Triton Update













Triton Review

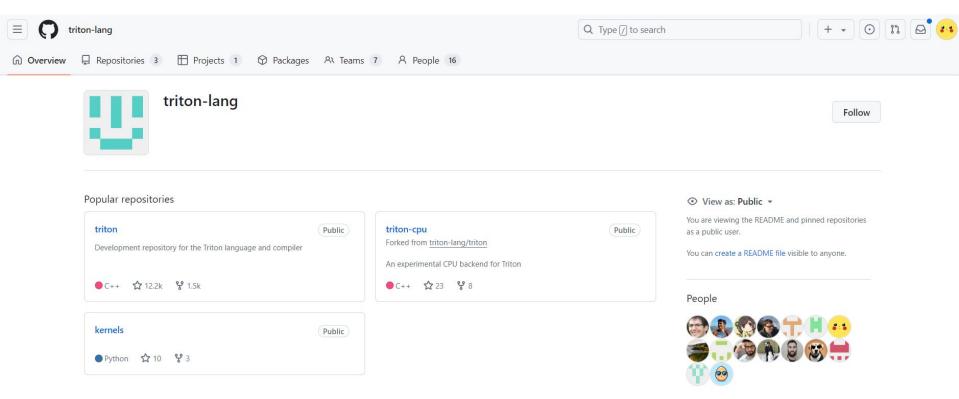
What is Triton

- A Python-like language
- A JIT compiler
- A PyTorch backend
- A set of MLIR dialects
- A community
- An organization

A Triton Program (Permutation)

```
vecAdd
                       import torch
                       import triton
Package import
                       import triton.language as tl
                       atriton.jit
Device function
                       def permute(x, index, SIZE):
 Triton operator
                           indicator = tl.arange(0, SIZE)[:, None] = index
                           return tl.sum(indicator * x, axis=1)
Kernel decorator
                       atriton.jit
                       def kernel(x_ptr, y_ptr, BLOCK_SIZE: tl.constexpr):
                           permute_tid = (tl.arange(0, BLOCK_SIZE) + 15) % BLOCK_SIZE
                           tid = tl.arange(0, BLOCK_SIZE)
    Kernel body
                           x = tl.load(x_ptr + tid)
                            x = permute1d(x, permute_tid, BLOCK_SIZE)
                           tl.store(y_ptr + tid, x)
                                                                              snappify.com
```

The Triton-Lang Organization



Triton Community

Third-party Repositories In-tree Modules Out-of-tree Modules Triton Intel Language **PyTorch CPU** JAX Profiler Interpreter Tools Triton-shared AMD **NVIDIA** Backend Deepspeed (accelerator) . . . \sim \sim

Beginner's Resources

- Triton Index
 - <u>cuda-mode/triton-index: Cataloging released Triton kernels. (github.com)</u>
- Awesome Triton Kernels
 - zinccat/Awesome-Triton-Kernels: Collection of kernels written in Triton language (github.com)
- Unsloth
 - unslothai/unsloth: Finetune Llama 3, Mistral & Gemma LLMs 2-5x faster with 80% less memory (github.com)
- Triton Puzzles
 - <u>srush/Triton-Puzzles: Puzzles for learning Triton (github.com)</u>
- Torchao
 - pytorch/ao: Native PyTorch library for quantization and sparsity (github.com)
- Attorch
 - <u>BobMcDear/attorch: A subset of PyTorch's neural network modules, written in Python using OpenAl's Triton.</u> (github.com)

Guide for Developers

- Read the Triton source code!
- Read the MLIR source code!
- I found a handful of Triton backend analysis articles on zhihu.com
 - But triton core developers may not have time to write any of these
 - We prefer to leaving comments to save time
 - Discussion
 - [QST] Triton MLIR · Issue #3 · srush/Triton-Puzzles (github.com)

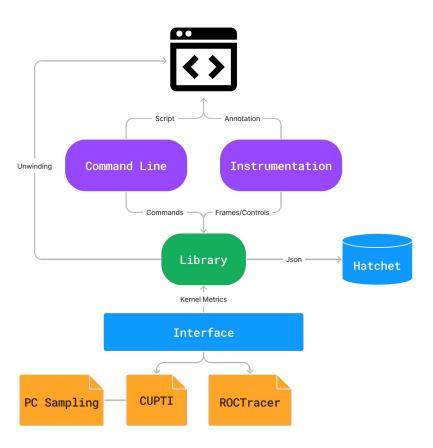
Proton

A Profiler for Triton

Proton

- Provide a quick, intuitive, and simple way to check kernel performance
 - Open source
 - Multiple vendor GPUs
 - Flexible metrics collection
 - Hardware metrics
 - Software metrics
 - Call path profiling

Design



Call Path Profiling

- Profile kernel running time

```
55.193 ROOT

→ 31.212 /home/kzhou6/qh200/triton/third_party/proton/tutorials/dynamic_net.py:<module>@98

  └─ 31.212 /home/kzhou6/gh200/triton/python/triton/profiler/profile.py:wrapper@151

→ 0.002 /home/kzhou6/qh200/triton/third_party/proton/tutorials/dynamic_net.py:run@51

        └ 0.002 _ZN50_GLOBAL__N__c922cf59_17_RangeFactories_cu_38772b0829elementwise_kernel_with_indexI
_clevENKUlvE0_clevEUlle_EEvT_T0_PN15function_traitsISD_E11result_typeE

→ 0.003 /home/kzhou6/gh200/triton/third_party/proton/tutorials/dynamic_net.py:run@52

□ 0.003 ZN2at6native29vectorized elementwise kernelILi4EZZZNS0 15sin kernel cudaERNS 18TensorI
_T1_

→ 19.610 /home/kzhou6/gh200/triton/third_party/proton/tutorials/dynamic_net.py:run@66

— 19.610 /home/kzhou6/qh200/pytorch/torch/nn/modules/module.py:_wrapped_call_impl@1532

└─ 19.610 /home/kzhou6/qh200/pytorch/torch/nn/modules/module.py:_call_impl@1541

→ 13.931 /home/kzhou6/gh200/triton/third_party/proton/tutorials/dynamic_net.py:forward@36

                 1.460 _ZN2at6native29vectorized_elementwise_kernelILi4EZNS0_53_GL0BAL__N__2ced54f@
18TensorIteratorBaseET0_EUlfE0_NS_6detail5ArrayIPcLi2EEEEEviS6_T1_

─ 1.479 _ZN2at6native29vectorized_elementwise_kernelILi4EZNS0_53_GLOBAL__N__2ced54f0

18TensorIteratorBaseET0_EUlfE_NS_6detail5ArrayIPcLi2EEEEEviS6_T1_

→ 6.022 _ZN2at6native18elementwise_kernelILi128ELi2EZNS0_22qpu_kernel_impl_nocastINS0_:

→ 2.025 _ZN2at6native18elementwise_kernelILi128ELi2EZNS0_22gpu_kernel_impl_nocastINS0_

    □ 2.945 ZN2at6native29vectorized_elementwise_kernelILi4ENS0_15CUDAFunctor_addIfEENS_6c
```

```
763 ROOT
  25,004 backward

─ 14.366 _ZN2at6native13reduce_kernelILi512ELi1ENS0.

   — 2.007 _ZN2at6native18elementwise_kernelILi128ELi2E
vEUlfffE_EEvRNS_18TensorIteratorBaseERKT_EUliE_EEviT1_

→ 2.461 _ZN2at6native29vectorized_elementwise_kernel

→ 5.725 _ZN2at6native29vectorized_elementwise_kernel

□ 0.446 _ZN2at6native29vectorized_elementwise_kernel

  19.399 forward

→ 7.961 ZN2at6native18elementwise kernelILi128ELi2E

EUliE_EEviT1_

→ 2.018 _ZN2at6native18elementwise_kernelILi128ELi2E

    4.415 _ZN2at6native29vectorized_elementwise_kernel
   1.455 _ZN2at6native29vectorized_elementwise_kernel
seET0_EUlfE0_NS_6detail5ArrayIPcLi2EEEEEviS6_T1_

→ 2.073 _ZN2at6native29vectorized_elementwise_kernel

seET0_EUlfE2_NS_6detail5ArrayIPcLi2EEEEEviS6_T1_

└─ 1.477 _ZN2at6native29vectorized_elementwise_kernel
seET0_EUlfE_NS_6detail5ArrayIPcLi2EEEEEviS6_T1_

→ 0.004 init

→ 0.003 _ZN2at6native29vectorized_elementwise_kernel

□ 0.001 _ZN50_GLOBAL__N__c922cf59_17_RangeFactories.

NKUlvE0_clEvEUllE_EEvT_T0_PN15function_traitsISD_E11resu
⊢ 4.412 loss
   — 2.949 _ZN2at6native13reduce_kernelILi512ELi1ENS0_8

└─ 1.462 _ZN2at6native29vectorized_elementwise_kernel
```

User Interface

- Lightweight source code instrumentation
 - Profile start/stop/finalize
 - Scopes
 - Hooks
- Command line
 - python -m proton main.py
 - proton main.py

Profile Start/Stop/Finalize

- Profile only interesting regions
 - proton.start(profile_name: str) -> session_id: int
 - proton.finalize()
- Skip some regions, but accumulate to the same profile
 - session_id = proton.start(...)
 - proton.deactive(session_id)
 - ... # region skipped
 - proton.activate(session_id)

Scopes

- Only collect the *Master Thread* scope
 - In PyTorch, the thread that train and test models

```
with proton.scope("test0"):
    with proton.scope("test1"):
        foo[1,](x, y)
with proton.scope("test2"):
    foo[1,](x, y)
```

Metrics

- Asynchronous metrics
 - Come from profilers
- Synchronous metrics
 - Come from users
 - Theoretical flops, bytes
 - Loss
 - Counts
 - Dict[str, Union[int, float]]

```
with proton.scope("test0", {"foo_metric": 1.0}):
    foo[1,](x, y)
```

"test0" scope ends with multiple metrics.

