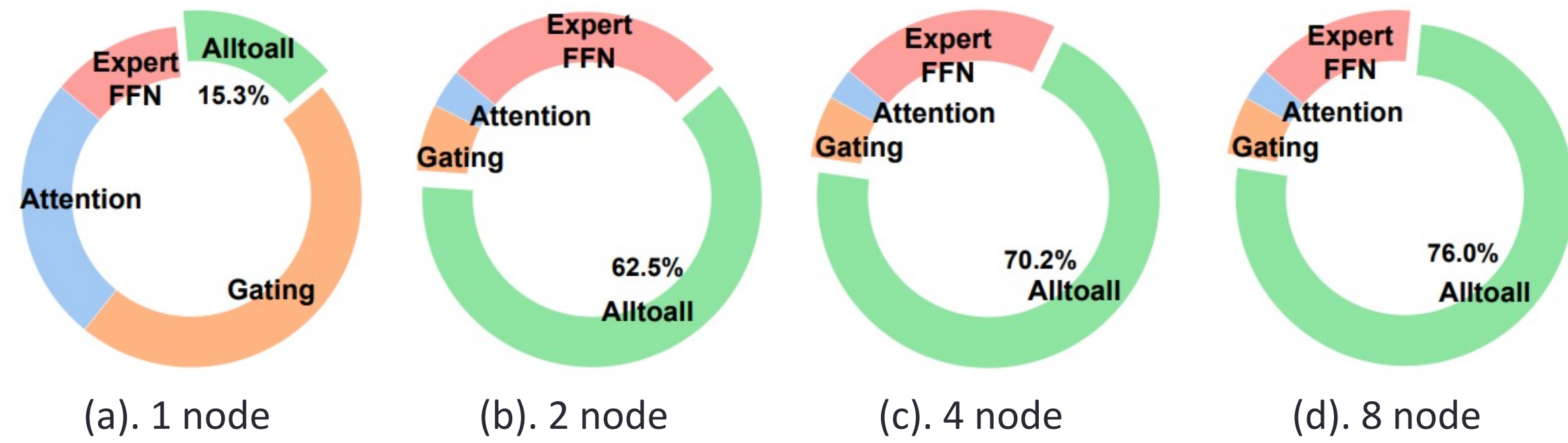


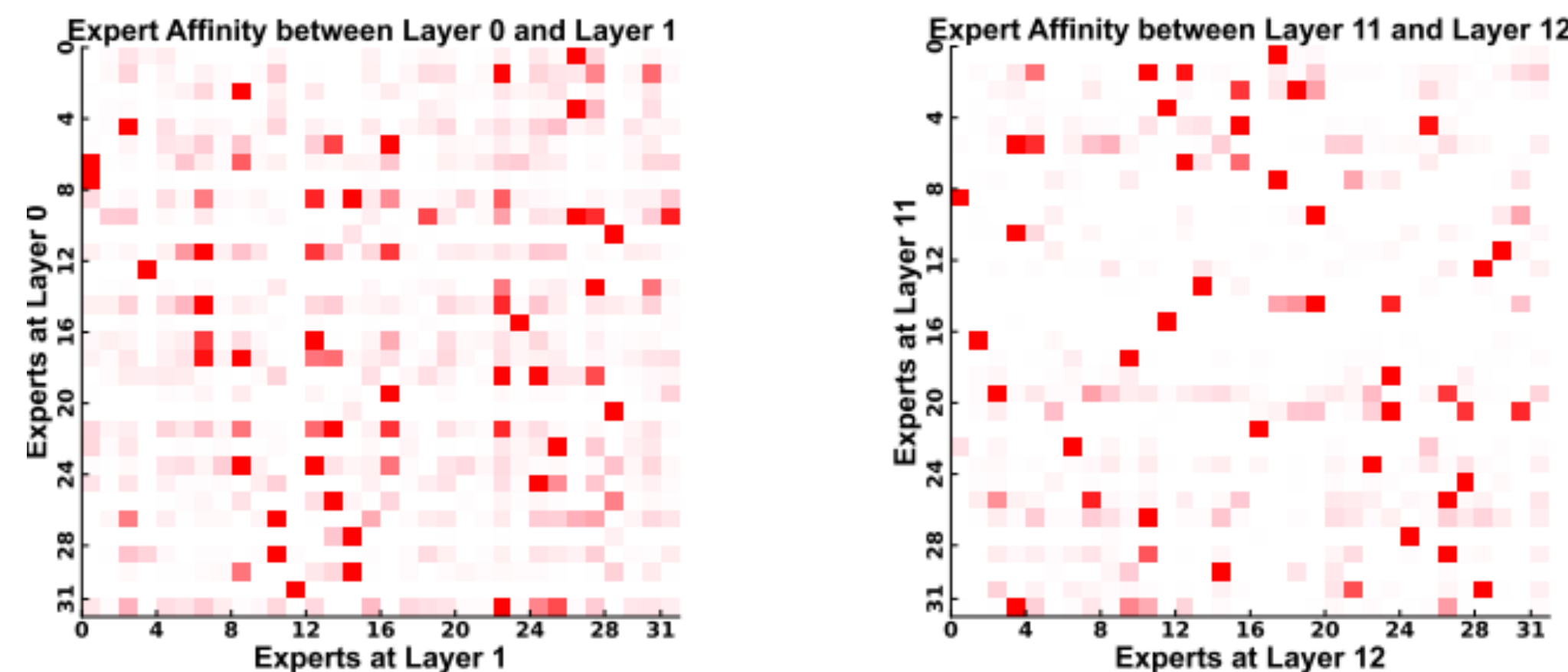
MOTIVATION

- Mixture-of-Experts (MoE) models are becoming popular in LLM/VLM, expert parallelism is widely used in both training and inference.
- Expert parallelism necessitates Alltoall communication to route tokens, which can introduce non-trivial overhead.



Proportion of Alltoall overhead to the time spent on computations. We only measure the most significant four operations in the MoE model.

