KI ABSICHERUNG: SAFE AI FOR AUTOMATED DRIVING

Dr. Sebastian Houben | Fraunhofer Institute for Intelligent Analysis and Information Systems | The Connected Car and Autonomous Driving, October 26th, 2020

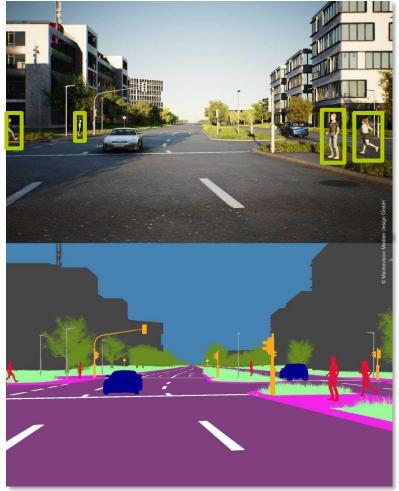




A Modern Application – Autonomous Driving AI-Based Pedestrian Detection

Setup

- An autonomously operating vehicle
- ... is crossing an intersection
- Al functionality for detecting pedestrians
 - Camera images processed by CNN
 - Output
 - Segmentation mask
 - Bounding box detections

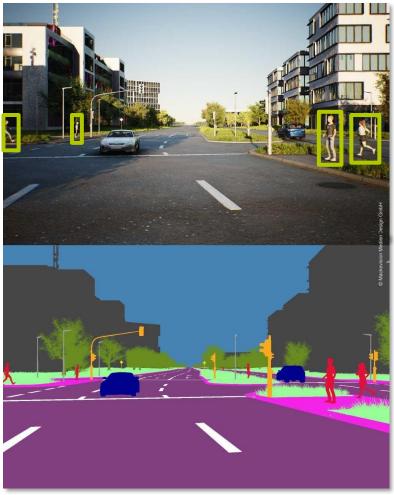


Synthetically generated intersection and corresponding semantic segmentation Project KI Absicherung - https://www.ki-absicherung.vdali.de



Can I Use a State-of-the-Art ML Model in an Automated Vehicle?

- Well, ML models (safety concerns)
 - don't work well on unseen data
 - are not robust to domain changes
 - may overfit to irrelevant correlations
 - are overconfident in their predictions

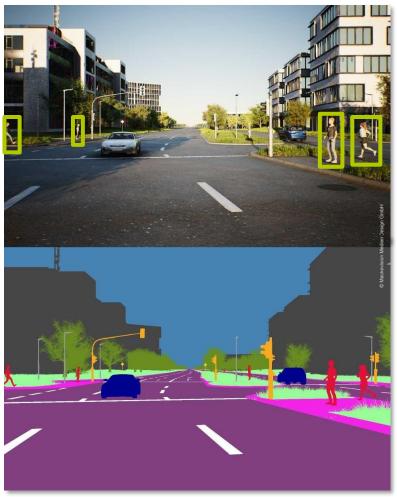


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Can I Use a State-of-the-Art ML Model in an Automated Vehicle?

- What can be done?
 - There are many attempts / research directions to alleviate these concerns.
 - One of our contributions is to investigate some of these methods more deeply.
 - Another one is to evaluate them w.r.t. to the saftey concerns
 - and find a plausible argumentation that they are circumvented or kept at bay.



