





Exercise 4: Implementing a New Training Loop



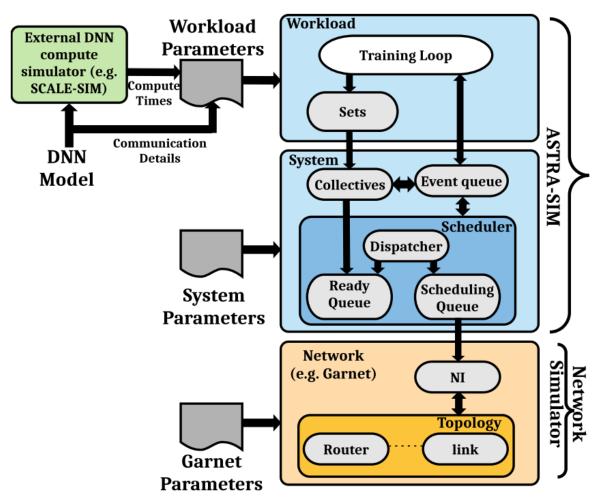
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Objective

To demonstrate how you can implement a new training loop in ASTRA-sim

Training Loops



- Training loop determines the behavior of a workload
 - Parallelization strategy
 - Computation order
 - Communication order
- Supported training loops
 - Data parallel
 - Model parallel
 - DLRM
 - Transformer
- You can implement a new training loop to support other models

Training Loop Analysis – Data Parallel

- Distribute Data across multiple nodes and replicate model (network) along all nodes.
- No communication during the forward pass.

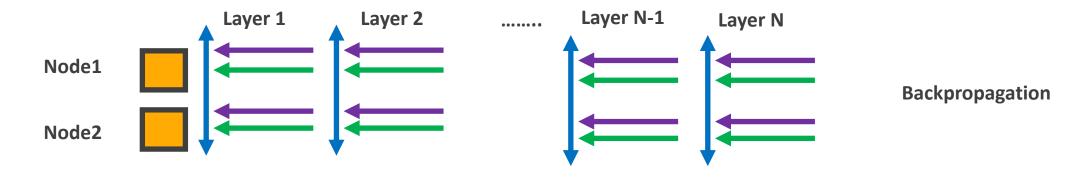


Flow-per-layer: 1.Compute output -> 2. go to the next layer



Training Loop Analysis – Data Parallel

- Distribute Data across multiple nodes and replicate model (network) along all nodes.
- Communicate weight gradients during the backpropagation pass.
 - Blocking wait during forward pass for collective of previous backpropagation for that layer.



Flow-per-layer: 1.Compute weight gradient-> 2.issue weight gradient comm -> 3.compute input gradient -> 4. go to previous layer

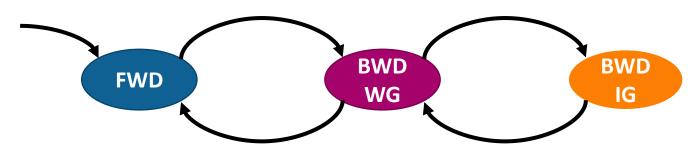


Training Loop Analysis – Data Parallel

Vanilla Data-parallel Training Schedule



FSM Diagram

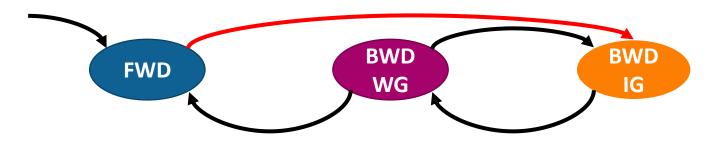


Exercise: Reorder Computation of Data-parallel Loop

Reordered Data-parallel Training Schedule



FSM Diagram



Adding a New Training Loop

- See astra-sim/workload/Workload.cc
- Vanilla data-parallel loop is iterate_data_parallel()
- Add a reordered version, iterate_data_parallel_reorder()

```
void Workload::call(EventType event, CallData* data) {
  if (counter > 0) {
    generator->try_register_event(
        this, EventType::Workload_Wait, NULL, counter);
    return;
}
if (parallelismPolicy == ParallelismPolicy::Data) {
    iterate_data_parallel();
} else if (parallelismPolicy == ParallelismPolicy::DataReorder) {
    iterate_data_parallel_reorder();
} else if (parallelismPolicy == ParallelismPolicy::Transformer) {
    iterate_hybrid_parallel_Transformer();
```

