1 Minute Madness



Order of presentations:

- 1. Preventing EFail Attacks with Client-Side WebAssembly
- 2. Differential Privacy for Private Pattern Protection
- 3. SLASH: Serverless Apache Spark Hub (demo)
- 4. Pattern-Level Privacy in DCEP
- 5. Goodbye Engineered ANNs, Hello Evolutionary Neural Networks
- 6. ComDeX Unveiled Demonstrating the Future of IoT-Enhanced ...
- 7. Practical Forecasting of Cryptocoins Timeseries using...
- 8. Thetacrypt: a distributed service for threshold cryptography...
- 9. No One Size (PPM) Fits All: Towards Privacy in Stream...
- 10. Cognitive Cyber Defense Dynamic, Adaptive Cyber...
- 11. StreamToxWatch Detector Architecture for Data Poisoning...
- 12. Handling Inconsistent Data in Industry 4.0
- 13. Privacy-preserving Transaction DAG
- 14. Decentralized Stream Reasoning Agents
- 15. Agent-based Orchestration on a Swarm of Edge Devices



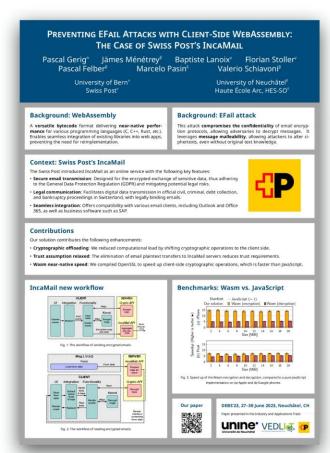




- Ever wished to send **digitally secure**, **legally-binding** letters in Switzerland?
- Worked with **Swiss Post** to design a **better** secure email transmission platform

Join us to: discover how + see our performance results!

#WebAssembly #Cryptography #OpenSSL



The Research Problem

trade-off

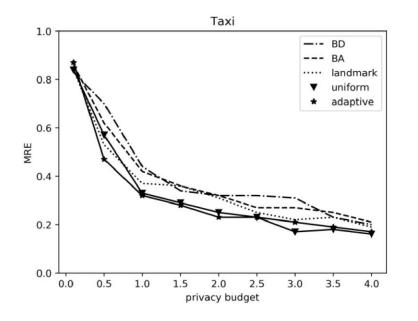
protect privacy protection under a required data utility in data streams

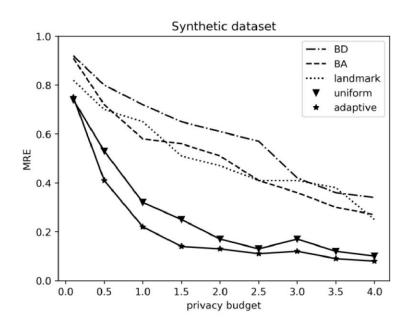
A few people have done something already, but not enough...

Contributions

• **break** the current performance **limitations** and **open up** a novel "level" of DP and PPMs

Early-stage verification results





Demo: SLASH: Serverless Apache Spark Hub



How do you slash operating expenses (and footprint) of big data infrastructure?

With SLASH!

As easy as 'import slash'.

SLASH is

- Distributed across Spark topology
- Event-based from multiple sources
 - scheduled usage
 - application requests
 - forecasting
- Adaptive to application needs: local mode, immediate scaling, lazy scaling

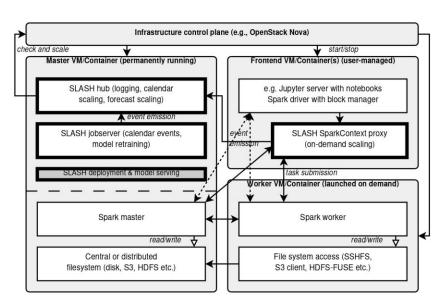
```
import sparky
import pyspark
import pyspark.sql

import slash

sc = sparky.connect("yourbigdatajob", 2)
spark = pyspark.sql.SparkSession.builder.getOrCreate()

# your custom code here...
#df = spark.createDataFrame([("x",), ("y",), ("z",)], ["reallybigdata"])
#df.show()

sc.stop()
```





