

Benchmarking **federated learning** approaches against **siloed** and **mega-analysis** regimes

Michelle Wang

PhD candidate, NeuroDataScience-ORIGAMI Lab (JB Poline)

2025 June 23

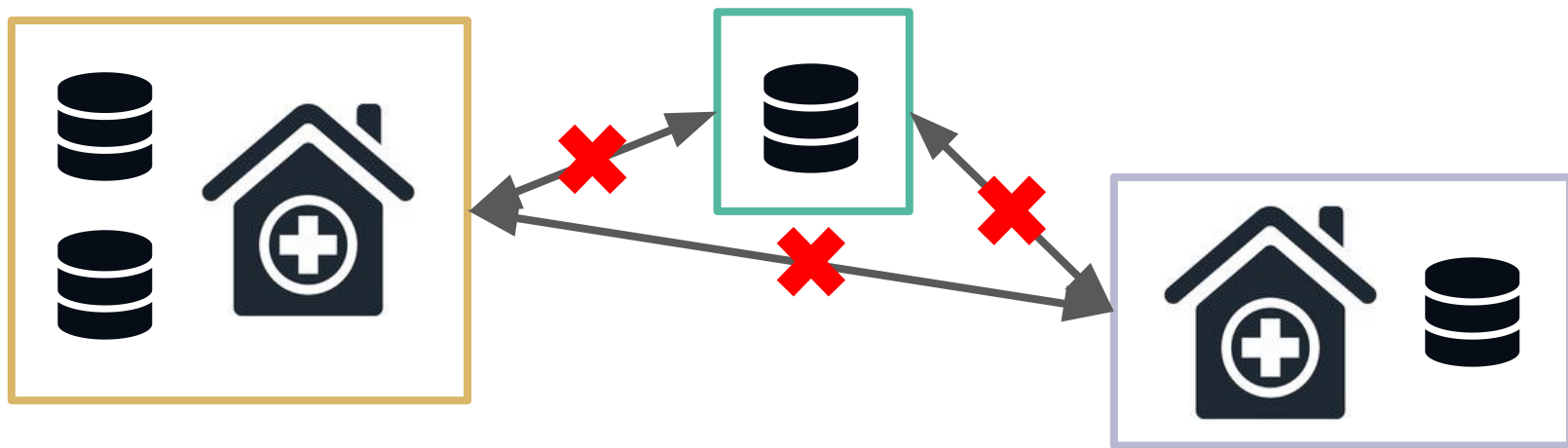


Data-sharing in studies involving clinical populations

Not all data can be openly shared

- Due to strong data privacy frameworks (e.g., the European Union's General Data Protection Regulation [GDPR]), etc.

Reality: Despite promotion of open science and data-sharing practices, **many datasets still live in so-called “silos”**



Decentralized methods for multi-centric analyses

Data **processing**

- Share software pipelines (including versions and runtime parameters)



Nipoppy framework for organization and decentralized processing of neuroimaging-clinical datasets

