

APPFLX: PROVIDING PRIVACY- PRESERVING CROSS-SILO FEDERATED LEARNING AS A SERVICE



APPFL



ZILINGHAN LI

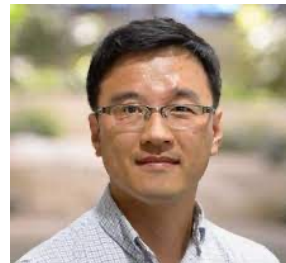
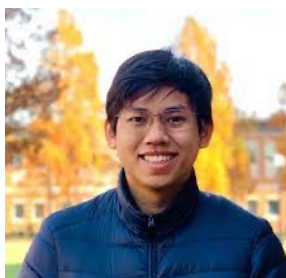
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TEAM



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MOTIVATION FOR FEDERATED LEARNING AS A SERVICE

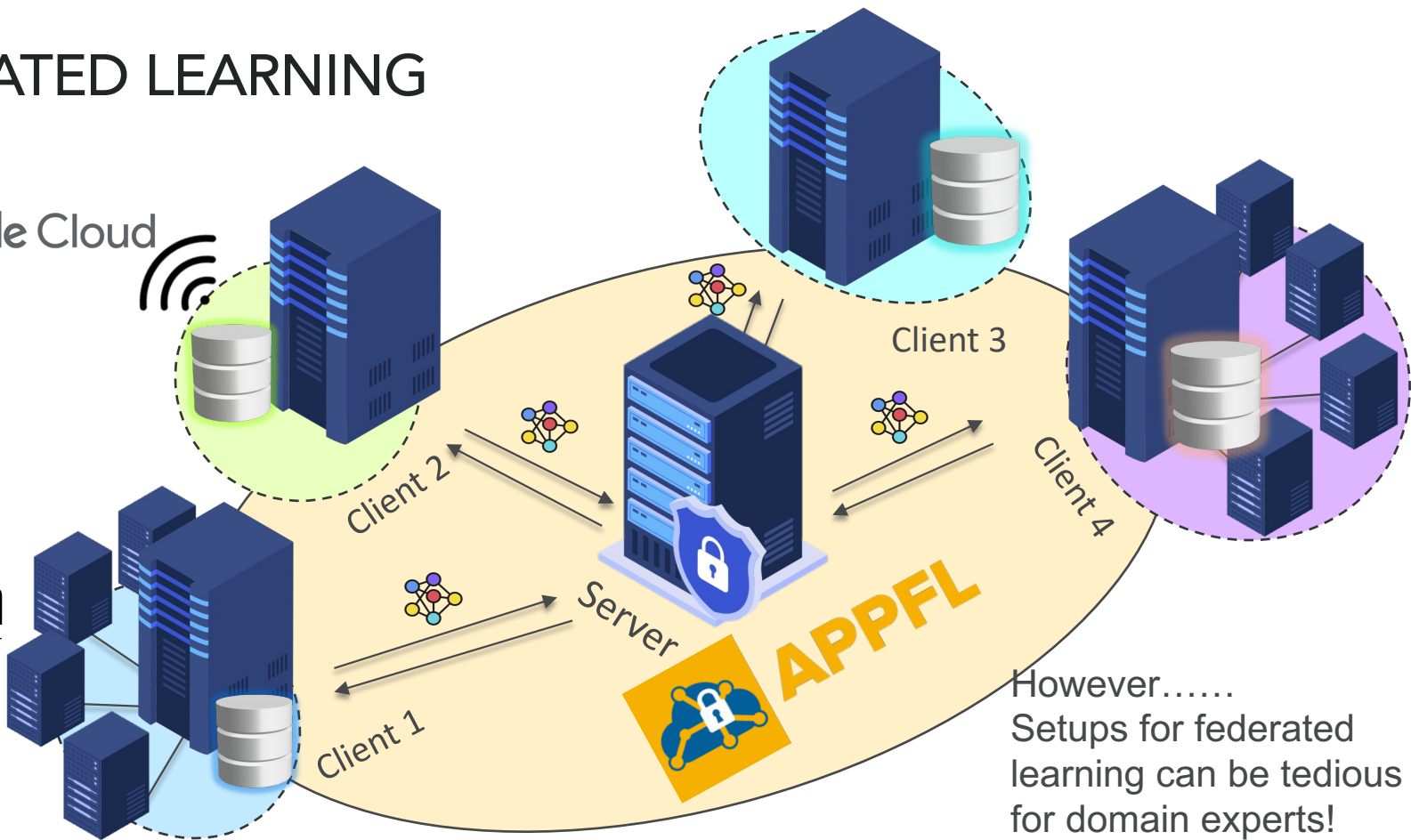


Data Shift in
Machine Learning



Privacy Concerns in
Biomedical Data

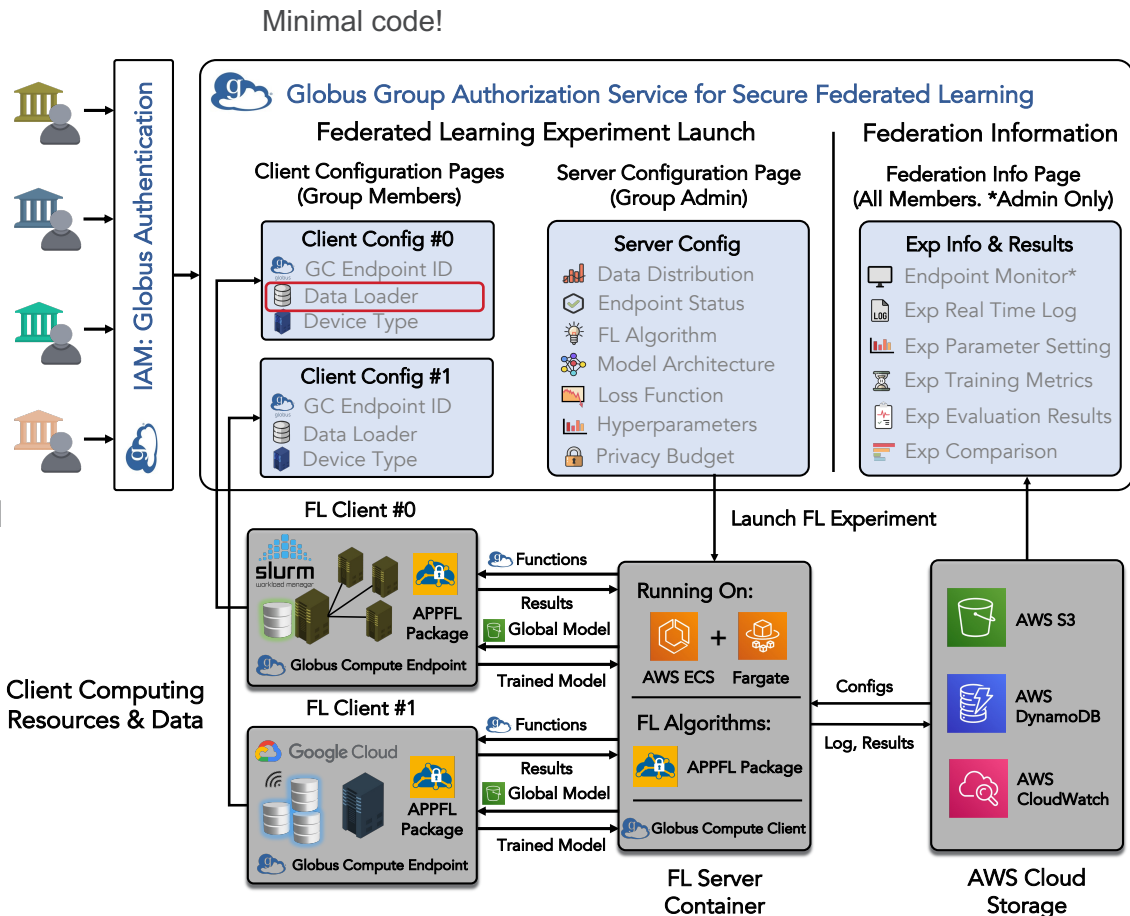
FEDERATED LEARNING



However.....
Setups for federated learning can be tedious for domain experts!

APPFLX WORKFLOW

- Login via Globus using institutional credentials
- Create a federation (FL group)
- Invite collaborators using institutional credentials
- Collaborators setup the globus compute endpoint
- Collaborators provide endpoint id and load data loader
- Configure and launch different FL experiments
- Monitor training in real-time, and obtain comprehensive reports
- Reason using data distribution visualization



GO BEYOND AN FL FRAMEWORK: WHY "AS-A-SERVICE"?

Comparison between a PPFL framework and APPFLx

Framework

- **Target users:** Developers for developing and simulating FL algorithms.
- **Authentication:** No client auth for most frameworks.
- **Launch Server:** Requires expertise to start federated learning experiments.
- **Results:** Server needs to manually share the whole results, which may require further post-process.
- **Connection:** Developed algorithms via the framework can be easily adopted to the service.

Service (APPFLx)

- **Target users:** Domain experts for applying FL.
- **Authentication:** Clients use institutional credentials via Globus Auth to setup a trust relationship
- **Launch Server:** Admin uses web UI to easily launch the FL experiment with different hyperparameters.
- **Results:** Comprehensive logs, reports, and visualizations shared among all clients on web UI.
- **Connection:** The service is built on the top of the APPFL framework
- **Misc:** Integrated with HuggingFace, GitHub for pre-trained models and pre-processing.

