

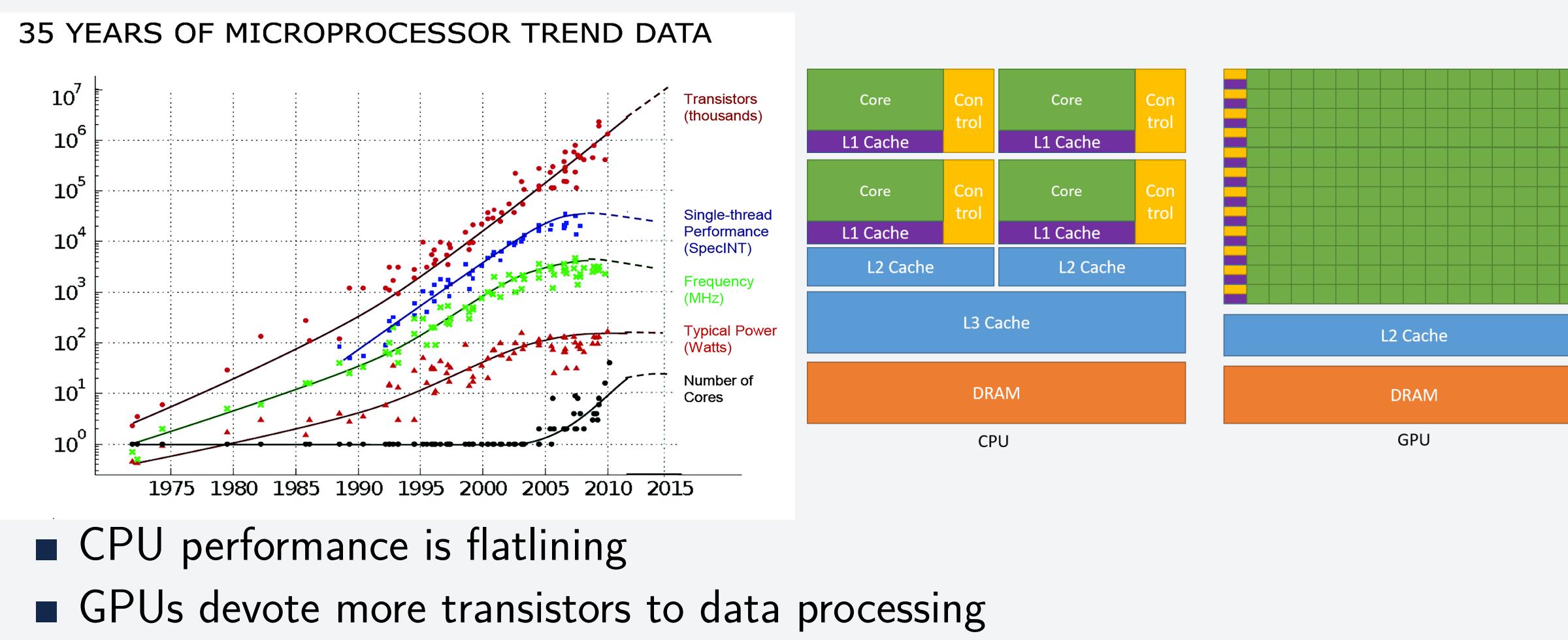
Efficiently Executing NumPy on GPUs via the CUDAGraph API

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Why GPUs?

