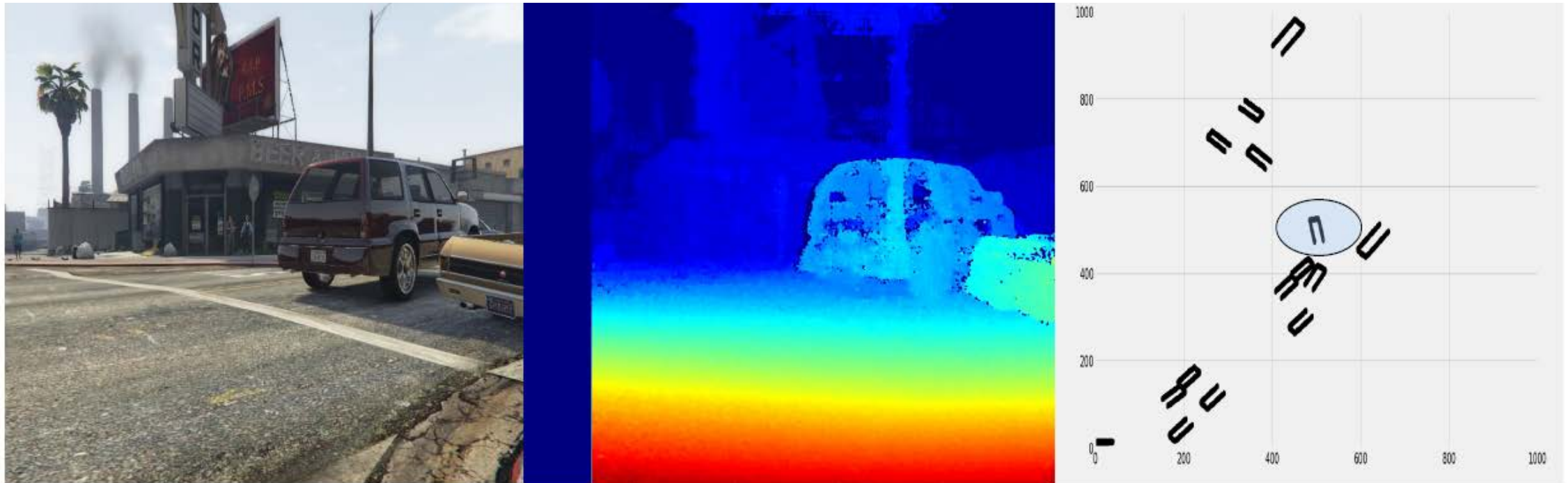


DYNAMIC RISK ASSESSMENT FOR VEHICLES OF HIGHER AUTOMATION LEVELS BY DEEP LEARNING

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* Author present

Motivation

WEDNESDAY SEPTEMBER 19TH

- With higher
- From a safety
- Assessment c
Dynamic Risk
- DRA in autor

08:00 - 08:45	Registration Desk
08:45 - 09:00	Welcome Note
09:00 - 10:00	Keynote Talk 1: Robyn Lutz (Session chair: Irfan Sljivo)
10:00 - 10:30	Coffee Break
10:30 - 12:00	Session 1 - Automotive safety standards and cross-domain reuse potential: (Session chair: Ewen Denney) Papers: <ul style="list-style-type: none">• Automotive Safety Practices vs. Accepted Principles (Practical Experience Report). <i>Philip Koopman</i>• A Generic Method for a Bottom-Up ASIL Decomposition. <i>Alessandro Frigerio, Bart Vermeulen and Kees Goossens</i>• Assurance Benefits of ISO 26262 Compliant Microcontrollers for Safety-Critical Avionics. <i>Andreas Schwierz and Håkan Forsberg</i>
12:00 - 12:30	Exposition overview
12:30 - 14:00	Lunch Break
14:00 - 15:30	Session 2 - Autonomous driving and safety analysis: (Session chair: Elena Troubitsyna) Papers: <ul style="list-style-type: none">• Structuring Validations Targets of a Machine Learning Function applied in Automated Driving. <i>Lydia Gauerhof, Peter Munk and Simon Burton</i>• Multi-Aspect Safety Engineering for Highly Automated Driving - Looking beyond functional safety and established standards and methodologies. <i>Patrik Feth, Rasmus Adler, Takeshi Fukuda, Tasuku Ishigooka, Satoshi Otsuka, Daniel Schneider, Denis Uecker and Kentaro Yoshimura</i>• A Model-Based Safety Analysis of Dependencies Across Abstraction Layers. <i>Christoph Dropmann, Eike Thaden, Mario Trapp, Denis Uecker, Rakshith Amarnath, Leandro Avila Da Silva, Peter Munk, Markus Schweizer, Matthias Jung and Rasmus Adler</i>

