

Detection of critical weather situations in scenario-based traffic simulations using optimization techniques

Daniel Grujic, Thies de Graaff, Eike Möhlmann, Günter Ehmen

Motivation

The performance of a neural network trained for **perception** is highly dependent on the training data set. This can lead to a weak performance of the network under specific circumstances if these circumstances occur only very rarely (or are not present) in the training data set.

Use case: Weather

We are looking at weather conditions. As weather is highly diverse, it is usually impossible to train a neural network for all possible weather conditions, which can affect the performance of the network. This is shown in Figure 1, where we used a specific traffic scenario and initialized it with different weather parameter values. For each set of weather parameter values we started a simulation run and measured the performance of the network using the mean IoU (intersection-over-union) metric.

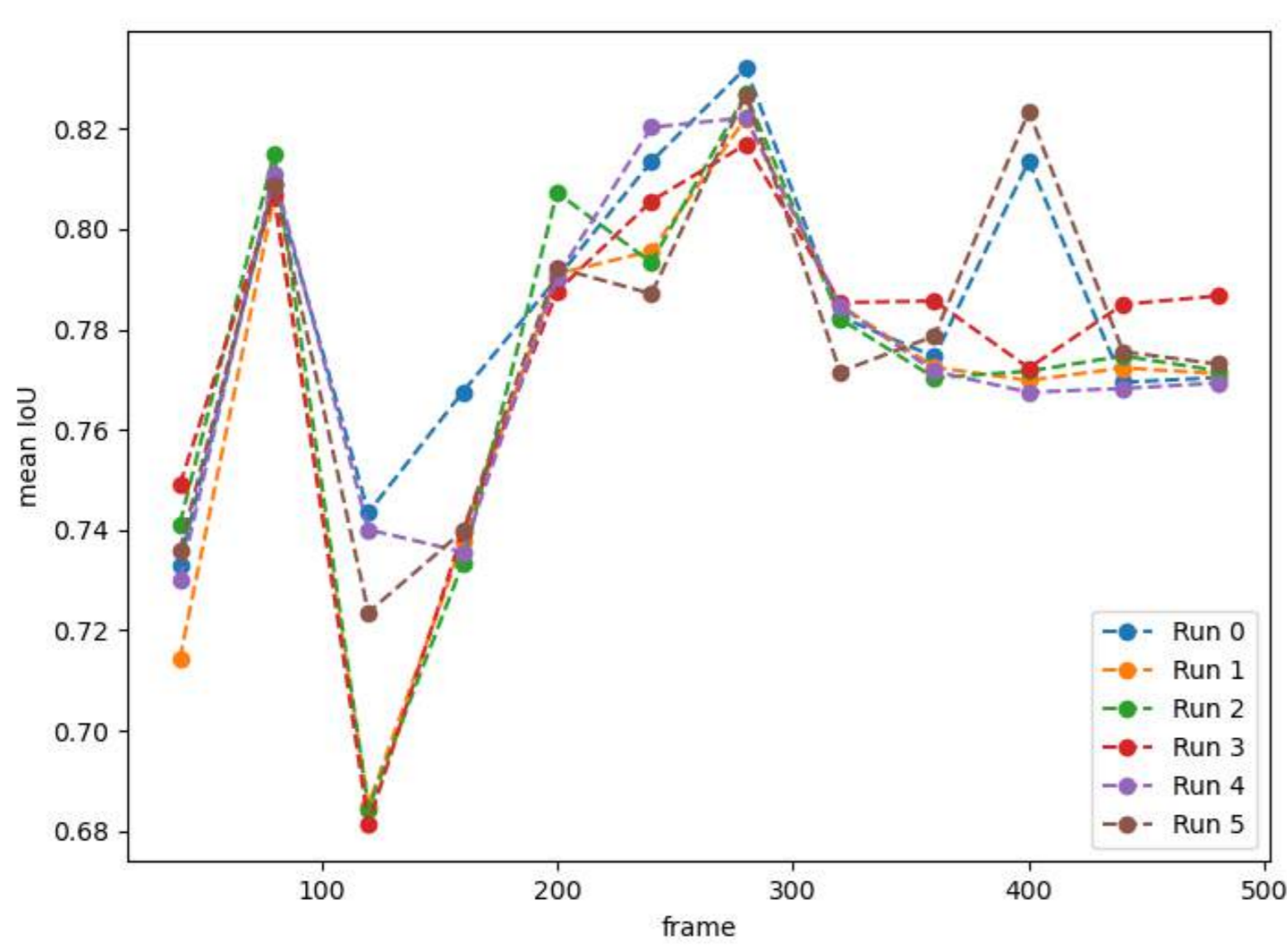


Figure 1: Depiction of six different simulation runs for a single scenario. A different weather situation was created for each simulation run, while everything else remained the same. The performance of the network was measured using the mean IoU (intersection-over-union) metric. (© Deutsches Zentrum für Luft- und Raumfahrt e.V.)

Task

We want to identify the most critical simulation parameters automatically and efficiently.



Figure 2: Depiction of the same scene in the CARLA traffic simulator for different weather parameters. (© Deutsches Zentrum für Luft- und Raumfahrt e.V.)

