



TNO



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Adapt*!Ve*

*Automated Driving Applications and
Technologies for Intelligent Vehicles*

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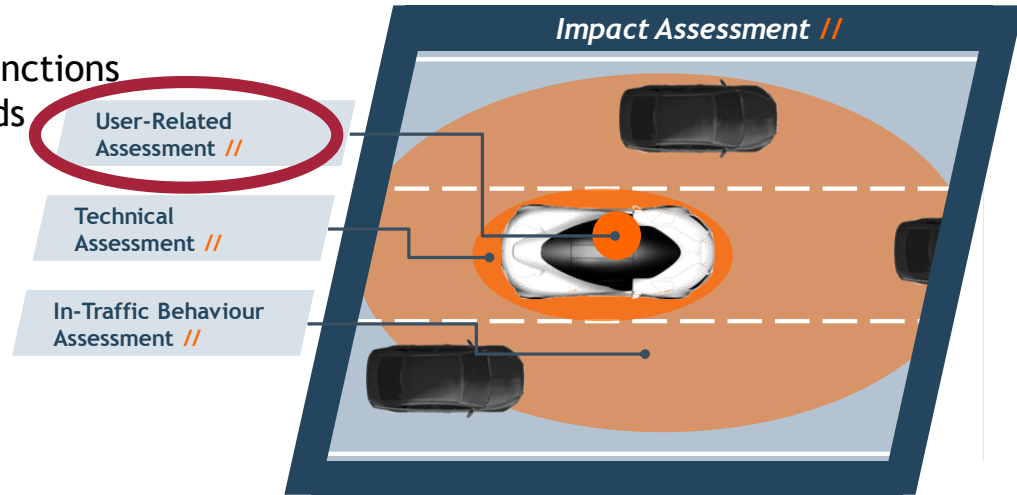
Evaluation of the Adapt*!Ve* functions
User-related assessment and
In-traffic behavior assessment

Final Event
Aachen, Germany
29 June 2017

// Evaluation of AdaptIVe functions

AdaptIVe Subproject „Evaluation“

- Main objectives:
 - Development of an evaluation framework for automated driving systems
 - Methodology for impact analysis of automated driving systems
- Detailed objectives:
 - Apply developed methods on selected functions in order to verify the evaluation methods
 - Benefit analysis with focus on safety and environmental impact
 - > Derive first recommendations and results on the impact of automated driving applications



// User-related Assessment - Evaluation tools and topics

“Highway Automation”

Real-life driving with/without

Driver behaviour

Workload

Understanding the system

Trust

Usability

Opinions about HMI

Experienced effects

Expected benefits

Willingness to pay

“Urban Automation”

Driving on test track

-

-

Understanding the system

Trust

Usability

Opinions about HMI

-

Expected benefits

Willingness to pay

//Methods

Behavioural observations - two observers in the car (“Highway Automation”)

Logging of driving data - speed, distance, lane keeping (“Highway Automation”)

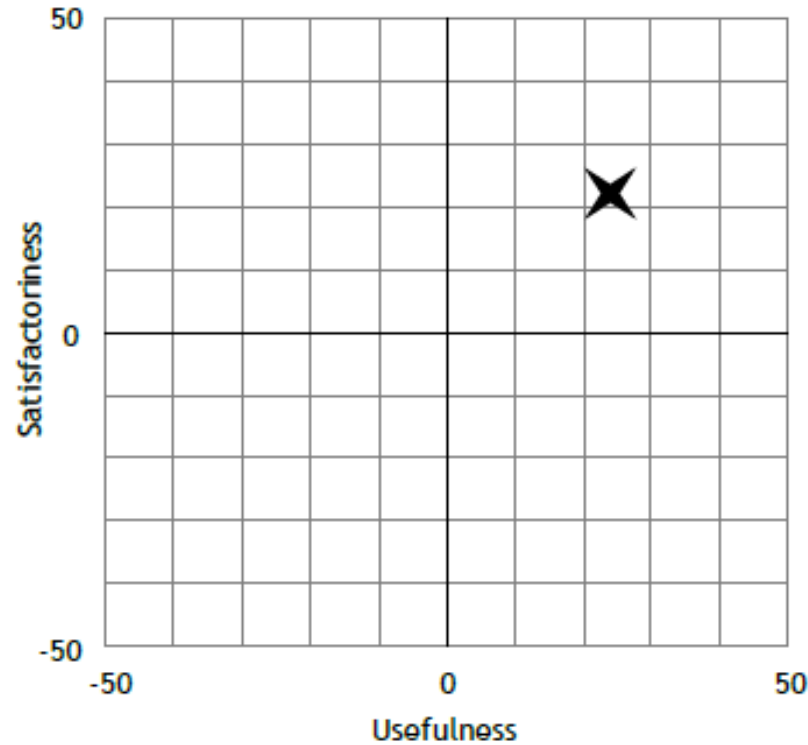
Questionnaires (both “Highway” and “Urban” Automation)

- **Mental workload**
- **Trust**
- **Usability**
- **Usefulness/Satisfactoriness**
- **Experienced effects**
- **Expected benefits/disadvantages**
- **Opinions about the HMI**
- **Willingness to pay**

// “Urban automation” - Driver experiences and opinions

- Most participants found the system easy to learn and use.
- High System Usability Scale (SUS) score: 80 (on a scale 0-100).
- The participants were not fully aware of the system’s limitations.
- The majority would be willing to pay between 1,000 and 4,000 Euros.
- There were clear expectations in decreased fuel consumption and increased driving comfort among the respondents.
- Some worries expressed:
 - “does the car constantly handle new and different situations consistently in real traffic with a lot of drivers around who cannot drive a car and do a lot of stupid things”?
 - “driving pleasure disappears with automated driving”.

