

FedScale

Benchmarking Model and System Performance of Federated Learning At Scale

fedscale.ai

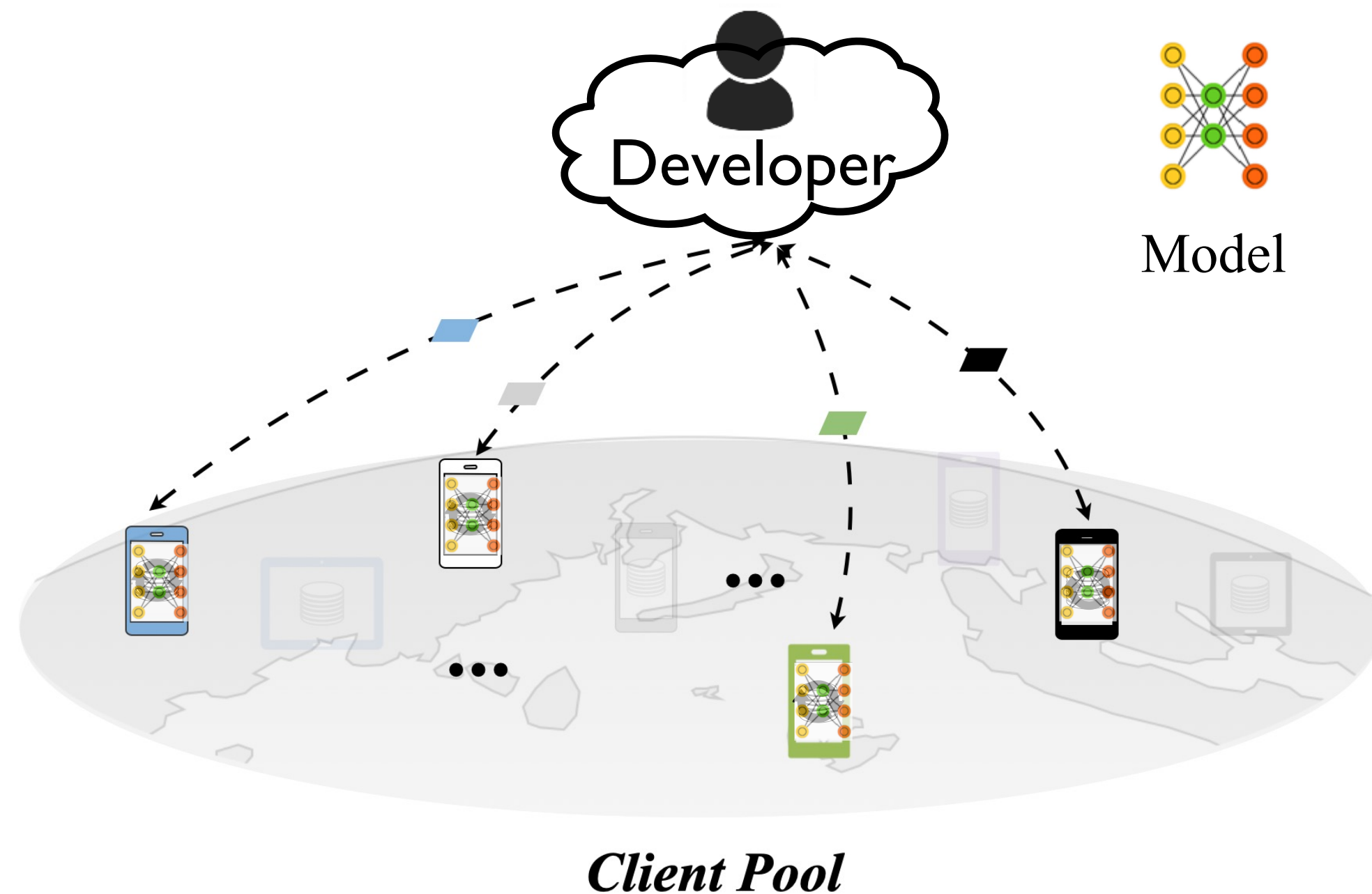
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Harsha V. Madhyastha, Mosharaf Chowdhury



