# ACCELERATED COMPLEX EVENT PROCESSING WITH GRAPHICS PROCESSING UNITS

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degree Master of Science in Computer Science

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#### **Declaration**

I declare that this is my own work and this dissertation does not incorporate without acknowledgement any material previously submitted for a Degree or Diploma in any other University or institute of higher learning and to the best of my knowledge and belief it does not contain any material previously published or written by another person except where the acknowledgement is made in the text.

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The above candidate has carried out research for the Masters Dissertation under my supervision.

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Dr. H.M.N. Dilum Bandara	Dr. Srinath Perera
Date	Date

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#### Abstract

As Big Data scenarios increasingly become common, a large number of distributed data processing systems require timely processing of high volumes of real-time data streams. Detecting complex correlations between incoming data streams in near real-time is at the heart of these data processing systems. Complex Event Processing (CEP) have been dominating in this domain since inception a decade back. But, growth of Big Data volumes demands for more performance and faster processing. CEP operators like stream join and event patterns require considerable processing power and have huge impact on the overall query processing performance. In some use cases these operators have to operate on lots of events simultaneously. Making parallel algorithms for these operators is a common approach for improving the individual operator performance.

A Graphics Processing Unit (CPU) provides a vast number of parallel computing cores and leverage new parallel agorithms which enables have problems vibration approaches for existing problems vibration the challenge is combining complex event processing and GPUs in the right way to get the maximum performance out of the this parallel hardware. There had been attempts to use parallel hardware in improving CEP performance in both commercial and academic implementations, and most of them uses multi-core approach. Only a very few researches had used GPUs for CEP. We believe the lack of GPU related CEP researches is that they are not designed to benefit from parallel processing in GPUs.

In this research we investigate how and when GPUs can be used to improve the query processing performance of a popular open source CEP implementation, Siddhi CEP. Siddhi, by design, supports for parallel query processing in multi-core CPUs. This work propose a novel approach for parallel event processing in GPUs with several GPU event processing algorithms. Performance evaluation on our implemented algorithms shows, for a mix of complex queries, parallel event processing on GPUs achieve more than ten times event processing throughput than the sequential processing in CPUs. Moreover, our approach helped to reduce event queuing at the incoming event queue when there are high frequent input event stream and several complex queries.

Keywords. Complex Event Processing, Parallel Hardware, GPGPU, Siddhi CEP.

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#### List of Abbreviations

**CEP** Complex Event Processing

**CMP** Chip Multi-processors

**CUDA** Compute Unified Device Architecture

**DBMS** Database Management Systems

**DSMS** Data Stream Management System

**EDA** Event Driven Architecture

FPGA Reld-Plogramonably Gate Monaytuwa, Sri Lanka.

GPGPU General Purpose Computing on the Graphics Processing Unit

**GPU** Graphic Processing Units

**HPC** High Performance Computing

JSON Javascript Object Notation

MIMD Multiple Instruction, Multiple Data streams

MISD Multiple Instruction, Single Data stream

POJO Plain Old Java Object

SIMD Single Instruction, Multiple Data stream

SISD Single Instruction, Single Data stream

**SMP** Symmetric Multiprocessing