Pattern-Level Privacy in DCEP

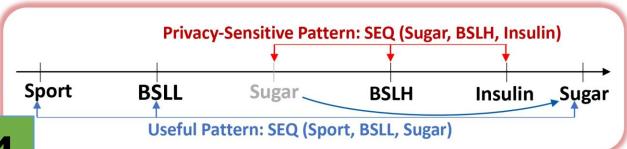
Majid Lotfian Delouee

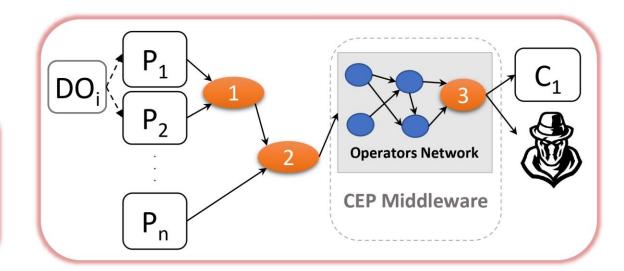
2 Research Questions

- Which obfuscation technique?
- Where to deploy the obfuscation operator?
- How to model adversary's background knowledge?

Event Obfuscation Techniques.

From Attributes to Patterns.





Goodbye Engineered ANNs, Hello Evolutionary Neural Networks

1. Nature-inspired Machine Learning:

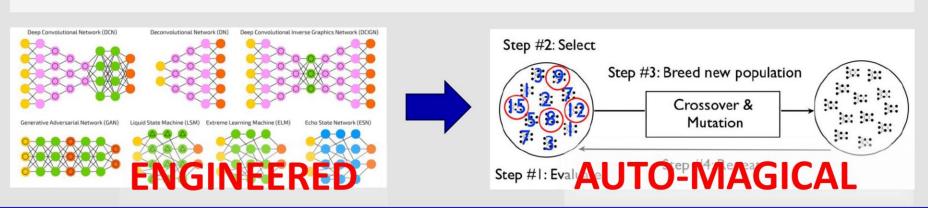
Artificial Intelligence + Evolution + Neuroscience

2. Scalable Neuroevolution:

Collaborating models for emerging new tasks

3. Universal Representation:

Data format for different input/output modalities











DEBS 2023

ComDeX: A Context-aware Federated Platform for **IoT-enhanced Communities**









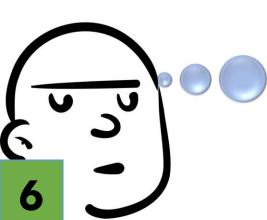
"Breaking the Silos"

How?



"Outperforming Alternatives"





How can stakeholders maintain their data sovereignty while participating in a data-sharing network?

Selective Sharing?

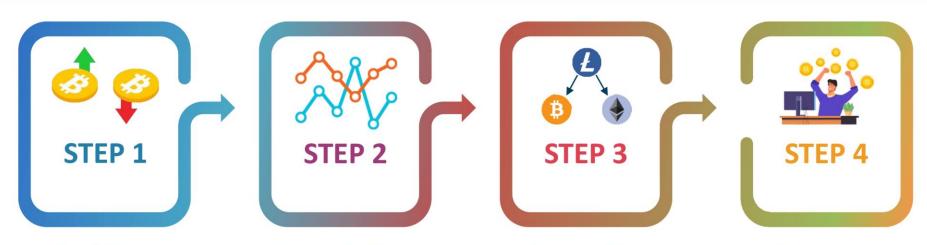
How to seamlessly connect diverse stakeholders and enable efficient data exchange?





Pasquale De Rosa, Pascal Felber and Valerio Schiavoni





Problem

Cryptocoins are extremely volatile assets. A real nightmare for potential investors! We propose a solution to predict... the unpredictable.

Correlation

We study correlations between altcoin and main coin (BTC/ETH) prices. Highly correlated altcoins are good predictors for BTC/ETH prices.

Causality

Correlation does not imply causation! We analyze causality between correlated altcoins and main coins to identify the best predictors.

Forecast

We use ML models to forecast BTC/ETH prices basing on these altcoins. Our predictions are highly accurate. Good news for the investors!