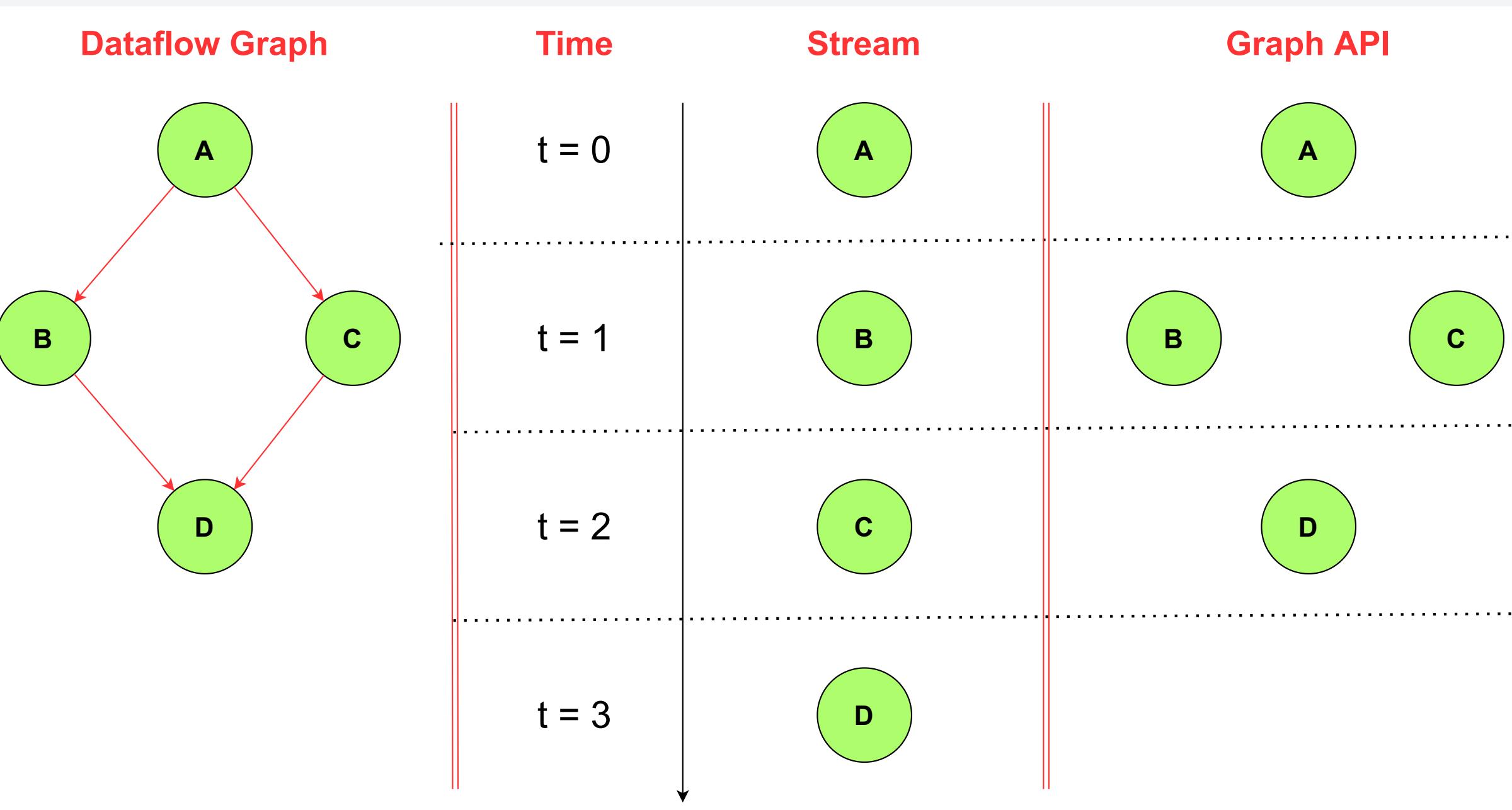


Abstract

- NumPy is the defacto standard for array-based numerical computations in scientific workloads (for ex. PDE solving, Image Processing, etc.)
- However, its ability to harness GPU acceleration is limited by its single-node, occasionally multi-threaded, CPU-only execution model
- Goal: Should be able to efficiently execute array operations on GPUs
- Challenges:
 - Saturating all available execution units to efficiently use the entire GPU
 - NumPy is a high-level programming interface, does not address performance directives such as:
 - Execution grid-size which affects data partitioning
 - User-specified local memory management
 - Kernel launch overhead costs
- Realizing concurrency across array operations**
- Our Approach: Use CUDAGraph API to concurrently launch array operations using tuned kernels that are accessed through PyCUDA
- Past work:
 - Legate: A runtime system for scheduling operations in a task graph
 - Representation lacks a global view of the program limiting the program optimizing space
 - Lazy evaluation: Theano | JAX | PyTorch
 - Requires expensive algorithms (ex. kernel/loop fusion)
 - Ex. Theano[3] claim to have super-linear codegeneration algorithm
 - Single Stream: cuPy | GPUArrays.jl

Why CUDAGraphs?