Data **discovery**

- Share information about availability of different datatypes/measures



Neurobagel ecosystem for decentralized metadata harmonization and search

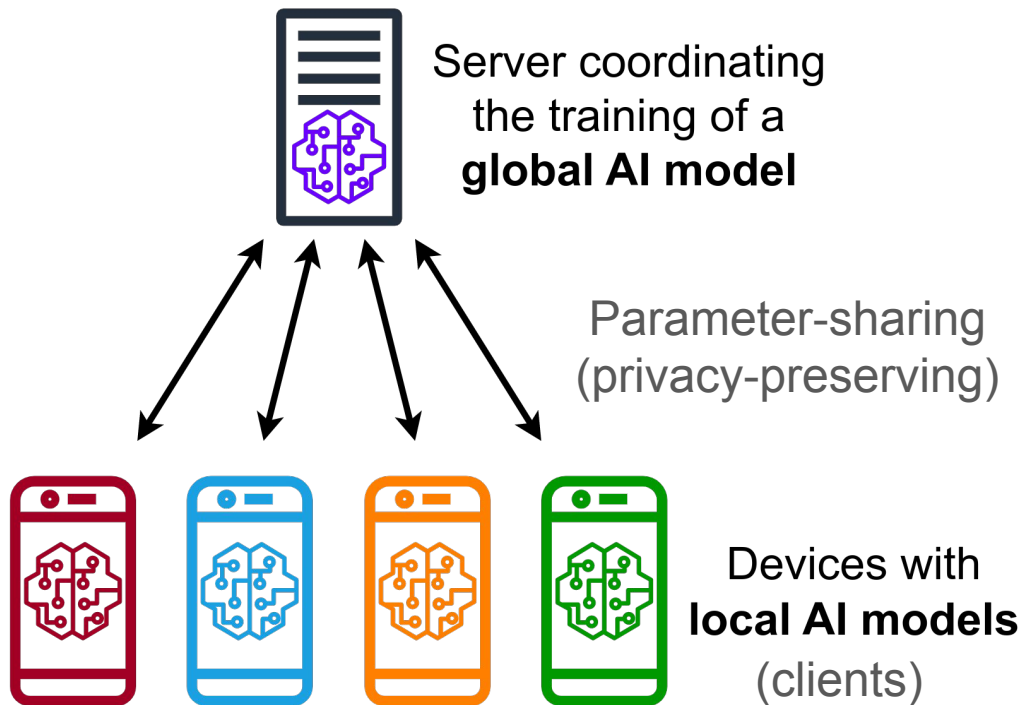
Data **analysis**

- Share fitted model parameters

Federated learning (McMahan *et al.*, 2016)

Federated learning

- Decentralized data
- Sharing of fitted model parameters to create a **global model**



We compared a simple **federated learning** setup with more traditional setups in clinical neuroimaging analyses

Data

Parkinson's/Alzheimer's disease (PD/AD) datasets:

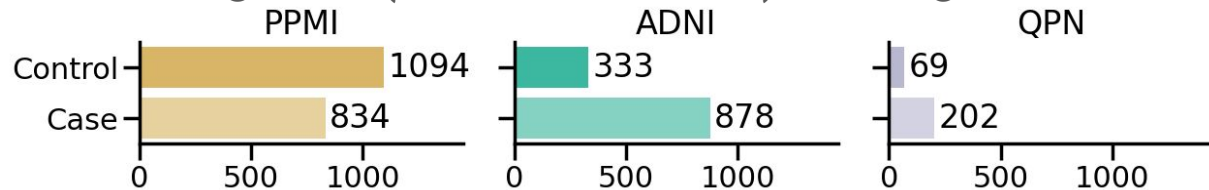
- Parkinson's Progression Markers Initiative (**PPMI**) (Marek et al., 2018)
- Alzheimer's Disease Neuroimaging Initiative (**ADNI**) (Jack et al., 2008)
- Quebec Parkinson Network (**QPN**) (Gan-Or et al., 2020)

Measures

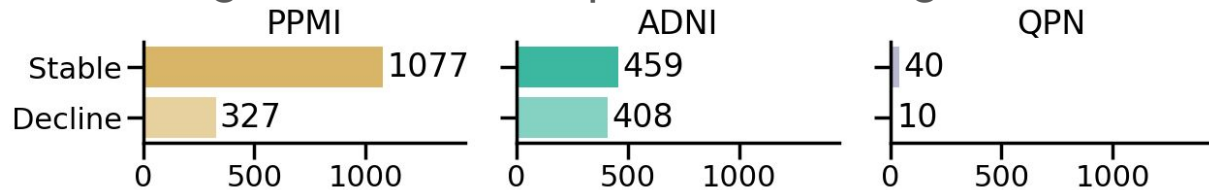
- Age
- Sex
- FreeSurfer cortical thicknesses (**CT**) and subcortical volumes (**SV**)
- Cognitive assessment score
 - PPMI and QPN: Montreal Cognitive Assessment (MoCA)
 - ADNI: Mini-Mental State Examination (MMSE)

Prediction tasks

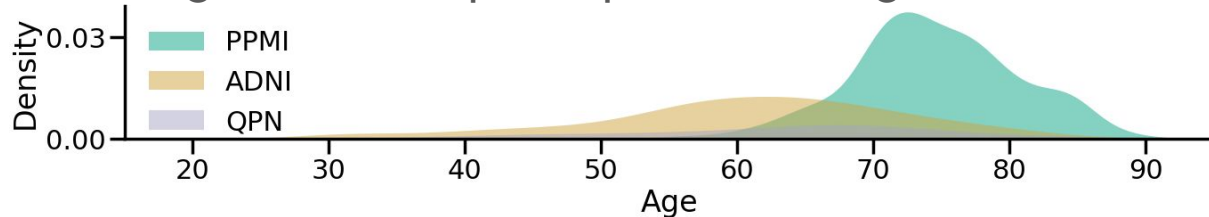
Predict **diagnosis** (PD/AD vs control) from **age + sex + CT + SV**



