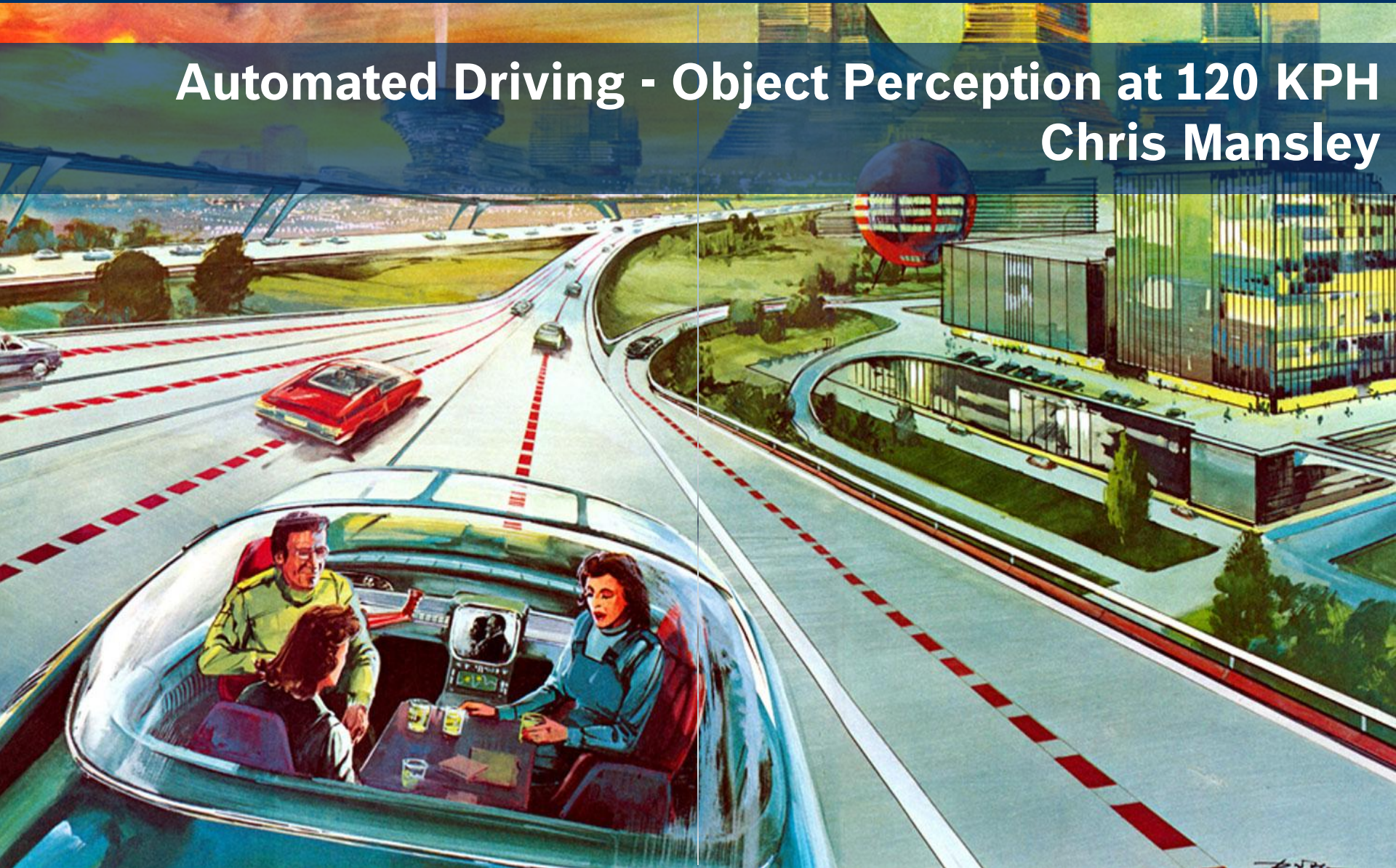


## Automated Driving - Object Perception at 120 KPH Chris Mansley

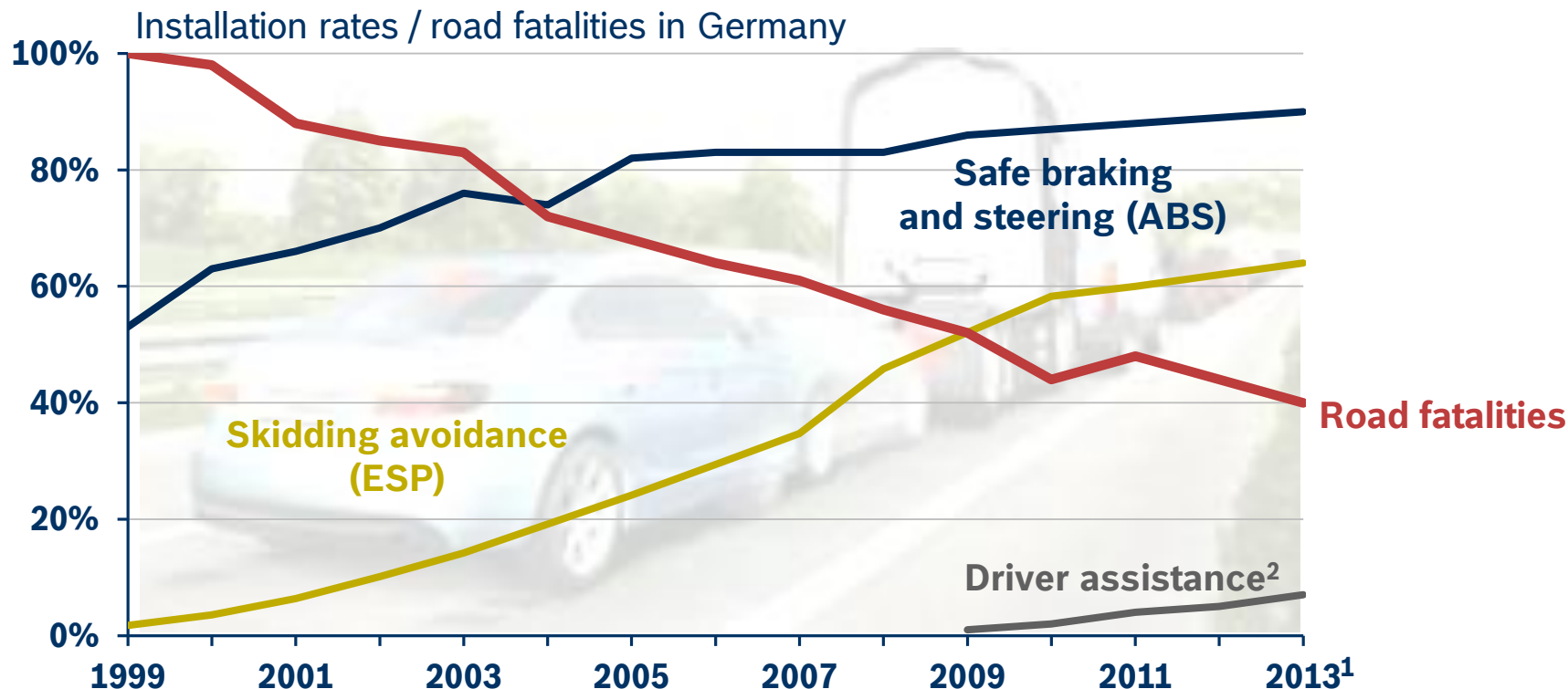


Chassis Systems Control



**BOSCH**

# Road safety – influence of driver assistance



## Number of road fatalities reduced by 60% within last 14 years

- 90% of all car accidents involving injury are caused by human error
- Introduction of further driver assistance systems will amplify positive trend

