



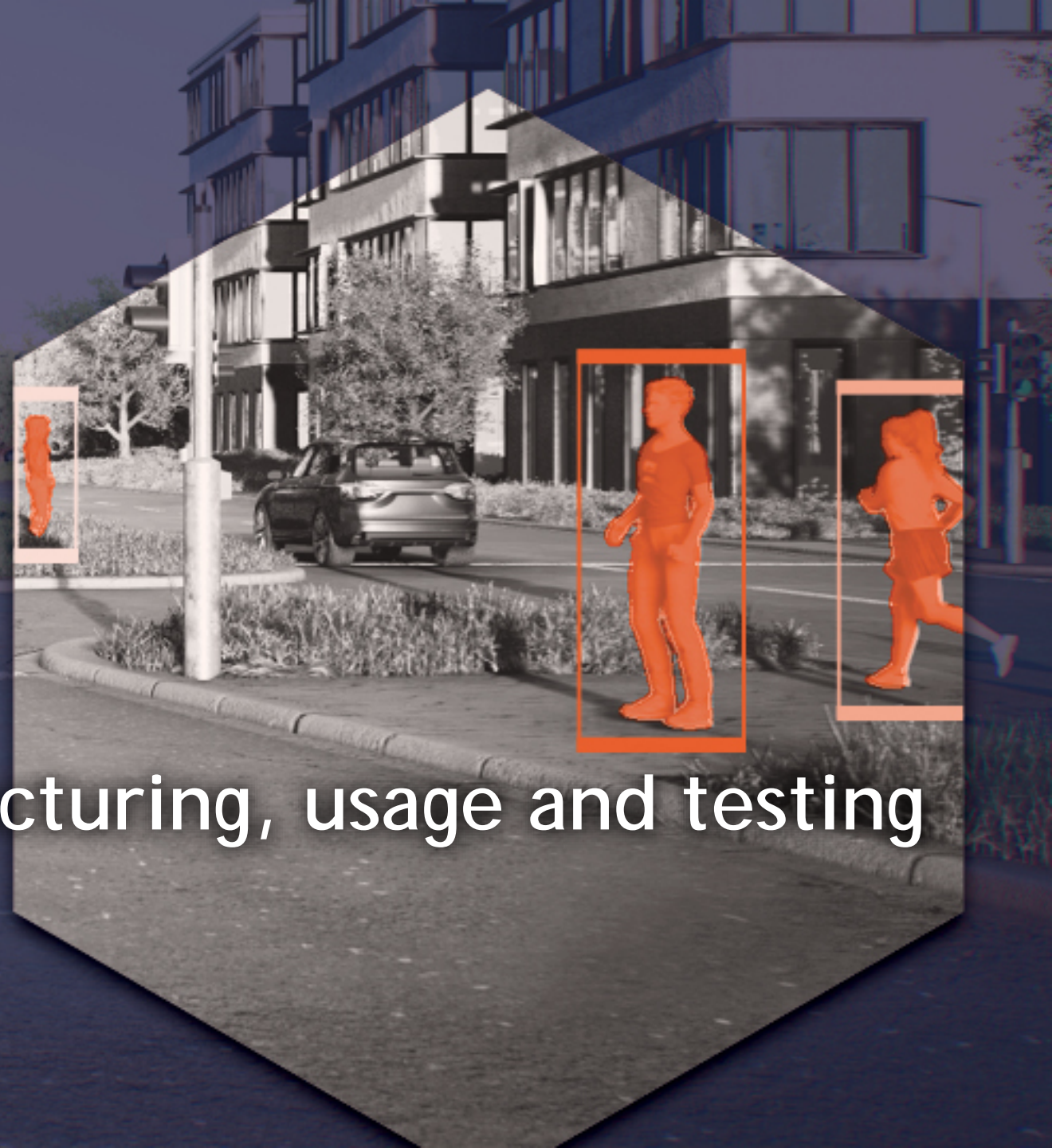
KI
ABSICHERUNG
Safe AI for Automated Driving

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Frédéric Blank (Robert Bosch GmbH)