- CUDA Streams
 - Operations are enqueued in-order into the stream object
- CUDAGraph API
 - Takes in a task dependency graph where each node corresponds to a CUDA kernel
 - Scheduler realizes concurrency across nodes in the graph through multiple streams



Interface Implementation

```
actx=PyCUDAGraphArrayContext()
x = actx.zeros((100,1))
tmp=x+1
result=actx.freeze(tmp)
cuGraphCreate(&m_graph, flags)
cuGraphAddMemAllocNode(&memalloc_x, m_graph)
cuGraphAddMemAllocNode(&memalloc_tmp, m_graph)
cuGraphAddMemSetNode(&memset_node, m_graph, [memalloc_x])
cuGraphAddKernelNode(&k_node, [memset_x, memalloc_tmp], 2)
cuGraphInstantiate(&m_exec, m_graph)
cuGraphLaunch(m_exec)
cuGraphExecDestroy(&m_exec)
cuGraphDestroy(&m_graph)
```

Figure: User Input Program

Figure: Driver C code

- Array operations realized as a composition of CUDA calls (memcpy, kernel_launches, memalloc) that are added onto a task dependency graph with precise edges.
- Object cleanup tied to lifetime of objects in array operations

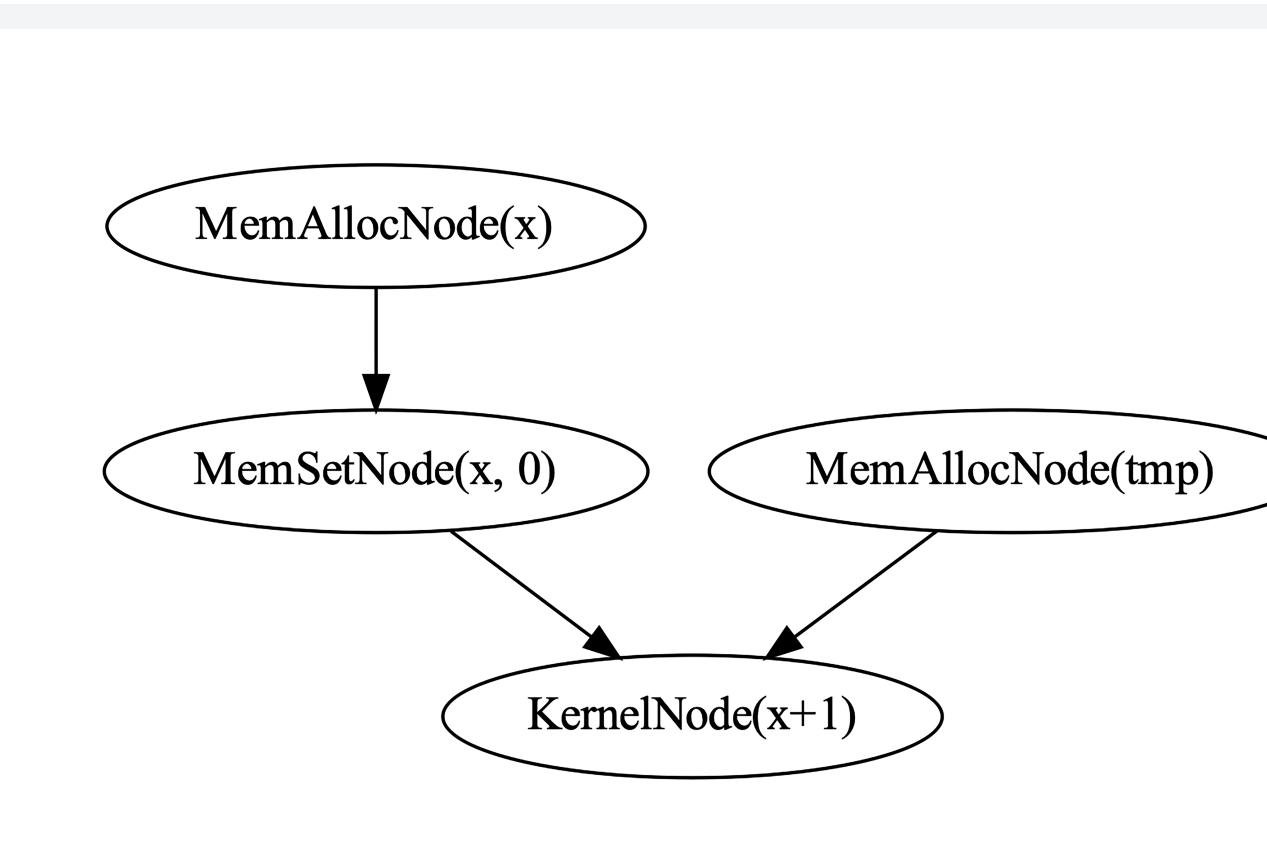


Figure: Generated CUDAGraph

Experimental Setup

We ran a set of image processing algorithms with and without CUDAGraph API on different image batches