- We found that in pretrained MoE models, experts among different layers exhibit a strong correlation. We name it as Expert Affinity.



Heatmaps illustrating the distribution of the conditional probability of expert routing in different layers of a pretrained GPT model. Color intensity represents the magnitude of the likelihood.

- Can we put experts with high affinity together to reduce the inter-layer Alltoall overhead?**

CHALLENGES

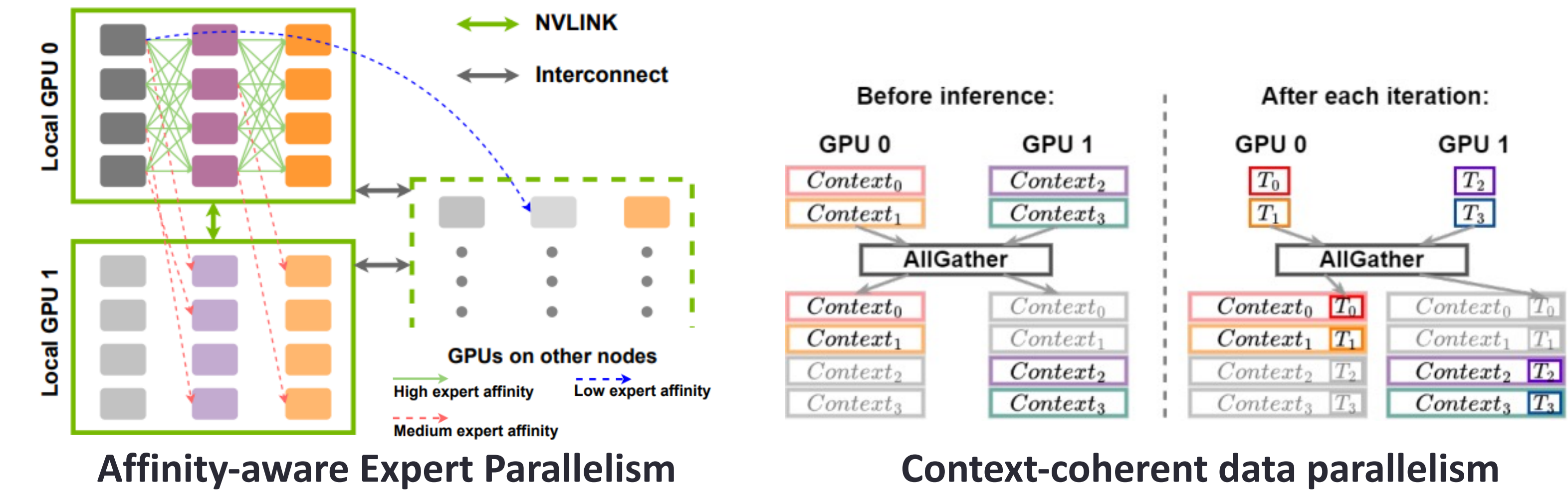
How to leverage expert affinity to accelerate MoE model inference?

