

# Computer Graphics Course Wrap-Up

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# Physically-Based Image Synthesis with Real-Time Ray Tracing

Intel Embree Wins Academy Scientific  
and Technical Award

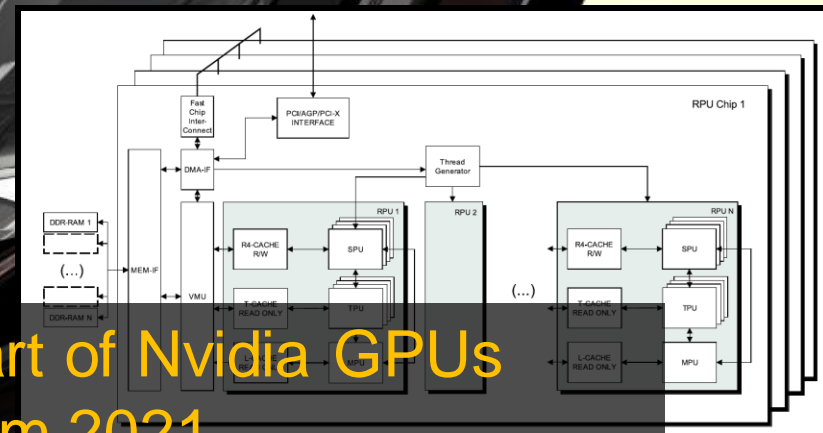
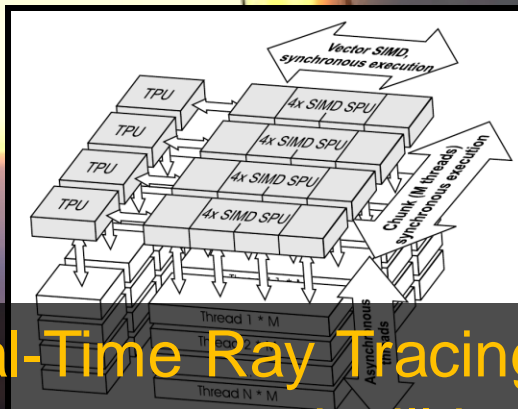
2.2.2021



"How to Train Your Dragon: The Hidden World." Property of DreamWorks Animation,  
MoonRay optimized with Intel Embree.

Key product offered now by all major HW vendors:  
e.g. Intel (Embree), Nvidia (OptiX), AMD (Radeon Rays) , ...

# Custom Ray Tracing Processor [Siggraph'05]



Real-Time Ray Tracing Hardware is part of Nvidia GPUs since 2018 and will be for all others from 2021



# Efficient Simulation of Illumination: Light Propagation and Sensor Models

VCM now part of most commercial renders:  
e.g. RenderMan, V-Ray, ...





# Recent Advances in Lighting Sim.



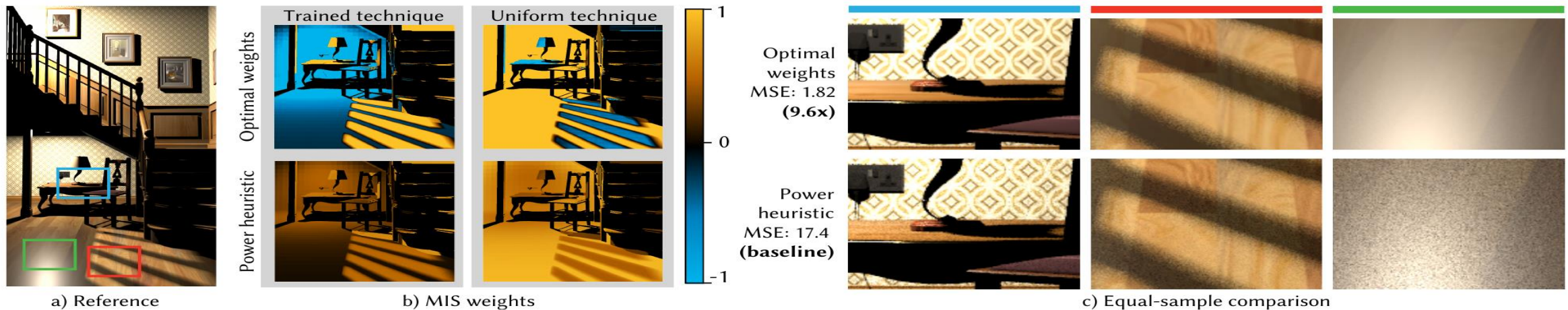
- **Light Transport Simulation with Vertex Connection and Merging (VCM)**
  - By Iliyan Georgiev et al., Siggraph 2012



# Recent Advances in Lighting Sim.



- **Optimal Multiple Importance Sampling**
  - By I. Kondapaneni, P. Vévoda, P. Grittmann, et al., Siggraph 2019





# Recent Advances in Lighting Sim.

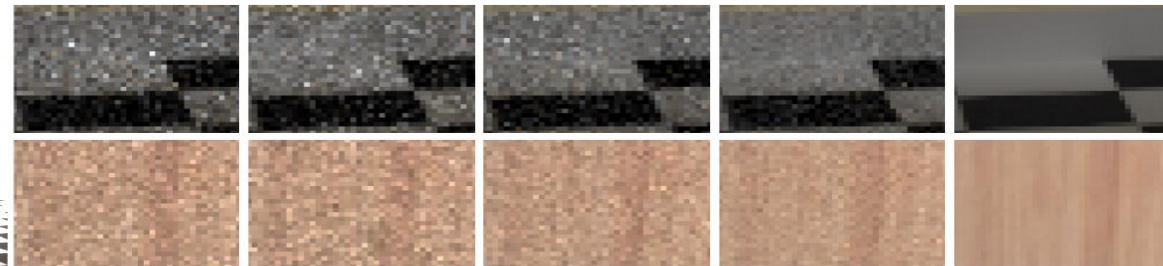


- Variance-Aware Path Guiding

- By A. Rath, P. Grittmann, S. Herholz, P. Vévoda, et al., Siggraph 2020



VEACH DOOR



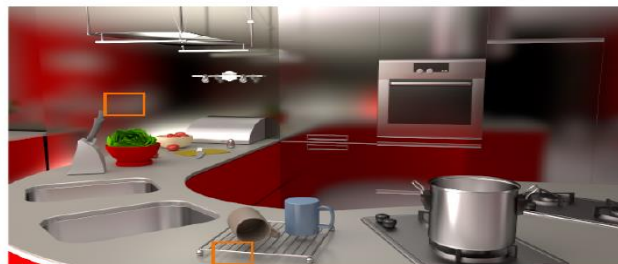
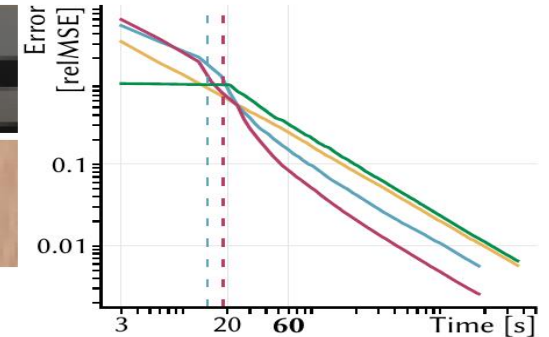
PT  
0.245 (0.6x)

VCM+MLT  
0.306 (0.5x)

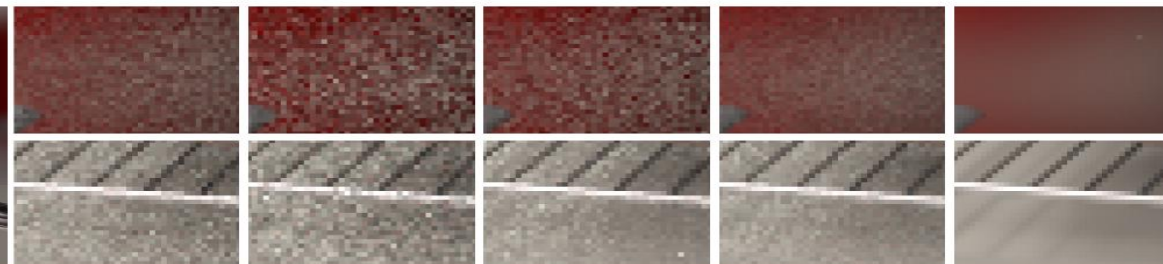
Müller et al.  
0.149 (baseline)

**Ours**  
0.084 (**1.8x**)

Reference  
*relMSE* (60s)



GLOSSY KITCHEN



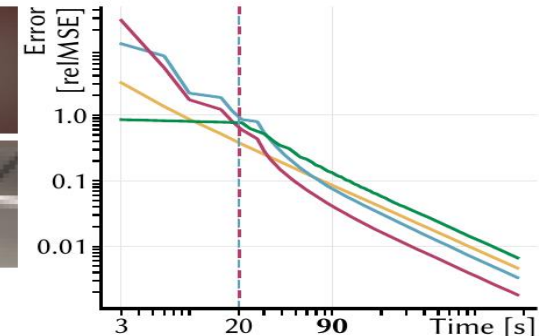
PT  
0.086 (0.9x)

VCM+MLT  
0.134 (0.6x)

Müller et al.  
0.076 (baseline)

**Ours**  
0.041 (**1.8x**)

Reference  
*relMSE* (90s)



# Challenge: Better Simulation (e.g. Radar Rendering)



- **Key Differences**

- Longer wavelength: Geometric optics (rays) not sufficient
- Need for *some* wave optics
  - Interference of multi-path interactions (coherent radiation, GO/PO)
  - Need for polarization and phase information
  - Diffraction from rough surfaces and edges
- Highly different goals
  - Optical: Focus on *diffuse* effects (+ some highlights, reflections, etc.)
  - Radar: Focus on *specular* transport only (i.e. caustic paths)

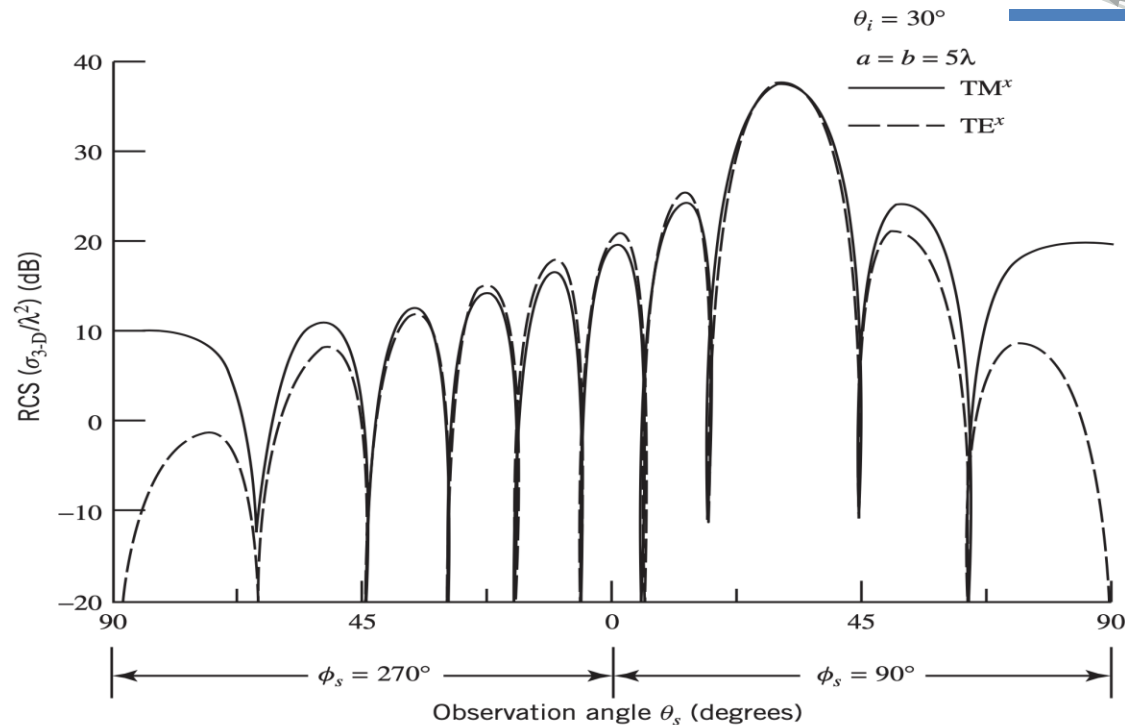
- **Completely novel approach (beyond ray tracing)**

- Using latest Monte-Carlo techniques (BiDir, MIS, VCM, ...)
- Using recent work on Path Guiding [Rath et al., Siggraph´19]