// “Urban automation” - Usefulness and Satisfactoriness

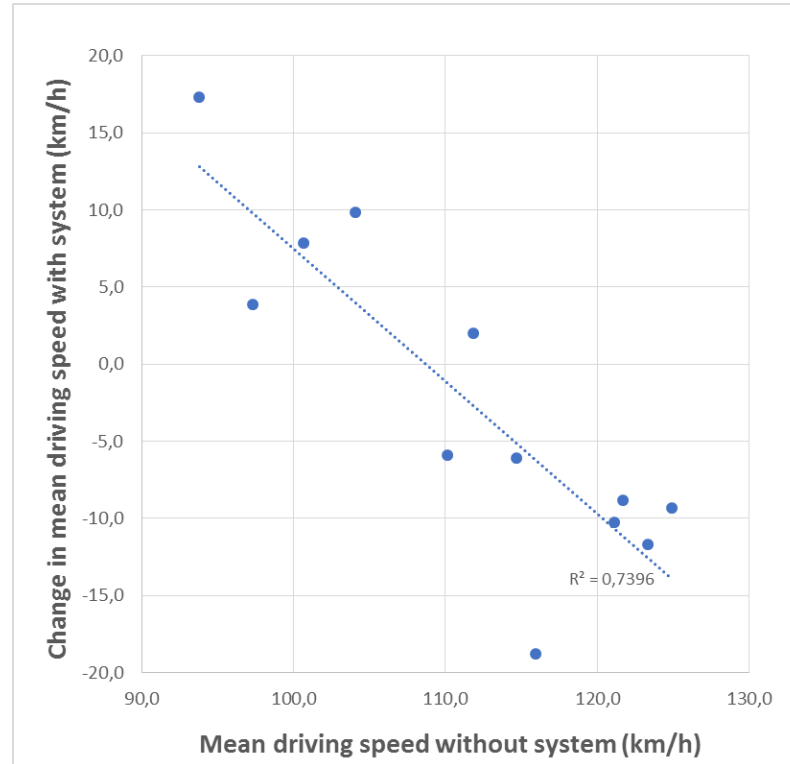


// “Highway automation” - Driving behaviour I

- The drivers used the system as it was intended to be used.
- The system affected driving positively in several ways:
 - + Better speed adaptation to speed limit and conditions, less speed variation
 - + Better distance keeping ahead
 - + Better lane choice (prescribed use of the right lane)
 - + Better indicator usage
 - + Fewer dangerous lane changes

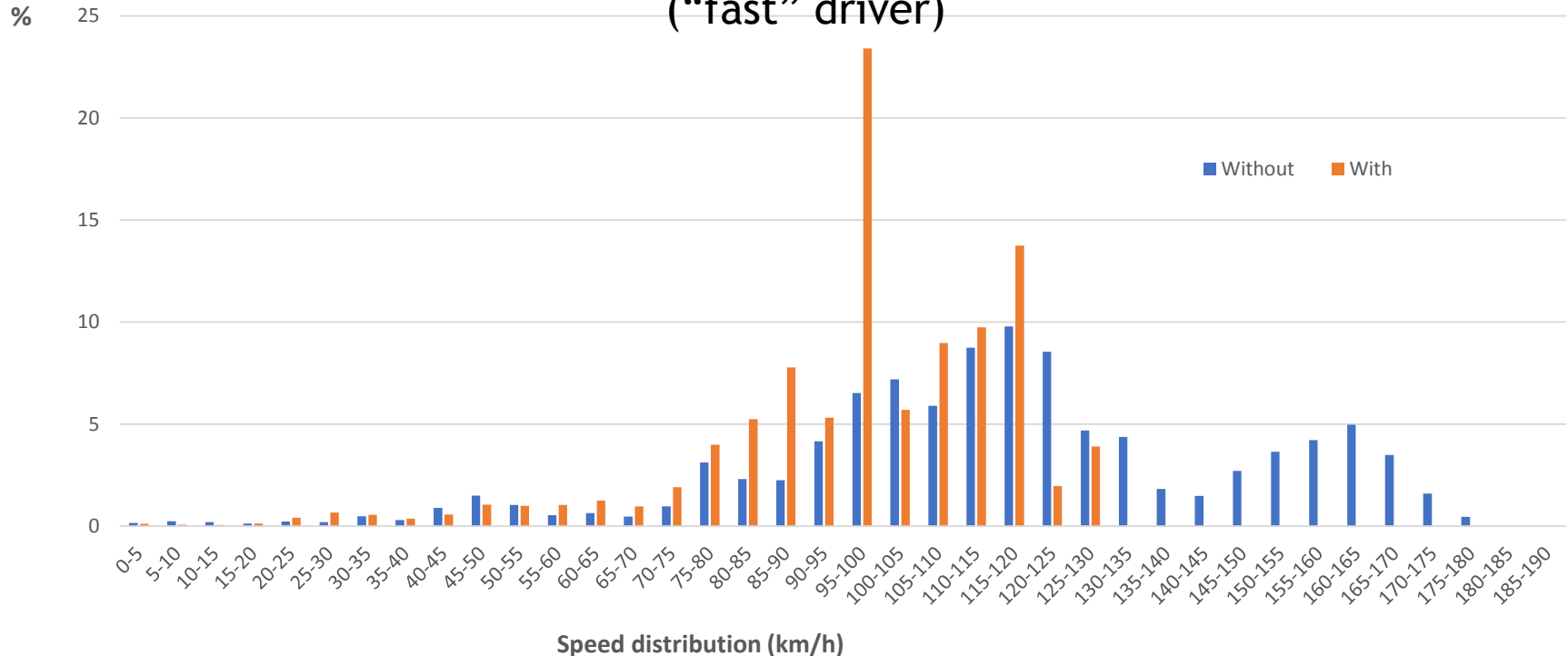
// “Highway automation” - Driving behaviour II

