

Using Trusted Execution Environments for Secure Stream Processing of Medical Data

DAIS'19 - 18-20 July 2019, Copenhagen (DK) - Use-Case Paper

C. Segarra¹ R. Delgado¹ M. Lemay¹ P. Aublin³ P. Pietzuch³ V. Schiavoni²

CSEM, Neuchâtel, Switzerland, {carlos.segarra,ricard.delgado,mathieu.lemay}@csem.ch

²University of Neuchâtel, Switzerland, valerio.schiavoni@unine.ch

³Imperial College London, United Kingdom, {p.aublin,prp}@imperial.ac.uk

Thursday, June 13, 2019

Imperial College London



Outline

1 Introduction

Motivation

2 Background

Technical Background
Medical Use-Case

3 Secure Stream Processing of Medical Data

Project Goals

Architecture

Evaluation & Results

Conclusions and Further Work



Why do we need secure big data processing engines?

We want to:

Process large amounts of **sensitive** information



Why do we need secure big data processing engines?

We want to:

Process large amounts of **sensitive** information



↓ Generally requires...

Outsourcing of data storage and processing



Why do we need secure big data processing engines?

We want to:

Process large amounts of **sensitive** information



Outsourcing of data storage and processing



Cloud tenant can access protected data



Why do we need secure big data processing engines?

We want to:

Process large amounts of **sensitive** information



Outsourcing of data storage and processing

↓ Generally means...

Cloud tenant can access protected data



UNACCEPTABLE FOR INDUSTRIES IN THE MEDICAL DOMAIN (and many others)



Trusted Execution Environments (TEE)

Formal definition, examples and availability

Trusted Execution Environment

A **TEE** is a secure area of a processor. It guarantees code and data to be protected with respect to **confidentiality** and **integrity**.



Trusted Execution Environments (TEE)

Formal definition, examples and availability

Trusted Execution Environment

A **TEE** is a secure area of a processor. It guarantees code and data to be protected with respect to **confidentiality** and **integrity**.

Available on a variety of commodity CPUs. For instance:



Trusted Execution Environments (TEE)

Formal definition, examples and availability

Trusted Execution Environment

A **TEE** is a secure area of a processor. It guarantees code and data to be protected with respect to **confidentiality** and **integrity**.



Intel SGX Available on consumer-grade CPUs starting from architecture codename *Skylake*.



Arm Trustzone Available on Cortex-A processors and v8 Cortex-M23 and Cortex-M33 (e.g. Raspberry Pi).



Intel Software Guard eXtensions Definition, Threat Model and Known Vulnerabilities

- Intel Software Guard eXtensions (SGX) are a set of instructions and memory access extensions that enable applications to create hardware-protected areas in their address space called enclaves.
- Security perimiter includes only the internals of the CPU package.
- An attestation protocol verifies that code is running in a genuine enclave and that it has not been tampered.

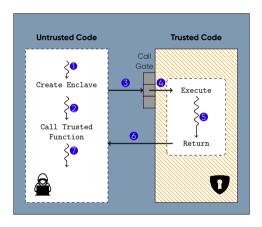


Figure 1: Intel SGX Operating Principles.



Intel Software Guard eXtensions Definition, Threat Model and Known Vulnerabilities

- Intel Software Guard eXtensions (SGX) are a set of instructions and memory access extensions that enable applications to create hardware-protected areas in their address space called enclaves.
- Security perimiter includes only the internals of the CPU package.
- An attestation protocol verifies that code is running in a genuine enclave and that it has not been tampered.

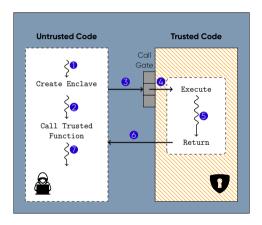


Figure 1: Intel SGX Operating Principles.



Intel Software Guard eXtensions Definition, Threat Model and Known Vulnerabilities

- Intel Software Guard eXtensions (SGX) are a set of instructions and memory access extensions that enable applications to create hardware-protected areas in their address space called enclaves.
- Security perimiter includes only the internals of the CPU package.
- An attestation protocol verifies that code is running in a genuine enclave and that it has not been tampered.

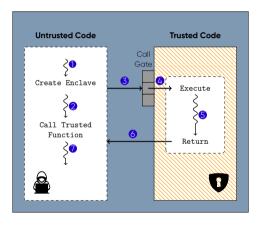


Figure 1: Intel SGX Operating Principles.



SGX-Spark Running Spark Jobs inside Enclaves

SGX-Spark

- SGX-Spark is a framework under-development at the Imperial College London to enable seamless deployment of Spark jobs inside enclaves.
- Protect confidentiality and integrity of existing Spark jobs without modifications to the applicatio code.
- Execute only sensitive parts of the application inside the enclave. Leave information outside the enclave encrypted.

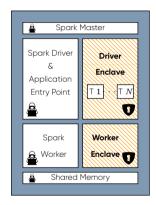


Figure 2: SGX-Spark Architecture.



