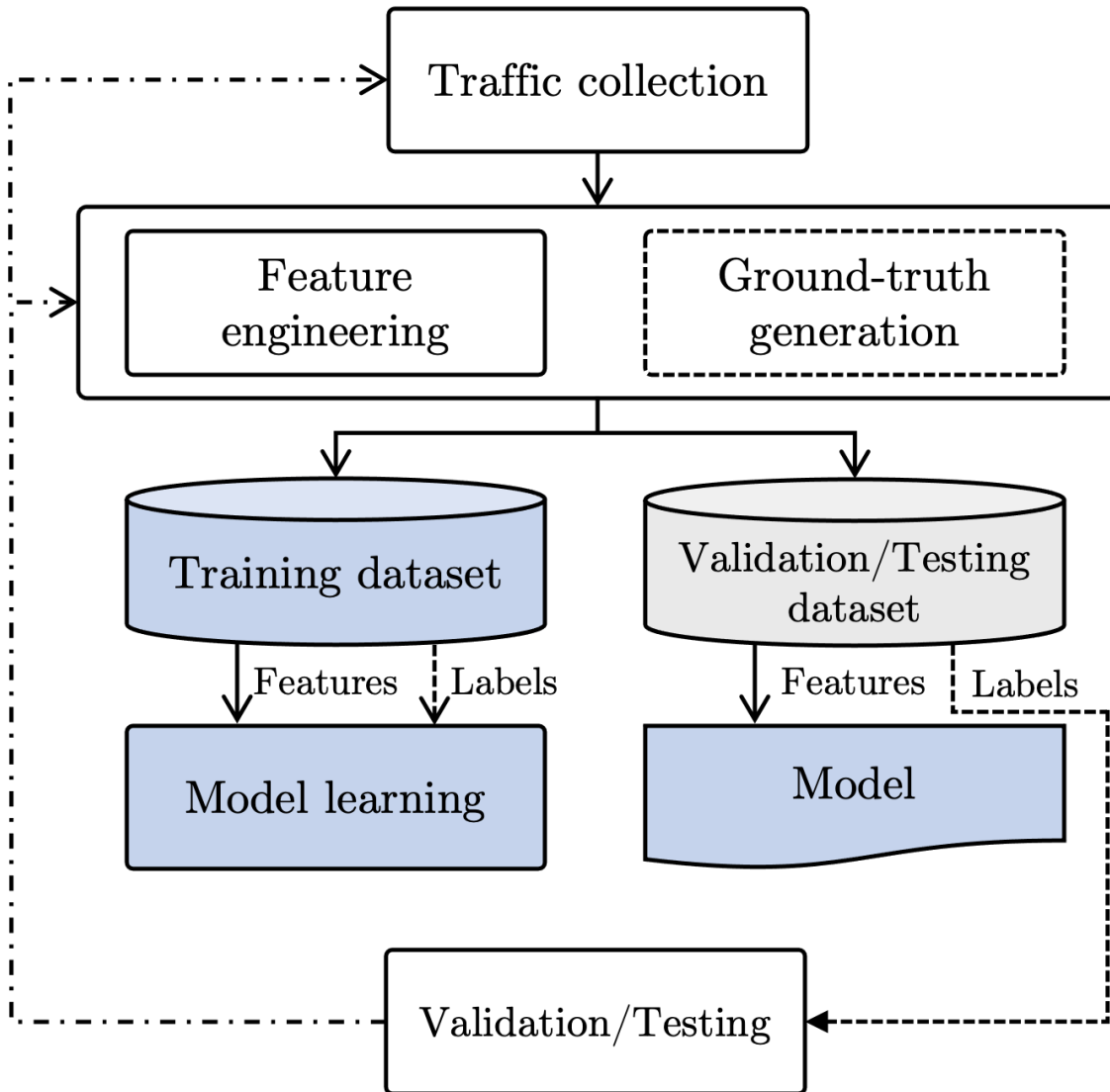


Privacy-Preserving Federated Learning: A New Horizon for Advanced Network Analytics

