



DistTrain: Addressing Model and Data Heterogeneity with Disaggregated Training for Multimodal Large Language Models

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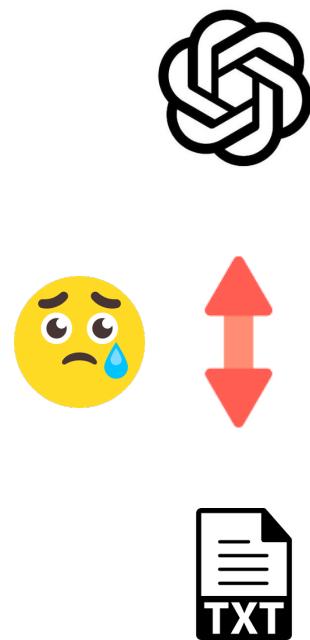
Zheng Ge, Yibo Zhu, Xin Jin



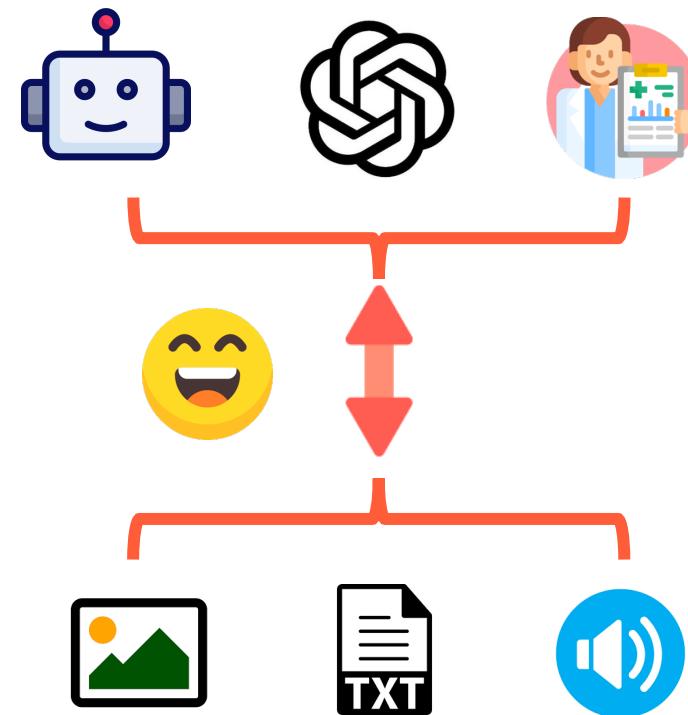
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Multimodal LLM



Unimodal

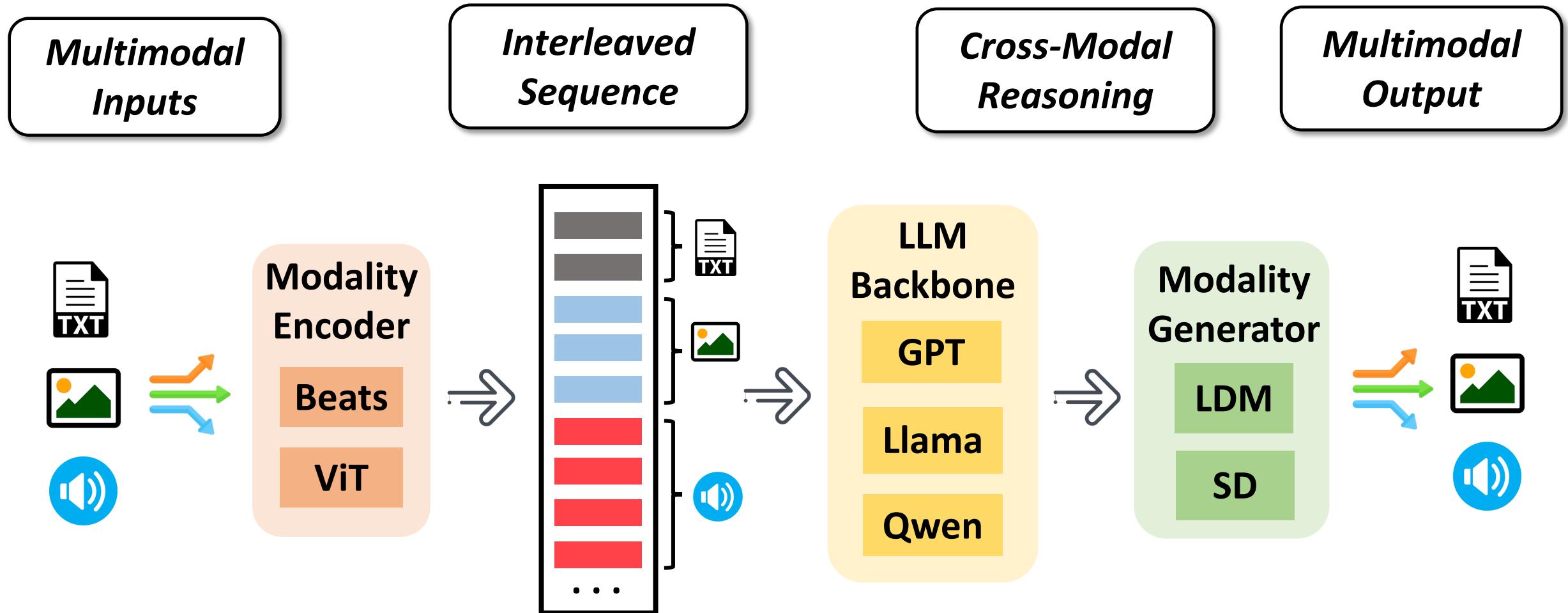


Multimodal

- *natural language*
- *code generation*

- *natural language, code*
- *photos, diagrams, charts*
- *speech, music, environmental sounds*

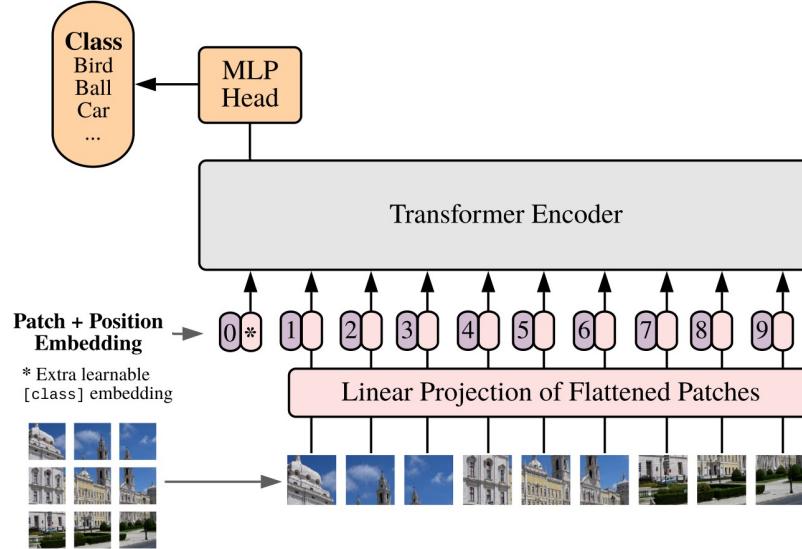
Multimodal LLM



Model Heterogeneity

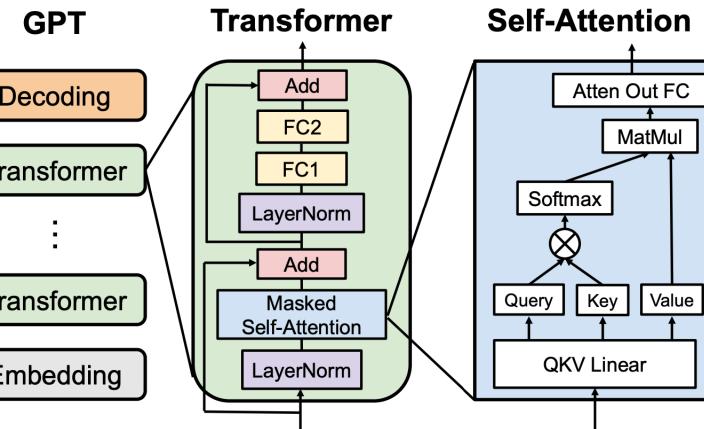
Modality Encoder

Vision Transformer (ViT)