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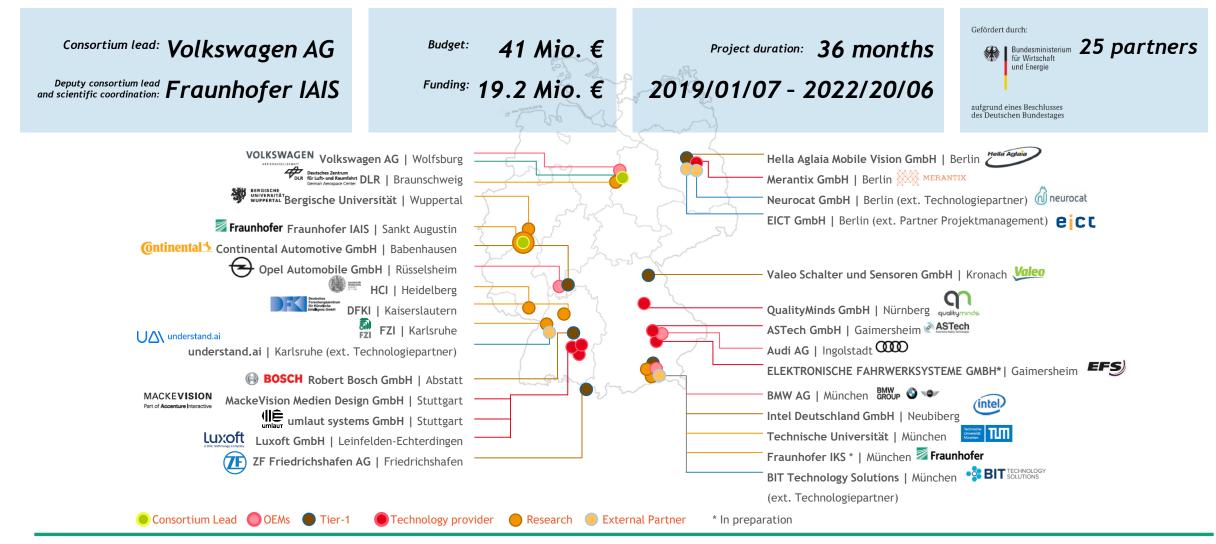




KI Absicherung is making the safety of AI-based function modules for highly automated driving verifiable.



The Project "KI Absicherung – Safe AI for Automated Driving"





KI Absicherung Main Goals

1. Methods for training and testing of Al-based functions

KI Absicherung develops and investigates means and methods for verifying AI-based functions for highly automated driving.

2. Safety argumentation

For the pedestrian detection use case, the project is developing an exemplary safety argumentation and methods for verifying a complex AI function.

3. Communication with standardization bodies on AI certification

The project's results will be used in the exchange with standardization bodies to support the development of a standard for safeguarding AI-based function modules.



KI Absicherung Main Goals

Today's Focus

1. Methods for training and testing of AI-based functions

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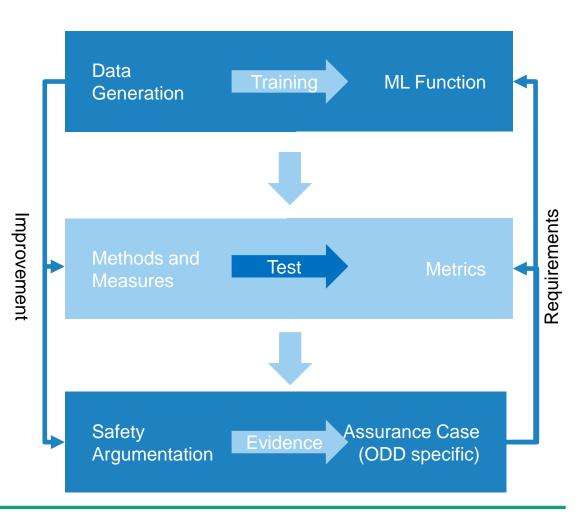
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From a Data-Driven AI Function to an Assurance Case Use Case: Pedestrian Detection