Dr. Adrián Pekár

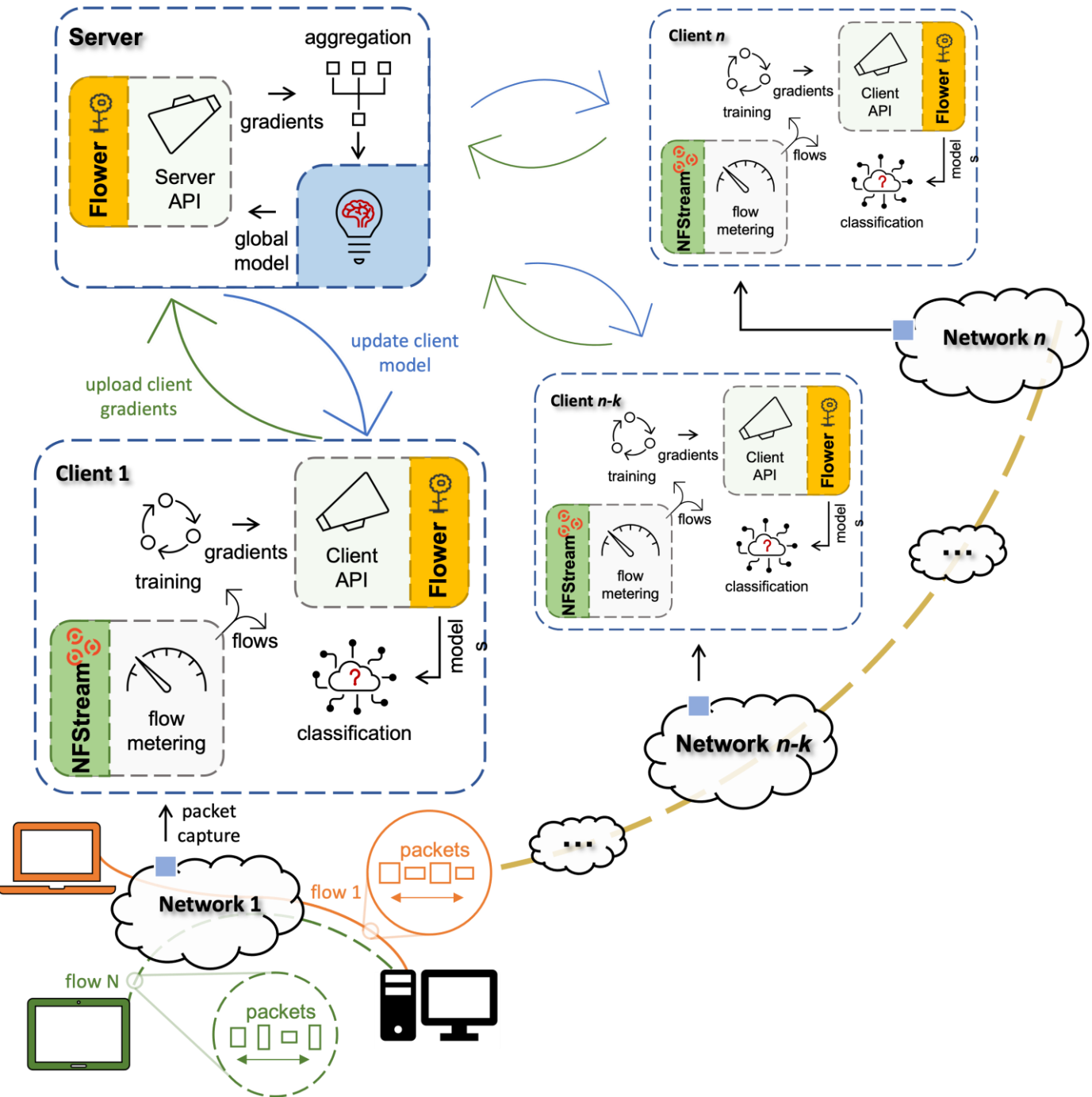


IP Flow Analysis Using Machine Learning

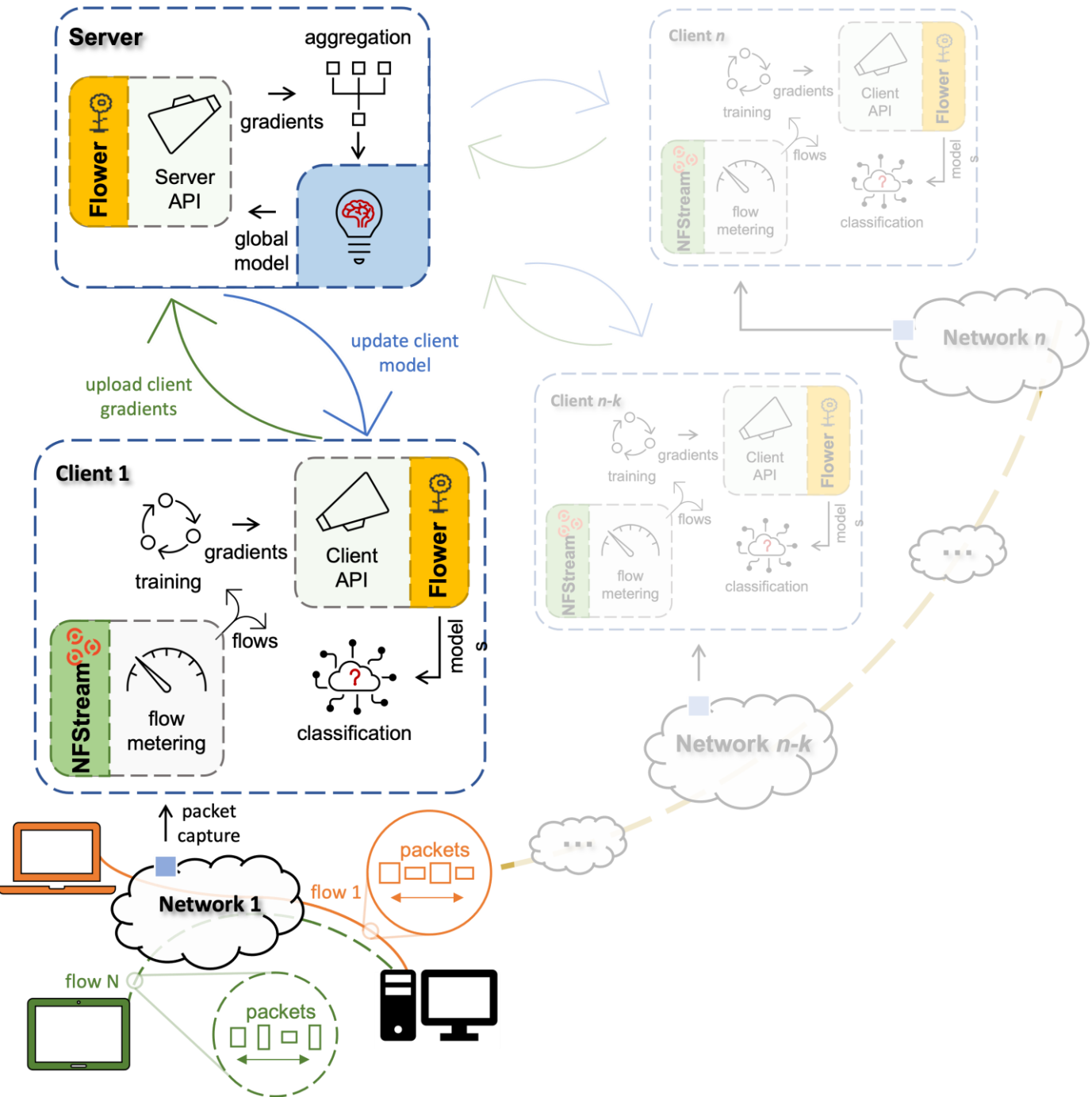
Leveraging
Federated Learning

for advanced network analytics,
all without sharing a byte of traffic data

Operational Principles

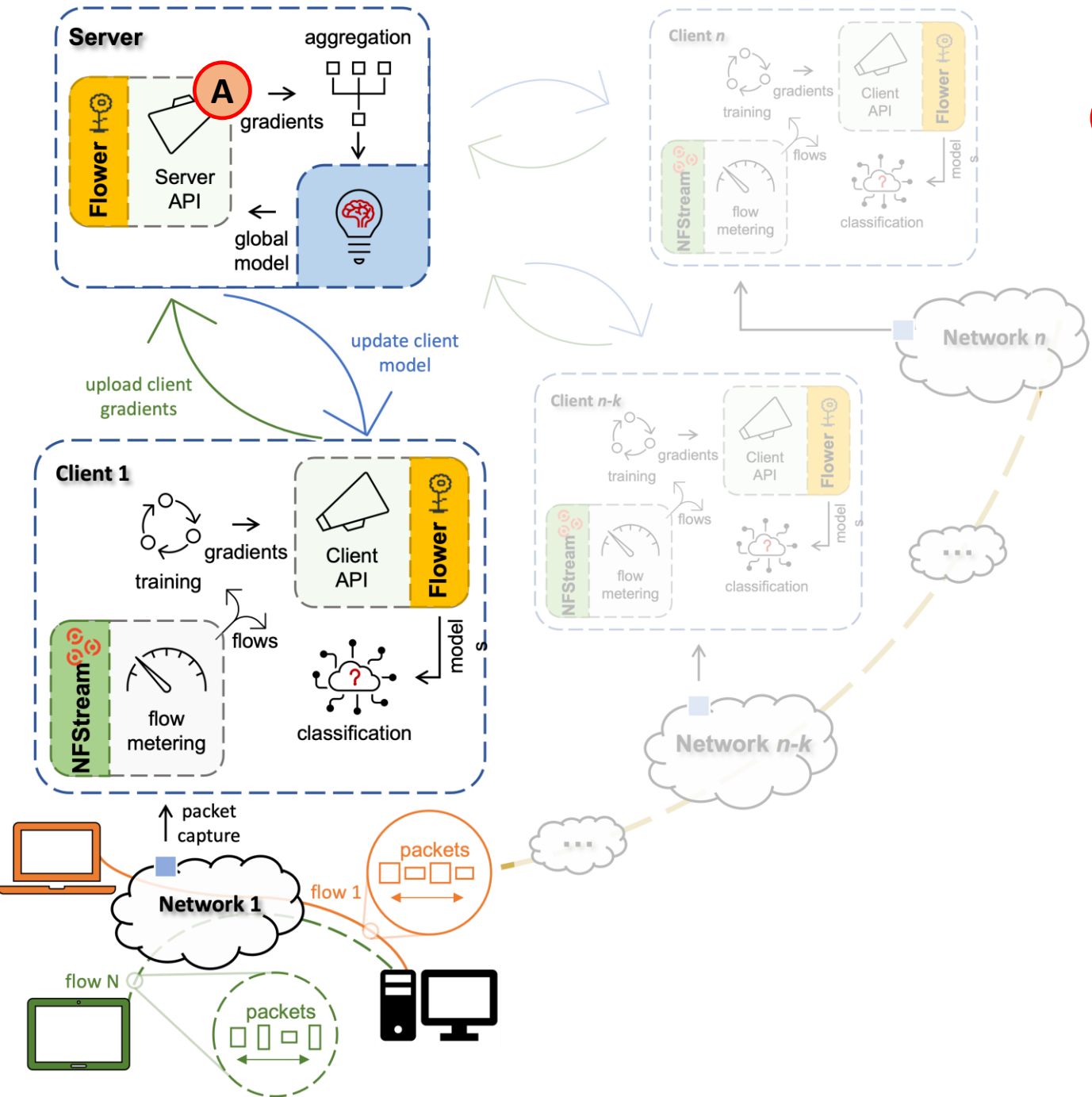


Operational Principles



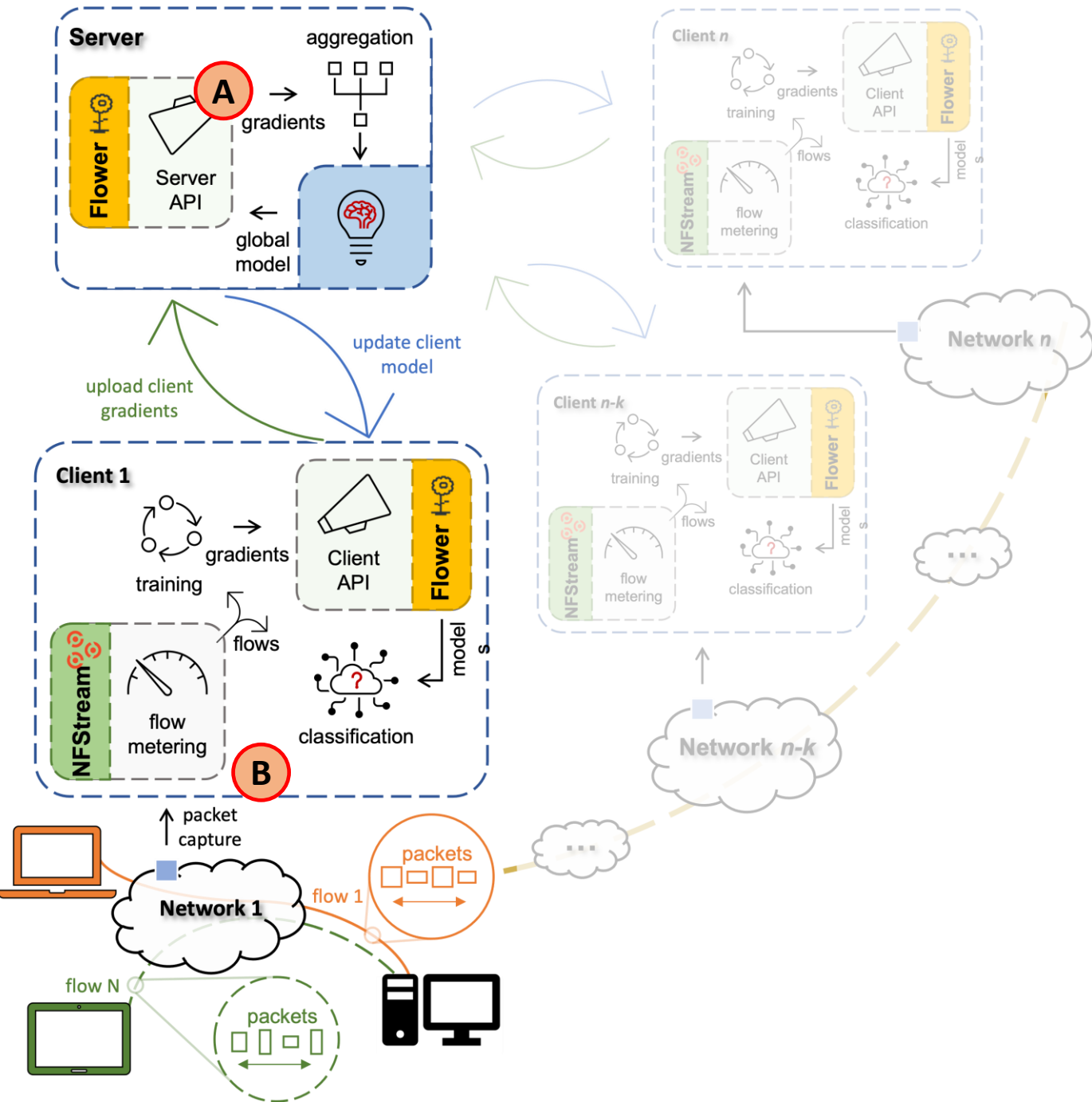
Operational Principles

A **Server Operation:** We run a secure server orchestrating the federated network, your cornerstone for collaborative learning.

