

Unsupervised Learning Algorithms for Denial of Service Detection in Connected Vehicles

Connected Cars Context

- ▶ 237 millions connected cars in the world in 2021 and over 400 millions by 2025.
- ▶ Global ITS (**Intelligent Transportation System**) market from \$22.91 billion in 2021 to \$42.80 billion in 2028
- ▶ Cybersecurity of the connected cars is highly critical.
- ▶ **Denial of Service** (DoS) attacks jam and flood the network by mass of malicious messages.

Parties prenantes



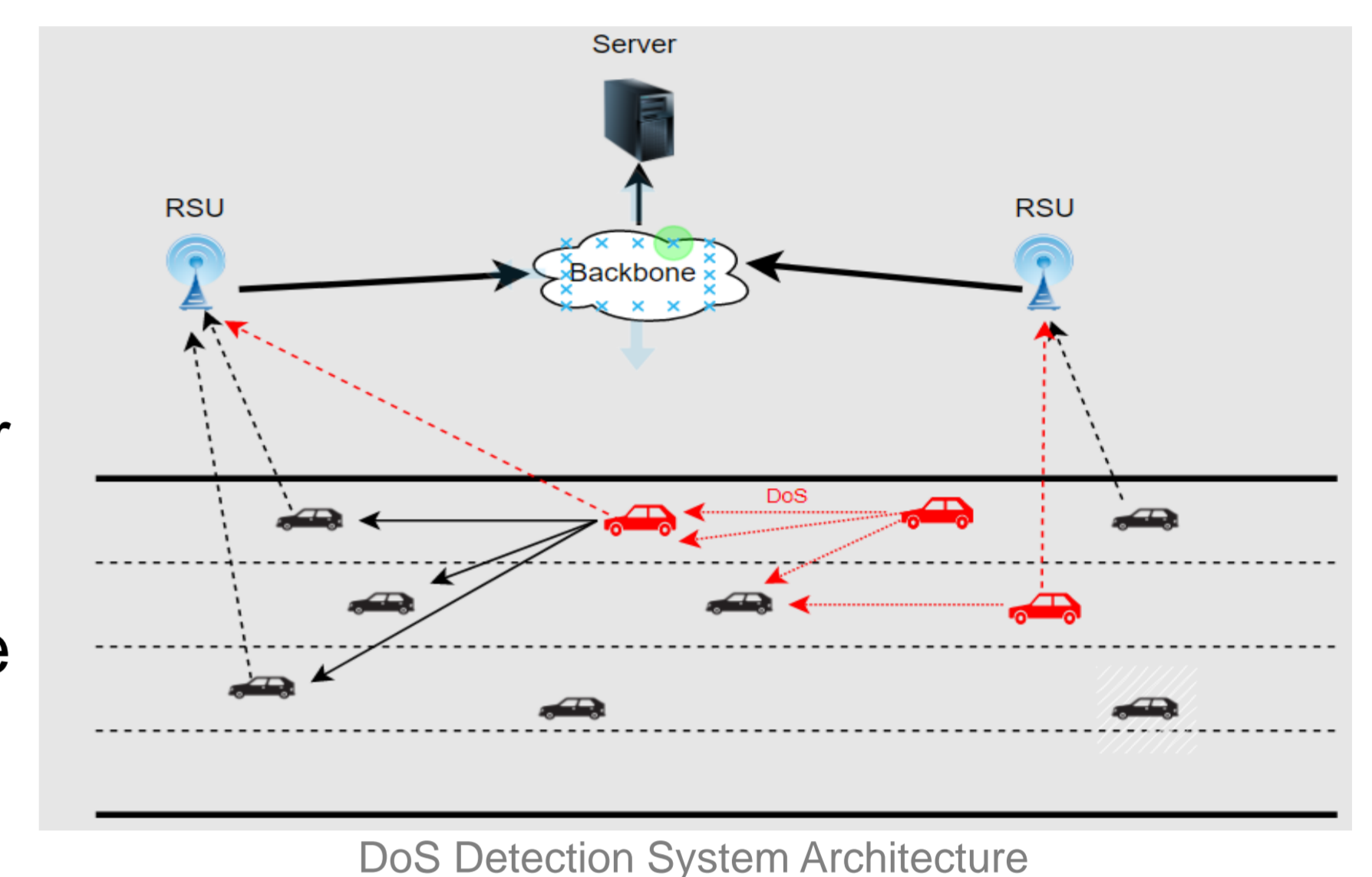
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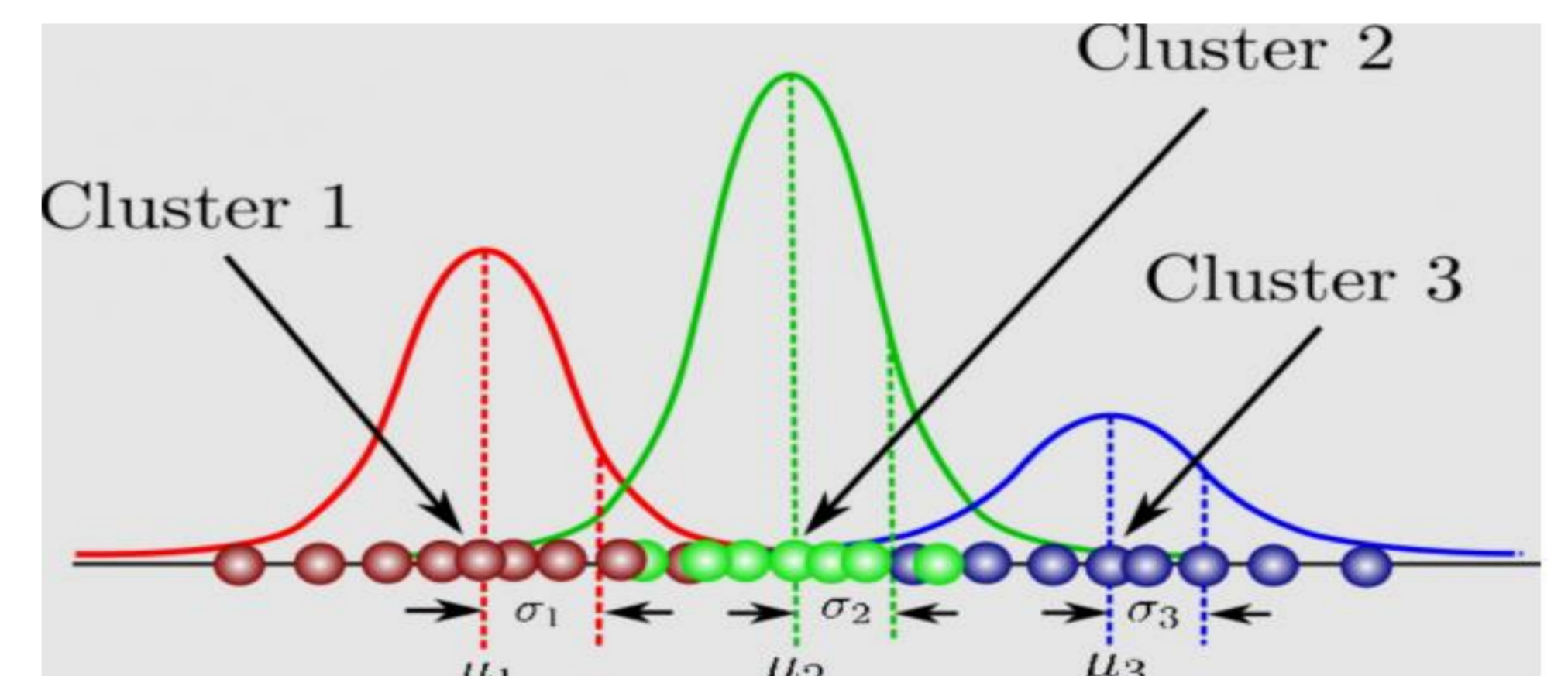
- ▶ Proposed DoS detection system architecture:
 - CAM DoS attack generated by one or more malicious vehicles.
 - CAM messages contain information (position, velocity, etc.).
 - CAM messages are sent to a central or distributed servers for analysis.
 - **Clustering-based** detection algorithms are applied to separate benign and malicious data.