Method

We use an optimization loop to steer the simulation towards (weather-) parameters that can lead to a critical situation. A visualization of this loop can be found in Figure 3. The steps are as follows:

- 1) Create an initial parameter set
- 2) Initialize the traffic scenario with the given parameter set
- 3) Simulate the scenario and retrieve the results
- 4) Use the neural network to create a semantic segmentation mask
- 5) Calculate a performance indicator (called criticality) for the current simulation run
- 6) Use optimization algorithms to create a new parameter set based on the criticality feedback
- 7) Go to (2).

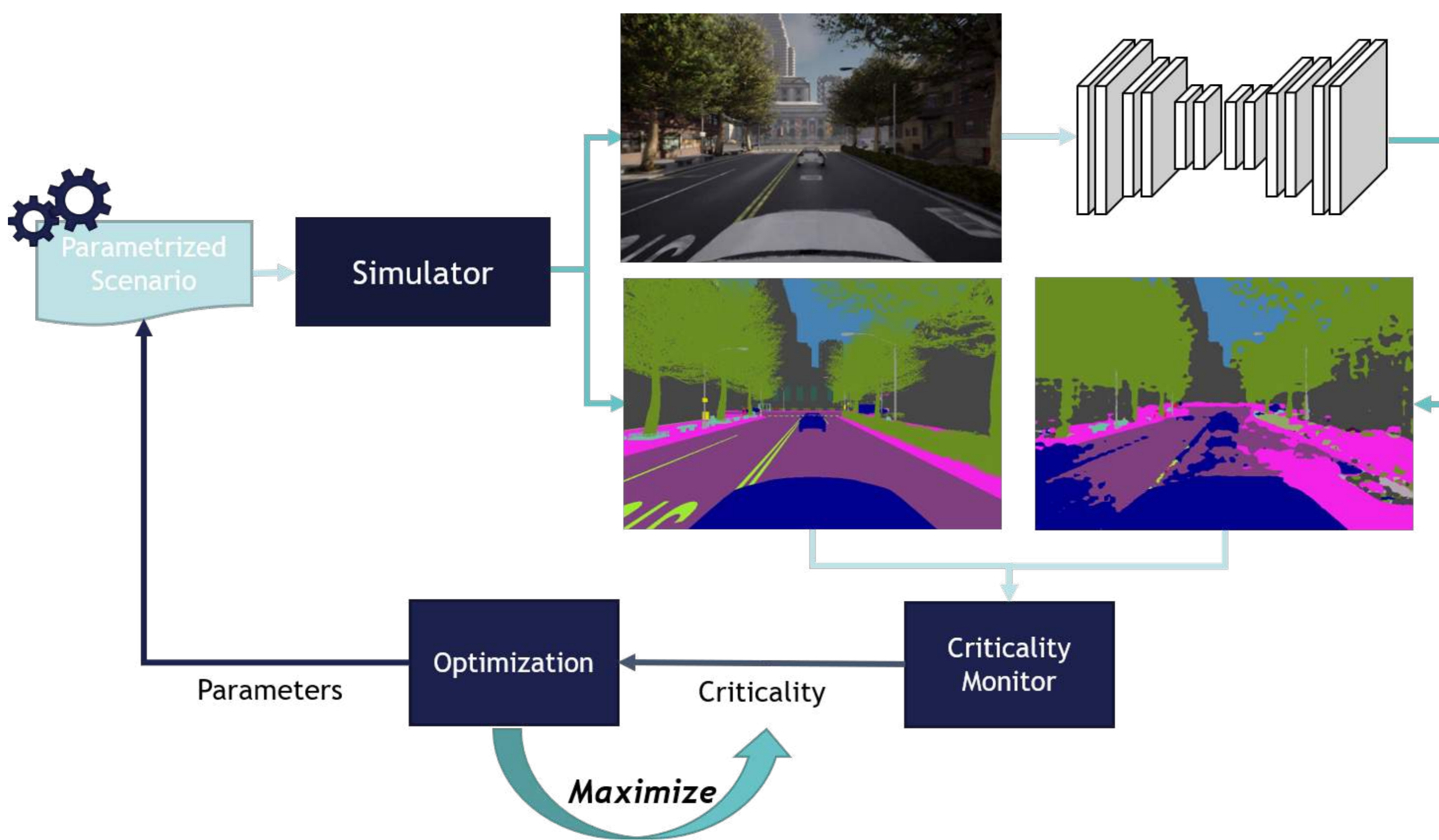


Figure 2: Visualization of the scenario-based optimization workflow. (© Deutsches Zentrum für Luft- und Raumfahrt e.V.)

Partners



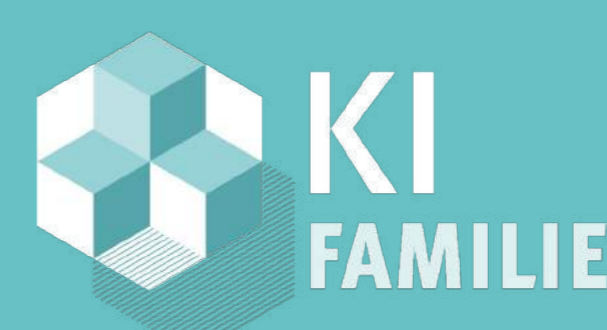
External partners



For more information contact:

daniel.grujic@dlr.de
thies.degraaff@dlr.de

KI Delta Learning is a project of the KI Familie. It was initiated and developed by the VDA Leitinitiative autonomous and connected driving and is funded by the Federal Ministry for Economic Affairs and Climate Action.



Supported by:

