

### Shared Memory Pool for Al Applications

August 2025

# :: 3.2 tbps acf-s "millennium"

## world's highest throughput ai supernic chip



#### ACF-S "Millennium" Chip

#### Resilient network

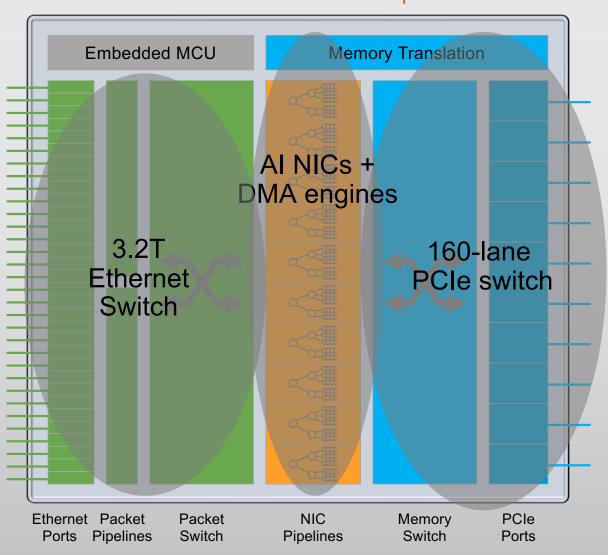
high port radix over fat (800G) or skinny (100G) links

#### **Full Router:**

consolidates NIC-TOR-PCIe fabrics with precise steering to/from queue pairs

#### **User programmable transport**

on scalable infra host cores at aggregate line rate PPS



Delivers elastic, peak aggregate

3.2 Tbps bandwidth to

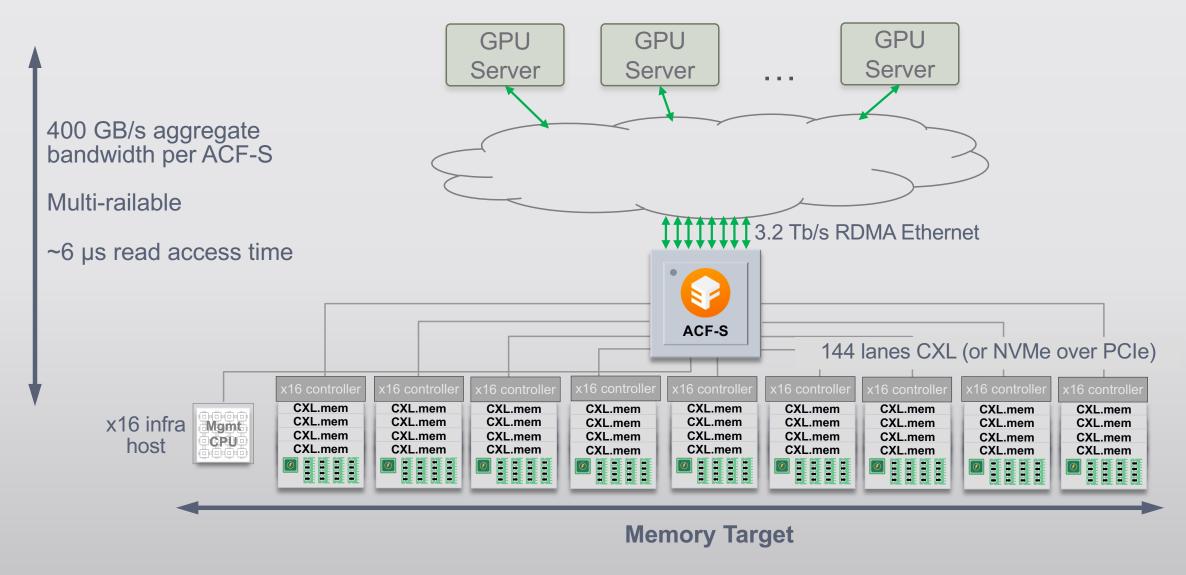
accelerator

Multi-planar internal switch fabric absorbs GPU incast, optimizes data placement and reduces data transfer time