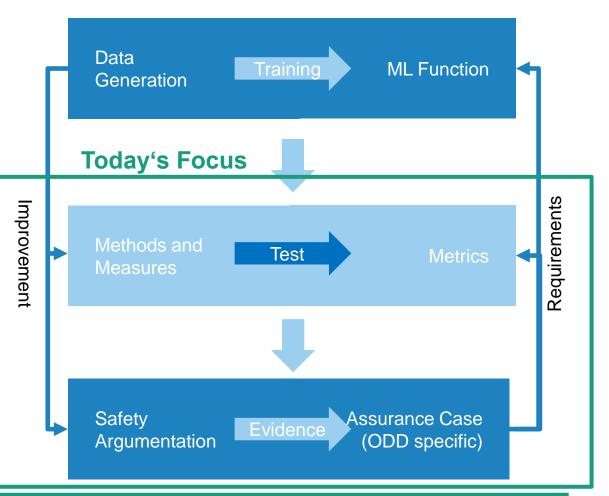
- Process-related generation of synthetic learning, testing and validation data.
- Development of measures and methods that improve the AI function over a wide array of metrics.
- Development and validation of testing methods for these metrics.
- Stringent argumentation for the AI function and its Operational Design Domain (ODD).





From a Data-Driven AI Function to an Assurance Case Use Case: Pedestrian Detection

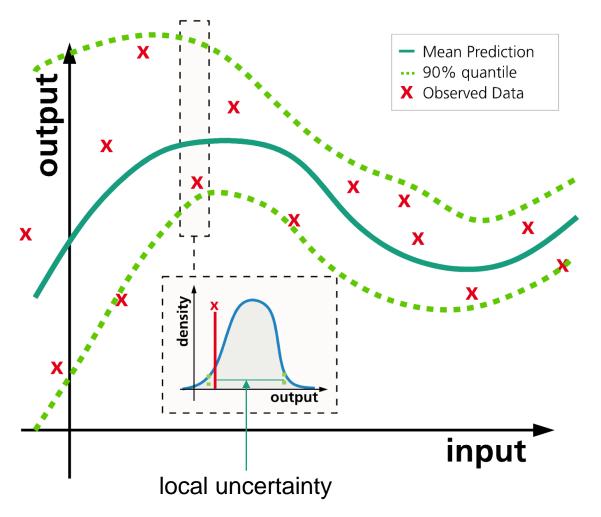
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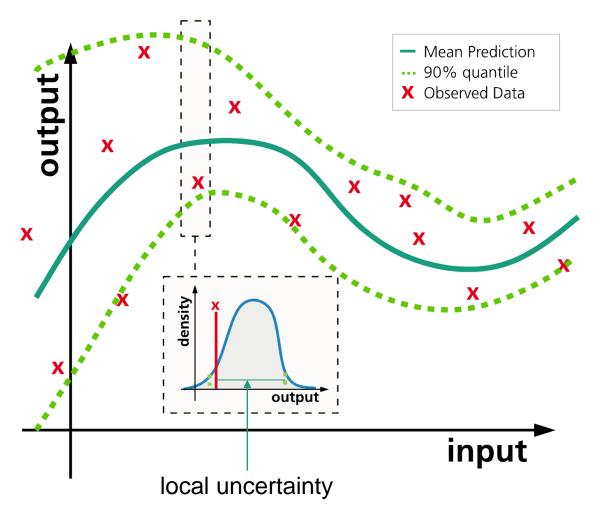
Realistic Uncertainty Estimation Know If You Know Nothing



- Local Uncertainty Estimation allows for a self-assessment of the neural network given its input, e.g., in order to detect out-of-distribution inputs
- **Increase safety** by discarding uncertain predictions
- **Optimize your dataset** by identifying data points with high uncertainty



Realistic Uncertainty Estimation Know If You Know Nothing



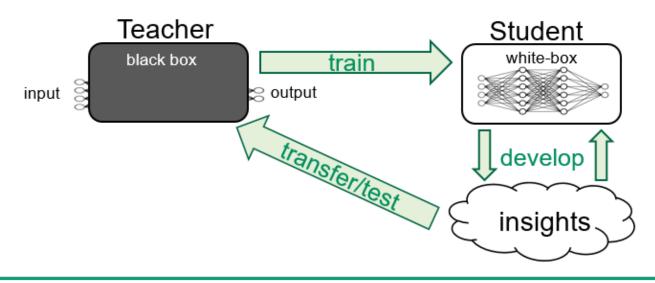
- State-of-the Art: Bayesian Networks, Deep Ensembles, MC Dropout
 - Poorly calibrated: Predictions are corrected by post processing
 - Yet, realistic local uncertainties are of minor quality
- Our approach:
 - Modify the loss function to provide realistic MC Dropout uncertainties
 - Formal understanding and proofs for uncertainty estimation in MC Dropout networks and deep ensembles