Change in mean driving speed versus mean driving speed without the system



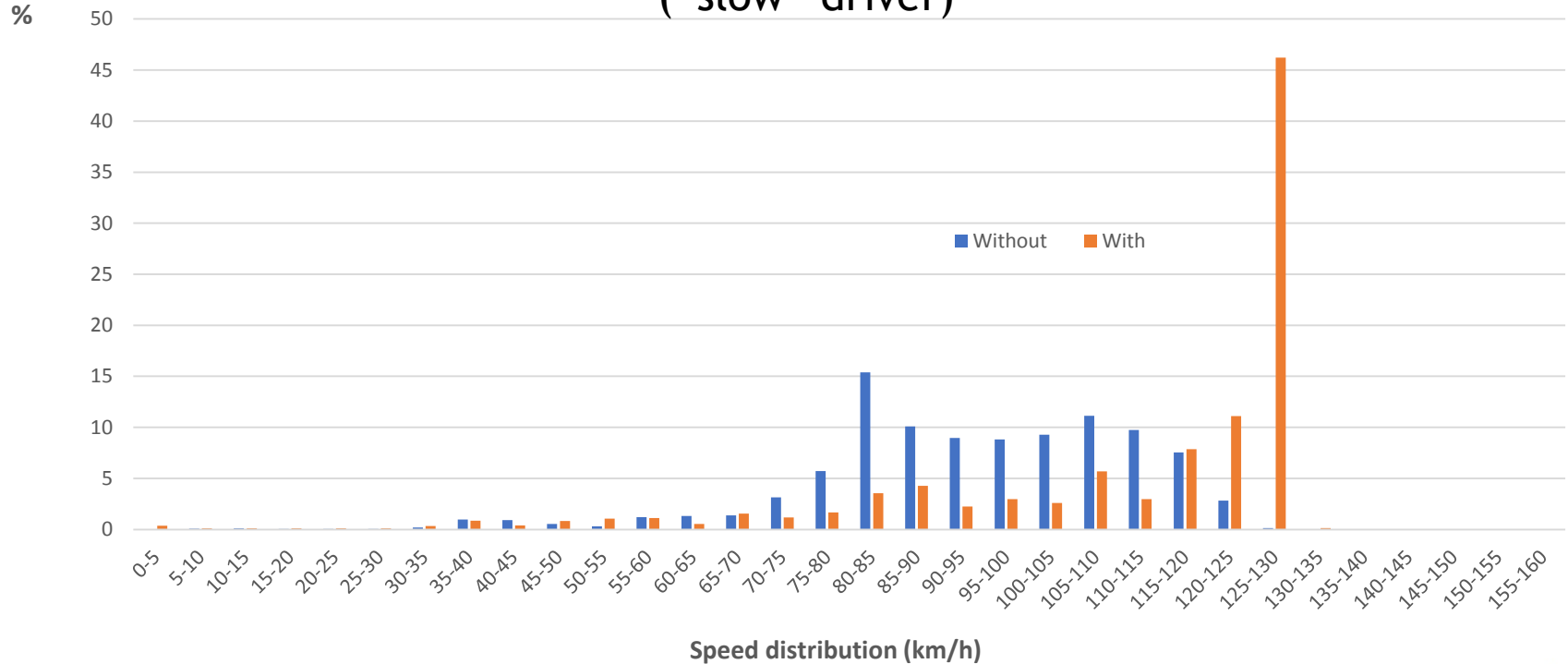
// “Highway automation” - Driving behaviour III

Distribution of driving speed when driving with- and without the system
 (“fast” driver)



// “Highway automation” - Driving behaviour IV

Distribution of driving speed when driving with- and without the system
 (“slow” driver)



// “Highway automation” - Driving behaviour V

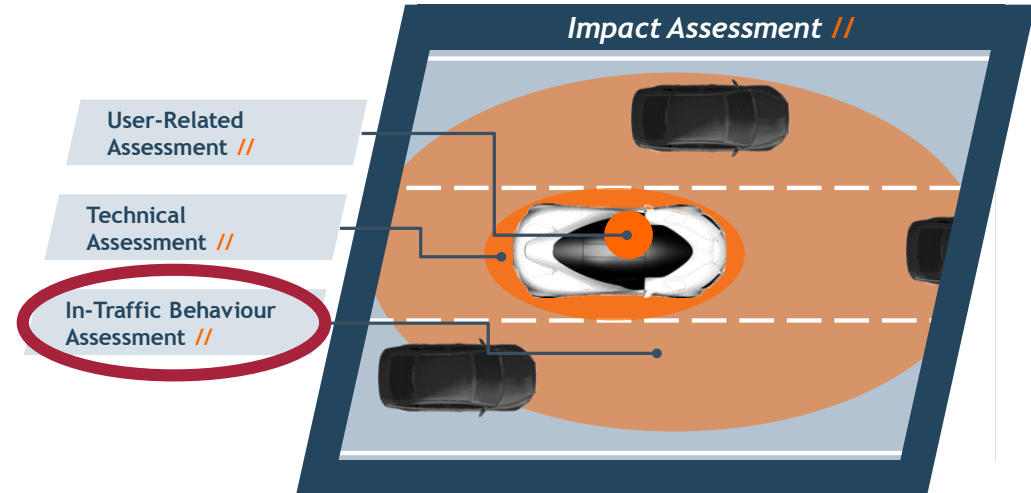
- Negative effects:
 - Not letting other drivers to make a lane change into own lane
 - Longer overtaking due to 130 kph system limit, hindering cars from behind
 - More conflicts due to losing the road markings due the reflection of the sun
 - Sudden braking manoeuvres due to not correctly recognising the surroundings

// “Highway automation” - Driver experiences

- Positive driver experiences:
 - + Driving comfort
 - + Trust
 - + Usability - High System Usability Scale (SUS) score: 82 (on a scale 0-100)
 - + Usefulness and Satisfactoriness
- No differences with regard to subjective workload
- Negative driver experiences:
 - Self-assessed driving performance decreased