Composable DMA & collective ops offload GPU SMs:40K copy engines / data movers







Can also aggregate NVMe Flash – up to 144 lanes

# :: shared memory pool benefits

### Very large memory space

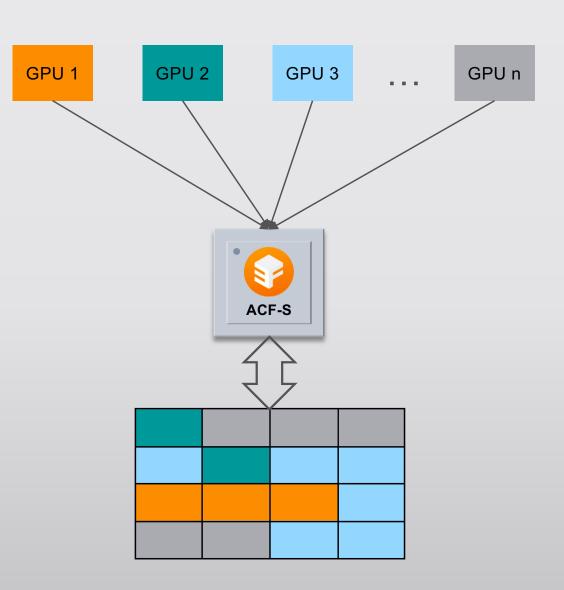
Can scale with additional systems

### Fast network access

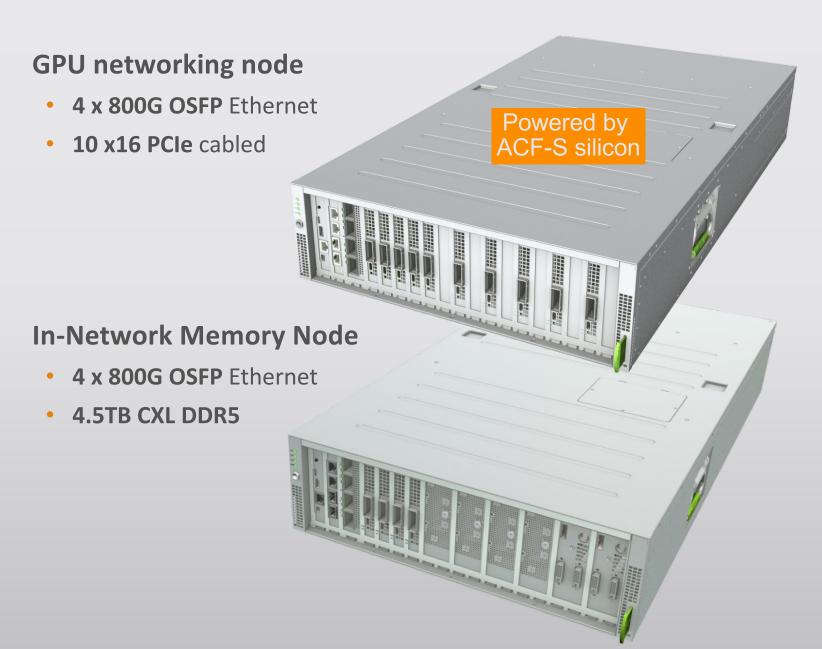
- Network access bandwidth matching memory
- One network hop away

### Shared across multiple clients

- Efficient memory utilization vs dedicated memory per client
- Ability to share data between clients



