

Comparing OTA/ViL and Real World Measurement Radar Data

German-Japan joint virtual validation methodology for intelligent driving systems – VIVID

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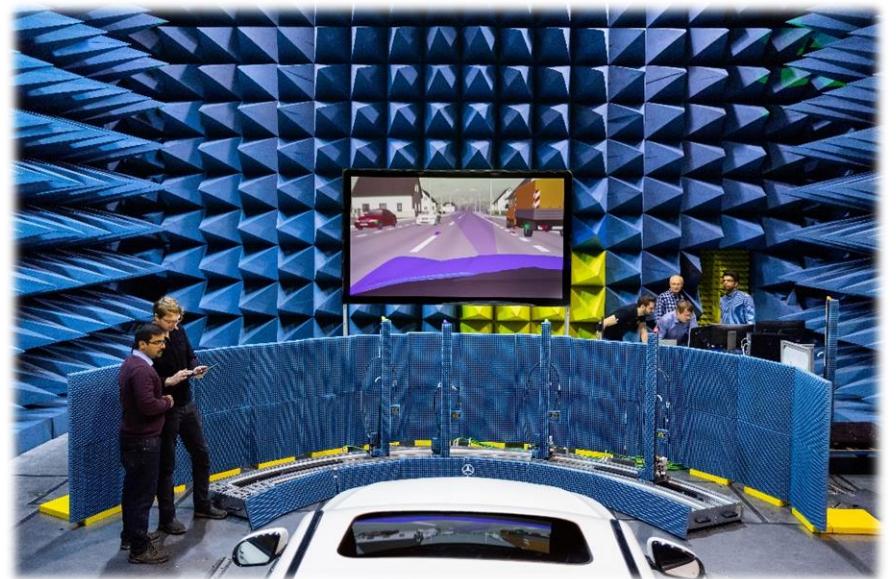
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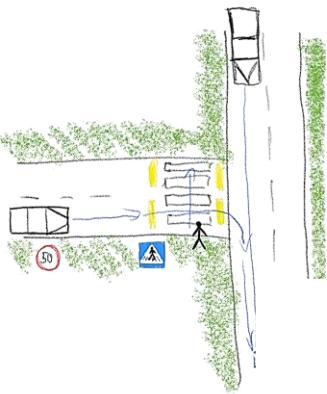


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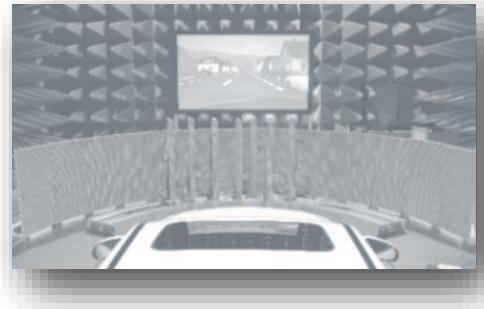
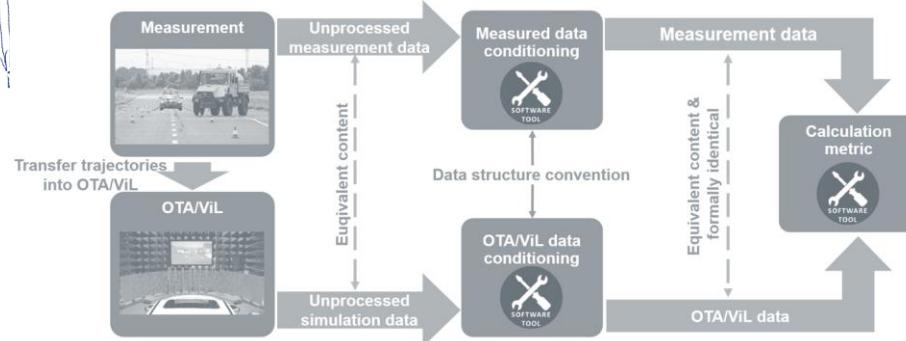


Outline

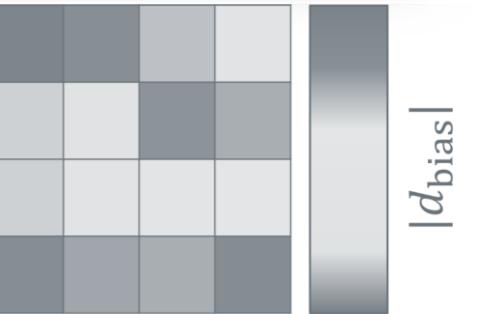
Introduction & Motivation



Methodology



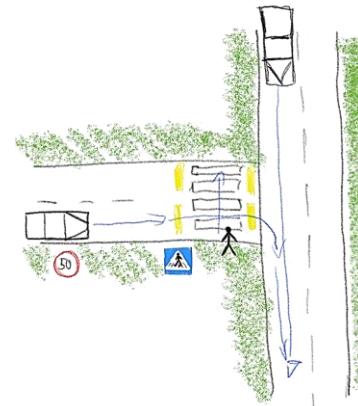
Experimental setup



Evaluation

Safety Assessment of Automated Driving Functions

1. Distance-based → scenario-based



2. Real environment → virtual environment



3. How X is X enough?

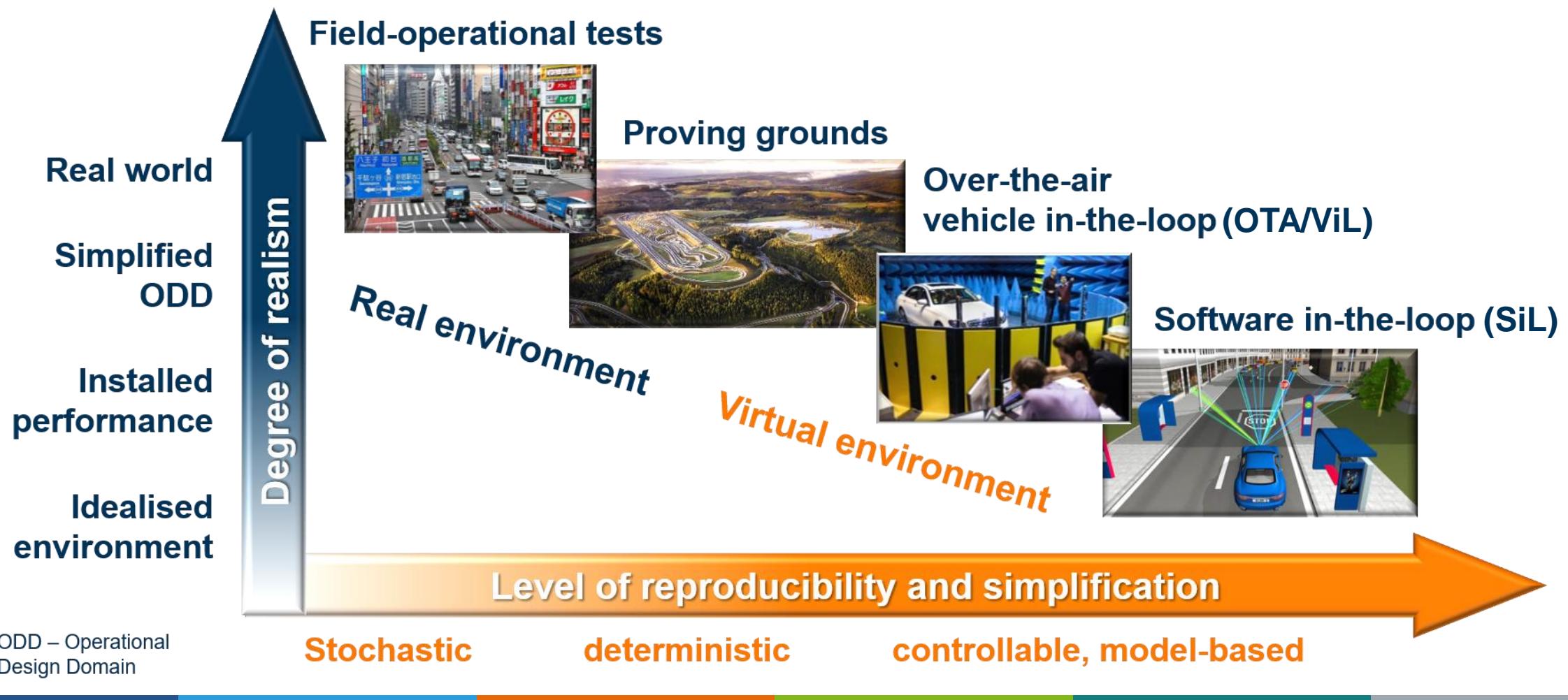
X = Realistic, evident, consistent, justifiable, credible, safe

4. Virtual verification and validation based on harmonized global R&D and standardization efforts



Adapted source: M. Hein et al.: German research approach towards AD safety assurance, Keynote SIPadus Workshop 2022, Kyoto, 2022

Environment Overview for Safety Assessment



State of the Art

Radar sensor in OTA/ViL for scenario-based virtual safety validation of automated driving

Validation procedure of radar sensors in OTA/ViL tests

Intended usage:

- Testing and development of algorithms, models, and functions (e.g. object detection, trajectory planner)
- Objective validity statement of radar sensors in OTA/ViL tests
- Safety assurance for automated driving functions

State-of-the-art:

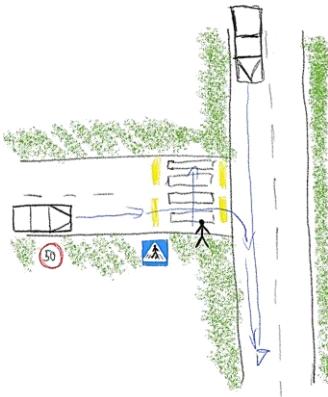
- Metric based analysis by comparison of real and OTA/ViL data
- Hypothesis tests to prove statistical properties
- Expert knowledge driven and subjective process due to result interpretation

→ How do we objectively measure the OTA/ViL capabilities based on measurements?