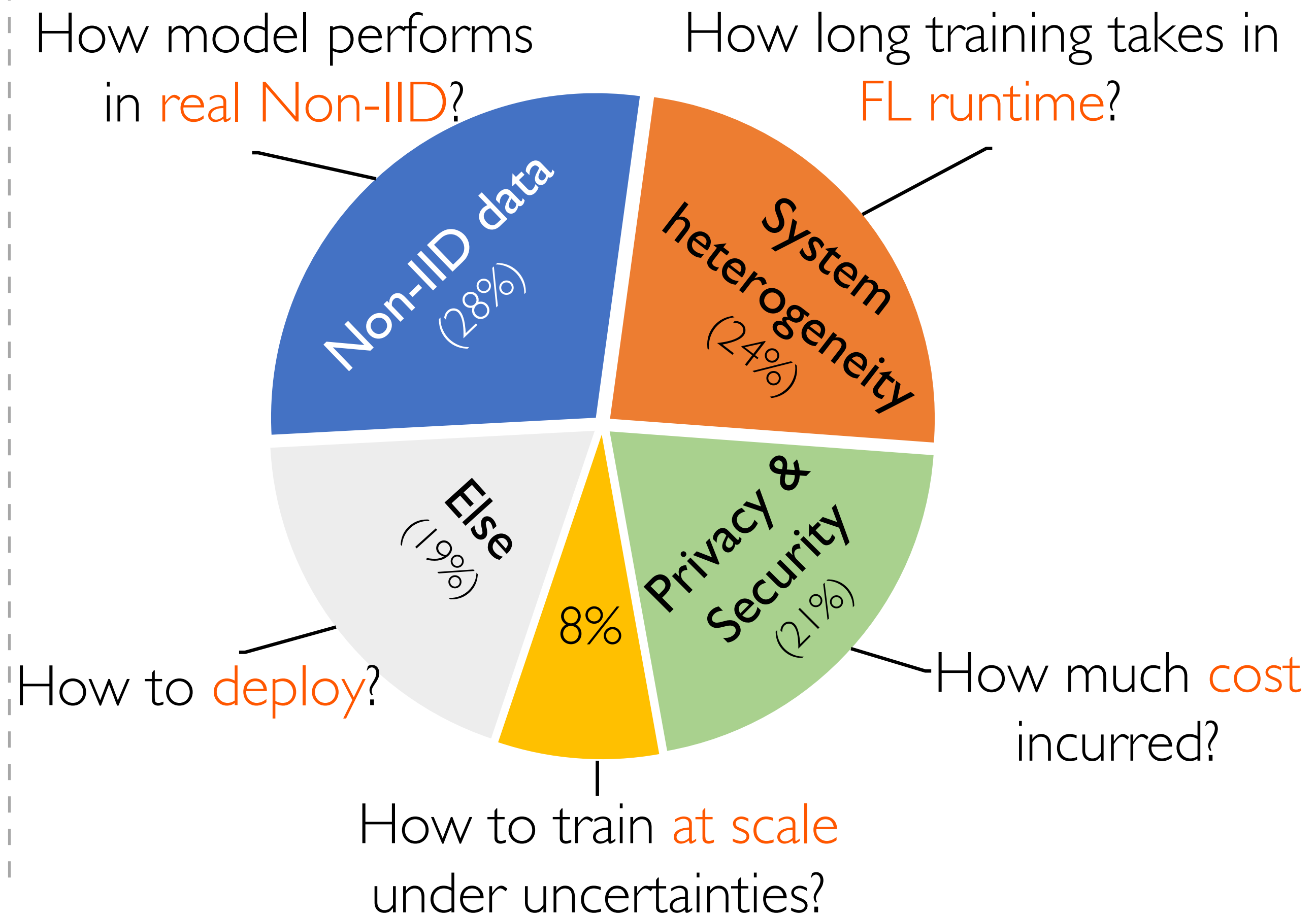
SymbioticLab

Federated Learning (FL) in Practice



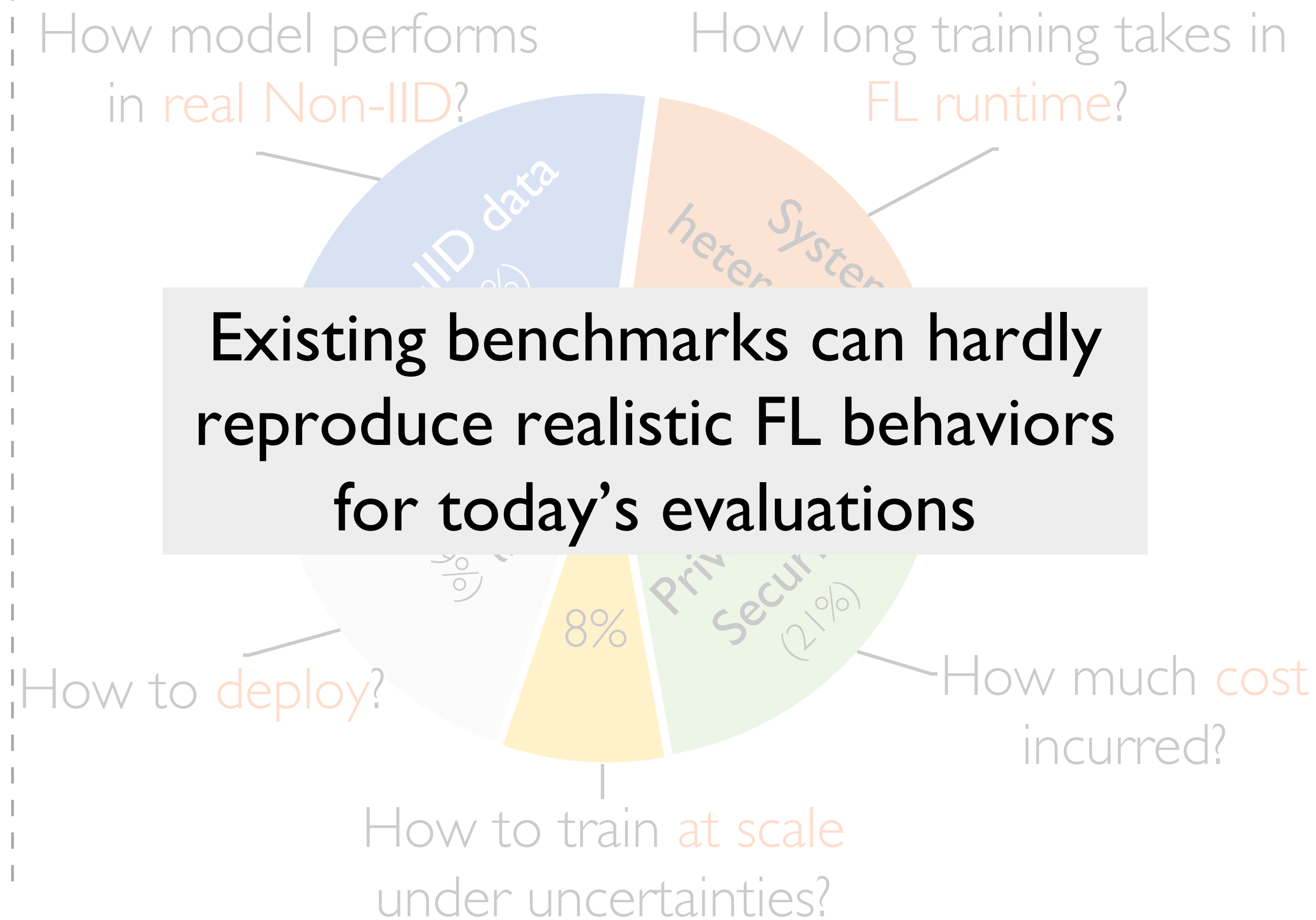
- @ Google, Meta, Apple, Nvidia, ...
- Training across millions of clients

Efforts for FL Challenges^[1]

