ML5G-PS-001: Federated Traffic Prediction for 5G and Beyond

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Base Station Traffic Forecasting

Main Challenges:

Data privacy

Real-time accurate predictions

Robust against non-iid data



^{*} This image has been generated using DALL·E 2

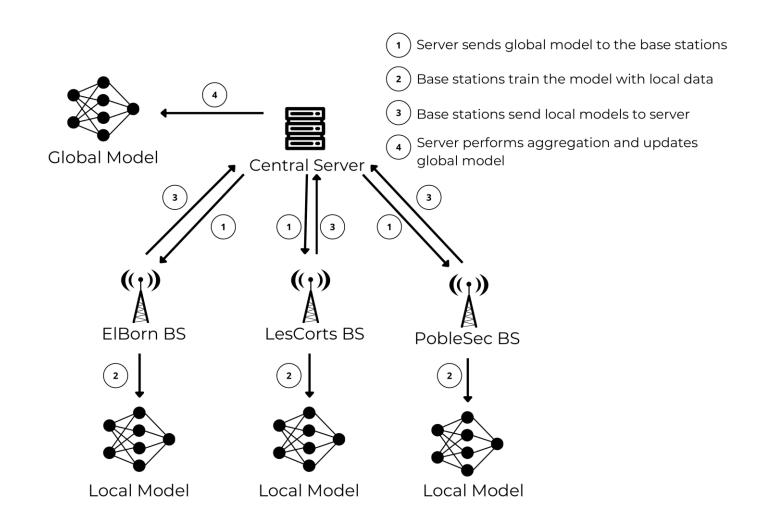
Federated Learning Approach

Advantages:

Privacy by-design

Low server latency

Generalization



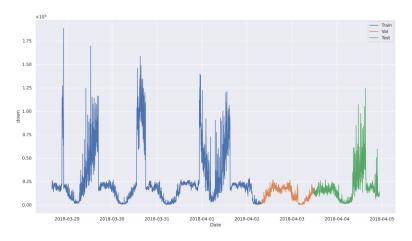
Challenges: non-iid data

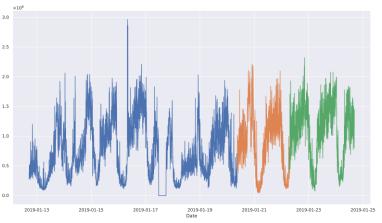
Three base stations in Barcelona, Spain. Differ in:

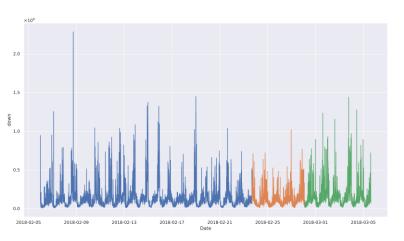
- 1) Quantity
- 2) Distribution
- 3) Temporality



How can we generalize a federated model?







Experiments

Key Observations

Pre-processing

Local pre-processing heavily influences the learning performance

Aggregation Algorithms

Aggregators specifically desinged to handle the non-iid data issue do not significantly outperform simple baselines

Predictive Accuracy

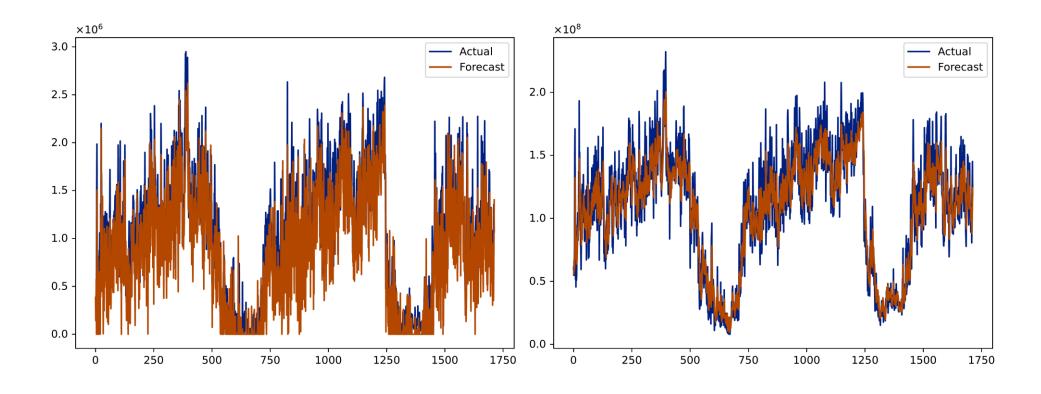
Models trained under the federated setting provide almost equivalent prediction errors to their centralized and local counterparts.

Carbon Footprint

Federated models leads to lower energy consumption and carbon footprint than centralized methods

REQUIREMENTS	FEDERATED	CENTRALIZED	LOCAL
Collaborative Training	~	~	
Privacy	~		~
Dynamic Execution	~		~
Generalization	~	~	

Federated Learning Results







Thank you!

Team Euclid

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