Parallel Algorithms for Kernel Density Estimation

Namo Wichitrnithed^a, Rylan Spence^a

^aOden Institute for Computational Engineering and Sciences, University of Texas at Austin, Austin, 78712, TX, USA

Abstract

Keywords: Kernel Density Estimation, Parallel computing, Cuda

1. Introduction

- Project Motivation
- Brief literary review KDE (See References Below)
- Brief literary review Parallized KDE (See References Below)

2. Multi-Core Programming Models

• MPI

Basic Review of MPI Architecture

• GPU

Basic Review of GPU Architecture

3. Serial vs. Vectorized vs. Multi-Core Kernel Density Estimation

- Serial Algorithm Pseudo Code
- SIMD Algorithm Pseudo Code
- MPI Algorithm Pseudo Code

Email addresses: namo@utexas.edu (Namo Wichitrnithed), rylan.spence@utexas.edu (Rylan Spence)

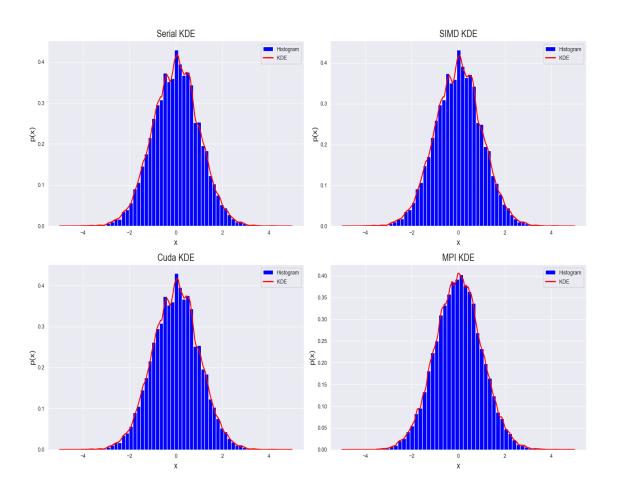


Figure 1: Kernel Density Estimates for all four proposed algorithms

• Cuda Algorithm Pseudo Code

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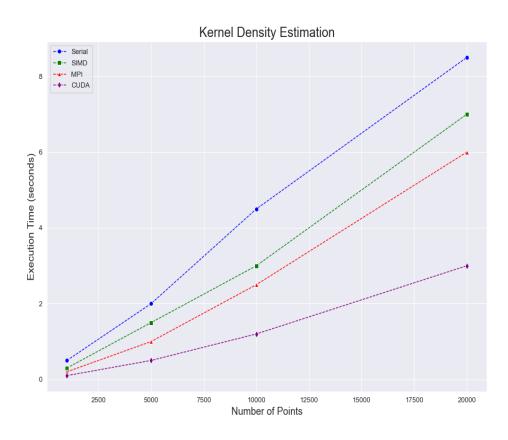


Figure 2: Execution times (in secs) of a kernel estimation as a function of the grid points.

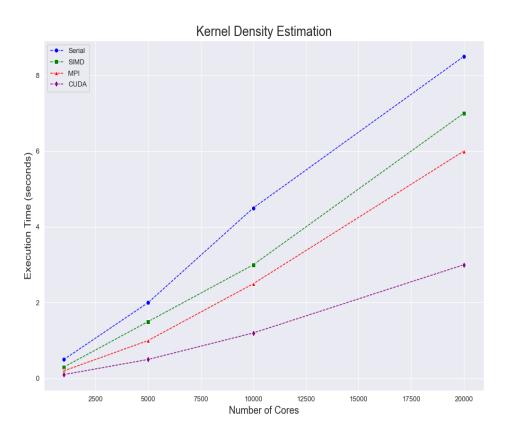


Figure 3: Execution times (in secs) of a kernel estimation as a function of the number of cores.

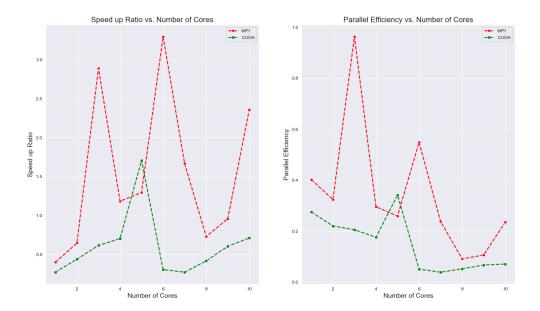


Figure 4: Performance of the MPI-parallel algorithm for KDE on the test data set (n=?, m=?): (a) speed up ratio and (b) parallel efficiency - defined as speedup ratio divided by number of CPU cores.

Kernel Density Algorithm	Serial	SIMD	MPI	Cuda
Lines of code	12	17	10	12

Table 1: Lines of code in Programming Models

4. Results

- 4.1. Quantititative Comparison
- 4.2. Qualitative Comparison

5. Conclusions

References

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- [2] P. D. Michailidis, K. G. Margaritis, Parallel Computing of Kernel Density Estimation with Different Multi-core Programming Models, in: 2013 21st Euromicro International Conference on Parallel, Distributed, and Network-Based Processing, 2013, pp. 77–85. doi:10.1109/PDP.2013.20.
- [3] G. Zhang, A.-X. Zhu, Q. Huang, A GPU-accelerated adaptive kernel density estimation approach for efficient point pattern analysis on spatial big data, International Journal of Geographical Information Science 31 (10) (2017) 2068–2097. doi:10.1080/13658816.2017.1324975.