- **Bringing together radar & latest research on MC rendering**

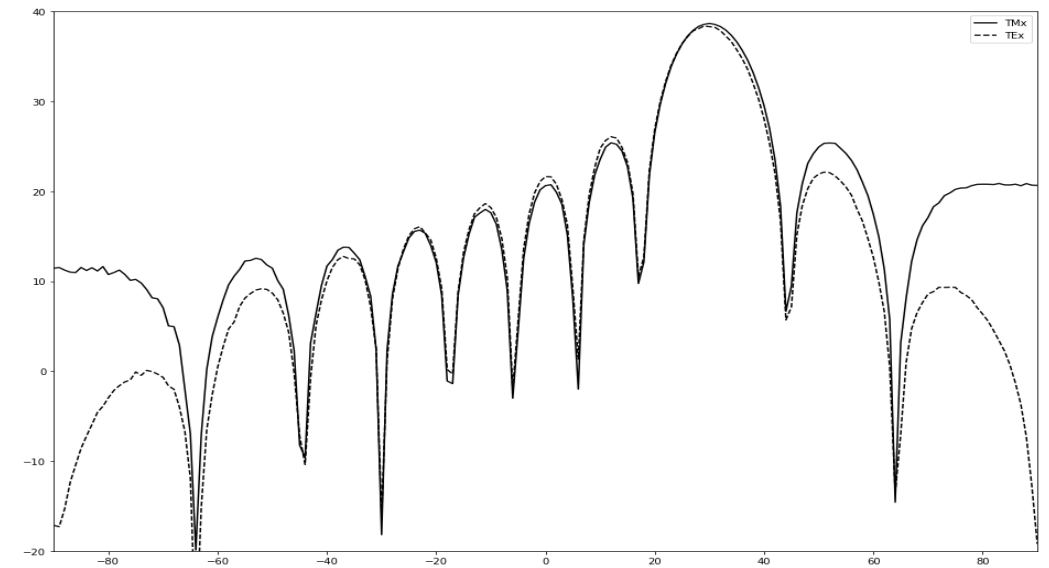


# Bi-Static Rectangle



## Analytic Solution

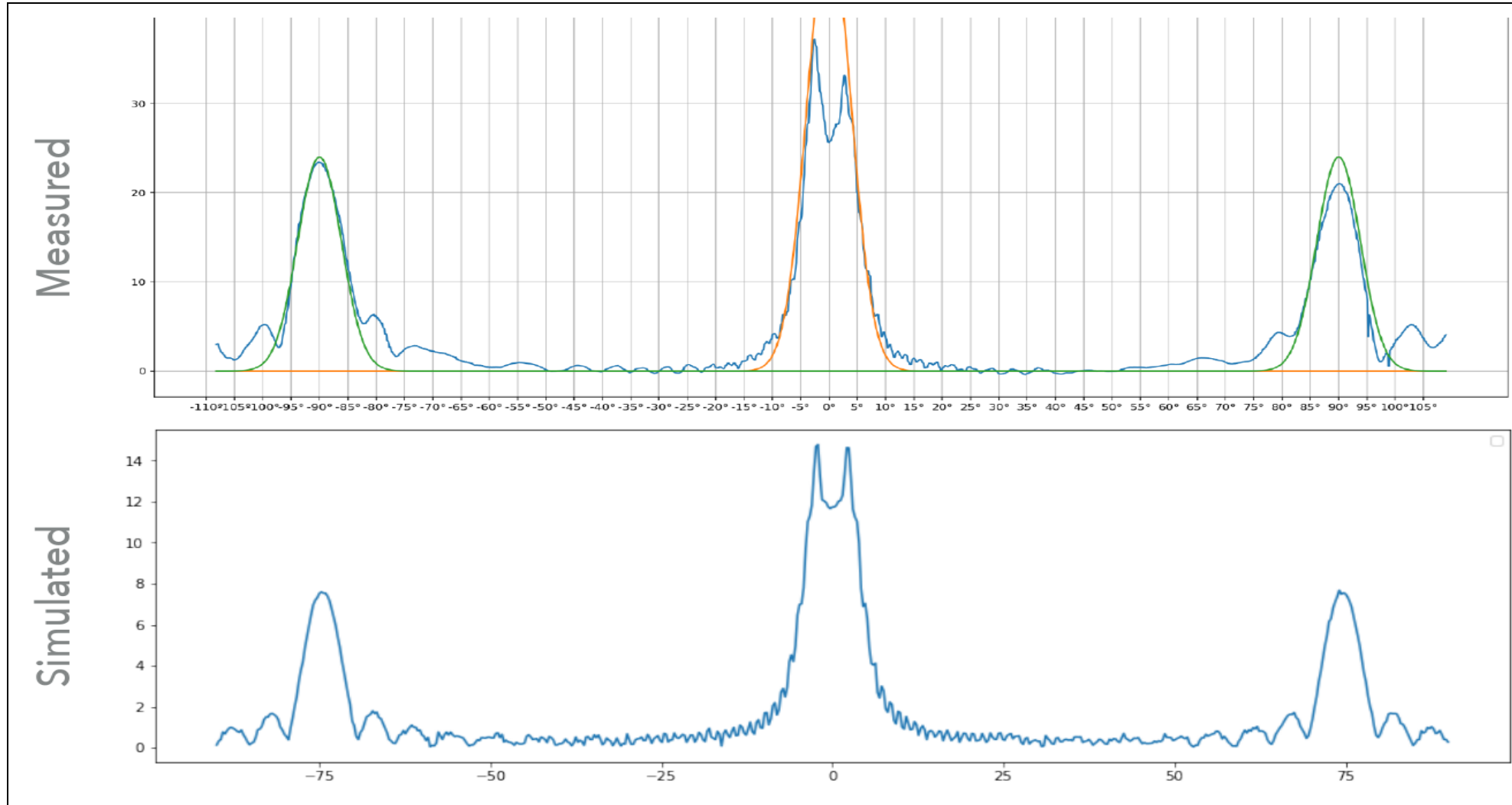
C.A. Balanis, Advanced Engineering Electromagnetics  
Chapter 11, pp. 591-599, John Wiley & Sons, 2012.



## Our Simulation

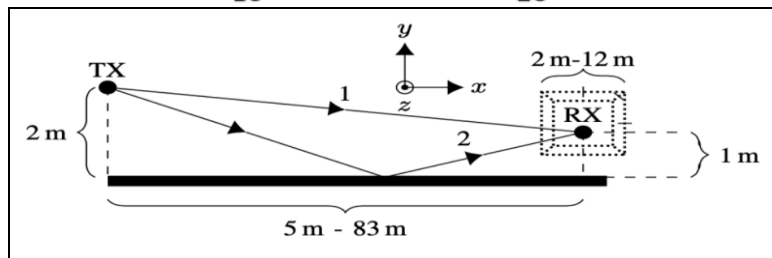
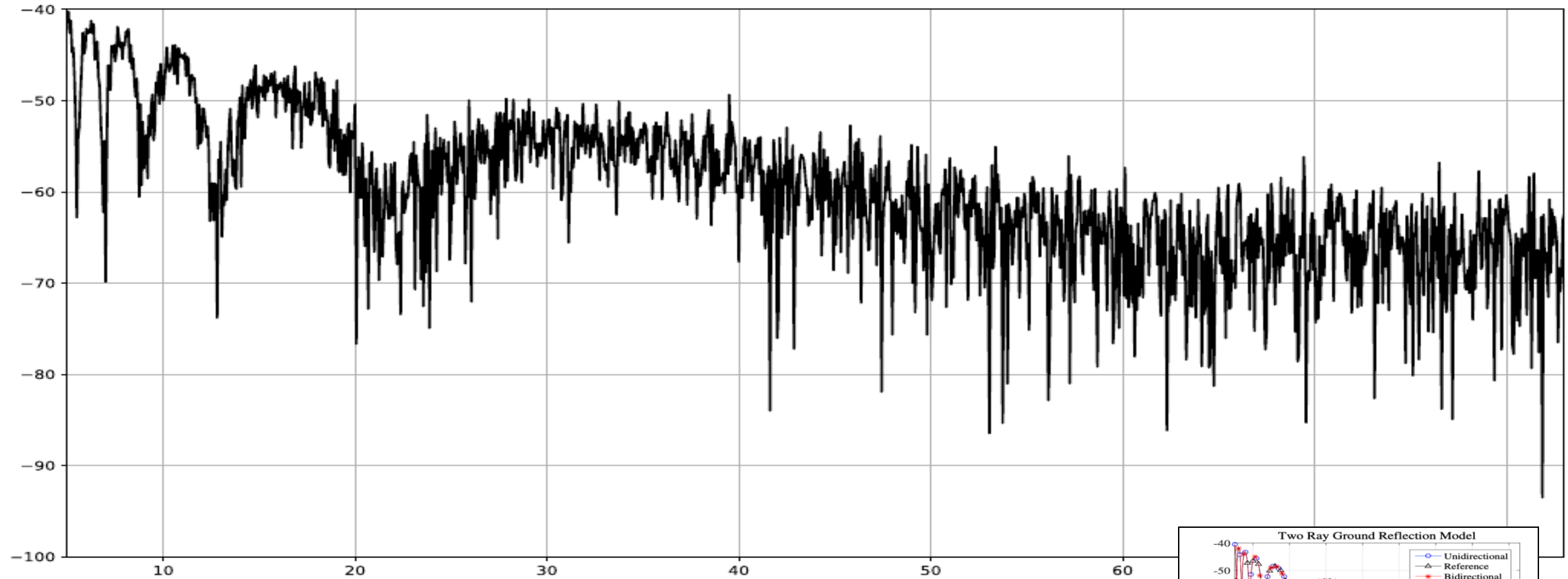
✓ perfect match

# First Results: Using Modern Monte-Carlo Algorithms

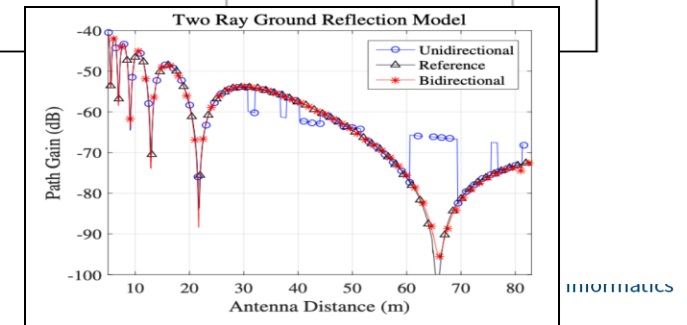




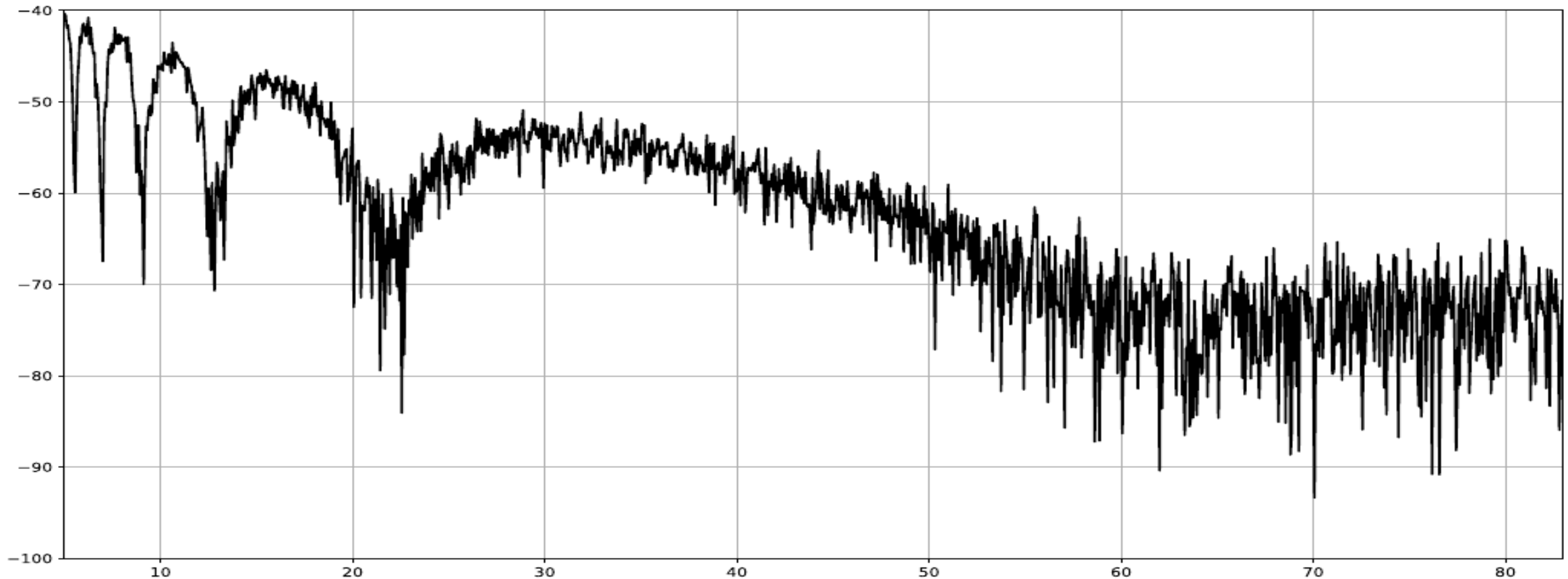
# Radar Simulation Using Modern Monte-Carlo Algorithms



