ASO LAB Seminar
#week6
Triton Server

엄소은

CONTENTS

01. Triton - Multiple Request

02. Local vs. Triton

1. Triton

Multiple Request - Memcpy

```
# and Initialize the data
input_data = np.ndarray([1, 3, 224, 224], dtype=np.float32)
for j in range(len(model_name)):
    inputs[j][0].set_data_from_numpy(input_data, binary_data=False)

cycle = 10
async_requests = [[] for _ in range(cycle)]
for c in range(cycle):
    for i in range(len(model_name)):
        # Asynchronous inference call.
        async_requests[c].append(
            triton_client.async_infer(
                model_name=model_name[i], inputs=inputs[i], outputs=outputs[i]
            )
        )

    for async_request in async_requests[c]:
        # Get the result from the initiated asynchronous inference request.
        # Note the call will block till the server responds.
        result = async_request.get_result()

        print(result.get_response())
```

- **3개의 다른 모델에 100번 request 를 보내는 경우**

Nsight 분석 결과

Memcpy from Host to Device (model parameters..)

input = [...]

for i in range(100):

Memcpy from Host to Device

infer(input, ...)

Memcpy from Device to Host

540	void cutlass::Kernel<cutlass_80_tensorop_s168...	3.63123s	57.856 µs	GPU 0	Stream 30
110	Memcpy HtoD (Pageable)	3.10545s	50.656 µs	GPU 0	Stream 7
206	Memcpy HtoD (Pageable)	3.1844s	50.496 µs	GPU 0	Stream 30
1	Memcpy HtoD (Pageable)	2.5147s	50.432 µs	GPU 0	Stream 7
209	void implicit_convolve_sgemm<float, float, (int)...	3.23443s	50.111 µs	GPU 0	Stream 30
210	void implicit_convolve_sgemm<float, float, (int)...	3.23487s	48.928 µs	GPU 0	Stream 30
213	void implicit_convolve_sgemm<float, float, (int)...	3.26022s	48.896 µs	GPU 0	Stream 30

1. Triton

Multiple Request - 개선여지?

input = [...]

Memcpy from Host to Device

for i in range(100):

    infer(input, ...)

