

- How to defend connected intelligent vehicles: Transferring established Information Security best practices to the vehicular world Miriam Gruber and Jan Lange, Volkswagen
- Detection is not enough: Low-cost Attack Recovery for Autonomous Robotic Vehicles (RVs)
  Karthik Pattabiraman, University of British Columbia
- Design and Assessment of Safe Autonomous Vehicles (AVs) Saurabh Jha, IBM T. J. Watson Research
- Session Chair: Andrea Ceccarelli, University of Florence
- Rapporteur: Homa Alemzadeh, University of Virginia

## ifip Transferring IT Security Best Practices to IVs WG.10.4

- Collision of two worlds: Information security and automotive safety
- From prevention to active defense
  - **Prevention:** Interface protection, SW integrity, authenticated communication
  - Defense: Intrusion detection, intrusion reaction, active defense, and recovery
  - Challenges:
    - New technology, timing constraints, increasing complexity, fixed rules
- Incident response:
  - Active Attack Detection
  - Response
- Al-based defense:
  - AI-based Detection:
    - Learn from real-world attack scenarios, not enough data
  - AI-based Response:
    - Too risky, needs absolute certainty, not enough real-world data to train on







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- Active Attack Detection
  - Steps:
    - Vehicle: Collect data from vehicle => Apply anomaly detection rules
    - Backend (Cloud): Aggregate data (fleet-wide) => More in-depth detection
  - Challenges:
    - What data? Data from ECUs, interfaces (e.g., Wifi, Bluetooth), V2V communications
    - How much data? Just enough to analyze the attacks and the infrastructure
  - Best practices:
    - Asset register (ECUs), asset use cases, review by service owners
- Response
  - Goals:
    - Contain or mitigate attacks => Stop incident => Recover => Lessons learned
  - Challenges:
    - Variety of attack models with different levels of intelligence and complexity
  - Best practices:
    - Safety-critical context/usage, context-specific fall-back, automated vs. manual response

### Low-cost Attack Recovery for RVs

Perception in RVs

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- Sensor attacks
- Can RVs continue to operate safely despite sensor attacks?
- State-of-the-art Attack Detection and Recovery
  - Detection: Invariant-based and model-based
  - Recovery: Fail-safe mechanisms (emergency landing)
- Attack Recovery without mission failure or crash
  - Prevent erroneous physical states AND prevent erroneous actuator signals
  - PID-Piper
    - **Problem:** PID overcompensation under attacks => good for faults, not for attacks
    - **Solution:** Redundant feed-forward controller (FFC)
  - DeLorean
    - **Problem:** Multiple sensors under attack
    - Solution: Identify attacked sensors, isolate them, substitute sequence, recover by replay





**R2:** Prevent erroneous

actuator signals

## Low-cost Attack Recovery for RVs

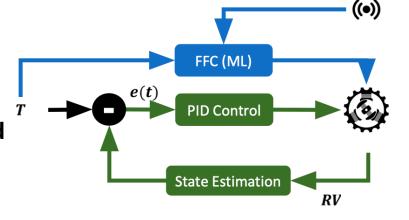
- Attack Recovery Methods
  - PID-Piper

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- Redundant feed-forward controller to address PID overcompensation
- ML trained on sensor and waypoint data to predict recovery actions
- Switched to upon attack detection and active for the attack duration
- Higher mission success, low false positives, negligible overhead

#### – DeLorean

- Detect the attacked sensors
- Prevent erroneous physical states: isolating sensor(s) from controller
- Prevent erroneous actuator signals: substituting input sequence
- Discard corrupted states and replay historic states
- First work to recover from multiple sensor attacks with little overhead



**Recovery Requirements** 

**R1: Prevent erroneous** 

physical states

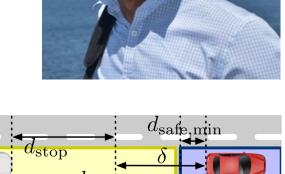


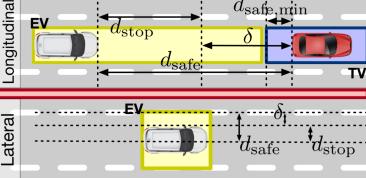
## **Design and Assessment for AV Safety**

Vulnerabilities in AVs

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- Much worse than non-AVs
- Increased attack surface: ML uncertainty, training data quality, unknown unknowns
- Identifying safety-critical vulnerabilities
  - Problem: State-space exploration to find the faults that lead to safety hazards
  - What/Where to inject faults?
    - Solution: Accelerate testing by only doing FI based on ML inference
      - Probabilistic Graph Models (PGMs) to model fault propagation
      - Training on observational data
      - Model fault injection as an inference query on PGM
  - What/How/When to launch attacks?
    - Solution: Design MI-driven attacks that can evade detection
      - Alter objects trajectories by corrupting pixels or perception output
      - ML inference of low safety potential and minimum time to hazards
  - Much faster and more efficient identification of safety-critical scenarios than random FI
- Runtime threat assessment for safety











#### Current Challenges

- IT to AV transfer of security and safety methods and best practices
- Lack of realistic incident data and labels for training detection and response models
- Effect of ML uncertainties and quality of training data
- Timing constraints, computational overhead, and side consequences of methods at runtime

#### Future Directions

- ML/AI driven models for fault injection, safety assessment, attack detection and recovery
- Combined model and data-driven methods, situationally-aware methods, both online and offline
- Simulation to real transfer of safety models, fault and driving scenarios, and datasets
- Community standards for quantifying the quality of ML models and datasets