

Electric Vehicles

CPS and IoT Security

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Master Degree on Cybersecurity



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- Recent climate crisis demands green alternatives to replace technologies with high environmental impact
- Electric Vehicles (EVs) have been proposed as a green alternative, where electric batteries are employed as a power source
- Governments are incentivizing the adoption of EVs thanks to the deployment of a large number of Electric Vehicle Supply Equipment (EVSE) in public charging infrastructure
- [Ban of gas-fueled cars](#)

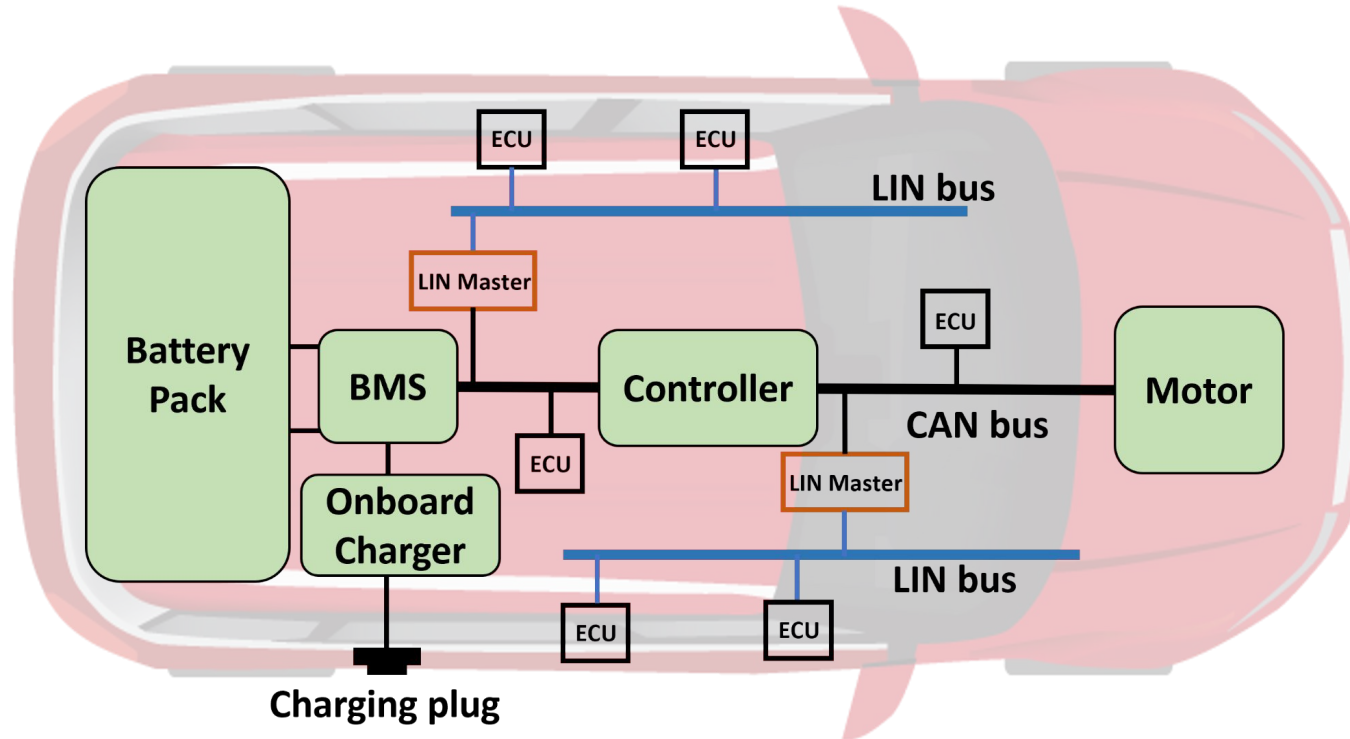
Components of EVs



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- **Battery:** The battery is where the charge is stored in the form of Direct Current
- Batteries are usually combined in packs and connected in series or parallel to increase the voltage and Ampere/hour they can deliver to the EV
- Batteries suitably combined are enclosed into a metal casing to prevent damage
- The case usually includes a cooling system to avoid damage due to batteries overheating



- **Battery Management System:** manages all operations regarding the battery
- Manages the current output and the charging and discharging of the battery by keeping it in a safe operating area
- Monitors each battery in the pack and measures each cell's voltage, current, and temperature
- Is instructed with a threshold limit for each of them and disconnects the load if values exceed the threshold value



- **Battery/On-Board Charger:** provides an interface between the charging system and the EV battery
- The charger converts the input voltage to DC and passes it to the battery for storage
- It prevents possible damages to the battery or the supply system (e.g., overheating) by limiting the power flow



- **Controller:** handles the flow of current from the battery to the EV associated with all operations, ranging from motors-related operations to powering the infotainment system
- Receives the input from the driver to control the acceleration, brake pressure, and driving mode and converts the energy in the battery from DC to AC
- The controller converts the generated AC to DC such that the energy can be stored in the battery