# Bosch in Automated Driving

- First involvement in automated driving 1990s
- DARPA Urban Challenge 2007
- Corporate Research, Palo Alto until 2011
- Engineering in Abstatt (DE) and Palo Alto (US) since 2011



## Development steps – automated driving

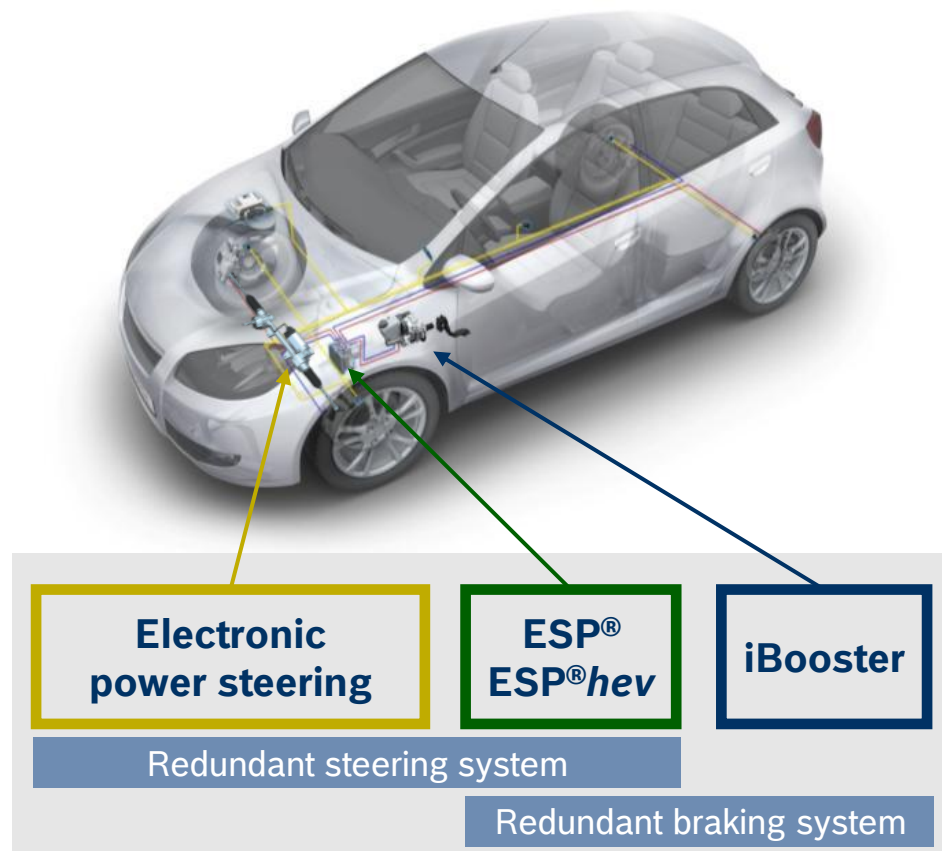
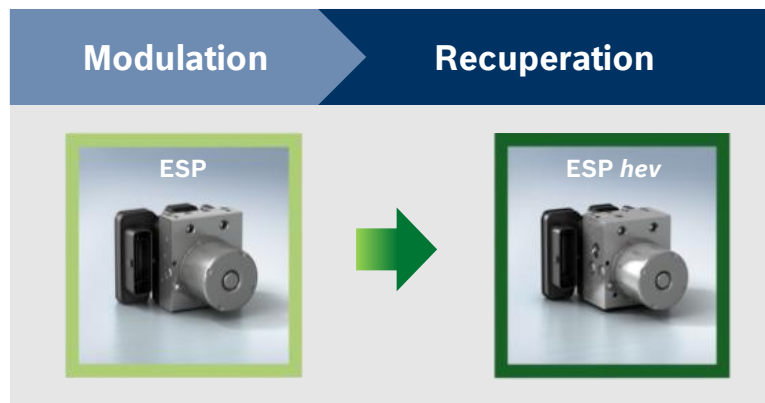
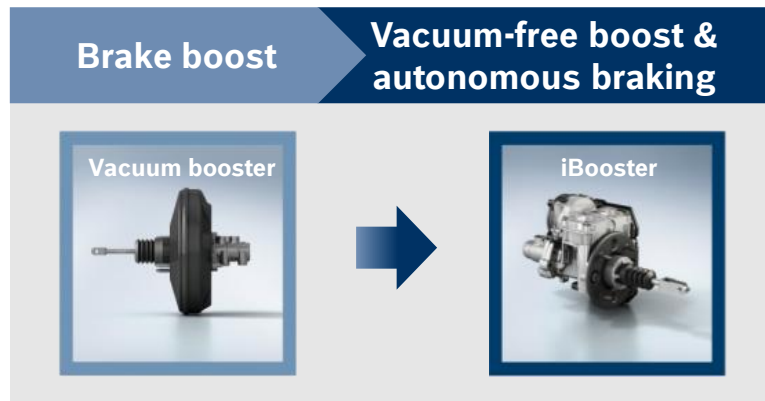