SGX-Spark Running Spark Jobs inside Enclaves

SGX-Spark

- SGX-Spark is a framework under-development at the Imperial College London to enable seamless deployment of Spark jobs inside enclaves.
- Protect confidentiality and integrity of existing Spark jobs without modifications to the application code.
- Execute only sensitive parts of the application inside the enclave. Leave information outside the enclave encrypted.

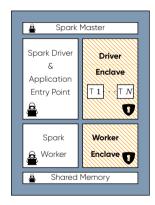


Figure 2: SGX-Spark Architecture.



SGX-Spark Running Spark Jobs inside Enclaves

SGX-Spark

- SGX-Spark is a framework under-development at the Imperial College London to enable seamless deployment of Spark jobs inside enclaves.
- Protect confidentiality and integrity of existing Spark jobs without modifications to the application code.
- Execute only sensitive parts of the application inside the enclave. Leave information outside the enclave encrypted.

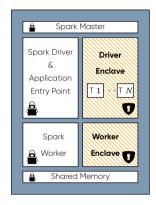


Figure 2: SGX-Spark Architecture.



- Our solution is independent of the chosen data streams, hence it could be leveraged in a variety of domains where privacy might be a concern. However,
- The data streams used for this project belong to the medical domain, in particular, they are obtained from human cardiac activity monitoring. The two most standard procedures are:
 - ECG: measure heart's electrical activity over time.
 E.g: chest straps.
 - PPG: measure blood's volume variation over time.
 E.g: smartwatches and pulse oximeters.





- Our solution is independent of the chosen data streams, hence it could be leveraged in a variety of domains where privacy might be a concern. However,
- The data streams used for this project belong to the medical domain, in particular, they are obtained from human cardiac activity monitoring. The two most standard procedures are:
 - **ECG:** measure heart's electrical activity over time. *E.g.* chest straps.
 - PPG: measure blood's volume variation over time.
 E.g: smartwatches and pulse oximeters.





- Our solution is independent of the chosen data streams, hence it could be leveraged in a variety of domains where privacy might be a concern. However,
- The data streams used for this project belong to the medical domain, in particular, they are obtained from human cardiac activity monitoring. The two most standard procedures are:
 - **ECG:** measure heart's electrical activity over time. *E.g.* chest straps.
 - **PPG:** measure blood's volume variation over time.
 E.g: smartwatches and pulse oximeters.





- We are interested in the **inter-beat intervals** from the generated diagram (23-69)B per second per user).
- From an ECG we can obtain obtain these intervals from the time between R peaks.
- With their timestamps we compute a live analysis of the Heart Rate Variability

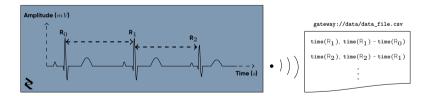


Figure 3: Schematic representation of an ECG signal and the data we gather.

- We are interested in the **inter-beat intervals** from the generated diagram (23 69) B per second per user).
- From an **ECG** we can obtain obtain these intervals from the time between **R peaks**, abbreviated as RR intervals (Figure 3).
- With their timestamps we compute a live analysis of the Heart Rate Variability (HRV).

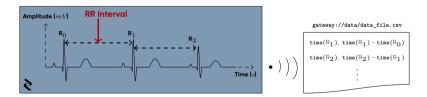




Figure 3: Schematic representation of an ECG signal and the data we gather.

- We are interested in the **inter-beat intervals** from the generated diagram (23-69)B per second per user).
- From an ECG we can obtain obtain these intervals from the time between R peaks, abbreviated as RR intervals (Figure 3).
- With their timestamps we compute a live analysis of the Heart Rate Variability (HRV).

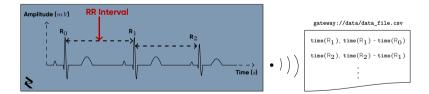


Figure 3: Schematic representation of an ECG signal and the data we gather.

Secure Stream Processing of Medical Data Project Goals

Project Goals

- Use SGX-Spark to implement a streaming platform that gathers data from sensors and securely outsources computation.
 - Deploy in an existing medical environment.
 - Perform stress tests to assess the system's robustness and reliability.
 - Evaluate and quantify the overhead of providing strong security guarantees.



Secure Stream Processing of Medical Data Envisioned Scenario

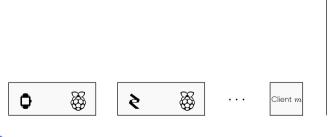
The **chosen scenario** involves:

- 1 ECG data streamed from a sensor to a gateway
- 2 Real-time processing with HRV algorithms: SDNN and HRV Bands analysis.
 - SDNN: rolling standard deviation of NN (RR) intervals.
 - HRV Bands: rolling Discrete Fourier Transform and low/high frequency component.
- 3 Support for **storage** and result **post-processing**

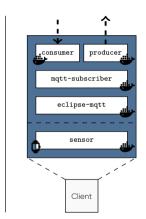




Figure 4: Wearable-enabled ecosystem of our platform.



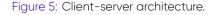












. . .