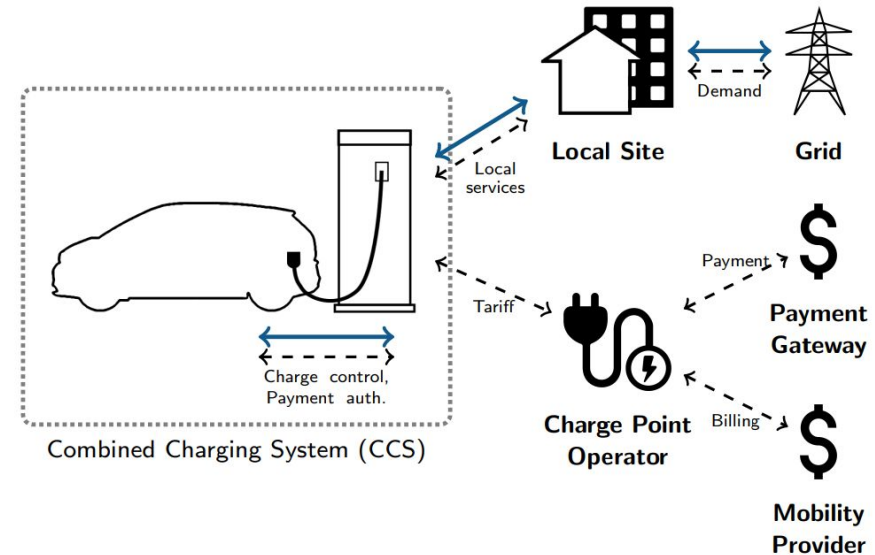


- **Electric Motor:** the motor is powered by the EV battery
- The electric motor communicates with sensors and actuators in the EV that control the amount of thrust required
- There exist many implementations of electric motors
- The most commonly used for EVs are AC induction due to their lower cost implementation thanks to the absence of permanent magnets

Losing the Car Keys



- Physical layer attack against the Combined Charging System
- Exploit electromagnetic side channel attacks to tamper with the Power Line Communication (PLC)
- The unintentional wireless channel is sufficient to recover messages in the vast majority of cases





- PLC design assumes differential signalling, wherein two transmission lines are driven with equal but opposite signals
- However, practical implementations of PLC circuit connect one of the transmission lines to the ground
- The resulting single-ended signalling creates a suitable antenna for emission or inference



- We consider a passive attacker, eavesdropping the charging process via unintended radiations
- Eavesdrop general purpose communication between EV and charger
- The attacker can get closer, but cannot modify or tamper the equipment
- The data collection can be done both in presence or remotely

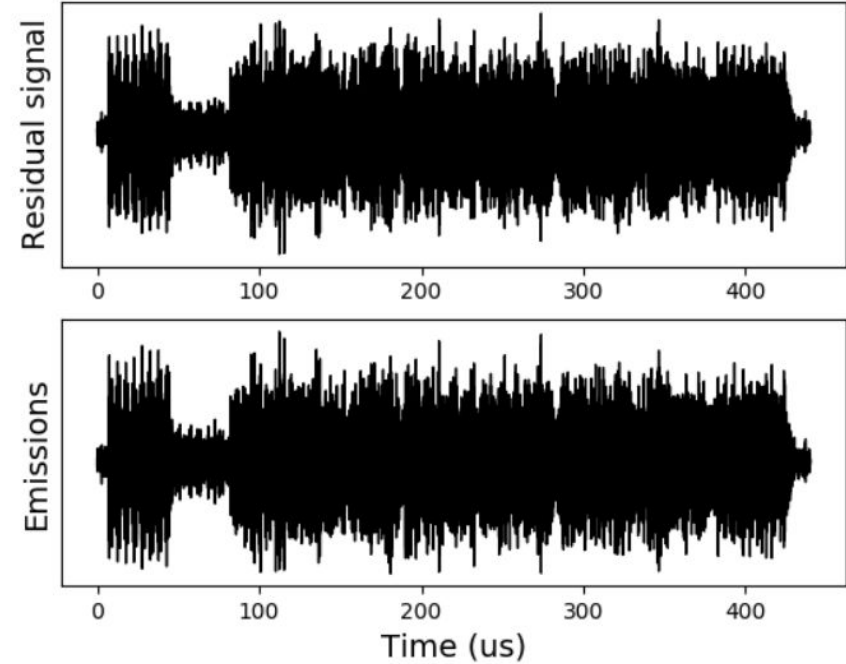
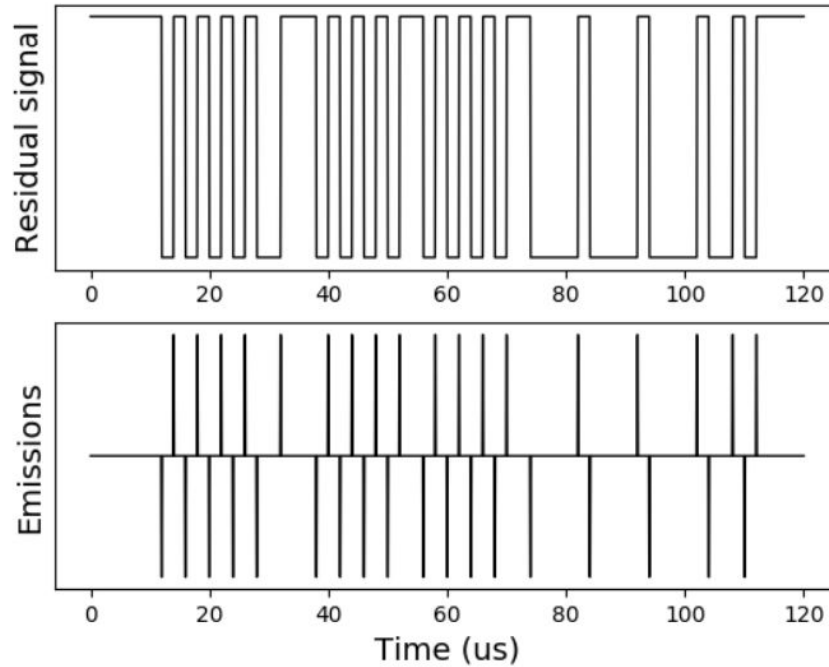
Emission Examples