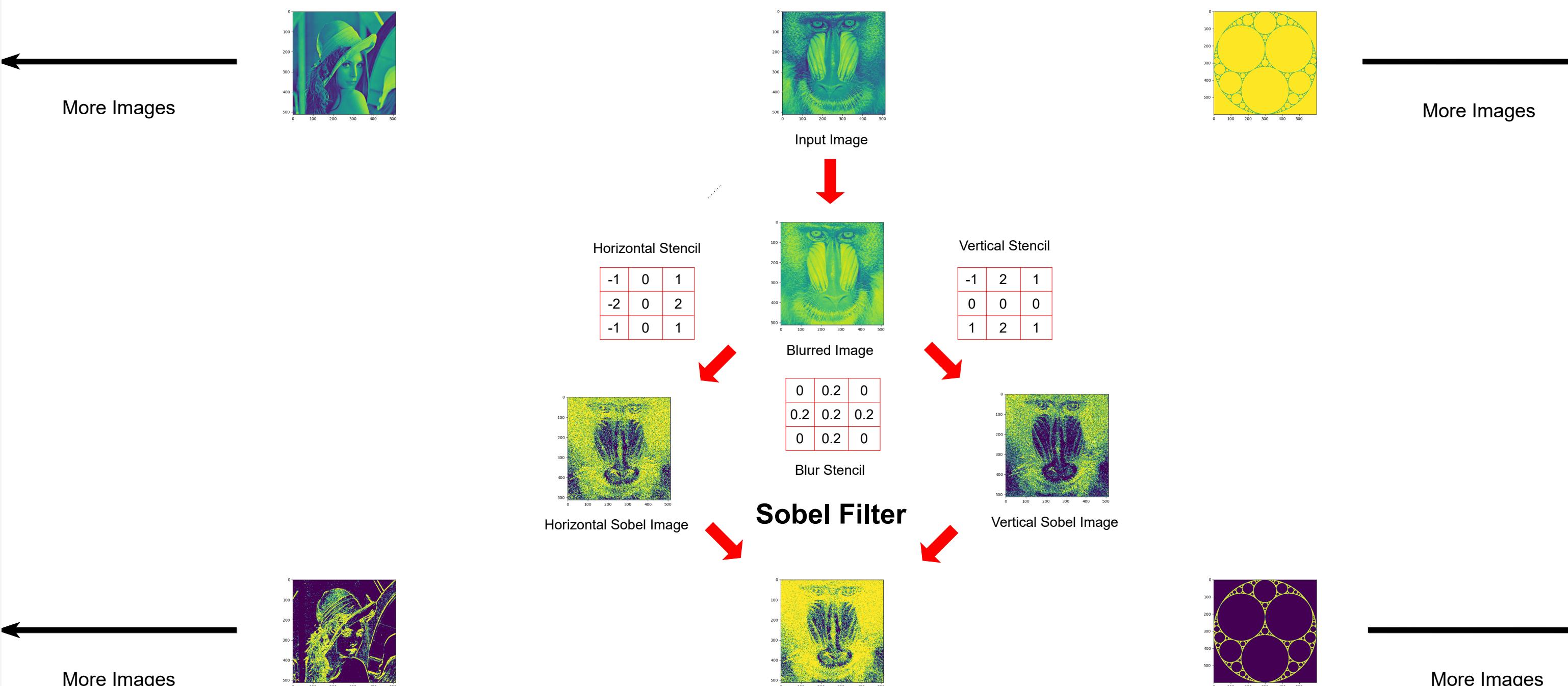


Figure: Sobel Filter Graph

Results

- We observed a speed up of upto 15% on NVIDIA Titan V for smaller problems which was attributed to high task parallelism
- We observed overlapping kernels across multiple streams for the CUDAGraph API program