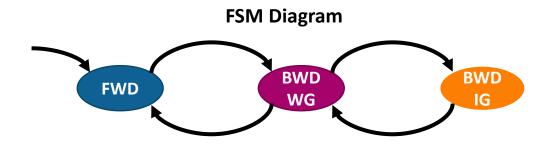
Reorder the Computation Schedule

```
void Workload::iterate data parallel() {
  assert(index >= 0);
  assert(index < SIZE);</pre>
  check for sim end();
  if (current state == LoopState::Forward Pass) {
 -- 31 lines: if (!layers[index]->is_weight_grad_comm finished blo
    if (index >= SIZE) {
      current state = LoopState::Weight Gradient;
      index--;
    generator->register event(this, EventType::General, NULL, 1);
    return;
    else if (current state == LoopState::Weight Gradient)
    if (index == 0) {
      pass counter++;
      current state = LoopState::Forward Pass;
     else {
      current_state = LoopState::Input_Gradient;
    generator->register event(this, EventType::General, NULL, 1);
    return;
    else if (current state == LoopState::Input Gradient) {
    11 lines: if (delay loaded == false) {------
    delay loaded = false;
    index--;
    current state = LoopState::Weight Gradient;
    generator->register_event(this, EventType::General, NULL, 1);
    return;
```

Training loop is implemented as a FSM

Vanilla Data-parallel Training Schedule





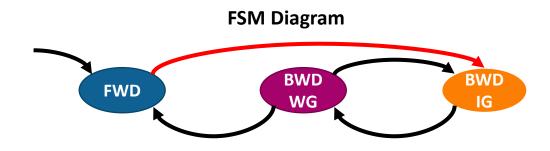
Reorder the Computation Schedule

```
void Workload::iterate data parallel reorder() {
 assert(index >= 0);
 assert(index < SIZE);</pre>
 check for sim end();
 if (current_state == LoopState::Forward_Pass) {
 - 16 lines: if (!layers[index]->is weight grad comm finished bl
   if (index >= SIZE) {
     current state = LoopState::Input Gradient;
     index--:
   generator->register event(this, EventType::General, NULL, 1);
   return;
   else if (current_state == LoopState::Weight_Gradient) {
   15 lines: if (delay loaded == false) {-------
   if (index > 1) {
     index--:
     current state = LoopState::Input Gradient;
    } else if (index == 1) {
     index--:
     current state = LoopState::Weight Gradient;
    } else if (index == 0) {
     pass_counter++;
     current_state = LoopState::Forward_Pass;
   generator->register_event(this, EventType::General, NULL, 1);
   return;
   else if (current state == LoopState::Input Gradient) {
   11 lines: if (delay loaded == false) {------
   delay loaded = false;
   current state = LoopState::Weight Gradient;
   generator->register_event(this, EventType::General, NULL, 1);
   return;
```

 You can reorder the computation schedule by tweaking the index and current_state

Reordered Data-parallel Training Schedule





Adding Debugging Messages

```
void Workload::iterate data parallel reorder() {
 assert(index >= 0);
 assert(index < SIZE);</pre>
 check_for_sim_end();
 if (current_state == LoopState::Forward Pass) {
   3 lines: if (!layers[index]->is_weight_grad_comm_finished_blockir
   if (delay loaded == false) {
     counter = layers[index]->get fwd pass compute();
     delay_loaded = true;
     if (generator->id == 0)
        std::cout << "[TUTORIAL] FWD[" << index <<"]" << std::endl;</pre>
  } else if (current state == LoopState::Weight Gradient) {
   if (delay loaded == false) {
     counter = layers[index]->get weight grad compute();
     delay loaded = true;
     if (generator->id == 0)
        std::cout << "[TUTORIAL] BWD_WG[" << index <<"]" << std::endl;</pre>
 } else if (current_state == LoopState::Input_Gradient) {
   if (delay loaded == false) {
     counter = layers[index]->get input grad compute();
     delay loaded = true;
     if (generator->id == 0)
        std::cout << "[TUTORIAL] BWD_IG[" << index <<"]" << std::endl;</pre>
    9 lines: if (counter > 0) {------
```

- You can add debugging messages to make sure that the training loop works as expected
- Make sure to print debugging messages only when (generator->id == 0)
 - Each processing element is a generator
 - If you don't filtering the ID, you will see debugging messages from all PEs