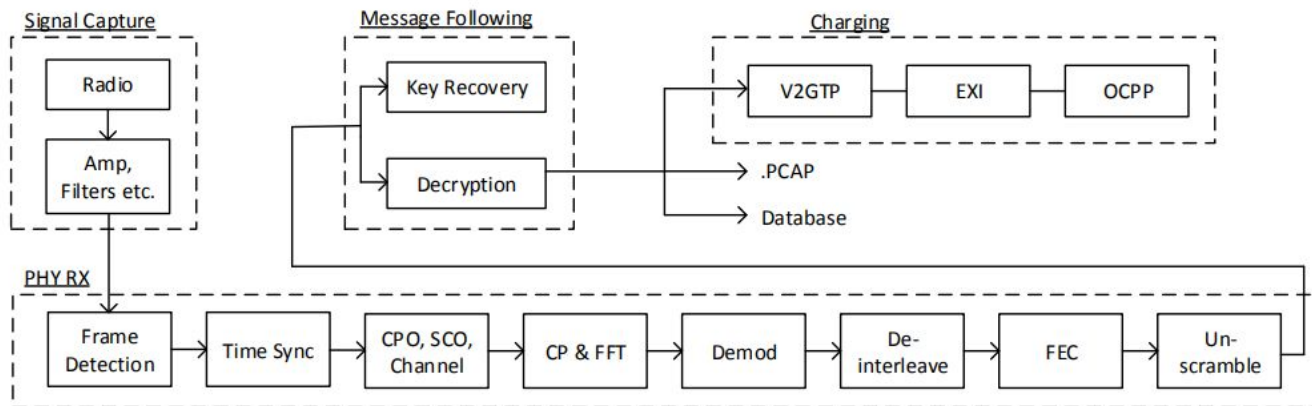
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- Task of the attacker: maximize the signal to noise ratio and bandwidth of the receiver
- Challenge: any exposed component might be generating unintended EM signals
- The tool resembles a HomePlug GreenPHY receiver





- Three different EV models, collecting a total of 54 unique charging sessions
- Use the eavesdropping tool to reconstruct the digital information
- The receiver antenna can be both inside the car, outside of it, or at a certain number of bays away