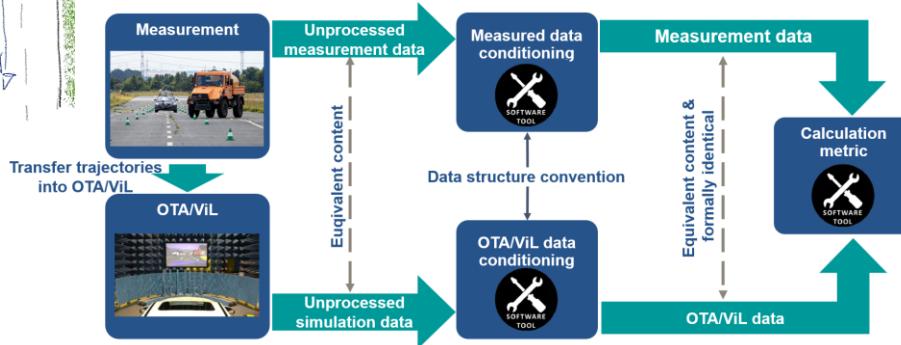
Adapted source: M. Viehof: *Objektive Qualitätsbewertung von Fahrdynamiksimulationen durch statistische Validierung*, PhD Thesis, TU Darmstadt, Darmstadt, 2018

Outline

Introduction & Motivation



Methodology

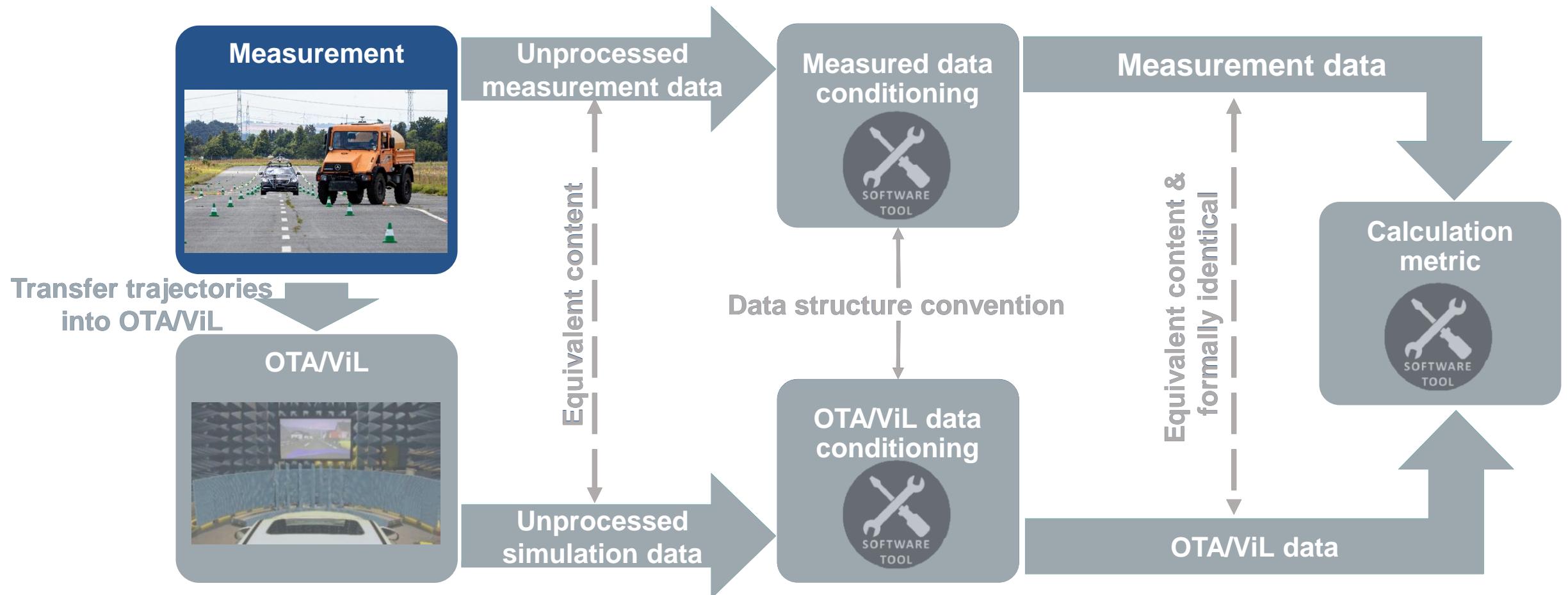


Experimental setup



Evaluation

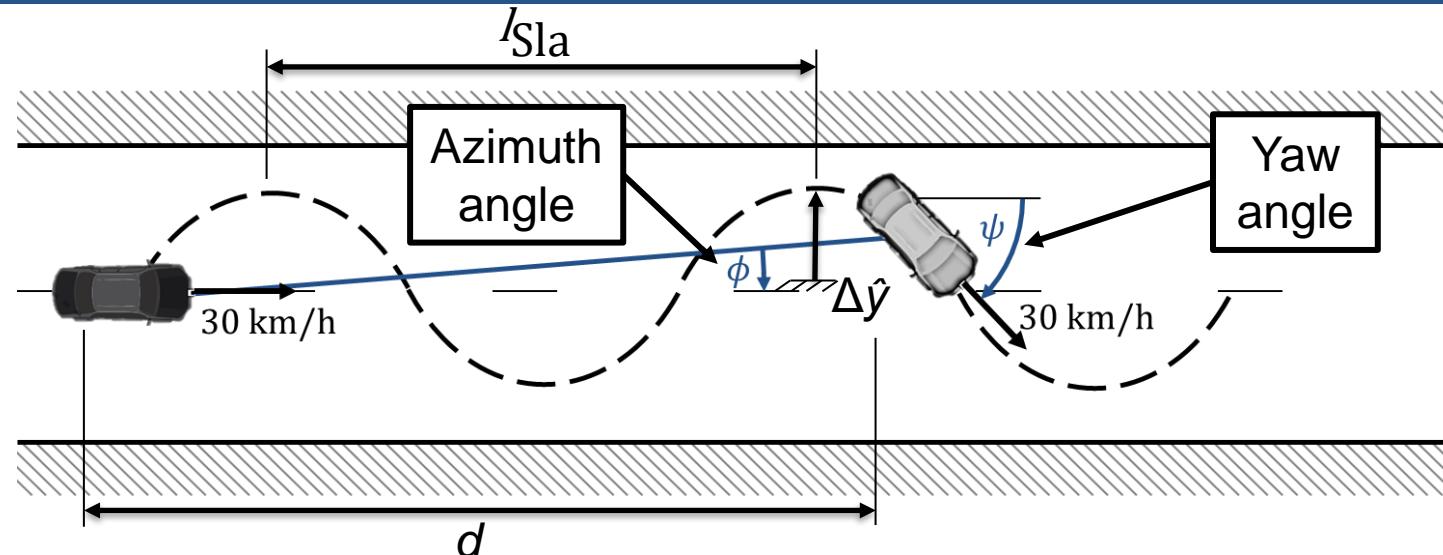
Validation Methodology of Radar Detection Data



Adapted source: M. Viehof: *Objektive Qualitätsbewertung von Fahrdynamiksimulationen durch statistische Validierung*, PhD Thesis, TU Darmstadt, Darmstadt, 2018

Experimental Setup OTA/ViL Validation

Source: M. F. Holder:
Synthetic Generation of Radar Sensor Data for Virtual Validation of Autonomous Driving,
 PhD Thesis, TU Darmstadt,
 Darmstadt, 2021



Parameter	Distance d	Amplitude $\Delta \hat{y}$	Period length I_{Sla}
Value	60 m – 80 m	5 m	71 m
Dependency to related value of interest	ϕ	ϕ	ψ

What is the effect of the slalom drive onto the RCS σ and position distribution of the radar detection data in an OTA/ViL testbed?

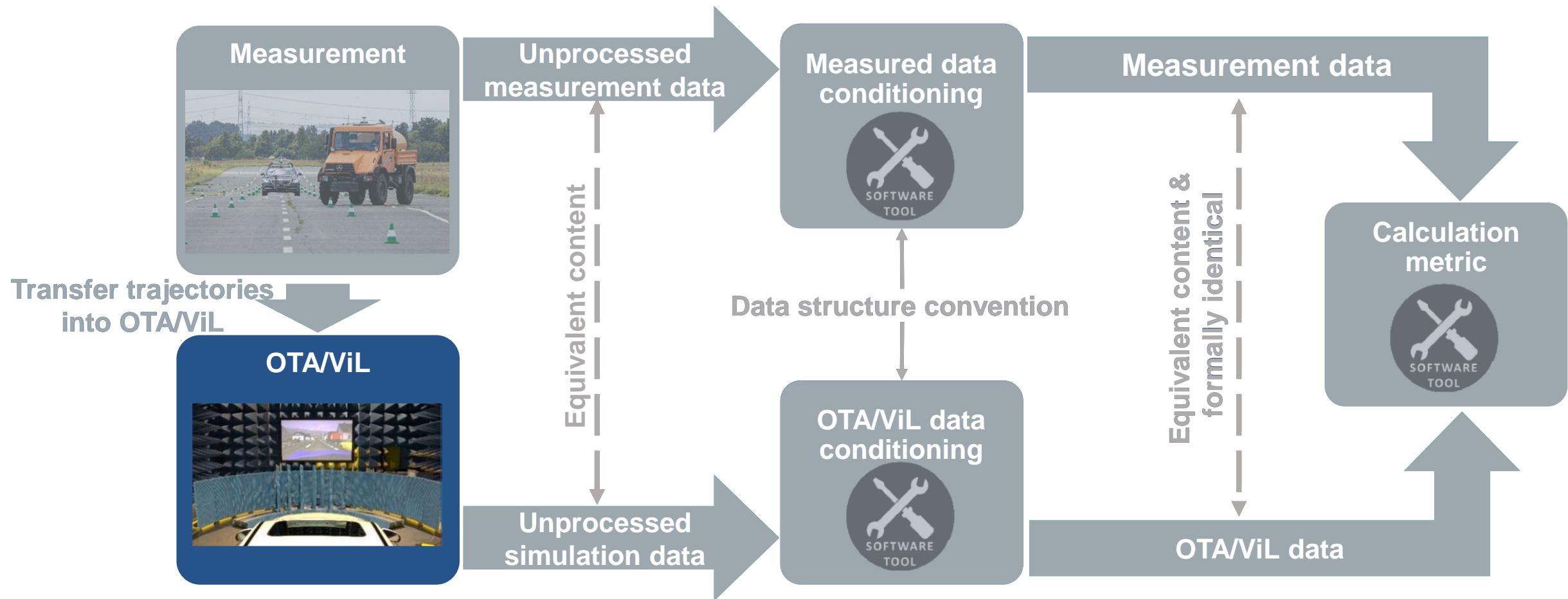
Source: M. Rapp: *Messkampagne zur Bestimmung der winkelabhängigen RCS-Profilen von Verkehrsteilnehmern*, B.Sc. Thesis, Darmstadt, 2021

Experimental Setup OTA/ViL Validation



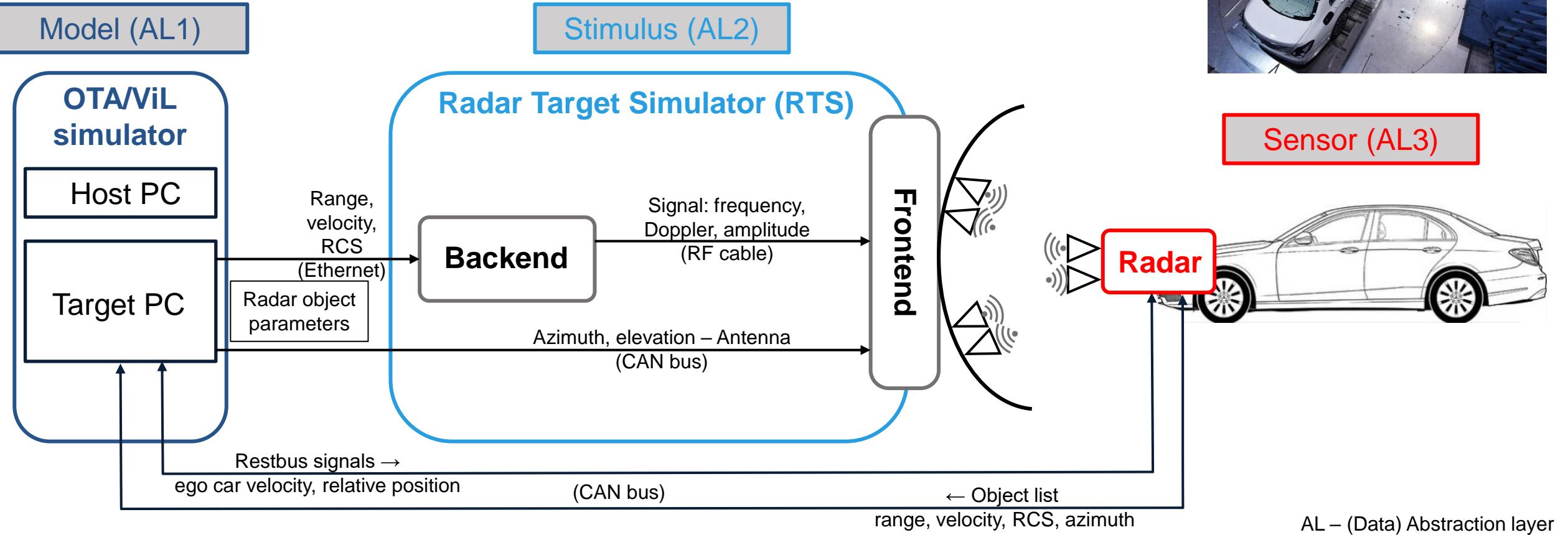
Picture: M. Holder

Validation Methodology of Radar Detection Data

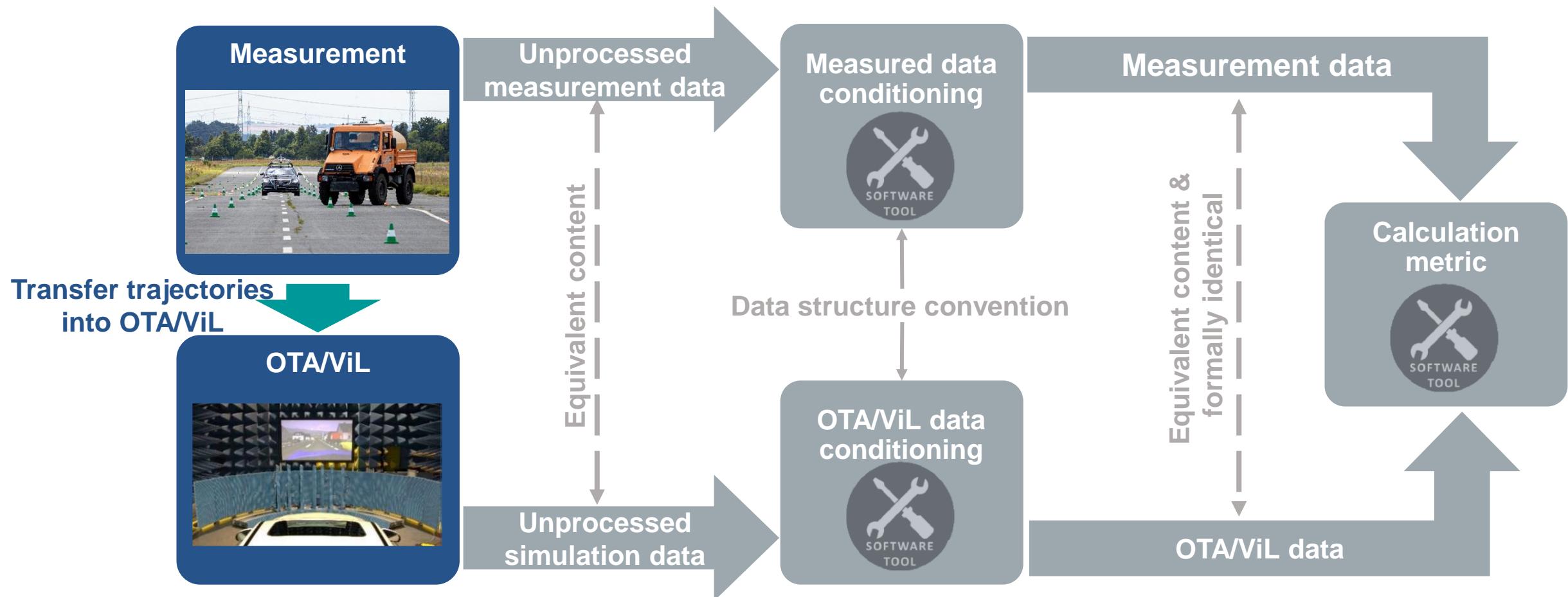


Adapted source: M. Viehof: *Objektive Qualitätsbewertung von Fahrdynamiksimulationen durch statistische Validierung*, PhD Thesis, TU Darmstadt, Darmstadt, 2018

OTA/ViL in Virtual Test and Simulation Area (ViSTA)

