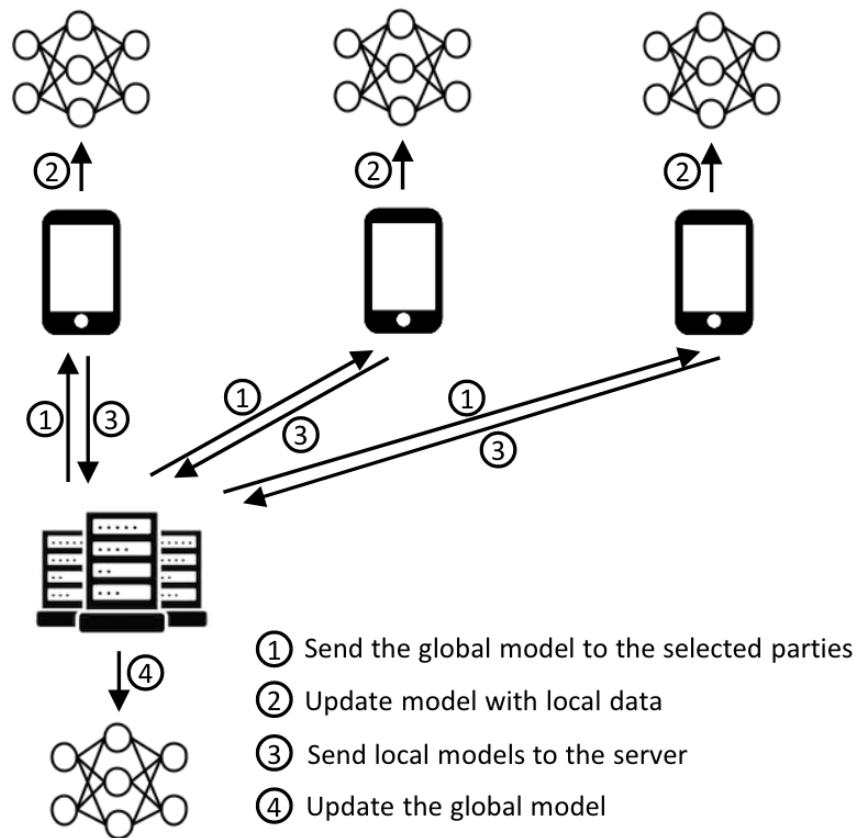
- **Introduction**
- **Experiments**
- **Future Work**
- **Conclusion**





Introduction

Federated learning (FL) is a decentralized approach to machine learning. It tackles the issues of centralized machine learning by allowing models to train on distributed data sources.



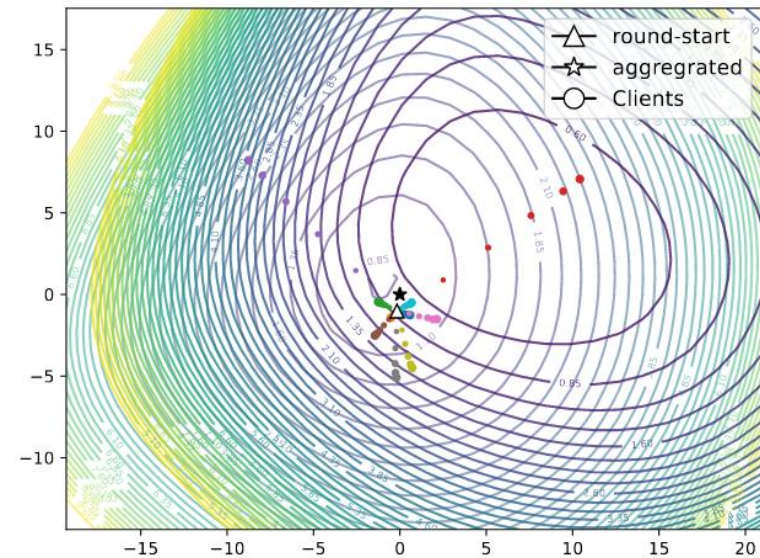
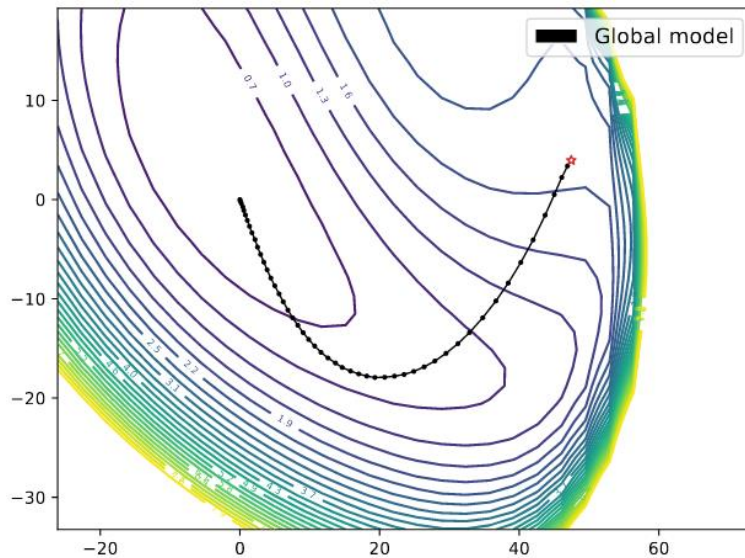
Underlying Architecture