- How do we model expert affinity in a deterministic manner?
- How to map expert affinity to a given hardware topology?
- How to eliminate data locality in expert + data parallel schemes?

DESIGN – ExFlow

We propose a novel expert parallelism optimization solution based on context coherence and expert affinity, named ExFlow.

- ExFlow defines the expert affinity property that exists implicitly in pretrained MoE models by capturing the combined conditional probability of the expert routing decisions among layers.
- Together with expert affinity, ExFlow introduces a context-coherent data parallelism to largely reduce the Alltoall overhead.



PERFORMANCE – End-to-end Inference

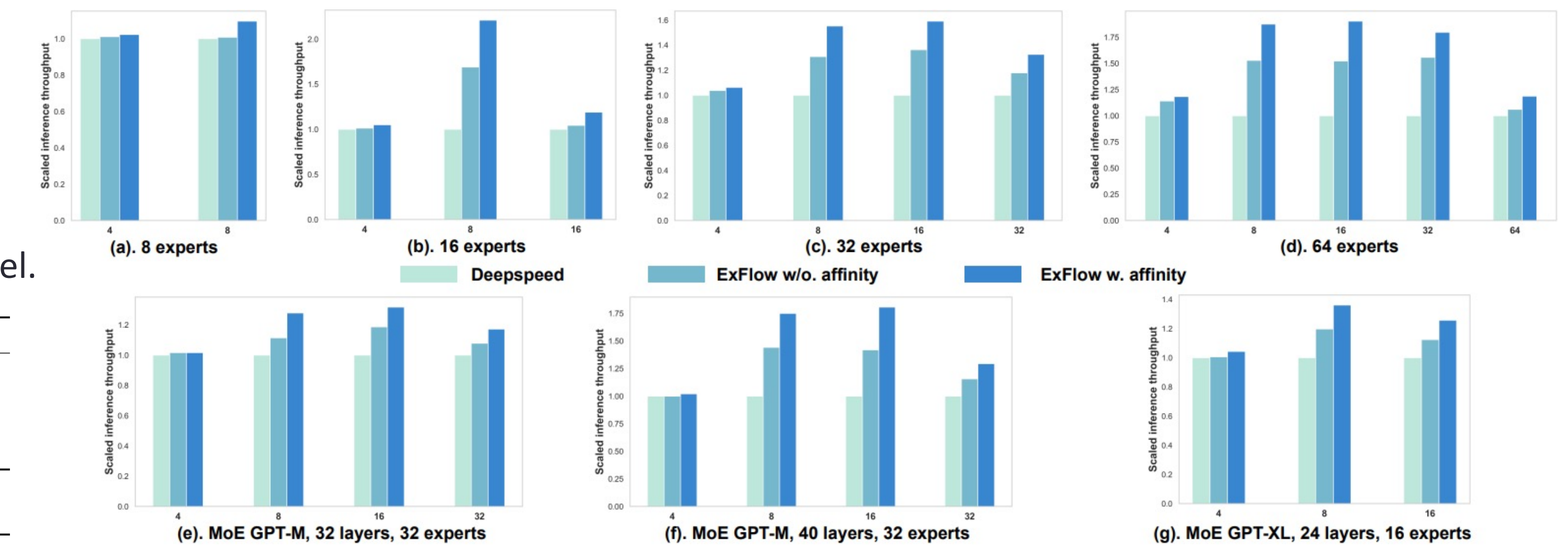
We test ExFlow based on DeepSpeed MoE implementation. We use Cambridge Wilkes-3 (A100 GPUs) for up to 64 GPUs with IB HDR 200.

- Baseline:** DeepSpeed MoE with Tutel optimization.
- ExFlow w/o. affinity:** Only use context-coherent data parallel.
- ExFlow w. affinity:** With context-coherent data parallel & expert affinity-aware parallel.

