Assessing the Impact of Network Bandwidth and Operator Placement on Data Stream Processing for Edge Computing Environments

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## Overview



#### Stream Processing Context

- Introduction
- Data Stream Processing Engines
- Application Graph
- Cloud Computing
- Fog Computing
- Contributions
- Goal

#### 2 Experiments Setup

- Assumptions
- Infrastructure Configuration
- Analysed Application
- Evaluated Scenarios



#### Results



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## Data are generated all the time!



- Today's instruments and services are producing ever-increasing amounts of data
- Data requires processing to provide insights or assist in decision making
- Big Data poses challenges regarding data transfer, storage, and processing
- MapReduce has been very popular for batch processing (High latencies)
- Data generated is received in (near) real-time and requires quick analysis(Low latencies)

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## Data Stream Processing Engines

The interest in **processing data events as they arrive** has led to the emergence of several Data Stream Processing Engine (DSPE)



DSPEs are part of larger architectures that comprise multiple tiers



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# Stream Processing Application Characteristics

Vertices are operators that execute a function over the incoming data and edges define how data flows between them Data Source generates the data events and Sink consumes them Data Events arrive continuously (It is not possible to predict when they ceased)



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# **Cloud Computing**



Clouds are often the target infrastructure for deploying such engines

- Pay-as-you-go business model
- Scalability
- Elasticity

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# Fog Computing



More modern solutions intend to exploit the edges of the Internet

- Reduce the application end-to-end latency and communication costs
- Enable services to react locally
- Offload processing from the cloud

[Chan, 2016]

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## Research Goals!



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Algorithms for Elastic Big Data Stream Analytics

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What do we really want to solve?

# **APPLICATION AUTO-PARALISATION**

- Profitability mechanism
  - Throughput
  - 2 Transferred Data





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# Are there benefits in deploying operators in Fog Computing?



We aim to evaluate multiple combinations of operators deployment, spreading them across the cloud and edge network

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Assumptions Infrastructure Configuration Analysed Application Evaluated Scenarios

## **Environment** restrictions

#### Heterogeneity

The heterogeneity taken into account for this first effort is the **network bandwidth between the edge-to-cloud** 

#### Application's operators

We consider stateless operators

#### Application's structure

We considered **pipelined applications** 

Assumptions Infrastructure Configuration Analysed Application Evaluated Scenarios

## Roles, Frameworks and Tools

#### 4 R410 Dell servers - Intel® Xeon® Processor E5506 4M Cache, 2.13 GHz, 4.80 GT/s Intel® QPI



- Python psutil 5.2.0 used for measuring resource consumption, CPU utilisation, memory usage and network I/O
- Linux Network Time Protocol (NTP) for syncronizing the time

• Off-the-shelf frameworks for DSPE and Queues ( ) ( )

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## Sentiment Analysis

- Compute the **polarity** of tweets
- Dataset with 1 GB of tweets
- A simple Natural Language Processing (NLP) technique
- Tweets are **JSON** dictionaries



Assumptions Infrastructure Configuration Analysed Application Evaluated Scenarios

### How the evaluations were carried out?



- Setup Edge-to-Cloud network bandwidth (10, 100, 1000, 10000 Kbps)
- 2 7 minutes considering the testbed specification
- 8 Eliminate warm up and cool down effects

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## What really happened?



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## What really happened?



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## Conclusion

In this work we evaluated the **impact in terms of number of tweets** handled by a stream processing application when **varying the network bandwidth** and the **deployment** of **operators across cloud and edge Computing** resources.

- graph behaviour has a high impact on the placement, mainly points with data reduction
- placing operators **both in cloud and edge** network brings some **important benefits**
- depending on the network bandwidth, the edge deployment improves the number of processed events

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## Future Work – Thank you for your attention – Questions!

- Consider resources heterogeneity
- Isolate operators processes to achieve a certain level of determinism
- Look for infrastructures that handle FOG Computing
- Improve our model to reduce its complexity

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