- Central Server
- Parties (Clients)
- Communication Framework
- Aggregation Algorithm
 - Parameter-based
 - Output-based



Introduction

Loss landscape analysis is an effective approach for studying the learning dynamics and generalization properties of neural networks in high-dimensional spaces.





Introduction

Our Tasks:

- 1 Study local and global loss landscapes in federated learning.
- 2 Compare local loss landscapes and global loss landscapes in federated learning.
- 3 Understand a local client's contribution to the overall loss landscape.
- 4 Provide insights to optimize federated learning systems.

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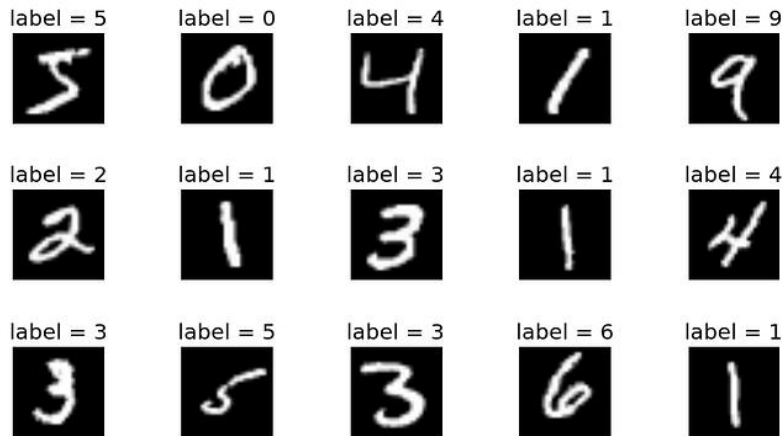
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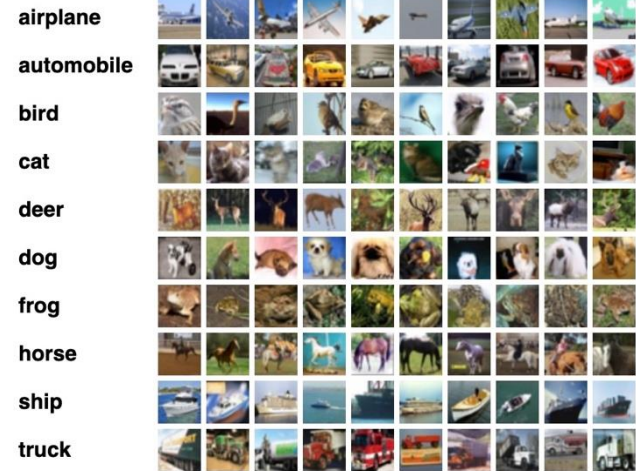


Experiments

Datasets



The **MNIST** dataset contains **70,000** images, each with dimensions of **28x28** pixels, resulting in **784** features per image. Each feature represents the intensity of a pixel, ranging from **0** to **255**.



The **CIFAR-100** dataset is a subset of the Tiny Images dataset, containing **60,000 32x32** color images across 100 classes grouped into 20 superclasses. Each class has 600 images, with 500 for training and 100 for testing. Images have both a "fine" label (class) and a "coarse" label (superclass).



Experiment FEDn Architecture





Experiments

Experiment Settings

Clients	Samples	Distribution	Experiments
2	Balanced	IID	.
3			.
4	Imbalanced	Non-IID	.
5			.

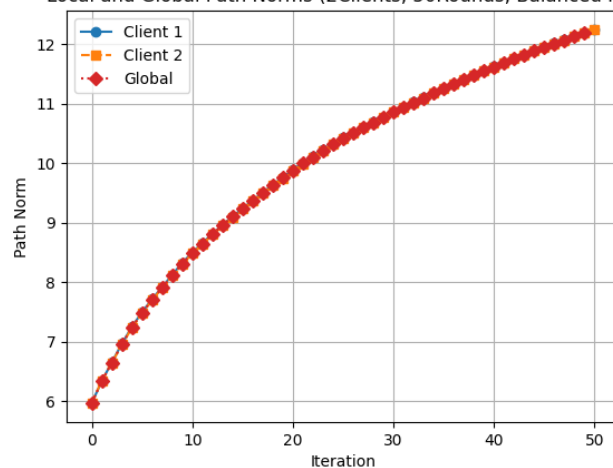
Clients + Samples + Distribution = Experiments



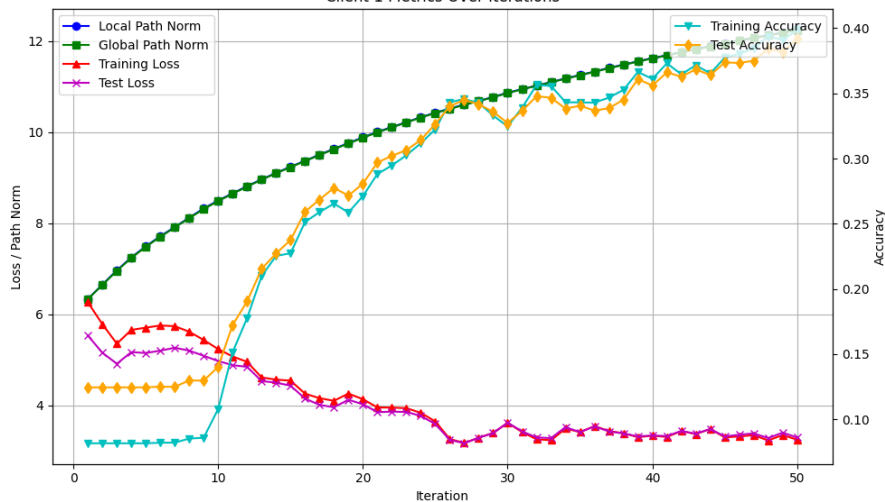
Experiments

2 Clients + Balanced + IID

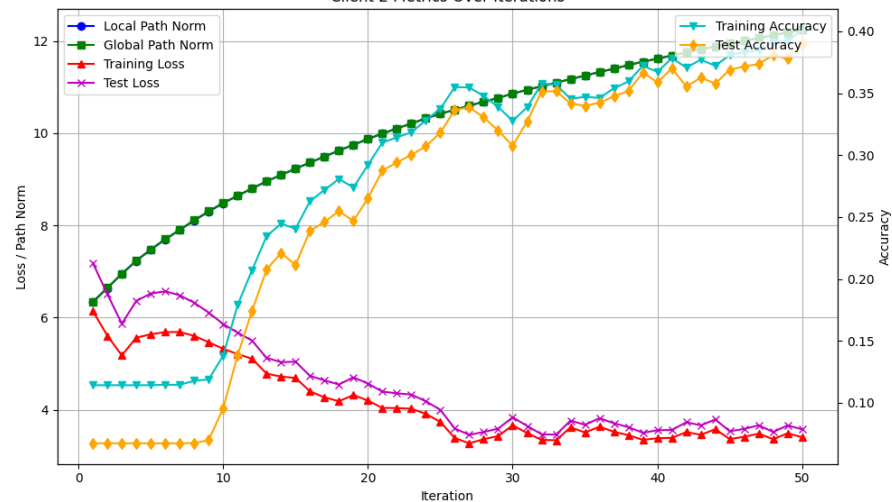
Local and Global Path Norms (2Clients, 50Rounds, Balanced IID)