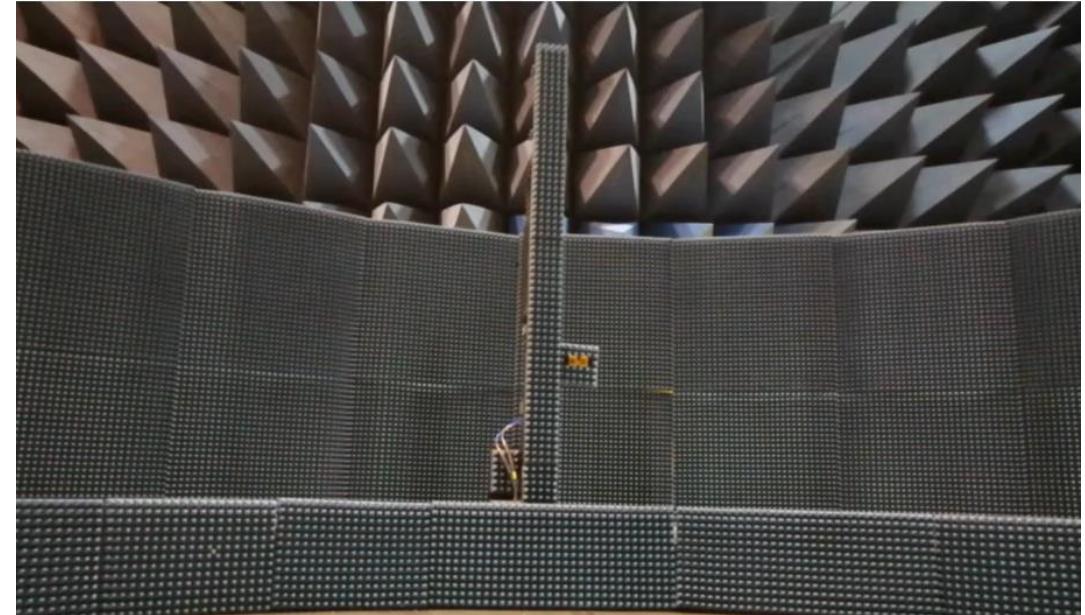
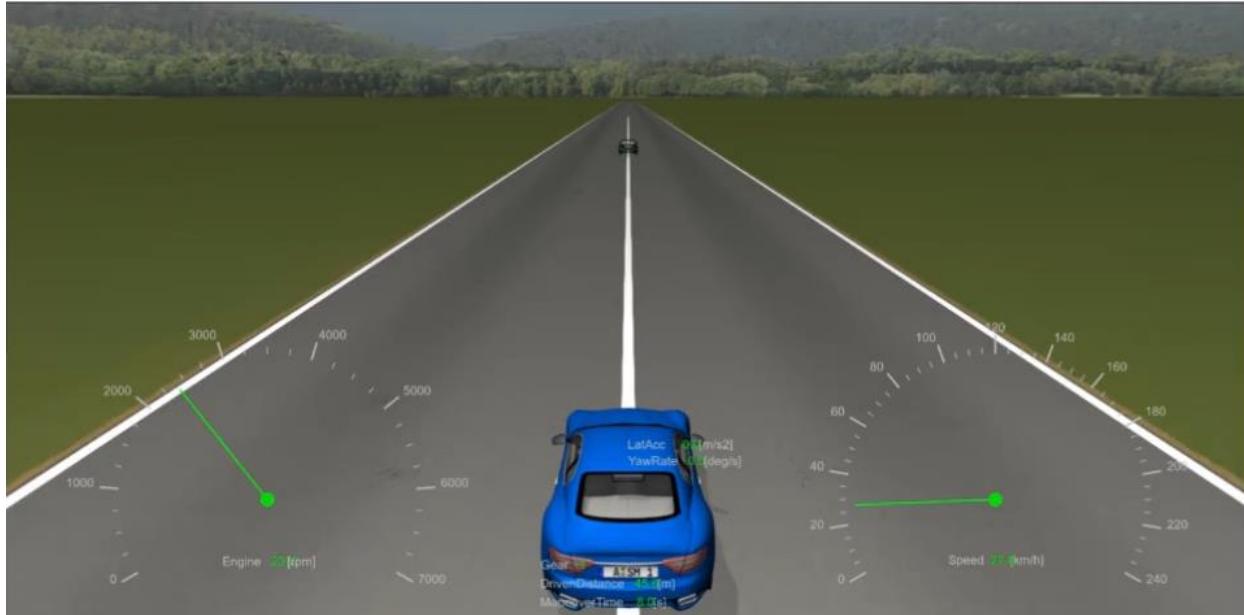


Validation Methodology of Radar Detection Data

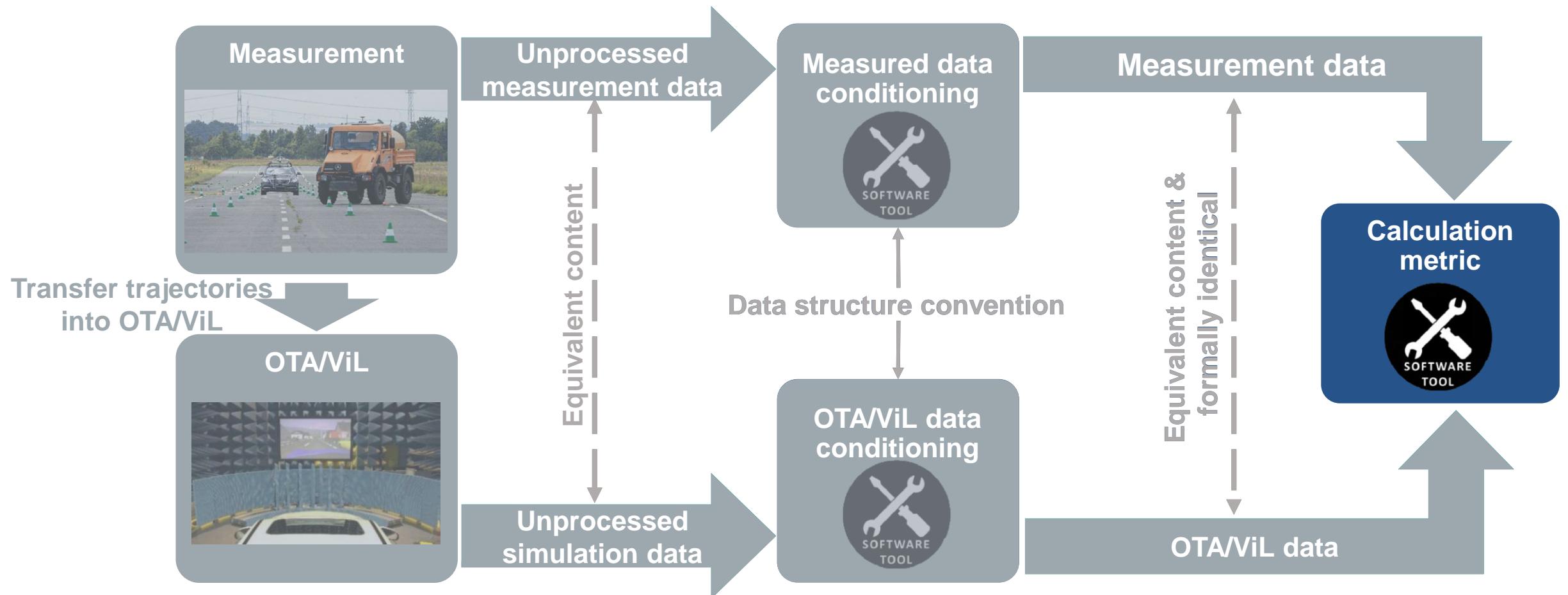


Adapted source: M. Viehof: *Objektive Qualitätsbewertung von Fahrdynamiksimulationen durch statistische Validierung*, PhD Thesis, TU Darmstadt, Darmstadt, 2018

Resimulate Slalom Drive in OTA/ViL Testbed

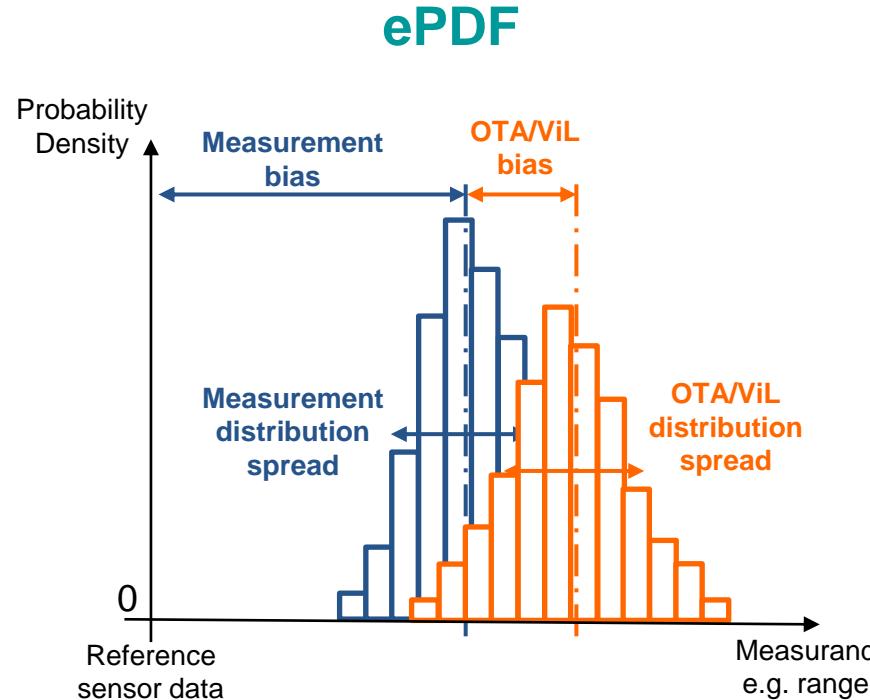


Validation Methodology of Radar Detection Data



Adapted source: M. Viehof: *Objektive Qualitätsbewertung von Fahrdynamiksimulationen durch statistische Validierung*, PhD Thesis, TU Darmstadt, Darmstadt, 2018

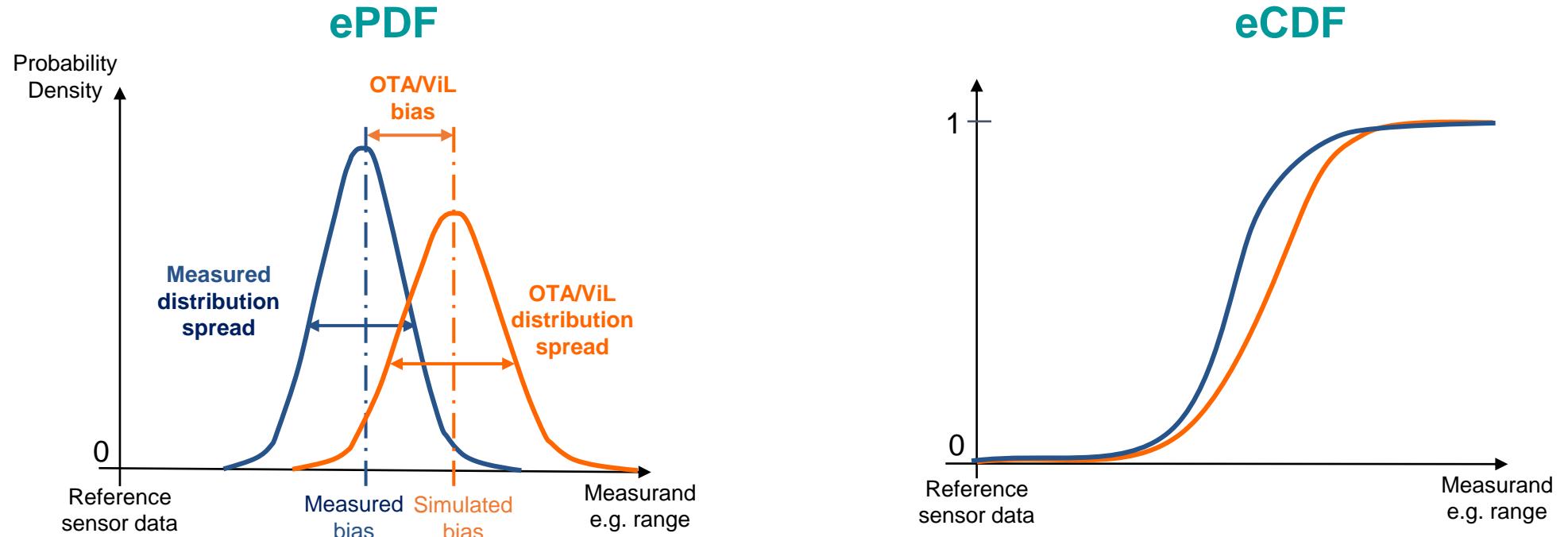
Empirical Probability Density Functions (ePDF)



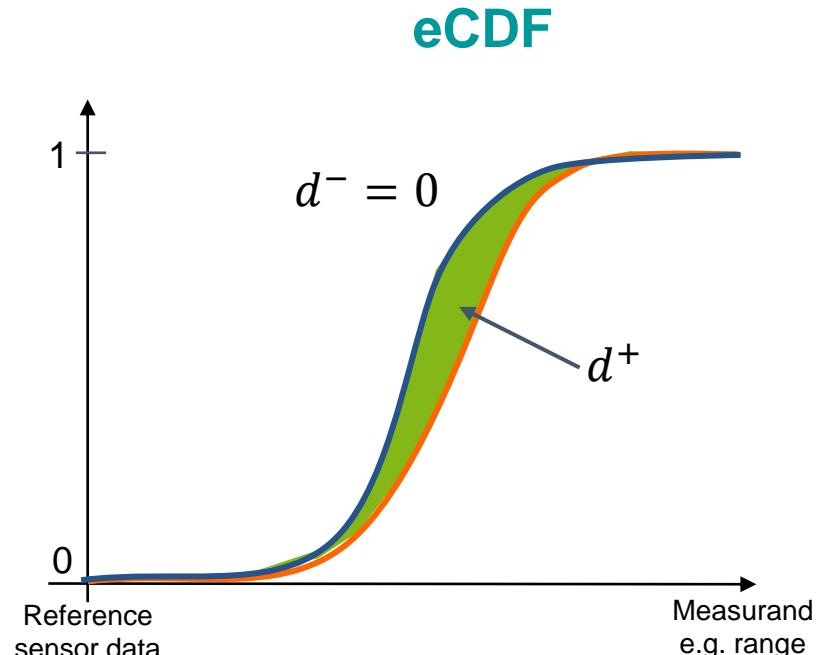
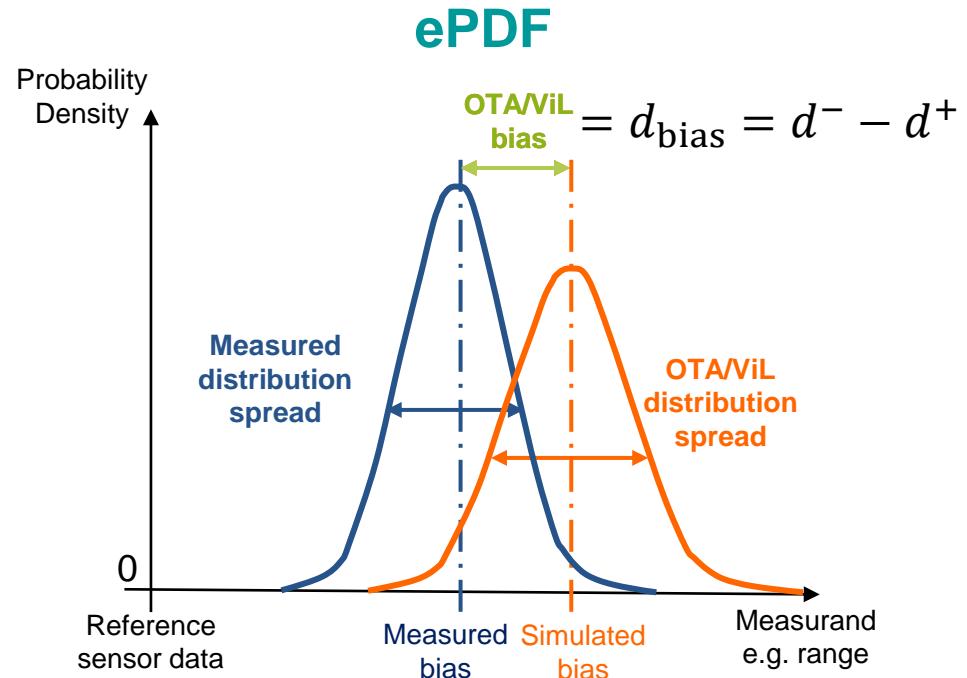
How can we compare the bias between the measured and OTA/ViL data?

How can we determine the deviation in the distribution spread between measurement and OTA/ViL?

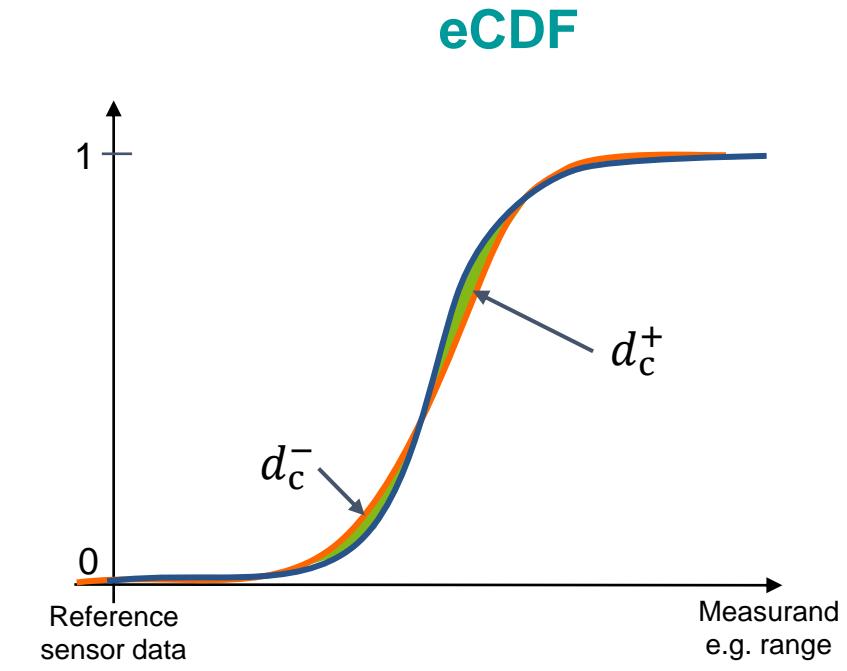
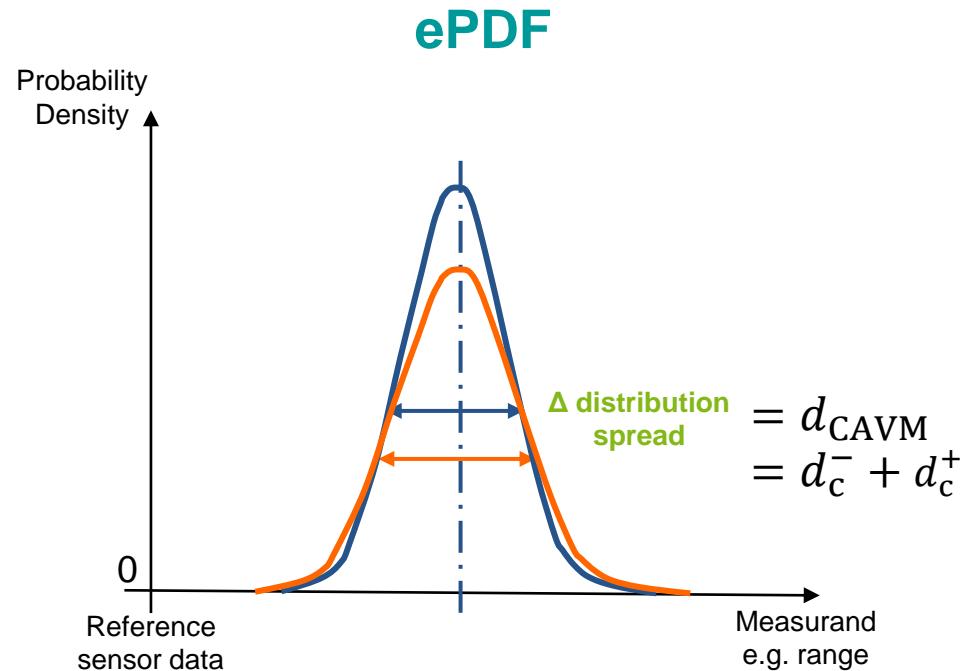
Empirical Cumulative Distribution Function (eCDF)