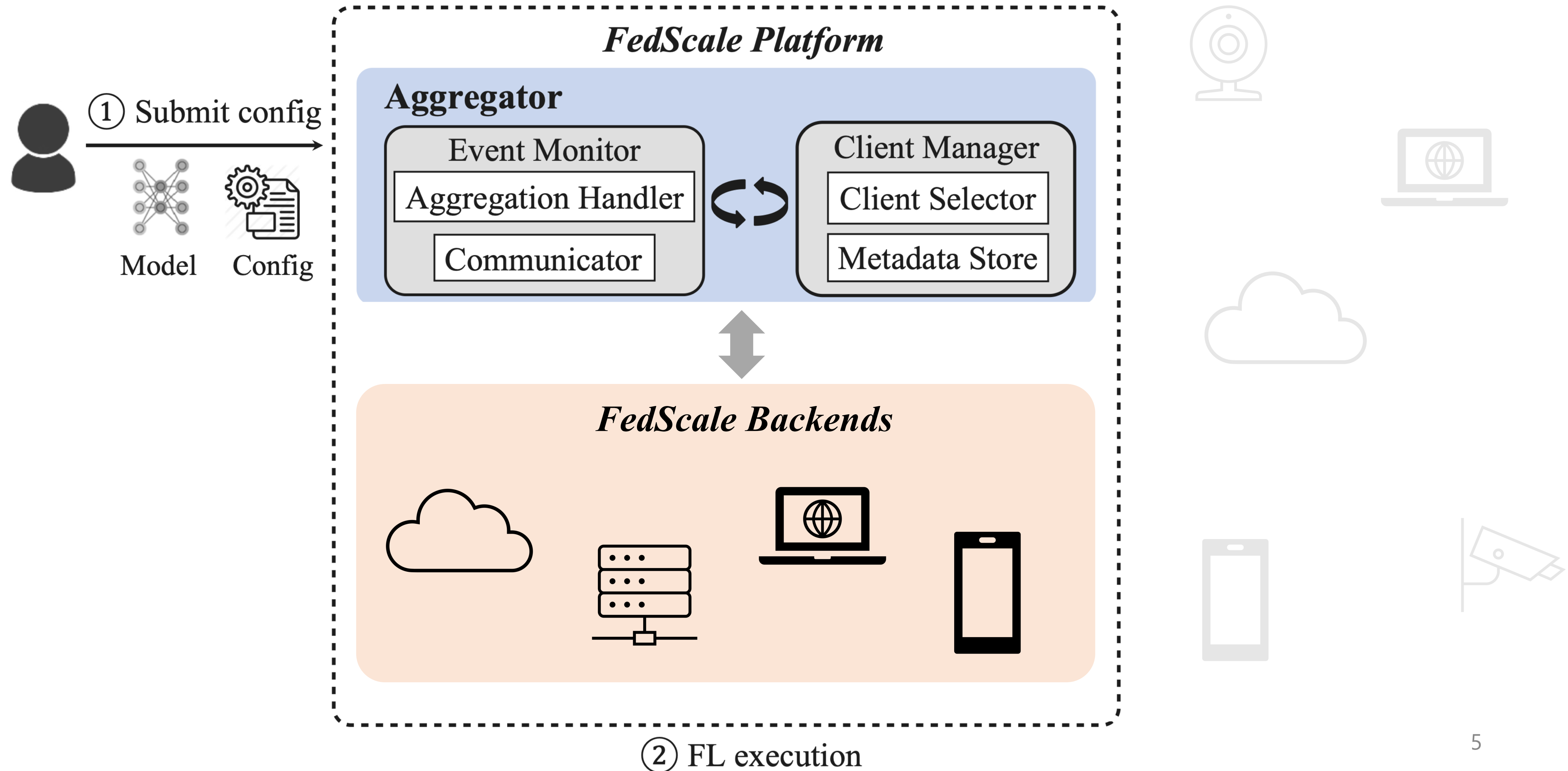


Federated Learning in Existing Benchmarks

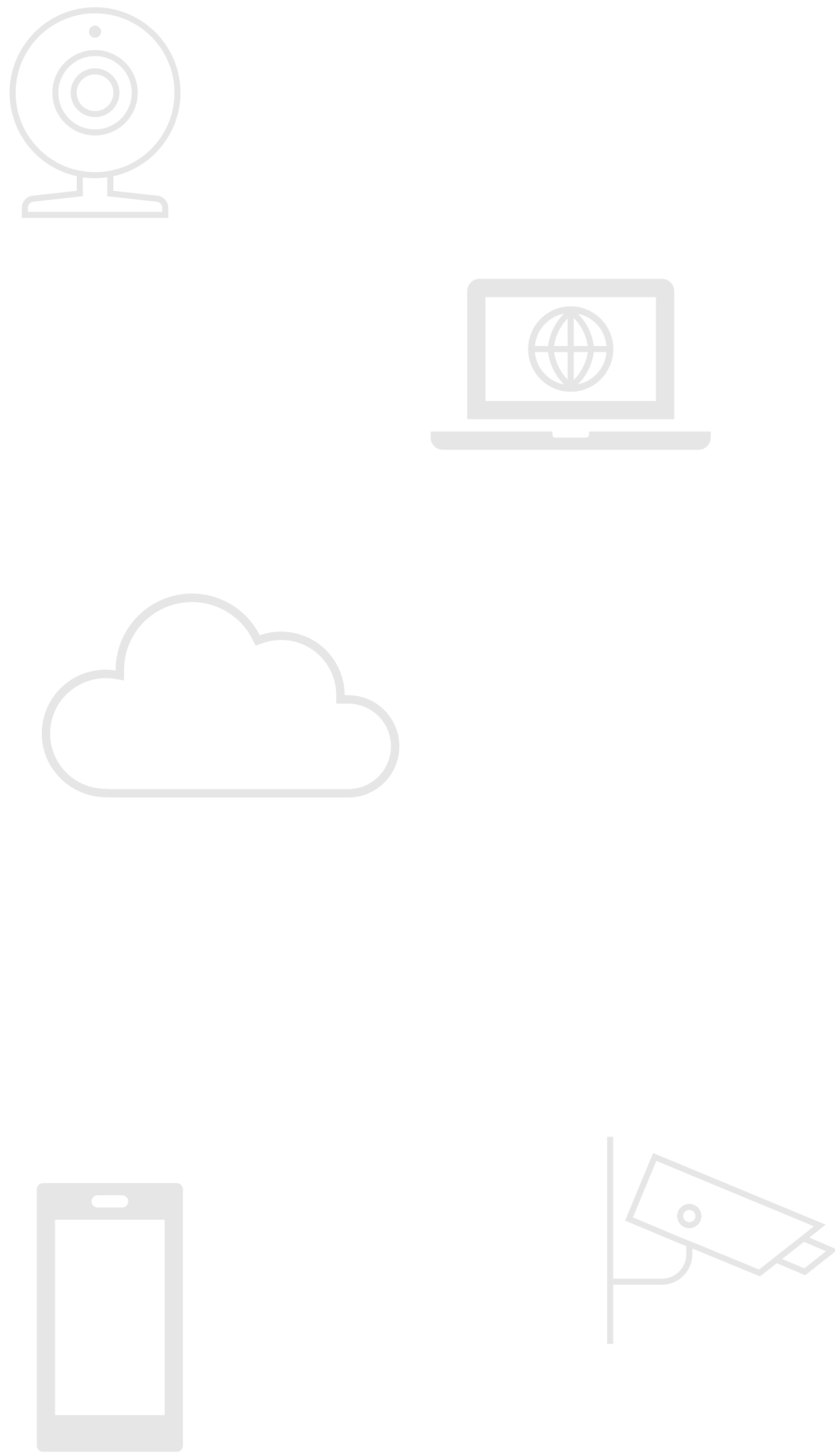
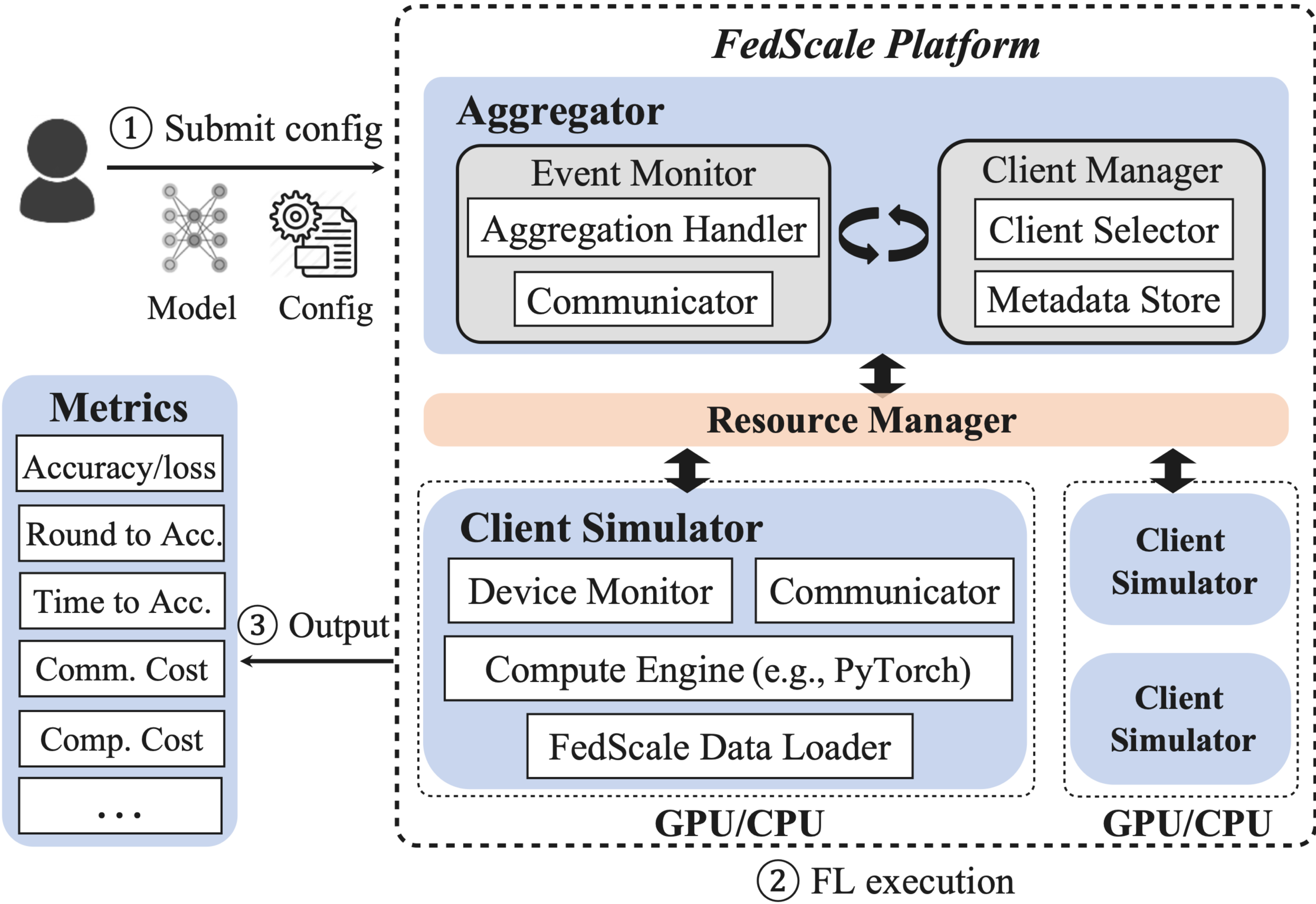
- **Few realistic datasets**
 - Synthesized data (e.g., CIFAR)
- **Missing system details**
 - Hard to evaluate MLSys optimization
- **Suboptimal scalability**
 - Hard to reproduce practical FL scale