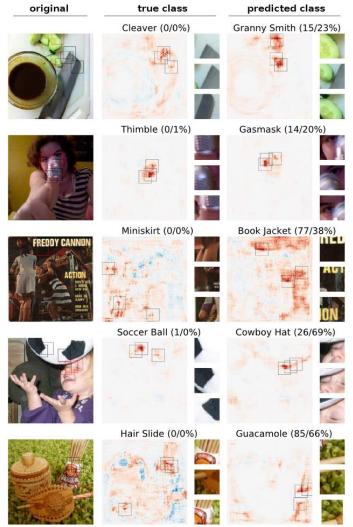


Teacher-Student-Methods

Gain Insight into the Inner Workings of a Neural Network

- Derive interpretable model (student) from a given black-box-model (teacher)
 - Identify erroneous "explanations"
 - Does the teacher suffer from the same problem?
- Enables analyses of the teacher model





1904.00760)

Image: BagNet (arXiv:



Teacher-Student-Methods Exploit Identified Insights

- Derive **interpretable model** (student) from a given black-box-model (teacher)
 - Identify erroneous "explanations"
 - Does the teacher suffer from the same problem?
- Enables analyses of the teacher model
- By these means we can construct semantic attacks

Binary classification: Is there a car in this image?



Student model considers traffic beacon an important hint Semantic attack: Add traffic beacons to an image



Student model predicts a car in this image Teacher model (a ResNet) does so, too.

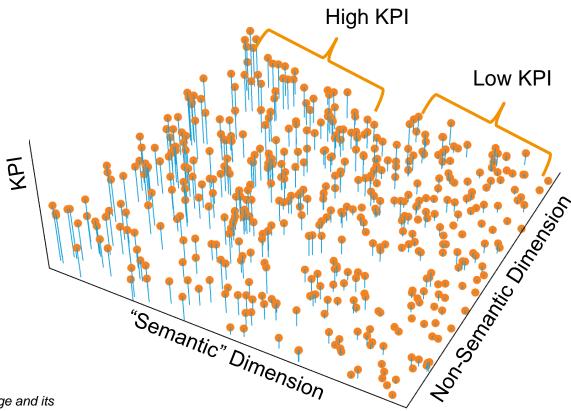


Assess Test Data Completeness

Find Situations with Systematically Low Performance

- Find correlations among semantic concepts (e.g., position / size of pedestrians)
 - and poor model performance
 - or pronounced and distinct safety concerns
- Reveal situations with poor prediction performance
- Reveal poor training procedures
 - E.g., with **Neuron Coverage** (Percentage of neurons that are sufficiently activated by at least one test example)

Abrecht, Akila, Gannamaneni, Groh, Heinzemann, Houben, Woehrle, "Revisiting Neuron Coverage and its Application to Test Generation". Third International Workshop on Artificial Intelligence Safety Engineering, SAFECOMP WAISE, 2020 (Best Paper Award)





Seite 15

Evaluation of Dependencies between Neural Networks and Data Visual Interactive Analysis of Semantic Features

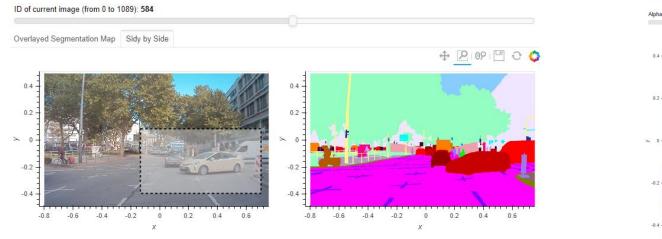
- Goal: finding correlated insufficiencies and gaining insight into the decision of networks
- Understanding semantic concepts of the data is the key to identifying & distinguishing outliers from systematic weaknesses (like shortcuts or data flaws)
 - But: automated analysis of semantics is difficult
 - Those semantic features are examined best visually by humans





Evaluation of Dependencies between Neural Networks and Data Finding Semantic Clusters in a Visual Interactive Interface

- Specific focus on enabling the human expert to:
- Interactively analyze the KPIs w.r.t. robustness
- Inspect image data sets to gain insights
 - E.g. into important image parts, hard or underrepresented images/image scenes ("corner cases"), unusual object appearances, data flaws etc.