# :: thames system



8 Tbps AI Networking Node

Connect any combination of GPUs, CPUs, CXL memory, SSD to network

Programmable Network Transport: RoCE, RDMA over TCP

Replaces NICs, PCIe switches, Ethernet TOR

800G server I/O

Composable, modular, production-grade

## :: enfabrica EMFASYS



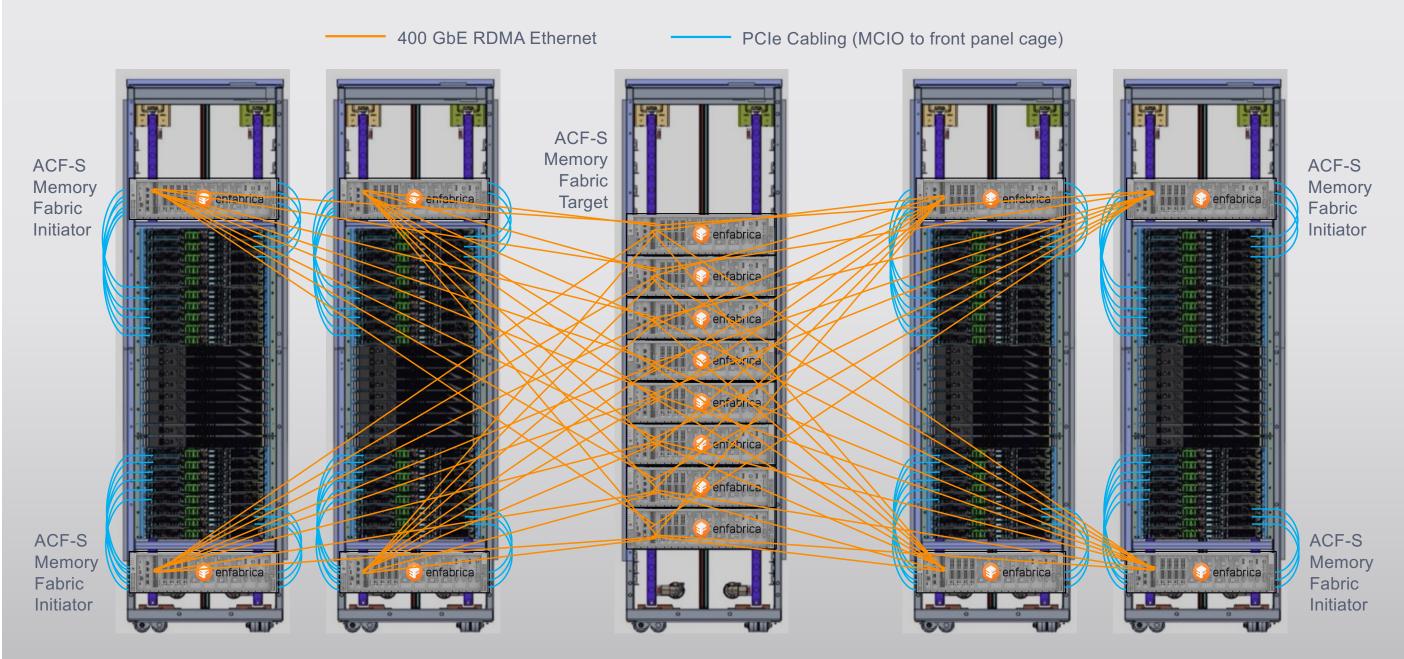


### **EMFASYS**: Enfabrica Elastic Al Memory Fabric

- 3.2Tbps RDMA Target with 4.5 TB -18 TB CXL DDR5 Store per node (50-100X HBM)
- Low Jitter 6µs RDMA read access time between GPU HBM and ACF-S Target Memory (50-200X latency reduction vs GPU Direct Storage)
- Unconstrained write/erase cycles vs Flash Storage for sustained high-throughput KV cache / token / activation data movement
- <\$15/GB fully loaded fabric-attached ACF-S memory vs ~\$100/GB for HBM3e stored</li>
   KV cache, token, or activation buffers → 5X cost reduction
- Memory Target to any initiator GPU server (H100/200, B200/300, MI3xx) via PCIe to drive down FLOPs/HBM consumption and cost per query up to 50%
- No CPU memory controller contention or CPU memory locality constraints,
   100% pooled, fabric-attached and headless