riate behavior.

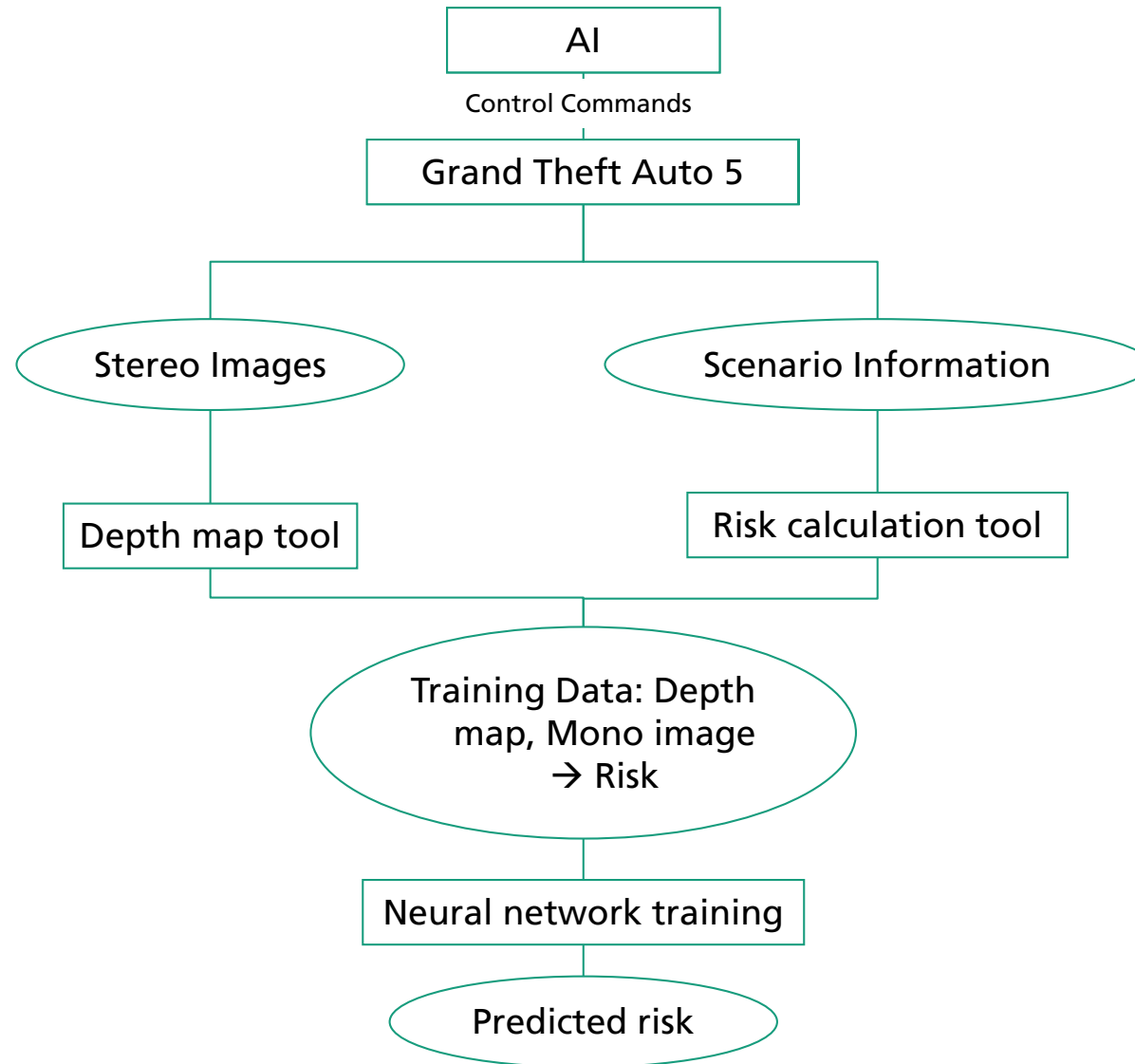
ation levels →

Motivation

- Risk metrics traditionally calculated based upon sensor data and algorithms