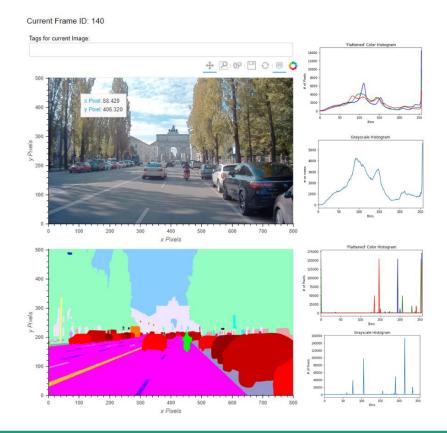




Evaluation of Dependencies between Neural Networks and Data Finding Semantic Clusters in a Visual Interactive Interface

- Enabling the human to understand semantic concepts of the data with additional information
 - E.g. metadata, histogram data
- Identify semantic clusters
 - Use VA to develop metrics incorporating human semantic understanding and DNN performance measures
 - E.g. by textual and visual querying ("query by example") and filtering
 - E.g. tagging, sorting and searching images







Evaluation of Dependencies between Neural Networks and Data Establishing a Feedback Loop

- These insights can then in turn be used to enhance
 - The training methods of the neural networks
 - The data set generation
- Establishing a feedback loop between data generation, neural network training and analyses of both

Data generation				Inge Inged: The data state and the data that is to be part ingeting the state and the data that is to be part ingeting the state and the stat	
Feedback: e.g. need for more images containing detected corner case	Mechanism developme	nt			
	Feedback, e.g. introspective heatmap mechanism should generate more fine- grained heatmaps	Data annotation Feedback, e.g. need for more detailed scene description to find data partitionings in diverse weather conditions Feedback, e.g. high uncertainty for predictions on pedestrians standing at bus stops		D of anomet mapping have 3 to 1009; 544 Developed Segmentation May Table 1000; 544 0 another the segmentation May Table 1000; 545 0 another the segmentation May Table 1000	



Beyond Absicherung Typical Challenges of the Individual Audit Areas





Certifying Artificial Intelligence Whitepaper Points out Audit Areas

- Collaboration of experts from Fraunhofer IAIS, Univ. Bonn and Univ. Cologne from the fields of
 - Machine Learning
 - Law
 - Ethics
 - IT Security
- Interdisciplinary initiative funded by the competence platform KI.NRW
- Audit areas for trustworthy AI
- www.iais.fraunhofer.de/ki-zertifizierung



Publication with high international attention





Advanced Trainings

Data Scientist Specialized in Trustworthy AI

- Advanced training offerd by Fraunhofer IAIS "Data Scientist Specialized in **Trustworthy Al"**
 - Audit areas of trustworthy AI
 - Methods for assessing and verifying AI applications
- Project "KI-Absicherung" **VDA** Leitinitiative "Autonomous and Connected Driving" www.ki-absicherung.vdali.de
- Point of contact: PD Dr. Michael Mock michael.mock@iais.fraunhofer.de

Data Scientist Advanced Trainings at Fraunhofer-Alliance Big Data and Artificial Intelligence



www.bigdata.fraunhofer.de/datascientist



THANK YOU FOR YOUR ATTENTION

Dr. Sebastian Houben | Fraunhofer IAIS | sebastian.houben@iais.fraunhofer.de The Connected Car and Autonomous Driving, October 26th, 2020





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