Path Tracing



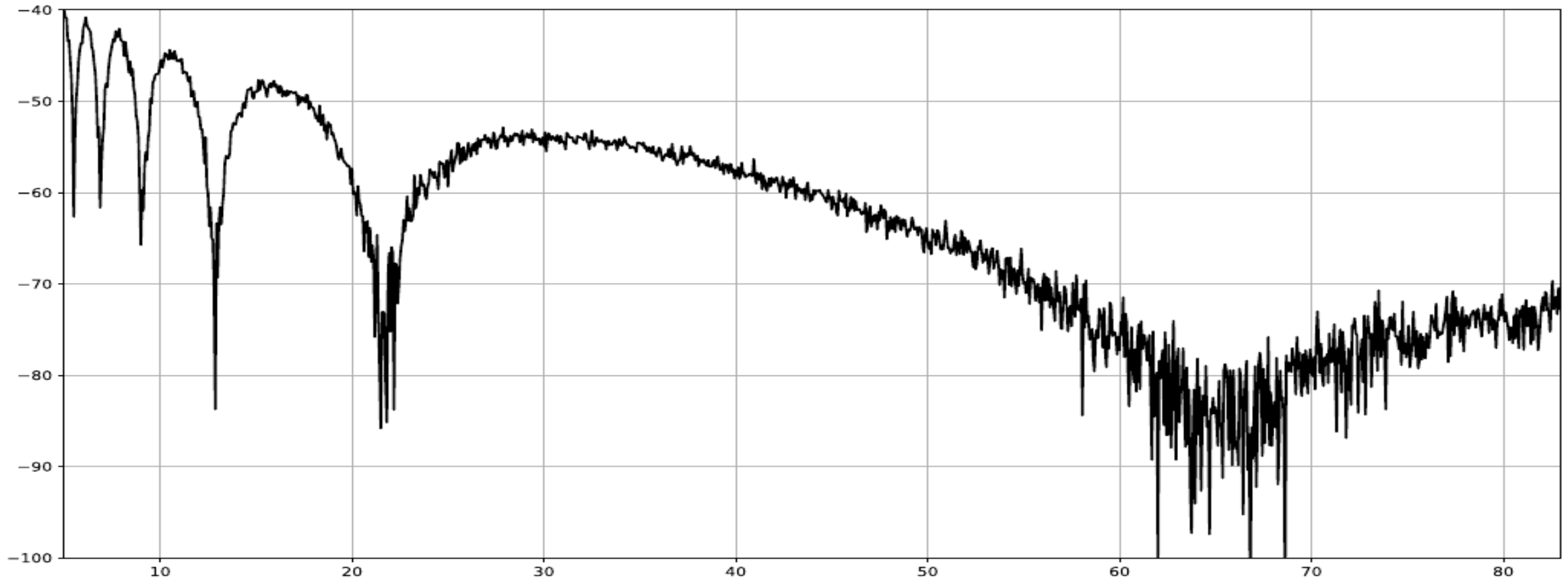
# Radar Simulation Using Modern Monte-Carlo Algorithms



Path Tracing + "Texture Filtering"

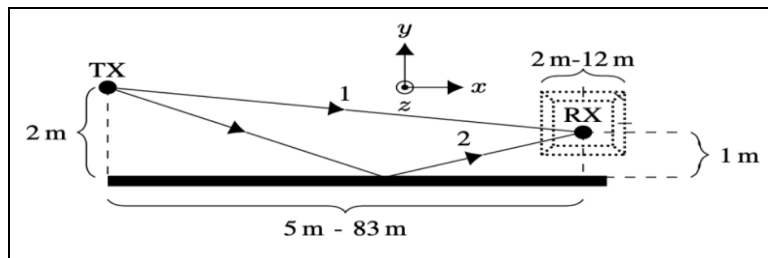
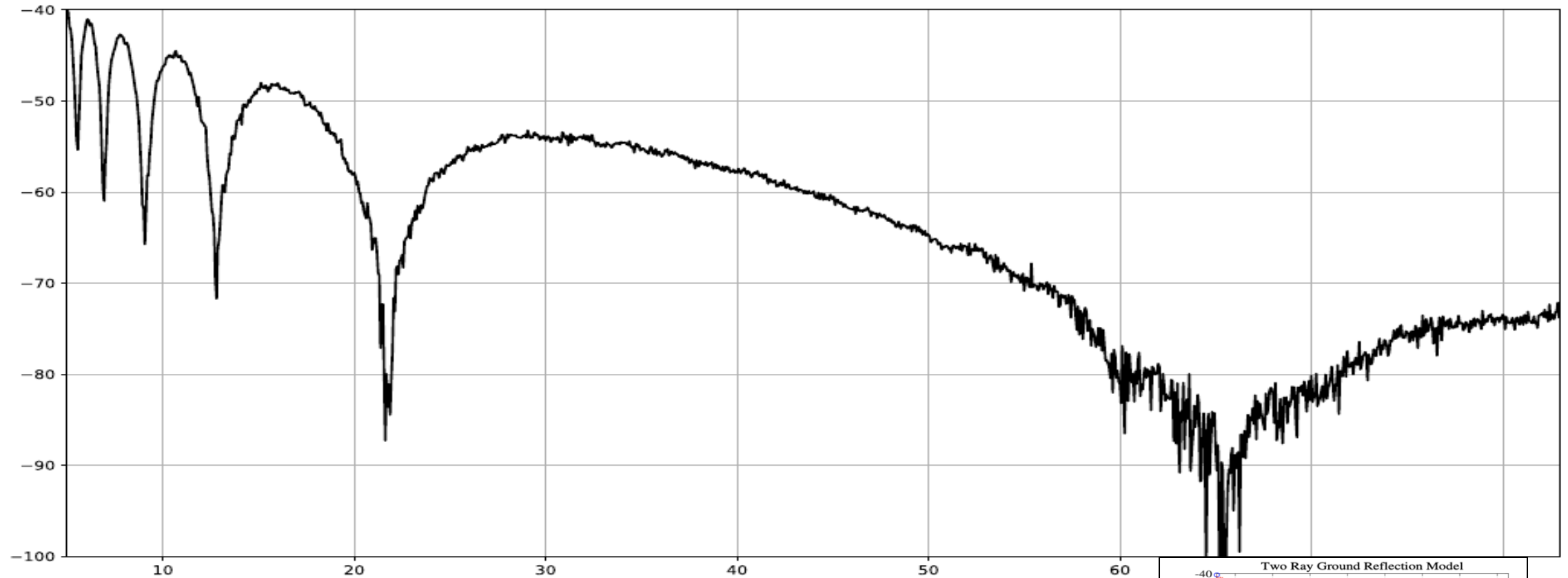


# Radar Simulation Using Modern Monte-Carlo Algorithms

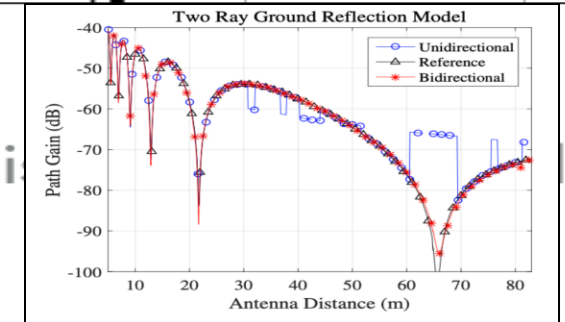


Path Tracing + "Texture Filtering" + Guiding

# Two-Way Ground Reflection

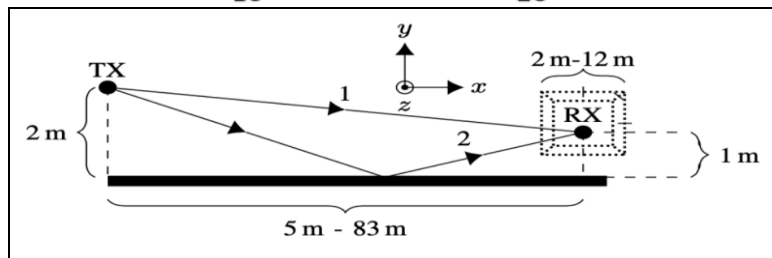
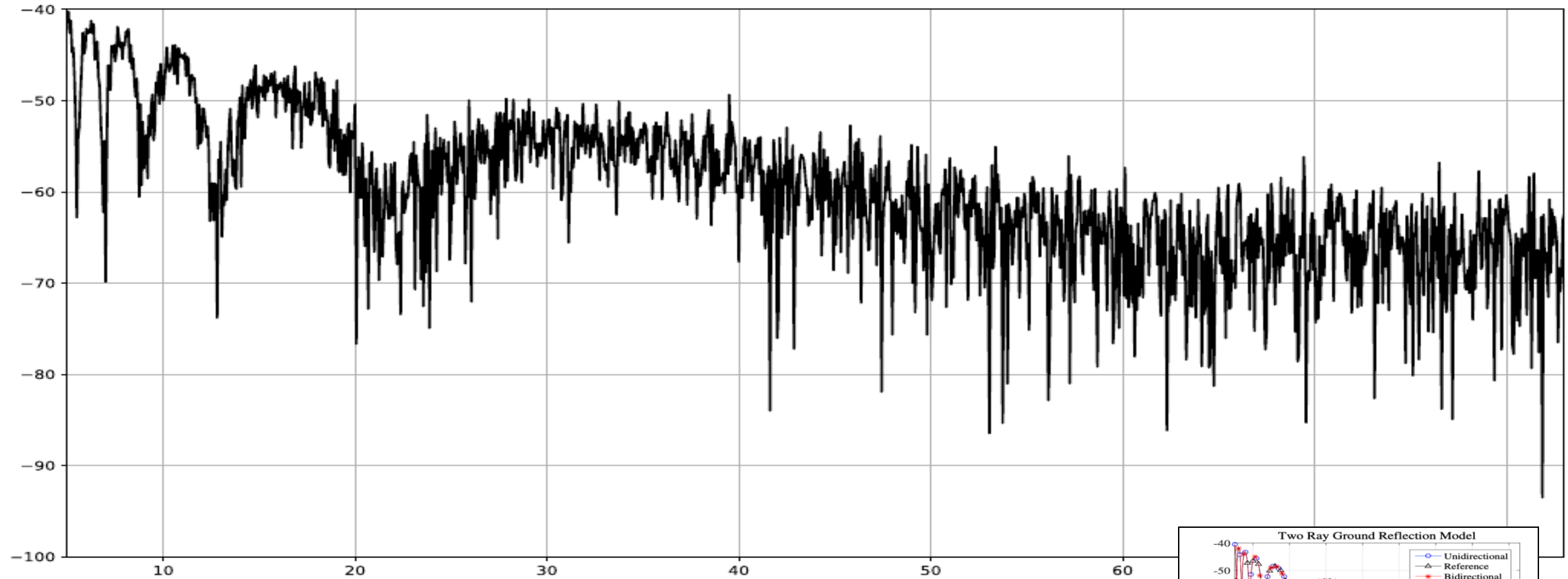


Filtering" + Guiding + Low Di ling

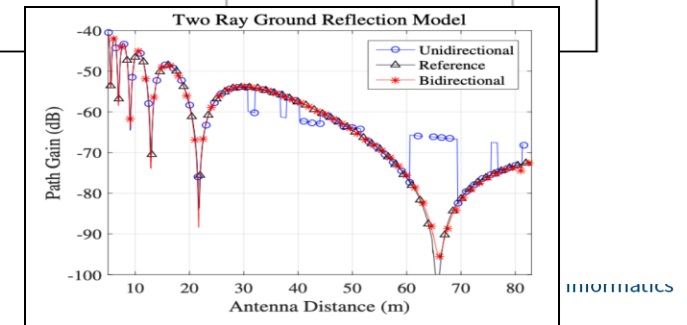




# Radar Simulation Using Modern Monte-Carlo Algorithms



Path Tracing



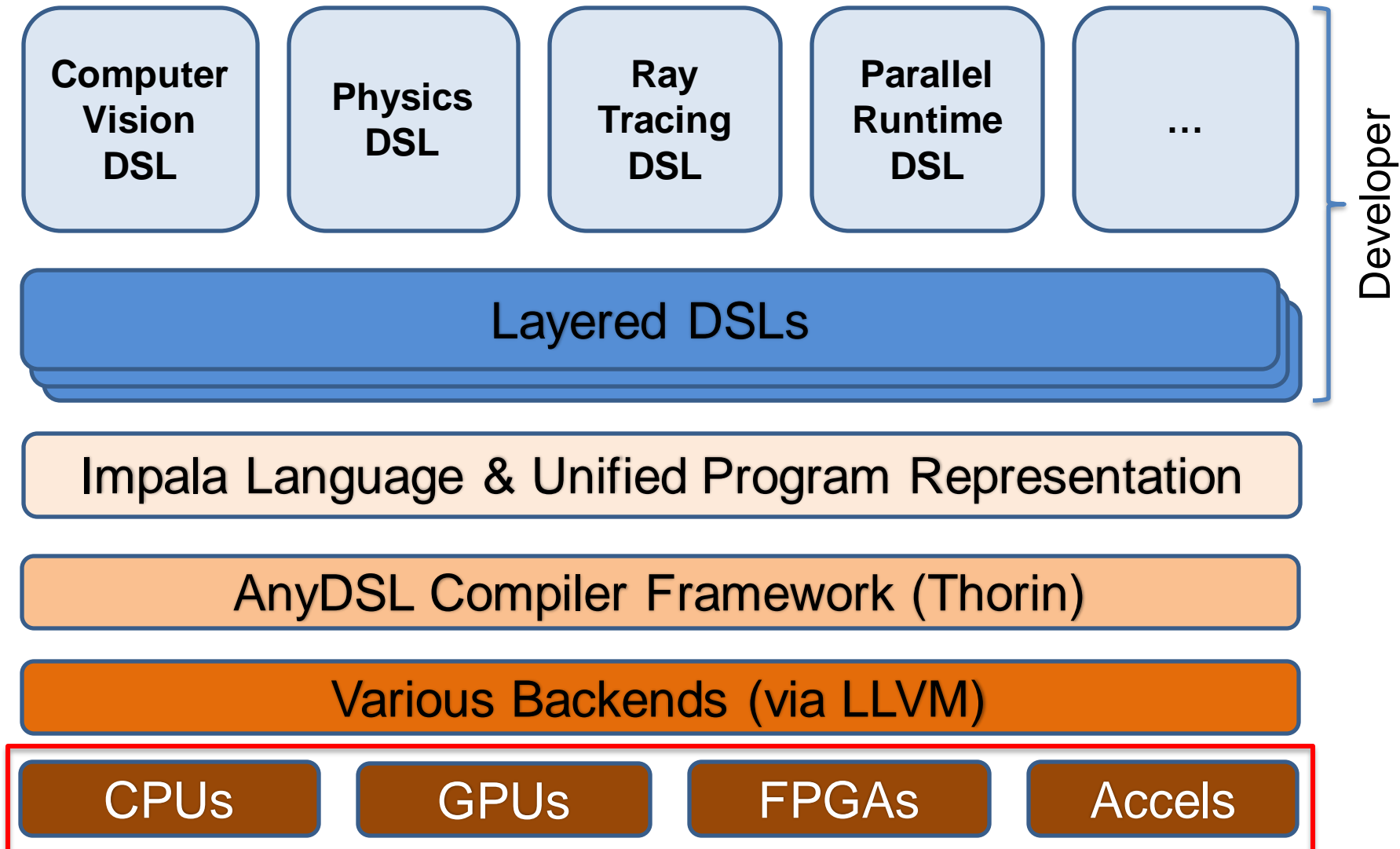
# Challenge: Do we Need a Better Basis for our Simulation?



