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ECE 408/CS483 Milestone 2 Report

1. Show output of rai running Mini-DNN on the basic GPU convolution implementation for batch size of 1k images. This can either be a screen capture or a text copy of the running output. Please do not show the build output. (The running output should be everything including and after the line "Loading fashion-mnist data...Done").

Loading model...Done

Conv-GPU==

Layer Time: 95.2482 ms Op Time: 4.65613 ms

Conv-GPU==

Layer Time: 96.4938 ms Op Time: 29.7385 ms

Test Accuracy: 0.886

real 0m9.719s user 0m9.309s sys 0m0.372s

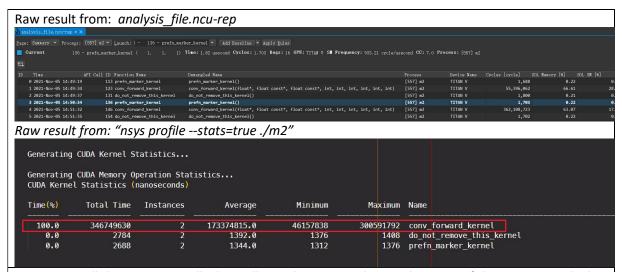
2. For the basic GPU implementation, list Op Times, whole program execution time, and accuracy for batch size of 100, 1k, and 10k images.

Batch Size	Op Time 1 (ms)	Op Time 2 (ms)	Total Execution Time (s)	Accuracy (%)	
100	0.479951	2.93208	1.169	0.86	
1000	4.65613	29.7385	9.719	0.886	
10000 46.1701		300.715	1m38.061s	0.8714	

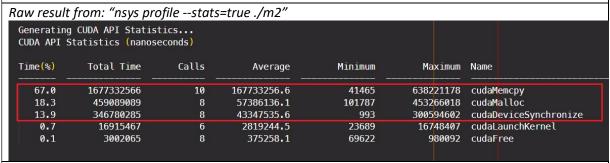
3. List all the kernels that collectively consumed more than 90% of the kernel time and what percentage of the kernel time each kernel did consume (start with the kernel that consumed the most time, then list the next kernel, until you reach 90% or more).

Ordered result:

	ID Time	API Call ID Function Name	Demangled Name	Process	Device Name	Cycles [cycle]	SOL Memory [%]	SOL SM [%]	time consume [%]
	4 2021-Nov-05 14:51:32	146 conv_forward_kernel	conv_forward_kernel(float*, float const*, float const*, int, int, int, int, int, int)	[557] m2	TITAN V	362,108,723	63.07	17.61	86.7302023
	1 2021-Nov-05 14:49:34	123 conv_forward_kernel	conv_forward_kernel(float*, float const*, float const*, int, int, int, int, int, int)	[557] m2	TITAN V	55,396,062	66.61	28.71	13.268146
	2 2021-Nov-05 14:49:37	131 do_not_remove_this_kernel	do_not_remove_this_kernel()	[557] m2	TITAN V	1,800	0.21	0	0.00043112
	3 2021-Nov-05 14:50:34	136 prefn_marker_kernel	prefn_marker_kernel()	[557] m2	TITAN V	1,703	0.22	0	0.00040789
	5 2021-Nov-05 14:51:35	154 do_not_remove_this_kernel	do_not_remove_this_kernel()	[557] m2	TITAN V	1,702	0.22	0	0.0004076
1	0 2021-Nov-05 14:49:19	113 prefn_marker_kernel	prefn_marker_kernel()	[557] m2	TITAN V	1,688	0.22	0	0.000404



4. List all the CUDA API calls that collectively consumed more than 90% of the API time and what percentage of the API time each call did consume (start with the API call that consumed the most time, then list the next call, until you reach 90% or more).