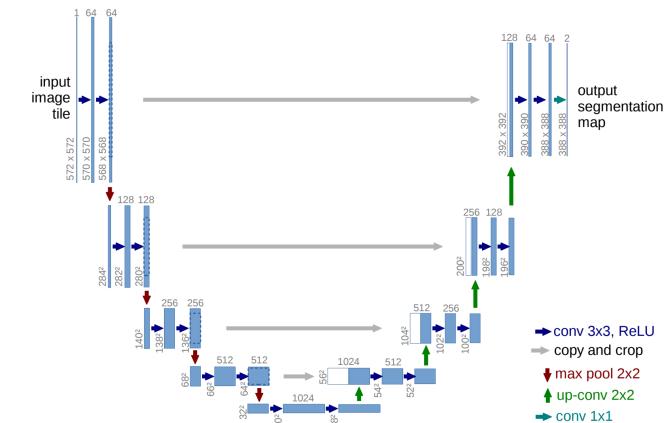
Lightweight Transformer

LLM Backbone



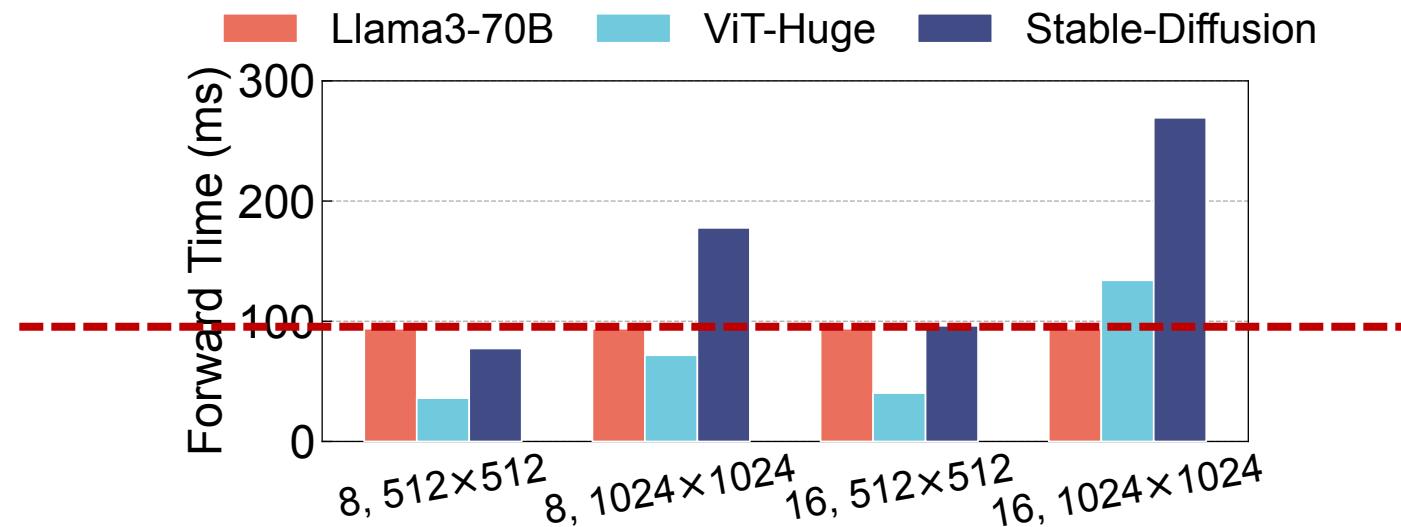
Heavyweight Transformer

Modality Generator

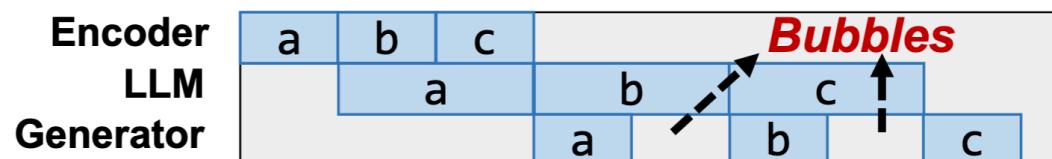


Convolutional U-Net

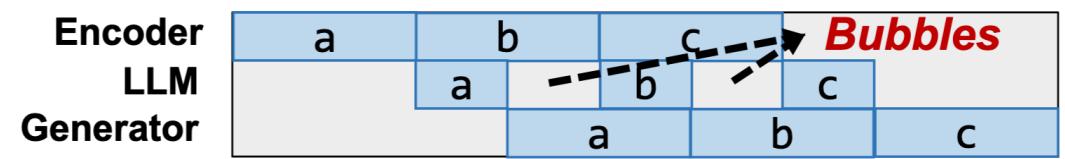
Model Heterogeneity



LLM Backbone is the PP straggler



Modality encoder/generator is the PP straggler



Data Heterogeneity

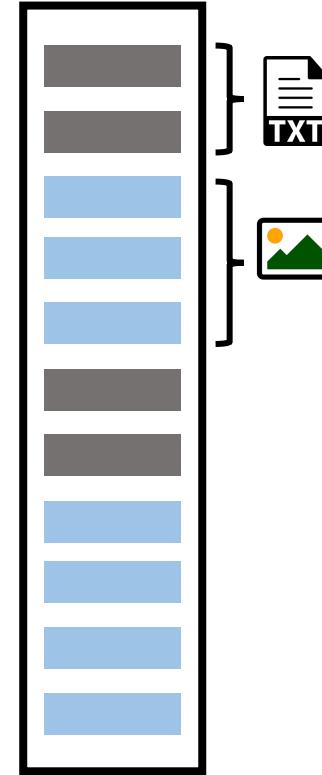
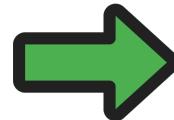
*What is the landmark, and what plants surround it?
Beside, remove the plants.*



*The landmark is the Great Wall, surrounded by flowers.
Here is the edited image.*

