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PRICAR: Privacy Framework for Vehicular Data Sharing with Third- Parties

IEEE SecDev 2023, Atlanta, GA
10/20/23



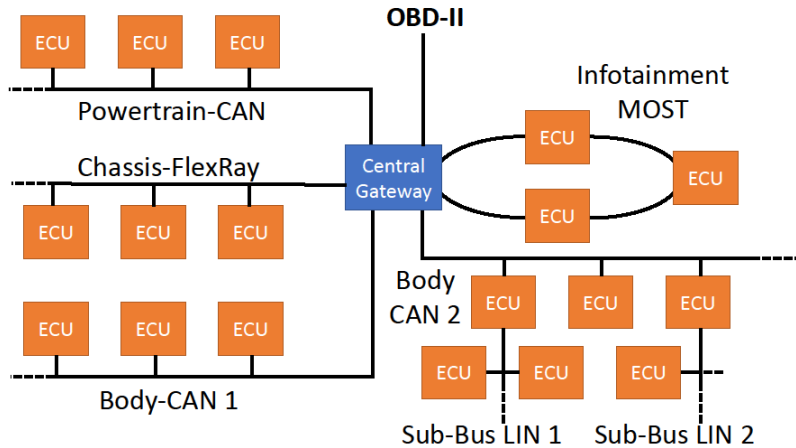
TigerSec Laboratory
@ Clemson University



REAL-TIME COMPUTING
LABORATORY
@ UNIVERSITY OF MICHIGAN, ANN ARBOR



Security and privacy landscape is changing



First-Generation Attacks (~2010-2015)

Using physical interfaces



Second-Generation Attacks (~2015-2020)

Using wireless interfaces (e.g., IVI and TCU)



Third-Generation Attacks (~2020-?)

Using app eco-system on IVIs

Scalability

Risk / Damage Potential

Increased connectivity comes at a price



Data Generation

~25 GB/h (total)
~200 MB/h (CAN)

Data Connectivity

- 2016: 20%
- 2020: 75%
- 2021: 98%



PROGRESSIVE

State Farm

verizon

otonomo



Privacy Concerns

- Rising Awareness (e.g., Facebook-Cambridge Analytica incident)
- General Data Protection Regulation (GDPR)
- Increasing Number of Privacy Attacks in Cars

ALLIANCE OF AUTOMOBILE MANUFACTURERS, INC.
ASSOCIATION OF GLOBAL AUTOMAKERS, INC.

Consumer Privacy
Protection Principles
PRIVACY PRINCIPLES FOR VEHICLE
TECHNOLOGIES AND SERVICES

November 12, 2014

Mert D. Pesé*, Xiaoying Pu, and Kang G. Shin

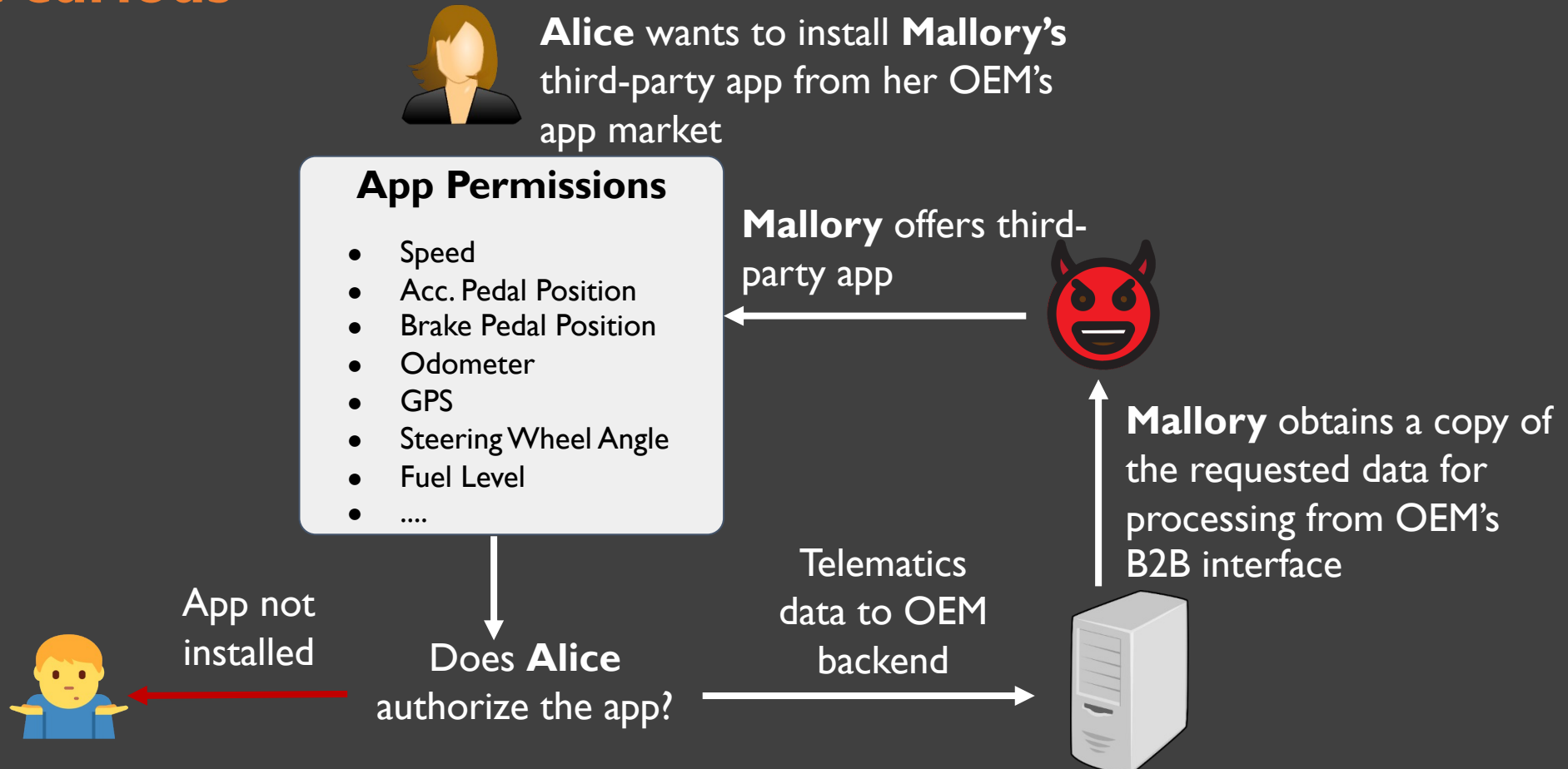
SPy: Car Steering Reveals Your Trip Route!