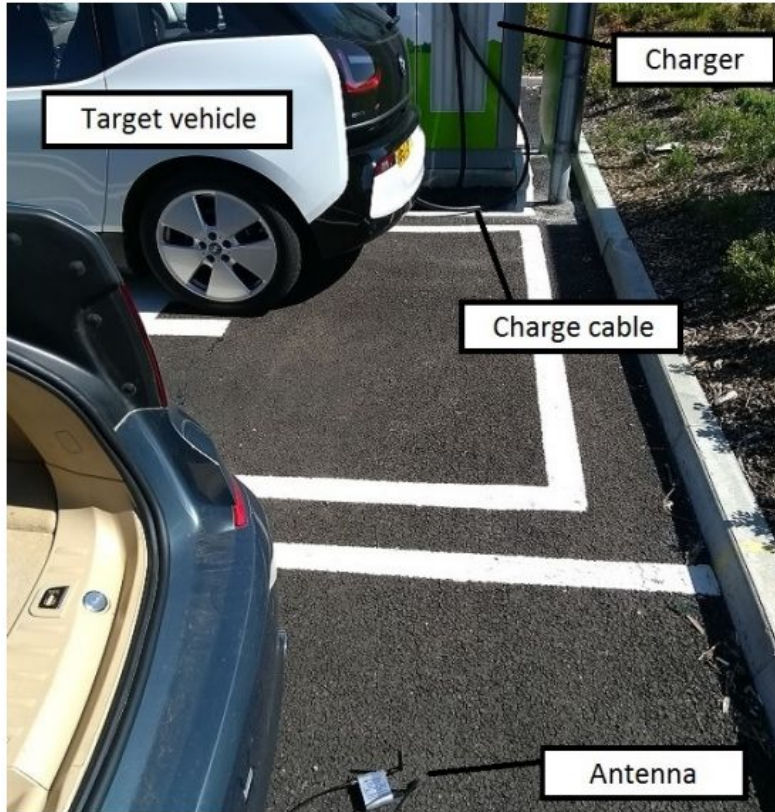
Measurement Campaign



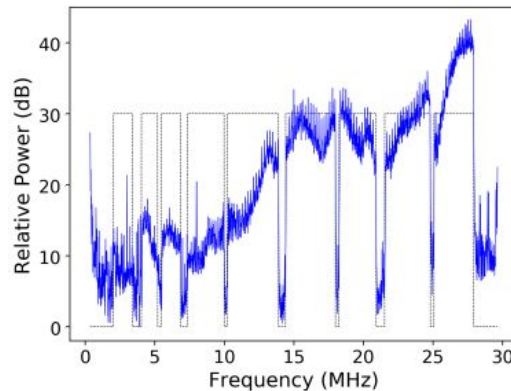
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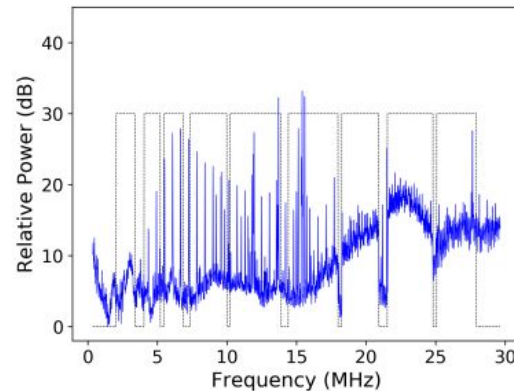
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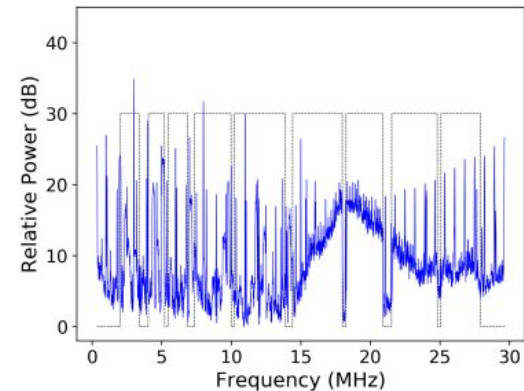
- Measured signals at different locations
- Overlaid HPGP spectrum to show where communication occurs
- Every site displays some form of unintentional communication
- We see the effect of distance from the transmitter



(a) Antenna by cable (site H)



(b) Antenna in bay behind (site F)



(c) Antenna in next bay (site G)



- Thanks to the received signals and its high SNR, decoding has a high success probability
- In the best case (in car, peak SNR = 20 dB, BW = 12 MHz) the rate of correctly decoded CRC is 100 %
- The proposed approach is also valid in case of multiple EVs simultaneously charging
- The success rate however varies from 24.3% to 94.8%



- The first phase is capturing the initialization of the charging session
- If that is the case, thanks to the attack it is possible to examine the network membership key exchange
- Since there is no encryption, if the message is received intact it is possible to determine the network membership key
- Successful in 31 cases when considering single EV
- When considering side-by-side vehicles, successful in extracting one NMK in 4 session, and both NMK in one



- As the vehicle and charger establish a network, the vehicle undertakes the discovery protocol to find a charge controller and the two established a TCP connection
- However, there is no establishment of TLS tunnel
- This leaves the high level protocols exposed
- There is still an open debate on the type of PKI to deploy, which limits the applicability of TLS
- Currently, infrastructures rely on the physical security of location and cabling



- Electric Vehicles and their charging infrastructure are cyber-physical systems
- We discuss how the different threat vectors can lead to multiple types of attack, undermining the communication security or jeopardizing the privacy of the users
- We discuss attacks based on the current exchanged during the charging process and the communication protocols needed to establish connections or to manage the charging process