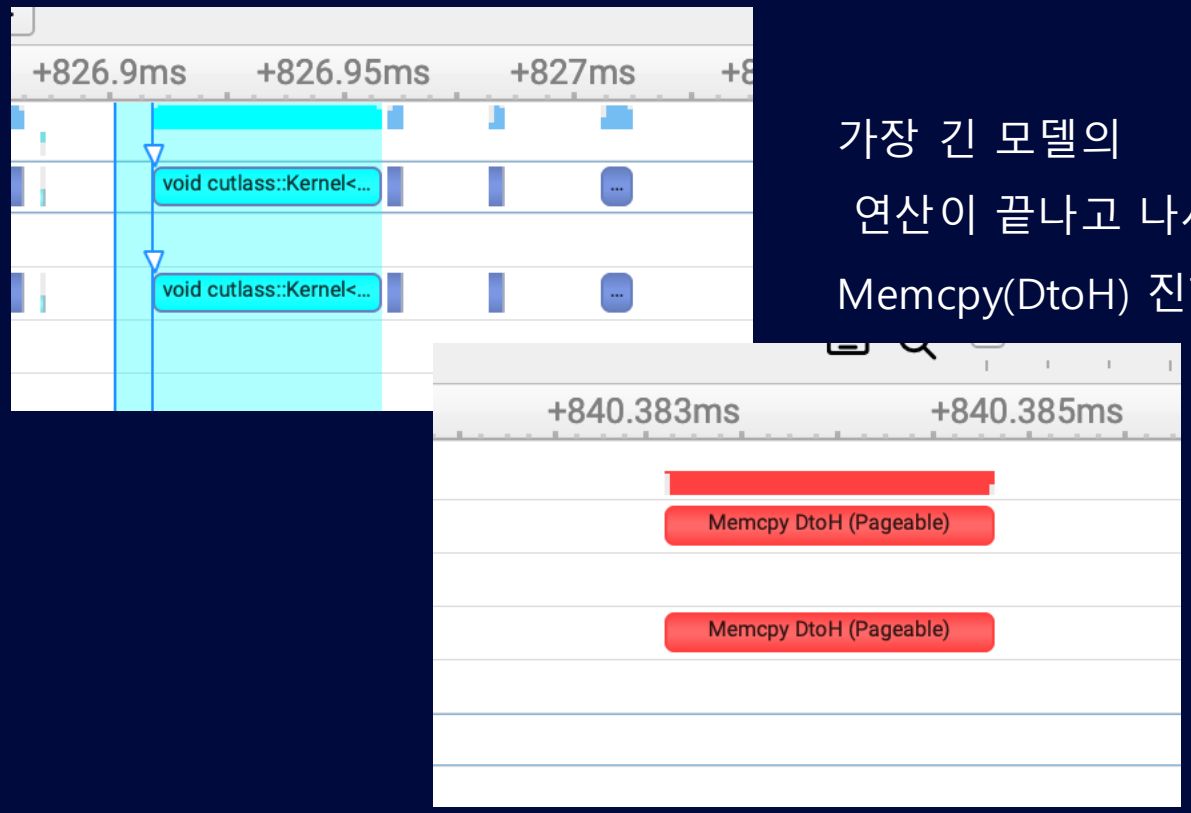
    Memcpy from Device to Host

1. Triton

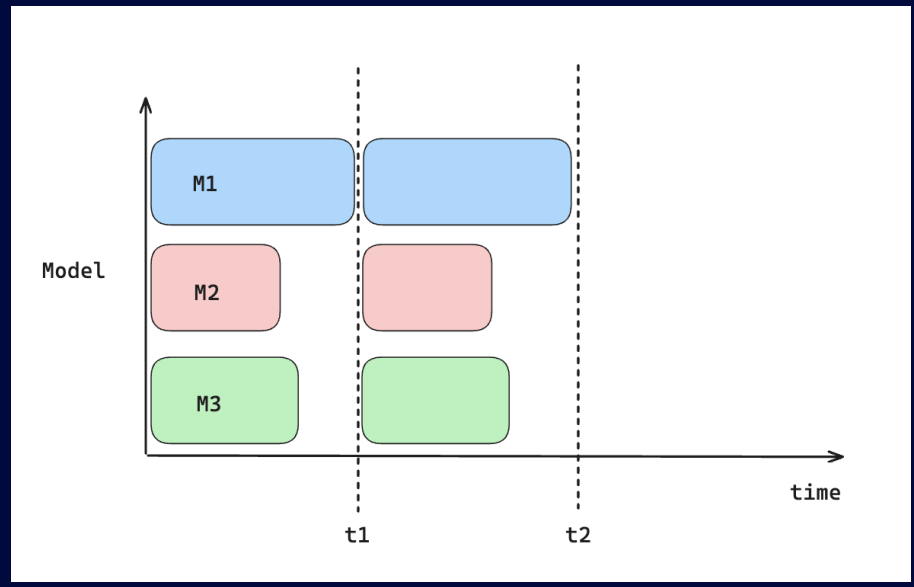
Multiple Request - Synchronization

- 3개의 모델에 100번 infer request 를 보냈을 때

(M1: Stable-Diffusion, M2 : Bert-base-uncased(110M), M3 : VIT-base-patch



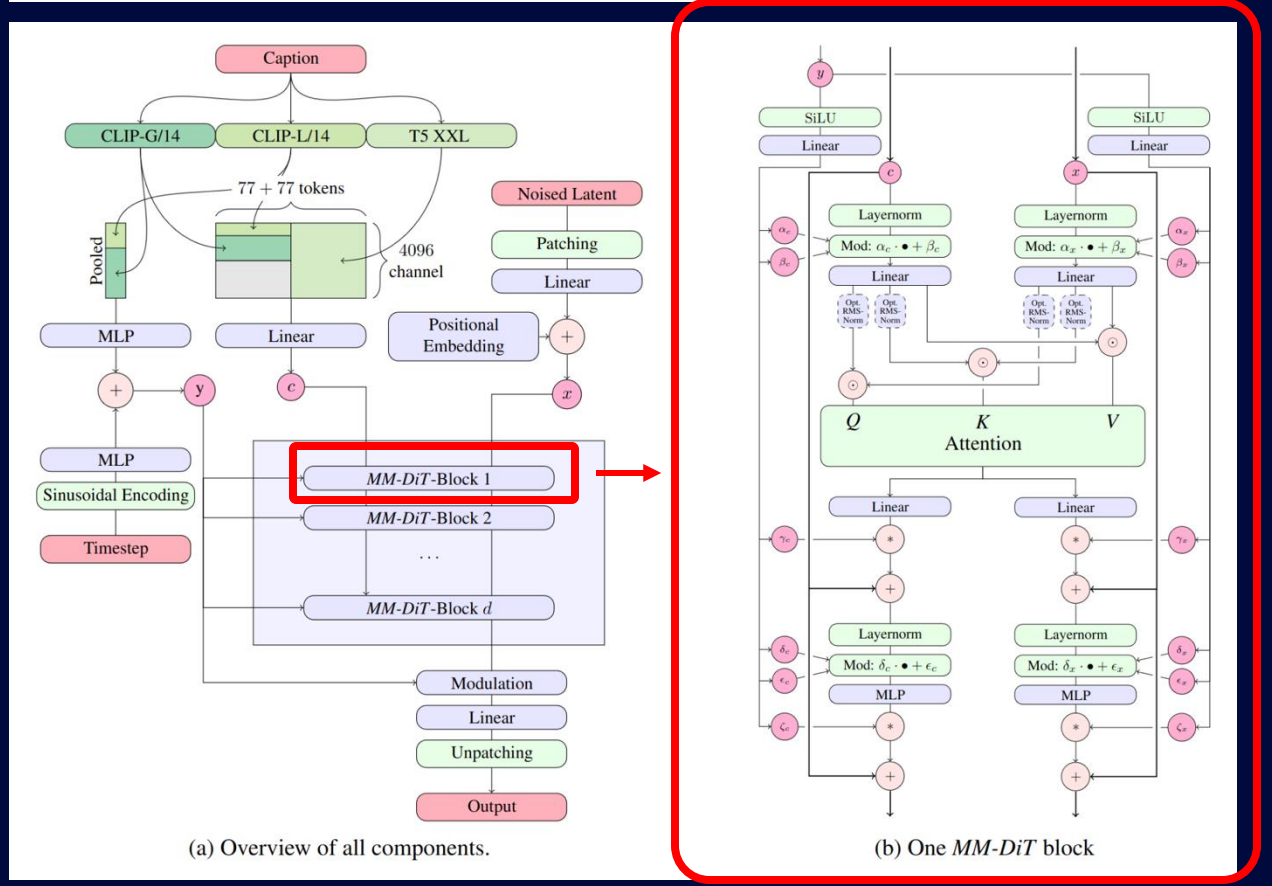
가장 긴 모델의
연산이 끝나고 나서야
Memcpy(DtoH) 진행



2. Nsight local vs. Triton

사용한 모델

stabilityai/**stable-diffusion-3-medium**

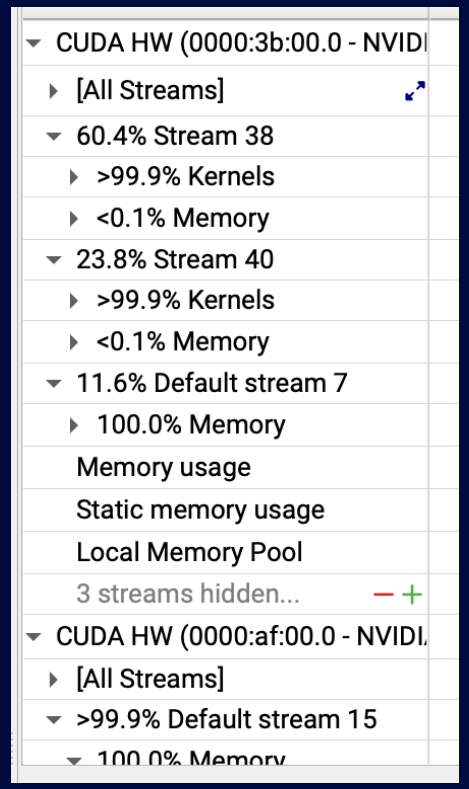


- text to image model

2. Nsight local vs. Triton

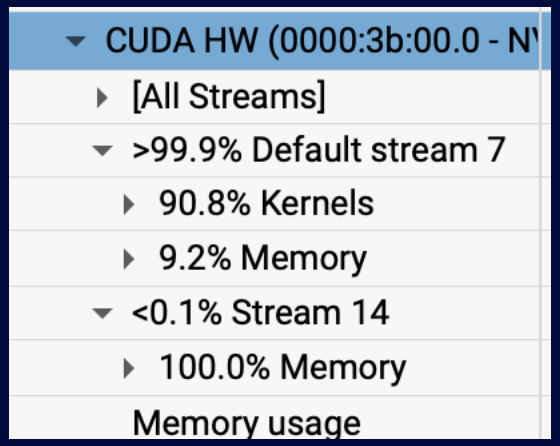
GPU Utilization

- Triton Server -> 2 streams 사용
- Local



Triton Server Profiling

-> 2개의 Stream이 kernel 을 나누어서 작동



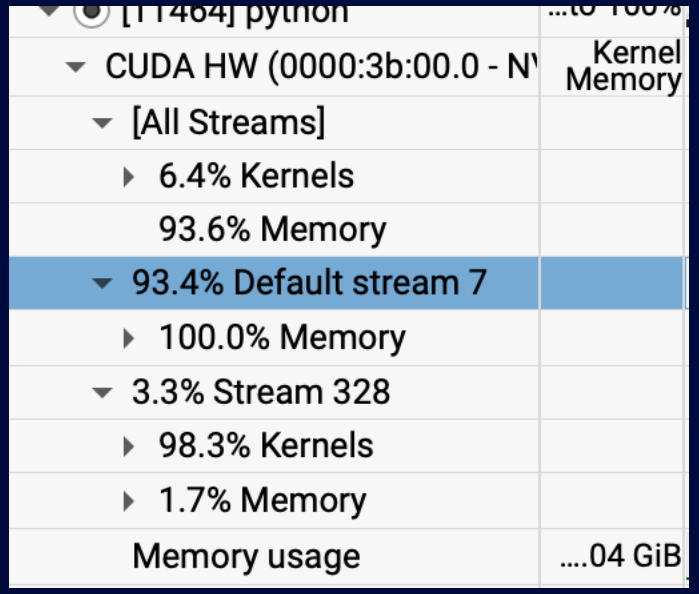
Local Profiling

-> Default stream 에서 커널 + 메모리 복사

2. Nsight local vs. Triton

GPU Utilization

- Local -> 2 stream to 2 device



Local Profiling

```python
# Create two CUDA streams
stream1 = torch.cuda.Stream()
stream2 = torch.cuda.Stream()

# Preprocess the image
input_data = preprocess_image(image_path)

# Convert input to torch tensor and move to GPU using stream1
with torch.cuda.stream(stream1):