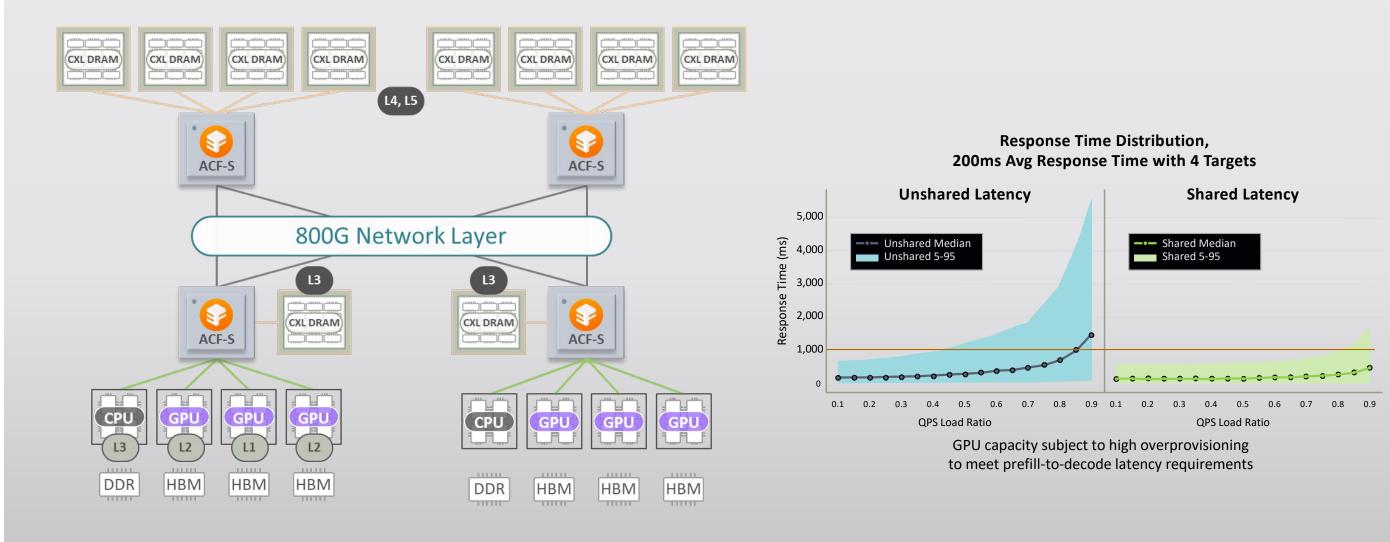
## :: cluster scale EMFASYS architecture







# :: impact of memory tiering on inference at scale



ACF-S can drive up to 50% fewer FLOPS required for large-sequence-length inference workloads at scale

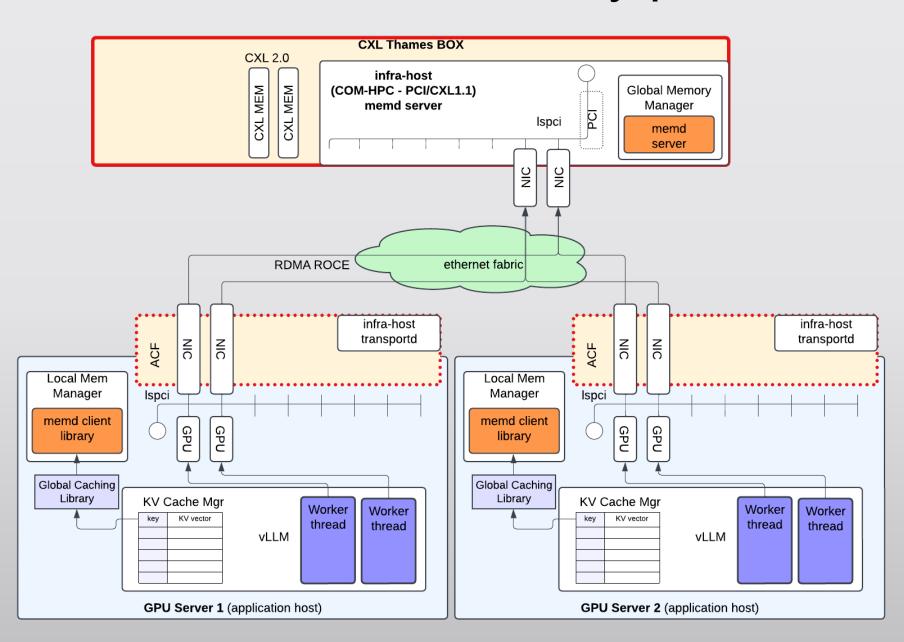
# :: inference acceleration with shared memory pool

### Testing environment

- Enfabrica Memory Box with up to 9 CXL memory cards
- 1 or more GPU servers
- 400Gb/s Ethernet network