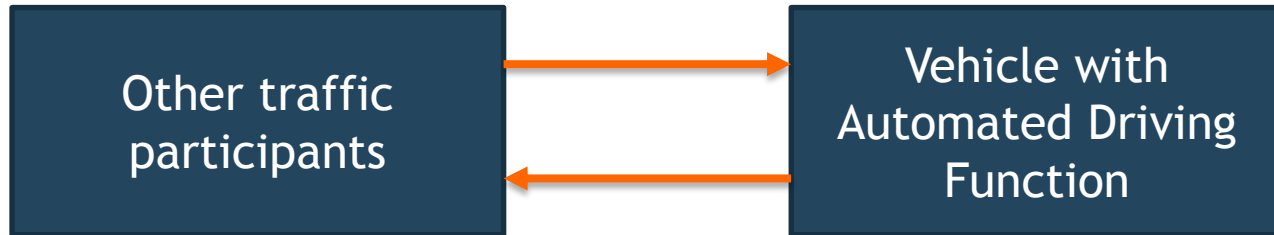
// In-traffic assessment

- What is in-traffic assessment
- General framework
- Method & Example
- Conclusion



// In-traffic Assessment

- How does the vehicle interact with other traffic participants?
- How do other traffic participants react on the (automated) vehicle?



// Solution proposal

- Just go on the road and see what happens.



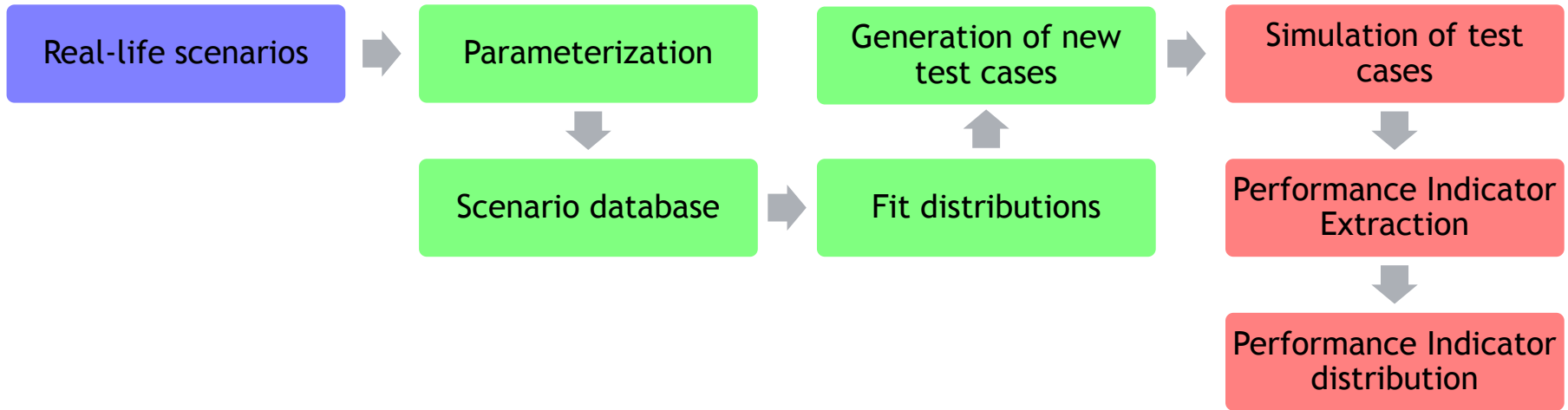
// Method - general framework

- Virtual testing
- Scenarios that resemble real-life traffic

Real-life data

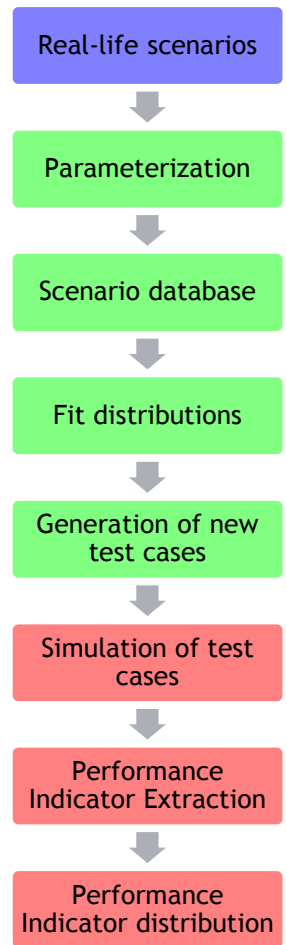
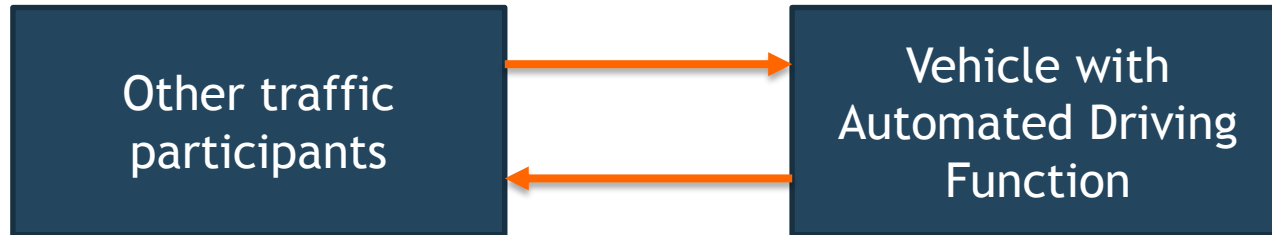
Test case generation