APPFLX CAPABILITIES

Creating Secure Federations

Dashboard

Federations

| Federation Name | | |
|--------------------------------|--------------|--|
| | | + Create Secure Federation |
| ANL_NCSA_LLNL | Group Manage | Create New Experiment |
| Shilan Test1 | Group Manage | Create New Experiment |
| B2AI/PALISADE-X/MGH | Group Manage | Create New Experiment |
| B2AI/PALISADE-X/MGH_FLAAAS_AWS | Group Manage | Create New Experiment |
| APPFLX-Demo | Group Manage | Create New Experiment |

Sites

| Site Name | | |
|---------------------|-------------------|-----------|
| ANL_NCSA_LLNL | Group Information | Configure |
| Shilan Test1 | Group Information | Configure |
| B2AI/PALISADE-X/MGH | Group Information | Configure |

<https://appflx.link/>

Federation Configuration

| Client Endpoints | Status | Email |
|---------------------|--------|-------|
| Jan F Nygård | | |
| Severin Langberg | | |
| Zilinghan Li | | |
| Zilinghan Li - NCSA | | |
| Ravi Madduri | | |
| Marcus Klarqvist | | |
| Jordan Fuhrman | | |

Federation Algorithm

Experiment Name

Server Training Epochs

Client Training Epochs

Server Validation Set for Benchmarking

None MNIST

Privacy Budget (ϵ)

Clip Value

Clip Norm

APPFLX CAPABILITIES

Comprehensive Experiment Reports

Federation Report

Print as PDF

Group Name: APPFLX-Demo
 Experiment Name: MNIST-FedAvgM-5Clients

Training Hyperparameters

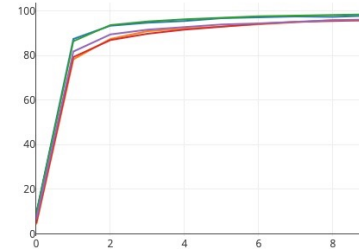
| hyperparameter | explanation | value |
|------------------------|---|----------------------------|
| Federation Algorithm | Server algorithm for the federated learning | Federated Average Momentum |
| Global training epochs | Number of global training epochs for the federation server | 10 |
| Local training epochs | Number of local training epochs for each federation site/endpoint | 2 |
| Privacy budget | Privacy budget used for privacy preserving | False |
| Clip value | Clip value for privacy preserving (TBF) | False |
| Clip norm | Clip norm for privacy preserving (TBF) | 0.0 |
| ▶ Model type | Type of trained model | CNN |
| Server momentum | Momentum of the federation server | 0.9 |
| Optimizer | SGD: Stochastic Gradient Descent Adam: Adaptive moment estimation | SGD |
| Learning rate | Client learning rate | 0.01 |
| Learning rate decay | Client learning rate decay | 0.975 |
| Client weights | How to assign weights for different clients in client model aggregation | sample_size |

Sites Validation

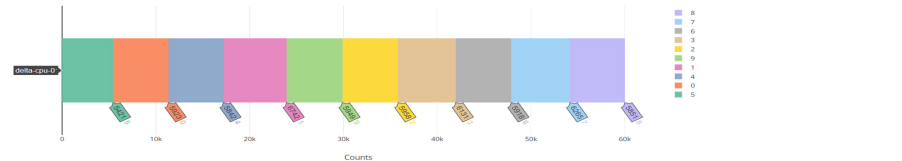
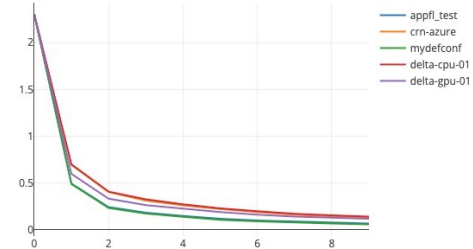
▶ [Click here to expand explanations:](#)

MNIST-FedAvg-5Clients

Accuracy vs. Step



Loss vs. Step



RESOURCES

- Privacy Preserving Federated Learning as a Service APPFLx - <https://appflx.link/> and instructions <https://ppflaas.readthedocs.io/en/latest/>
- GitHub for the APPFL framework: <https://github.com/APPFL/APPFL/>
- Globus Compute Communicator:
https://github.com/APPFL/APPFL/tree/main/src/appfl/comm/globus_compute
- APPFLx paper: <https://arxiv.org/pdf/2308.08786.pdf>
- FedCompass preprint: <https://arxiv.org/pdf/2309.14675.pdf>

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THANK YOU!

Q&A



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