Miro Enev*, Alex Takakuwa, Karl Koscher, and Tadayoshi Kohno

Automobile Driver Fingerprinting

Data Collection

- Driving data is shared with third-party entities who can be **benign, but curious**



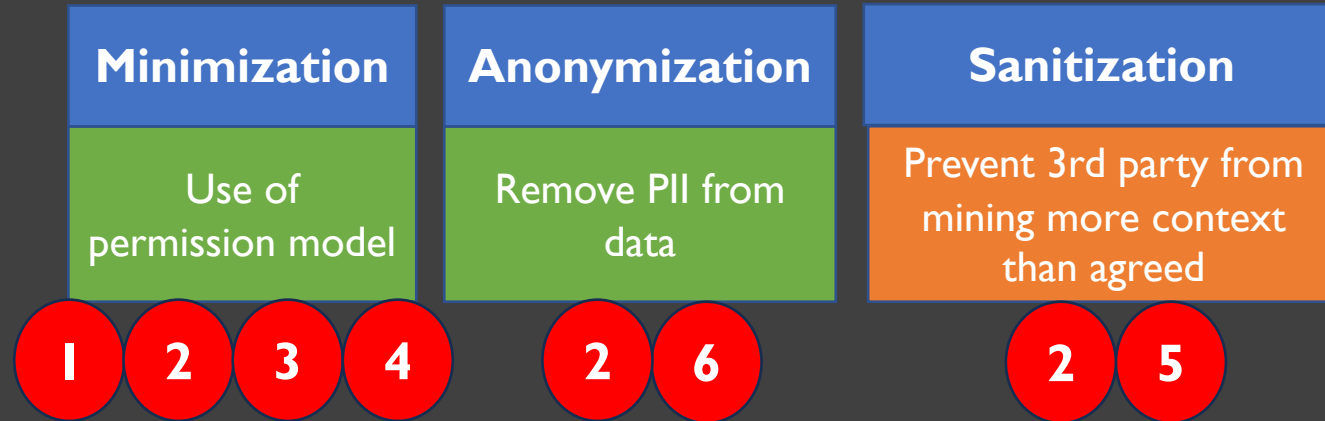
General Data Protection Regulation (GDPR)



1. Lawfulness, Fairness and Transparency
2. Purpose Limitation
3. Data Minimization
4. Accuracy
5. Storage Limitation
6. Integrity and Confidentiality
7. Accountability