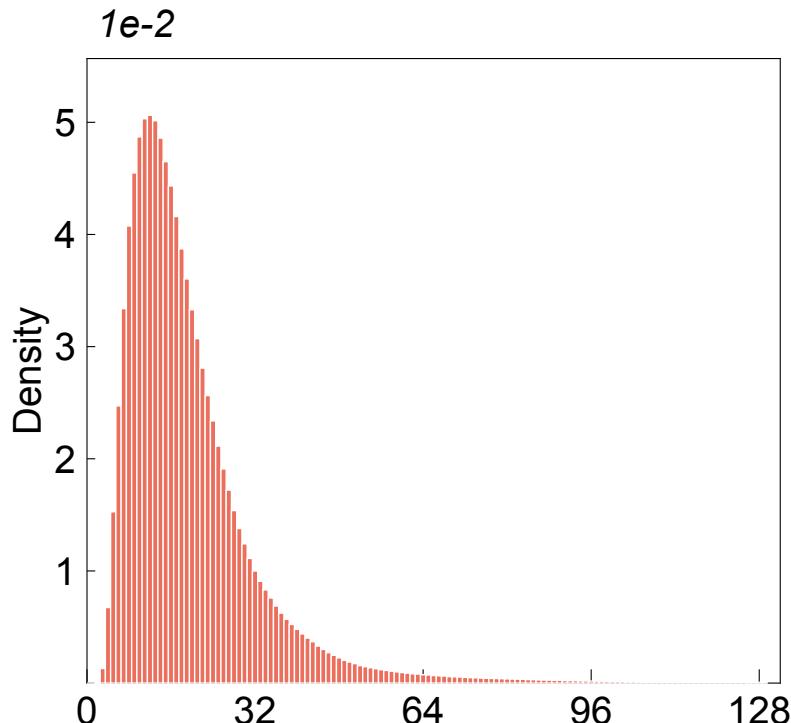


Training Sequence

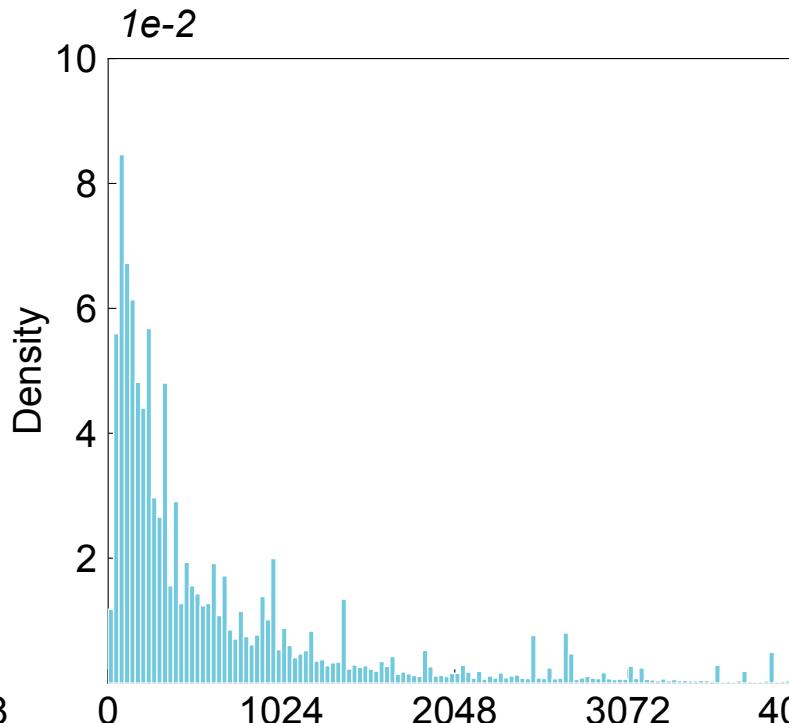


Data Heterogeneity

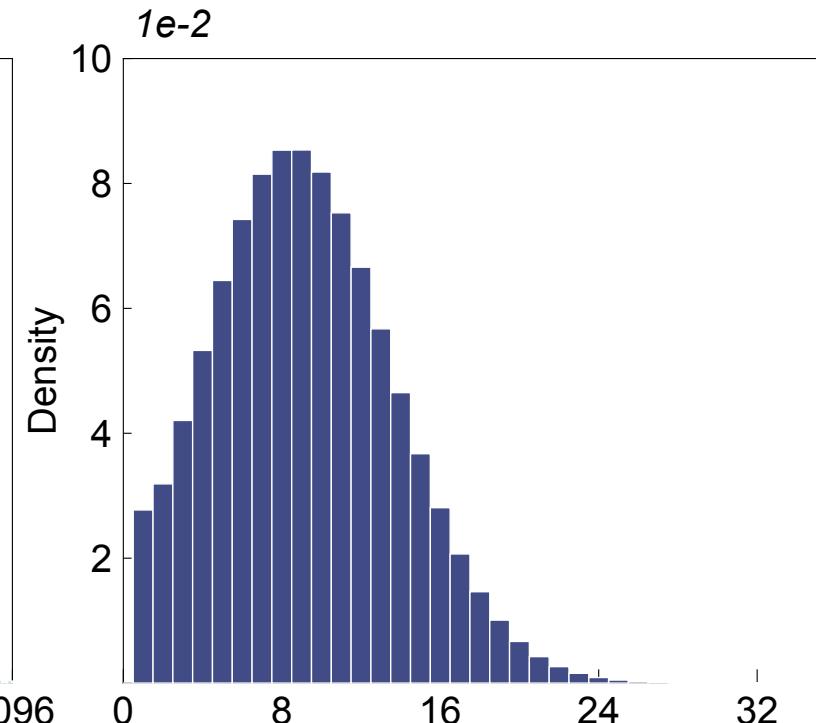
The sizes of text and image subsequences display highly skewed distributions



*# of tokens for one
text subsequence*



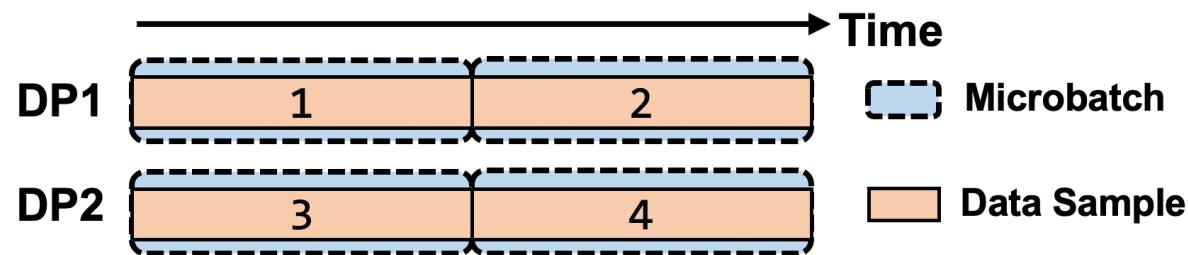
*# of tokens for one
image subsequence*



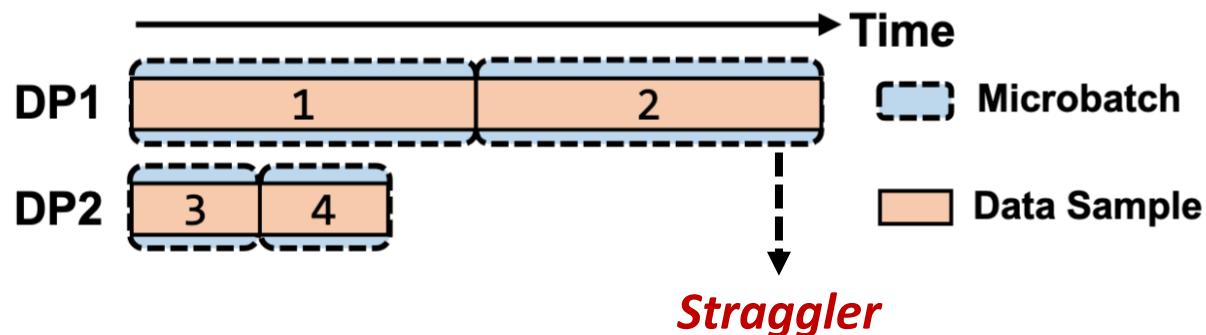
*# of image subsequence
in one 8K sequence*

Data Heterogeneity

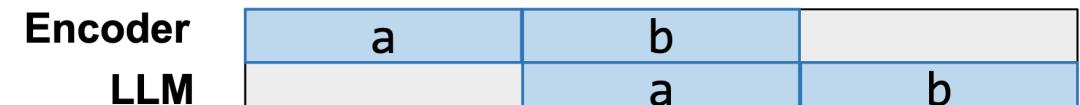
Intra-microbatch straggler



with data heterogeneity



Inter-microbatch straggler



with data heterogeneity



DistTrain design outline

Challenge 1: How to solve the model heterogeneity?

