

## Conclusion

- ZKPerf: a ZK-SNARK proving benchmark with crypto and ML tasks
- Most costly parts of ML circuits are nonlinearities (e.g., ReLU)
  - Need efficient lookups to prove larger models
- Circuit structure matters
  - One way: optimize row and column dimensions



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# Summit on Responsible Decentralized Intelli —— Future of Decentralization and Al

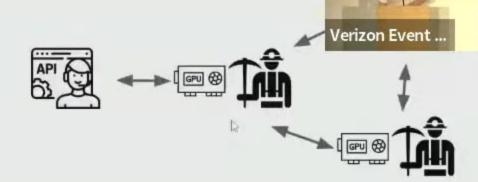


DeServe: Building Decentralized Services for Collaborative Large Language Model Inference

Xiaoyuan Liu, UC Berkeley, 2024



### Decentralized model serving



#### Problem scope

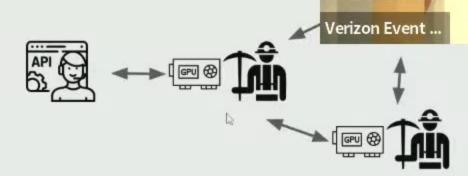
- Inference (fine-tuning, training)
- Decentralized computation resource: federated / personal device
- Offline serving (latency insensitive)

#### Key questions

- Can decentralized serving gain cost advantages?
- How to optimize throughput in high-latency heterogeneous env?
- How to prevent fraud and protect computation integrity?



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- 1. ChatGPT
  - a. gpt-40 \$5~\$15 / 1M token
  - b. gpt-40 mini (batched offline) \$0.075~\$0.3 / 1M token
- 2. Mining profit (Aug 2, 2024)
  - a. 1xRTX 4090 revenue:\$0.82 per day (NEXA)



- 3. LLM serving requirement
  - a. <u>Llama-3</u>-70b → mem: ~140G
  - b. RTX 4090: 24 $G \rightarrow \text{at least 6 (ideally >= 8)}_{Powered by Zoom}$



- ⇒ If you have 8×4090, mining for one day, you get \$6.56
- ⇒ Serving 21.9 M token ⇒ 253 token/s

#### Take away:

if you build a serving framework with throughput higher than this bar, you can convince miners to run decentralized serving to earn more







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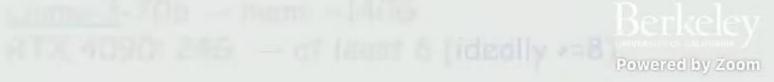


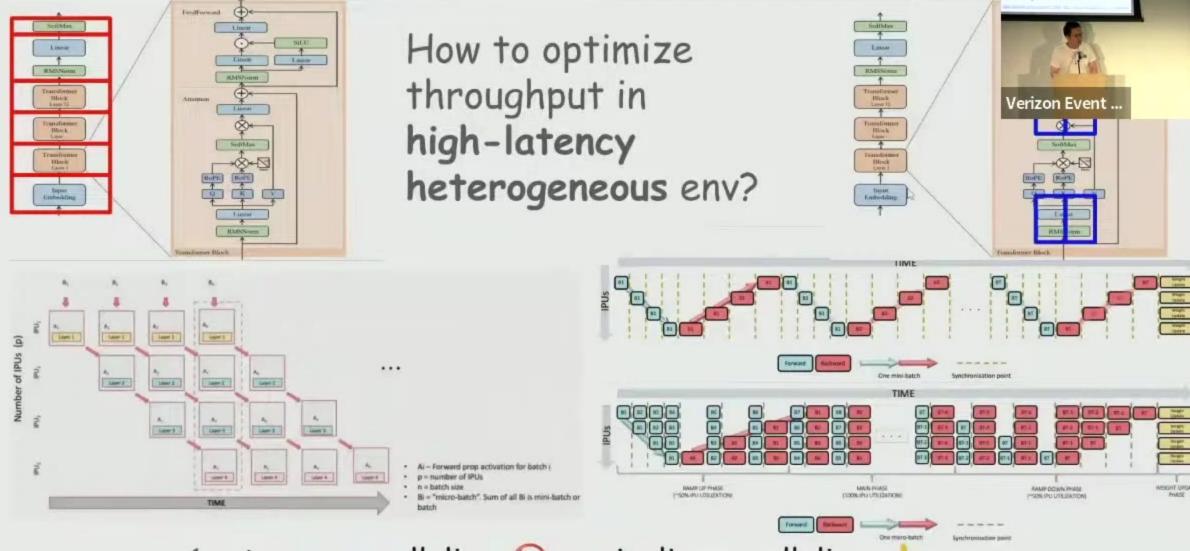


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1. tensor parallelism N vs pipeline parallelism

 Tensor parallelism requires frequent, high-bandwidth connection to be efficient (e.g. same machine) How to optimize throughput in high-latency heterogeneous env?

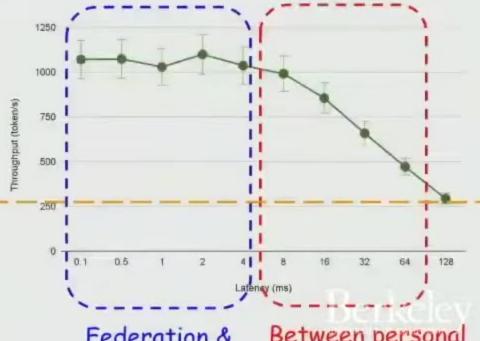


Applying optimization techniques:

- + Memory-aware (continuous) batching
- Pipelining (in circle)
- + Flash attention
- + Paged attention
- + Optimized tensor serialization, etc...