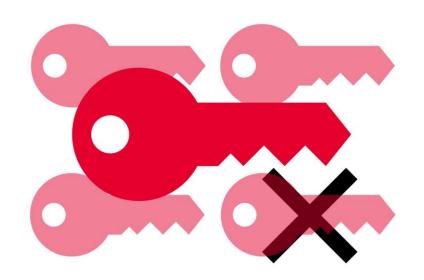


UNIVERSITÄT BERN

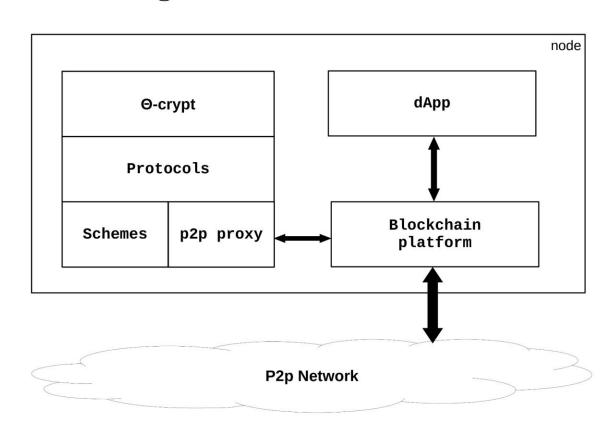
THETACRYPT: A DISTRIBUTED SERVICE FOR THRESHOLD CRYPTOGRAPHY ON CHAIN

Orestis Alpos, **Mariarosaria Barbaraci**, Christian Cachin, Noah Schmid, Micheal Senn, Nathalie Steinhauer University of Bern

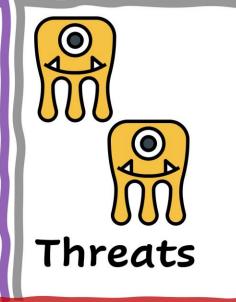
Threshold cryptosystems



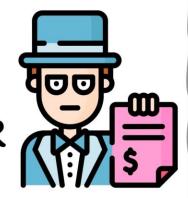
Integration on blockchains











Ooh

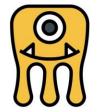
SPS

A tale of Privacy in SPSs

















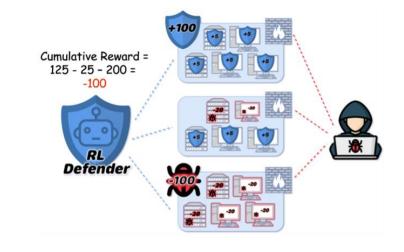
Cognitive Cyber Defense - Dynamic, Adaptive Cyber Defense Systems for Massively Distributed, Autonomous, and Ad-hoc Computing Environments - A New Brand of Cyber.

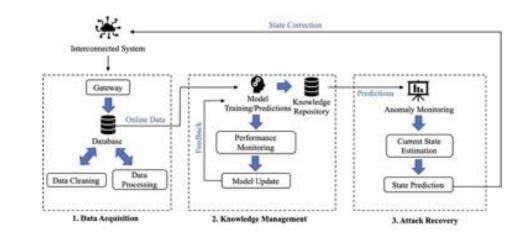
The unprecedented scale, speed, and scope of interconnectivity, ranging from the microsensors on the *edge* to the global networks, will be the prominent characteristics of the emerging computing environments.

Our current cyber defense methods, designed for the computing environments we know, will be mainly rendered inadequate for these future environments. To meet the emerging cyber defense demands, we need a new approach.

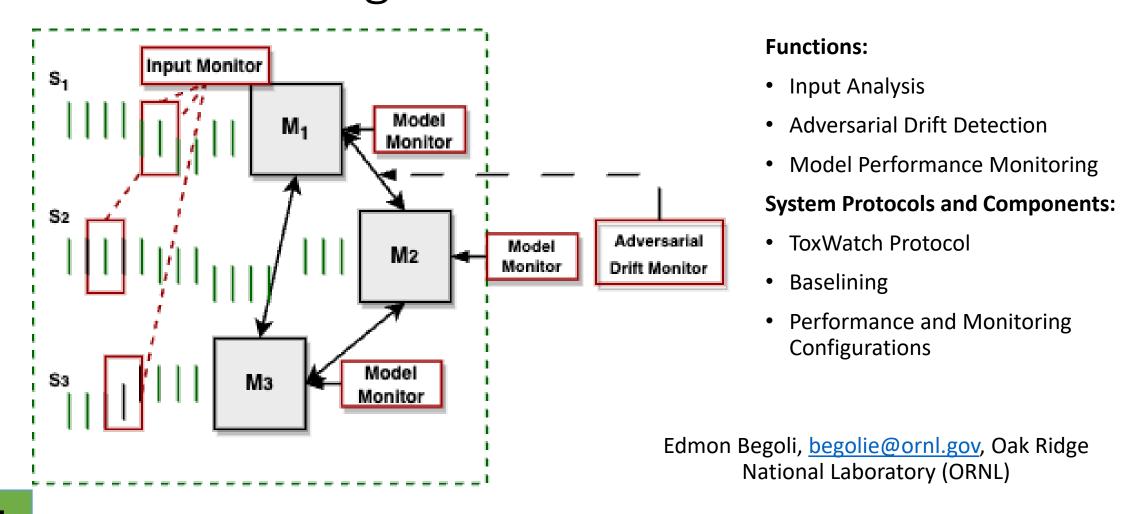
Cognitive Cyber Principles:

- (a) Multi-modal and multi-scale cyber defense
- (b) Dynamic adaptation
- (c) Autonomous responses and operations

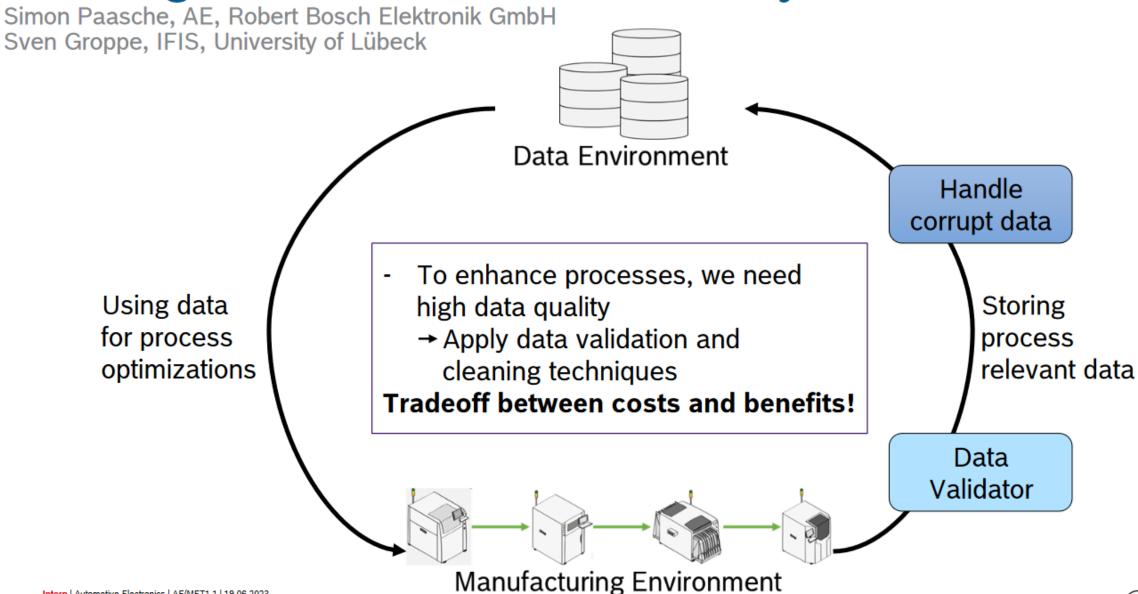




StreamToxWatch – Detector Architecture for Data Poisoning in Streams



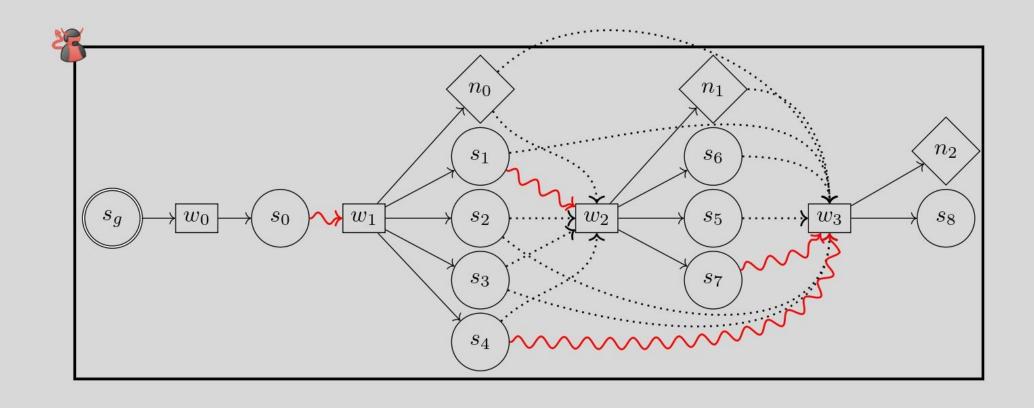
Handling Inconsistent Data in Industry 4.0





Privacy-Preserving Transaction DAG (PDAG)