    input_data = torch.tensor(input_data).cuda()

# Wait for stream1 to finish moving data to GPU
stream1.synchronize()

# Load the ONNX model with GPU support
ort_session = ort.InferenceSession("./model_repository/vit-base-patch/1/model.onnx", p

# Run inference using stream2
with torch.cuda.stream(stream2):
    ort_inputs = {ort_session.get_inputs()[0].name: input_data.cpu().numpy()}
    ort_outs = ort_session.run(None, ort_inputs)

# Wait for stream2 to finish inference
stream2.synchronize()
```

2. Nsight local vs. Triton

Kernel Execution Patterns

- regular_fft_pad (Fast Fourier Transform (FFT) process - 커널 연산 중 시간이 가장 오래 걸리는 연산)

void DSE::vector_fft<(int)1, (int)2, (int)128, (int)8, (int)8, (int)1, __half, float, float2>...	38.2033s	52.320 µs	GPU 0	Stream 38
void DSE::regular_fft_clip<(int)1, (int)2, (int)128, (int)16, (int)32, (int)1, __half, float,...	38.2034s	56.768 µs	GPU 0	Stream 38
void DSE::regular_fft_pad<(int)0, (int)1, (int)128, (int)16, (int)32, (int)1, __half, float...	38.2036s	121.439 µs	GPU 0	Stream 38
