

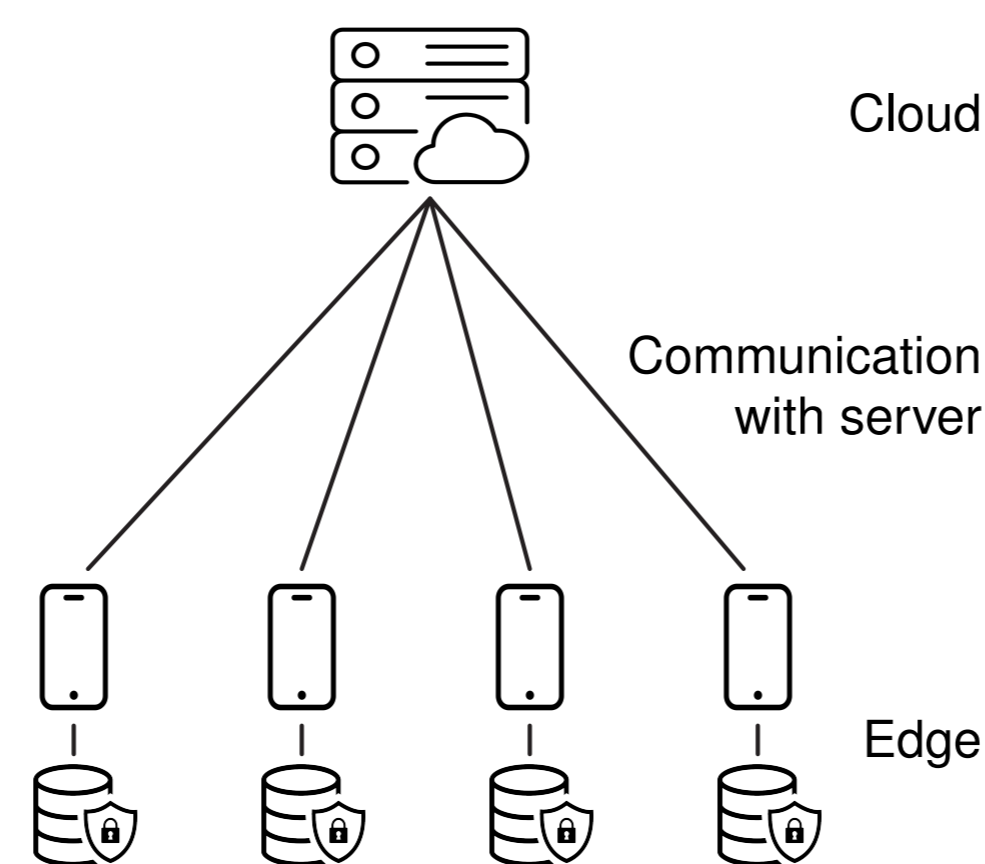
Tandem Outlier Detectors for Decentralized Data

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Scenario

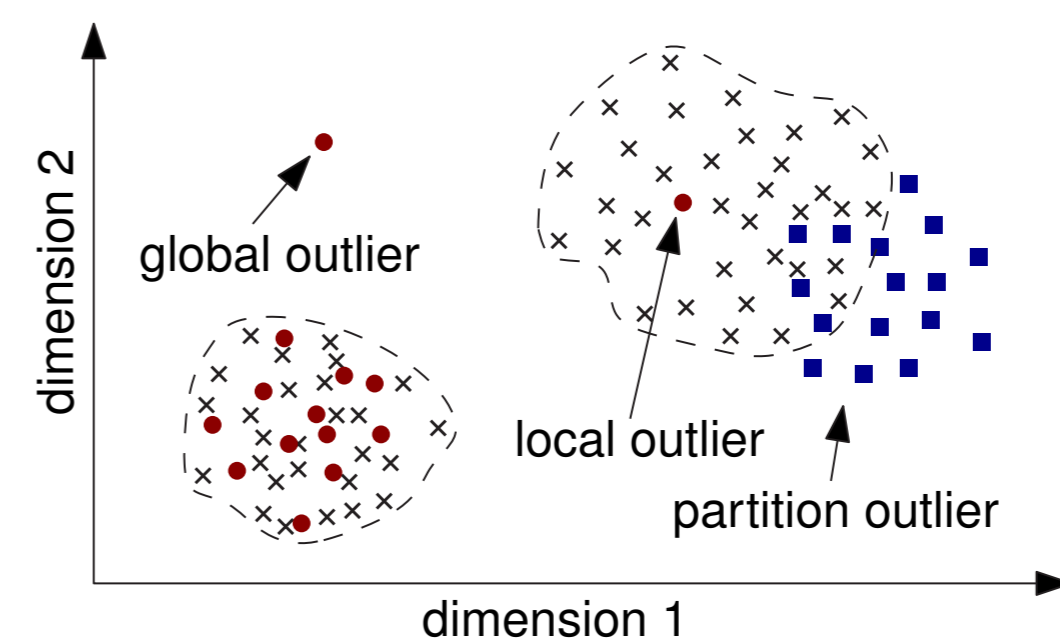
Finding outliers in decentralized data

- Data creation is often decentralized
 - Smart meters, powertools, electric cars
- Each device holds part of the data generated in the network (a *partition*)
- What outliers exist in decentralized data and how can we find them?