FedScale as a Comprehensive Benchmark



FedScale as a Comprehensive Benchmark



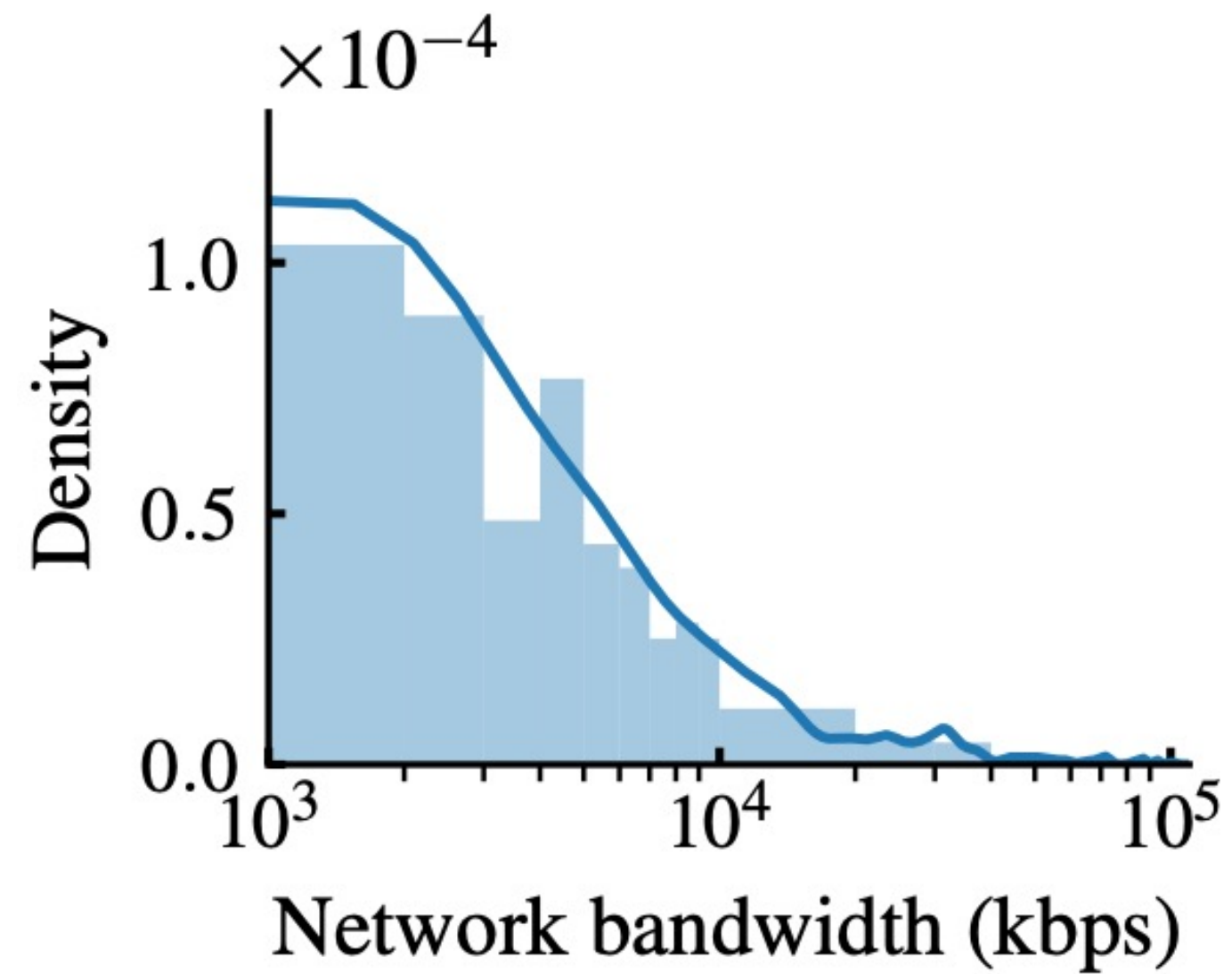
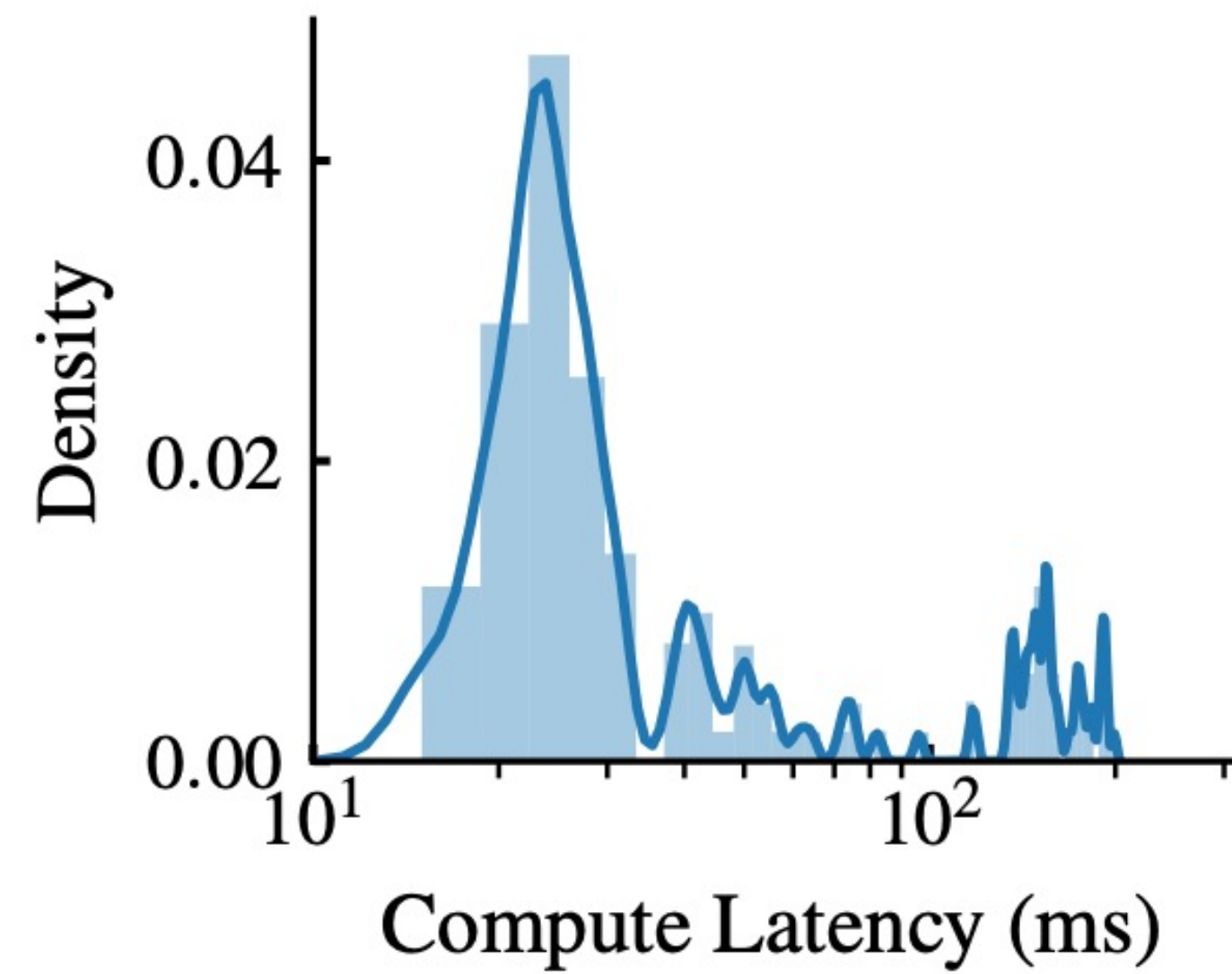
Realistic Client Datasets