Graph vs Non-graph comparision for batched sobel operator

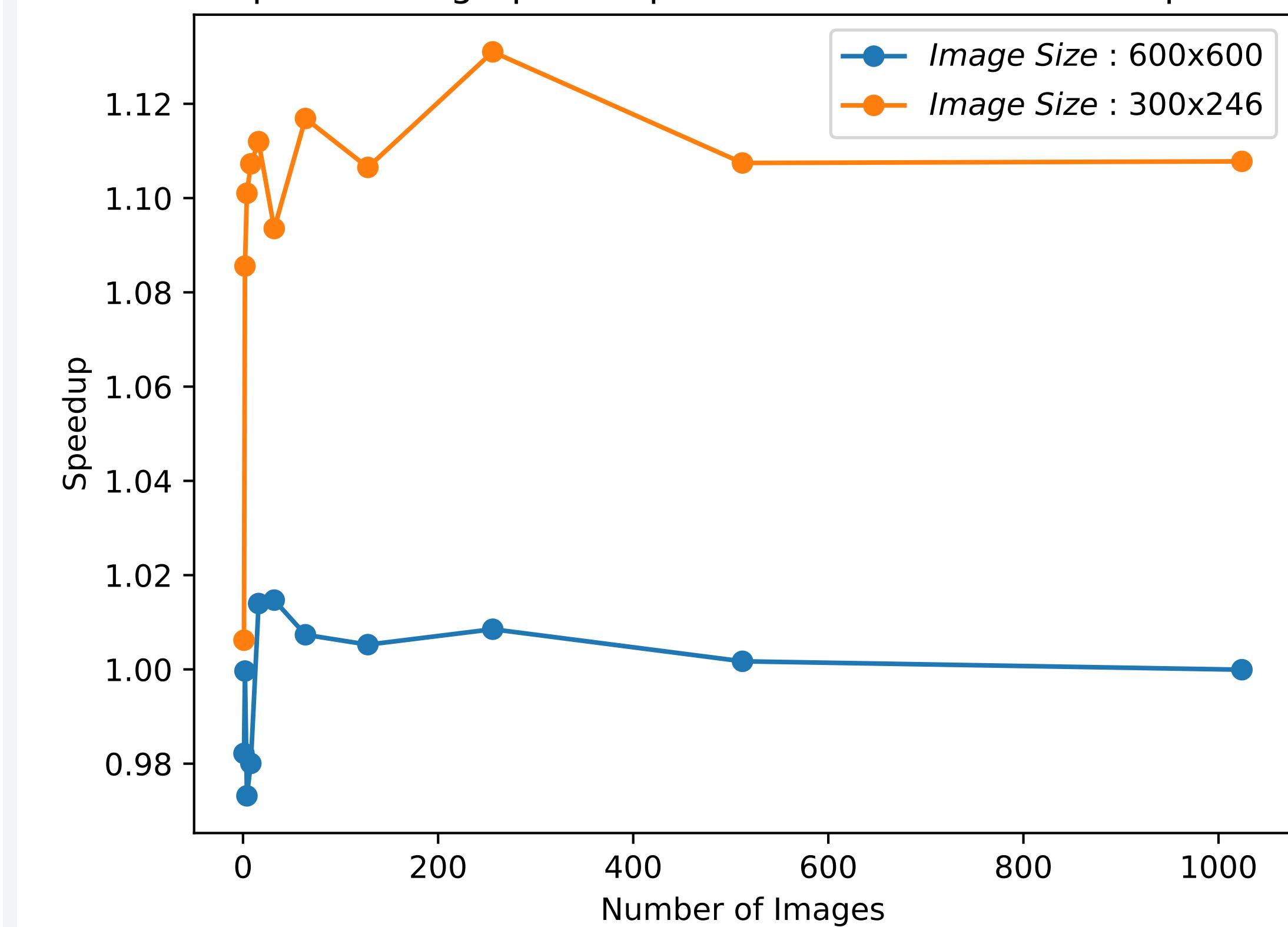


Figure: Sobel Filter on Image Batches

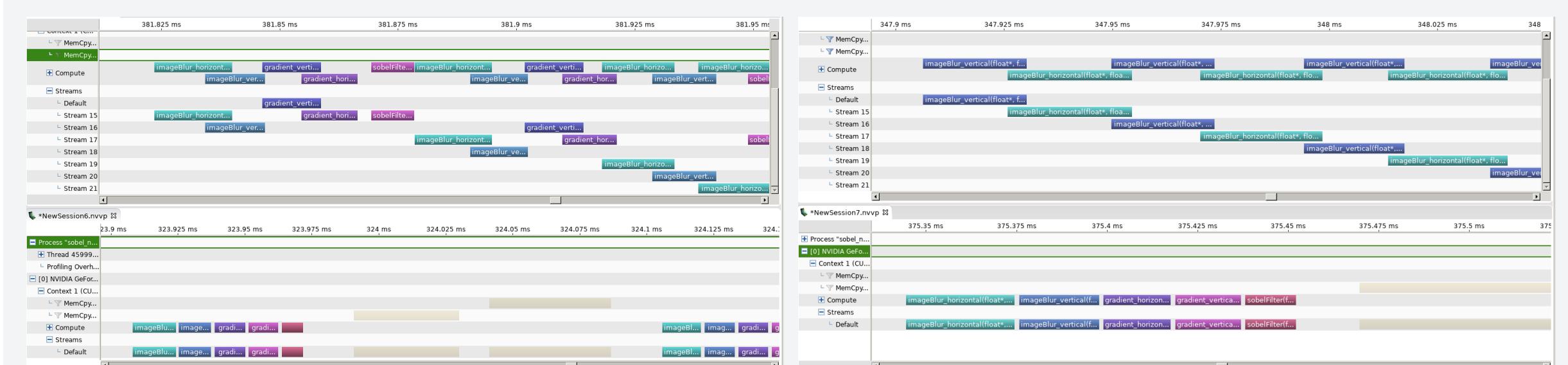


Figure: Kernel Execution timeline with (top) and without (bottom) CUDAGraph API Image size 300 x 246

Figure: Kernel Execution timeline with (top) and without (bottom) CUDAGraph API Image size 600 x 600

Results, run instructions at <https://github.com/mitkotak/sobel>



Future Work

- Upstream work: Integrate with PyCUDA and Arraycontext
- Evaluate this approach on large scale scientific simulations

Acknowledgements

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References

- CUDA Programming: A Developer's Guide to Parallel Computing with GPUs (<https://docs.nvidia.com/cuda/cuda-c-programming-guide/index.html>)
- Chapter 27 - GPU Scripting and Code Generation with PyCUDA (<https://doi.org/10.1016/B978-0-12-385963-1.00027-7>)
- Theano: A Python framework for fast computation of mathematical expressions (<https://doi.org/10.48550/arXiv.1605.02688>)