Ontology-based data structuring, usage and testing in KI Absicherung

October 7th, 2021





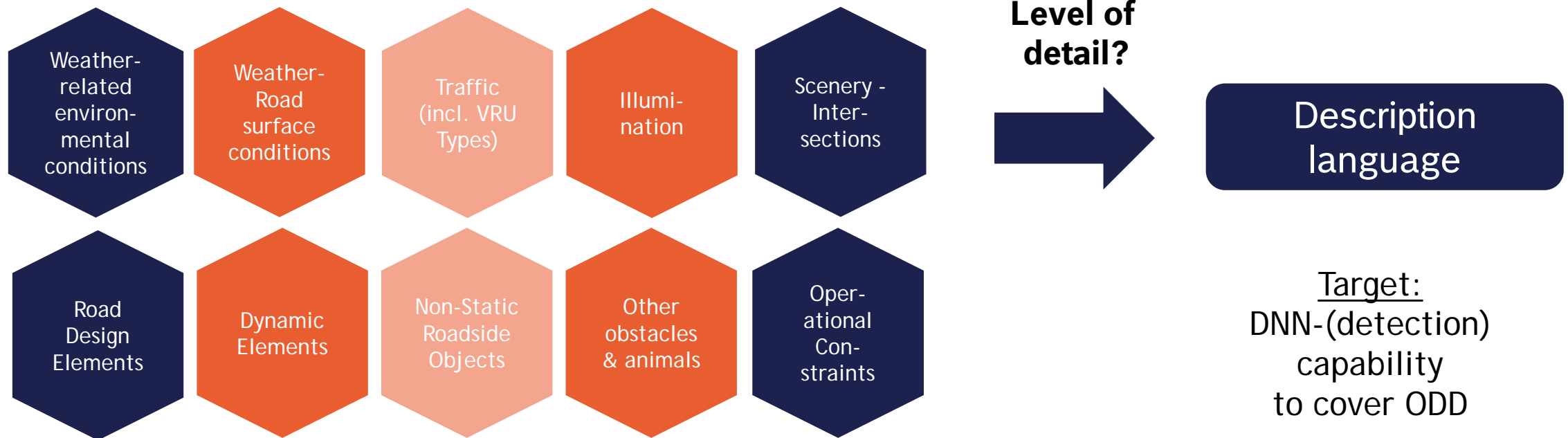
1

Description language and ontology development

Operational design domain (ODD)



- An ODD describes / specifies operating conditions under which a given driving automation system or feature is specifically designed to function [...]
- Taxonomy and Definitions for Terms Related to Driving Automation Systems (examples):



A description language & data input space modeling is needed to...



Complexity of language



Be able to describe / **specify operating conditions** (and edges of ODD*) as of PAS 1883:2020 and others



Systematically capture important knowledge and describe the (expected) **key input space dimensions** and their **possible variations** having an influence on the functional performance of a DNN-based function (→ Zwicky Boxes & Ontology)



Perform training and assurance **data coverage estimations** for data driven AI-based systems



Describe **Corner cases / rare critical situations** to be considered in training / test data sets



For synthetic perception data production & meta-data: describe data dimensions that should be varied & **incrementally generate new data** by analyzing coverage and generating missing combinations

DNN-specific Safety Concerns (examples)



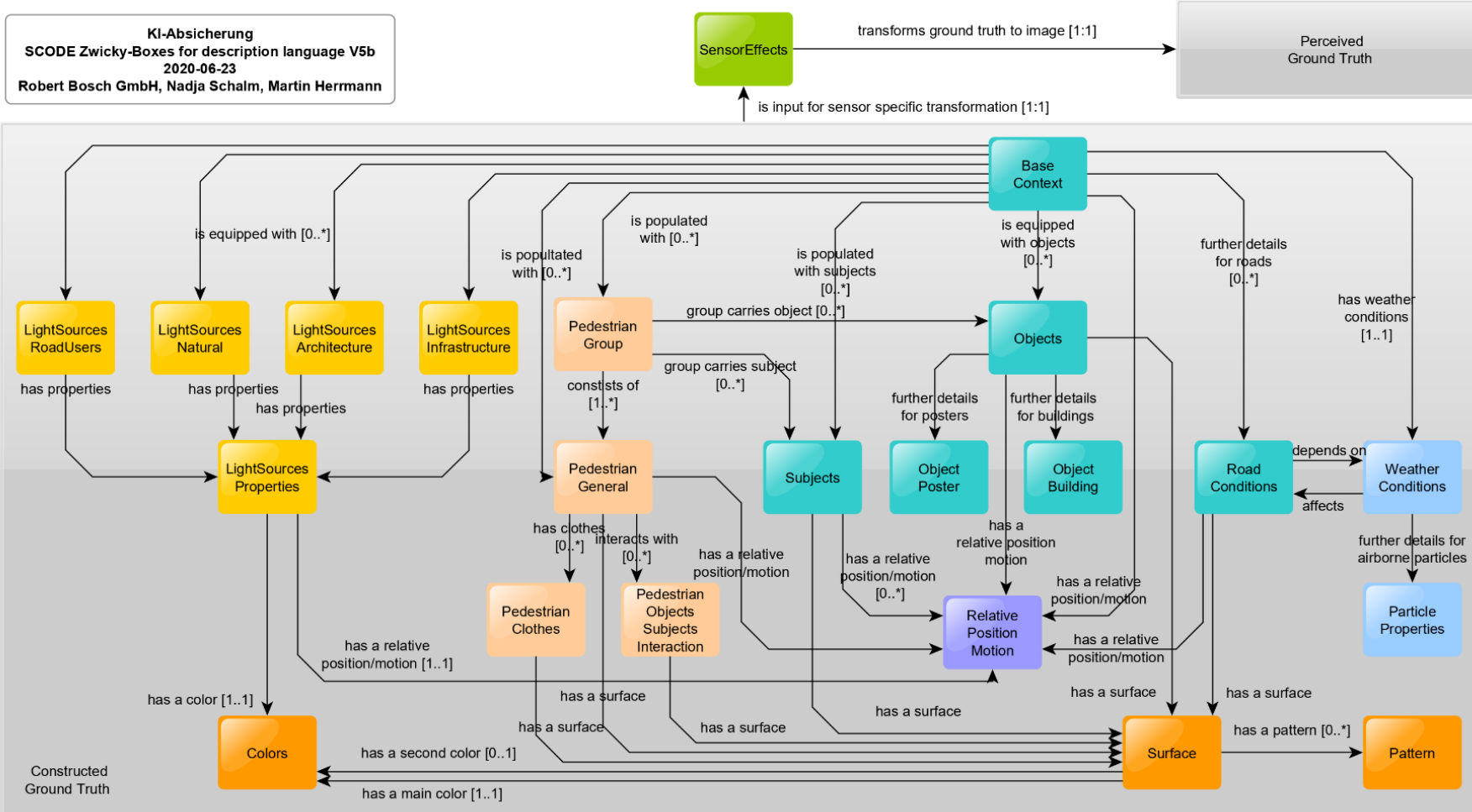
Data distribution is not a good approximation to target domain