- **In the past: Two big markets, focused on nice images**
  - *Gaming*: Very nice images (at 60+ Hz)
    - Must compromise realism for frame rate
  - *Film & Marketing*: Even nicer images (at hours per image)
    - Will compromise realism for the story and artistic expression
  - Both are being used for simulations for Autonomous Driving
- **But: Strong need for *correct* images**
  - Lidar, radar, multi-spectral, polarization, measured materials, ...
  - Need for “error bar per pixel” & validation
  - Existing engines unlikely to adapt to these fundamental changes
- **Towards “Predictive Rendering” engine**
  - Focused on physical accuracy (“sensor realistic”) & high throughput
  - Based on latest graphics research results (and GPU-HW)



# AnyDSL Compiler Framework



# AnyDSL Compiler Framework



- **Rodent: Generating Renderers without Writing a Generator**
  - By A. Perard-Gayot, R. Membarth, R. Leissa, S. Hack, P. Slusallek, Siggraph 2019



(a) Living Room



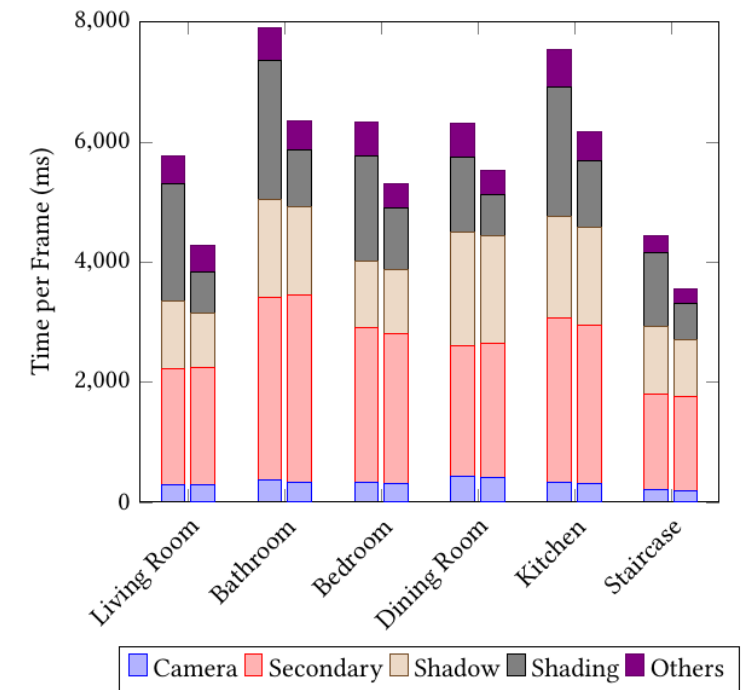
(b) Bathroom



(c) Bedroom



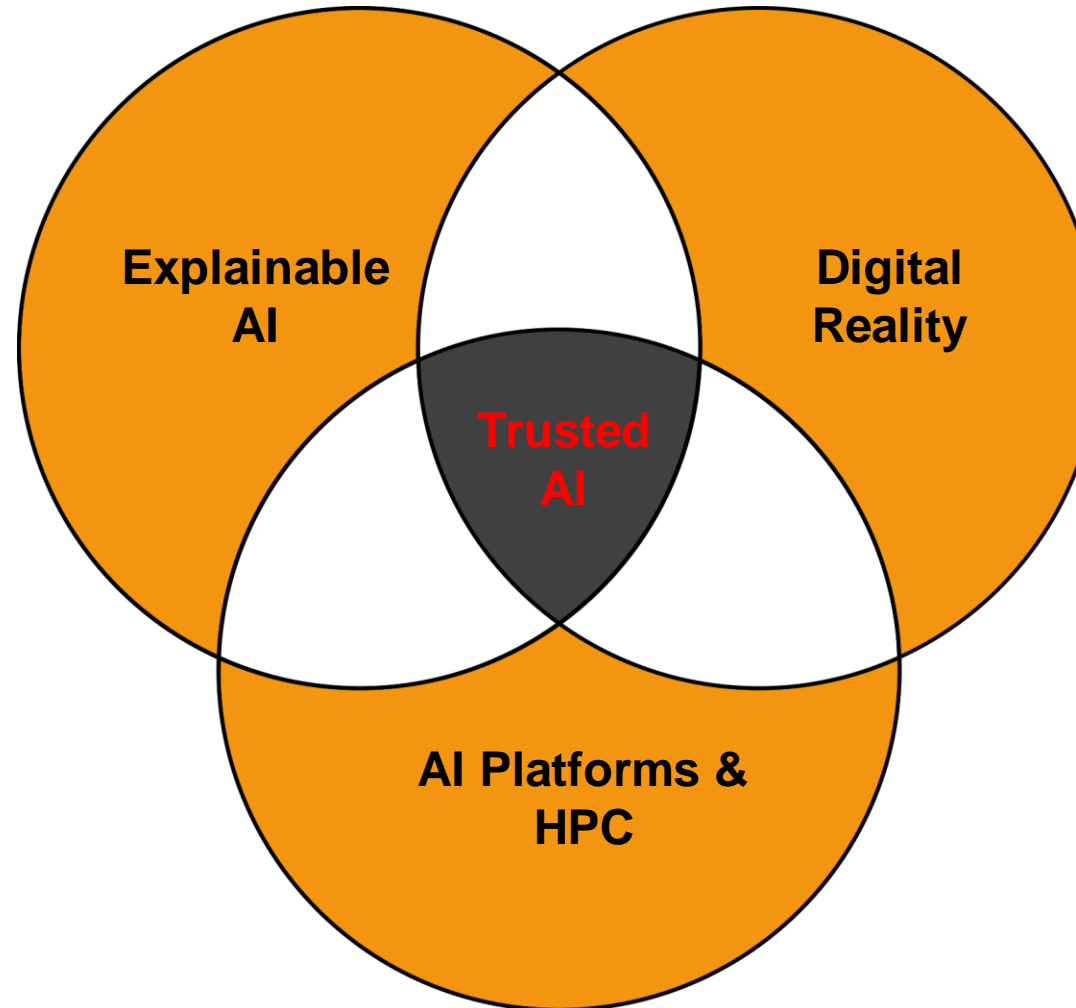
(d) Dining Room



# DFKI-ASR: Agents and Simulated Reality



**How to design *AI* systems  
that can provide guarantees  
and that *humans* can  
understand and trust?**



**How can *synthetic data* from  
parametric models and  
simulations be used for  
training, validating, and  
certifying AI systems?**

**How can AI-systems be realized technically  
in a *reliable and efficient way*?**

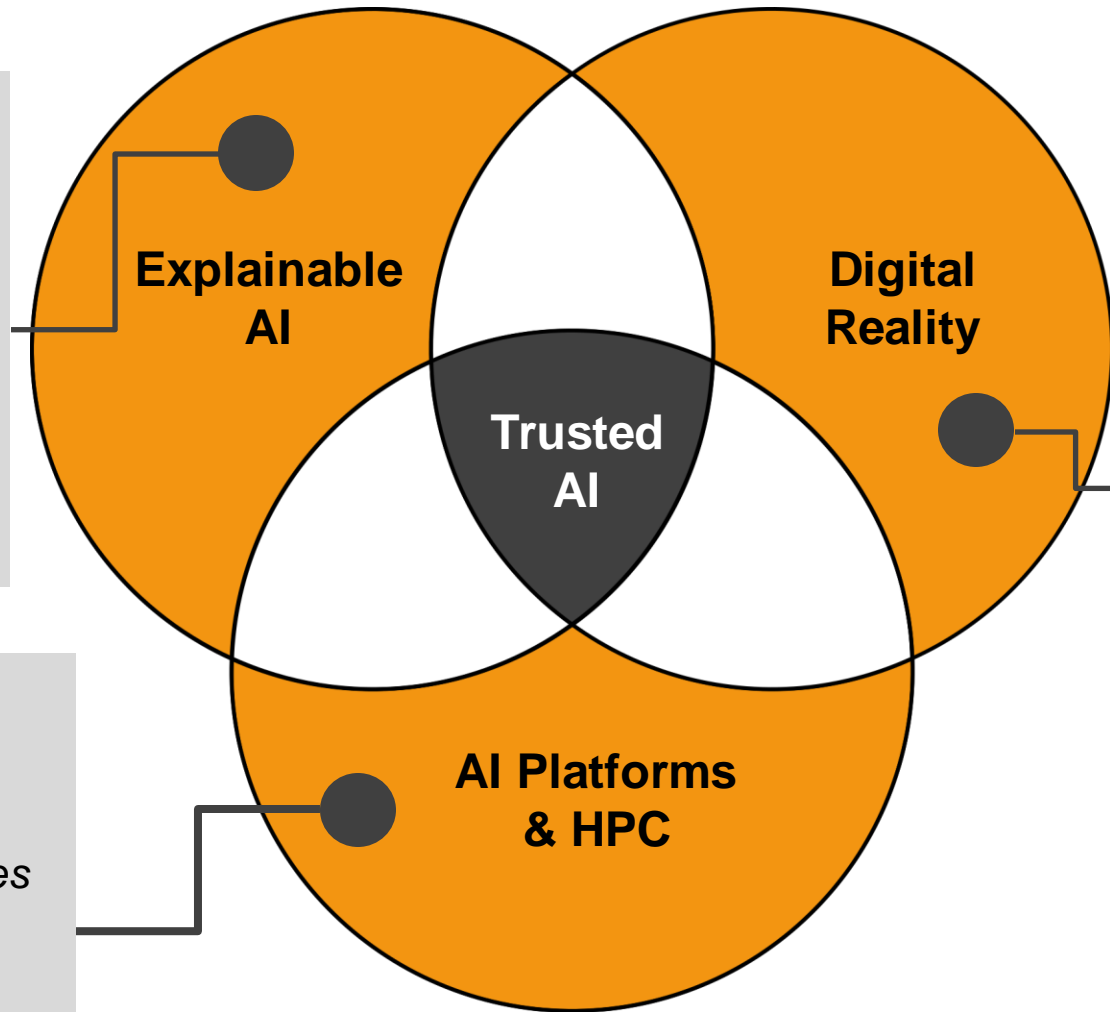


# ASR: Agents and Simulated Reality



- *Agent-Based Learning & Behavior*
- *Hybrid Learning & Planning in Interactive Environments*
- *Learning, Reasoning & Uncertainty*
- *AI & Ethics*

- *Semantics and Interoperability*
- *Service Architecture & Development*
  - *Machine Learning Services*
  - *Multi-Agents Services*
- *Efficient Management of AI Platforms*
- *High-Performance Computing & Compiler Technologies*



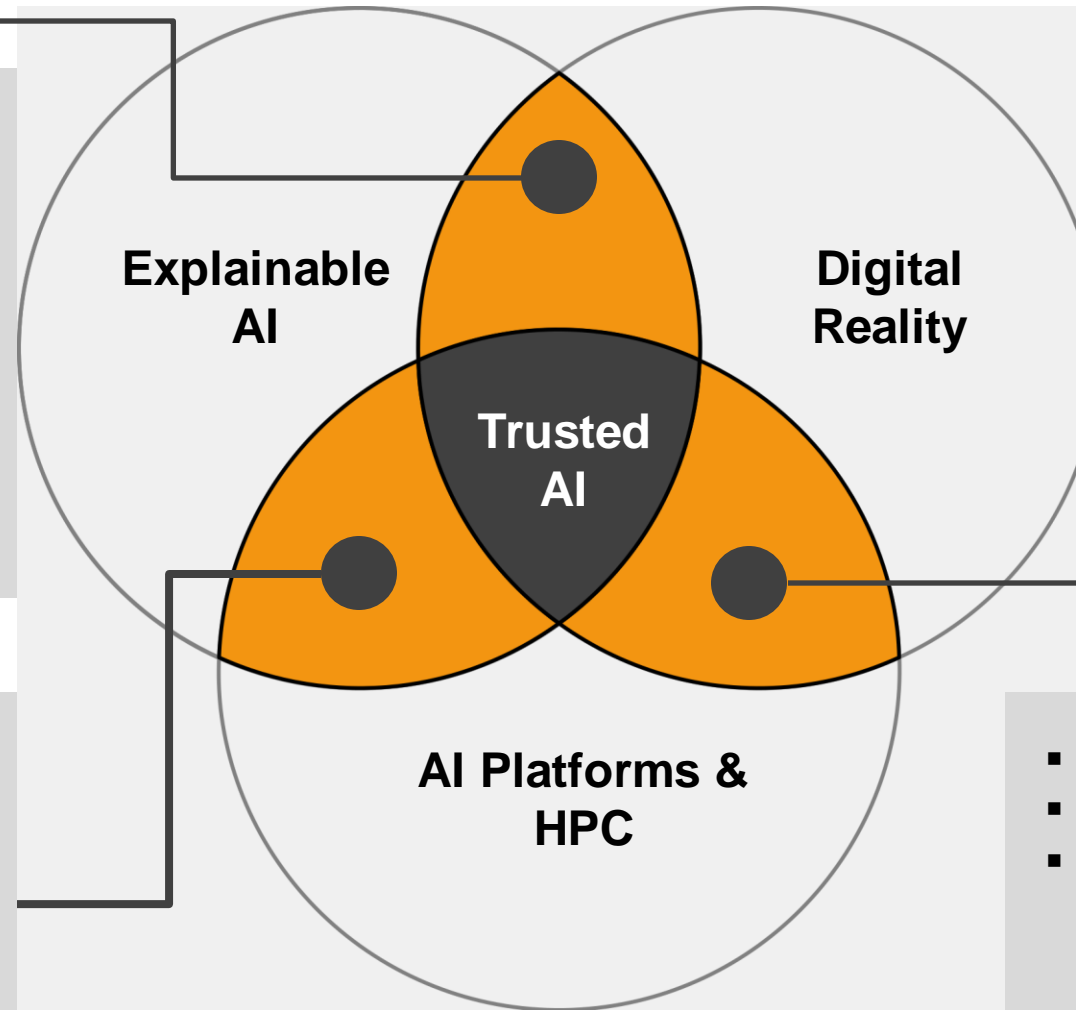
- *Model Capturing*
- *Model Learning or Extrapolation from Examples*
  - *Geometric Models*
  - *Appearance Models*
  - *Motion Models*
  - *Behavior Models*
- *Shallow (data-driven) to Deep (physical-driven) Modelling*
- *Parameter Space Sampling*
- *Simulation Techniques*
  - *Monte-Carlo Simulation*
  - *FEM Mechanics*