Client 1 Metrics Over Iterations



Client 2 Metrics Over Iterations

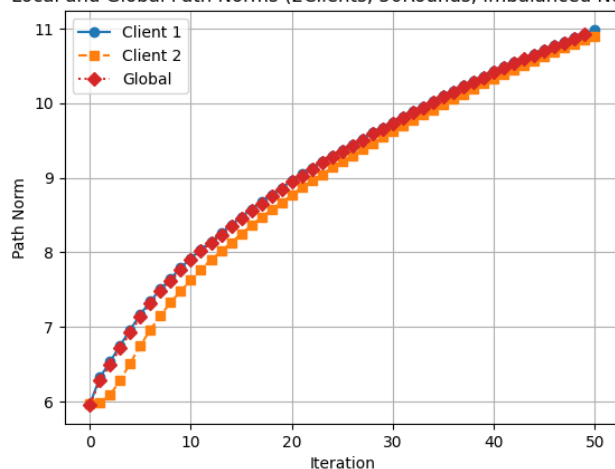




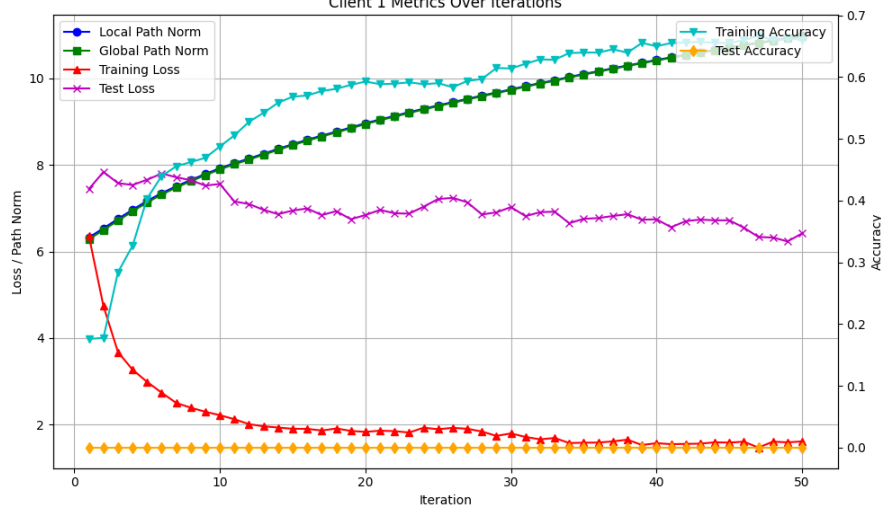
Experiments

2 Clients + Imbalanced + Non-IID

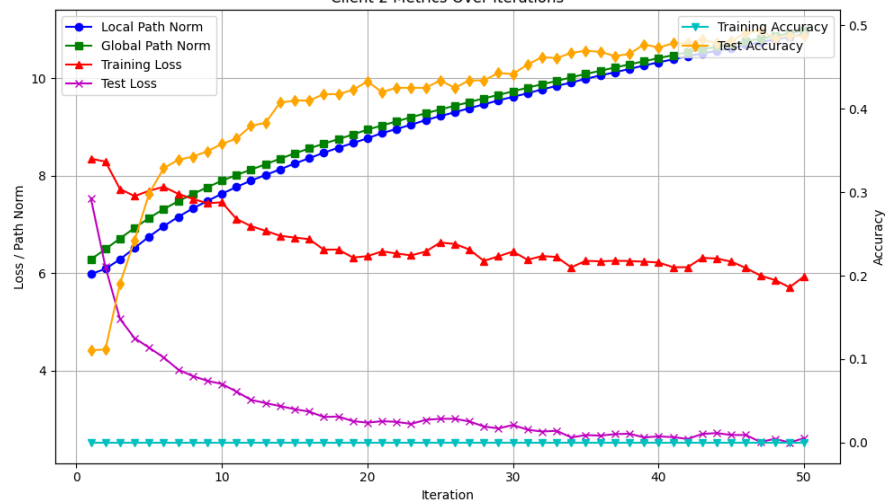
Local and Global Path Norms (2Clients, 50Rounds, Imbalanced Non-IID)