Predict **cognitive decline** in patients from **age + sex + CT**



Predict **age** of control participants from **age + sex + SV**



Model training setups

Setup	Train data	Model
Siloed (PPMI)	PPMI →	$M_{\text{siloed, ppmi}}$
Siloed (ADNI)	ADNI →	$M_{\text{siloed, adni}}$
Siloed (QPN)	QPN →	$M_{\text{siloed, qpn}}$
Federated	PPMI →	M_{ppmi}
	ADNI →	M_{adni}
	QPN →	M_{qpn}
		$\left. \begin{array}{l} M_{\text{ppmi}} \\ M_{\text{adni}} \\ M_{\text{qpn}} \end{array} \right\} \rightarrow M_{\text{federated}}$ (weighted avg. of params.)
Mega-analysis	PPMI	$\left. \begin{array}{l} \text{PPMI} \\ \text{ADNI} \\ \text{QPN} \end{array} \right\} \rightarrow M_{\text{mega-analysis}}$
	ADNI	
	QPN	

Base model pipeline:

- Z-score
- Logistic/linear regression

10-fold cross-validation

Test datasets:

- PPMI
- ADNI
- QPN
- PPMI+ADNI+QPN

Model training setups

Setup	Train data	Model
Siloed (PPMI)	PPMI →	$M_{\text{siloed, ppmi}}$
Siloed (ADNI)	ADNI →	$M_{\text{siloed, adni}}$
Siloed (QPN)		
Federated		
	QPN →	M_{qpni}
Mega-analysis	<div> <div>PPMI</div> <div>ADNI</div> <div>QPN</div> </div> →	$M_{\text{mega-analysis}}$

Hypothesis
 Siloed < Federated ≤ Mega-analysis

Base model pipeline:

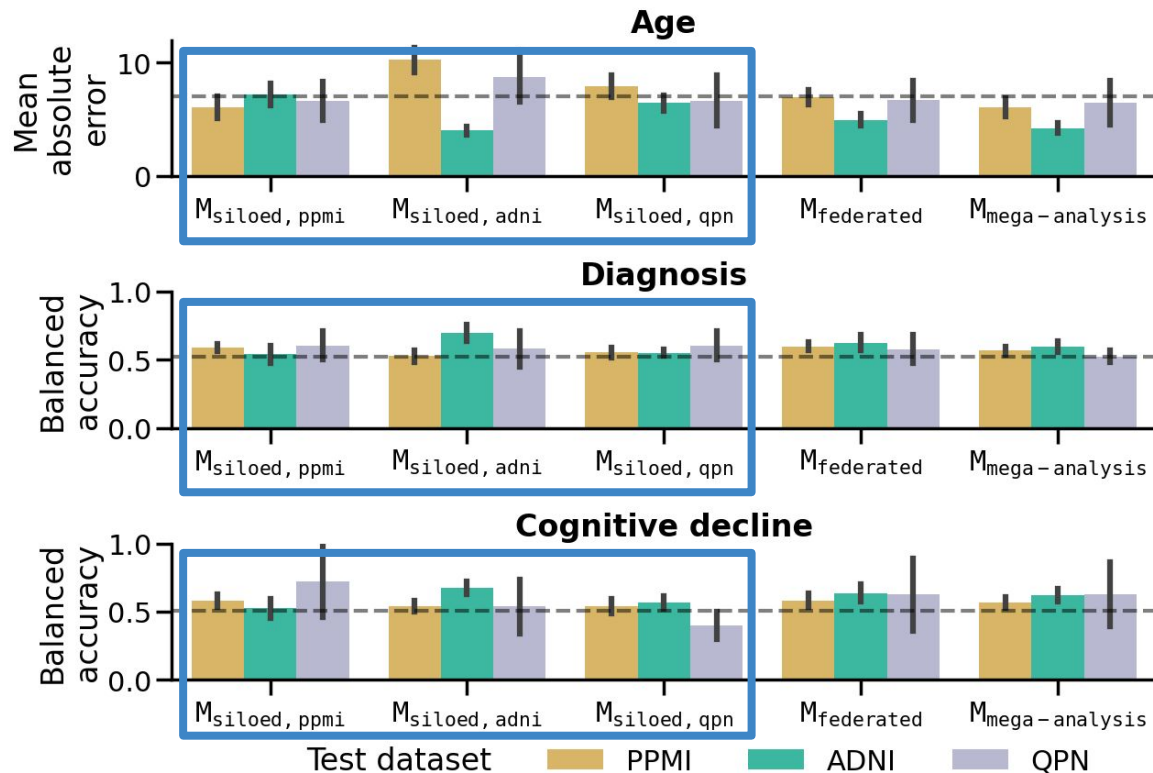
- Z-score
- Logistic/linear regression

5-fold cross-validation

Test datasets:

- PPMI
- ADNI
- QPN
- PPMI+ADNI+QPN

Preliminary results

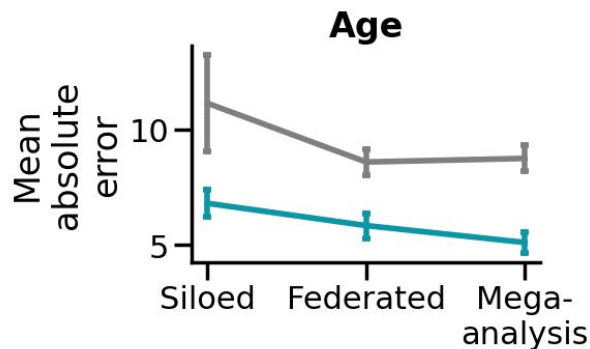
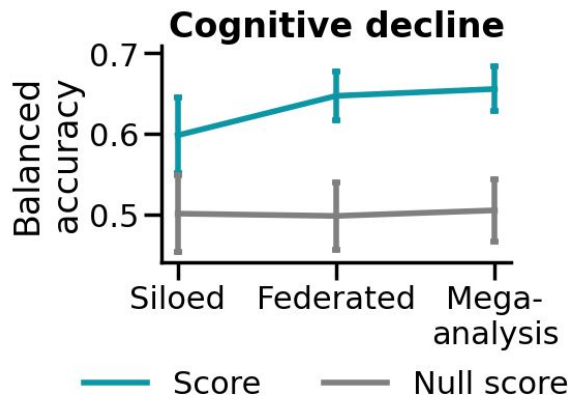
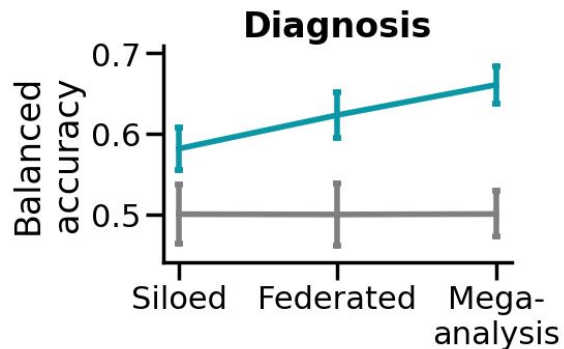


Effects of **dataset shift**:

- Poor generalizability of models trained with **Siloed** setup

Preliminary results

Test performance scores on **PPMI+ADNI+QPN** show **improvement from Siloed to Federated setup** in all three prediction tasks



Discussion

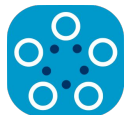
- Preliminary results suggest that, for the datasets and use-cases investigated, **federated setup shows generalizable and comparable performance to the mega-analysis approach**
 - But does not require sharing of sensitive clinical data
- Neuroinformatics infrastructure is needed for handling the practical aspects of this type of analysis



Nipoppy framework for organization and decentralized processing of neuroimaging-clinical datasets



Neurobagel ecosystem for decentralized metadata harmonization and search



Fed-BioMed framework for federated learning

Future!

Thank you!

- Contact: michelle.wang6@mail.mcgill.ca
- OHBM poster #1146 (June 27-28, 13:45-15:45)



Nikhil
Bhagwat



Alyssa
Dai



Sebastian
Urchs



Mathieu
Dugré



Arman
Jahanpour



Jean-Baptiste
Poline

NeuroDataScience-ORIGAMI Lab