void DSE::regular_fft_pad<(int)0, (int)1, (int)128, (int)16, (int)32, (int)1, __half, float...	38.2037s	26.591 µs	GPU 0	Stream 40
void DSE::vector_fft<(int)0, (int)1, (int)128, (int)8, (int)8, (int)1, __half, float, float2>...	38.2037s	202.590 µs	GPU 0	Stream 38
void DSE::vector_fft<(int)0, (int)1, (int)128, (int)8, (int)8, (int)1, __half, float, float2>...	38.2039s	13.856 µs	GPU 0	Stream 40

Triton : 38, 40 stream 에서 concurrent 하게 계산

void cutlass::Kernel<cutlass_80_wmma_tensorop_f...	7.08241s	7.009 µs	GPU 0	Stream 7
void cutlass::Kernel<cutlass_80_wmma_tensorop_f...	7.08253s	7.040 µs	GPU 0	Stream 7
void cutlass::Kernel<cutlass_80_wmma_tensorop_f...	7.08263s	6.849 µs	GPU 0	Stream 7
void at::native::vectorized_elementwise_kernel<(int)...	7.08281s	1.184 µs	GPU 0	Stream 7
void at::native::elementwise_kernel<(int)128, (int)4, ...	7.0831s	3.168 µs	GPU 0	Stream 7
fmha_cutlassF_f16_aligned_64x64_rf_sm80(PyTor...	7.08331s	12.224 µs	GPU 0	Stream 7
void cutlass::Kernel<cutlass_80_wmma_tensorop_f...	7.0836s	6.849 µs	GPU 0	Stream 7

Local : 하나의 스트림에서만 계산 (동기화 문제..)

3. Plan

- Triton Server code Build + (오류 해결하기)
- 앞의 문제 관련 코드 찾고 고쳐서 실행해보기