Client 1 Metrics Over Iterations



Client 2 Metrics Over Iterations

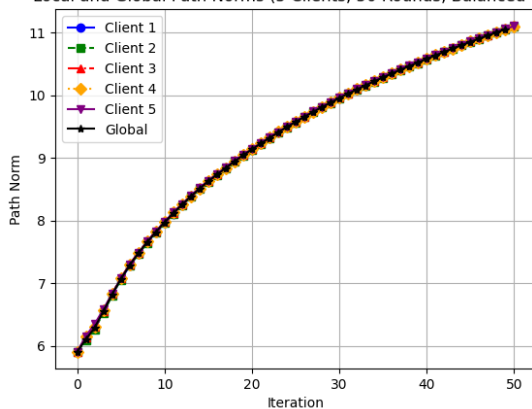




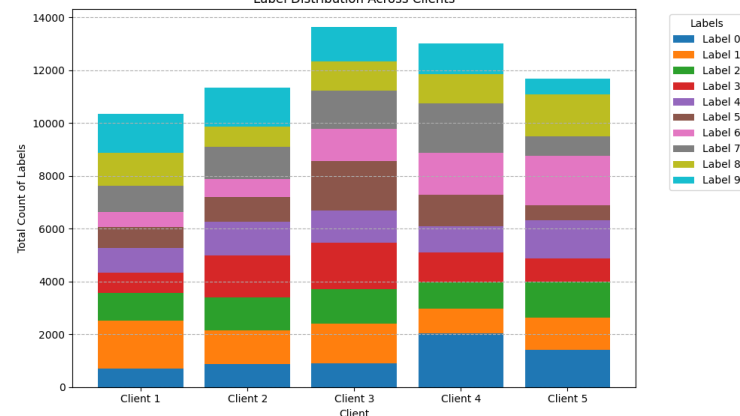
Experiments

5 Clients + Balanced + IID

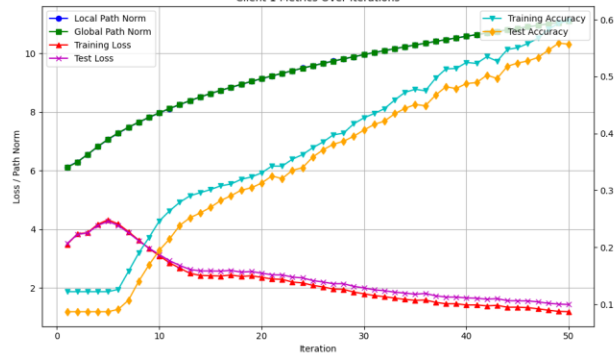
Local and Global Path Norms (5 Clients, 50 Rounds, Balanced IID)