Unknown behavior in rare safety-critical situations

Based on: O. Willers, S. Sudholt, S. Raafatnia, S. Abrecht: Safety Concerns and Mitigation Approaches Regarding the Use of Deep Learning in Safety-Critical Perception Tasks

High Level view of Ontology / Domain model derived from SCODE Zwicky-Boxes

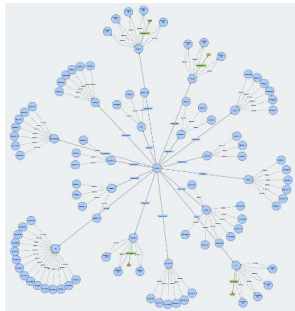


Quelle: Bosch

Data representations of the data input space aligned to ontology

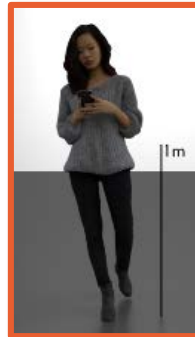


Ontology Graph (Relations)



Visualization of KI Absicherung pedestrian sub ontology

Asset & Object descriptions for data analytics



Pedestrian:Age "adult"
 Pedestrian:BodyHeight "160cm-200cm"
 Pedestrian:BodyShape "thin"
 Pedestrian:BodyType "hourglass"
 Pedestrian:FaceShape "oval"
 Pedestrian:Gender "female"
 Pedestrian:HairColor "black"
 Pedestrian:HairLength "long"
 Pedestrian:HairStyle "other"
 Pedestrian:Pigmentation "medium"
 Pedestrian:Pose "walking"
 Pedestrian:SkinModification "no"
 Pedestrian:SpecialHandicap "no"

Source: BIT-TS

Systematic Combination of variations

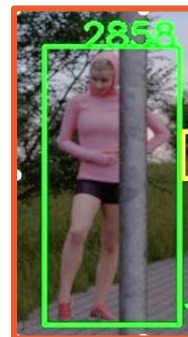
Dimension	Person1	Person2	Person3	...
Age	Child	Teenager	Adult	
Gender	Male	Female	Male	
Body height	80-120 cm	120-160 cm	160-200 cm	
Pose	Running	Lying	Walking	
Pedestrian Location	Middle of street	Left side walk	Right side walk	
...	

Representations of variations

DAYTIME	morning	day	evening	night	
HAZE/FOG	no		yes		
STREET CONDITION	dry	wet	icy	snow	broken
SKY	cloudy		no	clear	
RAIN	no		yes		
REFLECTION ON ROAD	no		yes		
SHADOW ON ROAD	no		yes		
VRU TYPE	adult		child		
VRU POSE	pedestrian	jogger	cyclist		
VRU CONTRAST TO BG	low		high		

Zwicky Box - Discretized variations of important dimensions (Bosch)

Object GT Annotations for DNN-Training & Testing



Height = 55 px
 Width = 10 px
 Occlusion_level: 80%
 Occluded_body_part: arm
 Occluder: lamp
 Within_breaking_distance_30kph: true

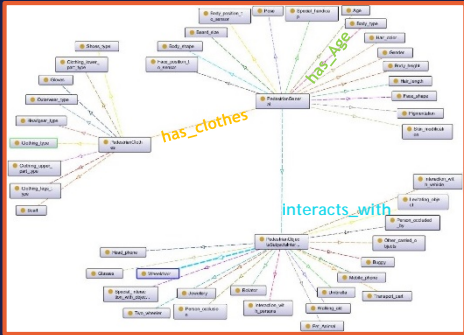
Source: BIT-TS

Systematically identify and describe the (known / expected) **key input space dimensions** and their **possible variations & combinations** having an influence on the functional performance of a DNN-based function

Structured Incremental dataset generation to boost data coverage (Vision)



Domain model / data input space



Known Performance Limitations

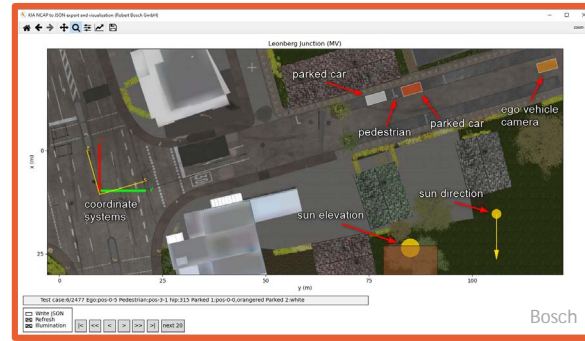
(expert know-how)

- Person occlusions
- Low contrast: similar color to background
- Uncommon person locations
- Uncommon poses

Possible Variations based on Ontology



Data Request tool



JSON



Synth. Data production



Annotations & Meta data



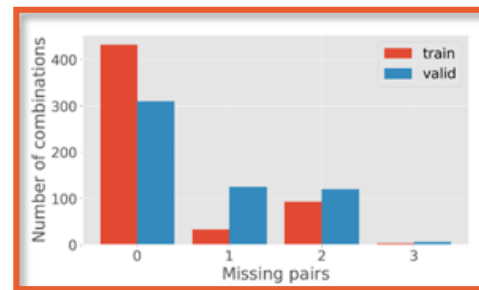
Data Coverage Analysis

Dimension	Person1	Person2	Person3
Age	Child	Teenager	Adult
Gender	Male	Female	Male
Body height	80-120cm	120-160cm	160-200cm
Pose	Running	Lying	Walking
Pedestrian Location	Middle of street	Left side walk	Right side walk
...