PRIVACY GOALS

— Frontend
— Backend



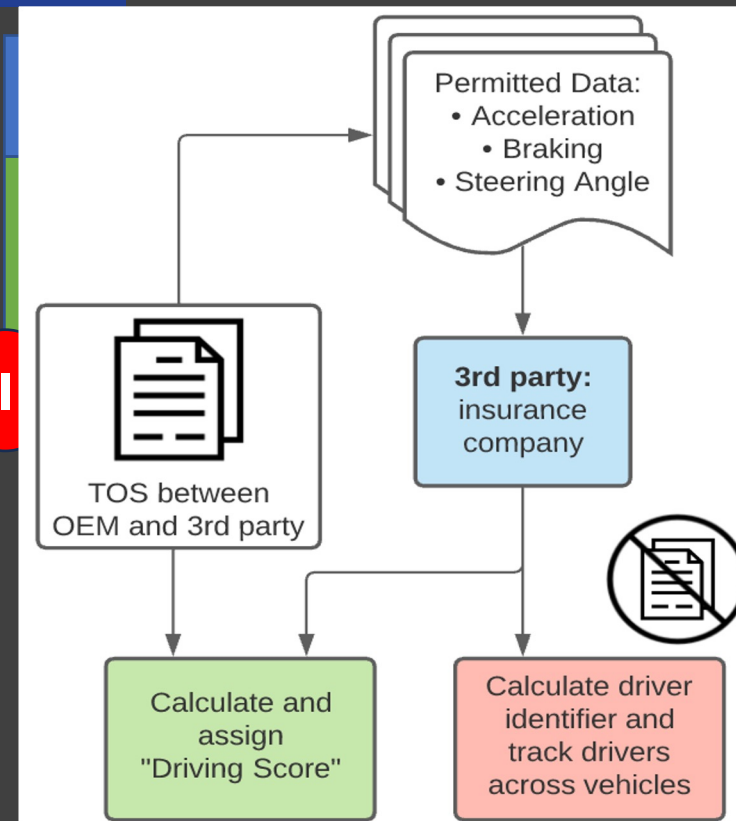
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PRIVACY GOALS

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Sanitization

Prevent 3rd party from mining more context than agreed

2

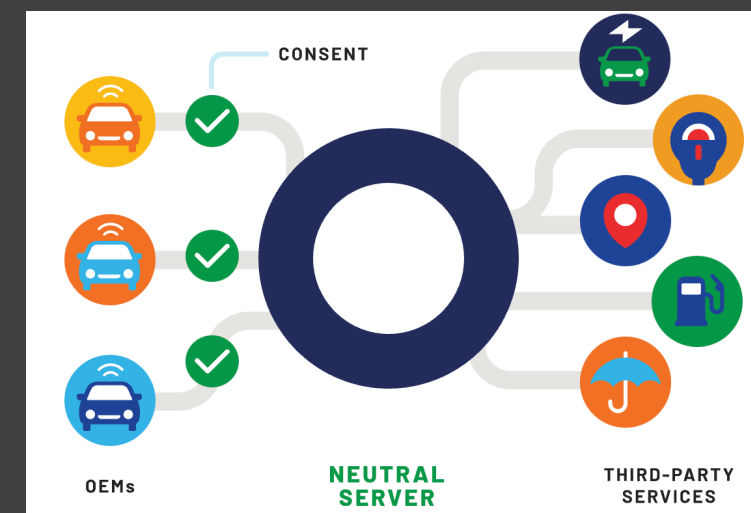
5

Data Sanitization

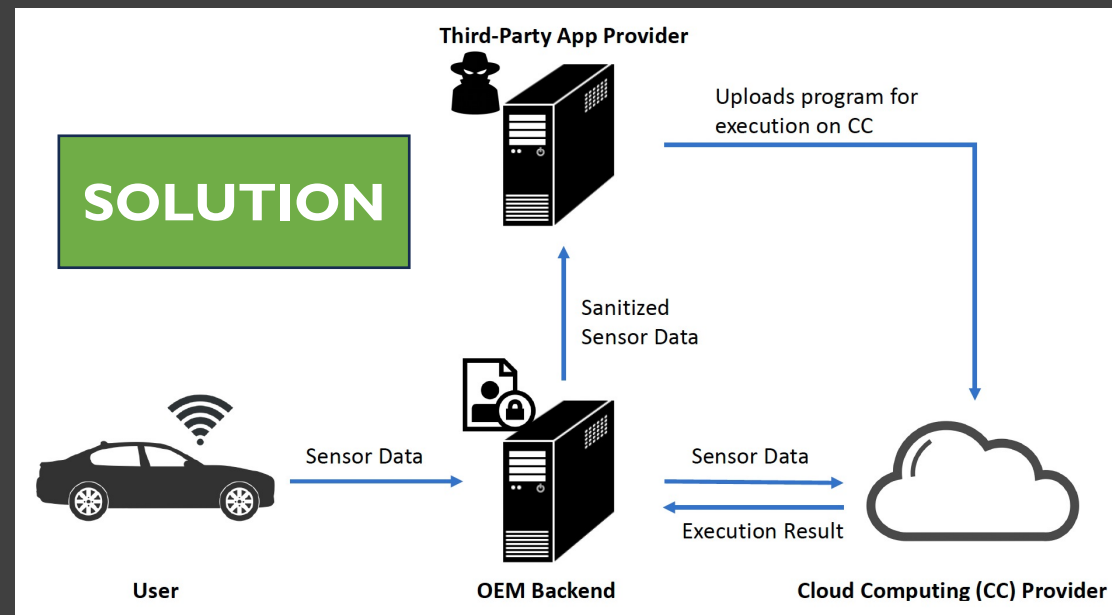
1. Lawfulness, Fairness and Transparency
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Existing Work