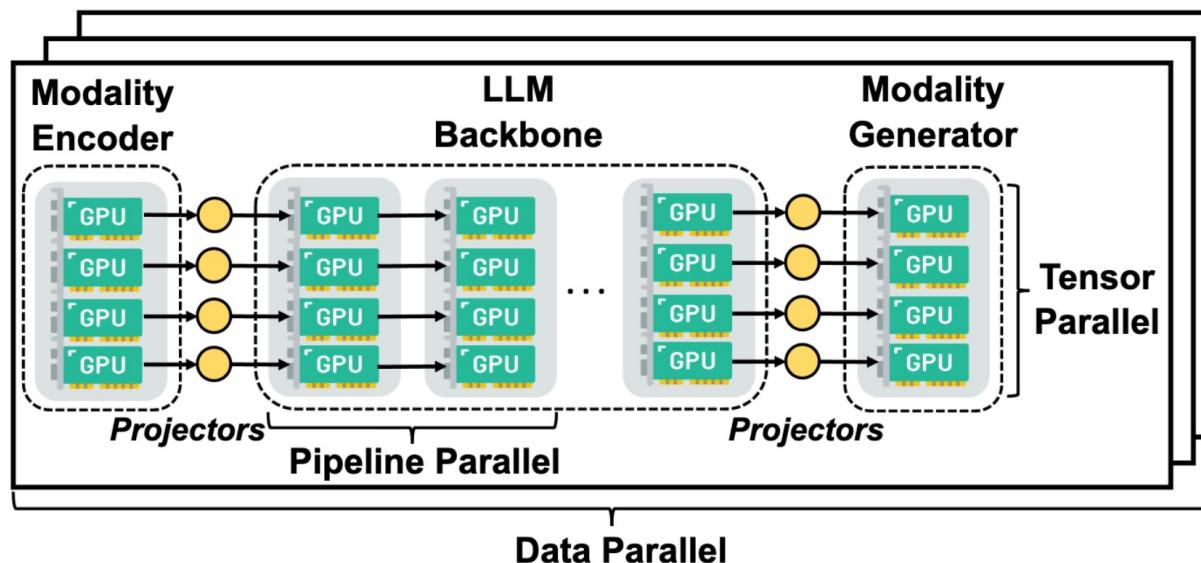
- Disaggregated model orchestration → Assign customized parallelism strategies to different modules
- Optimization formulation → Derive optimal parallelism configurations

Challenge 2: How to solve the data heterogeneity?

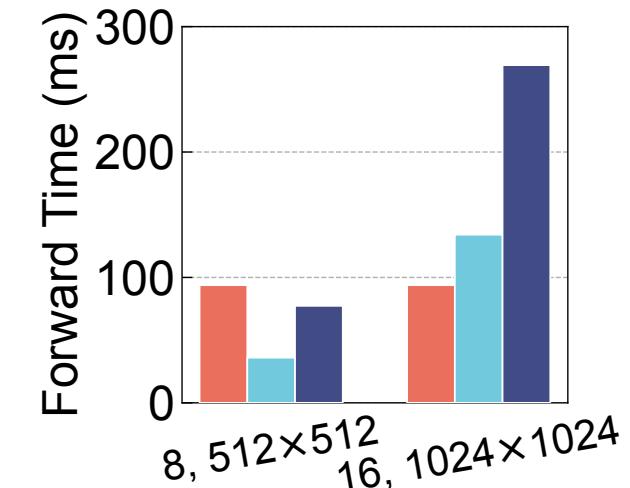
- Intra-microbatch reordering → Mitigate stragglers across data-parallel instances
- Inter-microbatch reordering → Mitigate stragglers within pipeline parallelism

Addressing Model Heterogeneity

The monolithic orchestration incurs severe imbalance across modules 😢

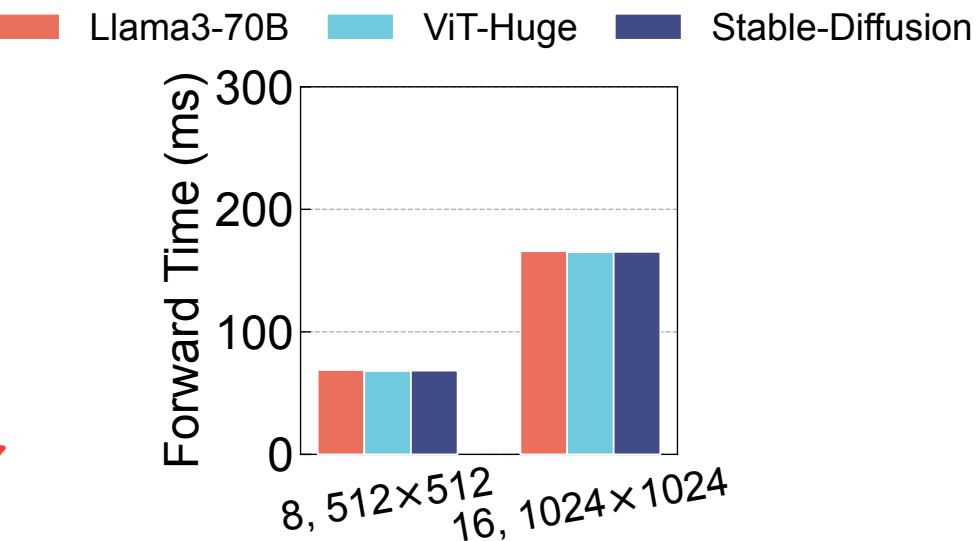
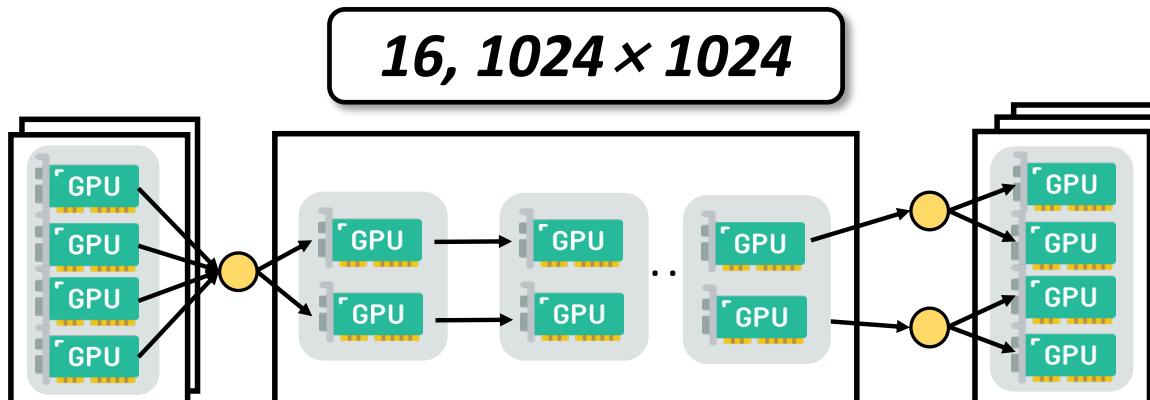
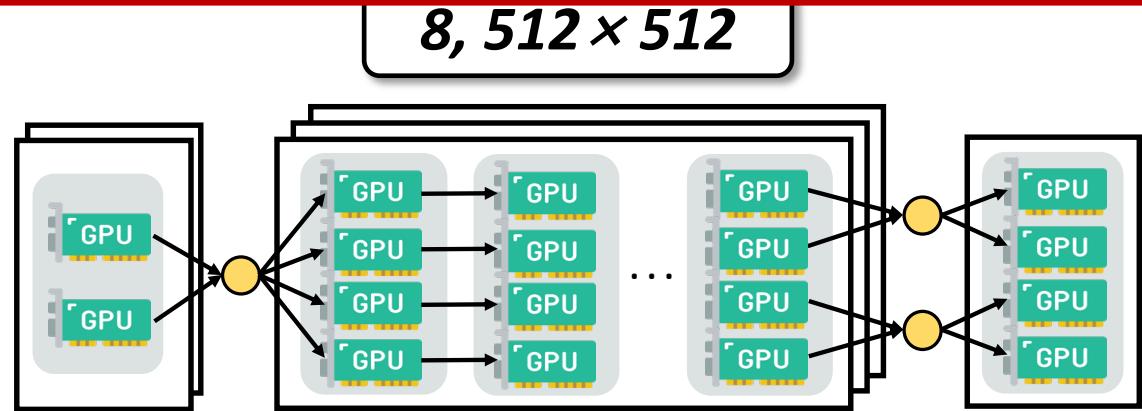