Simulation & Evaluation



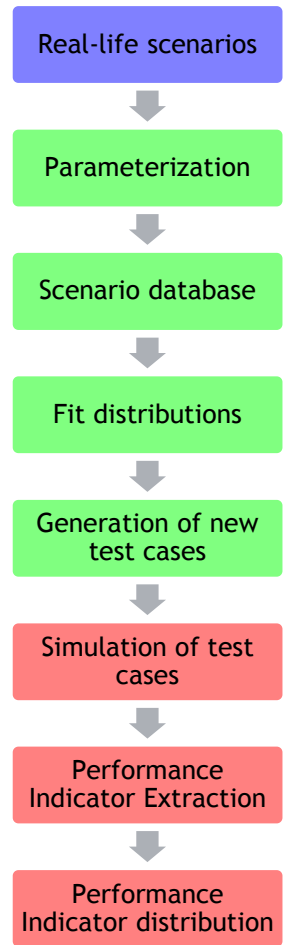
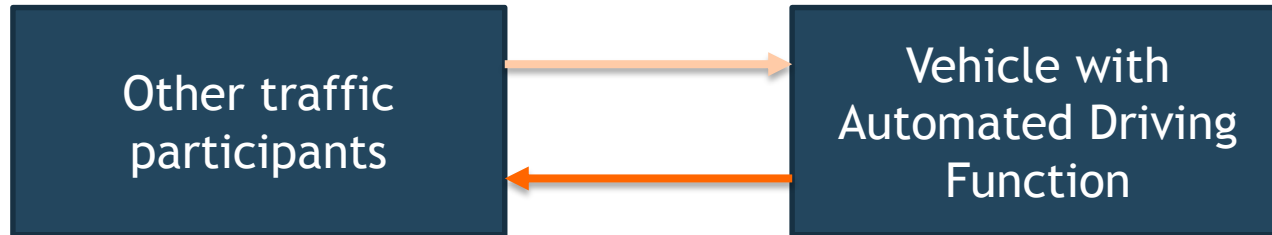
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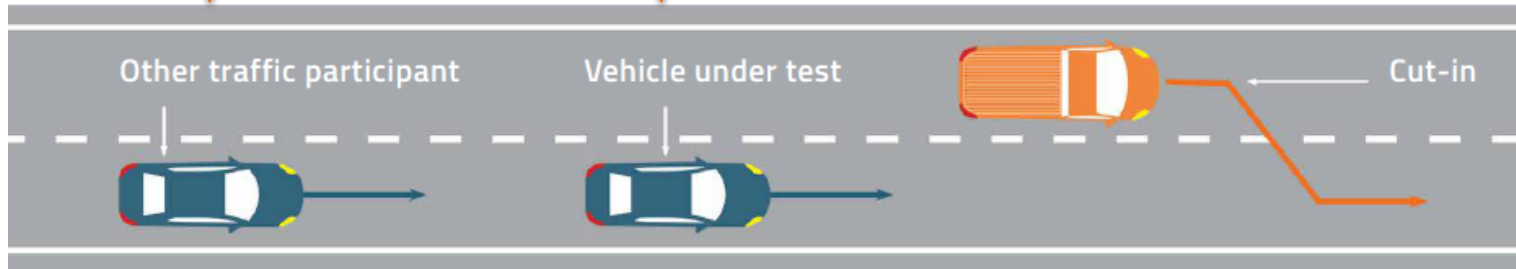


// Real-life scenarios

- Compare performance of third vehicle in two different configurations.

Intelligent Driver Model (IDM) [1]

IDM or Traffic Jam Assist (TJA)



Real-life scenarios

Parameterization

Scenario database

Fit distributions

Generation of new test cases

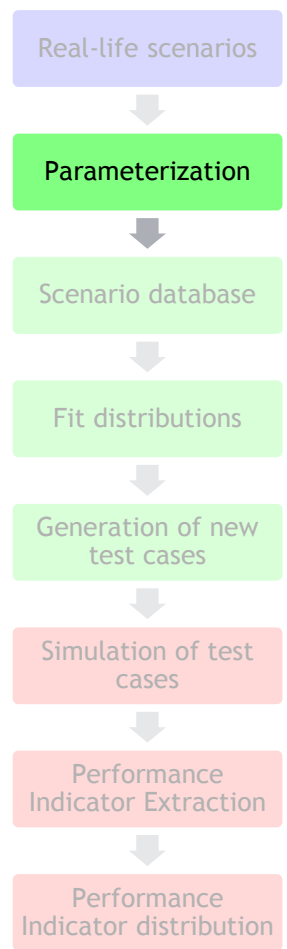
Simulation of test cases

Performance Indicator Extraction

Performance Indicator distribution

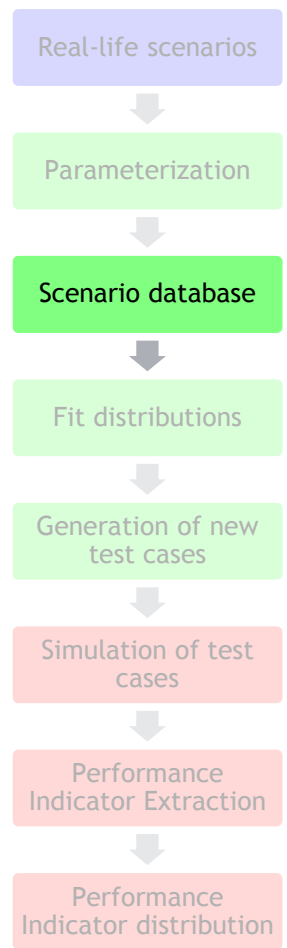
// Test case generation

- Summarizes scenario in only a few parameters.
- Why?
 - Probabilistic results
 - No need to ‘drive’ all kilometres to make claims!
 - Emphasize critical scenarios
 - Without *a-priori* knowledge of what might be critical
 - Prevent repetition
- Cut-in scenario → 5 parameters.
- Some assumptions, e.g. constant velocity



// Test case generation

- Store parameters in database.
- No need to store all data of a scenario.