- > 20 realistic datasets
 - For CV, NLP, ...
 - Small/Medium/Large scales

Category	Name	Data Type	#Clients	#Instances	Example Task
CV	<i>iNature</i> (ina)	Image	2,295	193K	Classification
	<i>FEMNIST</i> (Cohen et al., 2017)	Image	3,400	640K	Classification
	<i>OpenImage</i> (ope)	Image	13,771	1.3M	Classification, Object detection
	<i>Google Landmark</i> (Weyand et al., 2020)	Image	43,484	3.6M	Classification
	<i>Charades</i> (Sigurdsson et al., 2016)	Video	266	10K	Action recognition
	<i>VLOG</i> (Fouhey et al., 2018)	Video	4,900	9.6K	Classification, Object detection
	<i>Waymo Motion</i> (Ettinger et al., 2021)	Video	496,358	32.5M	Motion prediction
NLP	<i>Europarl</i> (Koehn, 2005)	Text	27,835	1.2M	Text translation
	<i>Blog Corpus</i> (Schler et al., 2006)	Text	19,320	137M	Word prediction
	<i>Stackoverflow</i> (sta)	Text	342,477	135M	Word prediction, Classification
	<i>Reddit</i> (red)	Text	1,660,820	351M	Word prediction
	<i>Amazon Review</i> (McAuley et al., 2015)	Text	1,822,925	166M	Classification, Word prediction
	<i>CoQA</i> (Reddy et al., 2019)	Text	7,189	114K	Question Answering
	<i>LibriTTS</i> (Zen et al., 2019)	Text	2,456	37K	Text to speech
	<i>Google Speech</i> (Warden, 2018)	Audio	2,618	105K	Speech recognition
	<i>Common Voice</i> (com)	Audio	12,976	1.1M	Speech recognition
Misc ML	<i>Taxi Trajectory</i>	Text	442	1.7M	Sequence prediction
	<i>Taobao</i> (tao)	Text	182,806	20.9M	Recommendation
	<i>Puffer Streaming</i> (Yan et al., 2020)	Text	121,551	15.4M	Sequence prediction
	<i>Fox Go</i> (go-)	Text	150,333	4.9M	Reinforcement learning