#### Software environment

- vLLM inference framework
- LMCache serving engine
- Enfabrica rmem layer

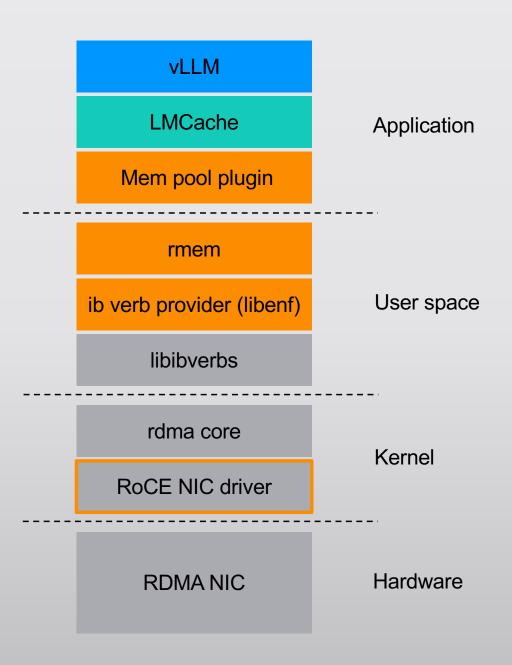


## :: initiator software stack

### Standard RDMA stack + Enfabrica libraries

- RDMA device driver (Enfabrica driver for ACF-S)
- IB verb provider (libenf)
- rmem client
- LMCache plugin

### Simple key-value store API



# :: let the fun begin!

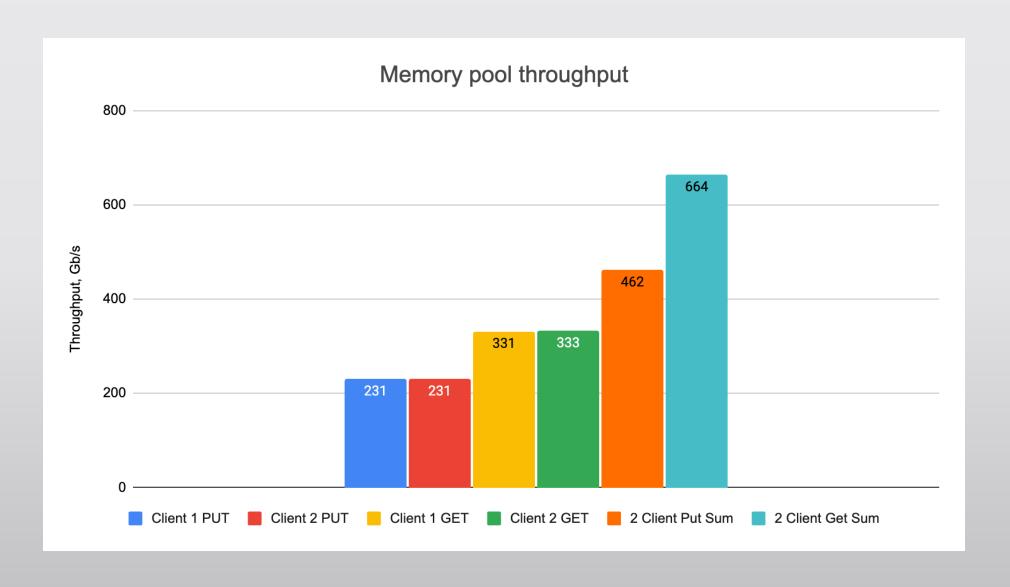






# :: memory pool performance

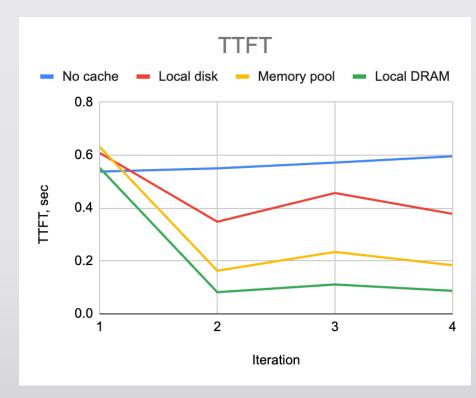
(optimizations ongoing)