- Here we analyze if the calculation of a risk metric is also possible with a Convolutional Neural Network



Overview

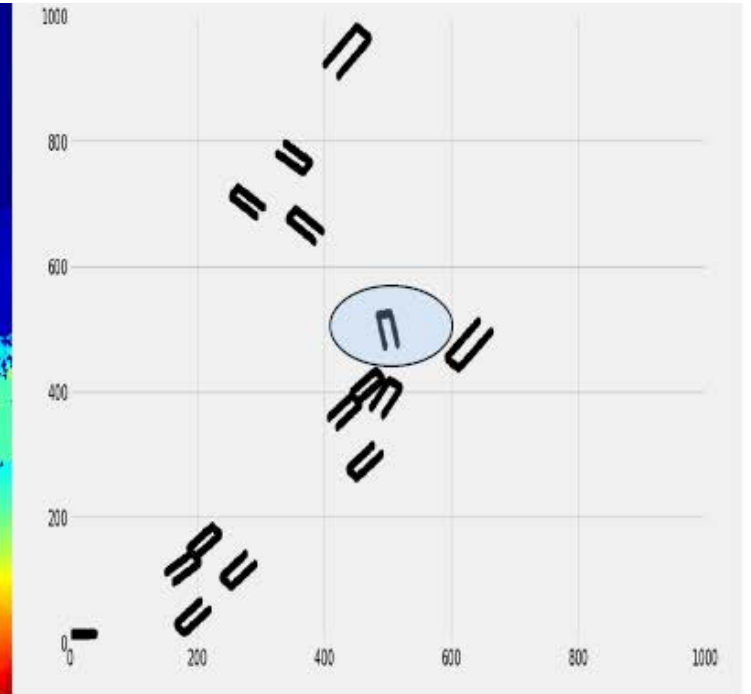
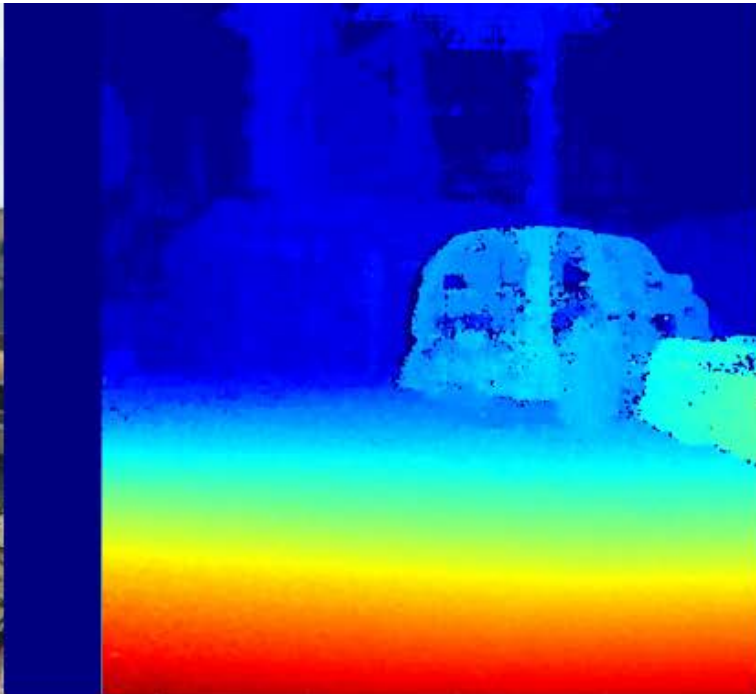