Determine Model Bias

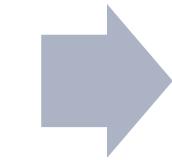


Determine Distribution Spread

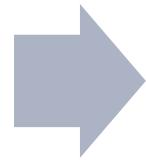


Double Validation Metric (DVM) Map

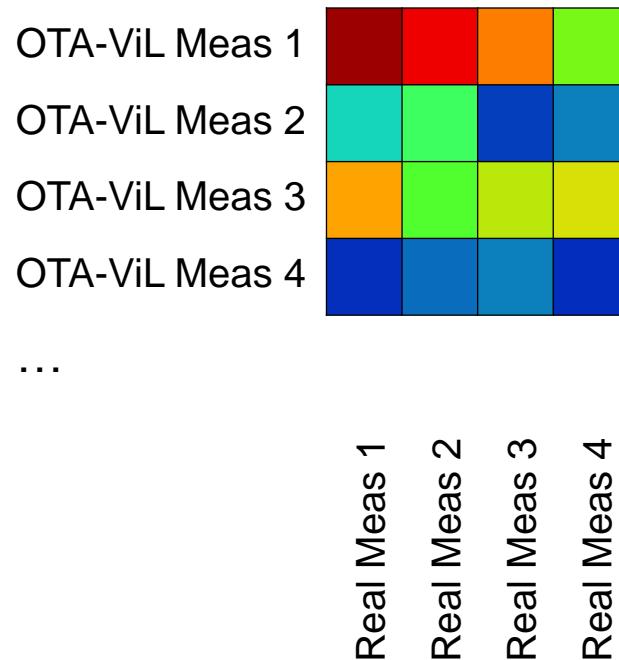
eCDF all OTA/ViL
and proving ground
measurements



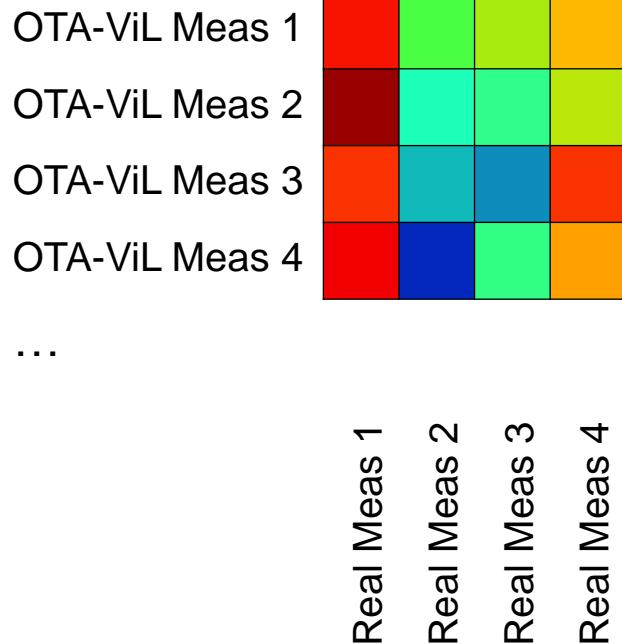
Calculate $|d_{\text{bias}}|$
and d_{CAVM}