We choose 6 variants of MoE models, as listed:

- Training:** We use DeepSpeed-Megatron for pretraining with 300B tokens.
- Inference:** All models are with Top-1 gating and lossless routing.
- We only use 5000 tokens to profile expert affinity.**

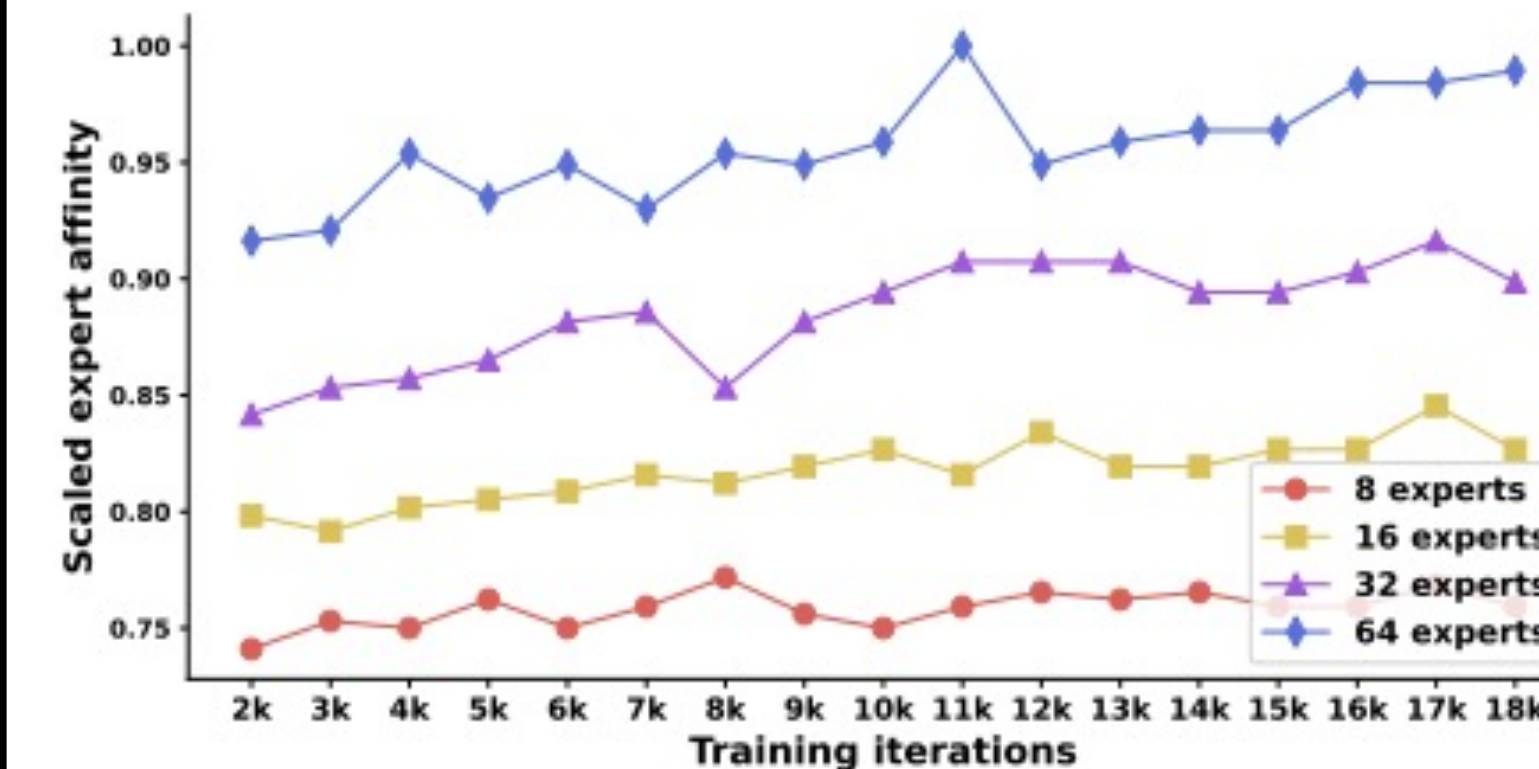
Model	Base	Experts	Layers	D
MoE GPT-M	350M	8, 16, 32, 64	24	1024
MoE GPT-M	470M, 590M	32	32, 40	1024
MoE GPT-XL	1.3B	16	24	2048