# Partial vs. Full automation

	Partial Automation	Full Automation
Execution /control	System	System
Monitoring	Driver	System
Driver Availability	Immediately	Not required
Failure to take over	Not acceptable	Safe state by system
System failure	Fail safe	Fail operational

- **Electronics:** Fail operational with redundant bus & power supply
- **Actuators:** Electronic/Electric fallback instead of mechanical fallback (driver)
- **Sensing:** Redundant sensing, multi-modal perception/localization
- **Computing:** Fail operational, automotive-grade (ECC memory, supervisor)
- **Functional Safety/Release Methods:** Novel system validation methods

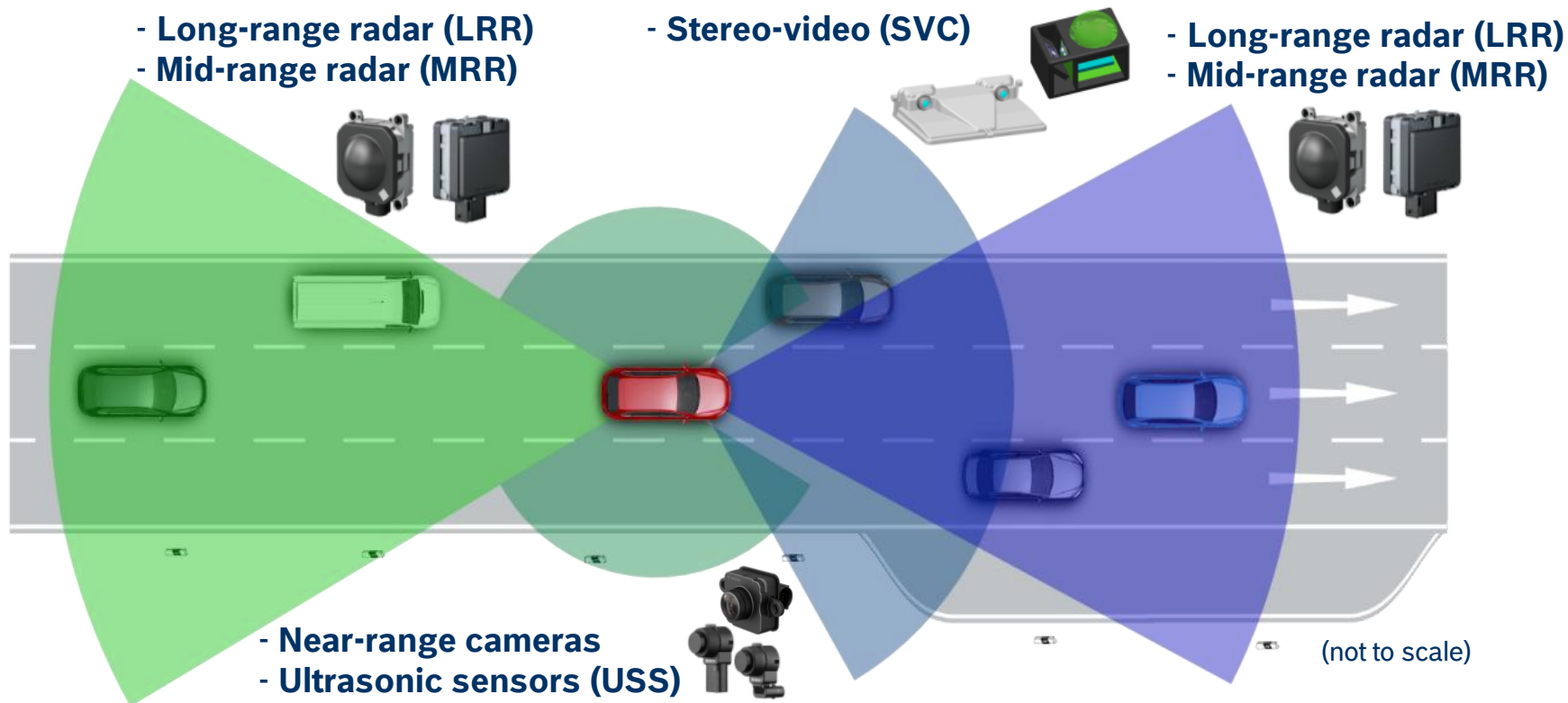
# Hardware Redundancy – Actuation



**Redundant steering, braking, and stabilization systems required**

- Modular actuation concept offers a perfect solution for automated driving