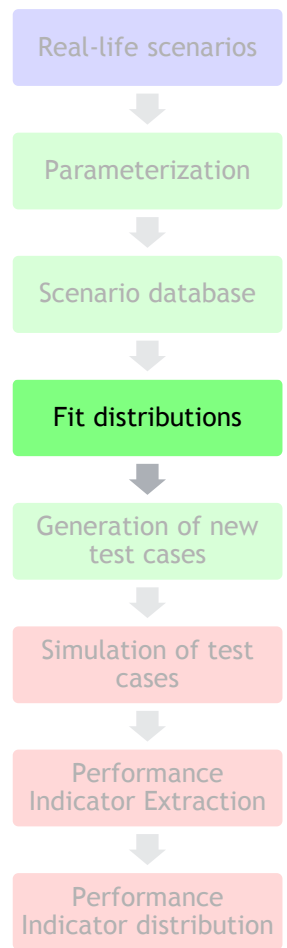
// Test case generation

- Kernel Density Estimation [2], [3]:

$$f_h(x) = \frac{1}{nh} \sum_{i=1}^n K\left(\frac{x - x_i}{h}\right)$$

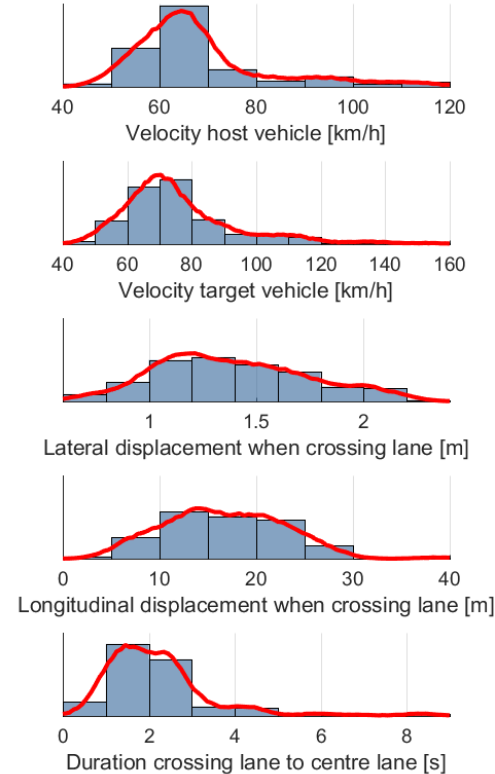
- Bandwidth $h \rightarrow$ cross-validation
 - Let the data speak for itself!

- ✓ No assumptions
- ✓ Multivariate data
- ✓ Easy to draw random samples



// Test case generation

- Histogram: original data
- Red lines: Kernel Density Estimation



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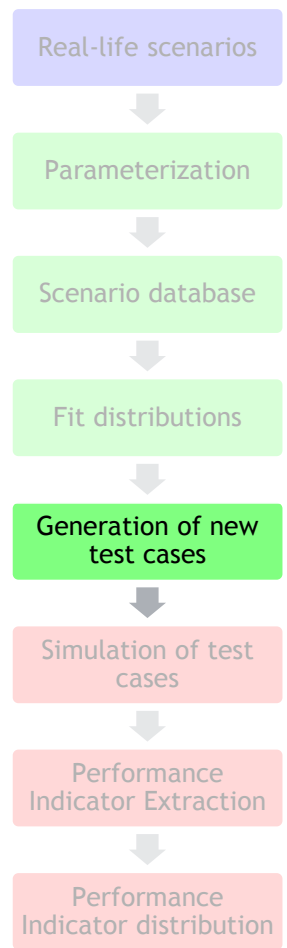
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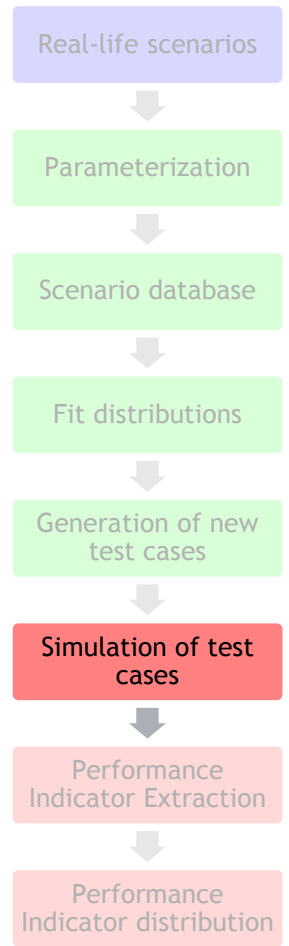
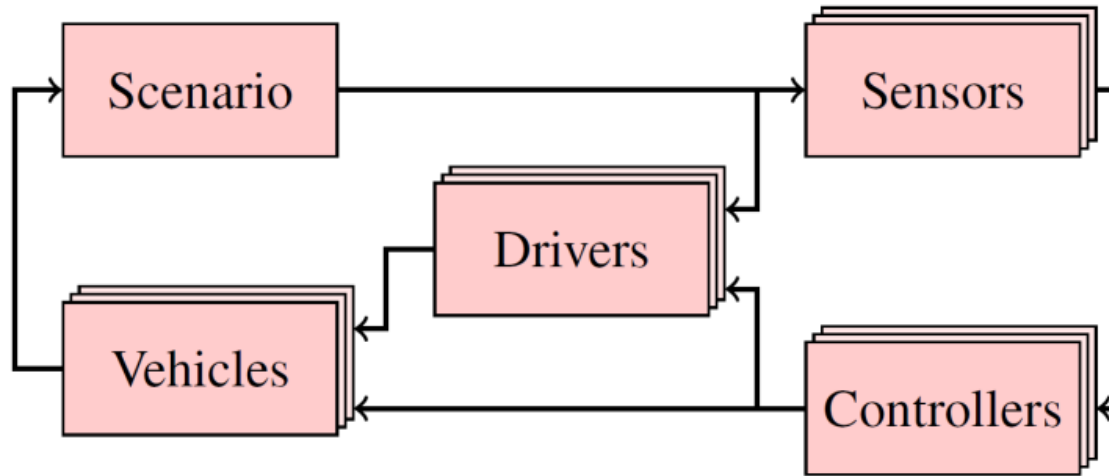
Performance Indicator distribution