Types of Outliers

Local, global, and partition outliers



● partition A ■ partition B ××× other partitions

Results

We could identify mistakes in operation and unusual user behavior

- Partition outlier:** User performed the task very well

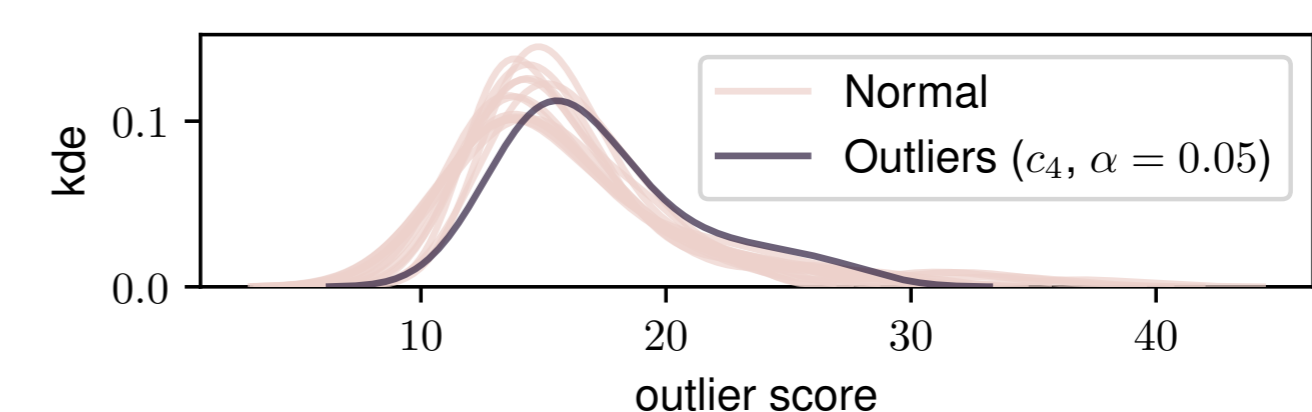


Figure: Partition outlier in powertool-data

- Local outliers:** Minor deviation from regular usage
- Global outlier:** User slipped off the screw-head

