GitHub: juliewiltbank

ECE 60827: Lab 1 Report

SAXPY Execution:

The SAXPY kernel profiling data is shown below for three different inputs of VectorSize (see Table 1). The inputs all vary dramatically ranging from 1024 to 131072.

		SAXPY Program			
Input Size	Kernel / API Call	Execution Time (ms)	Percentage of Total Time (%)		GPU
	1024 cudaMemcpy HtoD	2.59E-03		38.94	API
	1024 cudaMemcpy DtoH	2.02E-03		30.77	
	1024 saxpy_gpu()	2.05E-03		30.29	
	1024 cudaMalloc	110.43		94.21	
	1024 cudaLaunchKernel	6.40E+00		5.46	
	32768 cudaMemcpy HtoD	2.62E-02		66.21	
	32768 cudaMemcpy DtoH	1.11E-02		27.93	
	32768 saxpy_gpu()	2.34E-03		5.86	
	32768 cudaMalloc	92.967		93.32	
	32768 cudaLaunchKernel	6.0815		6.1	
	131072 cudaMemcpy HtoD	9.05E-02		66.99	
	131072 cudaMemcpy DtoH	4.12E-02		30.55	
	131072 saxpy_gpu()	3.33E-03		2.46	
	131072 cudaMalloc	103.37		92.32	
	131072 cudaLaunchKernel	6.7579		6.04	

Table 1: SAXPY Raw Data from nvprof Profiling Tool

This data is further synthesized into total execution time from both Kernel and API calls for each input size (see Table 2). A notable trend is that the GPU execution time scales with input size but the API calls took largely the same time despite scaling the input size. Indeed when profiling the data, the input size for API made no difference in the time it took and varied by ~20ms randomly each program run through.

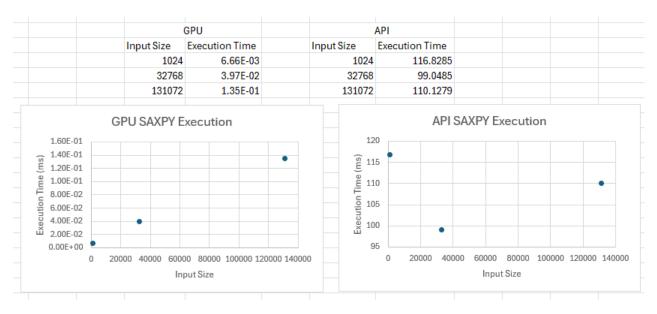


Table 2: SAXPY Kernel and API Execution Time for Various Input Sizes

The breakdown of top time consumers for Kernel and API for each input size is shown below in Graph 1. The biggest consumers across input size were memory operations. For the API cudaMalloc always took the most time and for the Kernel cudaMemcpy HtoD was the biggest consumer (though to a lesser extent).



Monte-Carlo Execution:

The raw data from profiling the Monte-Carlo program is shown below in Table 3. For three input sizes (sampleSize), the data ranges from 1000 to 1E9. The time execution took increased greatly between 1E6 and 1E9 but not between 1000 and 1E6.

Monte-Carlo Program		
Execution Time (ms)	Percentage of Total Time (%)	GPU
2.30E-03	1.06	API
8.45E-03	3.9	
2.06E-01	95.03	
104.48	93.96	
6.3282	5.69	
3.24E+00	4.88	
6.32E+01	95.11	
108.19	59.33	
67.58	37.06	
6.2055	3.4	
2.67E+03	4.52	
5.63E+04	95.46	
109.34	0.19	
5.89E+04	99.79	
	2.30E-03 8.45E-03 2.06E-01 104.48 6.3282 3.24E+00 6.32E+01 108.19 67.58 6.2055 2.67E+03 5.63E+04	Execution Time (ms) Percentage of Total Time (%) 2.30E-03 1.06 8.45E-03 3.9 2.06E-01 95.03 104.48 93.96 6.3282 5.69 3.24E+00 4.88 6.32E+01 95.11 108.19 59.33 67.58 37.06 6.2055 3.4 2.67E+03 4.52 5.63E+04 95.46 109.34 0.19

Table 3: Monte-Carlo Raw Data from nvprof Profiling Tool

This gap in size can be shown clearly in Table 4. The smaller two input sizes merged into one data point near the origin, dwarfed by the larger input size and it's much longer execution time. Interestingly, this program shows that varying sampleSize in the code affects both API and Kernel unlike the SAXPY program which largely only affected Kernel calls.

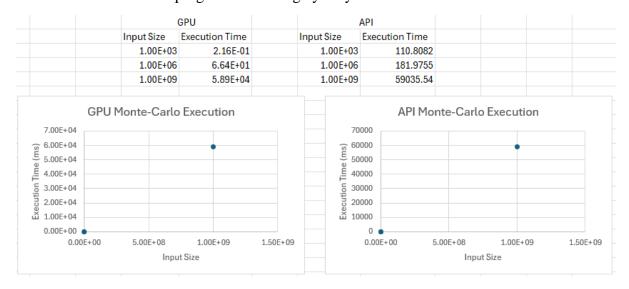


Table 4: Monte-Carlo Kernel and API Execution Time for Various Input Sizes

The breakdown of top time consumers for Kernel and API for each input size in the Monte-Carlo program is shown below in Graph 2. The biggest consumers across input size were not always memory operations unlike SAXPY. For the API cudaMalloc started out as the biggest consumer of time but then steadily decreases with input size and cudaMemcpy took over. For the Kernel generatePoints() was the biggest consumer every time.



Graph 2: Monte-Carlo Execution Time Breakdown by Top Consumers