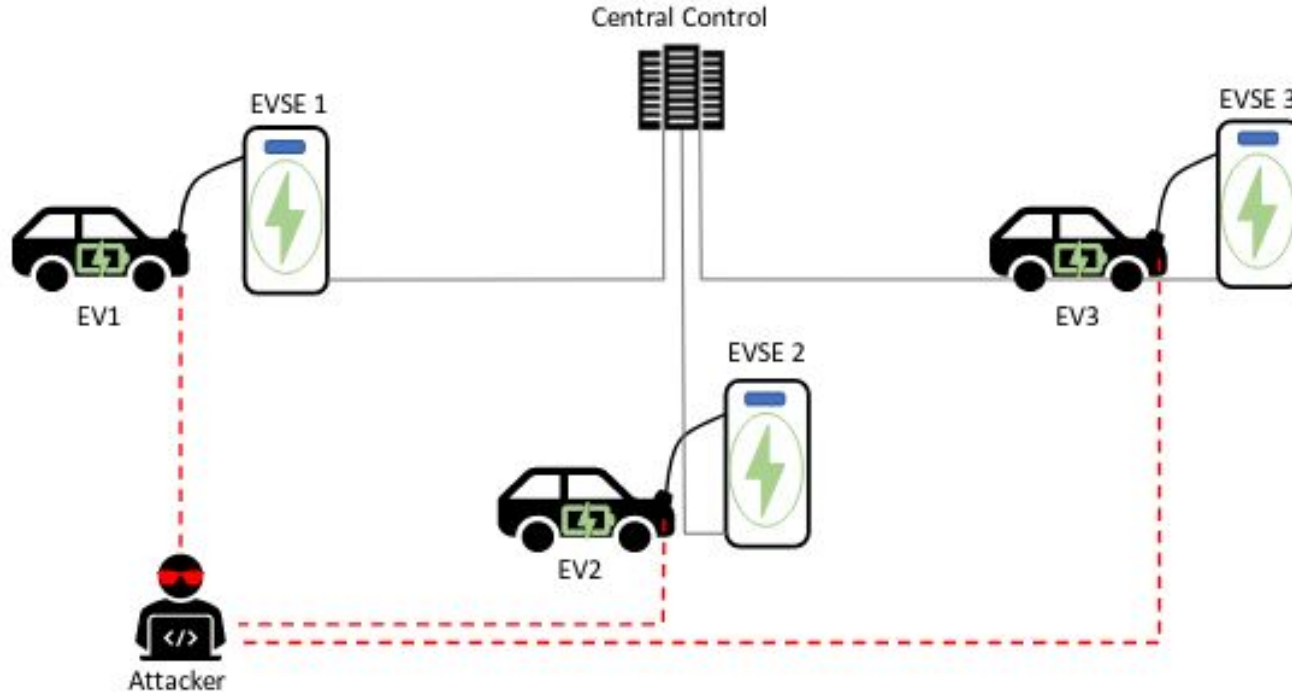


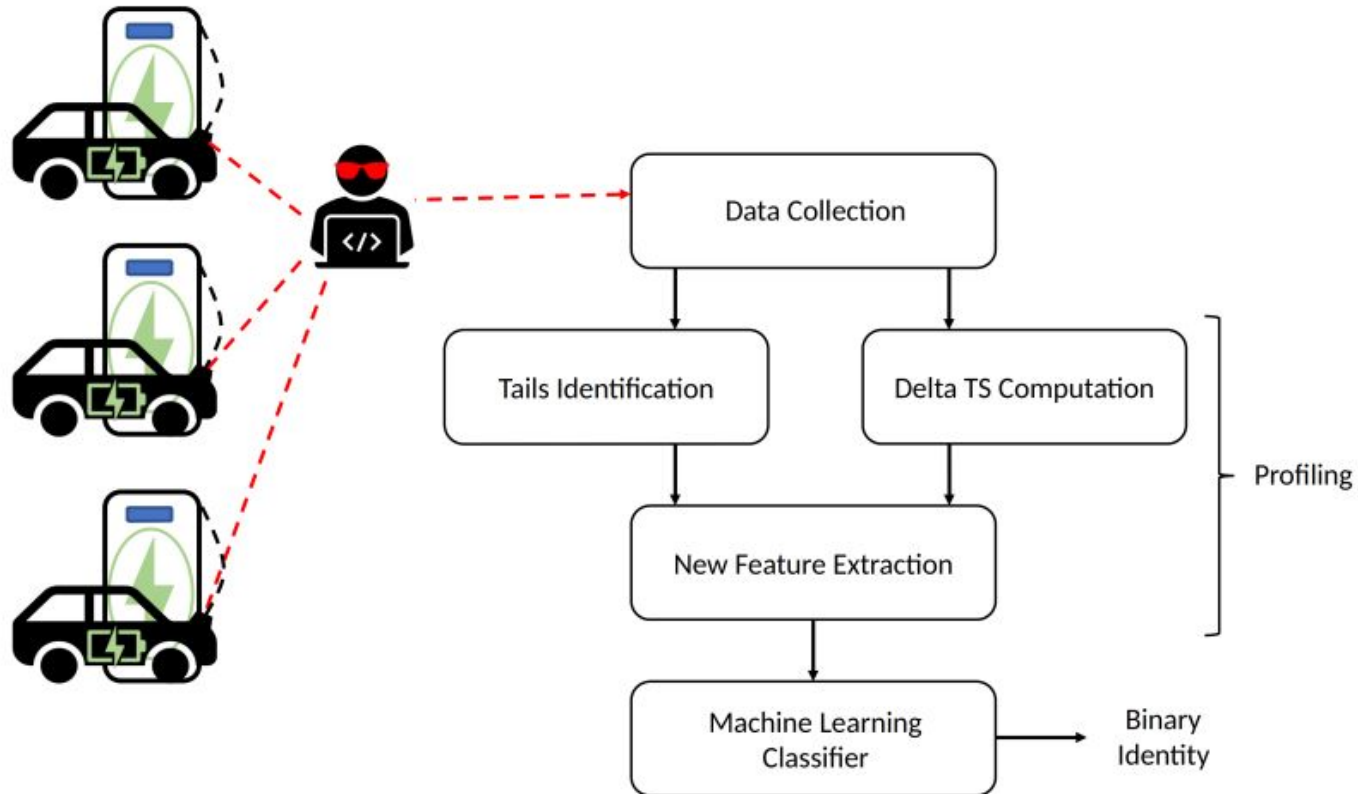
Motivation:

- Public infrastructures are easily accessible by anyone
- Attackers can physically tamper the charging infrastructure to collect sensitive data

Goal:

- Exploit physical charging information to identify vehicles
- Malicious purposes: e.g., Tracking
- Benign purposes: e.g., Authentication

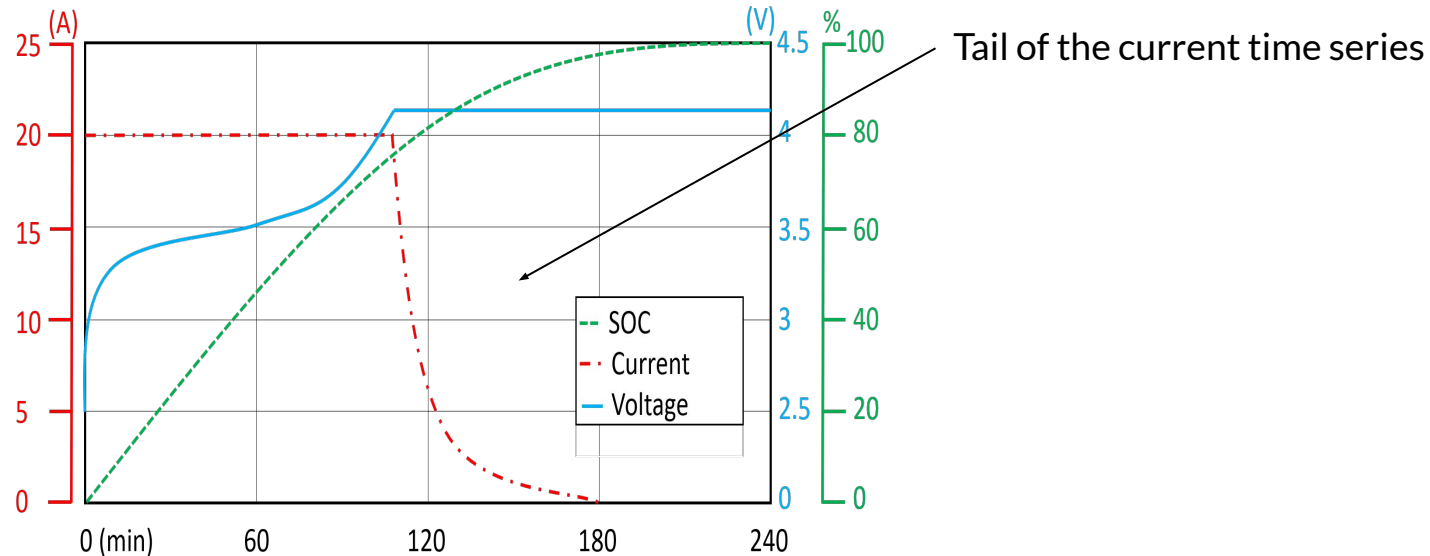




- Data of charging sessions of real devices from [1]
- TS data connect connected to users identifiers
- Total number of EVs considered: 187
- Information:
 - vehicle identifier
 - arrival and departure time
 - kwh delivered
 - current and pilot time series
 - disconnection time