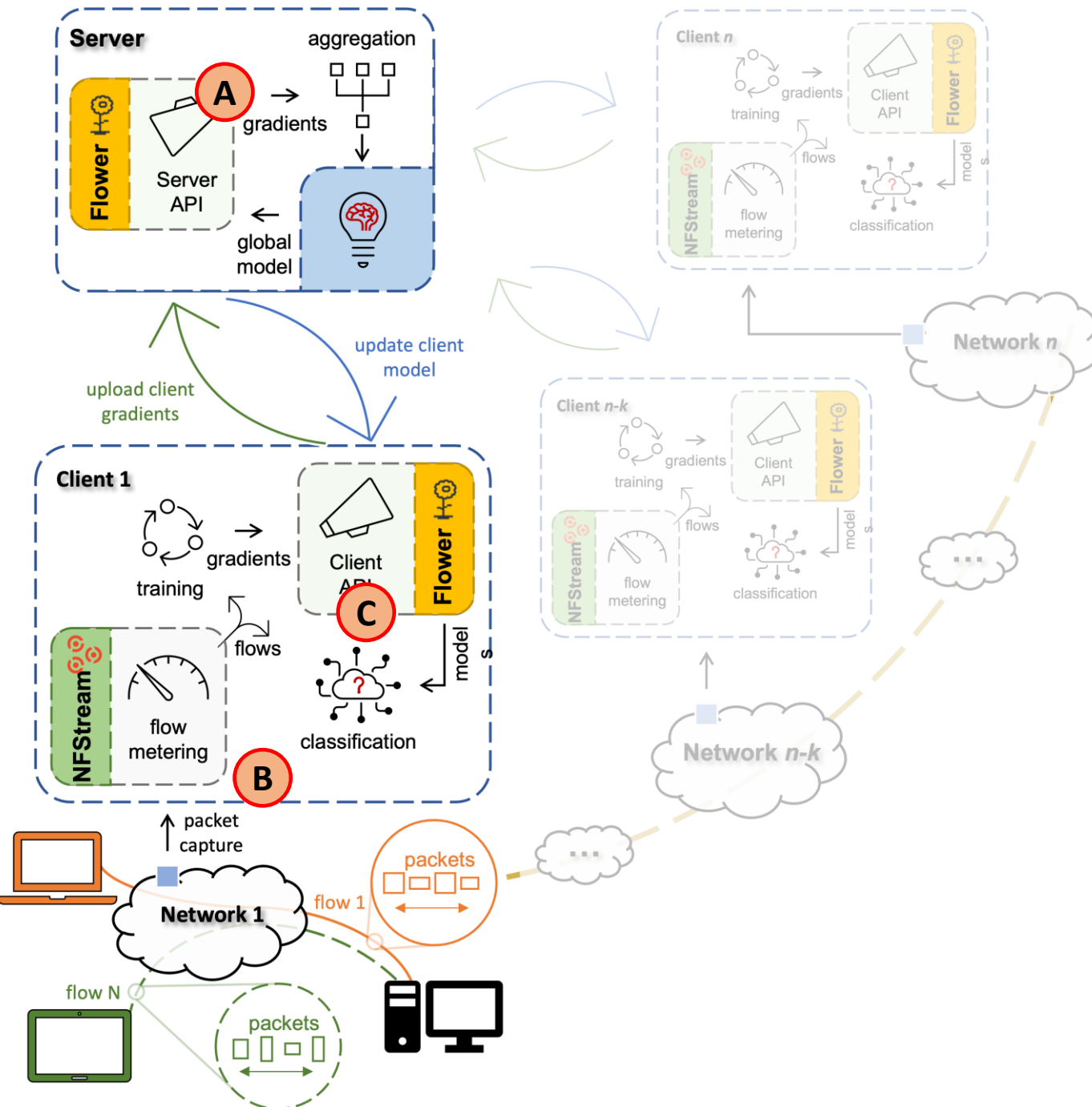


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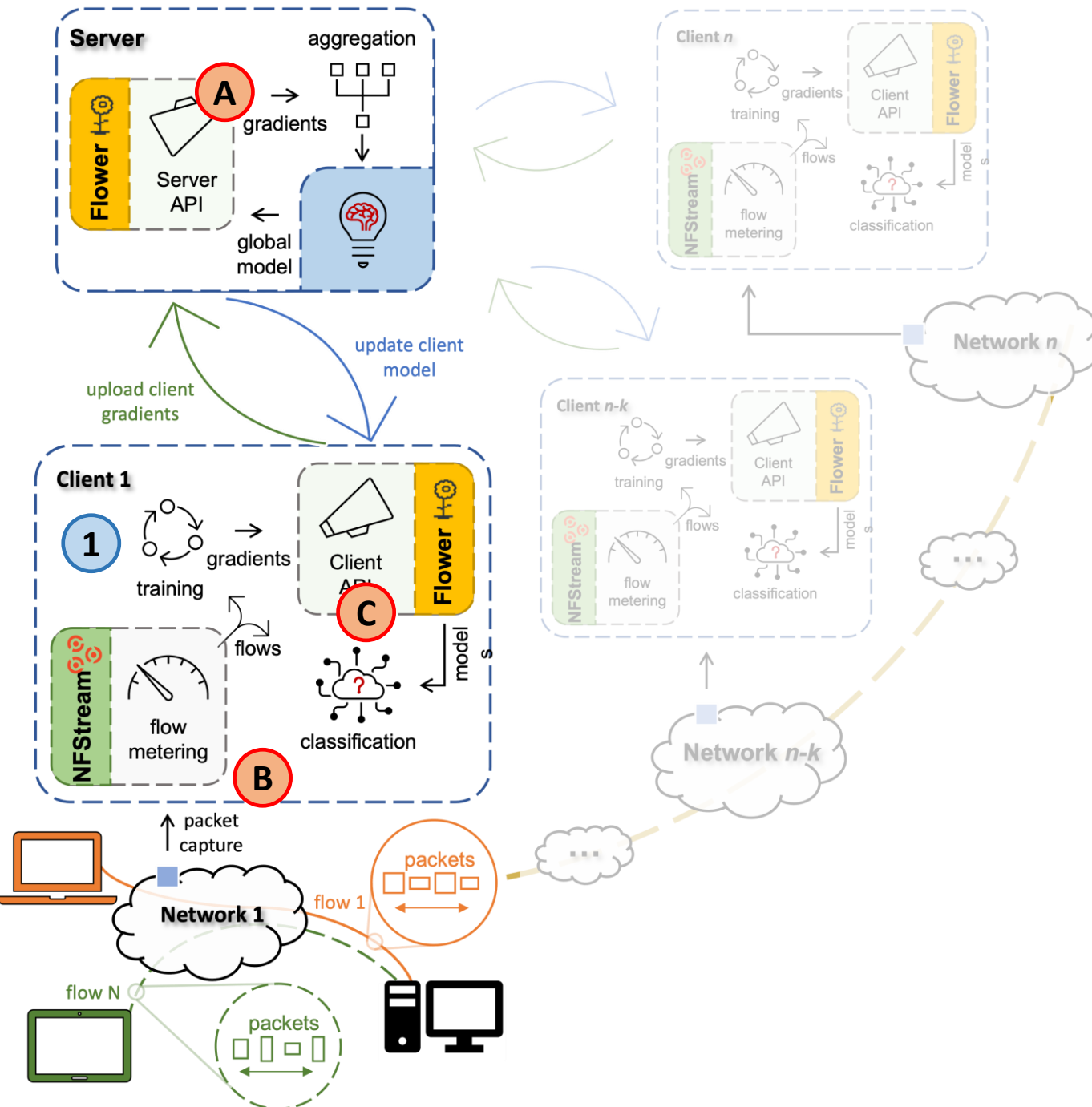