Incremental dataset generation



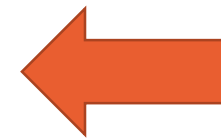
Missing Combinations



Constrained test-space



Optimized Combinatorial Testing (example)

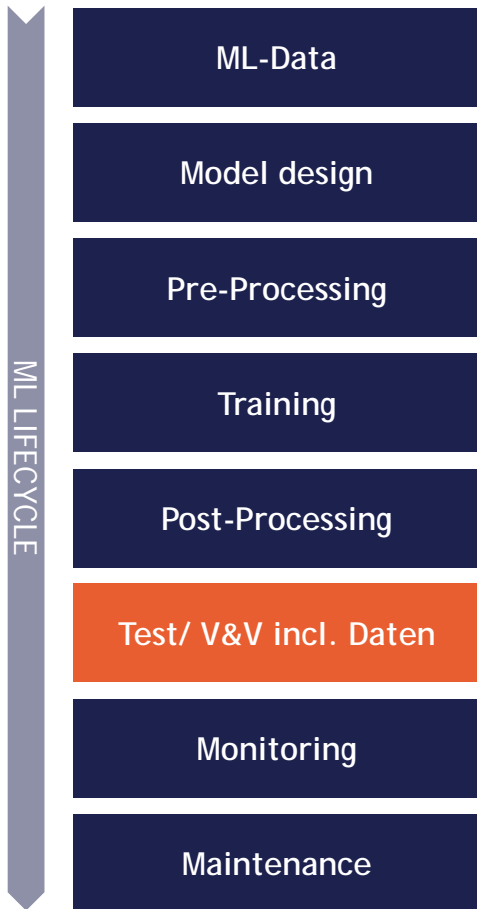




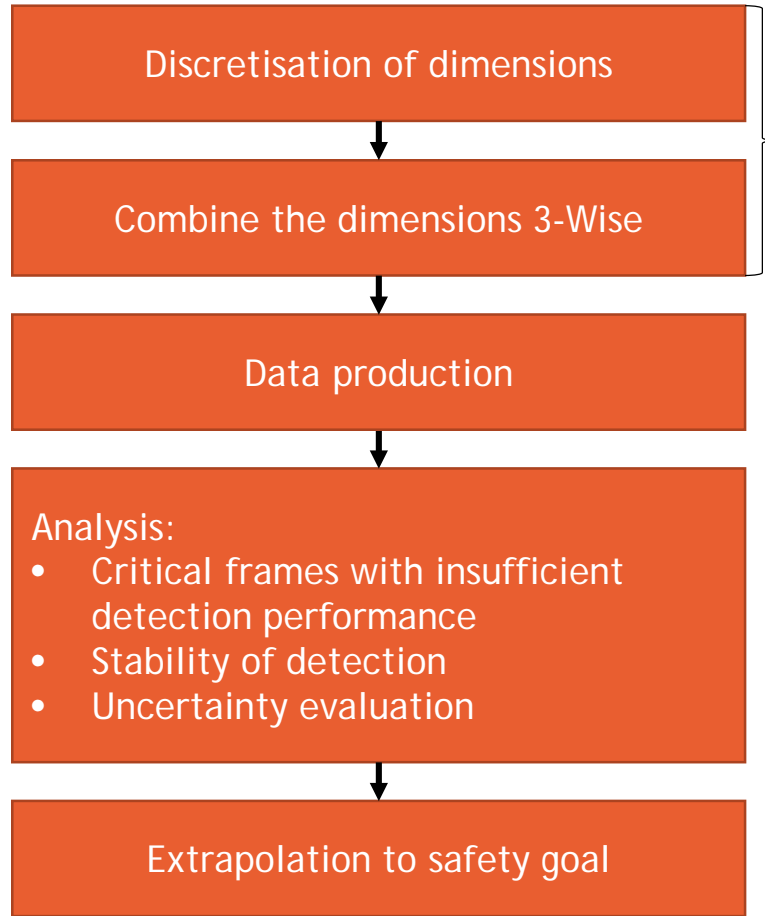
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NCAP inspired test data production process

ML-Lifecycle-Validation data



Relations to the safety assurance with NCAP like scenarios



Data request JSON

```

{
  "meta": {
    "uid": "72878ced-6535-4bfb-bb14-7e6c1debb106",
    "modelName": "V2X",
    "name": "High-Prio-Swicky-Boxes",
    "requester": "Kia@QualityIndia.de",
    "request": "Example request including high prio swicky boxes",
    "description": "A sample description.",
    "iteration": 0,
    "frame_rate": 0.0
  },
  "scene_context": {
    "open_drive_file_ref": "/Samples/Ref.opendrive",
    "road_types": [
      {
        "lane_id_ref": "A12",
        "surface": "Asphalt",
        "road_quality": "Good",
        "road_wetness": "NoWithPuddles",
        "reflection": "Medium",
        "reflection_directionality": "SlightlyGloss",
        "retro_reflectivity": true,
        "structure": "High",
        "translucence": "No"
      },
      {
        "lane_id_ref": "B100",
        "surface": "Asphalt",
        "road_quality": "Perfect",
        "road_wetness": "SlightlyWet",
        "reflection": "High",
        "reflection_directionality": "MicroGloss",
        "retro_reflectivity": false,
        "structure": "Medium",
        "translucence": "No"
      }
    ]
  }
}
  
```

Data production



Raw data

- Analysis:
- Critical frames with insufficient detection performance
 - Stability of detection
 - Uncertainty evaluation



- DS1 Information lossy data
- DS2 Noise generation
- DS3 Image corruptions
- DS4 Adversarial attacks
- DSx ...

