

**Problem:** Autonomous vehicles are engineered for specific scenarios. The use in an open world may exhibit new and unforeseen characteristics. Field observations are required to **identify new scenarios at runtime**. This means to observe the actual traffic on actual road segments. The observation should identify if new characteristics are relevant within a certain scenario. Due to the high dimensional observation space, it is likely that many new combinations of characteristics are recorded. Test vehicles or of in-use vehicle fleets use limited storage and bandwidth of recordings so a **filtering of runtime records** is required to determine if a combination of characteristics maps to a known case or if new edge case has been found, which should be feedback to development and testing.

**Idea:** At the safety department of the Fraunhofer IESE, we employ the use of auto-encoders to **distinguish in between relevant and irrelevant new scenarios**. The unsupervised learning allows to implement engineering agnostic detection capabilities, which require no explicit feature engineering. The identification of cause-effect relations goes beyond a pure anomaly detection. It enables an **automated reasoning** if the observation of a new combination of characteristics maps to a known cause-effect or not.

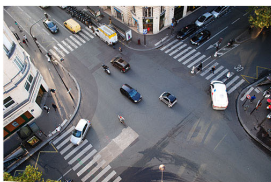
**Benefit:** The manufacturer of autonomous vehicles can fulfill its **product monitoring obligation** and automatically determine unexpected scenarios during the runtime of the autonomous vehicles using AI. Implementation takes place without feature engineering and requires no prior knowledge of the system architecture.

**Our solution:** The safety department of the Fraunhofer IESE developed the **tool AutoTestReduction to identify automatically new driving scenarios**. The system uses the available natural driving data of the development cycle as a reference and compares at runtime new driving scenarios against it. The learned representation incorporates for each scenario the set of characteristics that must, may or must not occur.

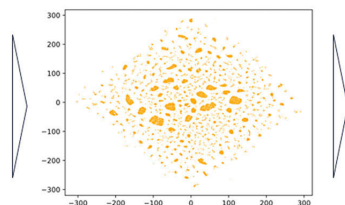
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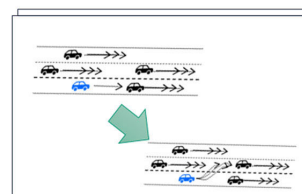
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Real-World recordings or simulator based recordings



Cluster recordings based on the learnt Auto-Encoder representation



Logical Scenarios with parameter ranges (one per cluster)

Clustering of traffic scenarios based on unsupervised learning with an auto-encoder. If new records are determined to be outliers and not to belonging to existing clusters, then a new relevant case has been identified.

We provide:

- The Tool **AutoTestReduction** to identify cause-effect relations based on natural driving data
- Support in the integration of **field observations** into the safety argumentation
- The integration of **dynamic risk management** into field operations