Operational Principles

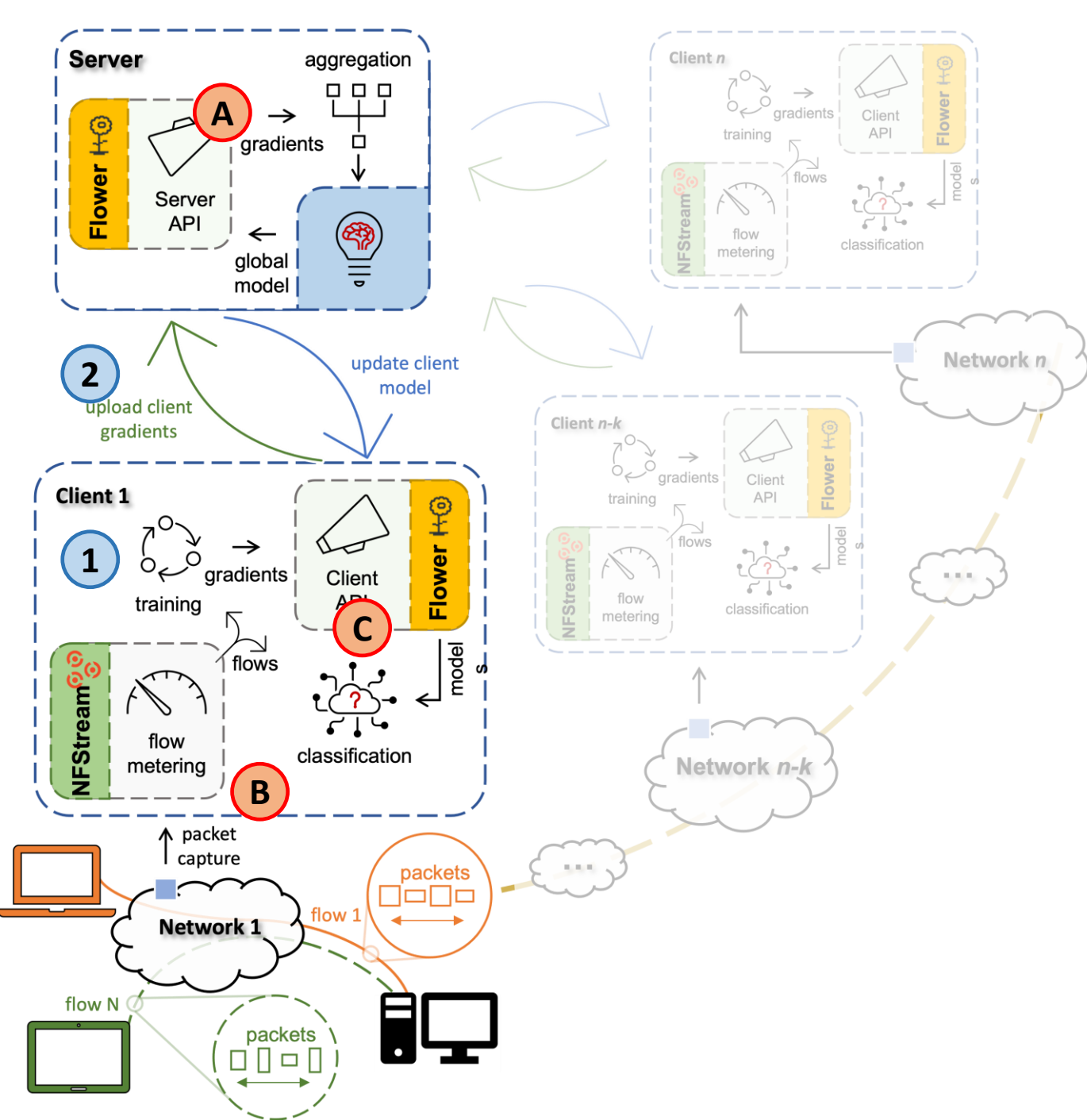


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Operational Principles

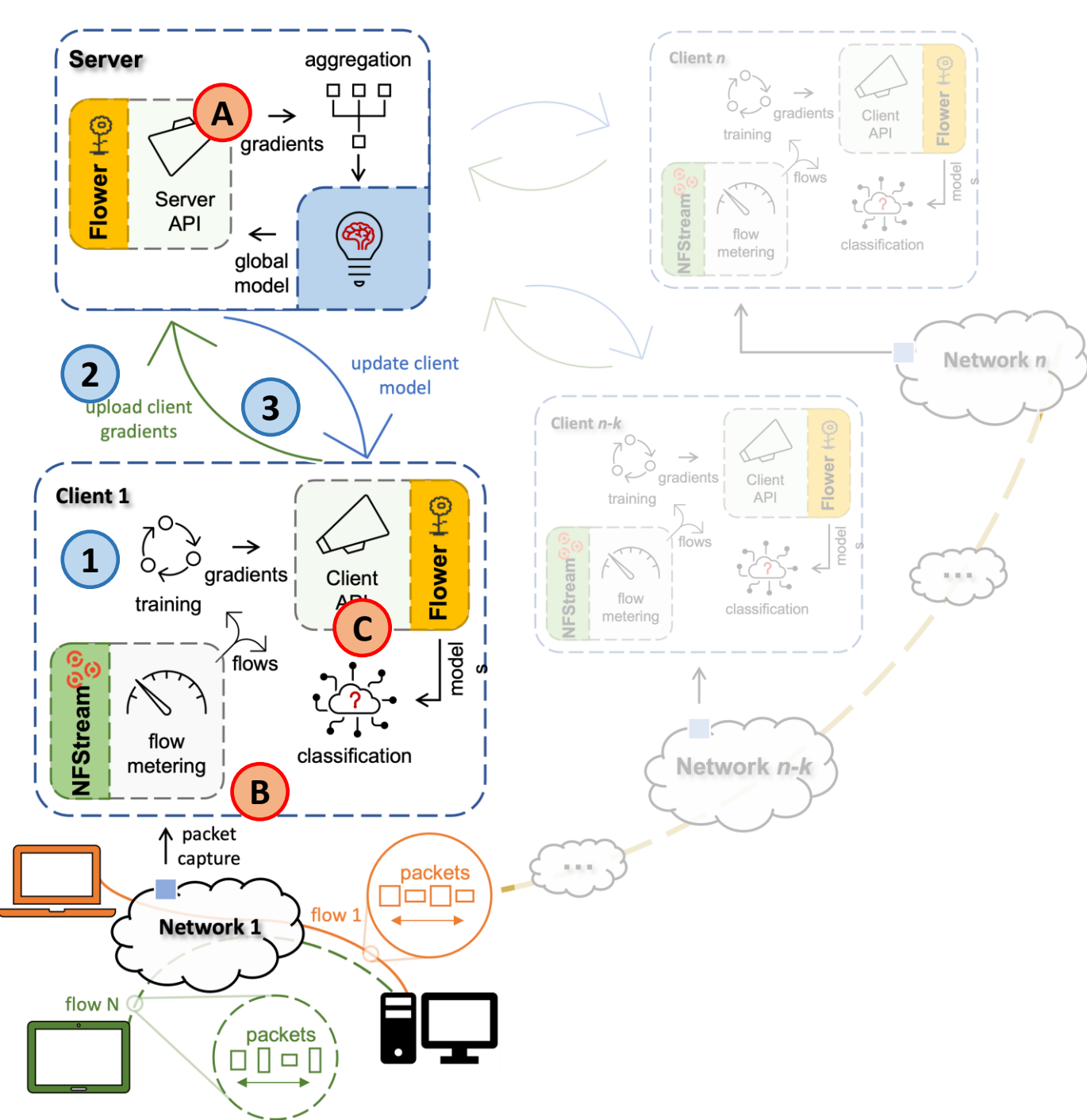


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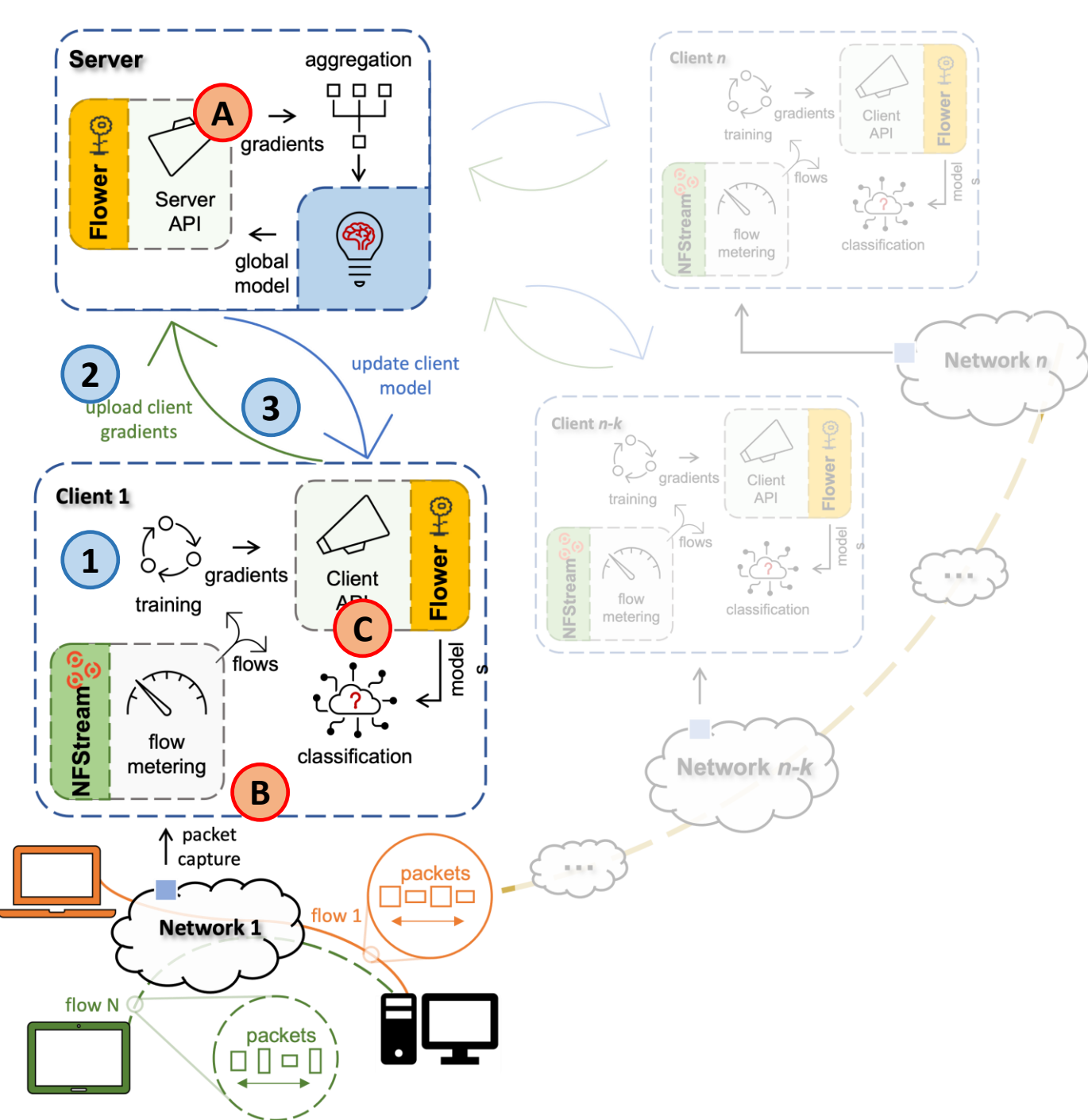