Modified data

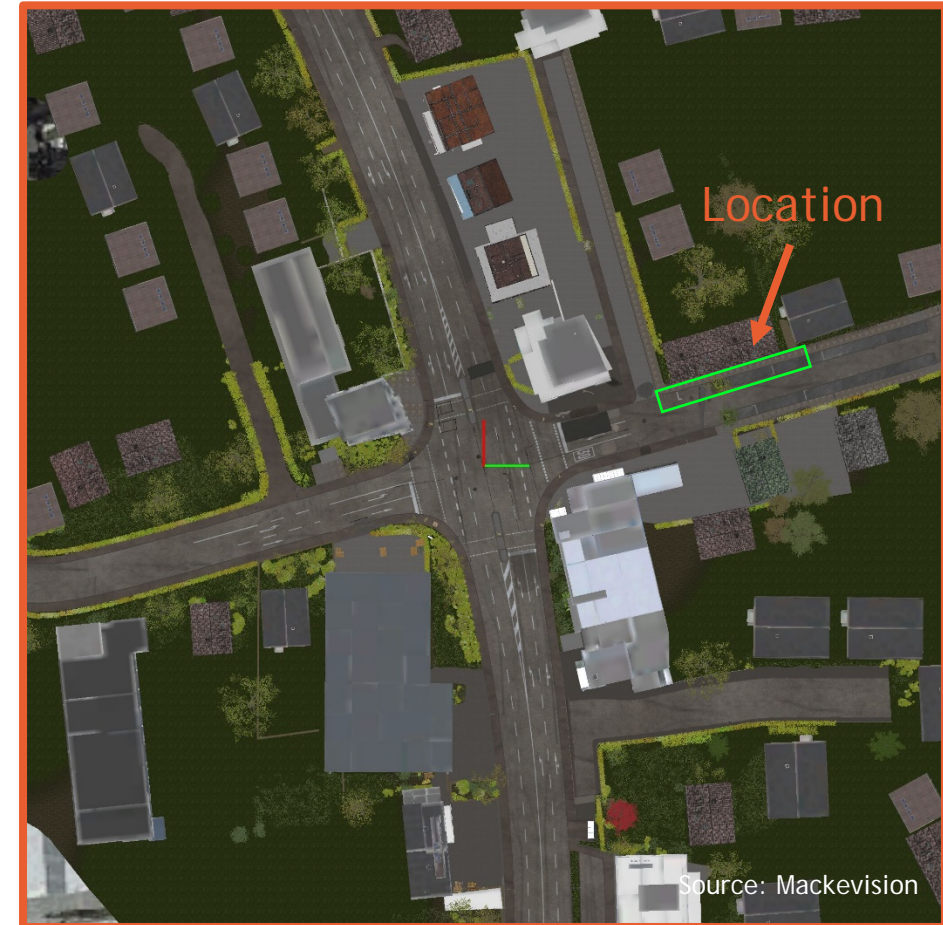
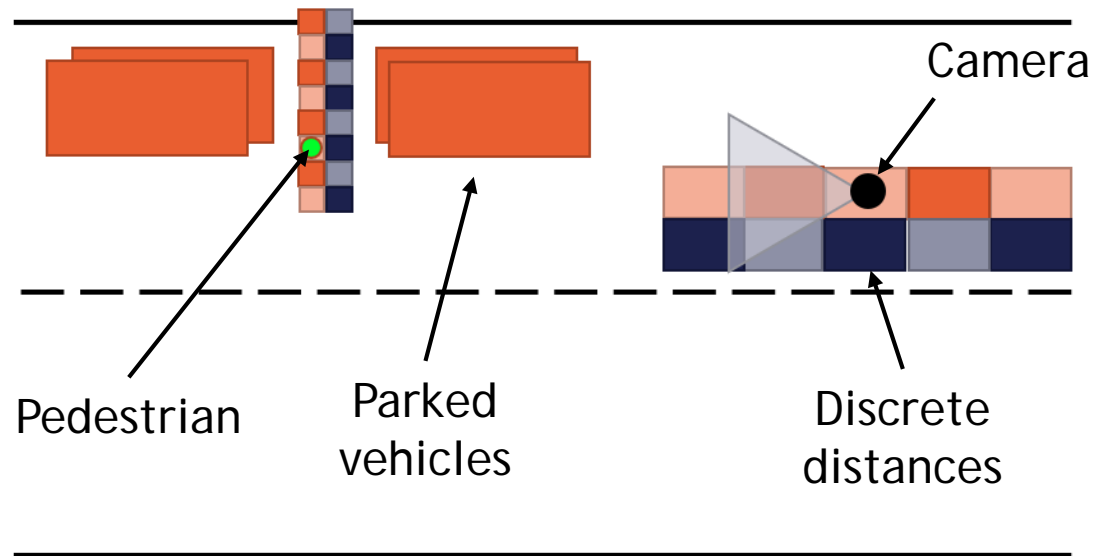
All relevant pedestrians are detected in the ODD

Definition of base scenario and location on base context



Story

A pedestrian is approaching the ego vehicle between two parking cars under different environment conditions



Discretization of dimensions in “Zwicky Boxes”