■ Llama3-70B ■ ViT-Huge ■ Stable-Diffusion



Addressing Model Heterogeneity

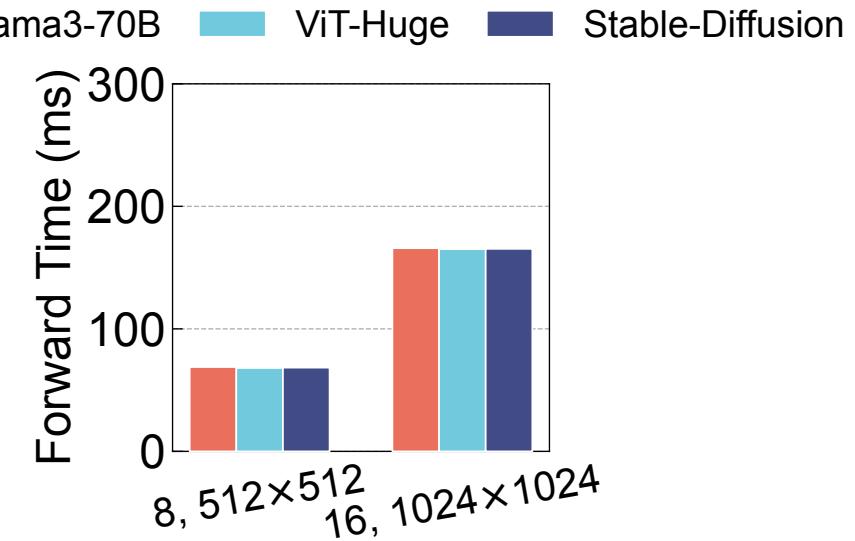
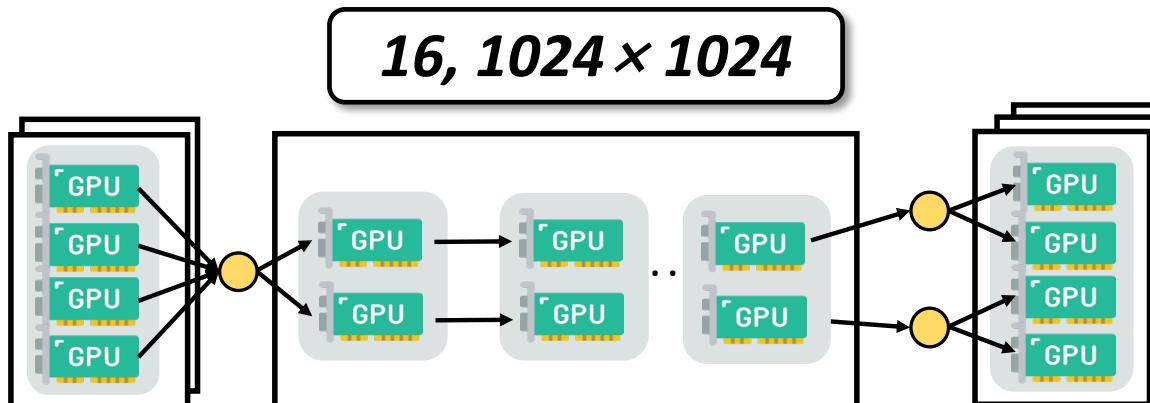
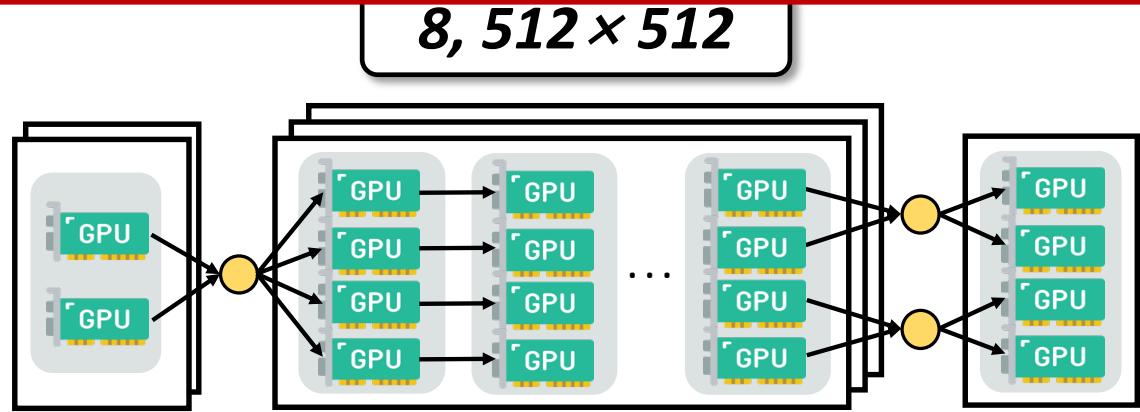
Our disaggregated orchestration balances computational loads 😊



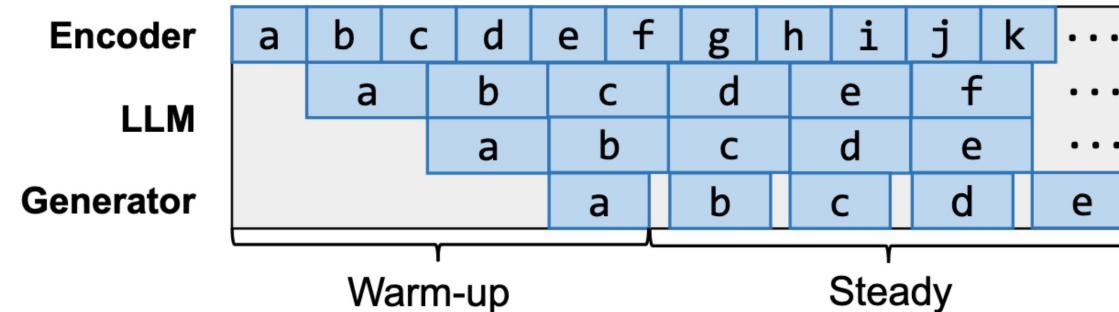
How to decide the detailed disaggregated parallelism strategy?

Addressing Model Heterogeneity

How to decide the detailed disaggregated parallelism strategy?



Addressing Model Heterogeneity



$$T_{warmup} = M \times C_{lm}(TP_{lm}) + \frac{DP_{lm} \times M}{DP_{me}} \times C_{me}(TP_{me}) \\ + \frac{DP_{lm} \times M}{DP_{mg}} \times C_{mg}(TP_{mg})$$