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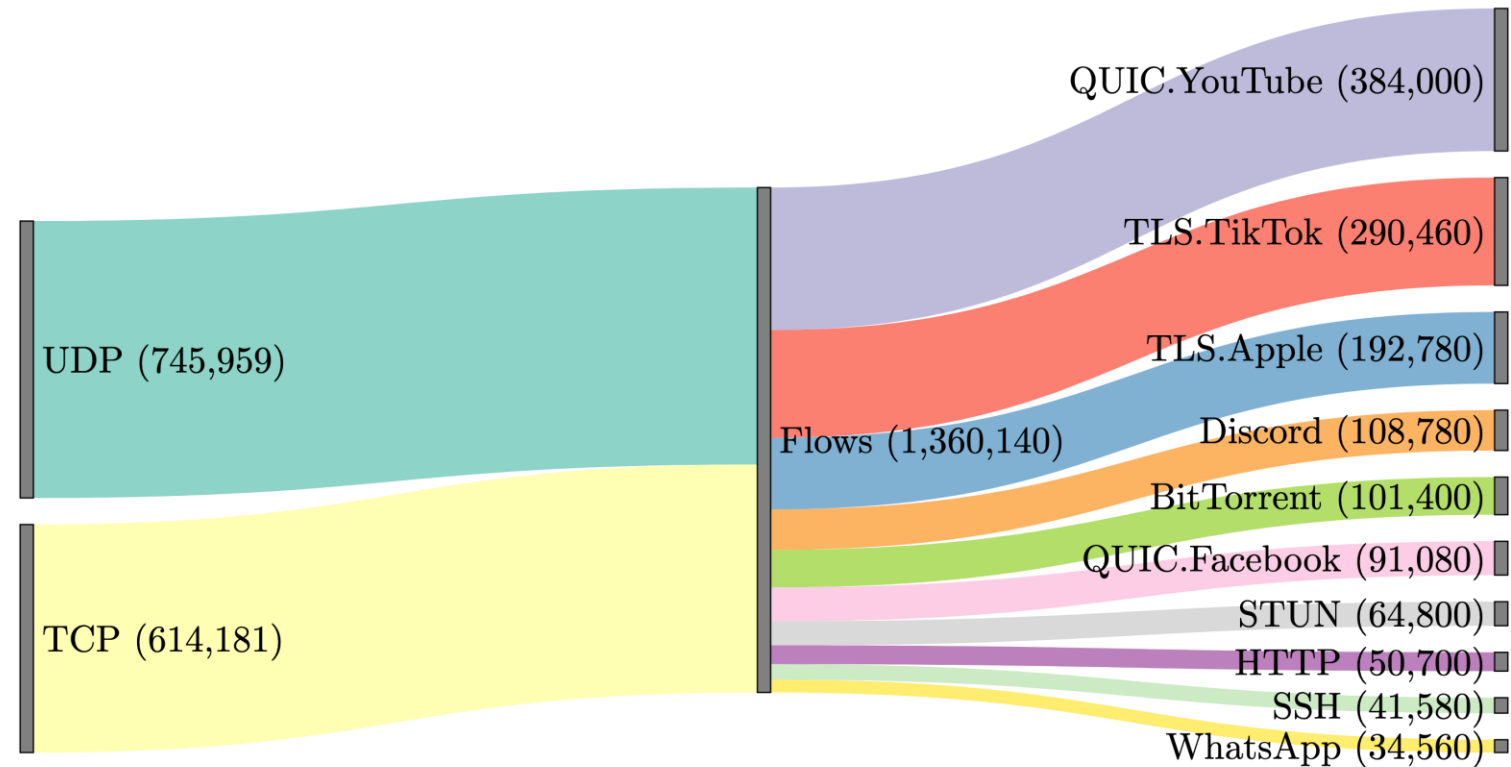
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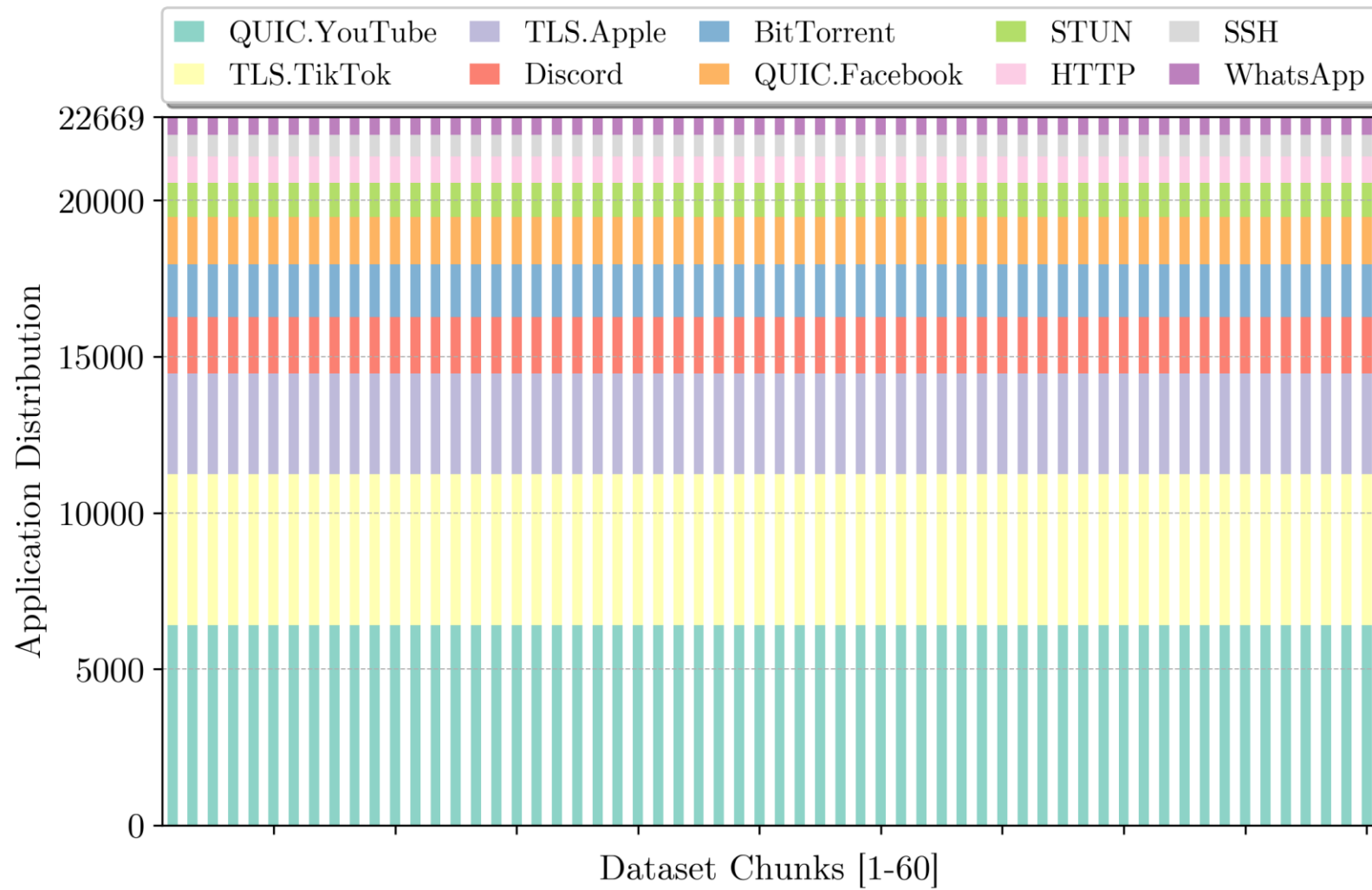
Learning & Privacy: By repeating these steps, we achieve a continuous learning cycle that keeps data local and privacy intact.