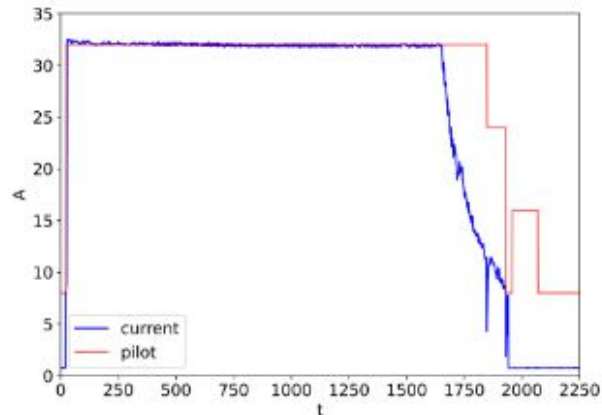
- Batteries charging model: constant current-constant voltage
- Tails include features to recognize the battery model [2]
- Objective: design an algorithm to identify and extract tails



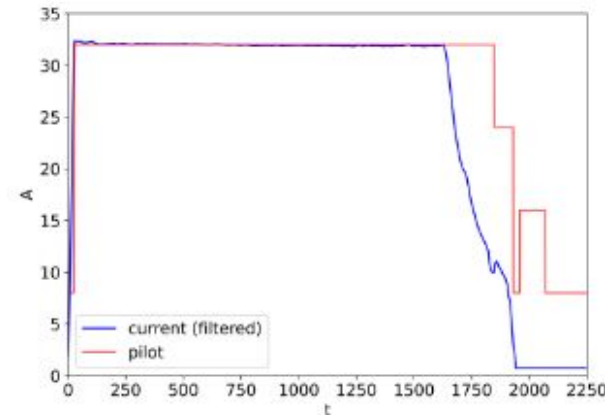


- Input: set C, P of filtered current and pilot time series, respectively
- Output: set TC , TP of current and pilot tails, respectively
- Steps:
 - Identify current TS ending point (zero values thereafter)
 - Identify backward ascending trend
 - Given the tail timestamp values, extract tail for both current and pilot
 - Add them to the sets TC , TP of current and pilot tails, respectively
- Feature set: 18 features including mean, variance, slop, time duration, min and max values.

- The charging itself also characterizes the battery
- Pilot \neq Current absorbed
- Objective: Characterize the normal charging behavior
- -> Compute Delta TS = Difference between pilot and absorbed currents



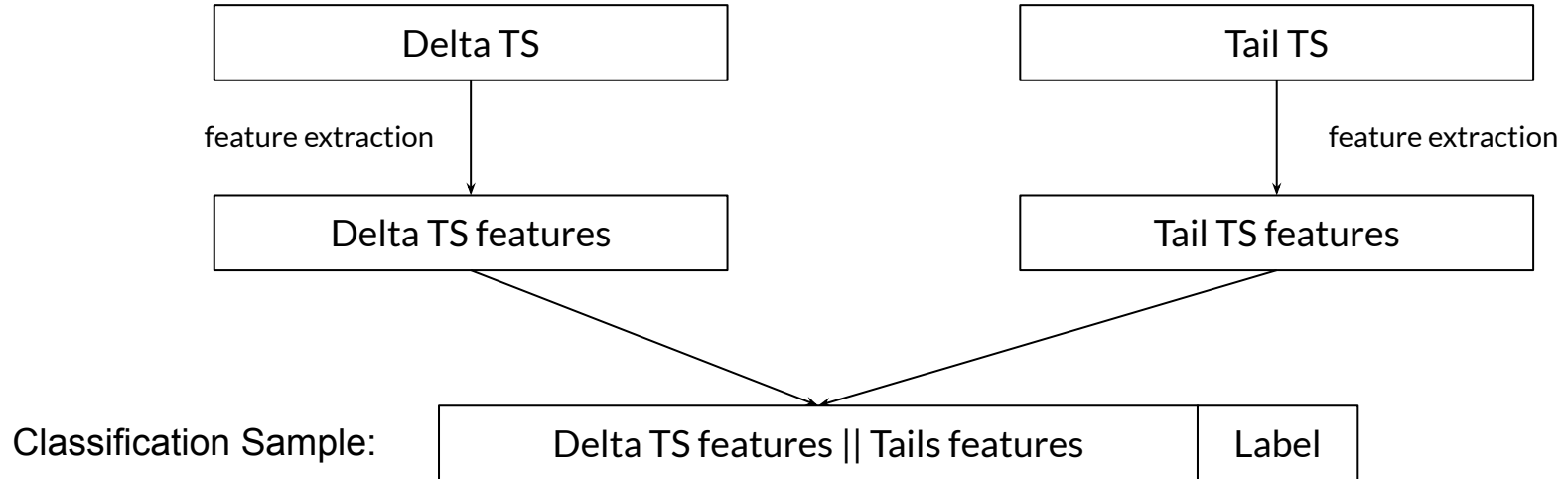
(a) Raw current profile.



(b) Current filtered with a moving average filter with $N_{avg} = 25$.