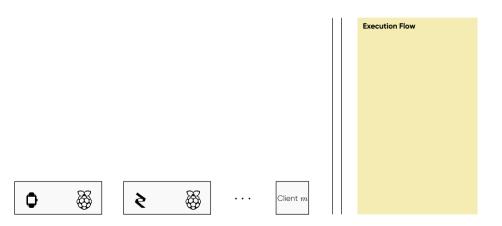




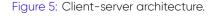
Figure 5: Client-server architecture.

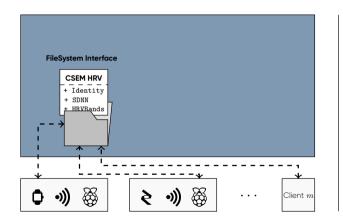










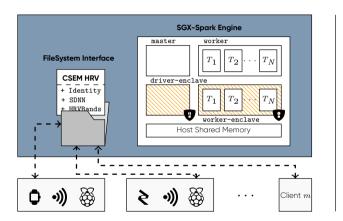


Execution Flow

- Sensors generate samples and streams them to the gateway over mqtt
- 2 Gateways aggregate data, encrypts it and sends it to the server's filesystem interface over SETP



Figure 5: Client-server architecture.

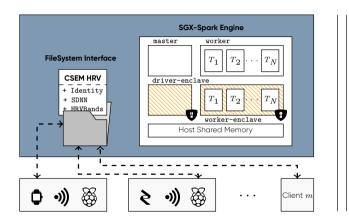


Execution Flow

- Sensors generate samples and streams them to the gateway over mqtt
- 2 Gateways aggregate data, encrypts it and sends it to the server's filesystem interface over SFTP
- 3 SGX-Spark streaming job monitors the FS, batch-processes data, and saves encrypted results back in the FS



Figure 5: Client-server architecture.



Execution Flow

- Sensors generate samples and streams them to the gateway over mqtt
- 2 Gateways aggregate data, encrypts it and sends it to the server's filesystem interface over SFTP
- 3 SGX-Spark streaming job monitors the FS, batch-processes data, and saves encrypted results back in the FS
- 4 Gateways fetch result files



Figure 5: Client-server architecture.

Evaluation & Results Experimental Setup & Metrics

Three Execution Modes:

- Vanilla Spark
- SGX-Spark w/o Enclaves
- **3** SGX-Spark w/ Enclaves

Two Algorithms:

- 1 Identity (Batch & Stream)
- SDNN (Batch & Stream)

Two Metrics:

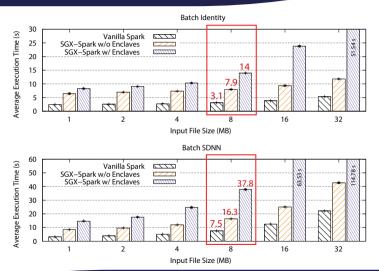
- Elapsed Time
- 2 Avg. Batch Processing Time

Workload	s_rate (samples / sec)	Input Load
Batch - Small	$\{44, 89, 178, 356, 712, 1424\}$	$\{1,2,4,8,16,32\}\ \mathrm{kB}$
Stream - Small	$\{44, 89, 178, 356, 712, 1424\}$	$\{1, 2, 4, 8, 16, 32\} \ kB \ / \ sec$
Batch - Big	$\{44, 89, 178, 356, 712, 1424\} * 1024$	$\{1,2,4,8,16,32\}\ MB$
Stream - Big	$\{44, 89, 178, 356, 712, 1424\} * 1024$	$\{1,2,4,8,16,32\}\ { m MB}\ /\ { m sec}$



Evaluation & Results

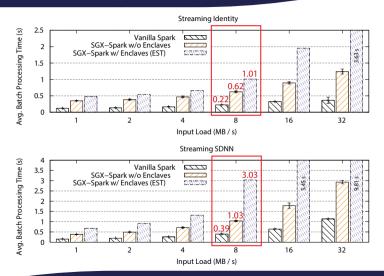
Results: Batch Execution - Big Load





Evaluation & Results

Results: Stream Execution - Big Load





Conclusion

- Introduced a PoC of a privacy-preserving streaming platform.
 - Introduces $4 \times -5 \times$ slowdown vs. **vanilla Spark Streaming** (load < 4 MB per second).
 - Requires **no changes** to the application source code.
- Further lines of research:
 - Perform an economical evaluation of the cost of deploying our system to the cloud: how expensive is privacy?
 - Reduce the TCB on the client side leveraging TEEs for low-power devices (e.g. ARM TrustZone).



Conclusion

- Introduced a PoC of a privacy-preserving streaming platform.
 - Introduces $4 \times -5 \times$ slowdown vs. **vanilla Spark Streaming** (load < 4 MB per second).
 - Requires **no changes** to the application source code.
- Further lines of research:
 - Perform an economical evaluation of the cost of deploying our system to the cloud: how expensive is privacy?
 - Reduce the TCB on the client side leveraging TEEs for low-power devices (e.g. ARM TrustZone).

Thank you very much for your attention! Questions?