Proof of Concept Validation

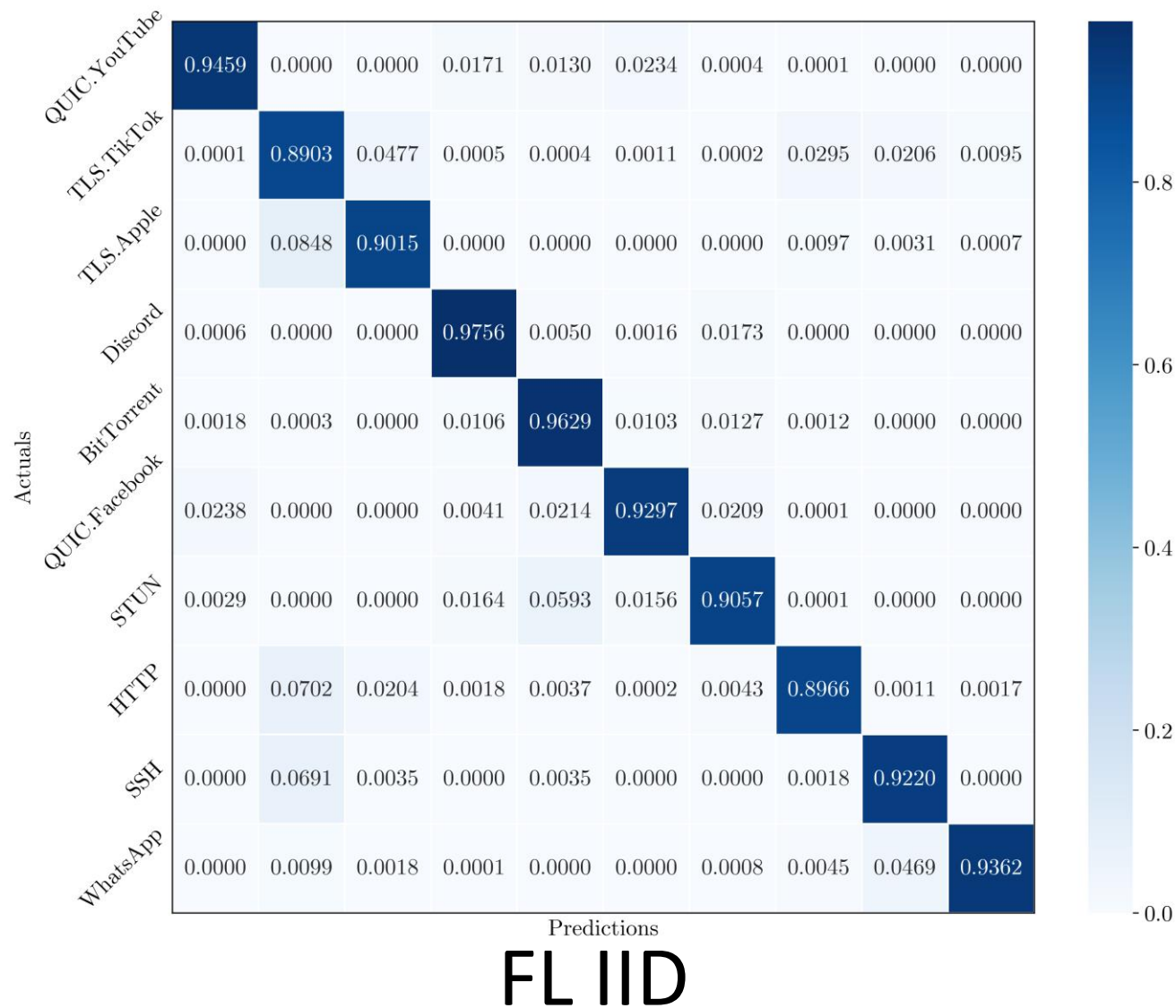
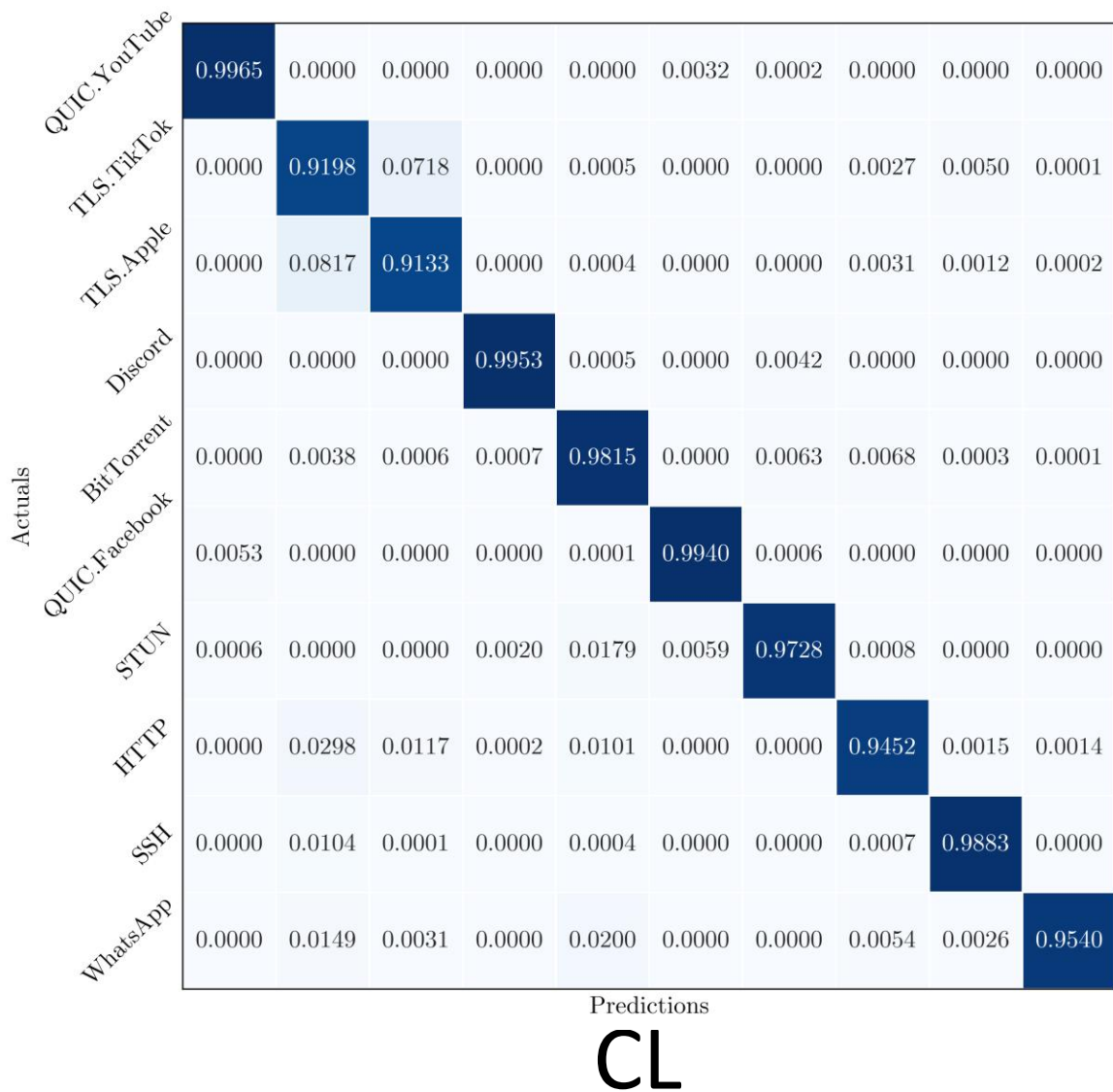
- L7 classification
- University network traffic
- 1 server
- 5 participants
- FNN: 18 (Input), 22 (ReLU), 18 (ReLU), 10 (Softmax/Output)
- CL vs FL IID vs FL non-IID