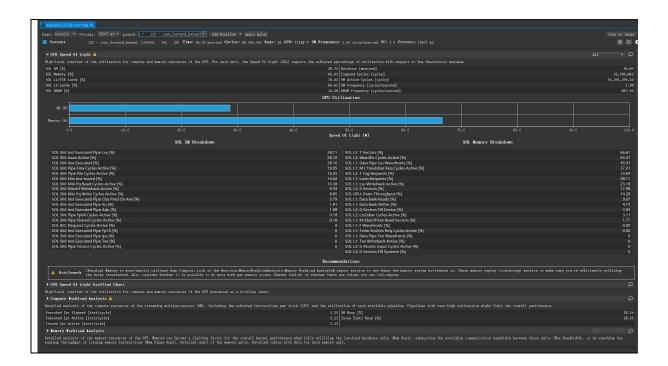


5. Explain the difference between kernels and CUDA API calls. Please give an example in your explanation for both.

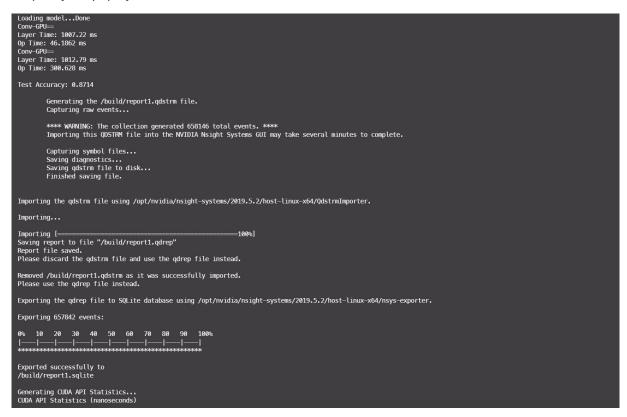
Kernels are the code that the GPU actually runs to realize certain functionalities. They will various significantly when the goals of projects change. For example, the function "conv_forward_kerne()!" used in m2.

CUDA API contains some functions that help to connect CPU (or your computer) and GPU. They do some basic things which make it possible for you to control the behavior of GPU, like allocate, free memory, copy data from one to another and Synchronize between devices. For example, the function "cudaDeviceSynchronize()" used in m2.

6. Show a screenshot of the GPU SOL utilization



Output of "nsys profile --stats=true ./m2"



Time(%)	Total Time	Calls	Average	Minimum	Maximum	Name				
67.0	1677332566	10	167733256.6	41465	638221178	cudaMemcpy				
18.3	459089089	8	57386136.1	101787	453266018					
13.9	346780285	8	43347535.6	993	300594602					
0.7	16915467	6	2819244.5	23689	16748407					
0.1	3002065	8	375258.1	69622		cudaFree				
	g CUDA Kernel S									
	g CUDA Memory (el Statistics		istics							
Time(%)	Total Time	Instances	Average	Minimum	Maximum	Name				
100.0	346749630		173374815.0	46157838	300591792	conv_forward_kernel				
0.0	2784		1392.0	1376	1408	do_not_remove_this_kernel				
0.0	2688		1344.0	1312	1376	prefn_marker_kernel				
CUDA Memo	CUDA Memory Operation Statistics (nanoseconds)									
Time(%)	Total Time	Operations	Average	Minimum	Maximum	Name				
67.4	1120586563	2	560293281.5	483320494	637266069	[CUDA memcpy DtoH]				
32.6	542066049	8	67758256.1	1216	269151198					
CUDA Memo	CUDA Memory Operation Statistics (KiB)									
	Total (Operations	Average	Minimum	М	Maximum Name				
2	261419.0	8	282677.0	0.004	100	00000.0 [CUDA memcpy HtoD]				
	722500.0	2	861250.0	722500.000		00000.0 [CUDA memcpy DtoH]				

Generating Operating	System Ru	untime API	Statistics
Operating System Pun	time ADT C	Statistics	(nanoceconde)

Time(%)	Total Time	Calls	Average	Minimum	Maximum	Name
33.3	94027562572	954	98561386.3	49708	100193755	sem timedwait
33.3	93894246699	952	98628410.4	75709	100259249	
21.6	60906123006	2	30453061503.0	22198308270		pthread_cond_wait
11.7	33009721948	66	500147302.2	500090916		pthread_cond_timedwait
0.0	115986185	912	127177.8	1078	21461719	
0.0	37690341	9072	4154.6	1179	8683539	read
0.0	20057132	26	771428.2	1487	19951598	fopen
0.0	3006839	98	30682.0	1427	1193676	
0.0	1294766	101	12819.5	7174	48971	open64
0.0	444303	2	222151.5	36594		pthread_mutex_lock
0.0	424552		84910.4	53027	139217	pthread create
0.0	187169		62389.7	55785	68150	fgets
0.0	185216		61738.7	8459	155851	fopen64
0.0	117639		6919.9	1293	21175	munmap
0.0	70904	15	4726.9	2288		write
0.0	55125		6125.0	1032		fflush
0.0	42396		8479.2	6001	11059	open
0.0	42275	20	2113.8	1078	7284	fclose
0.0	19699	8	2462.4	1045	10664	fcntl
0.0	18939		9469.5	7580	11359	socket
0.0	13015		4338.3	1137	10454	fwrite
0.0	10689		10689.0	10689	10689	connect
0.0	7226		7226.0	7226	7226	pipe2
0.0	5371		5371.0	5371	5371	pthread_cond_signal
0.0	1998		1998.0	1998	1998	bind
0.0	1433		1433.0	1433	1433	listen

Generating NVTX Push—Pop Range Statistics... NVTX Push—Pop Range Statistics (nanoseconds)