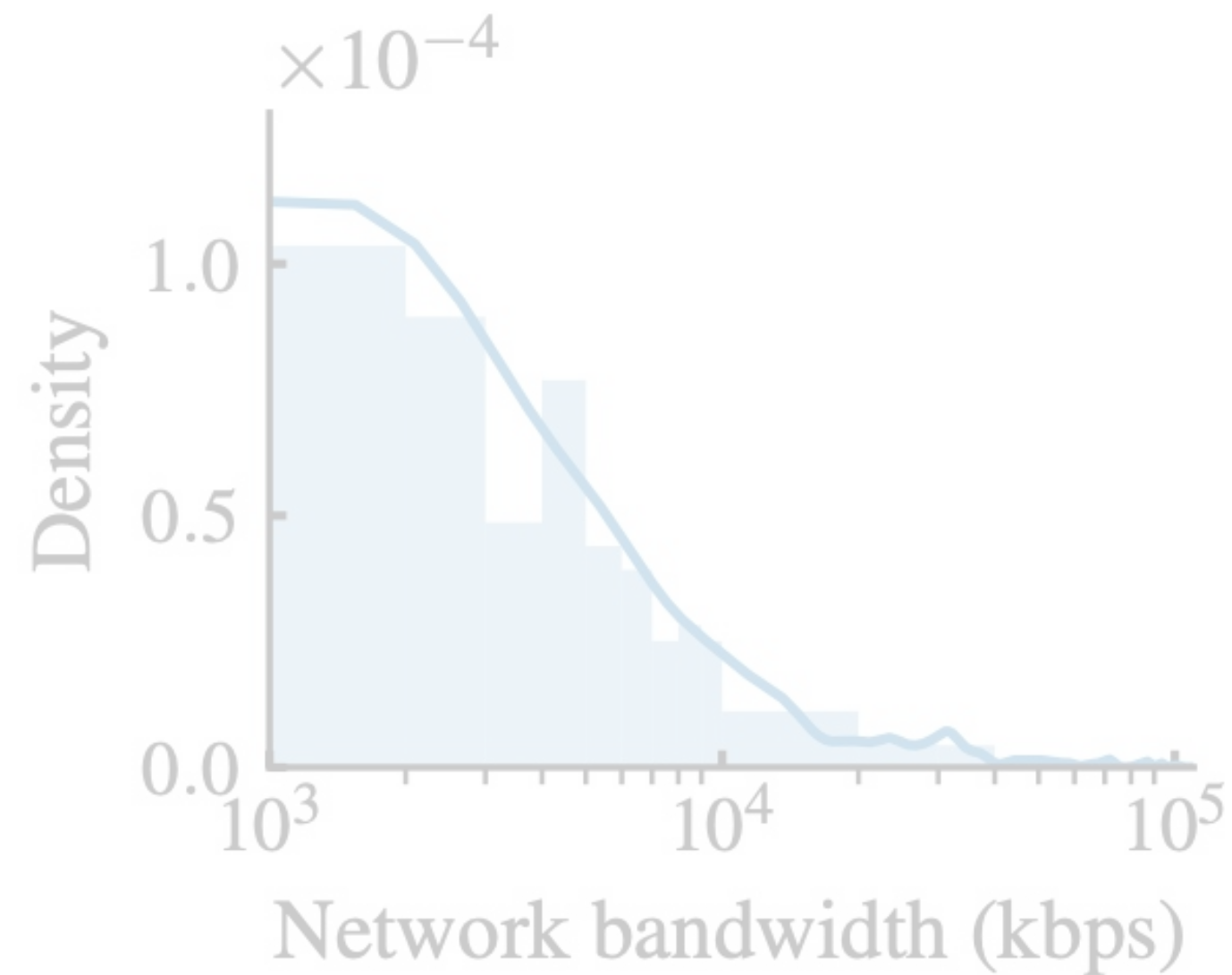
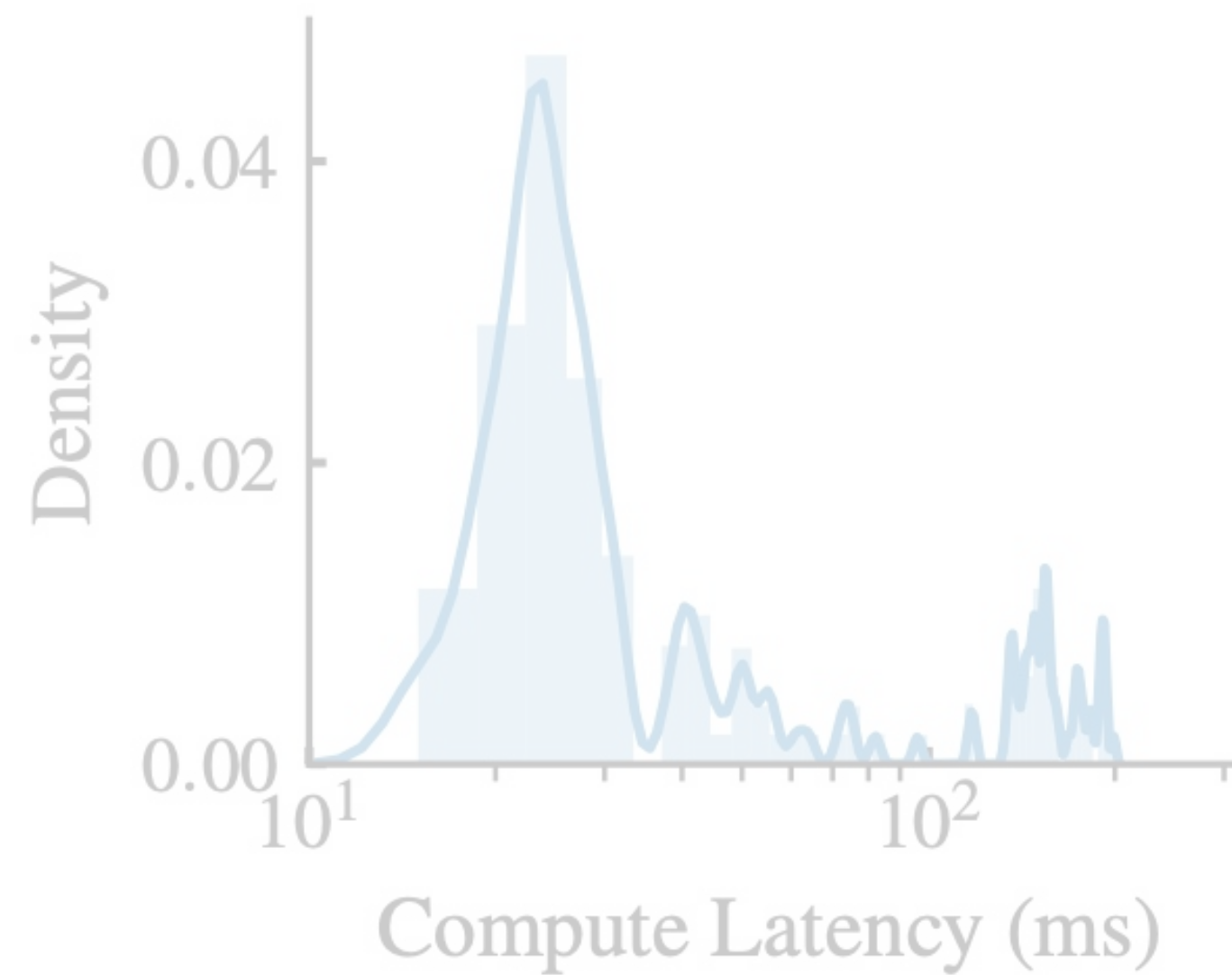
Some FedScale Datasets

