





ASTRA-SIM Description

Network Layer



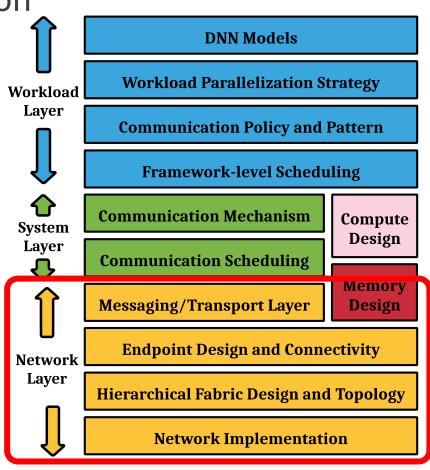
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Network Layer

- Workload layer manages workload, parallelization, training loop
- System layer initiates collective communication
- Network layer simulates actual network behaviors
 - Communication protocols (TCP, RDMA, etc.)
 - Network topology
 - BW/latency per link
 - In-network collective communication
 - NIC offloading
 - Compression
 - Buffering, Arbitration

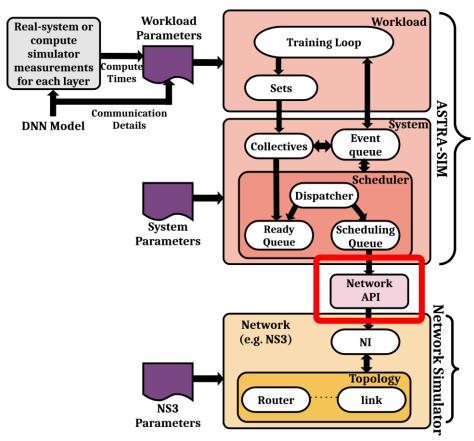


NetworkAPI

Interface between System layer and Network backend

Any network simulator implementing the NetworkAPI

could be used as ASTRA-sim backend



(HOTI '20) Scalable Distributed Training of Recommendation Models: An ASTRA-SIM + NS3 case-study with TCP/IP transport

Example NetworkAPIs

- sim_send(msg_size, src, dest, callback)
 - Simulate sending a message of size msg_size from src through dest and invoke callback function once transmission has finished
- sim recv(msg size, src, dest, callback)
 - Simulate receiving a message of size msg_size from src through dest and invoke callback function once transmission has finished
- sim_schedule(delta, callback)
 - Invoke callback function after delta time
- sim_get_time()
 - Return current time of simulation to the frontend

Simulation Control Flow

- The main file (simulation entry point) is implemented inside network layer.
- Network layer creates corresponding System and NetworkAPI instances.

• Each system layer instance **internally** creates its workload layer instance.

Network Layer

Where do instantiations happen?

Analytical backend (analytical): analytical/analytical/main.cc

Analytical backend (congestion): analytical/congestion/main.cc

Garnet backend: garnet/gem5 astra/src/mem/ruby/network/garnet2.0/NetworkInterface.cc

ns3 backend: ns3-interface/simulation/scratch/AstraSimNetwork.cc

```
void main(){
Instantiate NetworkAPI[];
Instantiate System[];
for(auto &s:system){
  s->workload.fire();
process all events();
return;
```

Available Network Backends

• Network backends are maintained separately and are imported as **submodule**.

We currently have 4 network backends which implement NetworkAPI

Backend	Purpose	Notable Feature
analytical/analytical	analytical equation-based simulation	fast simulation, hierarchical topologies
analytical/congestion	congestion-aware analytical simulation	first-order congestion (queueing) modeling
Garnet	on-chip/scale-up network simulation	packetization, flow control, congestion
ns-3	inter-network simulation	large parallel GPU clusters

Analytical Backend

• Leverages analytical equation to estimate communication delay

- sim send(msg size, src, dest, callback)
 - Estimate communication delay
 - Assign callback to event queue after delay
- No congestion modeling
 - Appropriate for topology-aware collectives without network congestion
- Fast simulation for large-scale systems

(ISPASS '23) ASTRA-sim2.0: Modeling Hierarchical Networks and Disaggregated Systems for Large-model Training at Scale

Congestion-aware Analytical Backend

- First-order congestion modeling by per-link queueing
- Per-link delay is calculated using analytical eqaution

```
e.g., send (msg_size: 1 MB, route: [1, 2, 3, 4, 5])
send (1 MB, 1 → 2)
send (1 MB, 2 → 3)
each send can be queued per each link
send (1 MB, 3 → 4)
link processes pending chunks in-order
send (1 MB, 4 → 5)
```

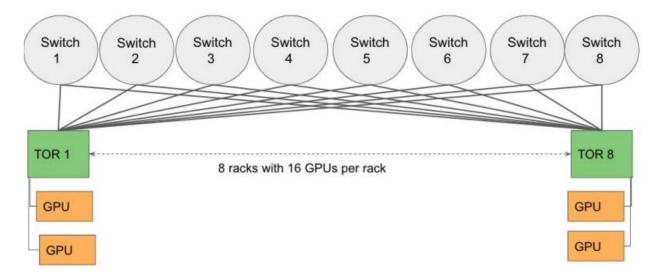
Fast simulation for large-scale systems with network congestion

Garnet Backend

- Leverages Garnet (interconnection network) simulator as backend
- Appropriate for on-chip/scale-up networks
- Simulates interconnection network behaviors:
 - Message Packetization
 - Credit-based flow control
 - Congestion modeling
 - etc.
- Slower than analytical backend for large systems/models
- Supports switch-based/torus-based topologies

NS3 Backend

- Network simulator for internet (inter-node) communication
- Used to model ML training in largely parallel GPU clusters
- 1 NPU in Analytcial/Garnet becomes **1 GPU** in NS3, connected with ToR/spine switch, etc.

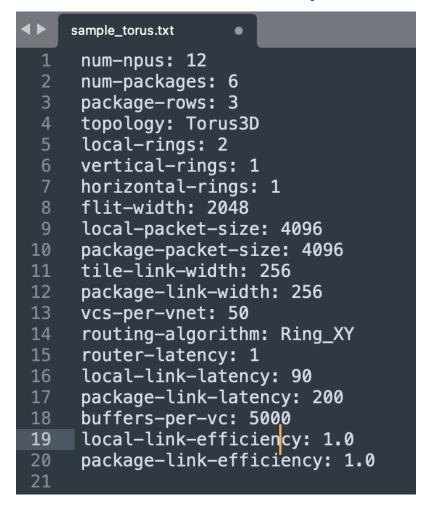


(HOTI '22) Current RoCE congestion control methods have little impact on ML training workloads

Slide courtesy of: Jinsun Yoo (Georgia Tech)

Sample Network Input

Garnet network input



```
Analytical network input
                                               Intra-package scale-up
                                          Inter-package scale-up [ ] Package
    sample_Torus3D.json
       "topology-name": "Hierarchical",
       "topologies-per-dim": ["Ring", "Ring", "Ring"],
       "dimension-type": ["N", "N", "N"],
       "dimensions-count": 3,
 6
       "units-count": [2, 2, 3],
       "links-count": [2, 2, 2],
 8
       "link-latency": [10, 100, 100],
       "link-bandwidth": [32, 16, 16],
10
       "nic-latency": [0, 0, 0],
       "router-latency": [0, 0, 0],
11
       "hbm-latency": [500, 500, 500],
12
       "hbm-bandwidth": [370, 370, 370],
13
14
       "hbm-scale": [0, 0, 0]
15
```