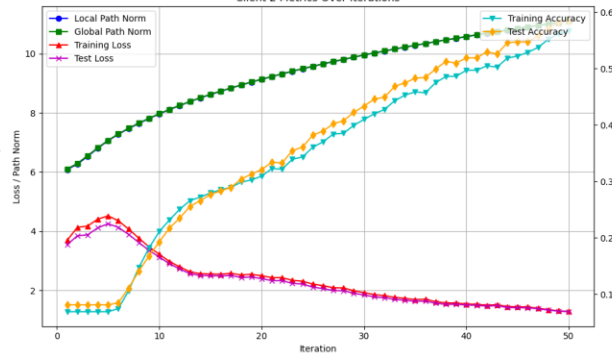
Label Distribution Across Clients



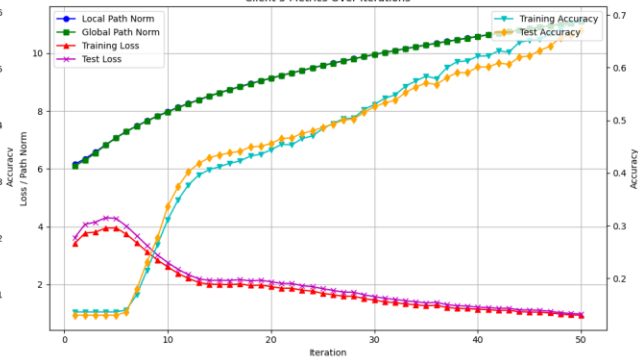
Client 1 Metrics Over Iterations



Client 2 Metrics Over Iterations



Client 5 Metrics Over Iterations

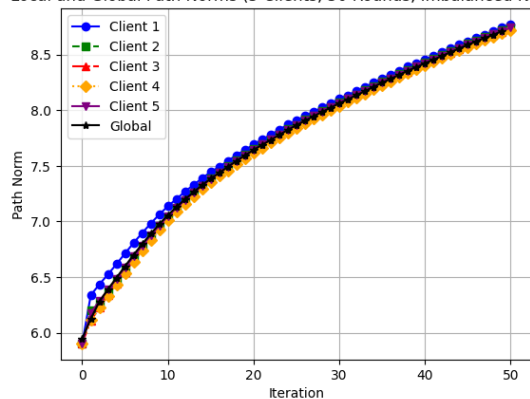




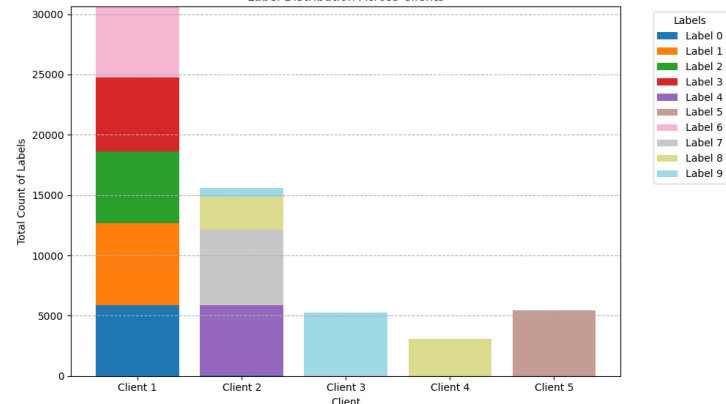
Experiments

5 Clients + Imbalanced + Non-IID

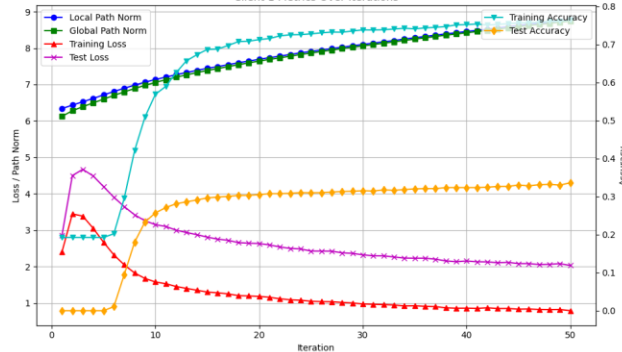
Local and Global Path Norms (5 Clients, 50 Rounds, Imbalanced Non-IID)



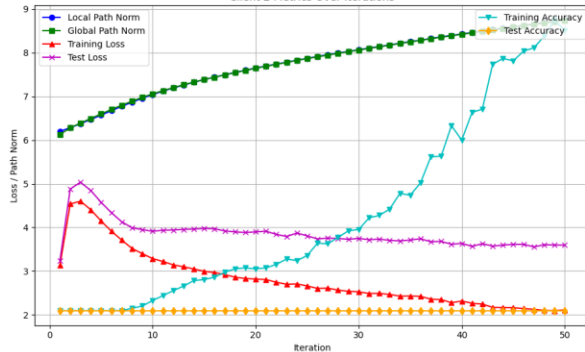
Label Distribution Across Clients



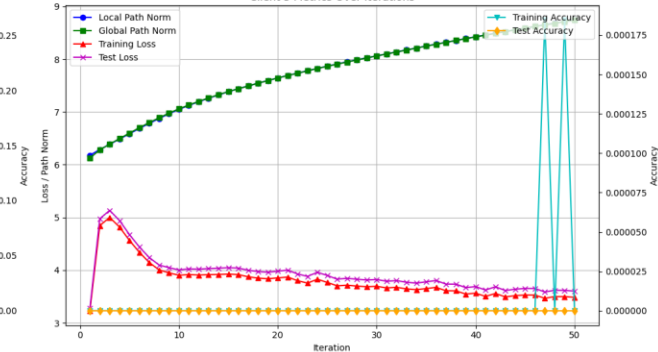
Client 1 Metrics Over Iterations



Client 2 Metrics Over Iterations



Client 5 Metrics Over Iterations





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Experiments

Experiment Results Analysis

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Future Wok

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Conclusion

Thanks



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