# Agents and Simulated Reality



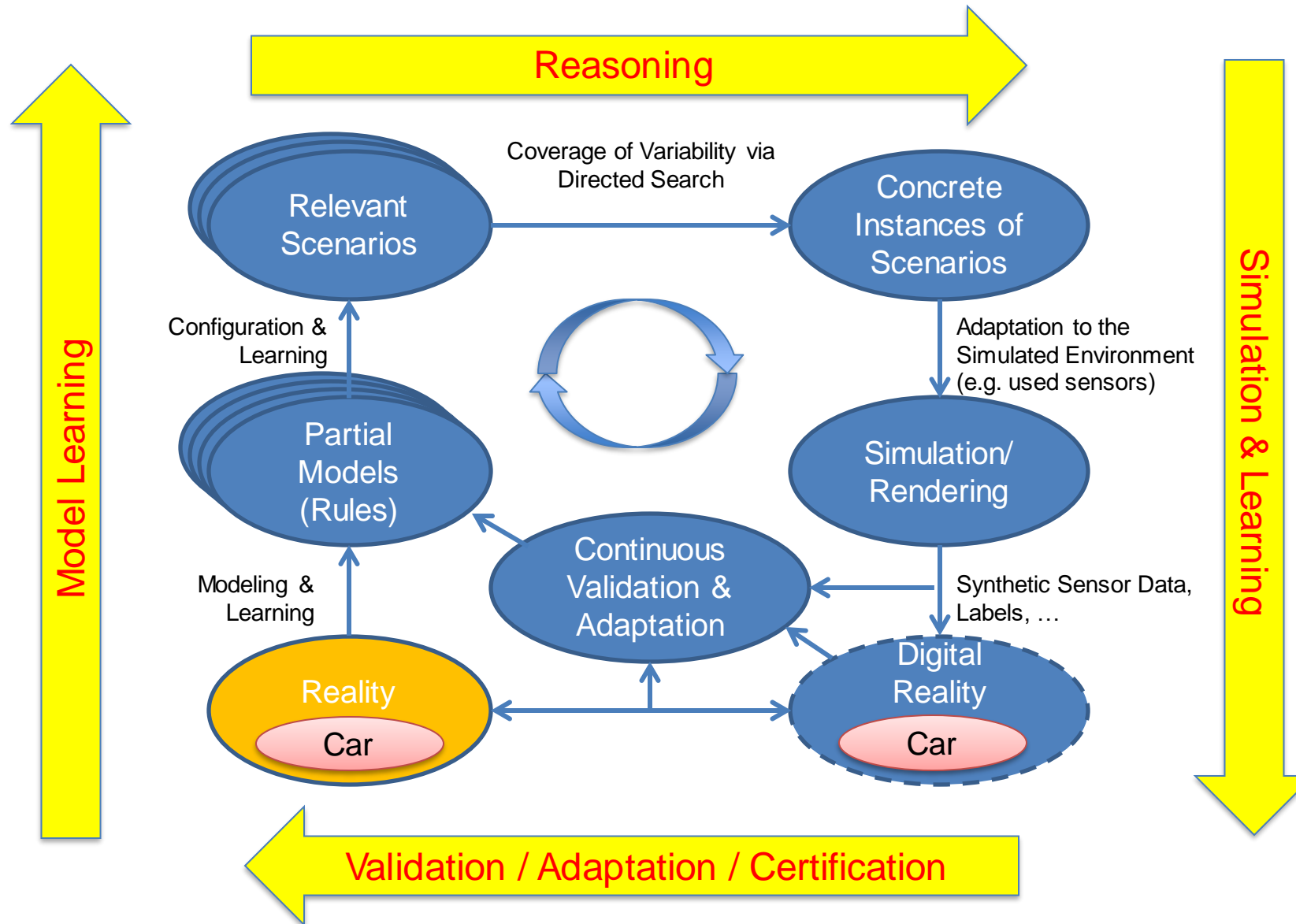
- *Novel AI techniques for DR*
- *Extracting semantics from DR (Re-Semantification)*
- *DR for explaining AI systems*
- *Adaptive sampling using corner case definitions*
- *GDPR conforming training data through abstract models*

- *Efficient AI-platform tools for explainable AI*
- *Ethics-driven requirements for AI platforms*
- *Handling of highly varying requirements in given HW systems*



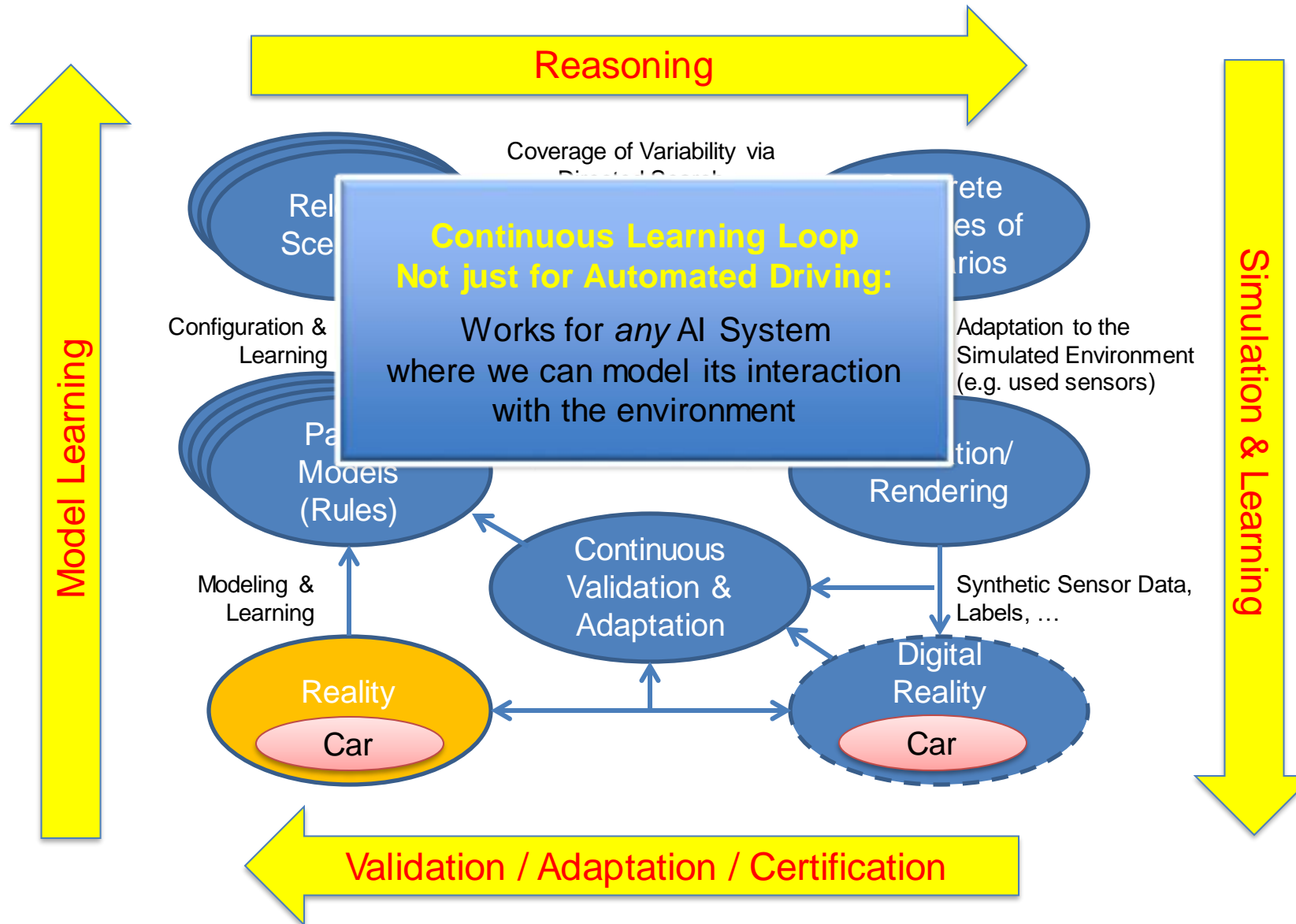
- *Distributed closed-loop learning*
- *Efficient simulation environments*
- *Behavior-driven generation & extrapolation of synthetic training data for ML*

# Digital Reality: AI to Optimize and Certify AI

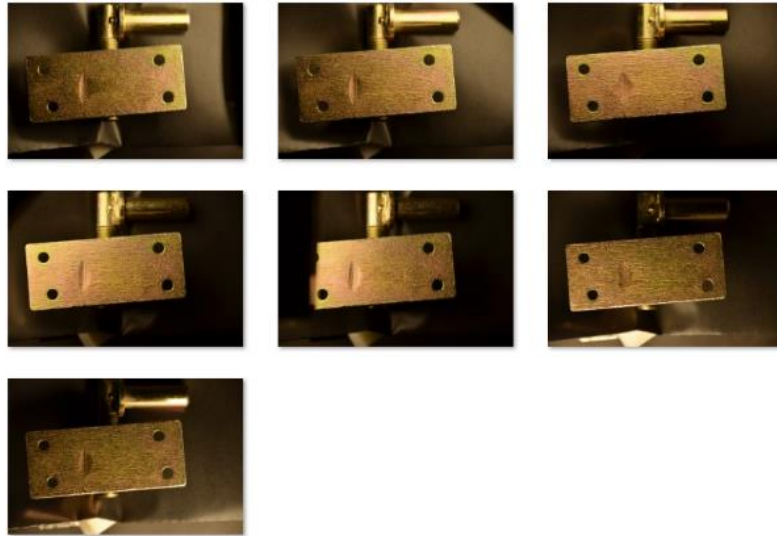




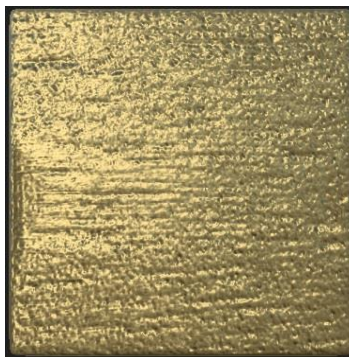
# Digital Reality: AI to Optimize and Certify AI



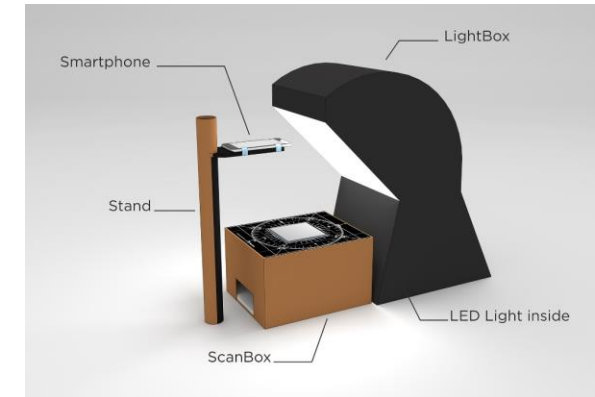
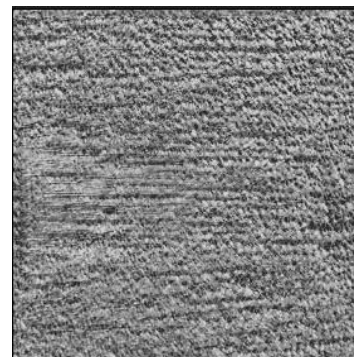
# Techniques: Capturing Material Properties



**Normal Map**



**Height Map**



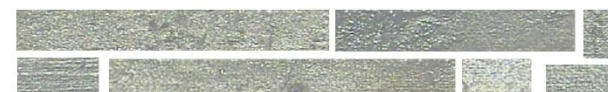
# Generating Synthetic Data



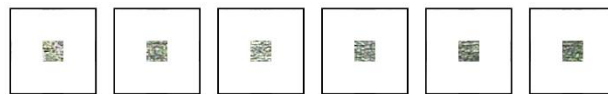
- **Parametrically generating synthetic textures**
  - Extraction of clean texture patches (as dictionary)
  - Exemplar-based in-painting for generating synthetic textures
- **Parametrically generating synthetic cracks**
  - Measured statistical distribution of intensities, widths, segment lengths, angles at bends
  - Used this parametric model to generate synthetic cracks



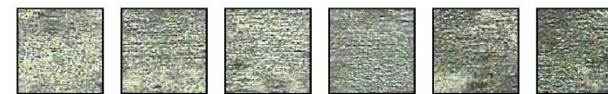
(a) Examples of real cracks.