⇒ With 8x4090, 1071 token/s (bar: 253 token/s)

#### When latency increases:



Federation & data center

Between personal

How to optimize throughput in high-latency heterogeneous env?

### Heterogeneity with pipelining:

Llama-3-8b: 32 layers

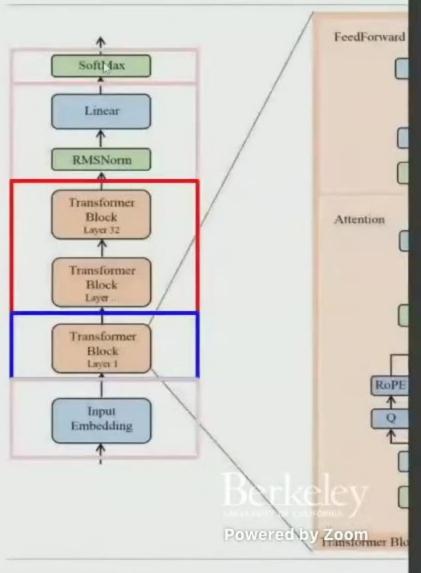
Llama-3-70b: 80 layers

Llama-3-405b: 126 layers

Faster device with more GPU mem

Slower device with less GPU mem

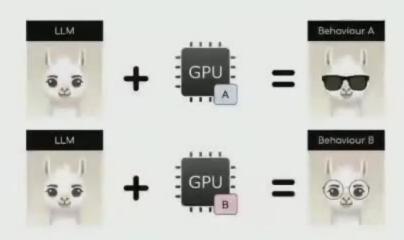




## How to prevent fraud and protect computation integrit



Definition of "correctness": close to the "standard" (precise) result



Changing the GPU is changing the behaviour of your LLM.



#### Different output (token prob), both benign

On the T4 side: "etc... This also means you can trust the output more since everything inside will be consistent across different runs!..."

On the A10G side: "etc... This also means you can be more confident when asking questions specifically related to topics covered within those texts..."

Reason: non associative float-point op

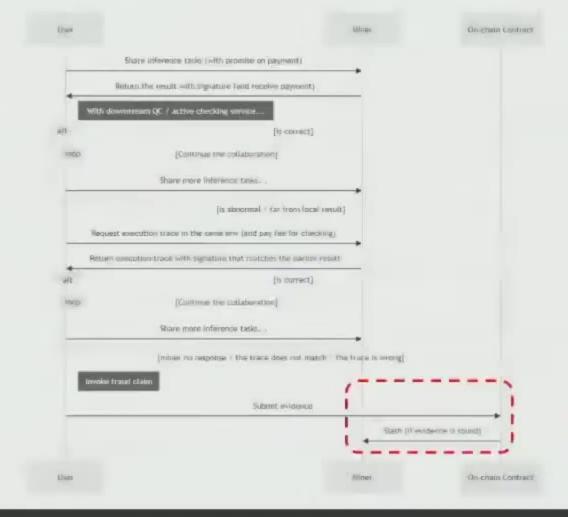
```
>>> sum([1e10] + [1e-10] * int(1e5))
100000000000.0
>>> sum([1e-10] * int(1e5) + [1e10])
100000000000.00001
```



## How to prevent fraud and protect computation integrit



A simple optimistic design: trace checking with a threshold on tensor difference



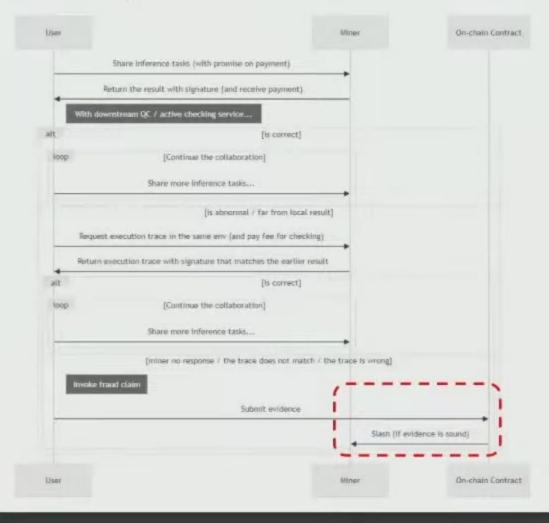
Save on-chain computation: Same design as step-proof

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# How to prevent fraud and protect computation integrit



A simple optimistic design: trace checking with a threshold on tensor difference



Save on-chain computation: Same design as step-proof

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## Thank you!

DeServe: Building Decentralized Services for Collaborative Large Language Model Inference

- Matiestiers pricing model - profet han 256 tilken/s/8+6090

Serving officionary optimization - model appairing through pateonic

- Correctness Suntegrity - optimistic & deformments serving

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## Thank you!

**DeServe**: Building Decentralized Services for Collaborative Large Language Model Inference

- Motivation: pricing model profit bar: 253 token/s/8x4090
- Serving efficiency optimization model pipelining through network
- Correctness & integrity optimistic & deterministic serving

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