Millions of Client System Traces

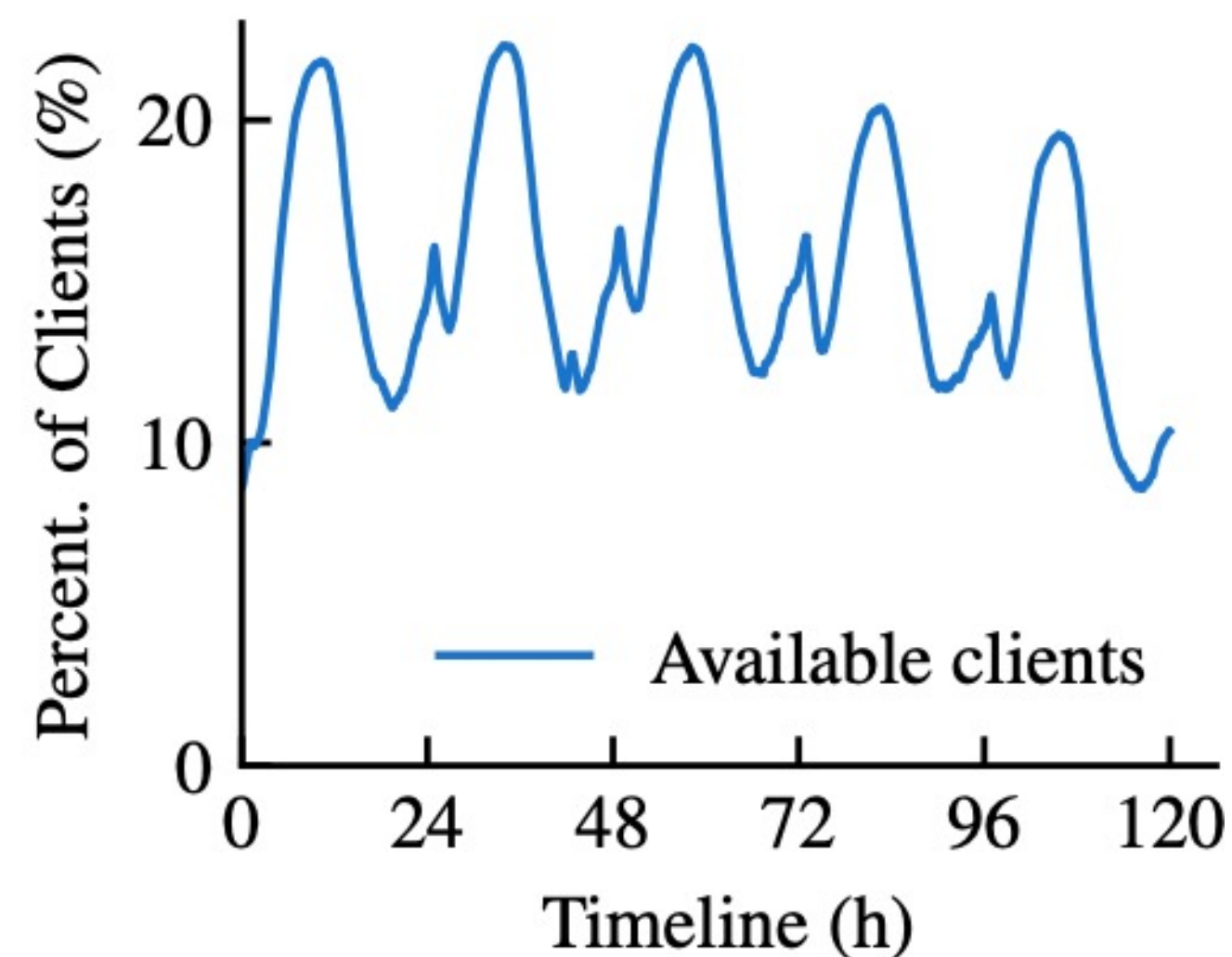
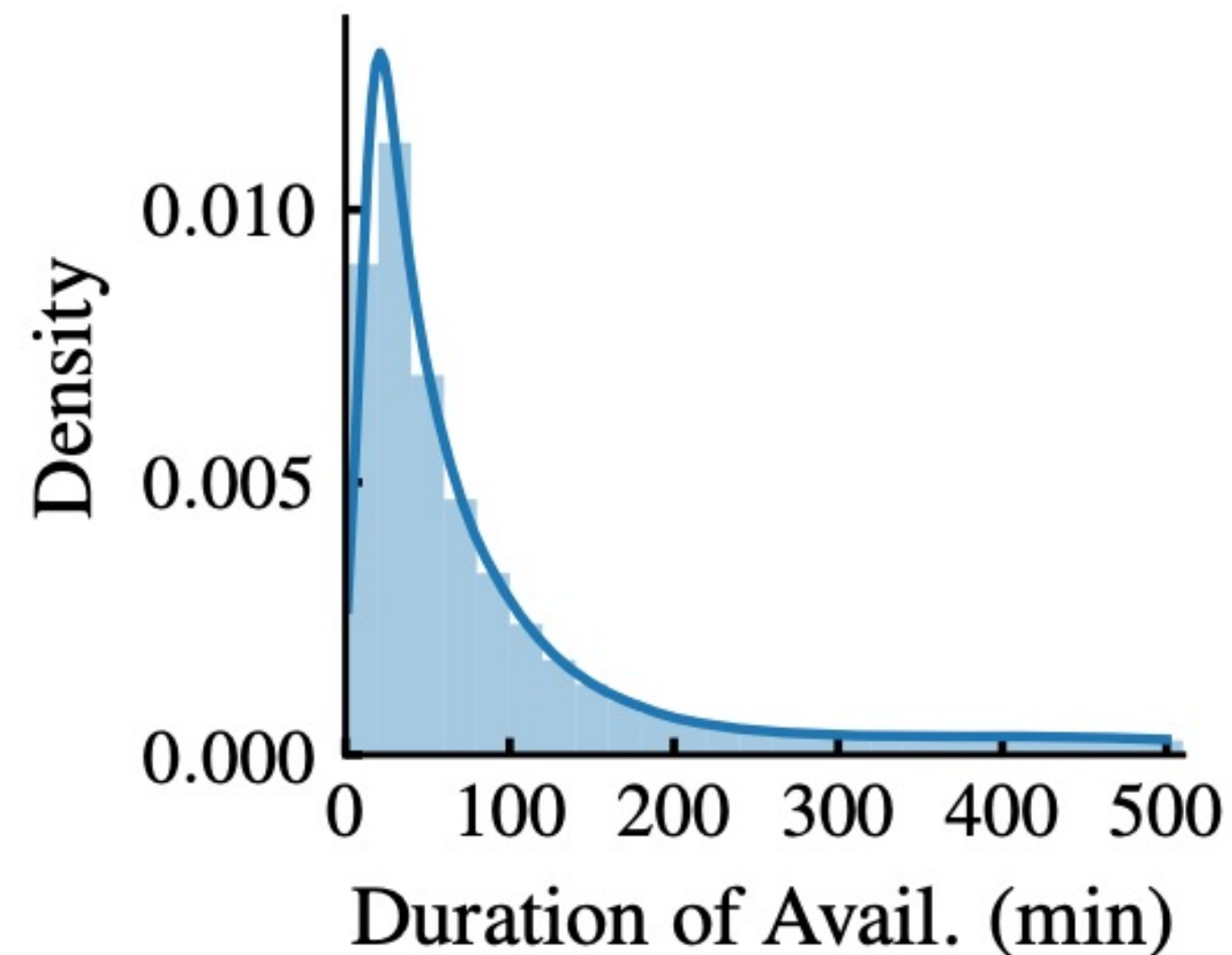