Brief methodology

- Virtual stereo camera simulation
- Data pre-processing:
 - Quality estimation
 - Disparity estimation
 - True risk estimation
- Training of neural network

Neural network

- Convolutional neural network
- By Nvidia DAVE-2



Confusion matrix of the CNN

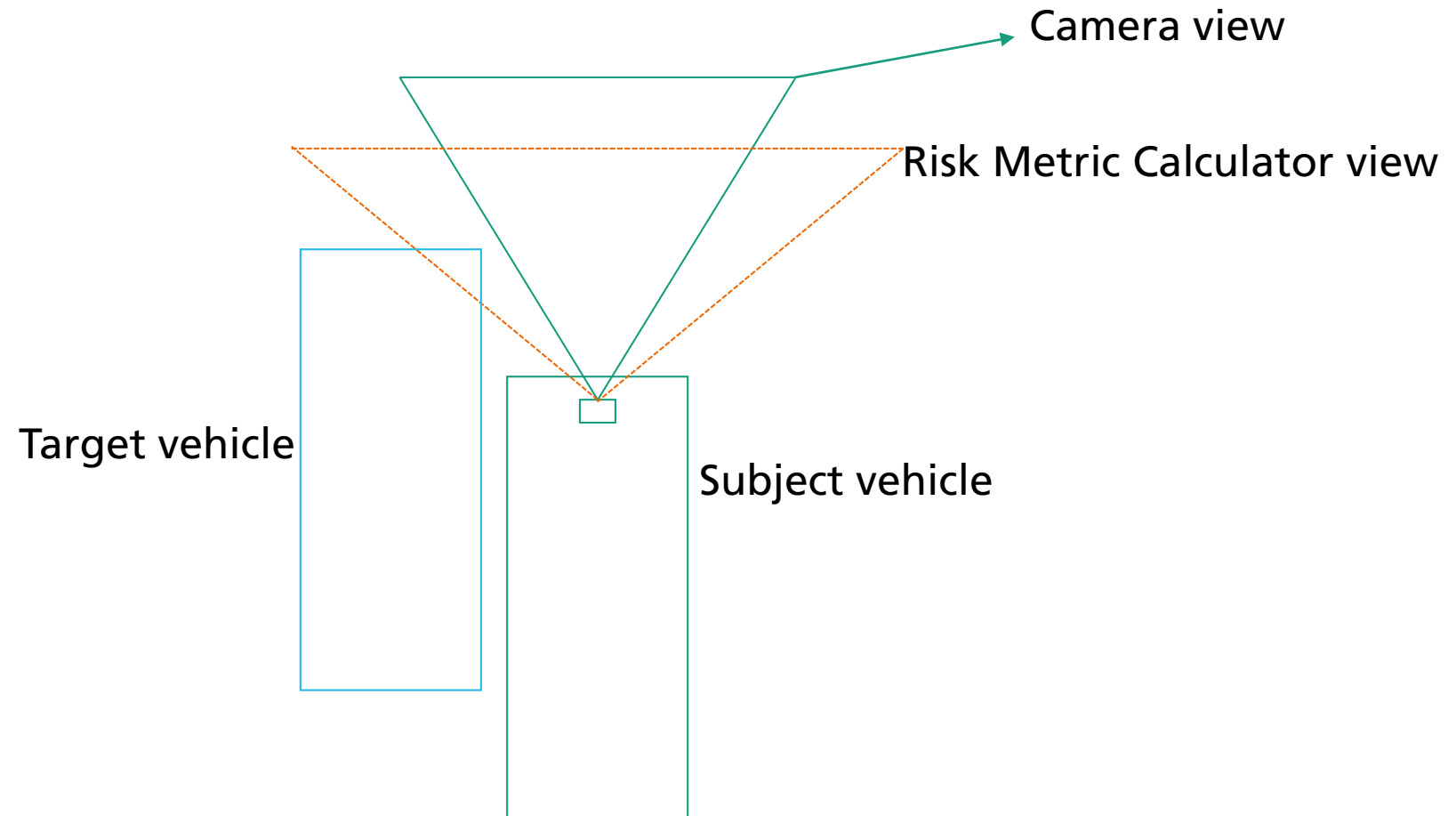
- Accuracy Paradox
- Threshold time headway: 1.5 seconds
- **72.87%**