Create DVM Map

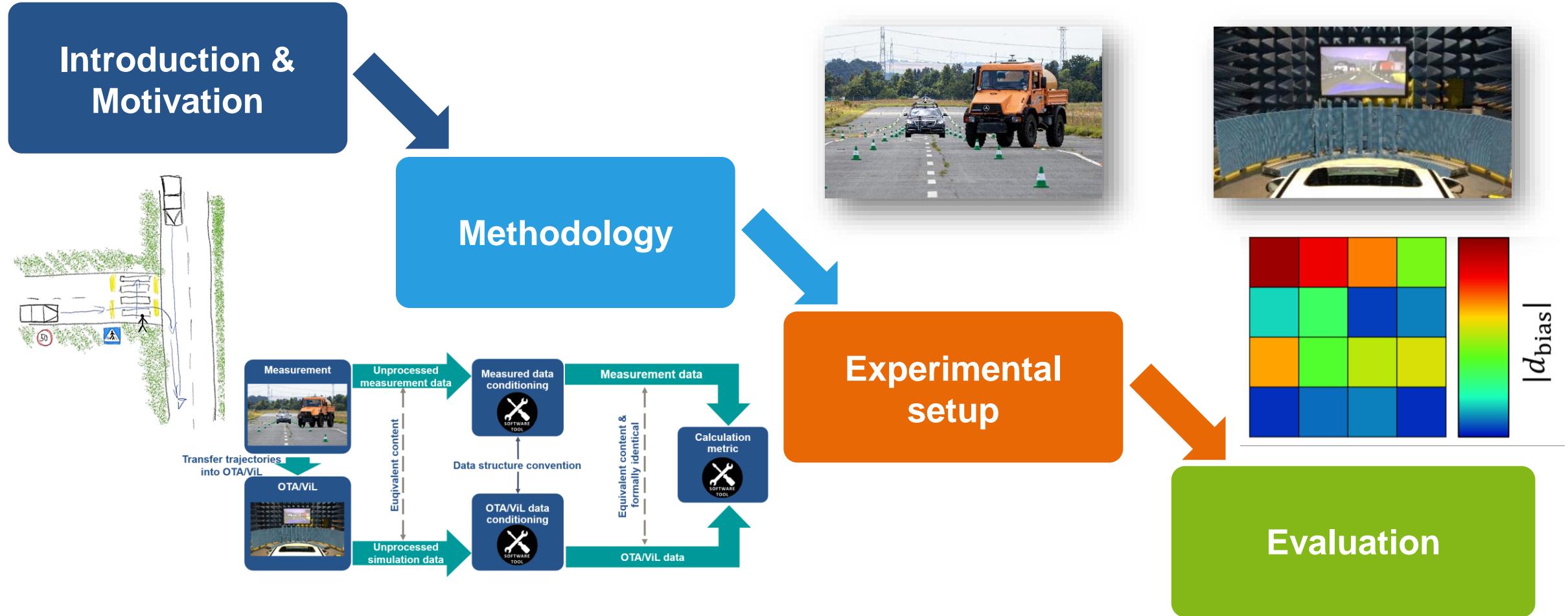


$|d_{\text{bias}}|$

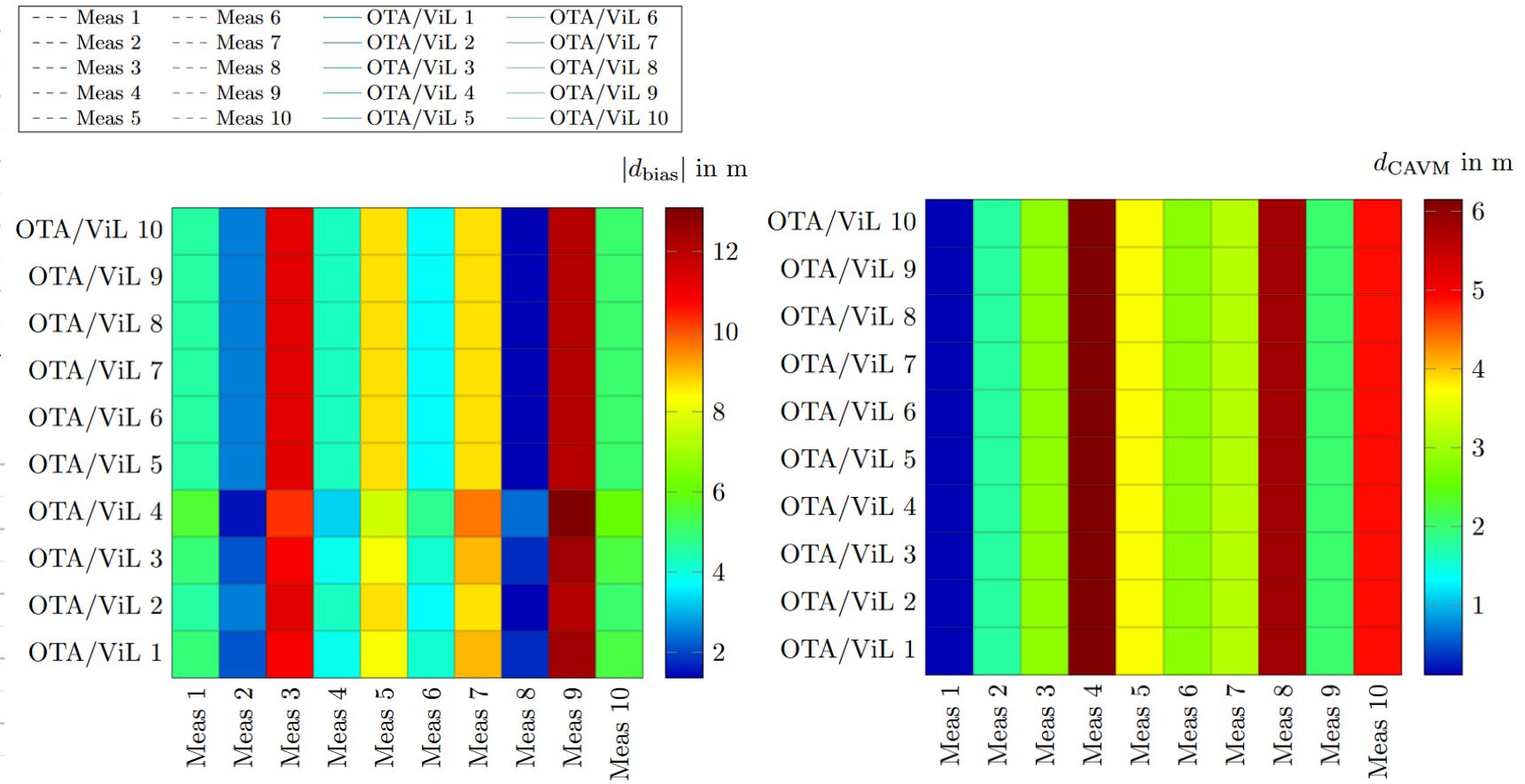
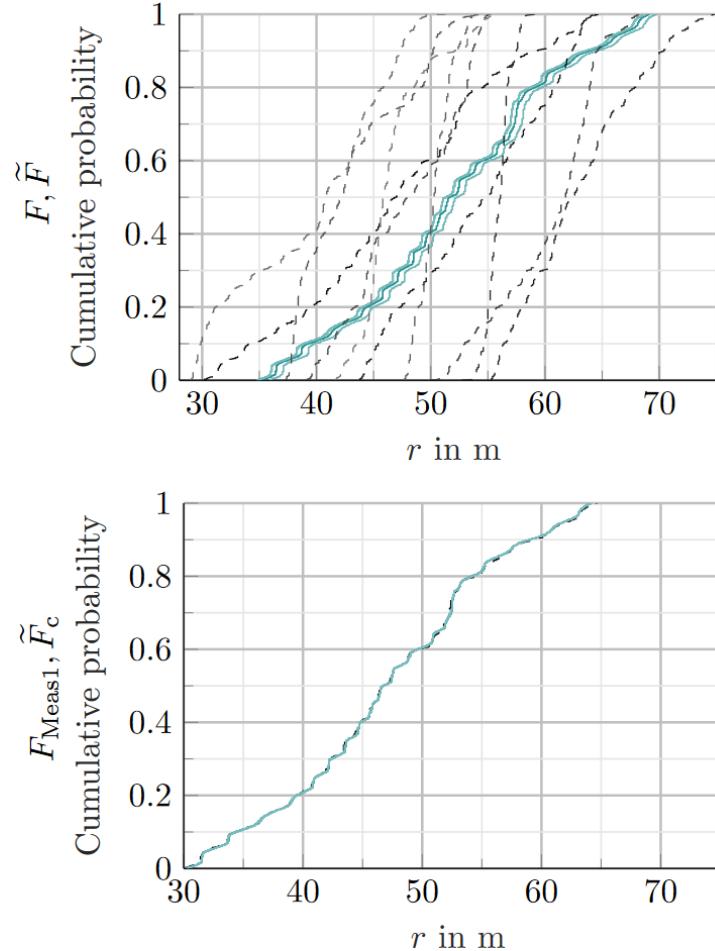


