

# BA-BFL: Barycentric Aggregation for Bayesian Federated Learning



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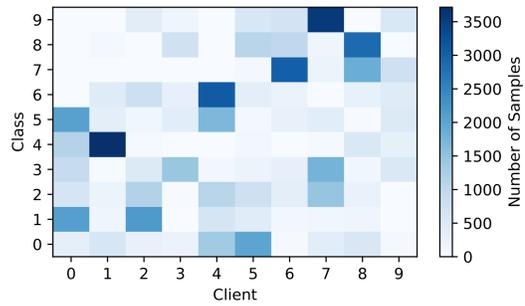
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## Introduction & Motivations

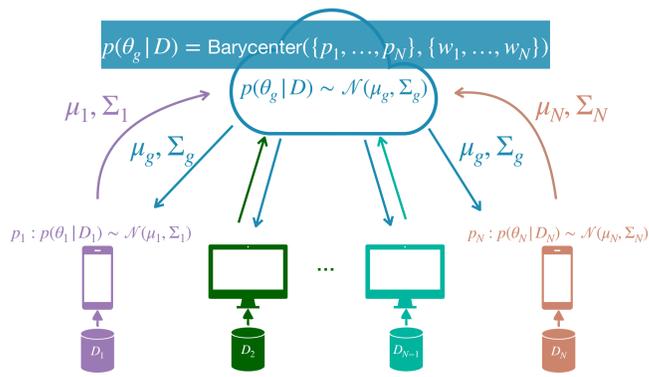
Federated Learning (FL) is a privacy-preserving decentralized learning paradigm consisting in 2 phases:

- **Learning Phase:** Each client trains its local model.
- **Aggregation Phase:** Local models are combined into a global model.

FL faces some challenges, like statistical heterogeneity, algorithmic fairness and uncertainty quantification.



## Main Contributions



The BA-BFL Framework.

Closed form solutions for:

- Reverse KL Barycenter of Gaussians
- Wasserstein-2 Barycenter of Independent Gaussians

A Unifying Framework:

**Deterministic Arithmetic Average Aggregations**  
*FedAvg, FedProx, PropFair, GiFair, FedLaw*

**WB / RKL**  
with variance  $\rightarrow 0$

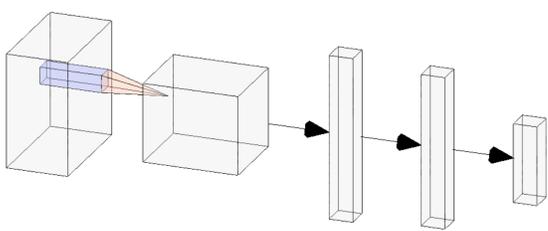
**Multiplicative Aggregations of Posteriors**  
*FedPA, FOLA, FedEP, FedIvon*

**RKL**

$$\Sigma_{RKL} = \left( \sum_{k=1}^N w_k \Sigma_k^{-1} \right)^{-1}, \quad \mu_{RKL} = \Sigma_{RKL} \sum_{k=1}^N w_k \Sigma_k^{-1} \mu_k.$$

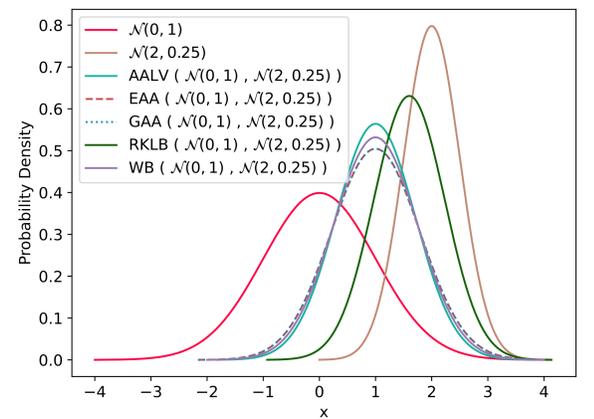
$$\Sigma_{W_2} = \left( \sum_{k=1}^N w_k \Sigma_k^{\frac{1}{2}} \right)^2, \quad \mu_{W_2} = \sum_{k=1}^N w_k \mu_k.$$

## Comparison to Statistical Aggregation Methods in a Heterogeneous Setting



- The models' architecture consists of two convolutional layers and three fully connected layers.
- The last  $n = 0, 1, 2, 3$  layers are implemented as Bayesian fully connected layers following the HBDL approach.
- The remaining layers are deterministic.

Nbl	Algorithm	FashionMNIST	SVHN	CIFAR-10
0	FedAvg	87.88 ± 0.79	86.06 ± 0.45	61.63 ± 3.11
1	AALV	88.22 ± 0.34	86.52 ± 0.29	63.42 ± 3.02
	EAA	88.07 ± 0.22	86.24 ± 0.20	63.69 ± 2.47
	GAA	88.15 ± 0.31	86.36 ± 0.28	63.66 ± 2.22
	RKLB	88.07 ± 0.36	86.26 ± 0.26	63.37 ± 2.62
	WB	<b>88.34 ± 0.30</b>	<b>86.55 ± 0.37</b>	63.91 ± 2.64
2	AALV	87.62 ± 0.45	85.46 ± 0.10	65.03 ± 2.92
	EAA	87.53 ± 0.57	85.64 ± 0.33	64.02 ± 1.99
	GAA	87.82 ± 0.64	85.54 ± 0.44	64.59 ± 3.51
	RKLB	87.59 ± 0.57	85.57 ± 0.45	<b>65.20 ± 3.99</b>
	WB	87.69 ± 0.74	85.57 ± 0.51	64.74 ± 3.29
3	AALV	88.07 ± 0.58	86.15 ± 0.80	63.71 ± 3.63
	EAA	87.81 ± 0.54	86.04 ± 0.62	64.45 ± 1.79
	GAA	88.02 ± 0.55	86.27 ± 1.02	64.40 ± 2.30
	RKLB	87.77 ± 0.80	86.53 ± 1.03	64.55 ± 2.97
	WB	87.54 ± 0.54	85.99 ± 0.68	64.30 ± 2.55



Comparison of statistical and barycentric aggregations using univariate Gaussian distributions.

## Effect of Bayesian Layers on Uncertainty Quantification, Model Calibration, Time-Effectiveness & Fairness