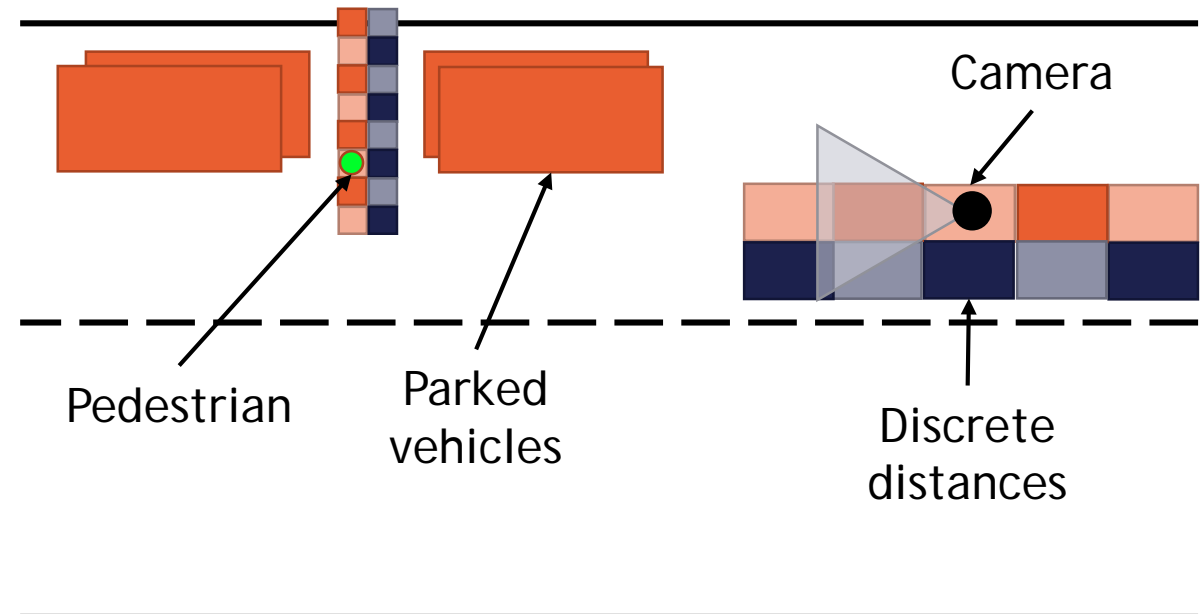
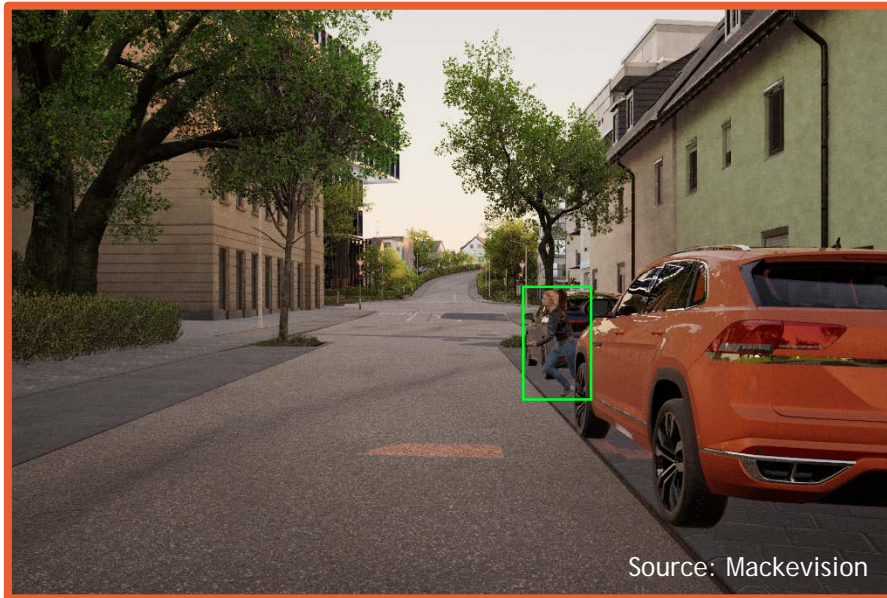


Ego XY position	pos-0-0		pos-0-1		pos-0-2		pos-0-3		pos-0-4		pos-0-5		pos-1-0		pos-1-1		pos-1-2		pos-1-3		pos-1-4		pos-1-5									
Pedestrian XY position	pos-0-0		pos-1-0		pos-2-0		pos-3-0		pos-4-0		pos-5-0		pos-6-0		pos-7-0		pos-0-1		pos-1-1		pos-2-1		pos-3-1		pos-4-1		pos-5-1		pos-6-1		pos-7-1	
Pedestrian pose	pose01					pose02					pose03					pose04					pose05											
Pedestrian asset	A1			A2			A3			A4			A5			A6			A7			A8			A9			A10				
Pedestrian hip direction	d0			d45			d90			d135			d180			d225			d270			d315										
Parked vehicle 1 type	BMW1					BMW2					BMW71					VW ID.3					VW Golf 8					VW Atlas						
Parked vehicle 1 XY position	pos-0-0			pos-0-1			pos-0-2			pos-1-0			pos-1-1			pos-1-2			pos-2-0			pos-2-1			pos-2-2							
Parked vehicle 1 color	BMW Black	BMW Cerium grey	BMW Melbourne red	BMW Mineral grey	BMW Misano blue	BMW Sao Paolo yellow	BMW Snapper Rocks blue	BMW Sunset orange	BMW White	VW Gletscher Weiss	VW Mangangrau	VW Mekana Turquoise	VW Mondsteingrau	VW Scale Silver	VW Stonewashed Blue	VW Energetic Orange	VW Deep Black	VW Delfingrau	VW Kings Red													
Parked vehicle 2 type	BMW1					BMW2					BMW71					VW ID.3					VW Golf 8					VW Atlas						
Parked vehicle 2 color	BMW Black	BMW Cerium grey	BMW Melbourne red	BMW Mineral grey	BMW Misano blue	BMW Sao Paolo yellow	BMW Snapper Rocks blue	BMW Sunset orange	BMW White	VW Gletscher Weiss	VW Mangangrau	VW Mekana Turquoise	VW Mondsteingrau	VW Scale Silver	VW Stonewashed Blue	VW Energetic Orange	VW Deep Black	VW Delfingrau	VW Kings Red													
Illumination	direct sun										diffuse light																					
Sun direction	d0			d45			d90			d135			d180			d225			d270			d315										
Sun elevation	low					medium										day																
Road surface	A					B					C					D																