Two metrics can be displayed the same time.

```
[(pytorch) kzhou6@gracehopper:~$ proton-viewer -l proton.hatchet
<IPython.core.display.Javascript object>
Warning: Roundtrip module could not be loaded. Requires jupyter
Available metrics:
  count
  time
  foo metric
(pytorch) kzhou6@gracehopper:~$ proton-viewer -m time,foo_metric
<IPython.core.display.Javascript object>
Warning: Roundtrip module could not be loaded. Requires jupyter

    1440.000 nan _ZN2at6native29vectorized_elementwise_kernelILi4

 — 1664.000 1.000 test0

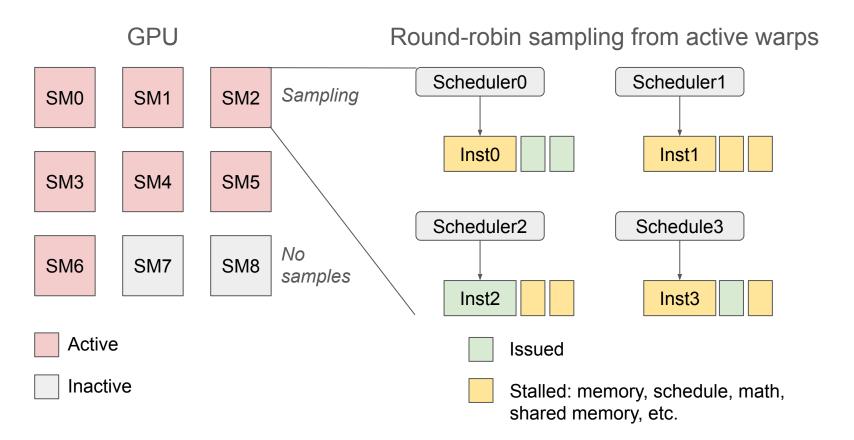
    □ 1664.000 nan foo

Legend (Metric: Time (ns) (inc) Min: 1440.00 Max: 3104.00)
  2937.60 - 3104.00
  2604.80 - 2937.60
  2272.00 - 2604.80
 1939.20 - 2272.00
  1440.00 - 1606.40
```

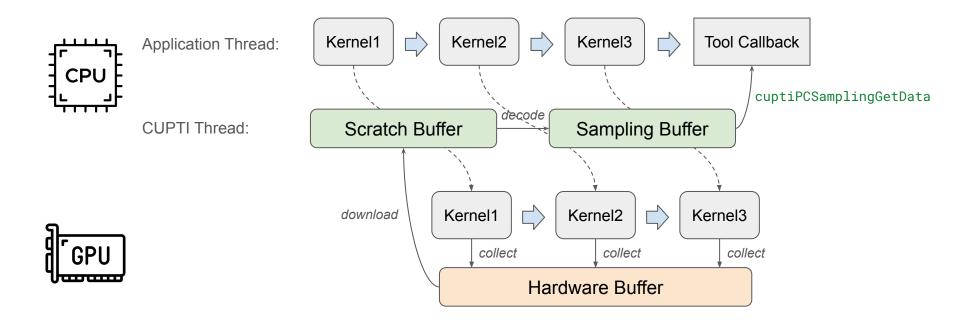
Triton Hooks

- Rename the triton function with a custom name
 - Append launch configurations
 - Append runtime dynamic
 - Append constants
 - e.g., foo_<num_warps:4>_<fast_math:4>_<branch_0:1>
- Supply custom metrics based on kernel arguments
 - flops{8, 16, 32, 64}
 - bytes

Instruction Sampling on NVIDIA GPUs



CUPTI Internals



Proton's View

109

110 111

112

Total samples and stalled samples

 $a = tl.load(a_ptrs, mask=offs_k[None, :] < K - k * BLOCK_SIZE_K, other=0.0)$

 $b = tl.load(b_ptrs, mask=offs_k[:, None] < K - k * BLOCK_SIZE_K, other=0.0)$

→ 8483.000 5349.000 matmul_<grid:18x1x1>_<cluster:1x1x1>_<warps:8>_<shared:147456>_<stages:3>

If it is out of bounds, set it to 0.

