Report Common Assignment 3



Counting sort

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Index

In	dex		i
1	Pro	blem description	1
	1.1	Experimental setup	1
		1.1.1 Hardware	2
		1.1.2 Software	9
2	Per	formance	10
	2.1	Case study n°1 - Global Memory	11
	2.2	Case study n°2 - Shared Memory	12
	2.3	Case study n°3 - Texture Memory	13
3	Con	nsiderations	14
	3.1	Comparison	14
	3.2	Premises	15
4	Нох	y to run	16

Problem description

Parallelize and Evaluate Performances of "Counting Sort" Algorithm , by using CUDA.

Counting sort is an algorithm for sorting integer numbers. The computational complexity is equal to O(n).

In this report we analyze counting sort algorithm: get execution times, bandwidth and other performances measures when of using:

- Global Memory
- Shared Memory
- Texture Memory

Evaluate performances with different block-grid configurations.

1.1 Experimental setup

All measurements were made using the Google Colab service, with the following setup.

1.1.1 Hardware

CPU

processor : 0 vendor id : GenuineIntel cpu family : 6 model : 63 model name : Intel(R) Xeon(R) CPU @ 2.30GHz stepping : 0 microcode : 0x1: 2299.998 cpu MHz cache size : 46080 KB physical id: 0 siblings core id : 0 cpu cores : 1 : 0 apicid initial apicid : 0 fpu : yes fpu_exception : yes cpuid level : 13 wp : yes flags : fpu vme de pse tsc msr pae mce cx8 apic sep mtrr pge mca cmov pat pse36 clflush mmx fxsr sse sse2 ss ht syscall nx pdpe1gb rdtscp lm constant_tsc rep_good nopl xtopology nonstop_tsc cpuid tsc_known_freq_pni_pclmulqdq_ssse3 fma cx16 pcid sse4_1 sse4_2 x2apic movbe popent aes xsave avx f16c

rdrand hypervisor lahf_lm abm invpcid_single

ssbd ibrs ibpb stibp fsgsbase tsc_adjust

bmil avx2 smep bmi2 erms invpcid xsaveopt

arat md_clear arch_capabilities

bugs : cpu_meltdown spectre_v1 spectre_v2

spec_store_bypass l1tf mds swapgs

bogomips : 4599.99

clflush size : 64 cache alignment : 64

address sizes : 46 bits physical, 48 bits virtual

power management:

processor : 1

vendor id : GenuineIntel

cpu family : 6

model : 63

model name : Intel(R) Xeon(R) CPU @ 2.30GHz

: 0 stepping

microcode : 0x1

cpu MHz : 2299.998

cache size : 46080 KB

physical id: 0 siblings : 2

core id : 0

cpu cores : 1

apicid : 1 initial apicid : 1

fpu : yes

fpu_exception : yes

cpuid level : 13

wp : yes

flags : fpu vme de pse tsc msr pae mce cx8 apic

sep mtrr pge mca cmov pat pse36 clflush mmx fxsr sse sse2 ss ht syscall nx pdpe1gb $rdtscp \ lm \ constant_tsc \ rep_good \ nopl$ xtopology nonstop_tsc cpuid tsc_known_freq pni pclmulqdq ssse3 fma cx16 pcid sse4_1 sse4_2 x2apic movbe popcnt aes xsave avx f16c rdrand hypervisor lahf lm abm invpcid_single ssbd ibrs ibpb stibp fsgsbase tsc_adjust bmi1 avx2 smep bmi2 erms invpcid xsaveopt arat md_clear arch_capabilities

bugs : cpu_meltdown spectre_v1 spectre_v2

spec store bypass l1tf mds swapgs

bogomips : 4599.99

clflush size : 64 cache alignment : 64

address sizes : 46 bits physical, 48 bits virtual

power management:

RAM

MemTotal:	13302920	kВ
MemFree:	236348	kВ
Mem A vailable:	12350076	kВ
Buffers:	283016	kВ
Cached:	11566856	kВ
Swap Cached:	0	kВ
Active:	3457004	kВ
Inactive:	8911716	kВ
Active (anon):	476972	kВ
Inactive (anon):	480	kВ
Active(file):	2980032	kВ
Inactive (file):	8911236	kВ
Une victable:	0	kВ
Mlocked:	0	kВ
SwapTotal:	0	kВ
SwapFree:	0	kВ
Dirty:	200	kВ
Writeback:	0	kВ
AnonPages:	518888	kВ
Mapped:	269548	kВ
Shmem:	1228	kВ
KReclaimable:	548408	kВ
Slab:	600016	kВ
${\bf SReclaimable}$:	548408	kВ
SUnreclaim:	51608	kВ
KernelStack:	5904	kВ
PageTables:	7172	kВ
$NFS_Unstable:$	0	kВ
Bounce:	0	kВ
Write back Tmp:	0	kВ