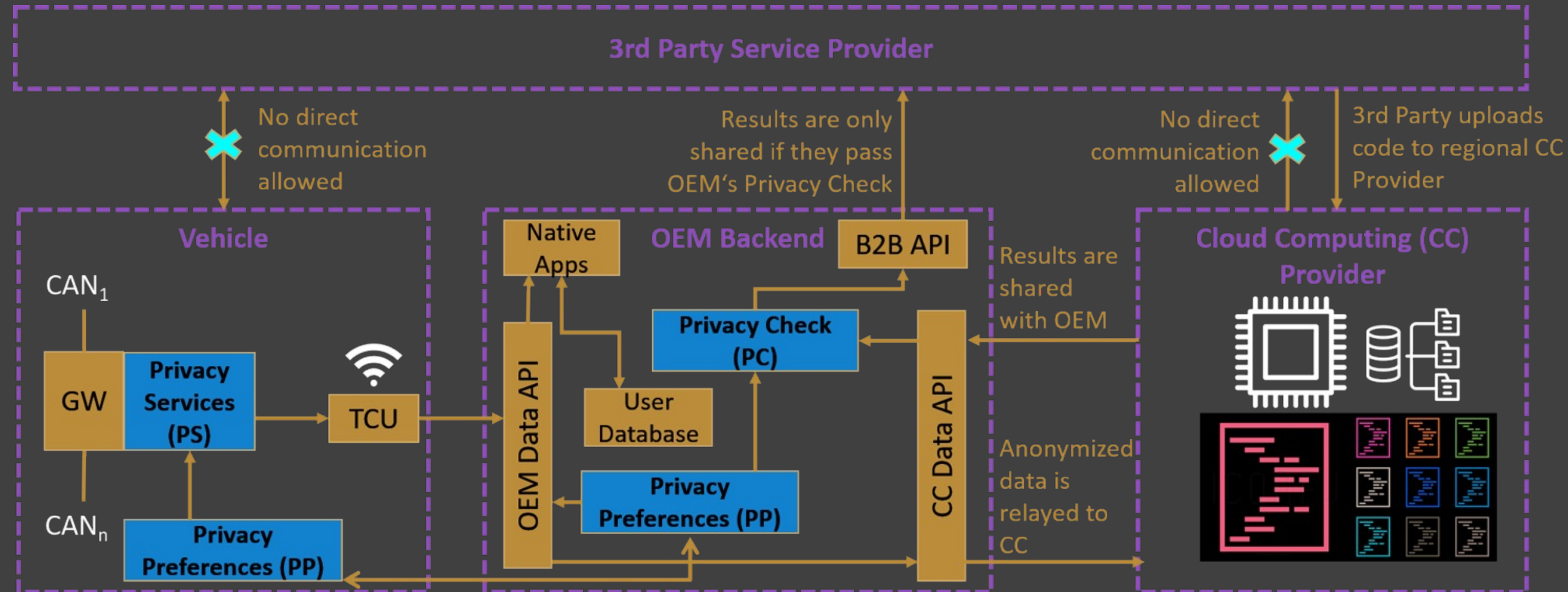
- Solid Project
- Neutral Server Concept (European Automobile Manufacturers' Association)



Problem: Rogue third-party can store data indefinitely and mine more context even if access is revoked!