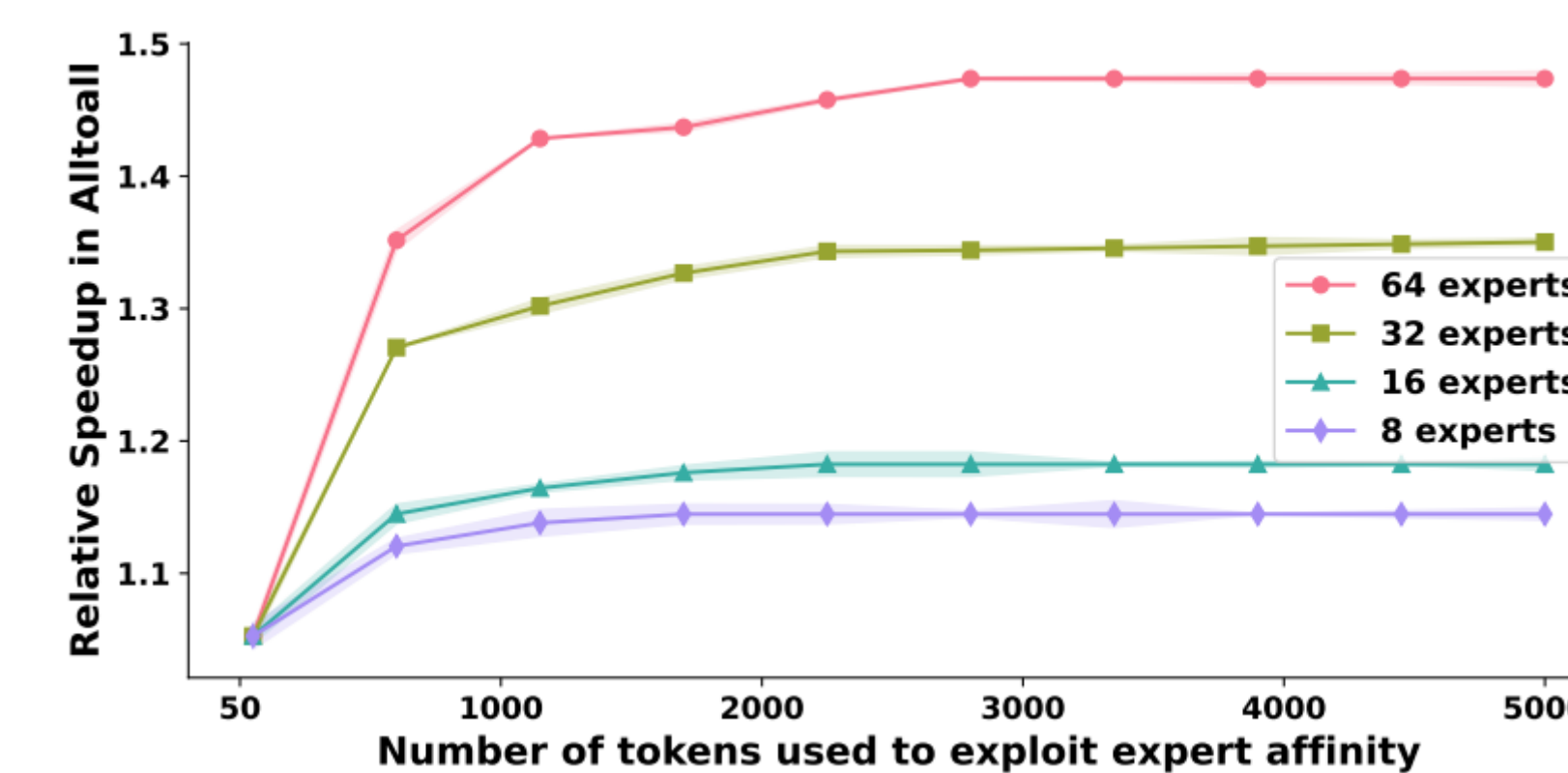
End-to-end GPT MoE model inference throughput. Results are normalized for better visualization.

PERFORMANCE – Expert Affinity

- Expert affinity gets more salient as the training goes on.**
- We design an efficient model to solve expert affinity. By only examine 5000 tokens, we can get an accurate estimation and gain significant speedup in Alltoall.**



Scaled expert affinity during training iteration 2000 to 18000. Oscillations are observed during the first 2000 iterations. (please refer to our paper for details)



Number of randomly sampled tokens used to estimate the expert affinity and its relative speedup during inference.

CONTRIBUTIONS

- Comprehensive profiling and evaluation of expert affinity in current MoE models.
- Exploit the expert affinity in pretrained MoE models and largely reduce the inference latency in the context of expert parallelism.
- Validate the intrinsic property of expert affinity and its independence and robustness to inference-time data distribution.
- Evaluate proposed designs compared to existing expert parallel libraries using **DeepSpeed Mixture-of-Experts on up to 64 A100 GPUs.**