Heterogeneous computation/
communication speed

Millions of Client System Traces

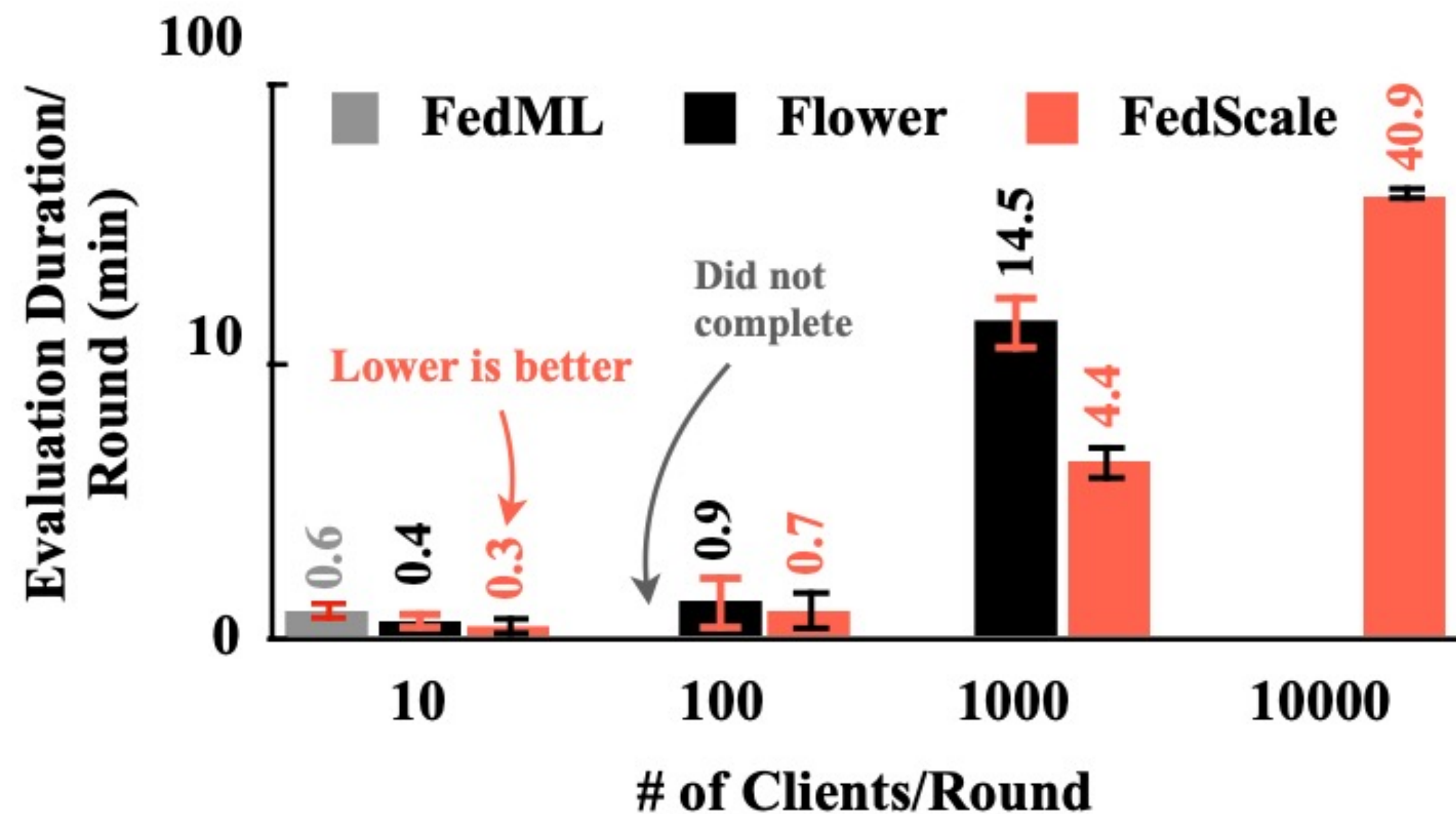


Heterogeneous computation/
communication speed



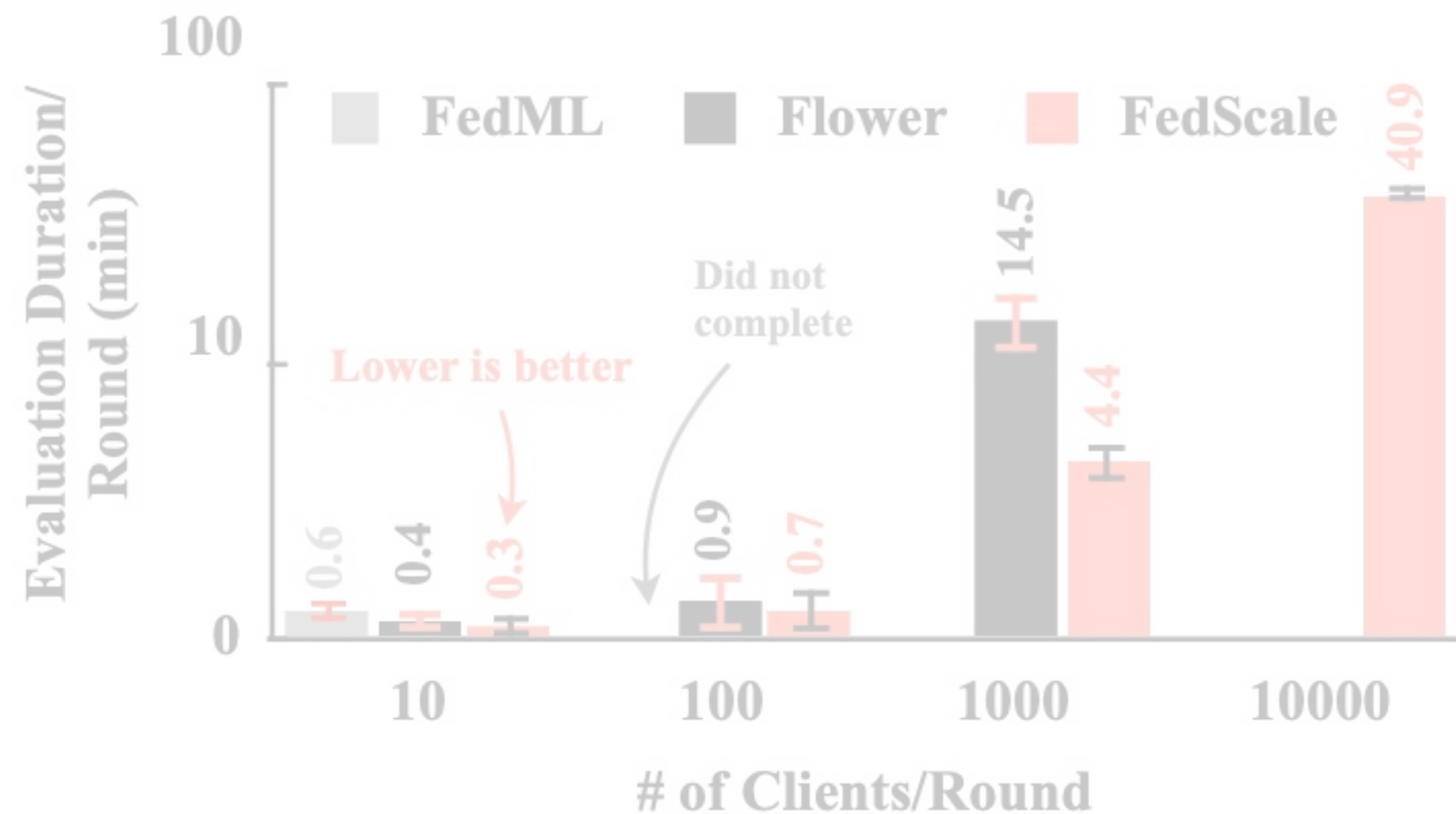
Dynamics of client availability
in the wild

Scalable & Extensible Runtime



FedScale supports **orders-of-magnitude** more clients on the same underlying cluster

Scalable & Extensible Runtime



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```
from fedscale.core.client import Client

class Customized_Client(Client):
    # Redefine training (e.g., for local
    #   SGD/gradient compression)

    def train(self, client_data, model, conf):
        # Code of plugin
        ...

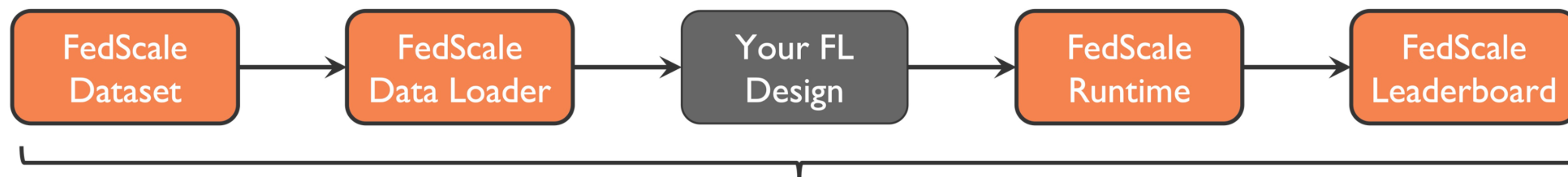
        # Results will be serialized, and
        #   then sent to aggregator
        return training_result
```

Implementing new FL designs only takes
a **few lines of code**

FedScale.ai

FL Benchmark & Platform:

- 20+ realistic datasets
- 70+ models
- Up to $O(10k)$ clients/round
- Sync/Async training mode
- Easy extension
- On-device deployment
- Practical FL runtime



FedScale automates FL benchmarking