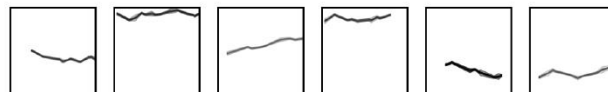
(b) Excerpt of texture dictionary.



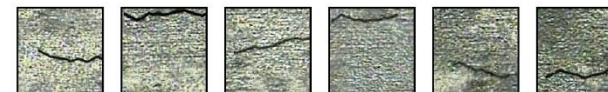
(c) Seeds for texture inpainting.



(d) Synthetically generated textures.



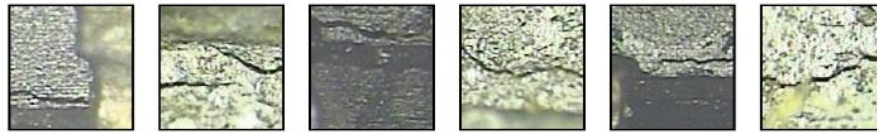
(e) Synthetically generated cracks.



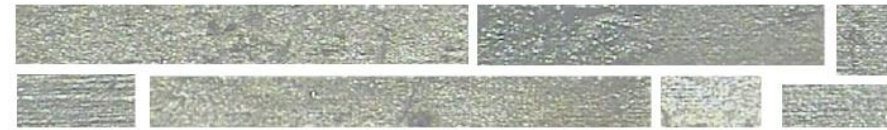
(f) Synthetically generated training data examples.



# Details of Data Generation



(a) Examples of real cracks.



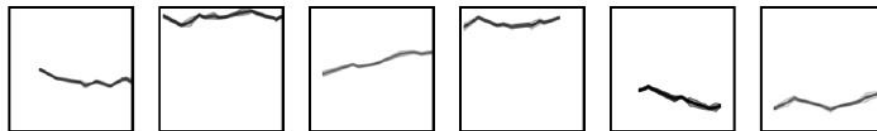
(b) Excerpt of texture dictionary.



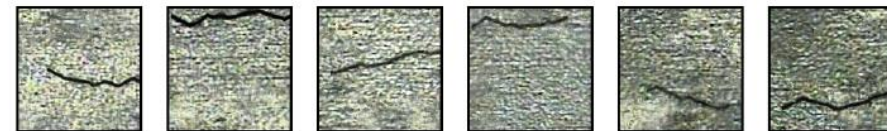
(c) Seeds for texture inpainting.



(d) Synthetically generated textures.



(e) Synthetically generated cracks.



(f) Synthetically generated training data examples.

# Autonomous Driving: Training using Synthetic Sensor Data (TÜV, OEMs, Suppliers, ...)

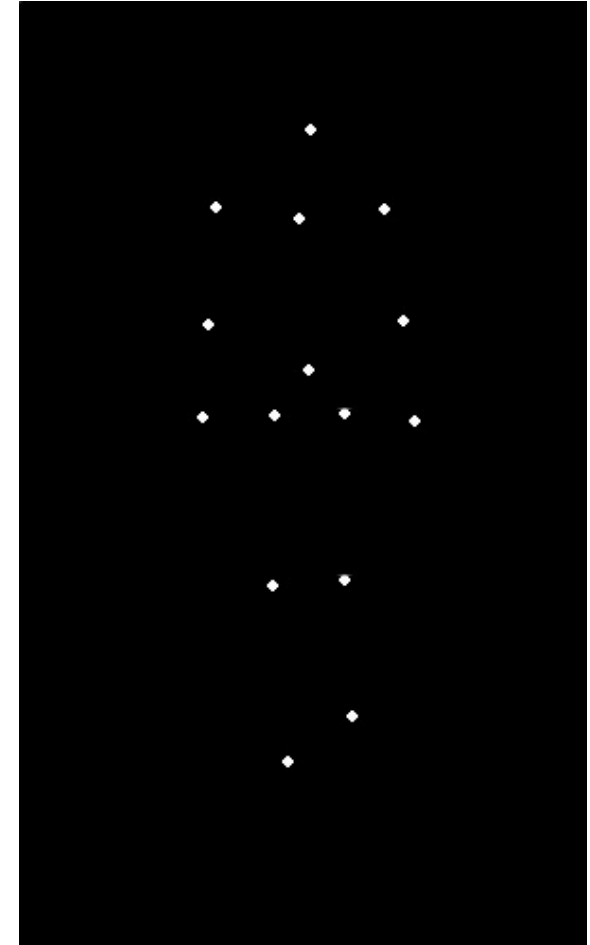




# Challenge: Better Models of the World (e.g. Pedestrians)



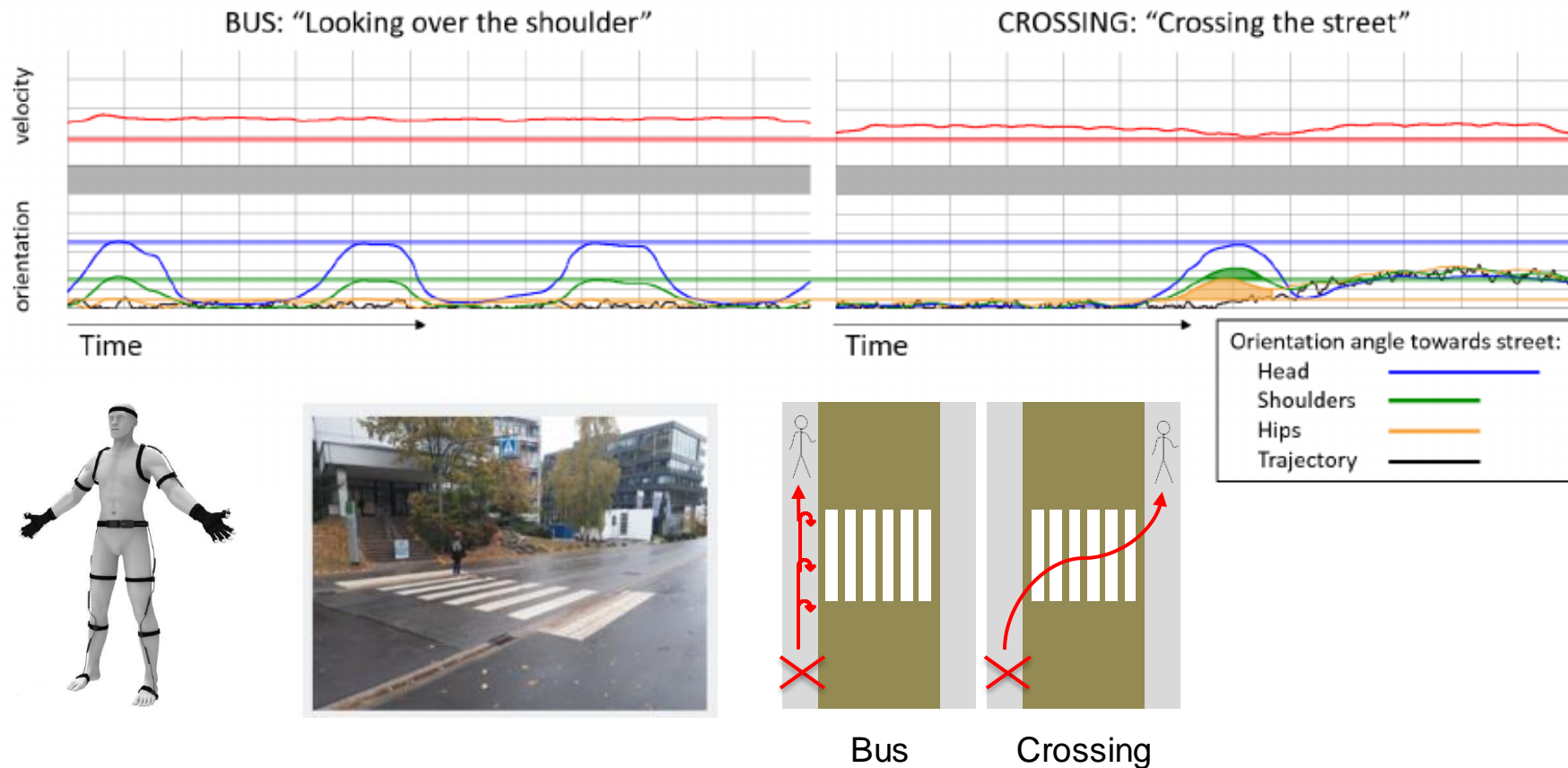
- **Long history in motion research (>40 years)**
  - E.g. Gunnar Johansson's Point Light Walkers (1974)
  - Significant interdisciplinary research (e.g. psychology)
- **Humans can easily discriminate different styles**
  - E.g. gender, age, weight, mood, ...
  - Based on minimal information
- **Can we teach machines the same?**
  - Detect if pedestrian will cross the street
  - Parameterized motion model & style transfer
  - Predictive models & physical limits



# Challenge: Pedestrian Motion



- Characterizing Pedestrian Motion
  - Clear motion differences when crossing the street

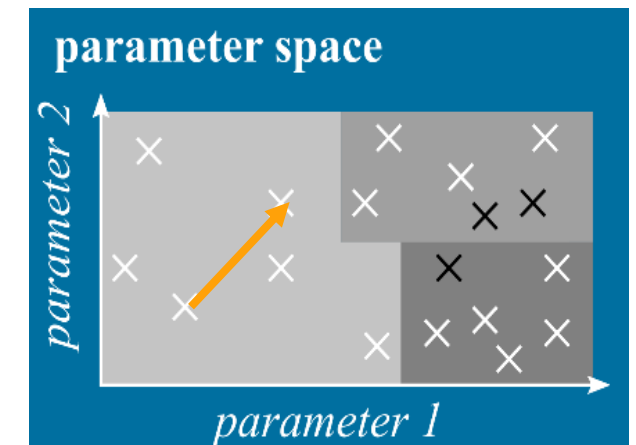
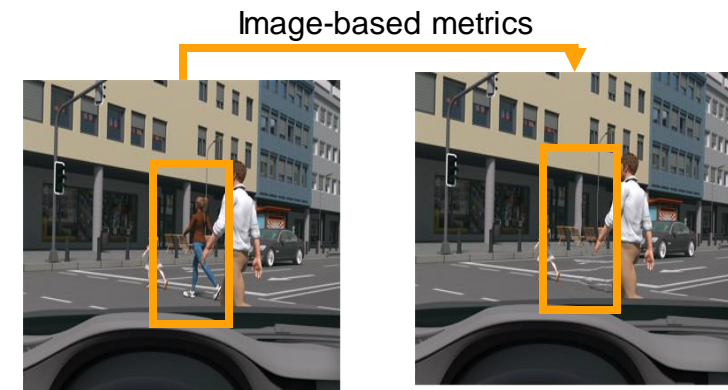




# Synthetic Training Data Generation: Parameter Space Characterization



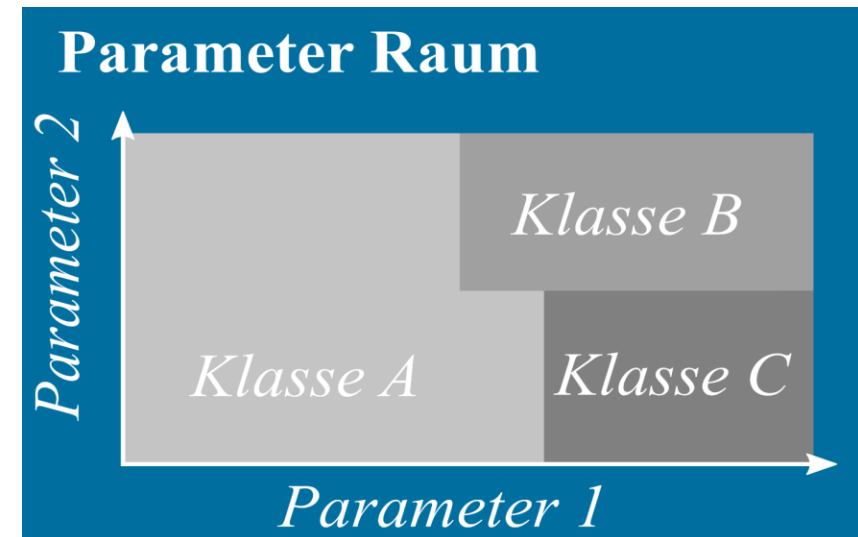
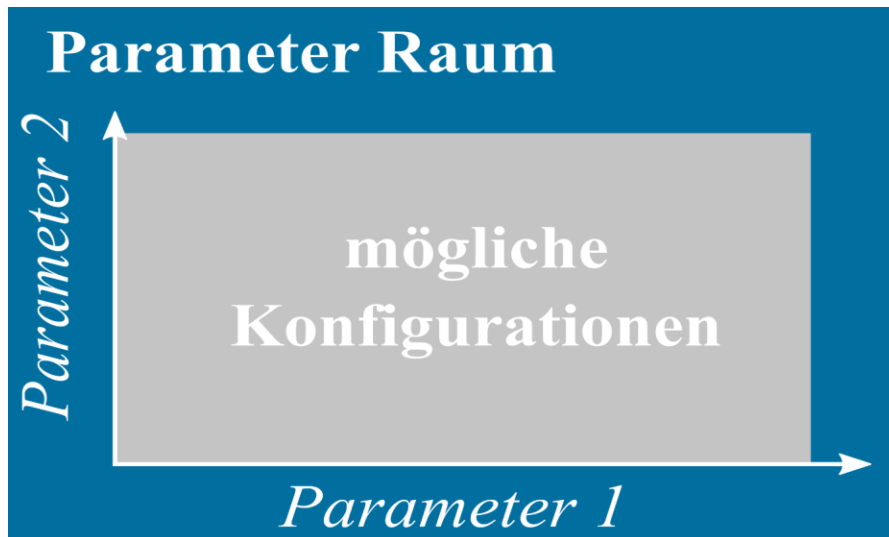
- **Goal: Need a metric for similarity of configurations**
  - Based on samples from high-dimensional parameter space
- **Allows for applying Monte-Carlo sampling approaches**
  - e.g. importance sampling
  - Provides statistical confidence and relevance of samples
- **Towards more semantically meaningful measures**
  - Class boundaries, input from NN, ...