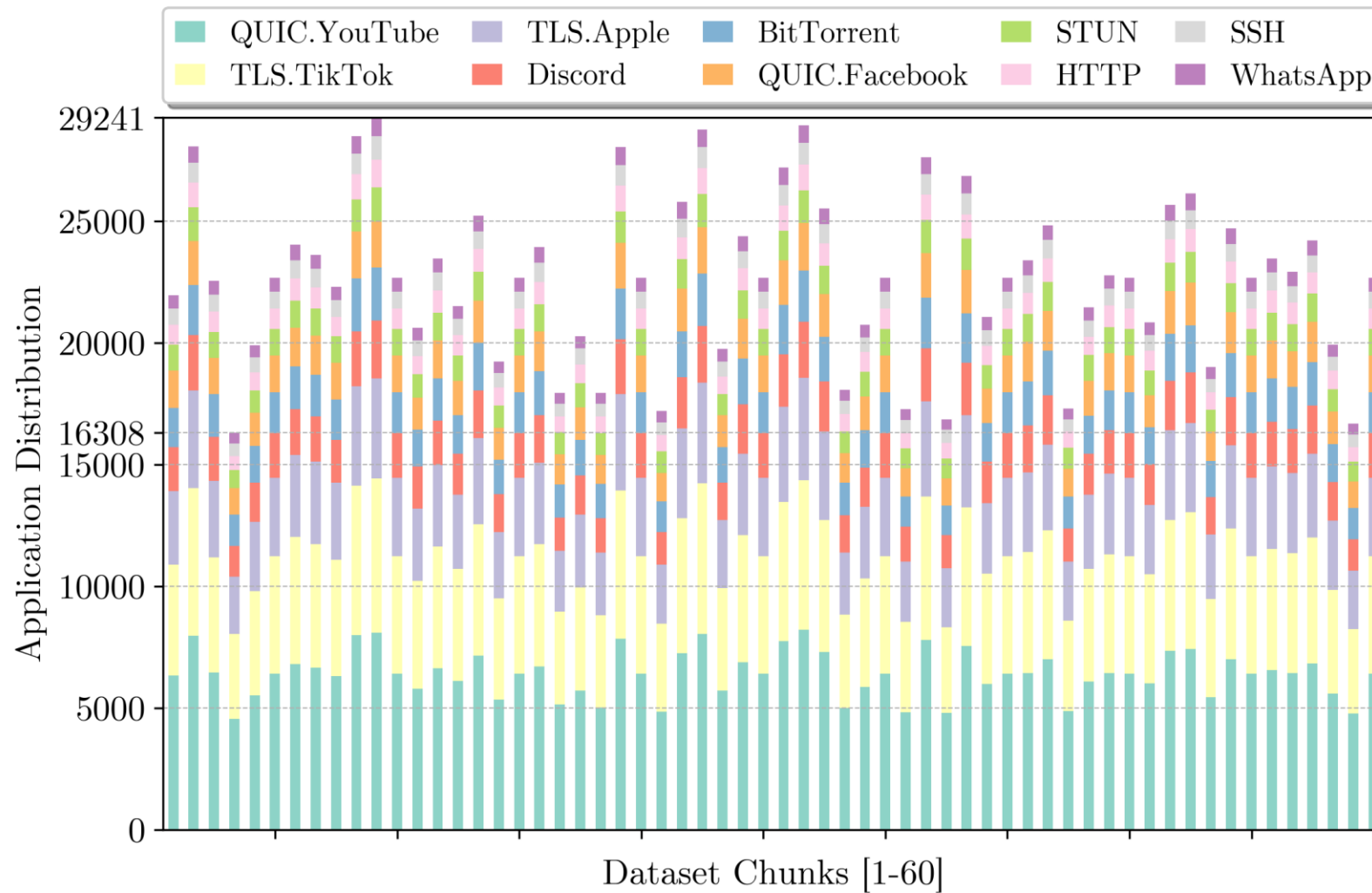
FL IID Scenario



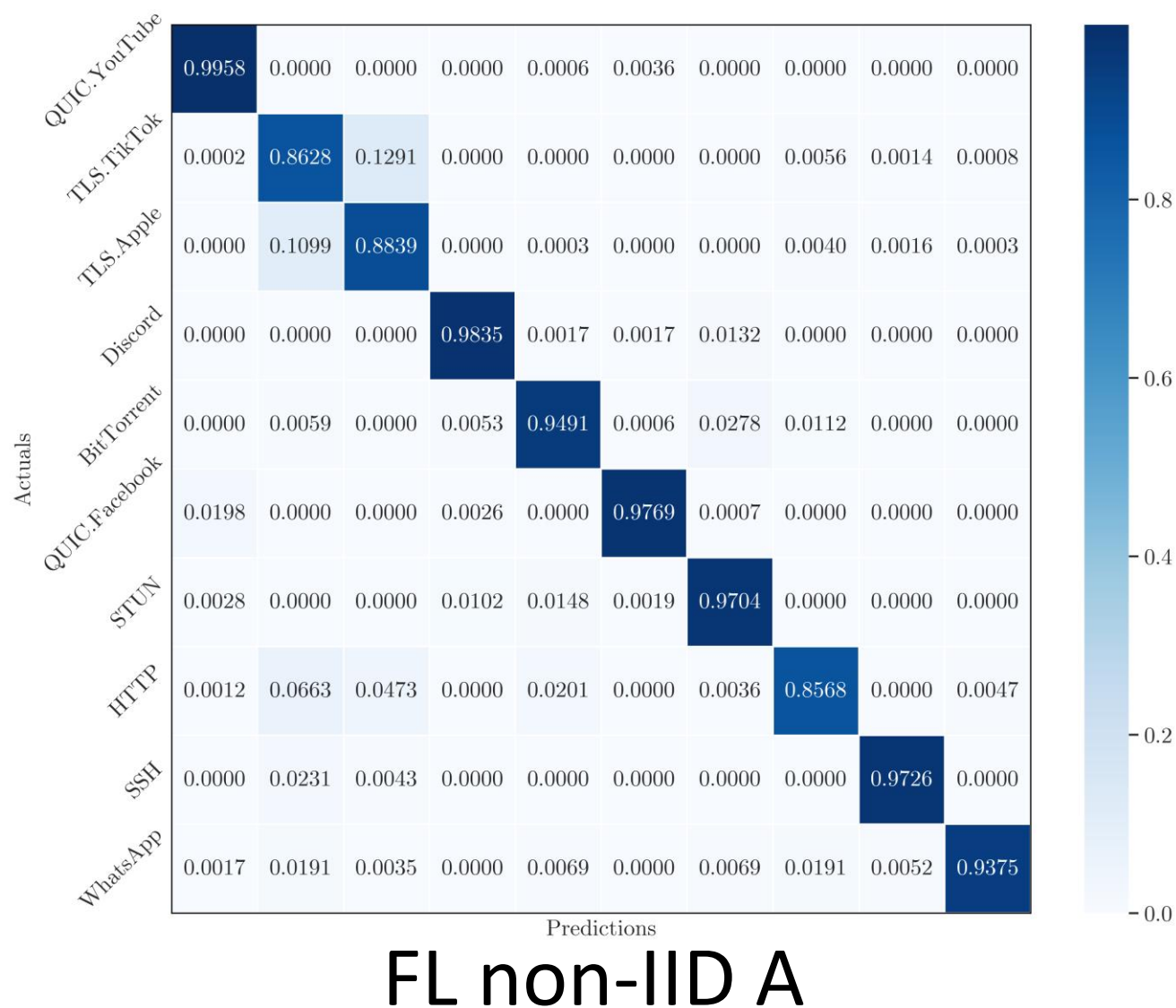
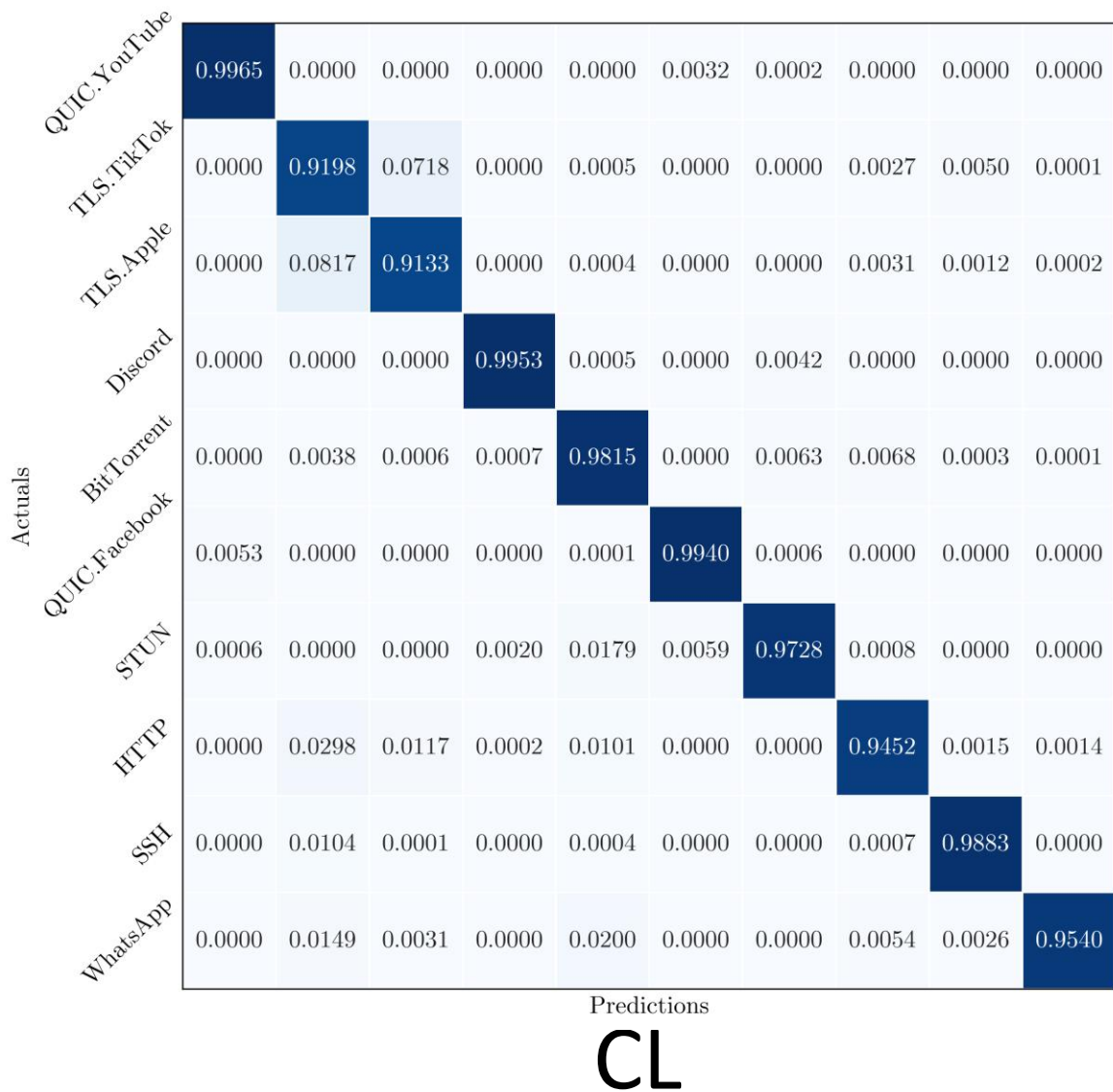
FL IID Scenario



