# Lab 0 Report

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## **Questions**

Input size is 1,000,000 and 1M % 512 = 64 so all warps are fully active at the beginning.

1. For the naive reduction kernel, how many steps execute without divergence? How many steps execute with divergence?

#### Answer:

Given the BLOCKSIZE is 512 (a.k.a. 2<sup>9</sup>), so 9 steps will be executed. Only the first step is without divergence since all the threads follow the same instruction to compute the sum. All other 8 steps are executed with divergence because 1 thread out of 2 will have nothing to do after the first step.

2. For the optimized reduction kernel, how many steps execute without divergence? How many steps execute with divergence?

### Answer:

Before the number of effective threads becomes smaller than a warp, there is no divergence at all. Each warp contains 32 threads therefore first 4 steps are without divergence and last 5 steps have divergence.

3. Which kernel performed better? (for both real GPUs and GPGPU-Sim) Answer:

Table 1 Execution Time (seconds)

	Naïve Optimized		
Bender	0.000281	0.000154	
GPGPU-Sim	113	89	

The optimized version is better in performance.

4. How does the warp occupancy distribution compare between the two Reduction implementations? Answer:

Table 2 Warp Occupancy Distribution

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	Stall	W0_Idle	W0_SB	W1	W2	W3	W4	W5	W6		
Naïve	146546	56748	315050	369306	187584	0	187584	0	0		
Opt-ed	93304	97096	378426	13678	7816	0	7816	0	0		
	W7	W8	W9	W10	W11	W12	W13	W14	W15		
Naïve	0	187584	0	0	0	0	0	0	0		
Opt-ed	0	7816	0	0	0	0	0	0	0		
	W16	W17	W18	W19	W20	W21	W22	W23	W24		
Naïve	187584	0	0	0	0	0	0	0	0		
Opt-ed	7816	0	0	0	0	0	0	0	0		

Table 2 Warp Occupancy Distribution (Cont.)

	W25	W26	W27	W28	W29	W30	W31	W32	
Naïve	0	0	0	0	0	0	0	2125882	
Opt-ed	0	0	0	0	0	0	0	2024274	

This table can reveal the reason of performance gap between two solution. The optimized version has less stalled warps and the warps that are not fully occupied. It means that the utilization rate is higher in optimized version.

### 5. Why do GPGPUs suffer from warp divergence?

#### Answer

The nature of SIMT/SIMD model is that all data are processed under the same code. If there are two possible paths to go through, which means divergence, they will be serialized. The serialization undermines the performance.