# Sampling of Parameter Spaces



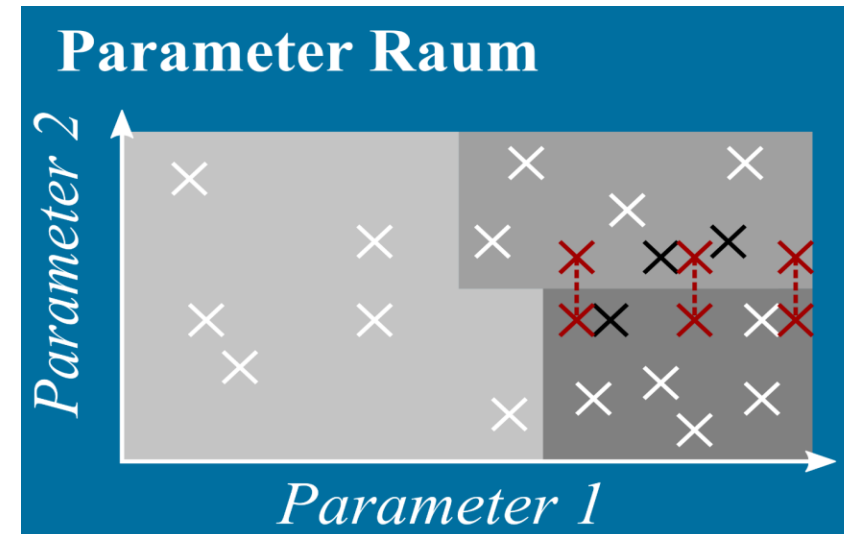
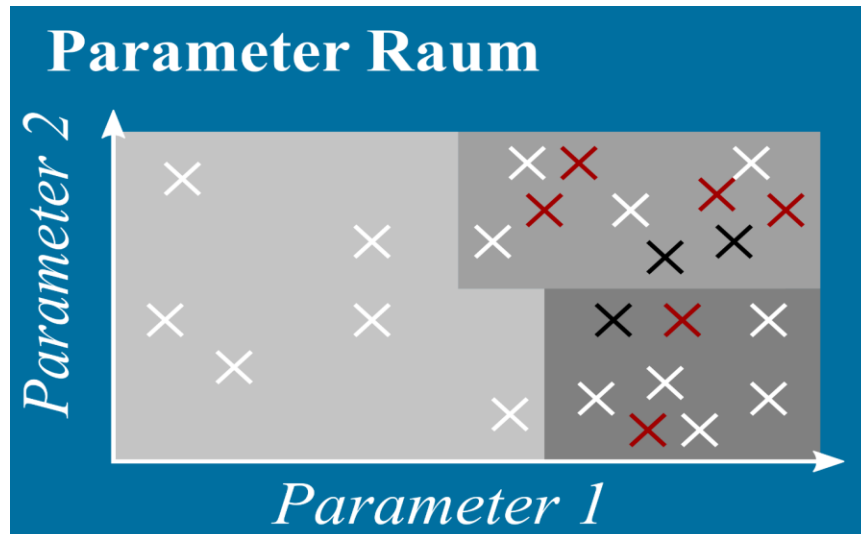
- Understand possibilities as a geometric space
- Possible configurations are points in this space
- High dimensional spaces of unknown characteristics
- Need to classify configuration (e.g. types of objects)



# Sampling of Parameter Spaces



- **Class Imbalance: Vastly different volume for classes**
  - Need metric of parameter space to compensate
- **Adaptives Sampling based on confusion matrix**
  - Iteratively refine based on metrics on results (loss function)
- **Direct comparisons: Swapping road sign and child**

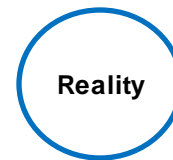




# Philosophy, Ethics, and AI

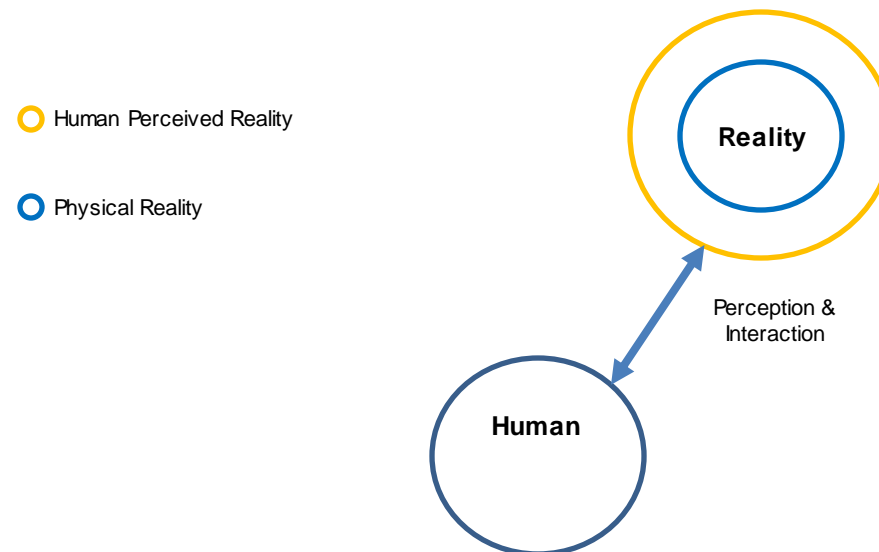


# Philosophy & AI: Relationship Between AI and Humans

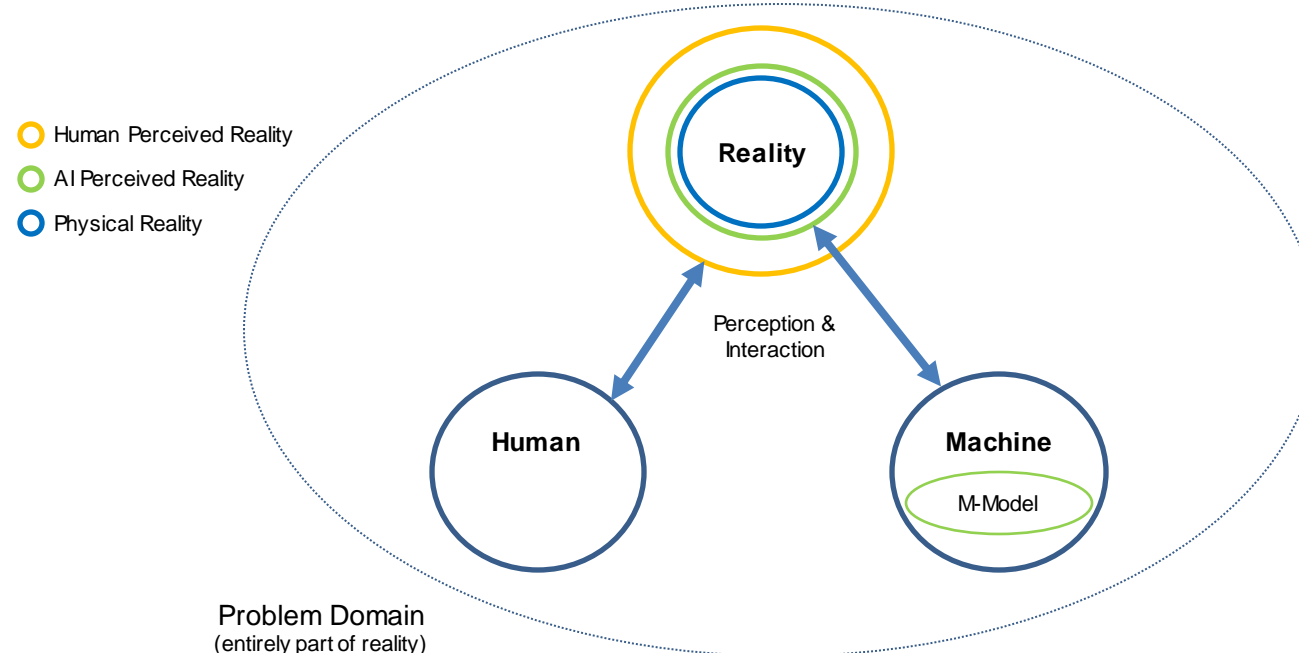


○ Physical Reality

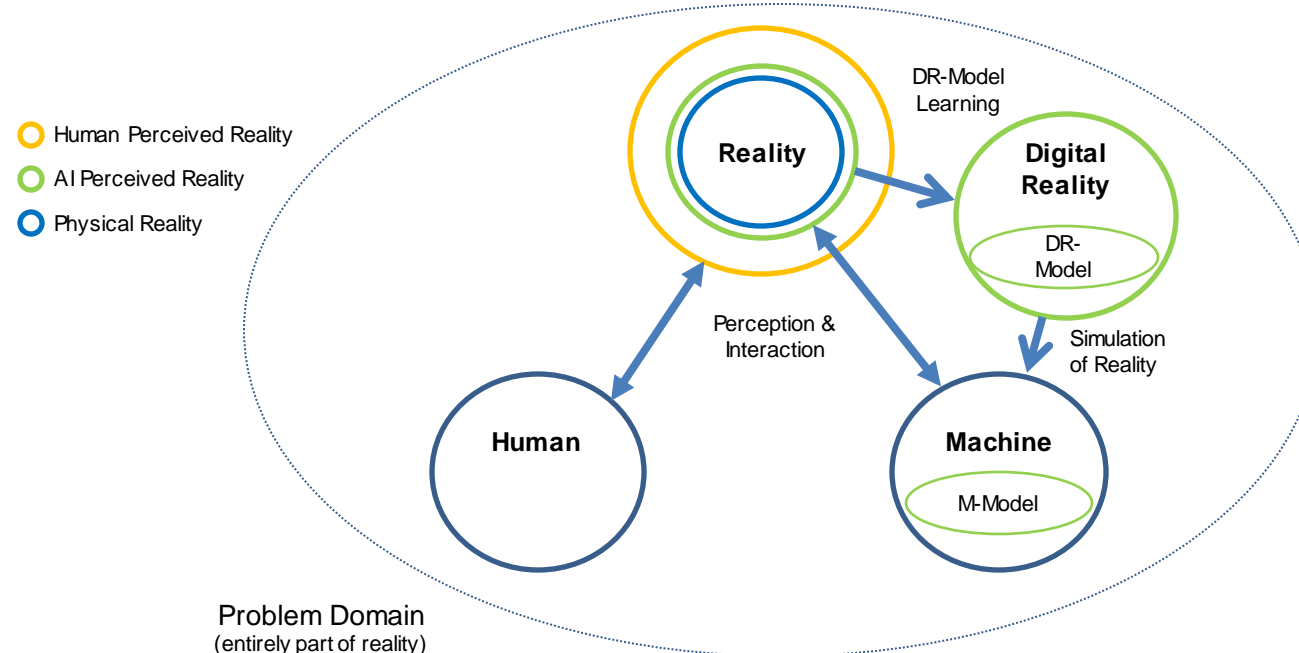
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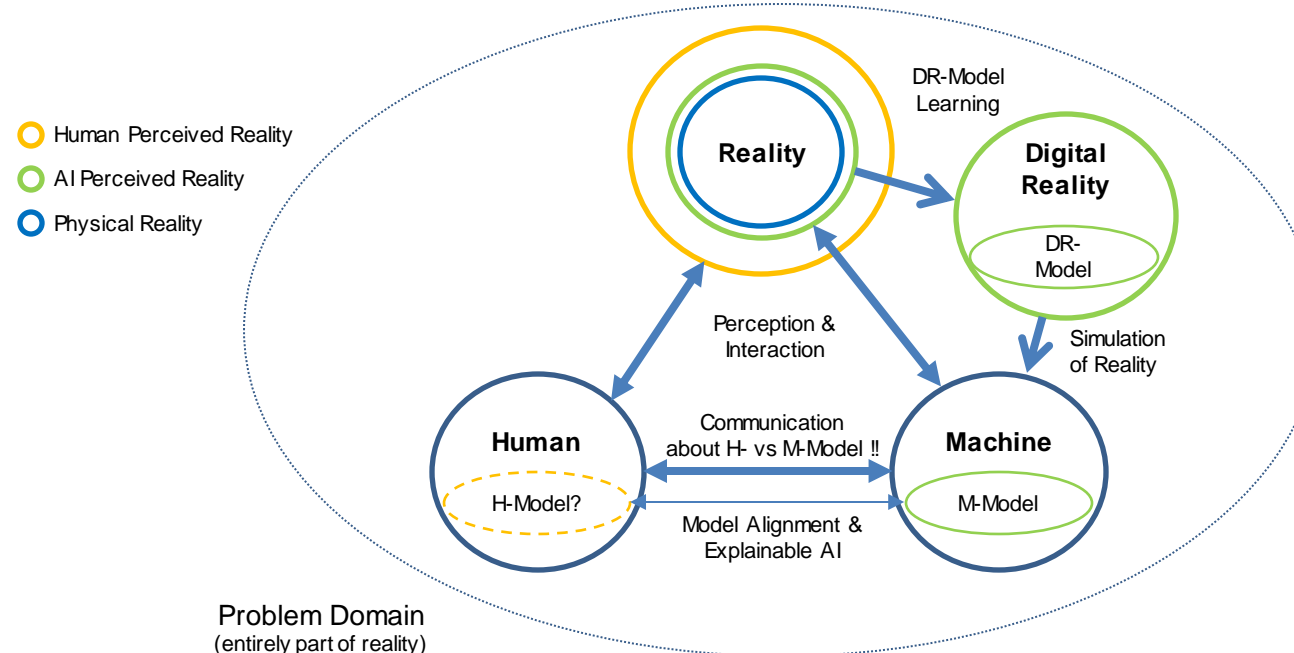