Source: Bosch

- **Discretization:** The most critical dimensions are identified and discretized
- **Test coverage:** With pairwise testing it's possible to achieve a high error coverage in traditional software testing

Data production - Example data snapshot 1

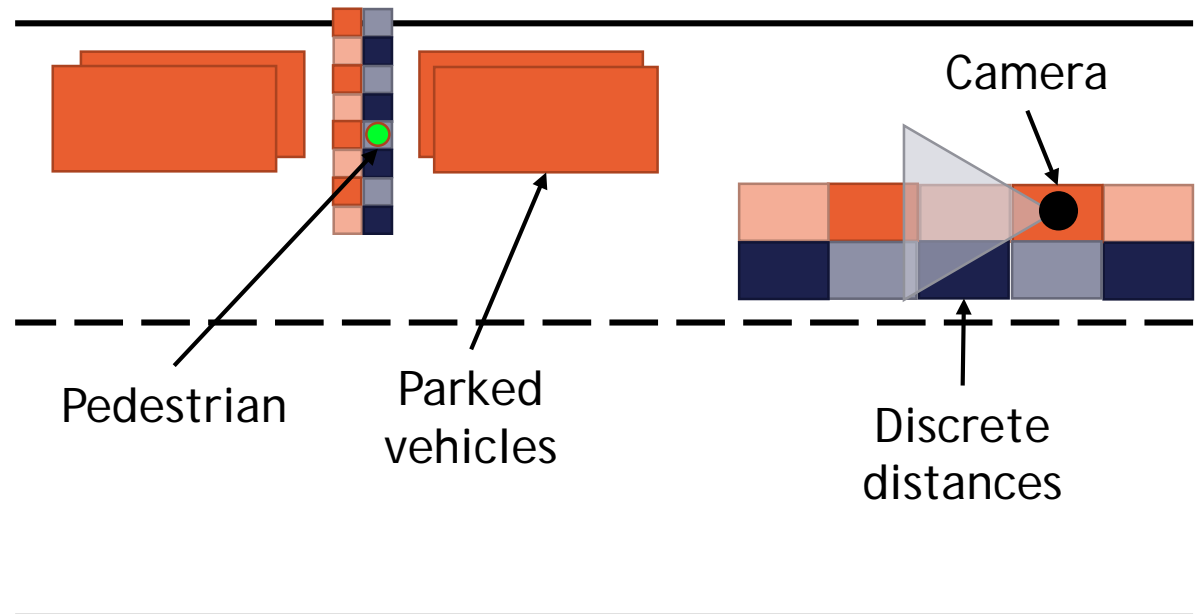


- **Safety critical:** Pedestrian has a running pose towards the camera
- The perception function shall be able to detect the pedestrian early enough without any image perturbations



- Those images are well suited as a reference for the analysis of brittleness in DNN's

Data production - Example data snapshot 1



- **Safety critical:** The legs are extended to the driving lane
 - **Uncommon pose:** Pedestrian lays between two vehicles and is difficult to see
- ➔
- In which combinations is the object detector **not** capable to perceive the pedestrian?