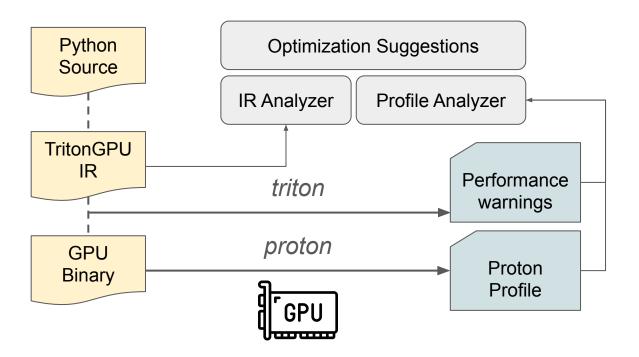
We accumulate along the K dimension.

Overhead

- NCU overhead >1000x
 - 4s -> 66 mins
 - time ncu --section SourceCounters python ./dynamic_net.py
- Proton overhead ~20x
 - Could be reduced to less than 5x
 - Many optimizations haven't been applied

Proton-Analyzer

Design for Torchinductor and Triton



Potential Views

- Interactive view
 - https://godbolt.org
- Terminal view
 - test.py@foo
 - test.py@kernel:1 (10%)
 - async_copy@prologue (5%)
 - async copy@body (5%)
 - test.py@kernel:3 (5%)
 - uncoalescd (5%)
 - test.py@kernel:4
 - test.py@kernel:5
 - test.py@kernel:6

- Diagnose 1:
 - Low utilization
 - Increase number of program instances

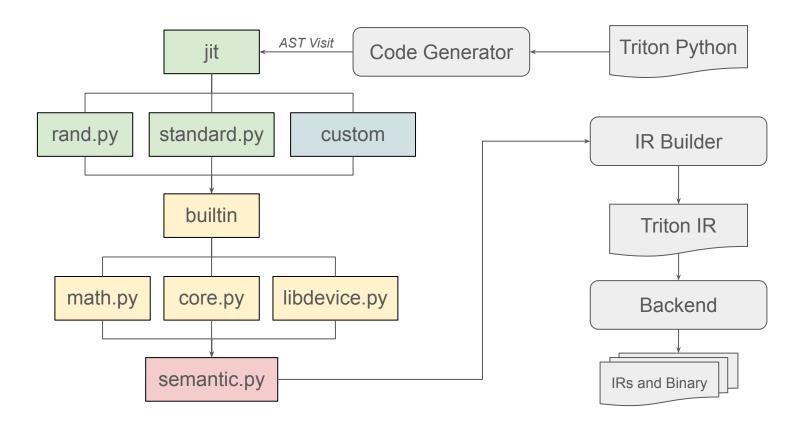
23

- Try triton autotune
- Diagnose 2:
 - No wgmma
 - Shape size not match
- Diagnose 3:
 - No wgmma
 - Shape size not map

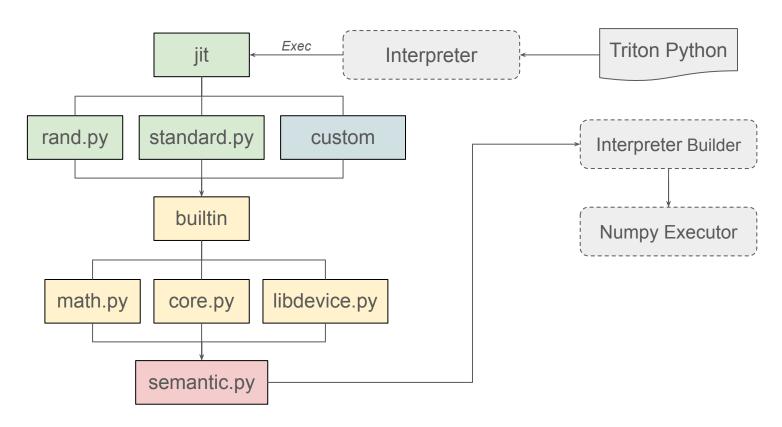
Profile Report Diagnostic Report

Interpreter

Revisit the Frontend



Interpreter



Usage

- Enable the interpreter mode
 - TRITON_INTERPRET=1 <your command>
- Debug with pdb
 - TRITON_INTERPRET=1 pdb test.py
 - b test.py:<line number>
 - r
- Highlights
 - You can set device='cpu' to execute code with the interpreter
 - You can print t1.tensor using the native python print and check all values of the tensor

Acknowledgement

- The Hatchet team
- Special thanks to lan Lumsden

Triton Conference 2024

- The Triton Conference is happening again on September 17th, 2024 in Fremont (CA)
- If you are interested in attending, please fill up this form
 - https://tinyurl.com/4rdfy8s9