AUTOTUNING GPU COMPILER PARAMETERS WITH OPENTUNER

Pedro Bruel phrb@ime.usp.br

Marcos Amarís amaris@ime.usp.br

Alfredo Goldman gold@ime.usp.br 29 de Setembro de 2015



Instituto de Matemática e Estatística Universidade de São Paulo

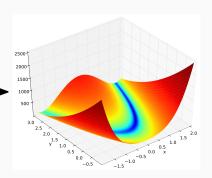
CONTRIBUTION

It is possible to optimize GPU applications for different devices by automatically tuning compilation parameters.

Configurations and Optimizations

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Search Space



AUTOTUNING: OPENTUNER



- · Autotuning framework
- Implements ensembles of search techniques
- Shares optimization results between techniques

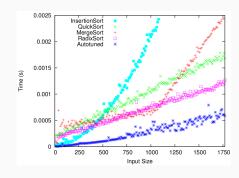


Figure 1: Autotuning recursive sorting algorithms for an 8-core machine.

Figure 1: Ansel, Jason, et al. "Opentuner: An extensible framework for program autotuning." Proceedings of the 23rd ICPAC. ACM, 2014.

HP THE MACHINE: POSSIBLE CONTRIBUTION

Optimizing programs for The Machine will cost a lot of time. Autotuning can help the programmer by:

- Adapting existing algorithms
- Pointing the way to the best optimizations

COMPILER FLAGS

Step	Options					
NVCC	<pre>prec-sqrt, relocatable-device-code, no-align-double, use-fast-math, gpu-architecture,</pre>					
PTX	<pre>def-load-cache, opt-level, fmad, allow-expensive-optimizations, maxrregcount</pre>					
NVLINK	preserve-relocs					

Options	gpu-architecture	opt-level	def-load-cache	maxrregcount	
Values	sm_20, sm_21, sm_30, sm_32, sm_35	0 - 1	ca, cg, cv, cs	16 - 64	

GPU TESTBED

Model	c.c.	Global Memory	Bus	Bandwidth	L2	SM/Cores	Clock
GTX-680	3.0	2 GB	256-bit	192.2 GB/s	512 KB	8/1536	1006 Mhz
Tesla-K20	3.5	4 GB	320-bit	208 GB/s	1280 KB	13/2496	706 Mhz
Tesla-K40	3.5	12 GB	384-bit	276.5 GB/s	1536 KB	15/2880	745 Mhz

EXPERIMENTS

All the results' data and the code for the experiments, the autotuner and the figures is hosted at github.com/phrb/gpu-autotuning, under the GNU GPLv3 license.

EXPERIMENTS: MATRIX MULTIPLICATION

Four optimizations of square matrix multiplication (N = 1024):

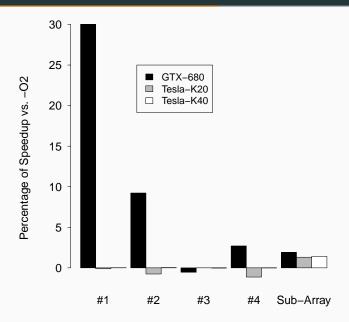
- #1: Non-Coalesced accesses to Global Memory
- #2: Coalesced accesses to Global Memory
- #3: #1, plus Shared Memory
- #4: #2, plus Shared Memory

EXPERIMENTS: MAXIMUM SUBARRAY

Find the maximum subsequence sum of an array (N = 134217728):

- · 4096 threads
- · 32 blocks of 128 threads

RESULTS: OPTIMIZATION



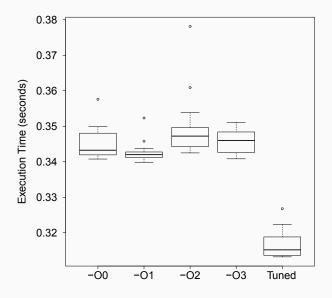
RESULTS: OPTIMIZATION

Why did the GTX-680 have the best results?

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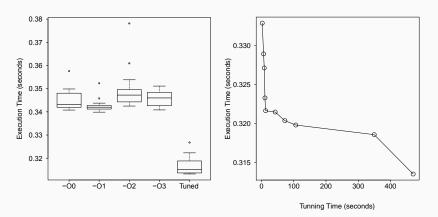
RESULTS: AUTOTUNER

Results for optimization #2 in the GTX-680:

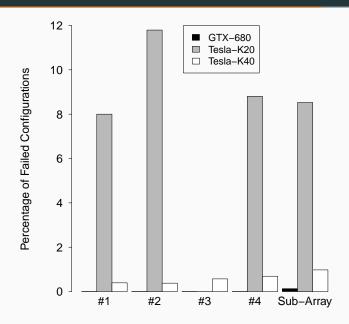


RESULTS: AUTOTUNER

Results for optimization #2 in the GTX-680:



RESULTS: FAILED CONFIGURATIONS



RESULTS: FAILED CONFIGURATIONS

Who's guilty?

• sm_32, for #1 in the K20 and K40

CONCLUSION

 \cdot 30% speedup for #1 in the GTX-680

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- 30% speedup for #1 in the GTX-680
- · Different parameters for each GPU
- · Always assert the results



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