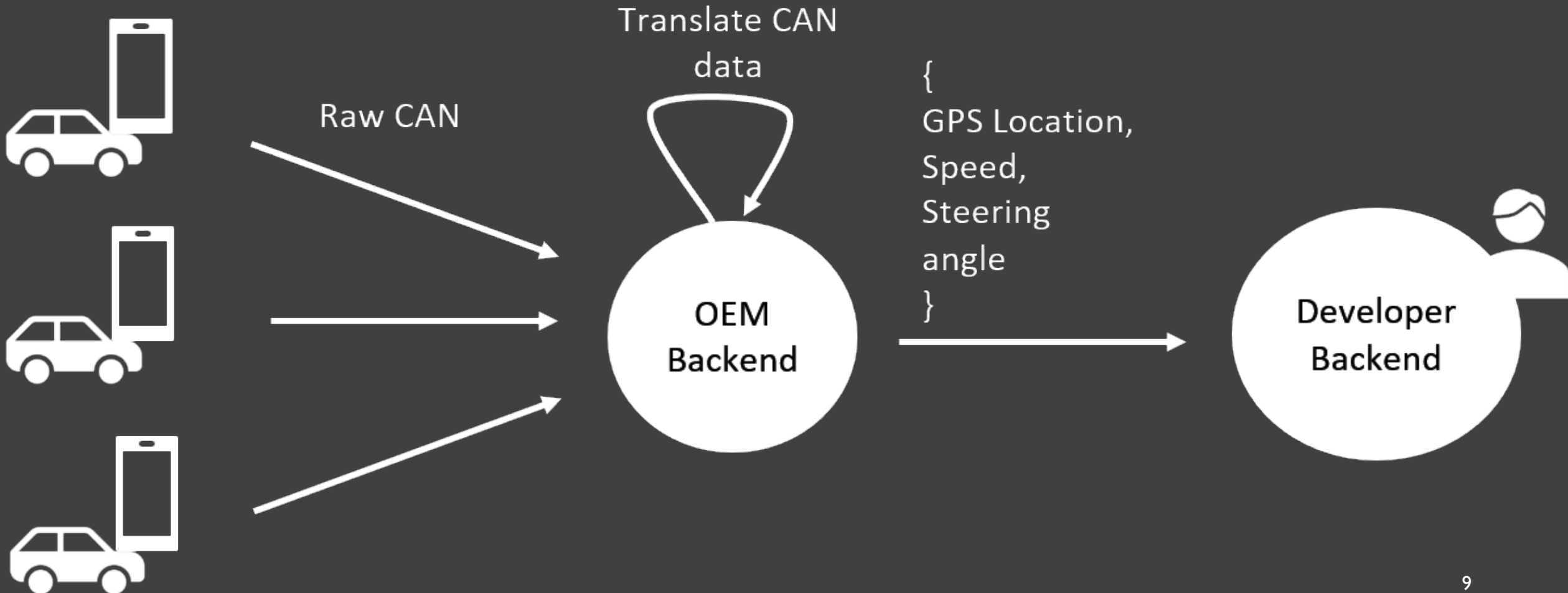


Reference Architecture



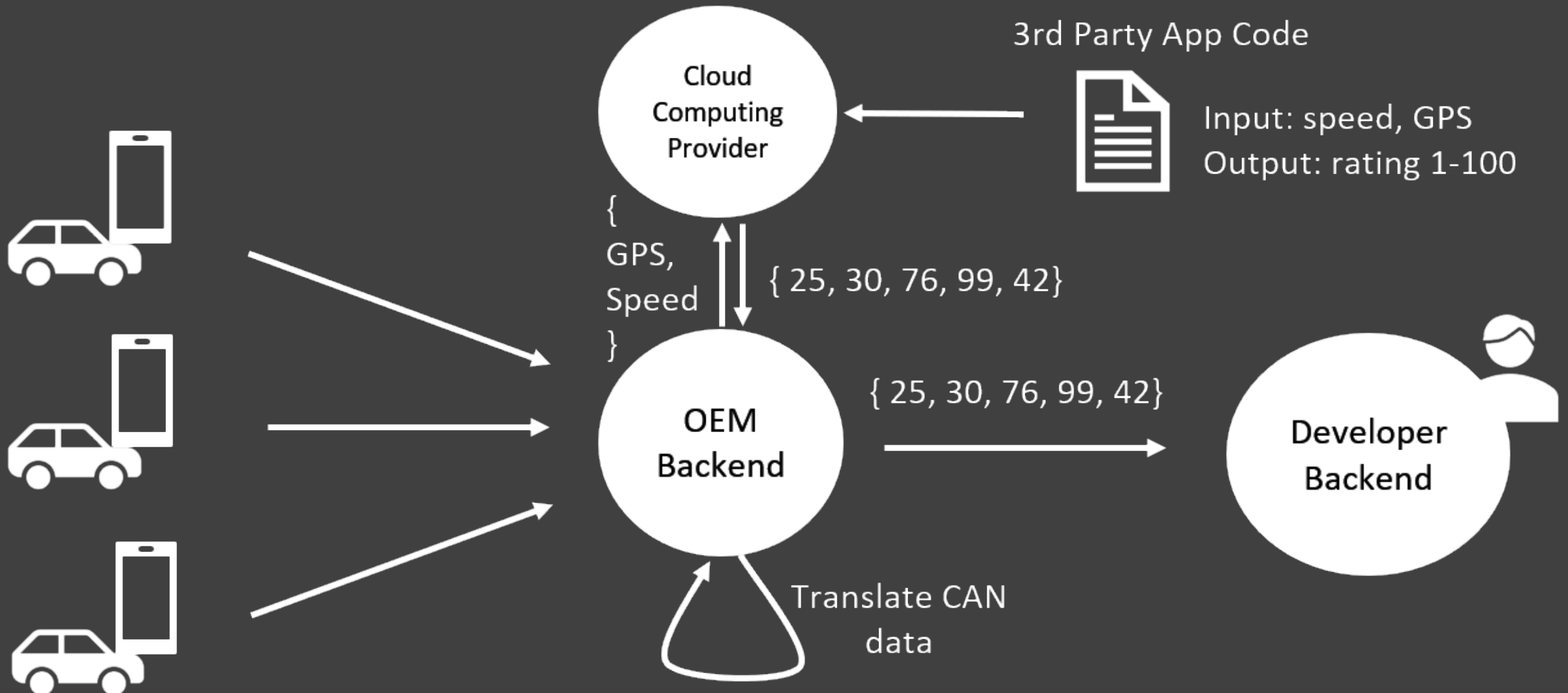
Example

Solves (2) Purpose Limitation +