Time for the first microbatch
to complete

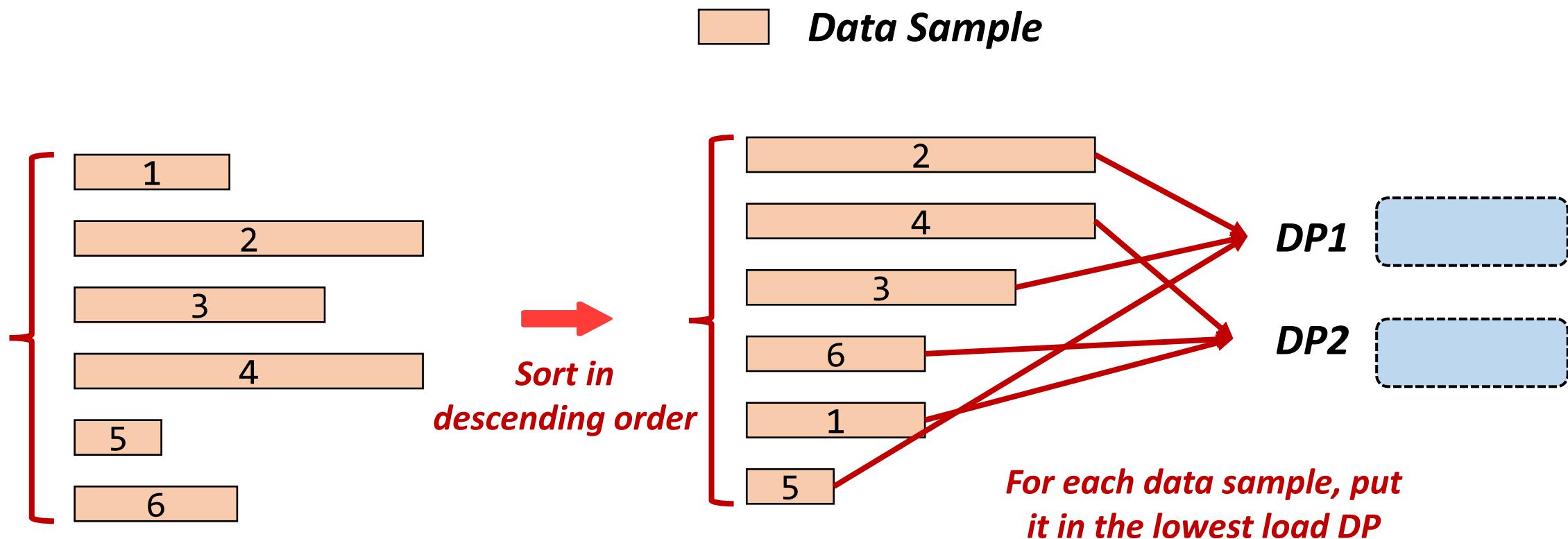
$$T_{steady} = \max \left\{ \begin{array}{l} \frac{DP_{lm} \times TP_{lm} \times M}{y} \times C_{lm}(TP_{lm}), \\ \frac{DP_{lm} \times TP_{me} \times M}{z} \times C_{me}(TP_{me}), \\ \frac{DP_{lm} \times TP_{mg} \times M}{x} \times C_{mg}(TP_{mg}) \end{array} \right\} \times \left(\frac{BS}{DP_{lm} \times M} - 1 \right)$$