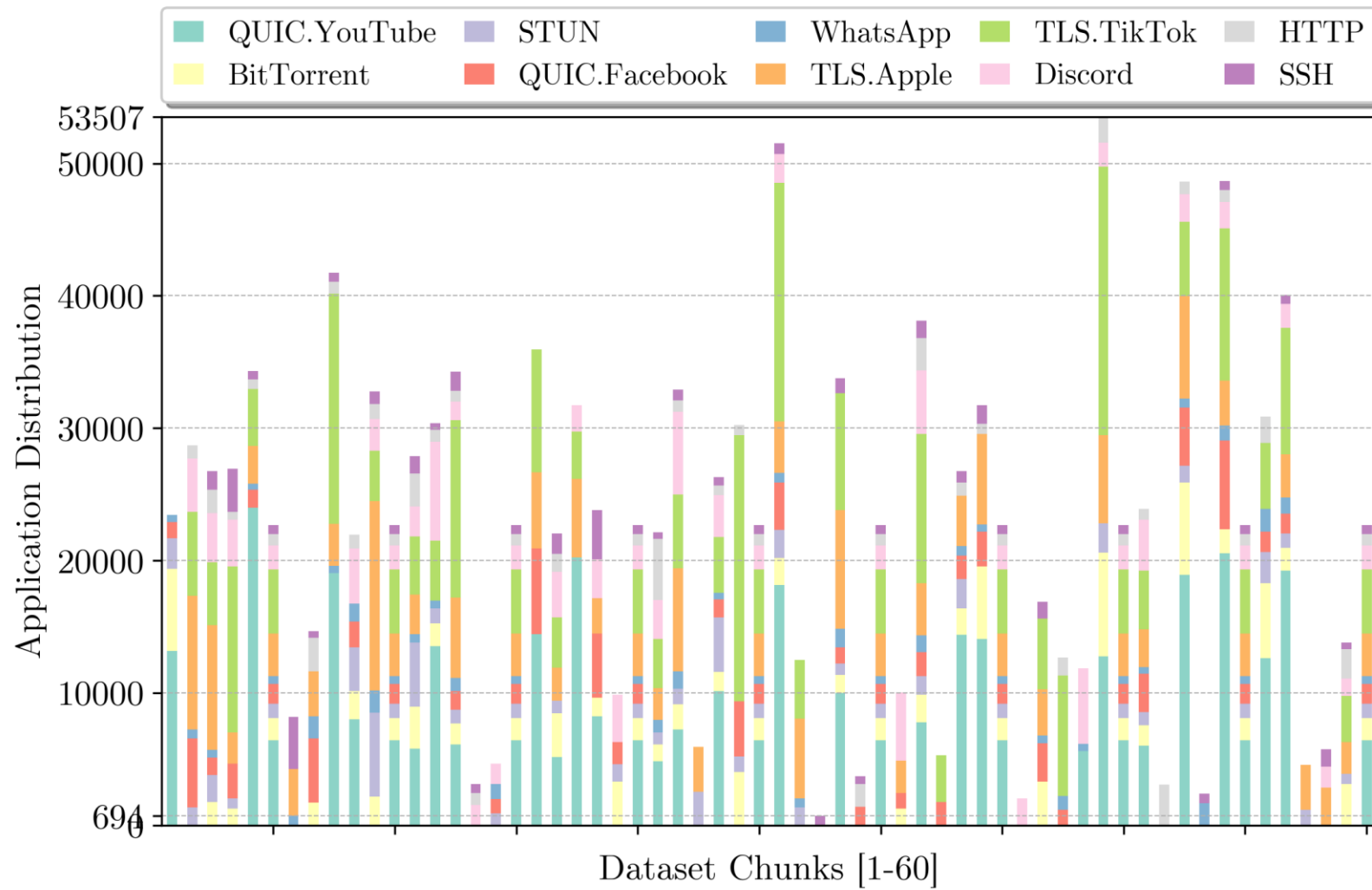
FL non-IID Scenario A



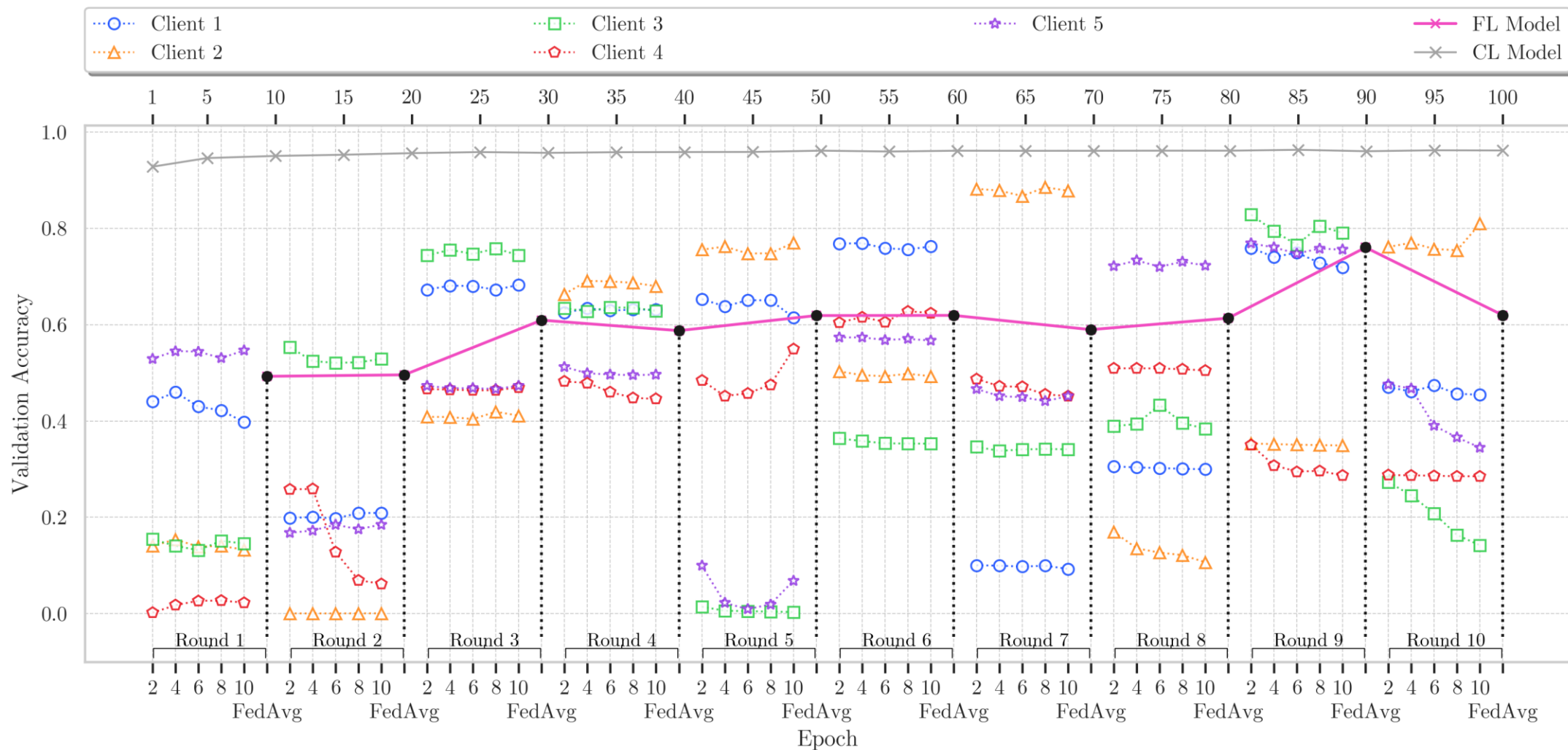
FL non-IID Scenario A



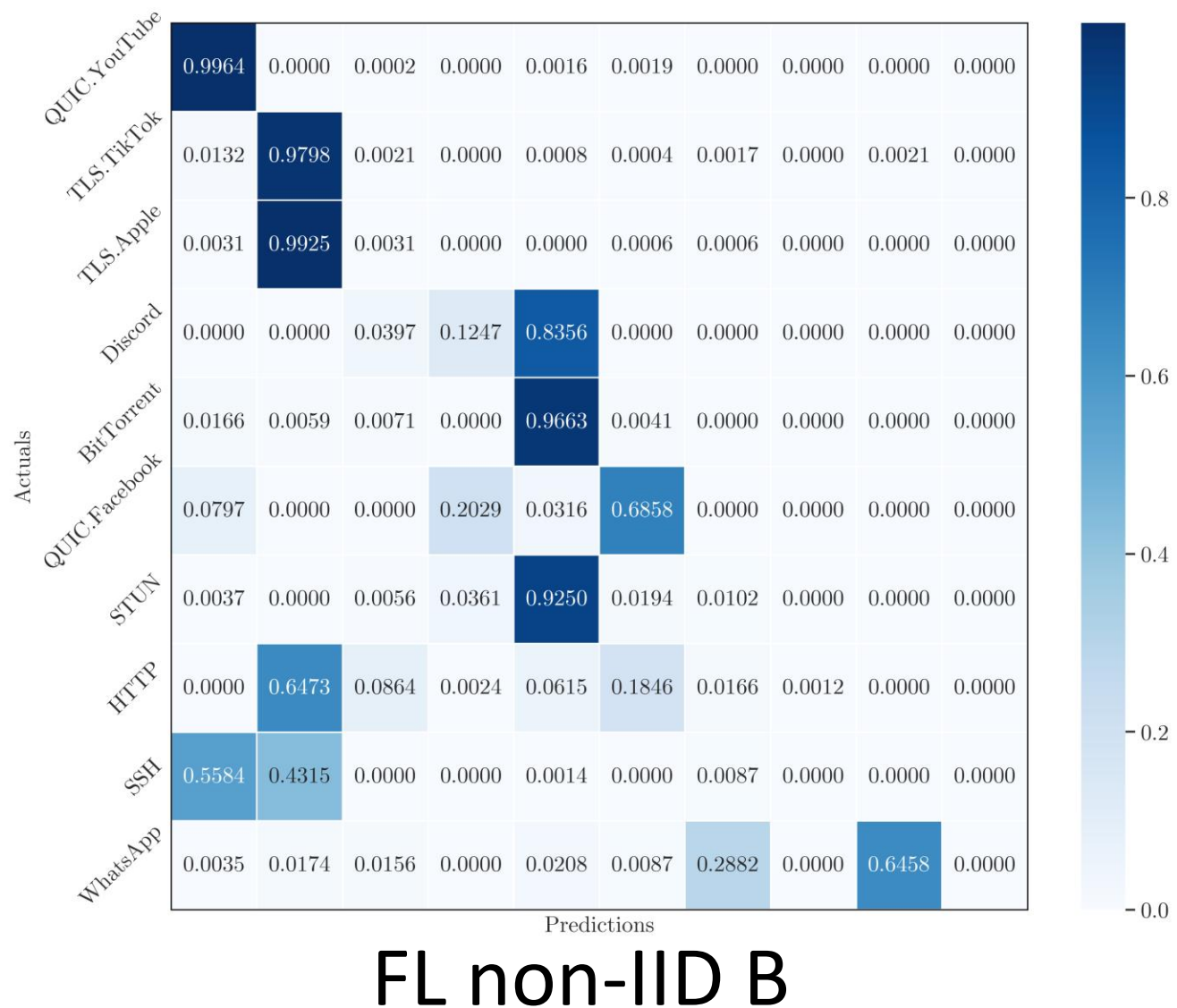
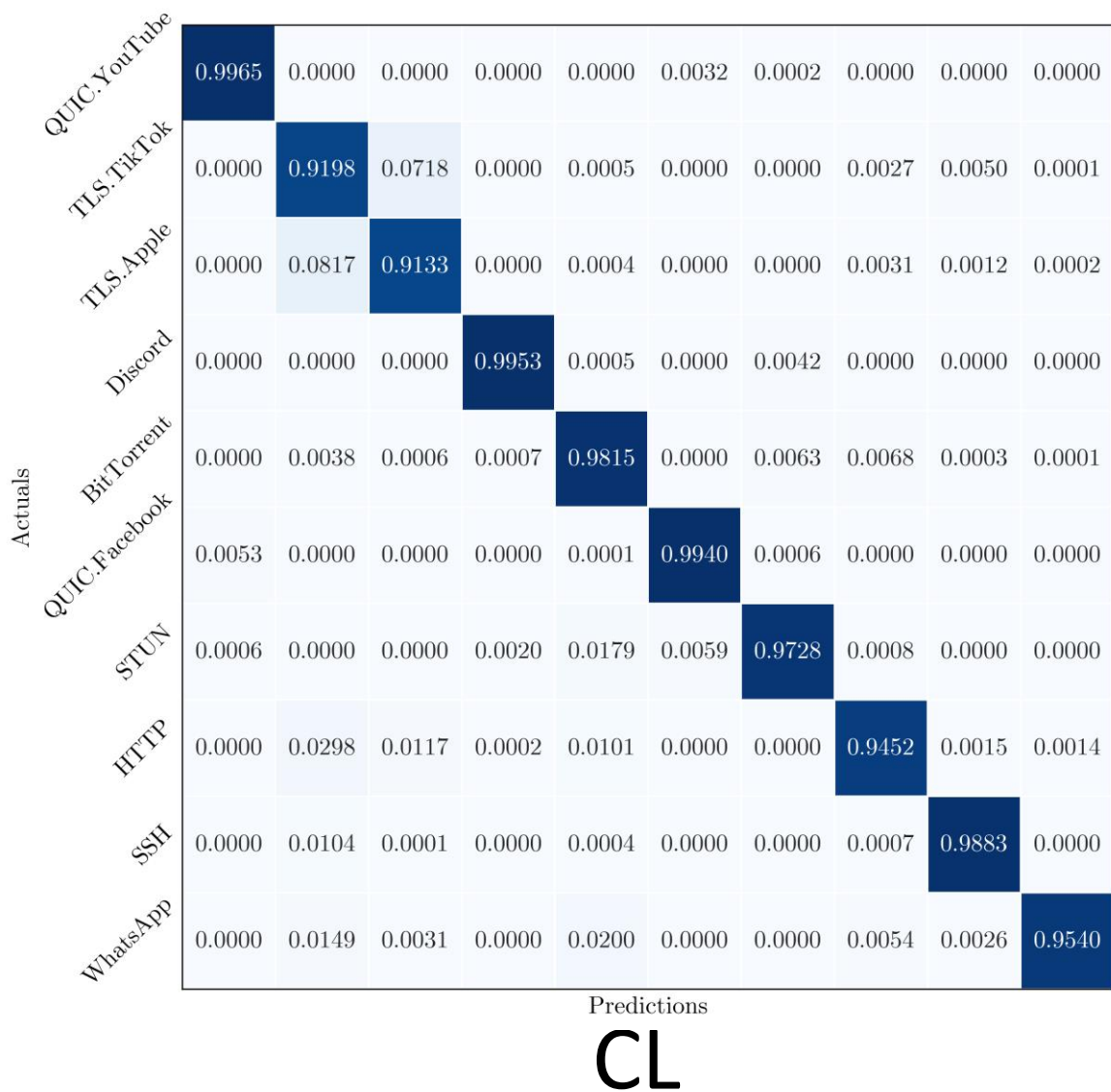
FL non-IID Scenario B



FL non-IID Scenario B



FL non-IID Scenario B



Prospects



CYBERSECURITY
THREAT DETECTION



PREDICTIVE
MAINTENANCE



INDUSTRY
NETWORK



LIVE TRAFFIC
ANALYTICS

Thank you

apekar@hit.bme.hu