(5) Storage Limitation



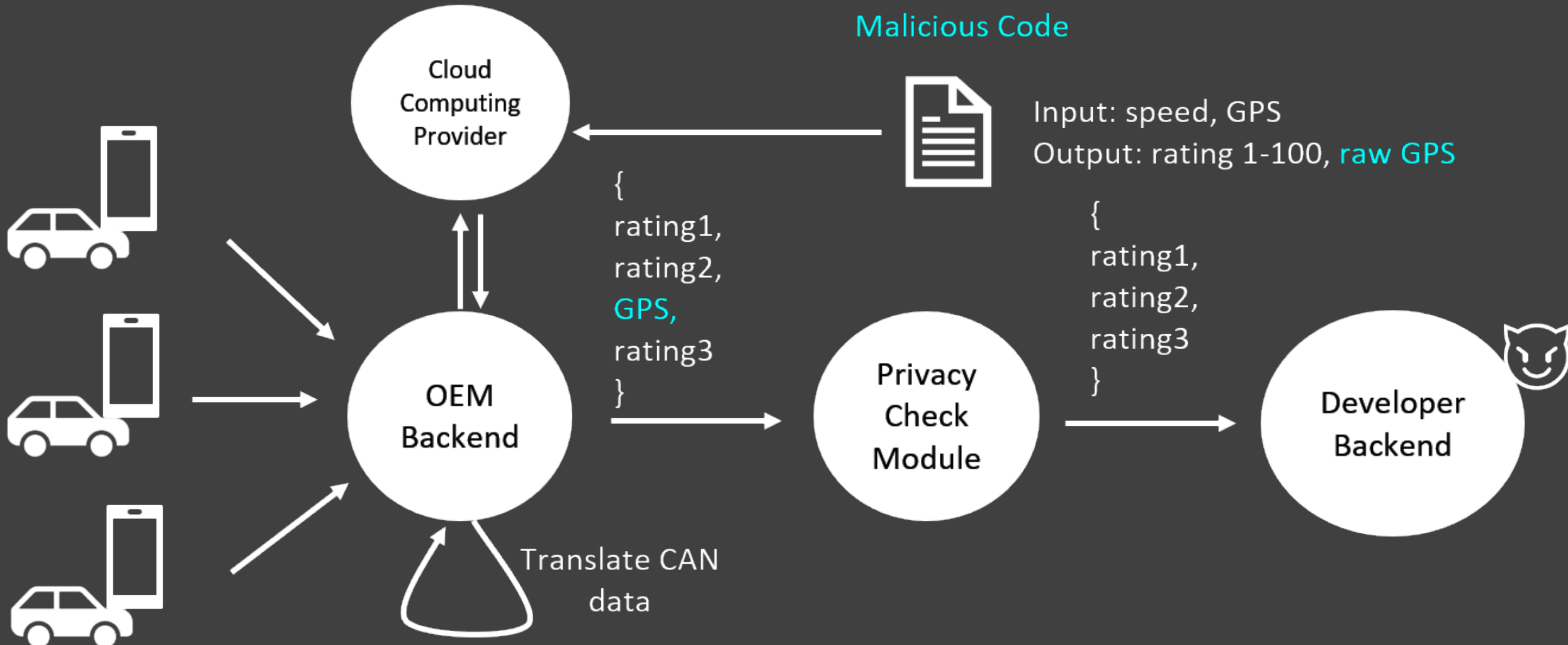
Example

Solves (2) Purpose Limitation +
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Example