- Deleted not complete charges
- Automated Features extraction from TS
 - TSFresh: Lot of features for time series classification
- Classification sample:
 - Delta TS features || Tails features





- Implemented classifiers:
 - Support-Vector Machine (SVM)
 - k-Nearest-Neighbours (kNN)
 - Decision Tree (DT)
 - Logistic Regression (LR)
 - Random Forest (RF)
 - ADA Boost (ADA)
- Unbalance ratio parameter Q
- Hyperparameter selected via grid search
- Performance measure: precision (P), recall (R), and F1
 - To cope with unbalancing: G-Mean
- Classification problem as Authentication

- $F1$ scores for different ratios Q
- Same model kNN
- Varying the NoF from 10 to 200.
- Negligible increase from 100 features up
- We selected $NoF = 100$

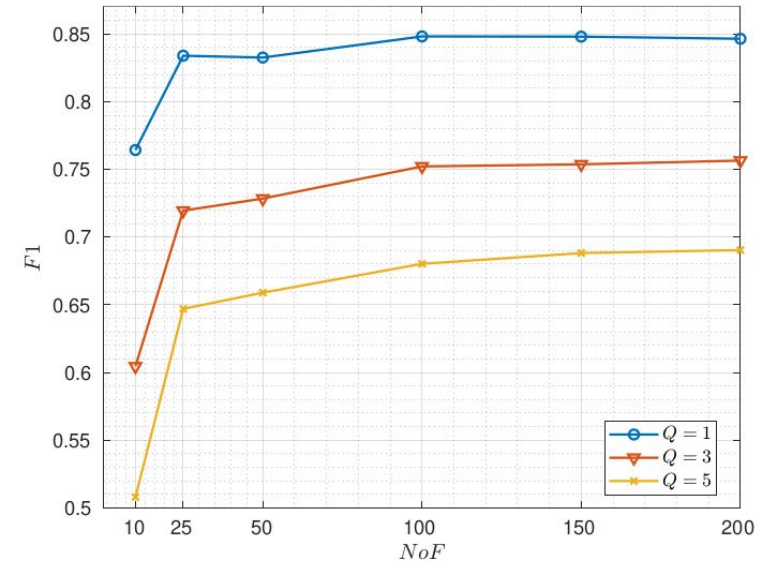
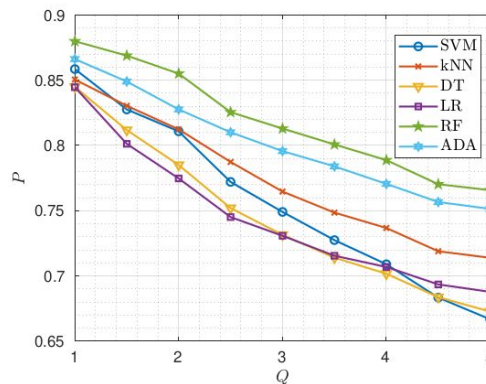
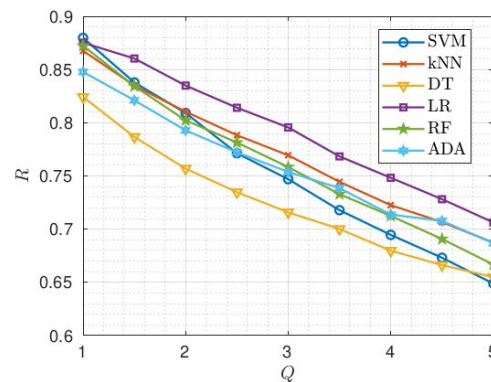


Fig. 5. $F1$ score of kNN classifier with different numbers of features $NoFs$.

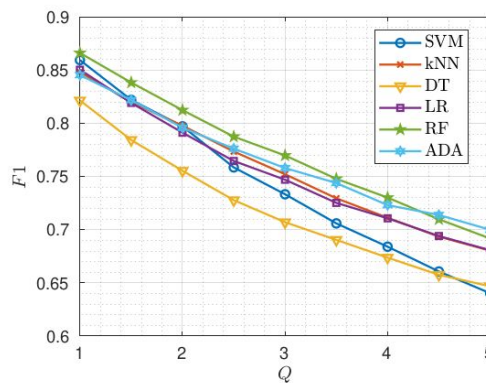
Results - Unbalance Effect



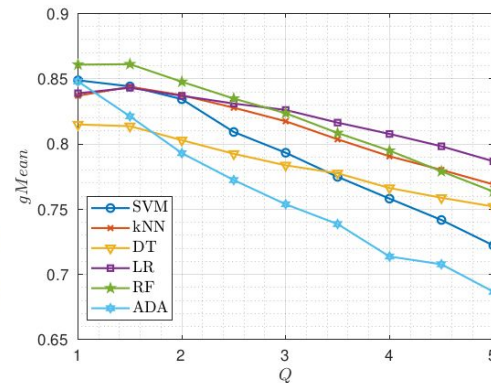
(a) precision



(b) recall



(c) F1



(d) G-Mean