Time of the slowest
pipeline stage

of
microbatches

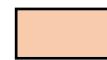
Addressing Data Heterogeneity

Intra-microbatch reordering



Addressing Data Heterogeneity

Intra-microbatch reordering



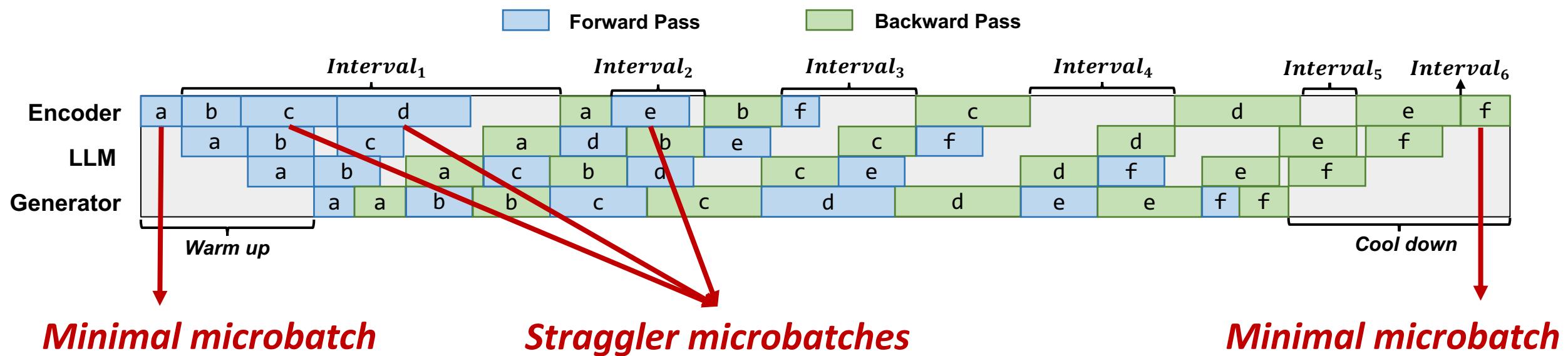
Data Sample



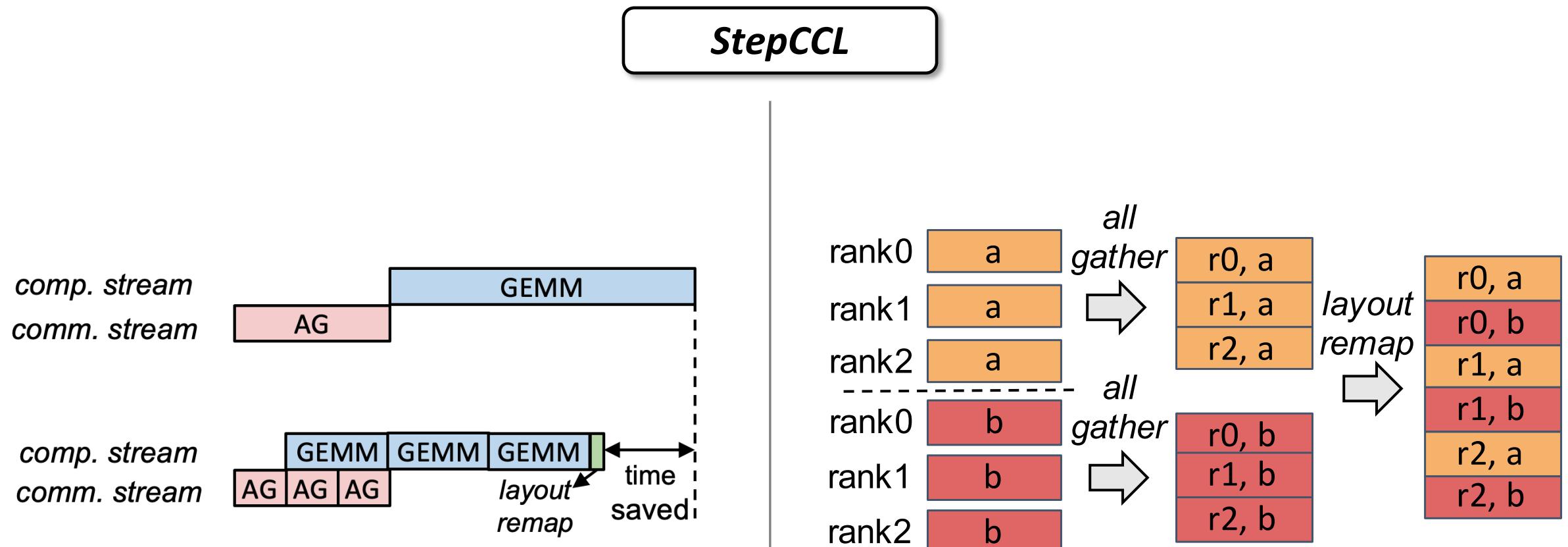
Addressing Data Heterogeneity

Inter-microbatch reordering

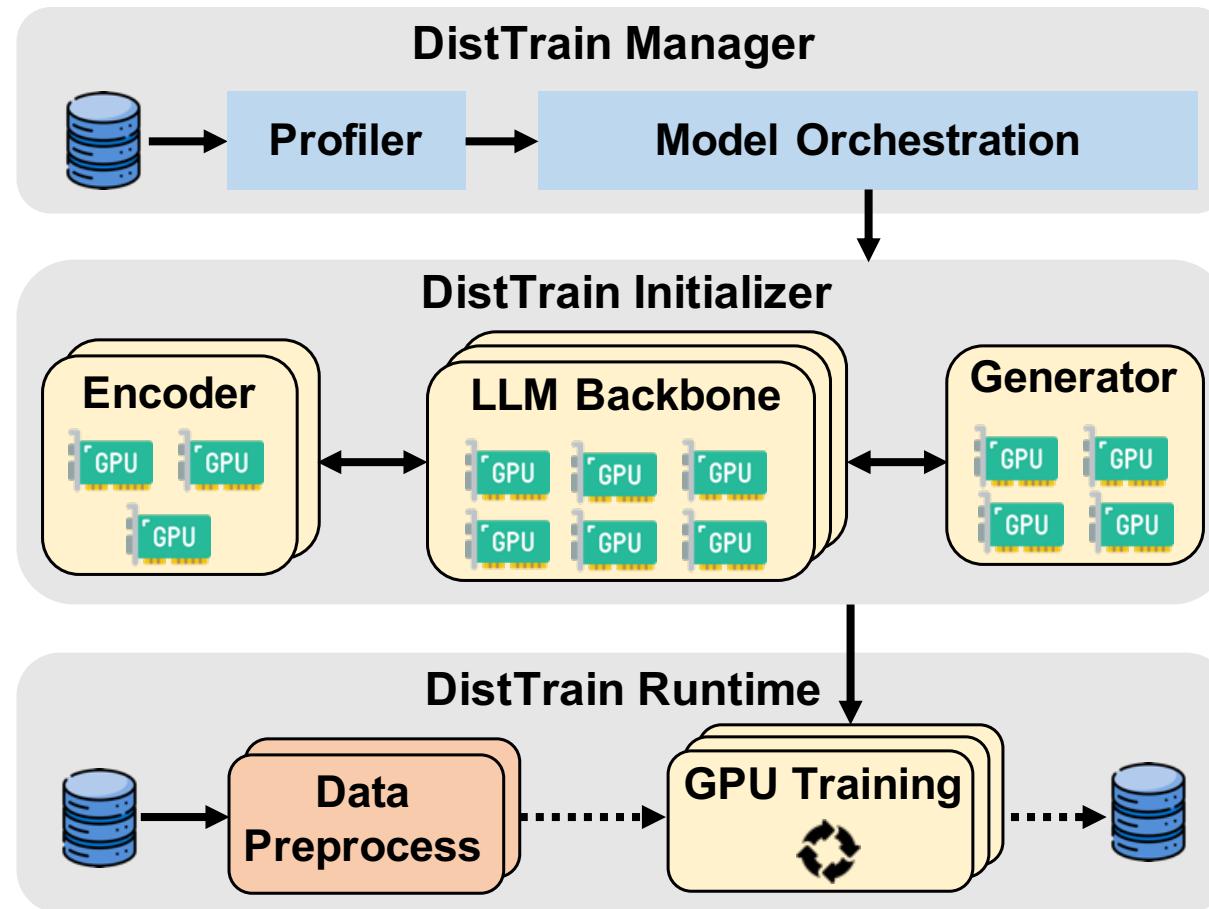
Hide the straggler within the large intervals in 1F1B pipeline



DistTrain System



DistTrain System



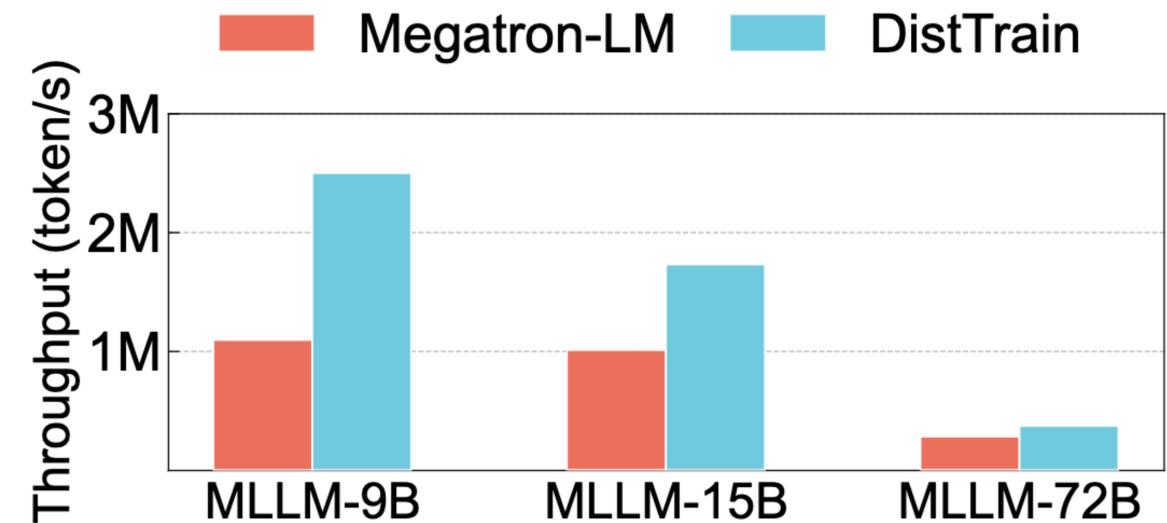
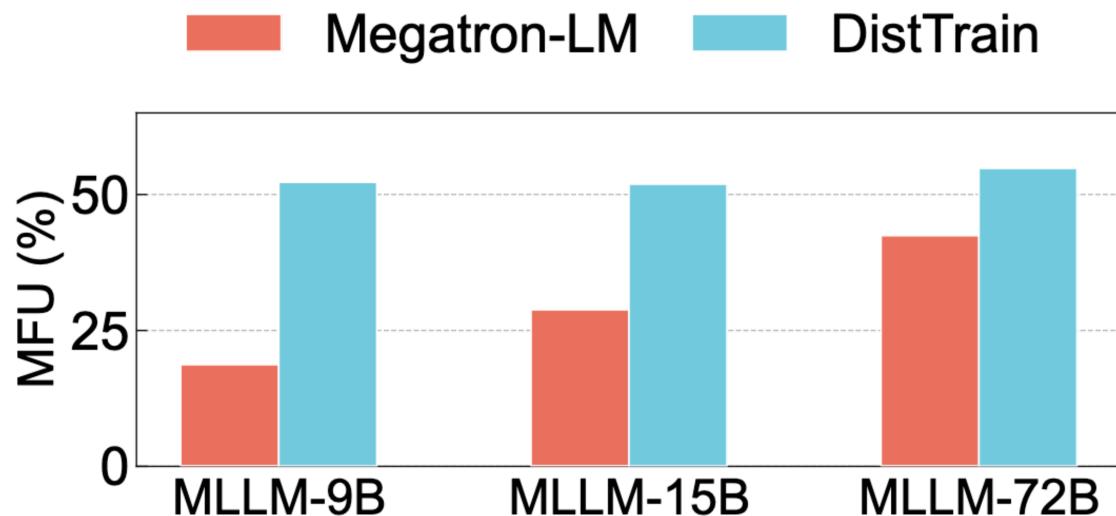
Evaluation

- Setup
 - Hardware: Production cluster with 1,296 A800 GPUs
 - Network: NVLink and RDMA based on RoCEv
 - Dataset: Production training data from LAION
 - Model: MLLM (Llama + ViT + Stable-Diffusion)

Models	# of Layers	Hidden Size	FFN Hidden Size	# of Heads	# of Groups
Llama3-7B	32	4096	11008	32	32
Llama3-13B	40	5120	13824	40	40
Llama3-70B	80	8192	28672	64	8