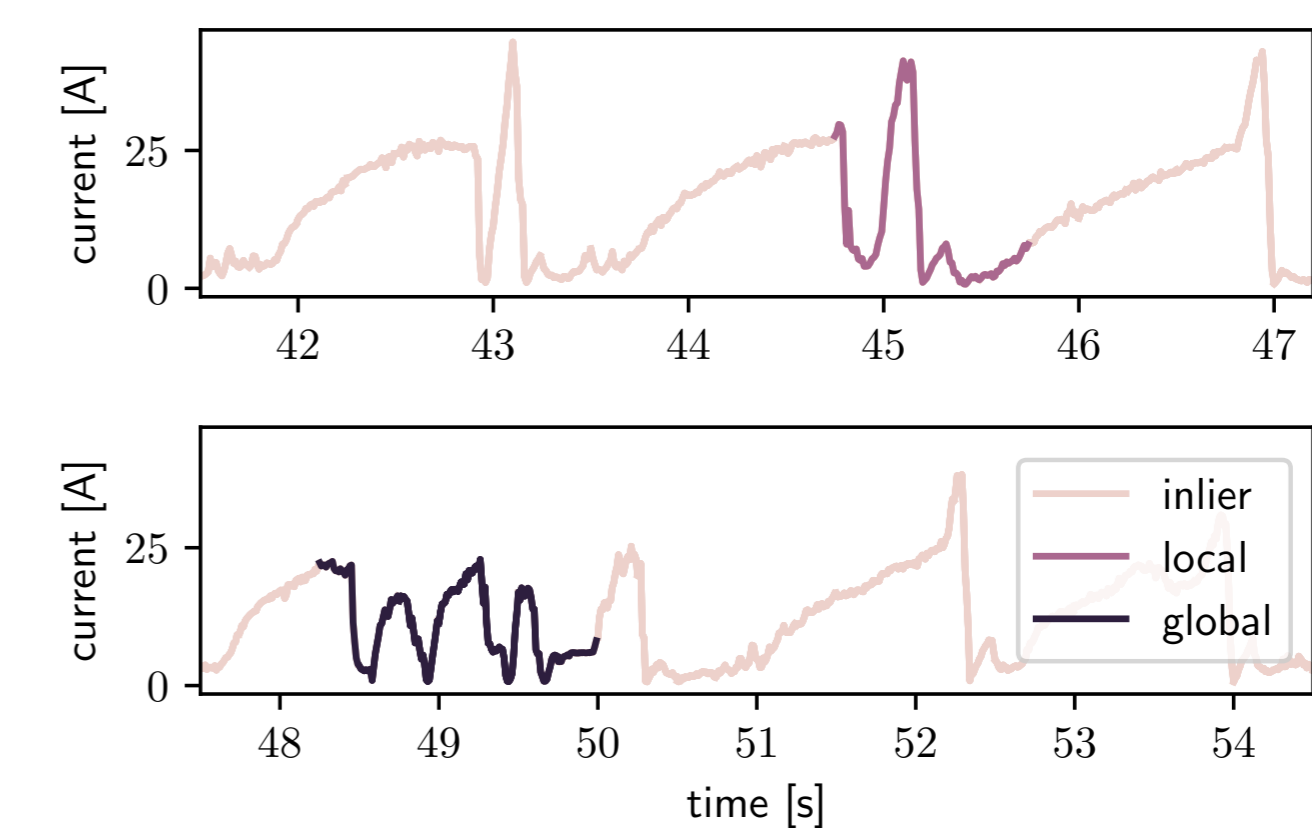


Figure: Local and global outlier in powertool-data

Local and centralized outlier detection are insufficient

- Centralized:**
 - Transfer data to central storage prior to outlier detection
 - Able to detect each outlier type
 - Problematic w.r.t. **privacy** and **efficiency**
- Local:**
 - Identify outliers in a single partition
 - Data remains on the device → **efficient and privacy-preserving**
 - Only **limited data available** for outlier detection

Research Question

How can we find **local, global, and partition outliers in decentralized data without sharing raw data** with other devices?

Solution Sketch

Combining federated and local outlier detectors

- Two outlier detectors L_i and F :
 - F : on single partition d_i
 - L_i : on entire network N with data $D = d_1 \cup d_2 \cup \dots \cup d_{|N|}$ using Federated Learning
- Outlier scores for device i :
 - $os_i^F = \{os_{i,1}^F, os_{i,2}^F, \dots, os_{i,d_i}^F\}$
 - $os_i^L = \{os_{i,1}^L, os_{i,2}^L, \dots, os_{i,d_i}^L\}$
- Distinguish point outliers:
 - Global outlier if $os_{ij}^L > \epsilon_i$ and $os_{ij}^F > \epsilon_i$
 - Local outlier if $os_{ij}^L > \epsilon_i$ and $os_{ij}^F \leq \epsilon_i$
- Identify partition outliers
 - Sort os_i^F , create bins of size b , compute averages
 - Set of average scores is os_i^*
 - Mann-Whitney-U test between os_i^* and $\bigcup_{k \in N \setminus i} os_k^*$

User Study

Connected Powertools

Protocol

- 15 participants
- 20 screwing tasks; 20 drilling tasks
- Prespecified goals (e.g., target depth)
- Acceleration, sound, magnetic field, gyroscope data

Outlier detectors

- F and L_i : single-layer Autoencoders
- Reconstruction error = Outlier score
- Trained on features extracted from sensor data in sliding windows of 25 ms



Related Work

- Server-coordinated [4] and peer-to-peer [6] Federated Learning
- Federated Learning for outlier detection [8, 9, 7, 5]
- Other definition of "local": outlier occurs multiple times in a small geographical region [3, 2, 10]
- Wireless Sensor Networks [1, 2]

Data and Code

<https://github.com/heymarco/TandemOutlierDetection>

References

- S. Bharti, K. K. Pattanaik, and A. Pandey. "Contextual outlier detection for wireless sensor networks". In: *J. Ambient Intell. Humaniz. Comput.* (2020).
- H. H. W. J. Bosman, G. Iacca, et al. "Spatial anomaly detection in sensor networks using neighborhood information". In: *Inf. Fusion* (2017).
- N. Giatrakos et al. "TACO: tunable approximate computation of outliers in wireless sensor networks". In: *SIGMOD*. 2010.
- B. McMahan et al. "Communication-Efficient Learning of Deep Networks from Decentralized Data". In: *AISTATS*. 2017.
- T. D. Nguyen, S. Marchal, et al. "D²IoT: A Federated Self-learning Anomaly Detection System for IoT". In: *ICDCS*. 2019.
- Abhijit Guha Roy et al. "BrainTorrent: A Peer-to-Peer Environment for Decentralized Federated Learning". In: *CoRR* abs/1905.06731 (2019).
- R. A. Sater and A. B. Hamza. "A Federated Learning Approach to Anomaly Detection in Smart Buildings". In: *CoRR* (2020).
- J. Schneible and A. Lu. "Anomaly detection on the edge". In: *MILCOM*. 2017.
- S. Singh, S. Bhardwaj, et al. "Anomaly Detection Using Federated Learning". In: *ICAIA*. 2021.
- X. Yu, H. Lu, et al. "An adaptive method based on contextual anomaly detection in Internet of Things through wireless sensor networks". In: *IJDSN* (2020).