// Test case generation

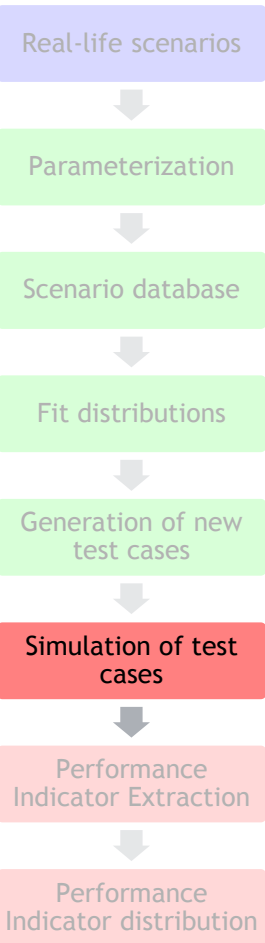
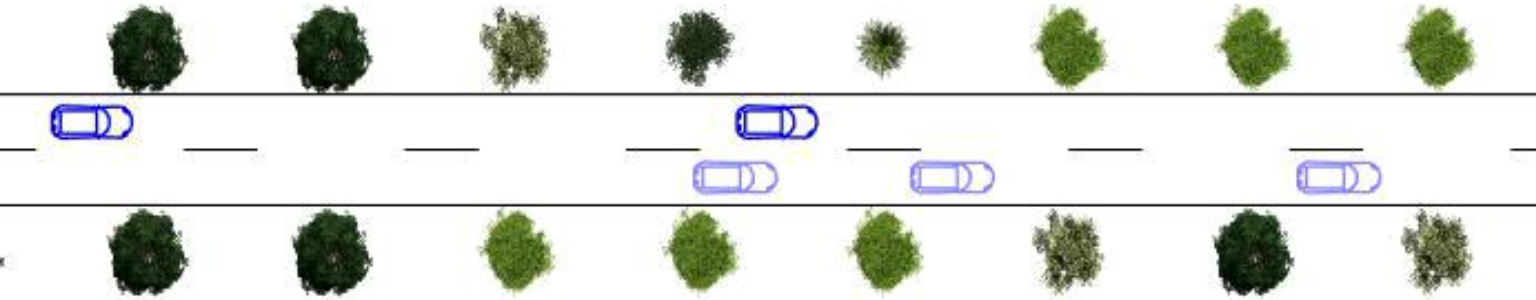
- Generation of new test cases:
 - Draw sample
 - Transform to real-life test case
- Importance sampling → emphasize performance-critical scenarios
 - Ask me for more details
 - See also *Assessment of Automated Driving Systems using real-life scenarios*, de Gelder, E. and Paardekooper, J.-P. (2017)



// Method - simulation and evaluation

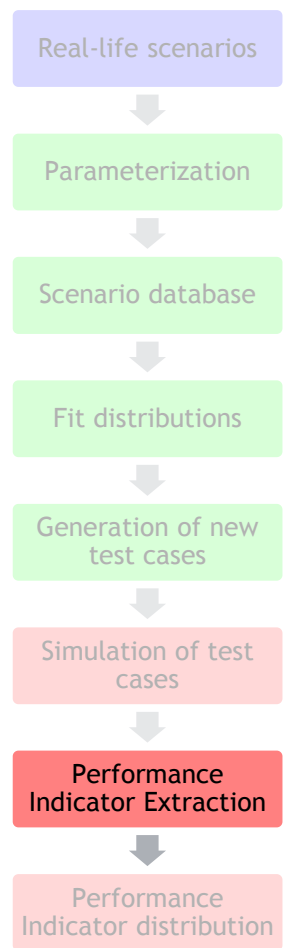


// Example - simulation and evaluation



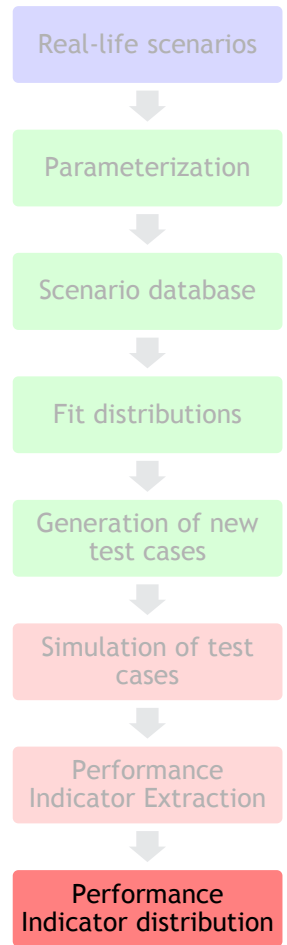
// Method - simulation and evaluation

- Performance Indicators are extracted from a simulation, e.g.
 - Time Headway (THW)
 - Time-To-Collision (TTC)
 - Distance
 - Velocity
 - Acceleration
 - etc.



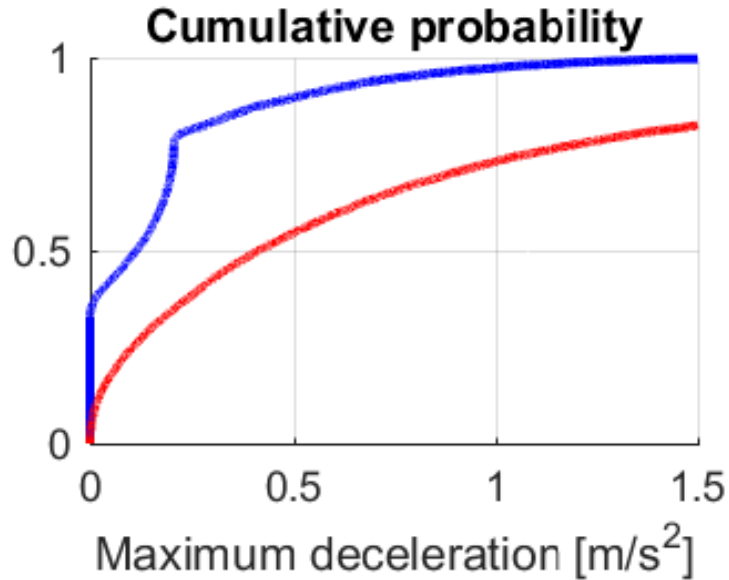
// Method - simulation and evaluation

- When a large number of simulations are performed, we can make distributions of the resulting Performance Indicators.