Examples for data post processing



original



fog



frost



brightness



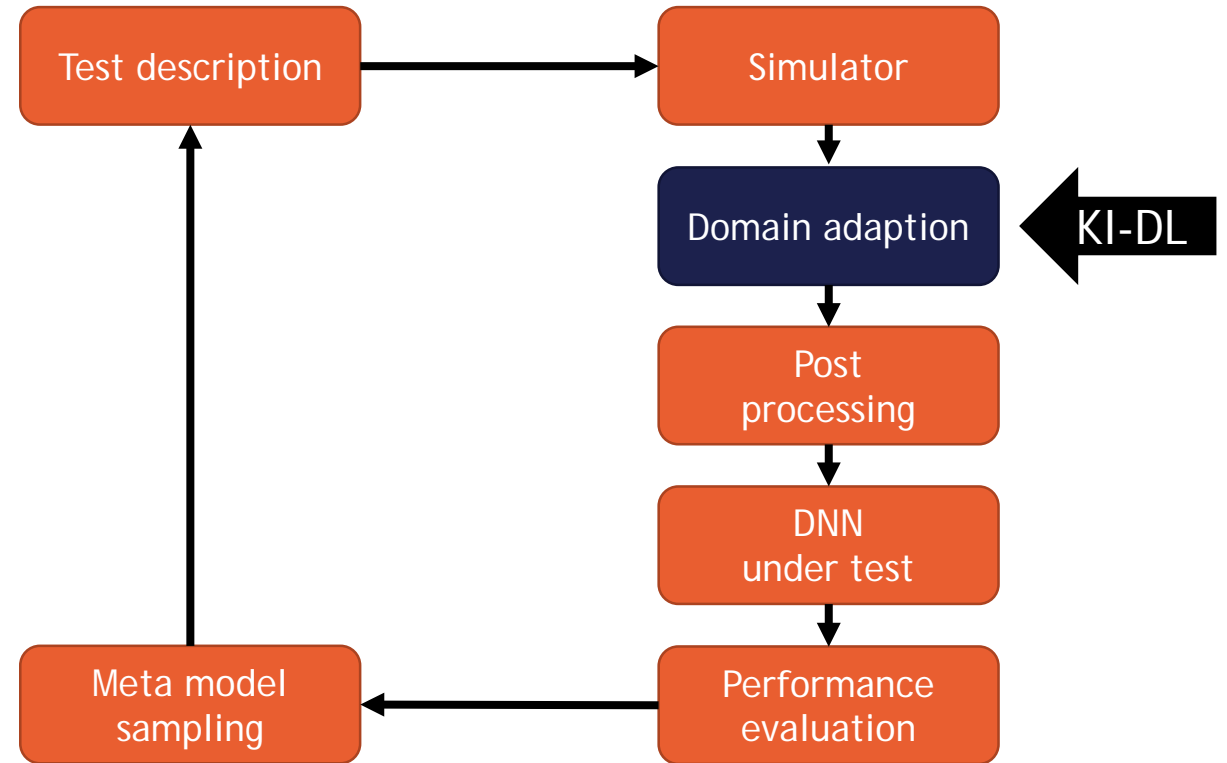
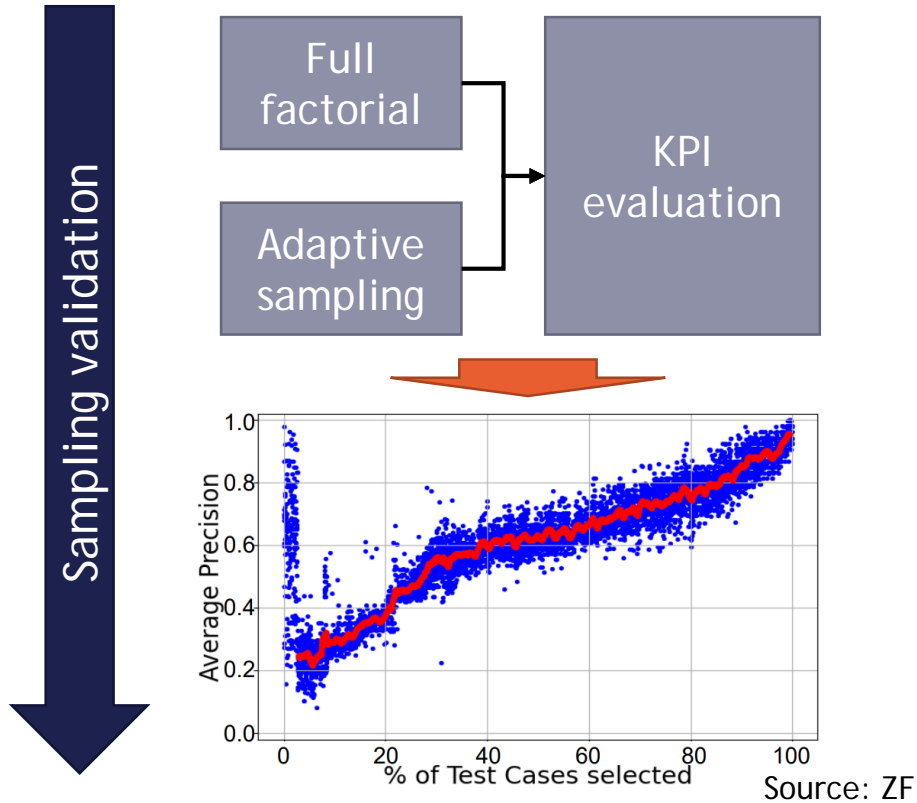
contrast



Motion blur



Test space exploration optimization



→ The most performance critical test cases are identified early in the test exploration
"Adaptive test case selection for DNN-based perception functions"

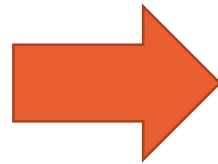
Paper release: [ISSE 2021 - 7th IEEE International Symposium on Systems Engineering \(ieeexpp.org\)](https://ieeexpp.onlinelibrary.wiley.com/doi/10.1109/ISSE46011.2021.9622100)

Synthetic video data for domain adaptation studies @ KI-Absicherung



Variation of sensor parameters

- High vs. Low resolution
- Camera opening angle (field of view)
- Height of camera over ground



Conclusion & Outlook



Conclusion

- Systematic structuring of training, test and assurance data is expected to be a crucial basis for safety assurance
- Challenging and safety critical scenes are structured based on expert knowledge
- An initial systematic coverage of test cases is provided by a combinatorial testing approach and further improved by an adaptive sampling strategy
- The “good” test cases are challenged by further augmentations and corruptions
- The newly created data set can be used for a benchmarking of DNN based perception algorithms

Proposals for KI-DL

- Embed the KI-A ontology knowledge into semi-supervised and unsupervised learning approaches from KI-DL
- Integrate domain adaption methods from KI-DL into the test space exploration pipeline