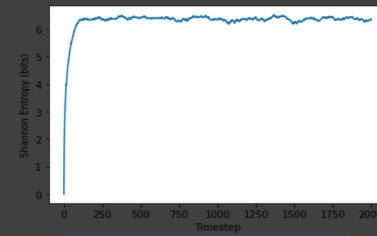
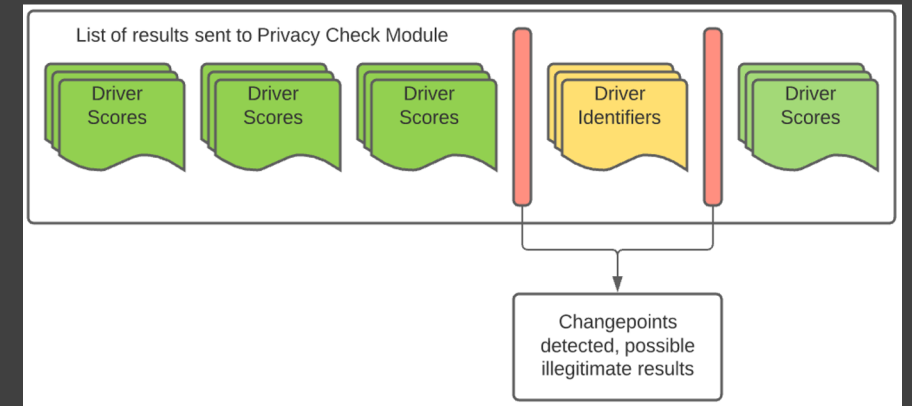
Solves (2) Purpose Limitation +
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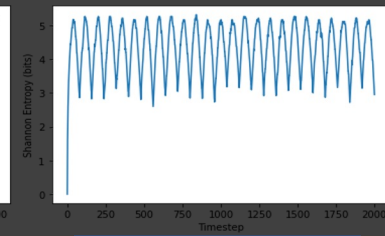
Privacy Check Module

- Goals
 - Approve legitimate results
 - Detect illegitimate results and stop transmission
 - Detect illegitimate results, even if disguised as legitimate
- Change-Point Detection (CPD)
 - Technique used for anomaly detection
- Entropy-Based Detection (EBD)
 - Use changes in the calculated entropy over time to find anomalies in input data

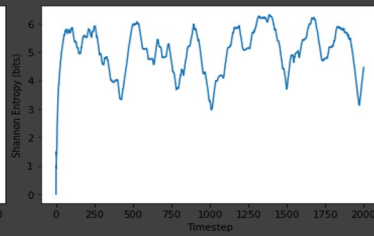
$$H(X) = - \sum_{i=1}^n P(x_i) \log P(x_i)$$



No anomalies



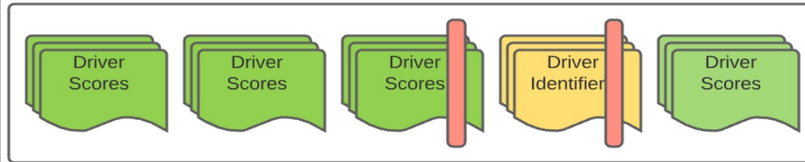
Periodic anomalies



Random anomalies

Evaluation: CPD

Coverage Score: Percent overlap between algorithm output and correct output



Detected changepoints are offset, reducing overlap with the correct output

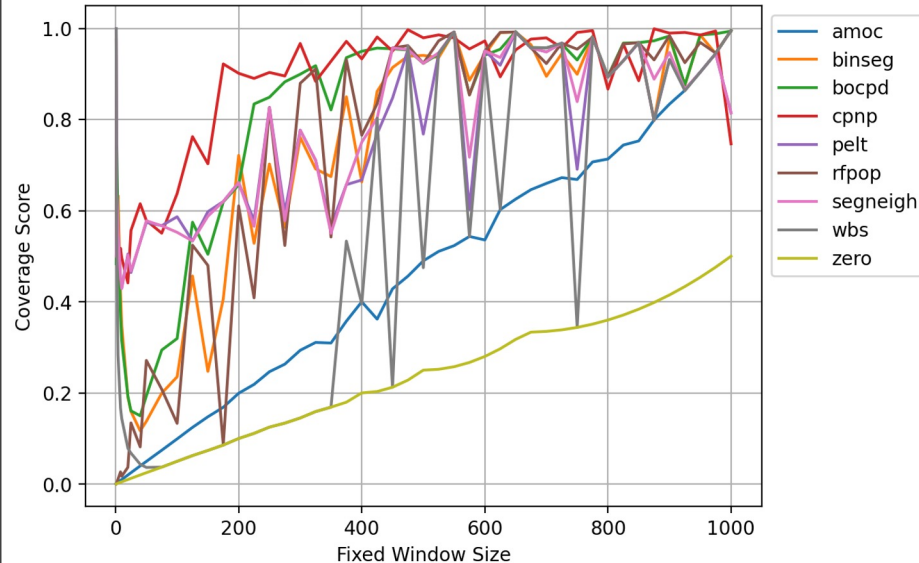
F1 Score: Measures the accuracy of the algorithm