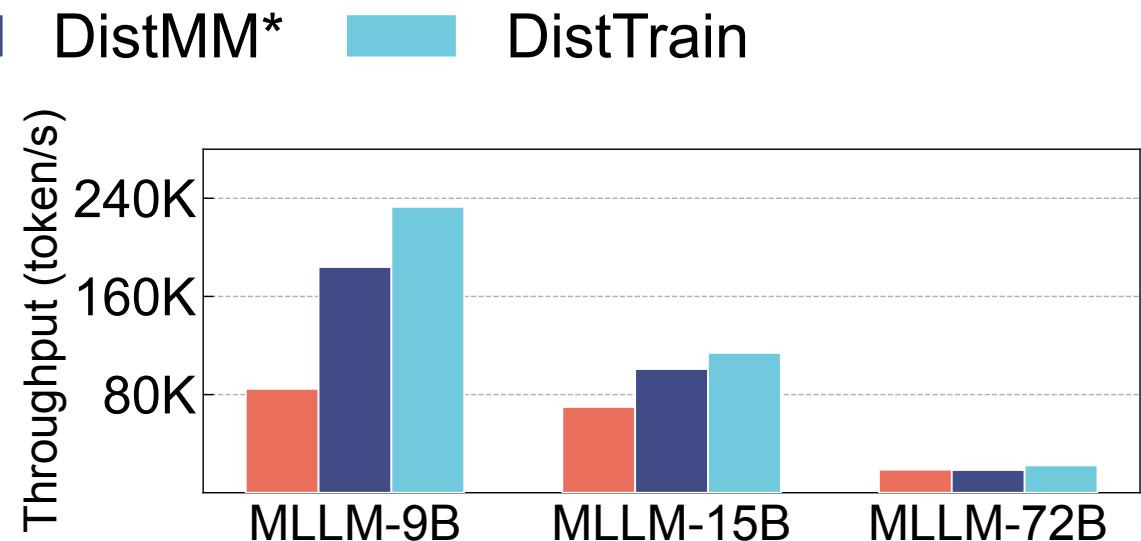
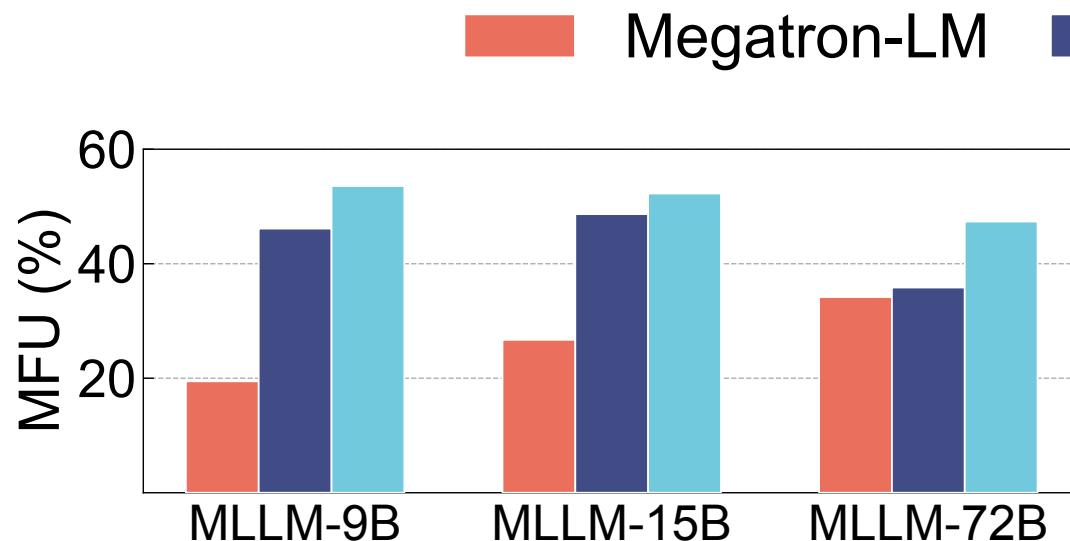
Evaluation

- DistTrain outperforms Megatron-LM on large-scale MLLM training



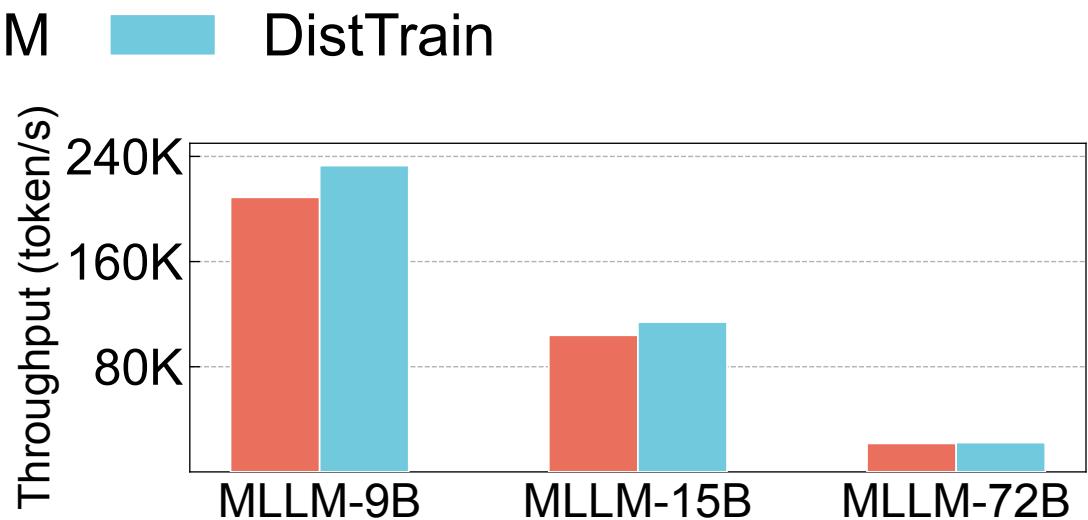
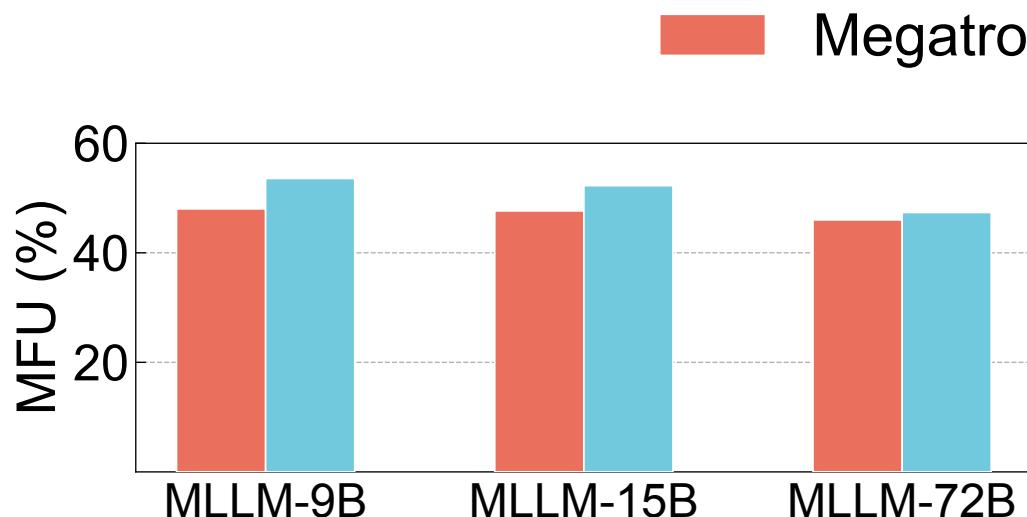
Evaluation

- DistTrain's disaggregated model orchestration is effective



Evaluation

- DistTrain's disaggregated data processing (reordering) is effective



Evaluation

- Performance under module freezing
- Performance of StepCCL
- System overhead of DistTrain

Conclusion

- Multimodal LLM training suffers from **model** and **data heterogeneity**
- DistTrain introduces two techniques:
 - *Disaggregated model orchestration* → tackles model heterogeneity
 - *Disaggregated data preprocessing* → tackles data heterogeneity
- DistTrain outperforms Megatron-LM by up to $2.2\times$ on training throughput and achieves 54.7% MFU
- DistTrain supports large-scale open-sourced [MLLM](#) (130B) training

Thank you!

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