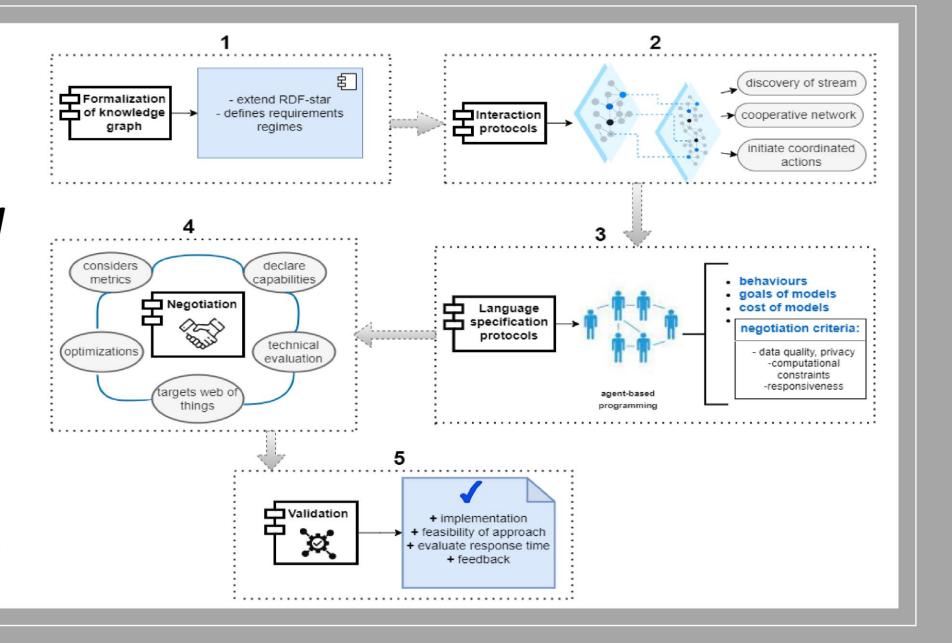


Decentralized Stream Reasoning Agents

Gozde A. Tataroglu Ozbulak

University of Applied Sciences and Arts Western Switzerland HES-SO

DEBS 2023



Hes·so WALAIS

By Banani

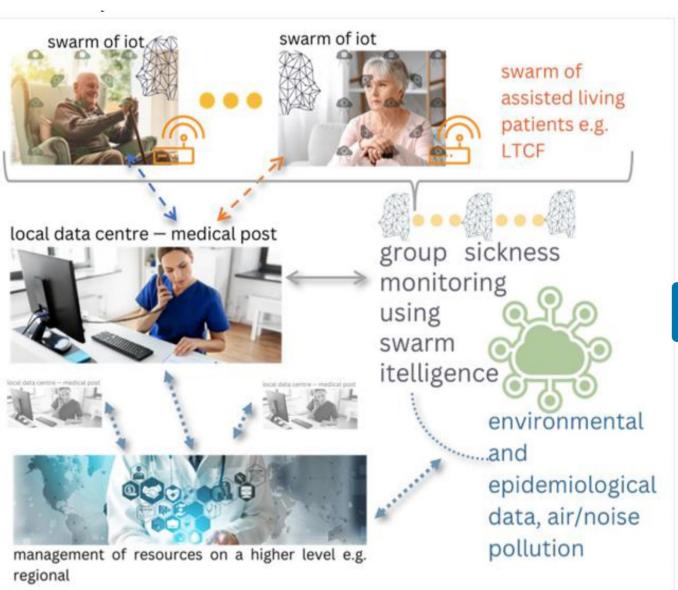
by Jean-Paul

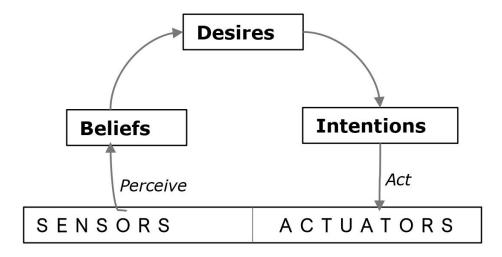
Calbimonte)

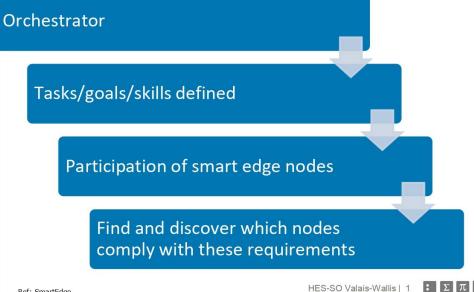
Anuraj (Supervised

Agent-Based Orchestration on a Swarm of Edge Devices









Documentation