DoS Clustering-based Model Detection

Several Clustering algorithms are applied:

- ▶ Centroid-based clustering (e.g. **K-Means, CLARA**)
- ▶ Density-Based clustering (e.g. **DBSCAN**)
- ▶ Probabilistic Model-based Clustering (e.g. **GMM**). EM (Expectation-Maximization) algorithm to estimate the parameters of GMM, it alternates between two steps:
 - **E step**: compute the log likelihood expectation using the current parameter estimation.
 - **M-step**: estimate of the parameters by maximizing the expected log likelihood found in the (E) step.



Example of clustering with a Gaussian Mixture Model of three components

Results and Analysis

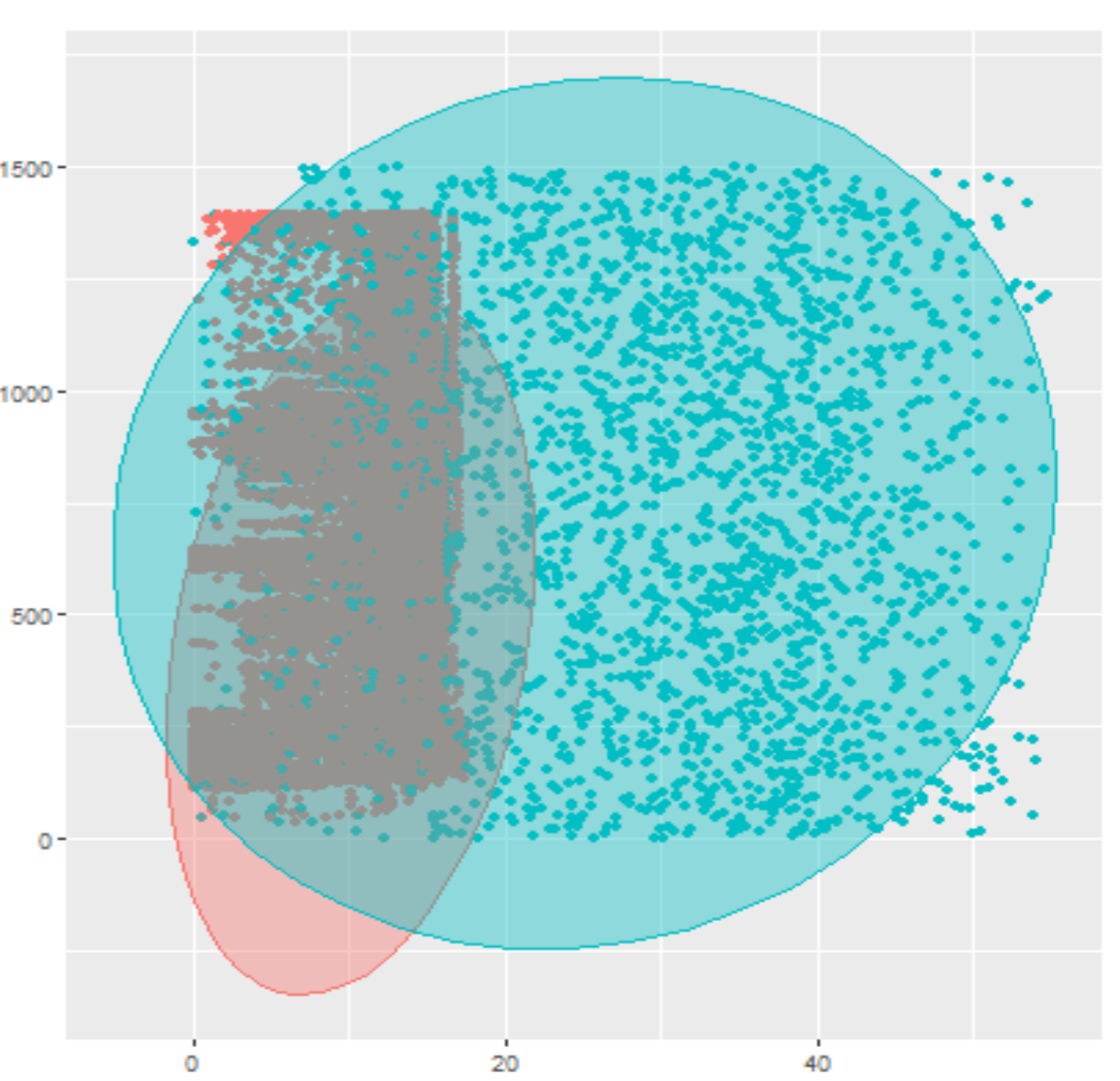
- ▶ Evaluation based on Vehicular Reference Misbehavior (**VeReMi**) dataset, containing the message logs for each vehicle in the simulation:
 - Position, velocity, acceleration and heading vectors.
 - Position noise, velocity noise, acceleration noise.
 - Reception timestamp, transmission time, sender, ID message, etc.
- ▶ GMM produces the best results in terms of **F-measure**. It is approximately 95% for the three subsets (**DoS (D1)**, **random DoS (D2)** and **disruptive DoS (D3)**), with high precision and recall values.

| Criterion | K-means | GMM | Clara | DBSCAN |
|-------------|---------|--------|--------|--------|
| Specificity | 0.6179 | 0.8103 | 0.5079 | 0.0647 |
| Sensitivity | 0.9856 | 0.9541 | 0.5049 | 0.8993 |
| Precision | 0.8823 | 0.936 | 0.7489 | 0.7365 |
| Recall | 0.9856 | 0.9541 | 0.5049 | 0.8993 |
| F-Measure | 0.9311 | 0.945 | 0.6031 | 0.8098 |

Detection Results on (B,D1,D2,D3)

| Criterion | K-means | GMM | Clara | DBSCAN |
|-------------|---------|--------|--------|--------|
| Specificity | 0.5084 | 0.5137 | 0.4965 | 0.0806 |
| Sensitivity | 0.5061 | 0.9714 | 0.5037 | 0.8961 |
| Precision | 0.8785 | 0.9334 | 0.8754 | 0.8725 |
| Recall | 0.5061 | 0.9714 | 0.5037 | 0.8961 |
| F-Measure | 0.6422 | 0.952 | 0.6394 | 0.8841 |

Detection Results on (B,D1,D3)



Results of GMM clustering (cluster 1:malware data, cluster 2: benign data)