# **Towards Federated Foundation Models**

Scalable Pipelines for Group-Structured Learning

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## Federated learning research has a small data problem

#### Research datasets for FL are often:

- Small
- Difficult to create/customize
- Unsuitable for foundation models, especially LLMs

#### Need for large-scale, group-structured datasets:

+ scalable, flexible and efficient pipelines

#### Dataset Grouper

Library for creating group-structured datasets.

- Scalable: can handle millions of clients 🔽
- Flexible: any custom partition function on any TFDS/HuggingFace dataset
- Platform-agnostic: works with TF, PyTorch, JAX, NumPy, ... 🗸

# Federated training of O(100M) and O(1B) parameter models

What happens in FL at LLM-scale?

- FedSGD vs. FedAvg
- Global vs. local performance

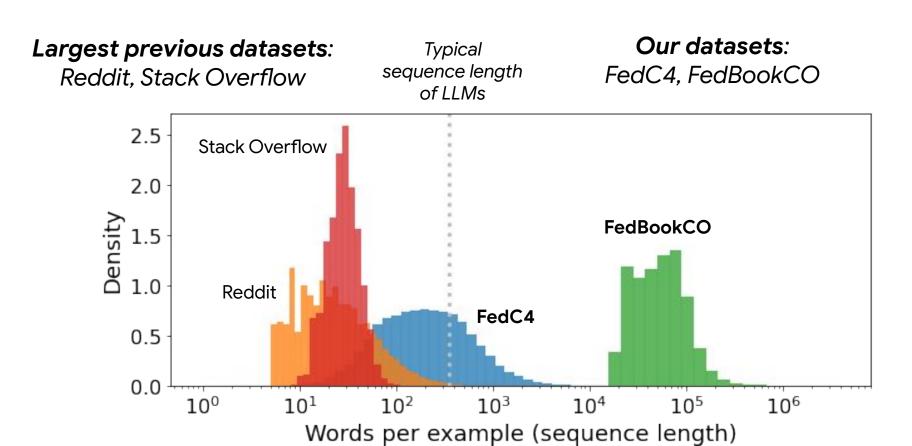
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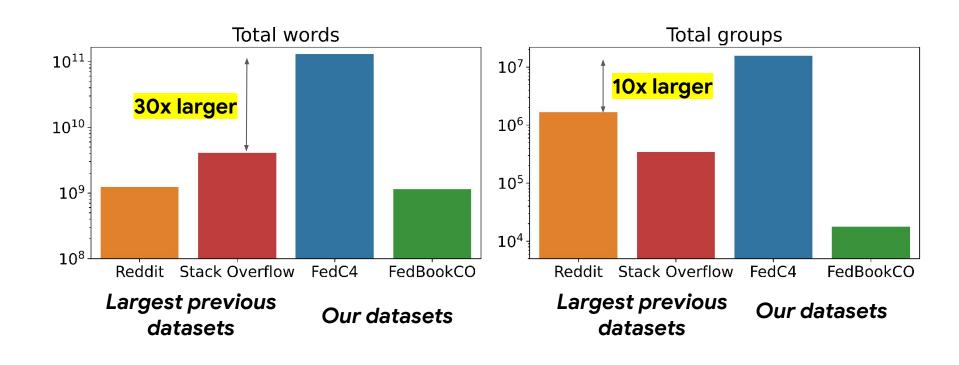
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**Scalable:** largest federated datasets to-date

### New federated LLM datasets: longer sequences



### New federated LLM datasets: more words & groups



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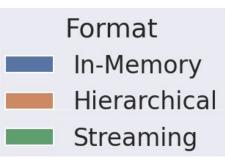
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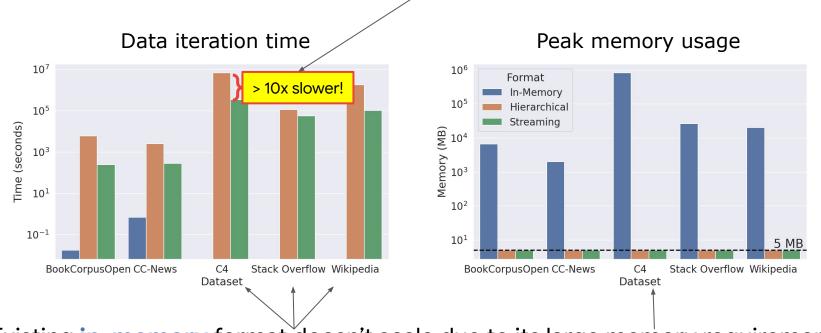
Scalable: fast data iterators

# Scalable streaming data loaders

Centralized → federated learning

Existing hierarchical format is much slower





Existing in-memory format doesn't scale due to its large memory requirement

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Flexible partitioning of existing datasets

```
import dataset_grouper as dsgp
import tensorflow_datasets as tfds
```

#### Load any TFDS/HuggingFace dataset

```
dataset_builder = tfds.builder("mnist")
```

#### Any user-defined partition function

```
def get_label_fn(x):
   label = x["label"].numpy()
   return str(label).encode("utf-8")
```

```
import dataset_grouper as dsgp
import tensorflow_datasets as tfds
```

```
with beam.Pipeline() as root:
    mnist_pipeline(root)
```

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**Platform-agnostic** group iterators

#### Load a partitioned dataset

#### **Platform-agnostic iterators**

```
for client_dataset in partitioned_dataset.build_group_stream():
    # client_dataset is an iterable of examples.
    for example in client_dataset.as_numpy_iterator():
        # Process this example.
```

**Empirical investigations** 

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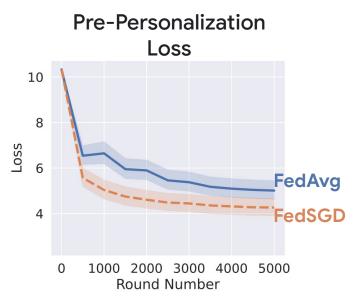
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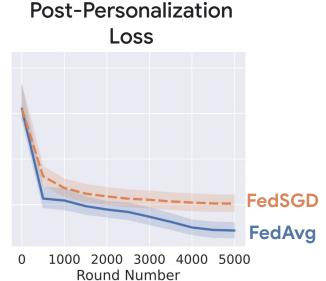
# FedAvg is a meta-learner!

Model: 128M param LM

**Train:** FedC4

**Eval:** FedBookCO





FedSGD learns a better global model than FedAvg

FedAvg learns a model that personalizes better than FedSGD

# Thank you!

https://github.com/google-research/dataset\_grouper Pull requests welcome!

pip install dataset-grouper

# Thank you!