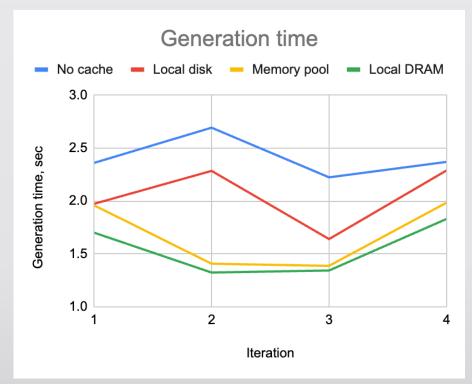


## :: inference test results: burst workload

#### Test conditions

- 1x GPU (H100)
- GPU HBM size: 43GB
- Local DRAM size:
  - 50GB (no L3 tier available)
  - 8GB (disk/memory pool present)
- Memory pool size: 50GB
- Concurrent users: 20
- Input tokens per query: ~5K-8K
- Output tokens per query: 100
- Iterations per user: 3
- Workload: all users issue their queries in parallel





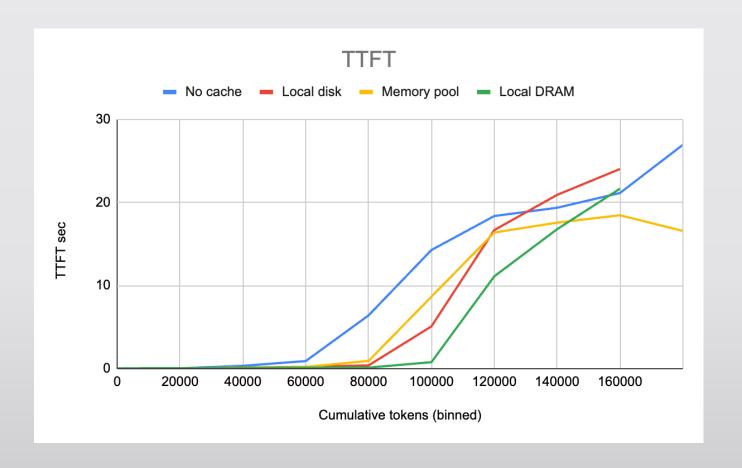
#### Results

- Without additional memory GPU HBM is too small resulting in KV cache recalculation for every new query since
- Large local DRAM shows best performance (as long as it can contain all generated data)
- Remote memory pool shows much better results than local disk due to its speed

## :: inference test results: iterative workload

#### Test conditions

- ° 1x GPU (H100)
- GPU HBM size: 43GB
- Local DRAM size:
  - 50GB (no L3 tier available)
  - 8GB (disk/memory pool present)
- Memory pool size: 50GB
- Concurrent users: 10
- Input tokens per query:
  - First prompt ~200
  - Last prompt: ~14K
- Output tokens per query: 600
- Iterations per user: 26
- Workload:
  - All users start at the same time
  - Gradually increase the prompt over 26 iterations



#### Results

- GPU starts running out of HBM capacity around 60K tokens served
- Remote memory pool outperforms local disk and GPU w/o additional memory after 120K tokens served
- Remote memory pool outperforms local DRAM after 150K tokens served

## :: conclusions

### Large RDMA accessible memory pool

- Accelerates inference response times
- Decouples KV cache storage from the GPU server
- Allows dynamic GPU allocation anywhere in the cluster (decoupled from server's local DRAM)
- Can be utilized to store other components (model data, encodings, intermediate storage for reinforced learning and more)

### **Key EMFASYS characteristics**

- Network throughput matched with CXL memory bandwidth allows fully non-blocking get/put access
- Low access latency results in much faster response times compared to flash storage
- Simple get/put API over standard RoCEv2 protocol makes it easy to use

## :: call for action

Interested in experimenting with Enfabrica Memory Pool solution?

- Want to explore new use cases?
- Compare performance to alternative solutions?
- Collaborate with Enfabrica on application software development/integration?

Contact Boris Shpolyansky boris@enfabrica.net

# :: Thank You