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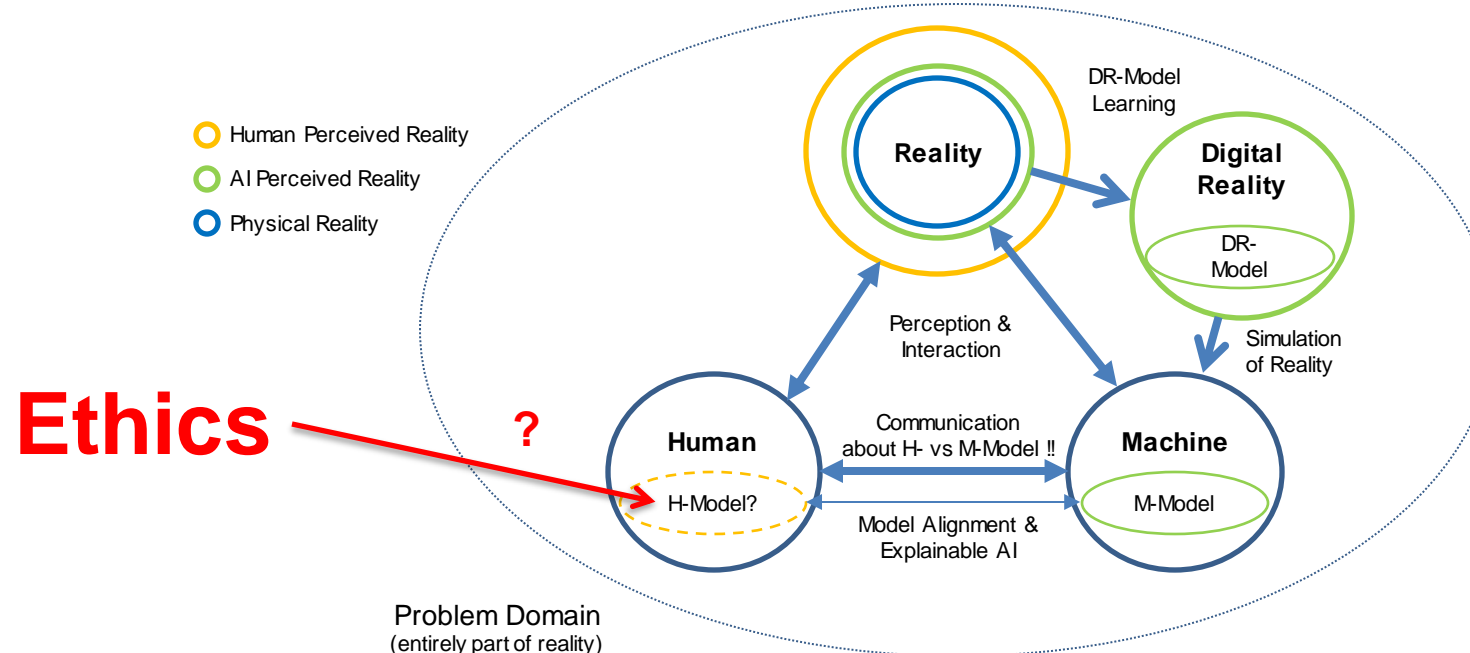




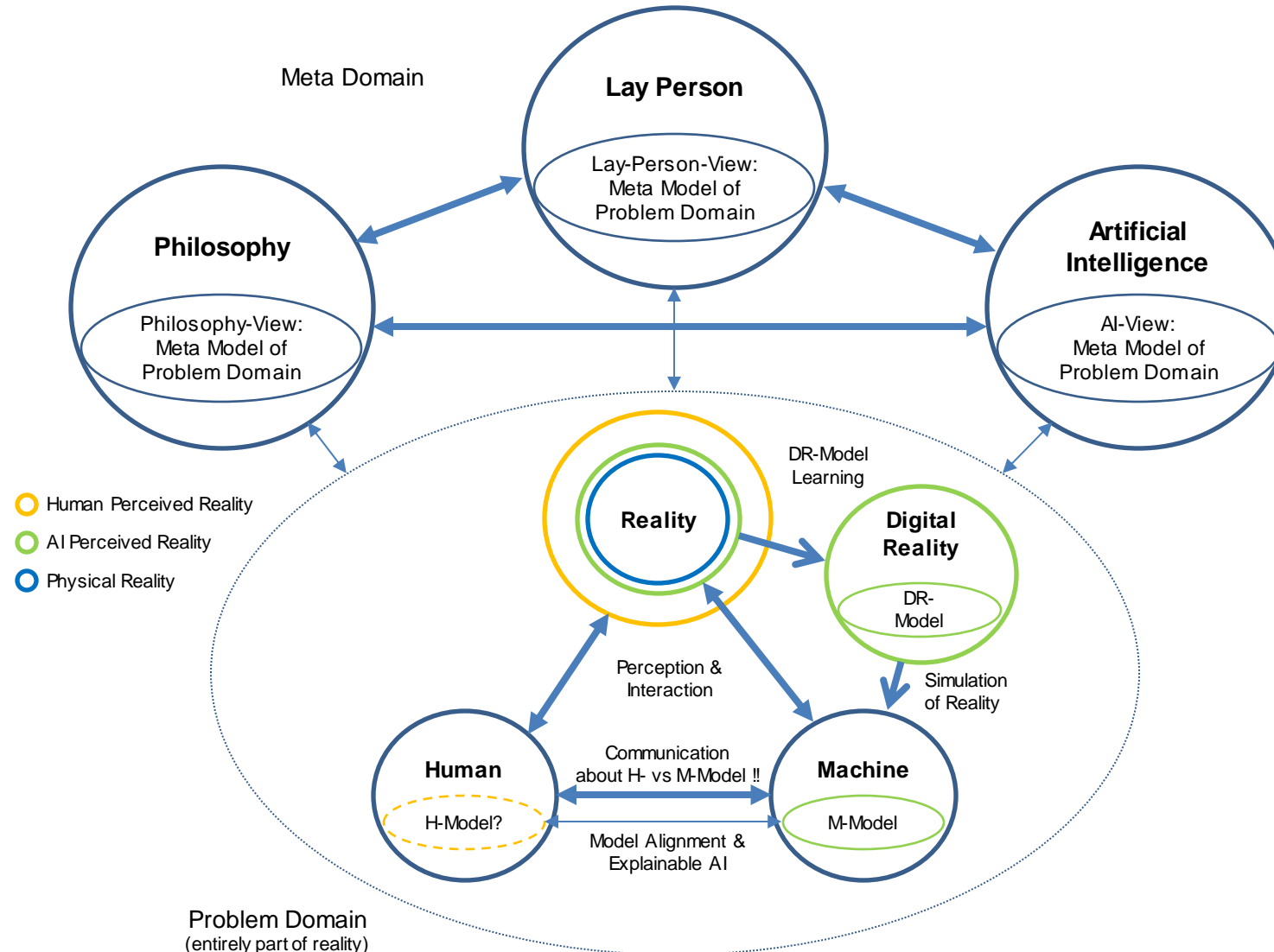
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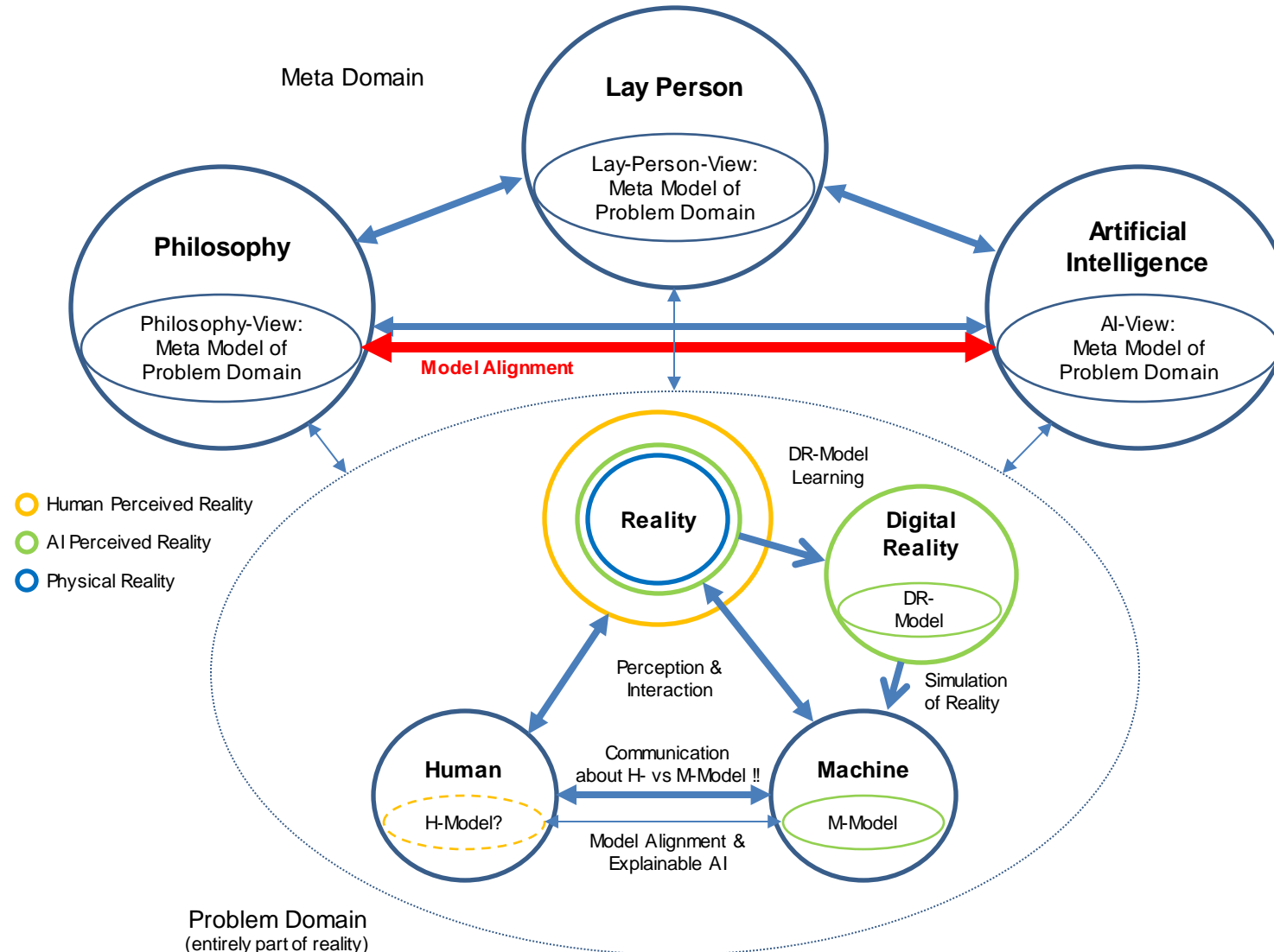
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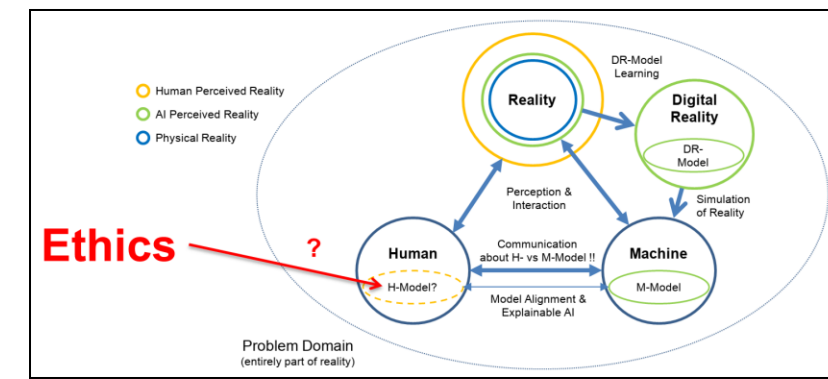
# Philosophy & AI: Relationship Between AI and Humans





# Ethics/Morality & AI (Provocative!)

- **Common Agreement on a Goal**
  - *Wellbeing* of the individual and society [e.g. Sam Harris: The Moral Landscape]
- **Evaluating the (likely) outcome of our actions in reality** (with respect to goal)
  - E.g. by simulation of actions with respect to a *model of reality*
- **I argue, this is also what we humans/societies do** (*at least in a rational setting*)
  - Model might be implicit/emotional (slow & fast thinking [Kahnemann])
- **In principal, AI could/should do the same – but ...**
  - Our models of the world are mediocre (but so were early weather models ...)
  - AI cannot deal with the complexity (yet ...)
  - The metric for evaluation is not clear (but might be a function of the goal ...)



# Ethics/Morality



- **Where are we going?**
  - Slowly converging on a better and better understanding of our world & society
  - Our models of the world need to be based on clear evidence & arguments
  - All models are open for revision as new evidence comes in (science & skepticism)
  - Our believes/models matter – they drive our actions
  - Humans matter – because we matter to other humans
  - We share this one world & the models of others affect all of us
  - Need to all work towards a common understanding
  - “There is nothing good, unless you do it” [Erich Kästner]