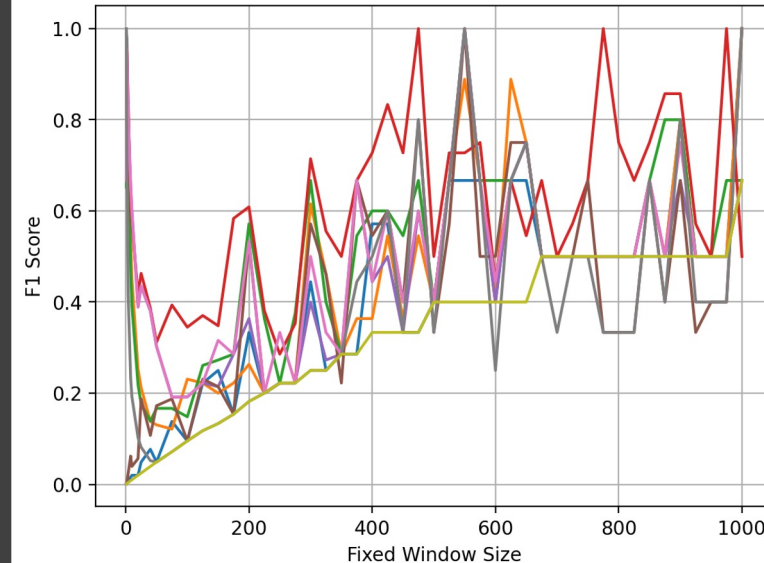
Algorithm incorrectly detected changepoint

Algorithm failed to detect changepoint

Coverage Score vs Fixed Window Size



F1 Score vs Fixed Window Size

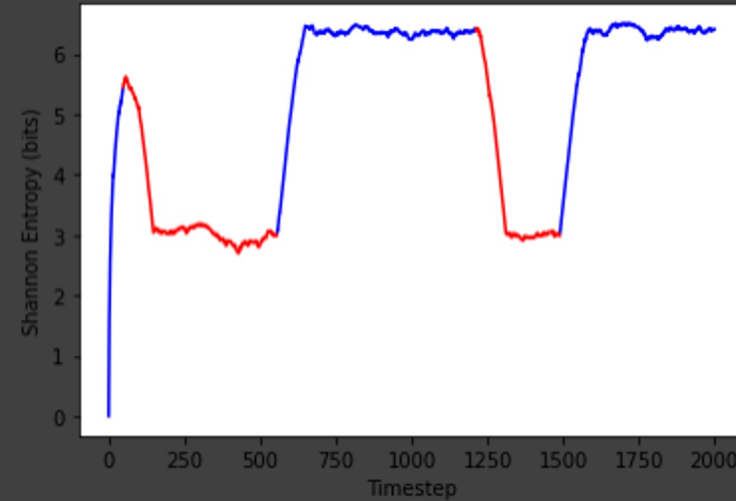


TPCDBench

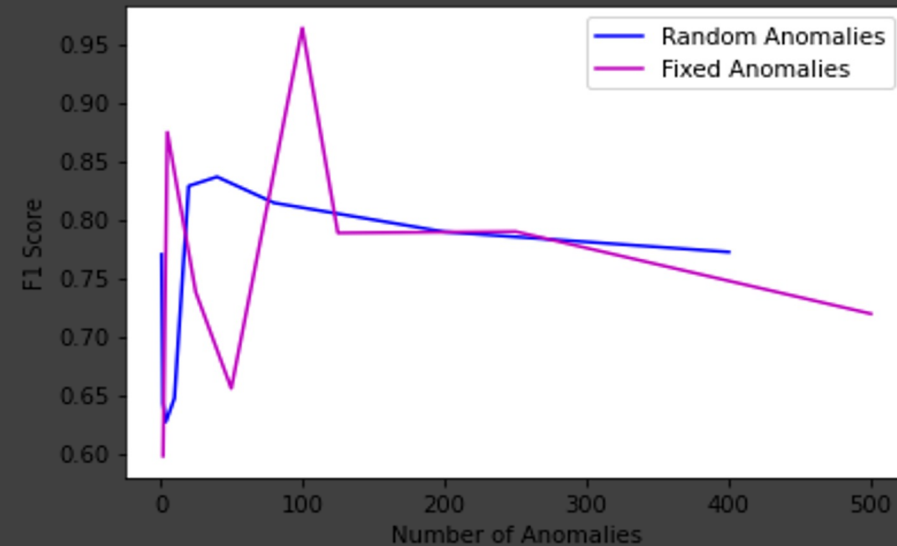
- Changepoint Detection can be used!
- Best average performance: CPNP and Pelt

Evaluation: EBD

- Negative gradient changes in entropy correspond to changepoints
- Good performance for both fixed and random anomalies



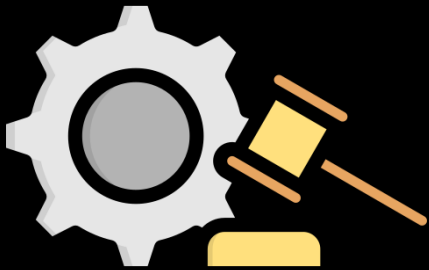
Red segments of data correspond to malicious data, while blue segments of data correspond to normal/benign data



Conclusion

First Privacy-Preserving Vehicular Data Collection and Sharing Platform

Privacy Goals



Definiton of three privacy goals that satisfy GDPR regulation

Architecture Design



Production-ready reference architecture for OEMs

Data Sanitization



Adapted and evaluted two anomaly detection techniques for data sanitization

Q & A

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