// Example - simulation and evaluation

- Red: both following vehicles are human driven (IDM)
- Blue: second car equipped with Traffic Jam Assist



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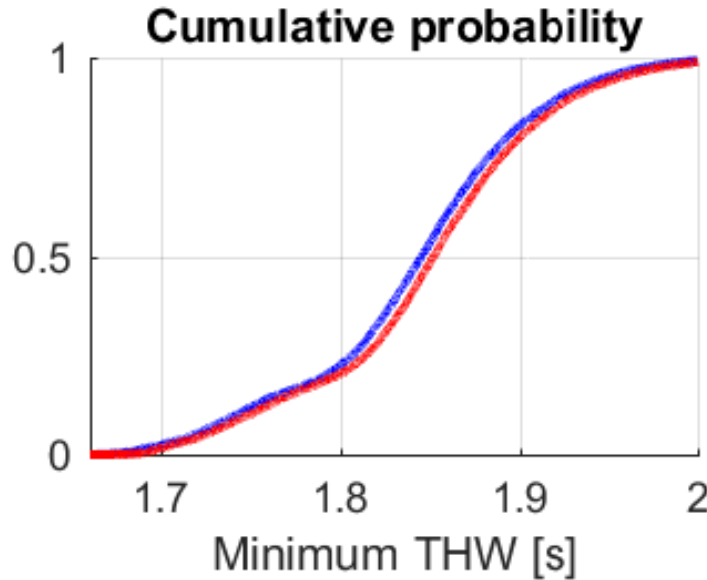
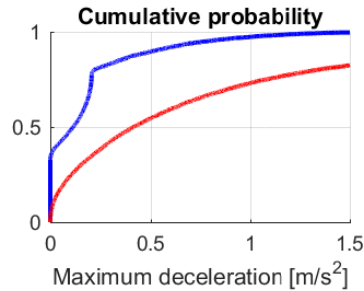
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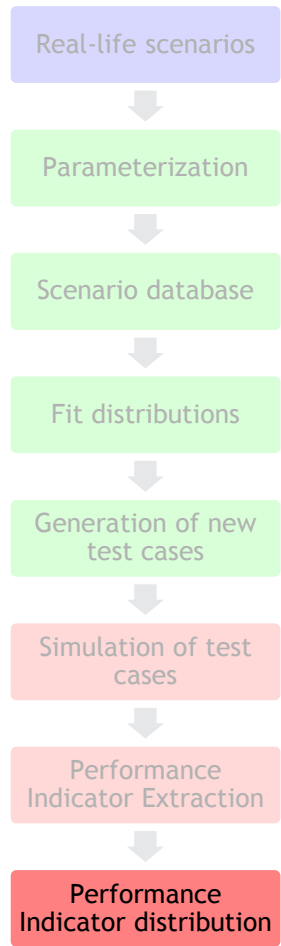
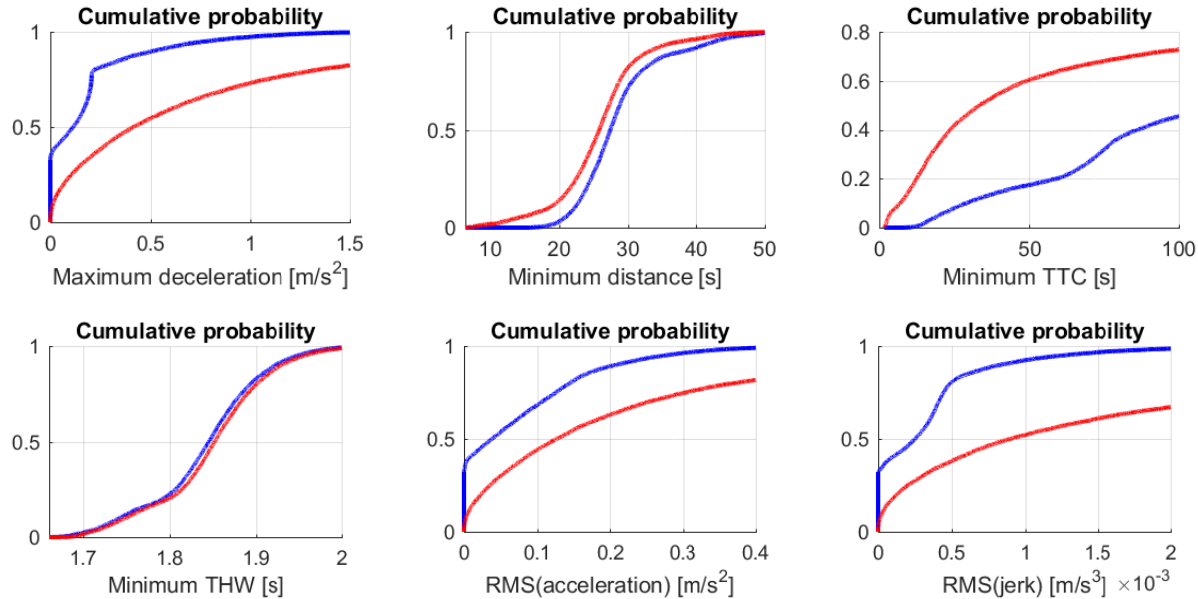
Performance
Indicator Extraction



Performance
Indicator distribution

// Example - simulation and evaluation

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// Conclusion

- Methodology to assess the performance of an Automated Driving Function has been developed.
- Through parameterization of the real-life scenarios, test cases are generated.
- The framework is mostly data driven.
- It provides quantitative results on
 - how a system will perform in real-life traffic and
 - how other traffic participants will react on the system.
- More information → check our stand



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*Automated Driving Applications and
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Thank you.