	Actual critical situation	Actual uncritical situation
Predicted critical situation	67052 (63.16%)	19069 (17.96%)
Predicted uncritical situation	9737 (9.17%)	10312 (9.71%)

Is this accuracy enough?

- Depends on application. warning system vs emergency breaking systems.

Reason for inadequate performance



Reason for inadequate performance (Cont . . .)

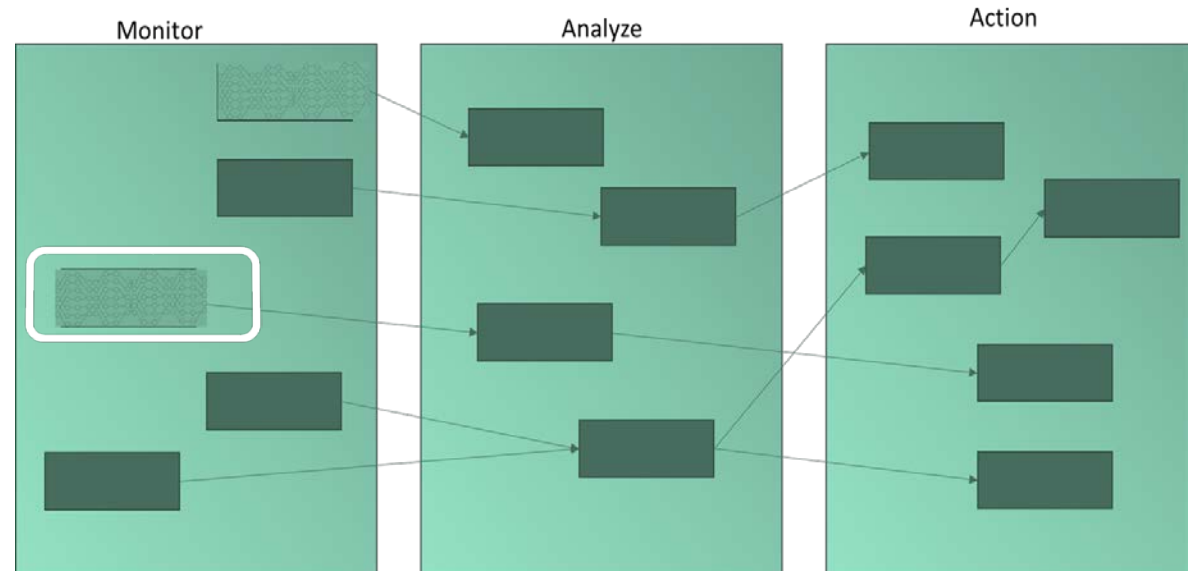


Risk from RMC: 0.36, CNN Prediction: 0.38

Risk from RMC: 0.75, CNN Prediction: 0.39

Conclusion

- Novel approach for the dynamic risk assessment from images
- The trained network achieved an accuracy of **72.87%** on our test set
- Heterogeneous redundancy



Future work

- Ground truth data should map the visible area for the CNN
- Using more sophisticated risk metrics
- Using more advanced neural networks, for example LSTM
- Using an open source simulation environment like CARLA

Thank You

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