CommitLimit: 6651460 kB

Committed_AS: 3798912 kB

VmallocTotal: 34359738367 kB

VmallocUsed: 46004 kB

VmallocChunk: 0 kB

Percpu: 1448 kB

AnonHugePages: 0 kB

ShmemHugePages: 0 kB

ShmemPmdMapped: 0 kB

FileHugePages: 0 kB

FilePmdMapped: 0 kB

CmaTotal: 0 kB

CmaFree: 0 kB

HugePages_Total: 0

HugePages_Free: 0

HugePages_Rsvd: 0

HugePages_Surp: 0

Hugepagesize: 2048 kB

Hugetlb: 0 kB

DirectMap4k: 279360 kB

DirectMap2M: 9154560 kB

DirectMap1G: 6291456 kB

GPU

Device number: 0

Device name: Tesla K80 Compute capability: 3.7

Clock Rate: 823500 kHz

Total SMs: 13

Shared Memory Per SM: 114688 bytes

Registers Per SM: 131072 32-bit

Max threads per SM: 2048

L2 Cache Size: 1572864 bytes

Total Global Memory: 11996954624 bytes

Memory Clock Rate: 2505000 kHz

Max threads per block: 1024

Max threads in X-dimension of block: 1024

Max threads in Y-dimension of block: 1024

Max threads in Z-dimension of block: 64

Max blocks in X-dimension of grid: 2147483647

Max blocks in Y-dimension of grid: 65535

Max blocks in Z-dimension of grid: 65535

Shared Memory Per Block: 49152 bytes

Registers Per Block: 65536 32-bit

Warp size: 32

BANDWIDTH

Device 0: Tesla K80

Range Mode

Host to Device Bandwidth, 1 Device(s)

PINNED Memory Transfers

Transfer	Size	(Bytes)	$\operatorname{Bandwidth}\left(\operatorname{MB/s}\right)$
1000			278.7
101000			6240.3
201000			6613.4
301000			6923.8
401000			7064.9
501000			7197.9
601000			7046.2
701000			7217.7
801000			7333.4
901000			7454.2

Device to Host Bandwidth, 1 Device(s)

PINNED Memory Transfers

Transfer Size (Bytes)	$\operatorname{Bandwidth}\left(\operatorname{MB/s}\right)$
1000	419.3
101000	6563.5
201000	7137.2
301000	7351.7
401000	7361.0
501000	7443.2
601000	7505.5
701000	7519.7
801000	7579.2
901000	7571.1

Device to Device Bandwidth, 1 Device(s)

PINNED Memory Transfers

Transfer Size (Bytes)	$\operatorname{Bandwidth}\left(\operatorname{MB/s}\right)$
1000	251.6
101000	25527.1
201000	44132.9
301000	57341.3
401000	75400.8
501000	83089.5
601000	89331.3
701000	102785.9
801000	91342.3
901000	80354.0

1.1.2 Software

• NVIDIA-SMI 510.39.01

• Driver Version: 460.32.03

• CUDA Version: 11.2

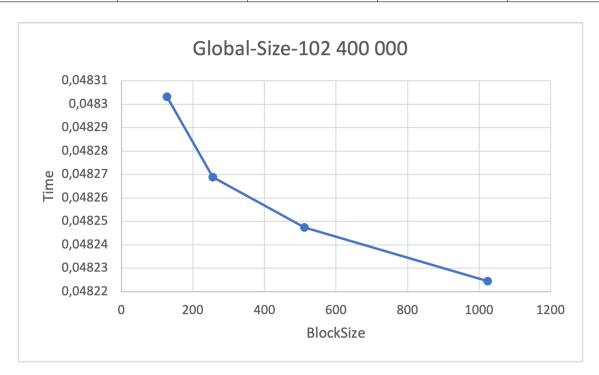
Performance

2.1 Case study n°1 - Global Memory

In this case study, we have analyzed the counting sort algorithm using global memory. The size of the problem is 102 400 000 and than we evaluate the performances with block size of:

- 128: with this choice we have that 2048/128 = 16 blocks. The occupancy is 100%.
- 256: with this choice we have that 2048/256 = 8 blocks. The occupancy is 100%.
- 512: with this choice we have that 2048/512 = 4 blocks. The occupancy is 100%.
- 1024: with this choice we have that 2048/1024 = 2 blocks. The occupancy is 100%.

N	BlockSize	GridSize	mips	time_sec
102 400 000	128	800 000	1 945 600 655	0,04830324
102 400 000	256	400 000	1 945 601 167	0,04826894
102 400 000	512	200 000	1 945 602 191	0,04824748
102 400 000	1024	100 000	1 945 604 239	0,04822454

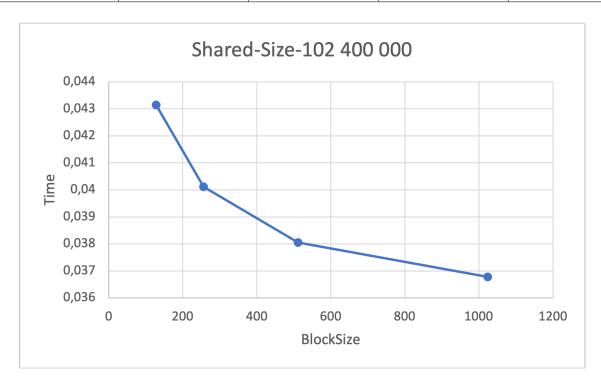


2.2 Case study n°2 - Shared Memory

In this case study, we have analyzed the counting sort algorithm using shared memory. The size of the problem is 102 400 000 and than we evaluate the performances with block size of:

- 128: with this choice we have that 2048/128 = 16 blocks. The occupancy is 100%.
- 256: with this choice we have that 2048/256 = 8 blocks. The occupancy is 100%.
- 512: with this choice we have that 2048/512 = 4 blocks. The occupancy is 100%.
- 1024: with this choice we have that 2048/1024 = 2 blocks. The occupancy is 100%.

N	BlockSize	GridSize	mips	time_sec
102 400 000	128	800 000	2 272 804 239	0,04314758
102 400 000	256	400 000	2 292 802 191	0,04011825
102 400 000	512	200 000	2 332 801 167	0,03805279
102 400 000	1024	100 000	2 412 800 655	0,03677322

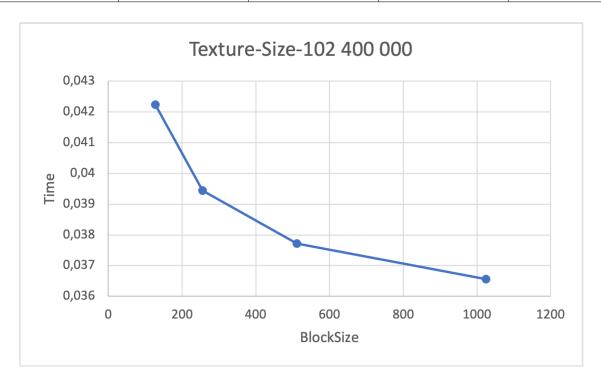


2.3 Case study n°3 - Texture Memory

In this case study, we have analyzed the counting sort algorithm using texture memory. The size of the problem is 102 400 000 and than we evaluate the performances with block size of:

- 128: with this choice we have that 2048/128 = 16 blocks. The occupancy is 100%.
- 256: with this choice we have that 2048/256 = 8 blocks. The occupancy is 100%.
- 512: with this choice we have that 2048/512 = 4 blocks. The occupancy is 100%.
- 1024: with this choice we have that 2048/1024 = 2 blocks. The occupancy is 100%.

N	BlockSize	GridSize	mips	time_sec
102 400 000	128	800 000	2 272 804 239	0,04223858
102 400 000	256	400 000	2 292 802 191	0,03944138
102 400 000	512	200 000	2 332 801 167	0,03771372
102 400 000	1024	100 000	2 412 800 655	0,03656407



Considerations

3.1 Comparison



As can be seen from the graph above, shared and textured versions are influenced by the BlockSize: as the size increases, the time decreases

Comparing the various versions, a clear improvement can be seen in the use of shared com-

pared to global as the shared has a very low latency. In the texture, there is an improvement over the global and shared versions, but only a small improvement in terms of time compared to the shared.

3.2 Premises

- Global memory is L1/L2-cache and texture memory is a separate texture cache.
- L2 is global so each access to global memory goes through L2 first, L1 is local to the multiprocessor.
- Texture memory is read-only, our program cannot dictate what is in the texture cache, so it is transparent to the program. (Behaves much like a CPU cache)
- We used atomic functions because the counting sort algorithm required that several threads access the same locations but above all it avoids data inconsistency due to the race condition.

Referring to the documentation, we realized that the atomic function was useful for our purpose because:

an atomic function performs a read-modify-write atomic operation on one 32-bit or 64-bit word residing in global or shared memory.

The operation is atomic in the sense that it is guaranteed to be performed without interference from other threads. In other words, no other thread can access this address until the operation is complete.

Atomic functions do not act as memory fences and do not imply synchronization or ordering constraints for memory operations.

In particular, we used atomicAdd and atomicSub function, the following URLs are referred to the specified functions documentation:

https://docs.nvidia.com/cuda/cuda-c-programming-guide/index.html#atomicadd https://docs.nvidia.com/cuda/cuda-c-programming-guide/index.html#atomicsub

How to run

Use the colab notebook in the folder or go to this URL:

https://drive.google.com/drive/u/1/folders/10ZMNoKnH-foAqqGiVjpBPBVHRx-Z5TAw

You can find the complete project on GitHub: https://github.com/scov8/CommonAssignment3

The starting point for the global and shared counting sort was taken here:

https://github.com/avtrunov/CUDA/blob/500bf0f1ca9af066f8dba6572d7b11121046d3a6/CountingSort.cu

https://github.com/jtruong27/ECEC413/blob/9713c61a9c5dc10c77b5d79def7ea83abc21fcb0/Homework_5/CountingSort_CUDA/counting_sort_kernel.cu

The previous year's group 02 files proposed by the professor during the course were used for file generation and extraction.

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