d_{CAVM}

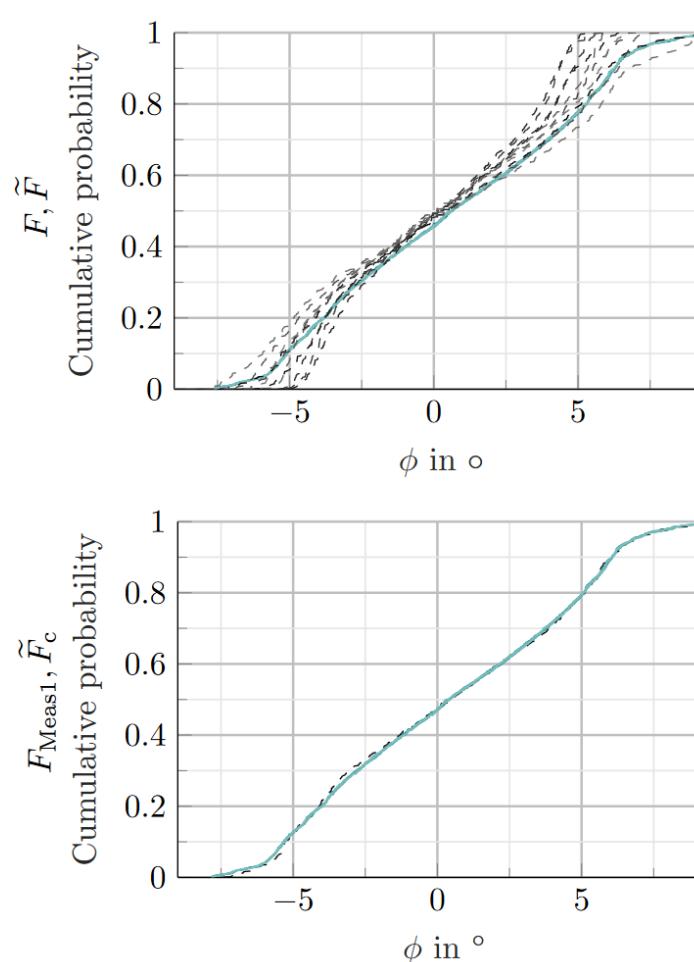
Outline



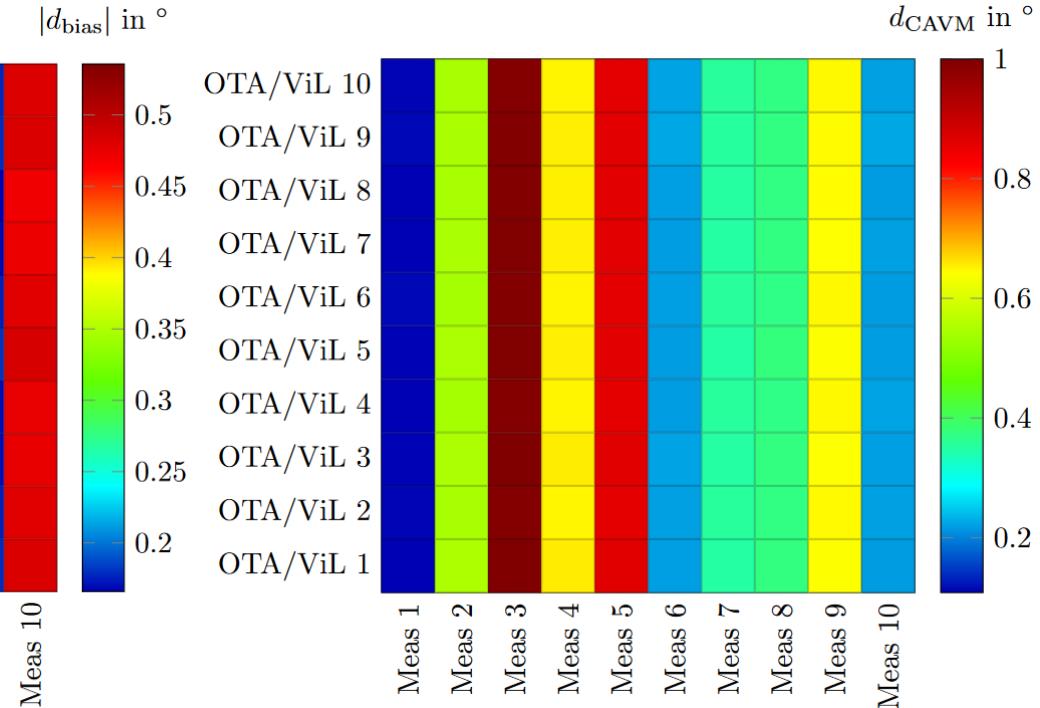
DVM Map Detection Data Range



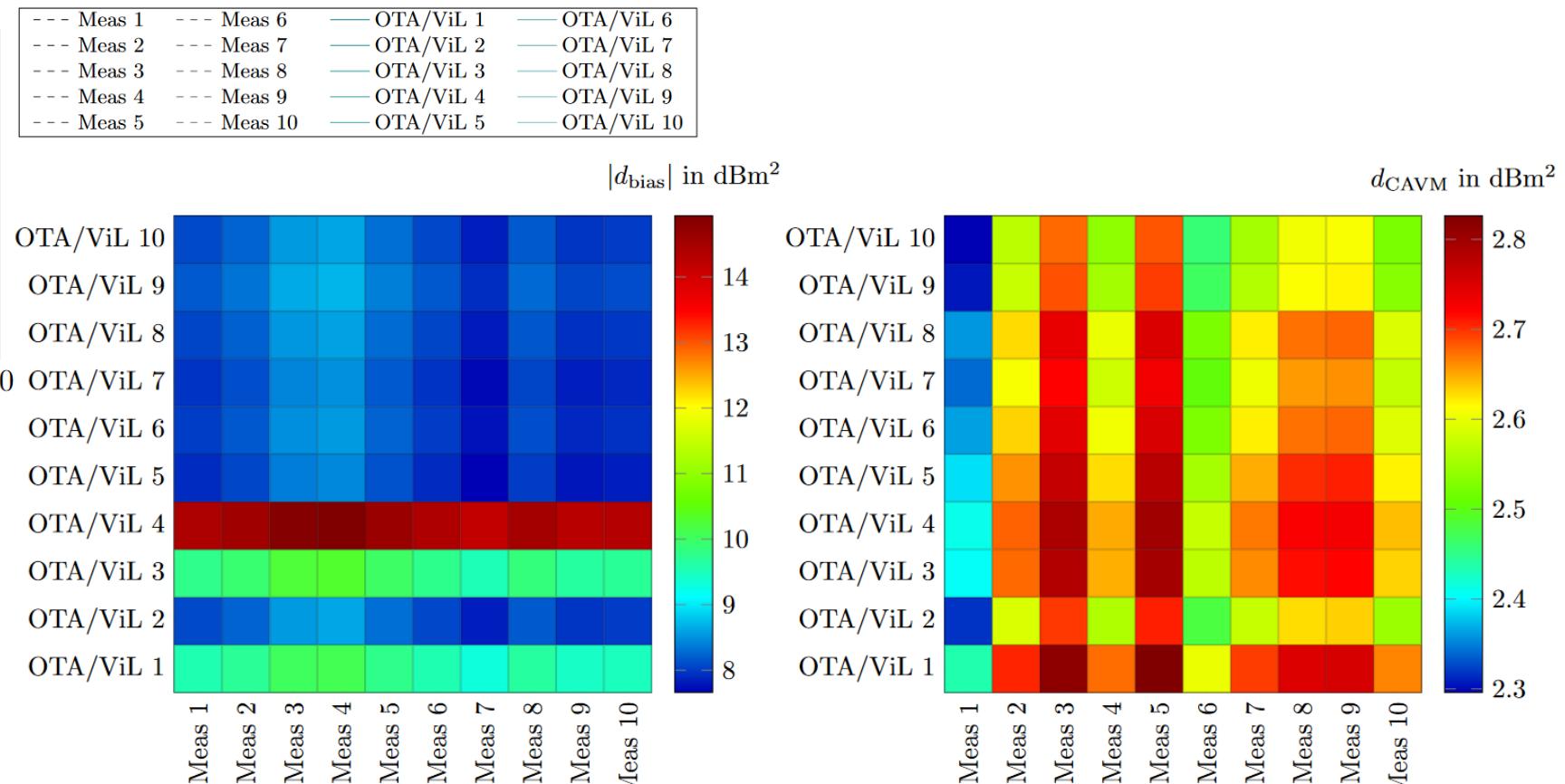
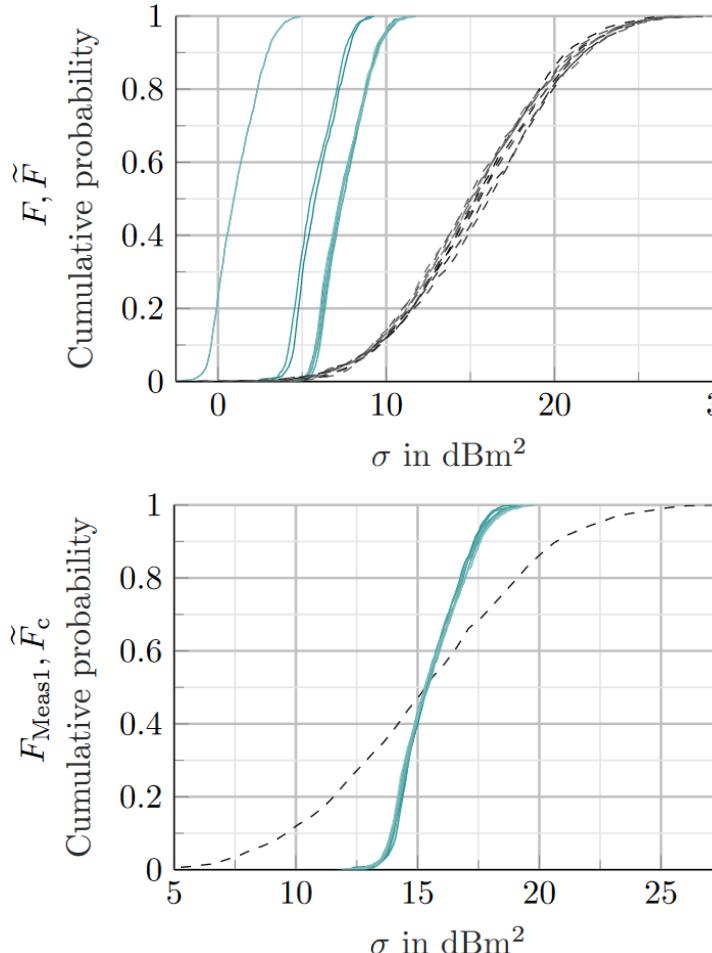
DVM Map Detection Data Azimuth



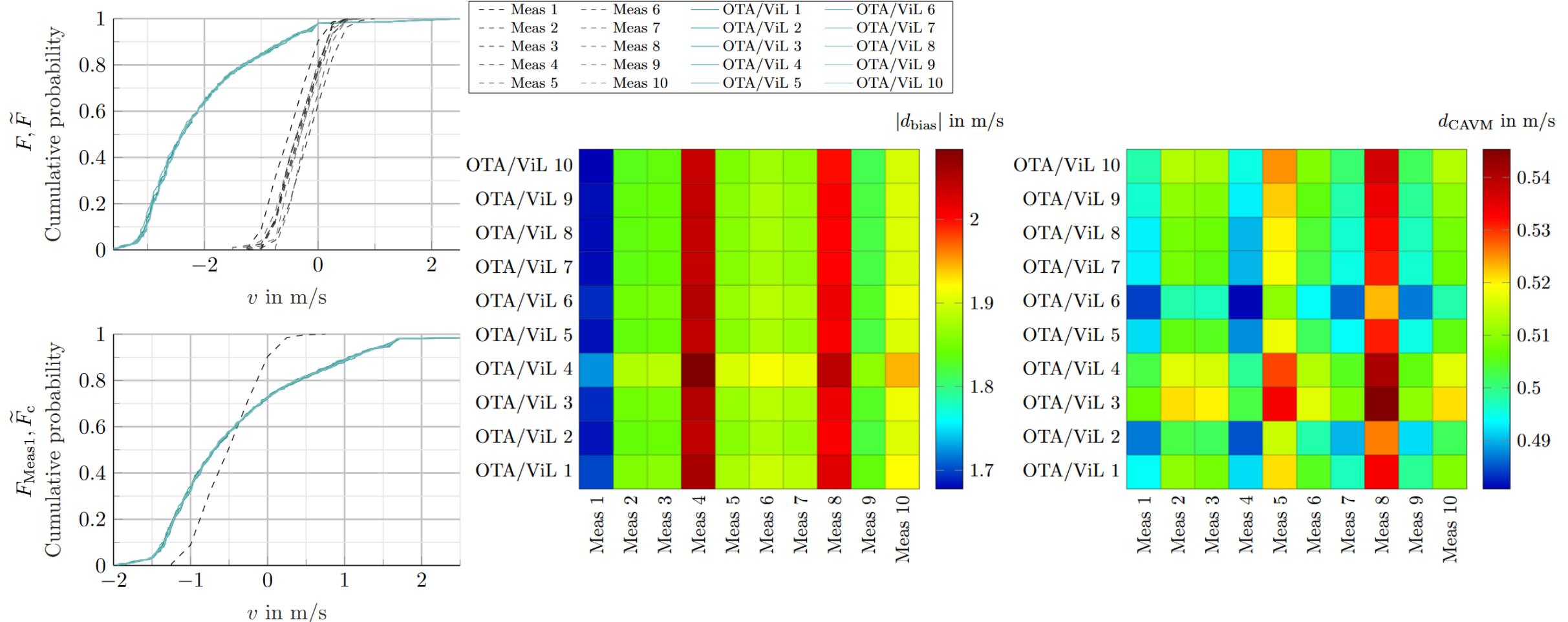
--- Meas 1	--- Meas 6	OTA/ViL 1	OTA/ViL 6
--- Meas 2	--- Meas 7	OTA/ViL 2	OTA/ViL 7
--- Meas 3	--- Meas 8	OTA/ViL 3	OTA/ViL 8
--- Meas 4	--- Meas 9	OTA/ViL 4	OTA/ViL 9
--- Meas 5	--- Meas 10	OTA/ViL 5	OTA/ViL 10



DVM Map Detection Data RCS



DVM Map Detection Data Velocity



Summary

How can we compare the bias and distribution spread between the measured and OTA/ViL data?

→ DVM Map enables the unit conform comparison of measured and OTA/ViL data

What is the effect of the slalom drive onto the RCS σ and position distribution of the radar detection data in an OTA/ViL testbed?

- OTA/ViL testbed shows high validity for RCS and azimuth
- Lower validity in case of velocity and range
- Temporal information get lost

How do we objectively measure the OTA/ViL capabilities based on measurements?

→ DVM Map is objective metric with high interpretability

Conclusion & Outlook

Conclusion

- DVM Map enables comparison of proving ground measurement and OTA/ViL data
- Valuable information about realism gap
- Interpretability given by unit conformity

Outlook

- Application on measurement data with reference data uncertainties
- Comparison of different slalom periods