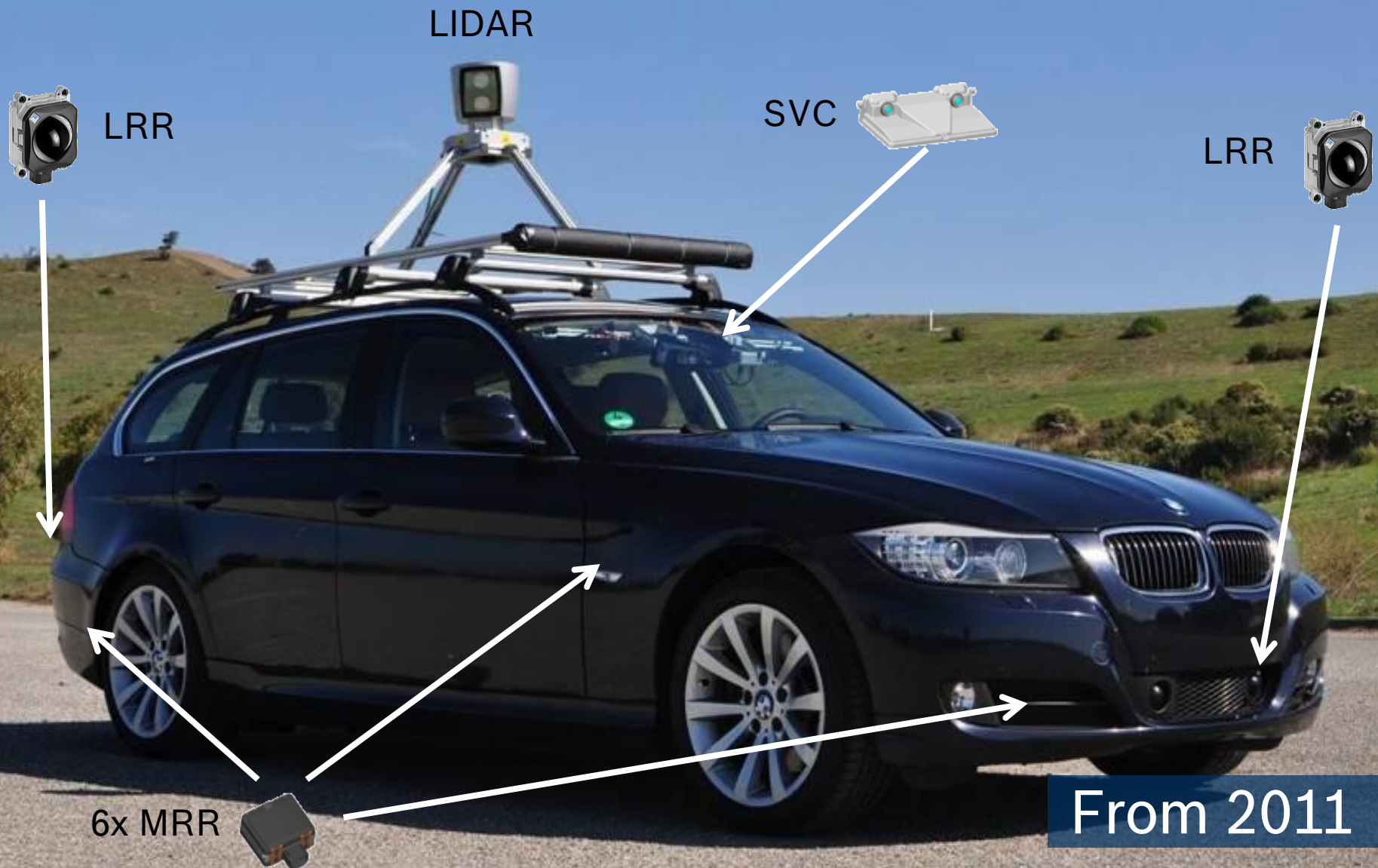
# Hardware Redundancy – Sensing



## 360° surround sensing by combination of different sensors

- Long- and mid-range radar prerequisite for driving at higher speed
- Satisfy reliability requirements by using multiple sensors for each area

# Object Perception @ 120 KPH



Chassis Systems Control



**BOSCH**

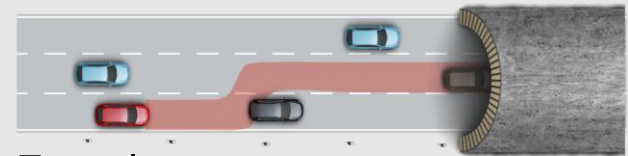
# Requirements for Sensing

- Automated driving use cases require
- 360° surround view
  - 3D information
  - Shape and surface measurement
  - High reliability
  - Low sensitivity to weather and light
  - Physical redundancy

### Example use cases:



Environment conditions (low sun)



Tunnel entrances



Uncommon obstacles (lumber truck)

**Highly automated driving raises new challenges for sensor concept**

# Example: Perception in High-speed Traffic

### → Challenge

- Timely response to fast approaching traffic

### → Example scenario:

- Other German highway drivers at up to 250 km/h (70 m/s)
- Assuming a perception cycle time of (say) 25ms
- Assuming a need for multiple detections to achieve object presence confidence and to converge to velocity estimate

### → At (say) 4 cycles with instantaneous (and in-step) decision making **the object has traveled 7 meters.**

- Not accounting for object prediction and trajectory computation

## Approach

### Surround sensors



Precise and reliable information on vehicle surroundings, e.g.

- Obstacle positions and velocity
- Obstacle classes, (vehicle, pedestrian, ....)
- Object shape

### Situational Data



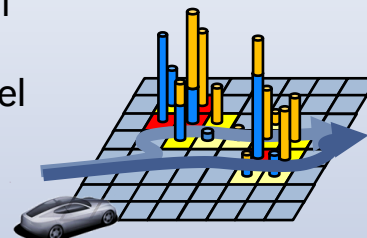
Additional (long-term, long-range)  
e.g.

- speed limits
- intersections
- road course



### Perception

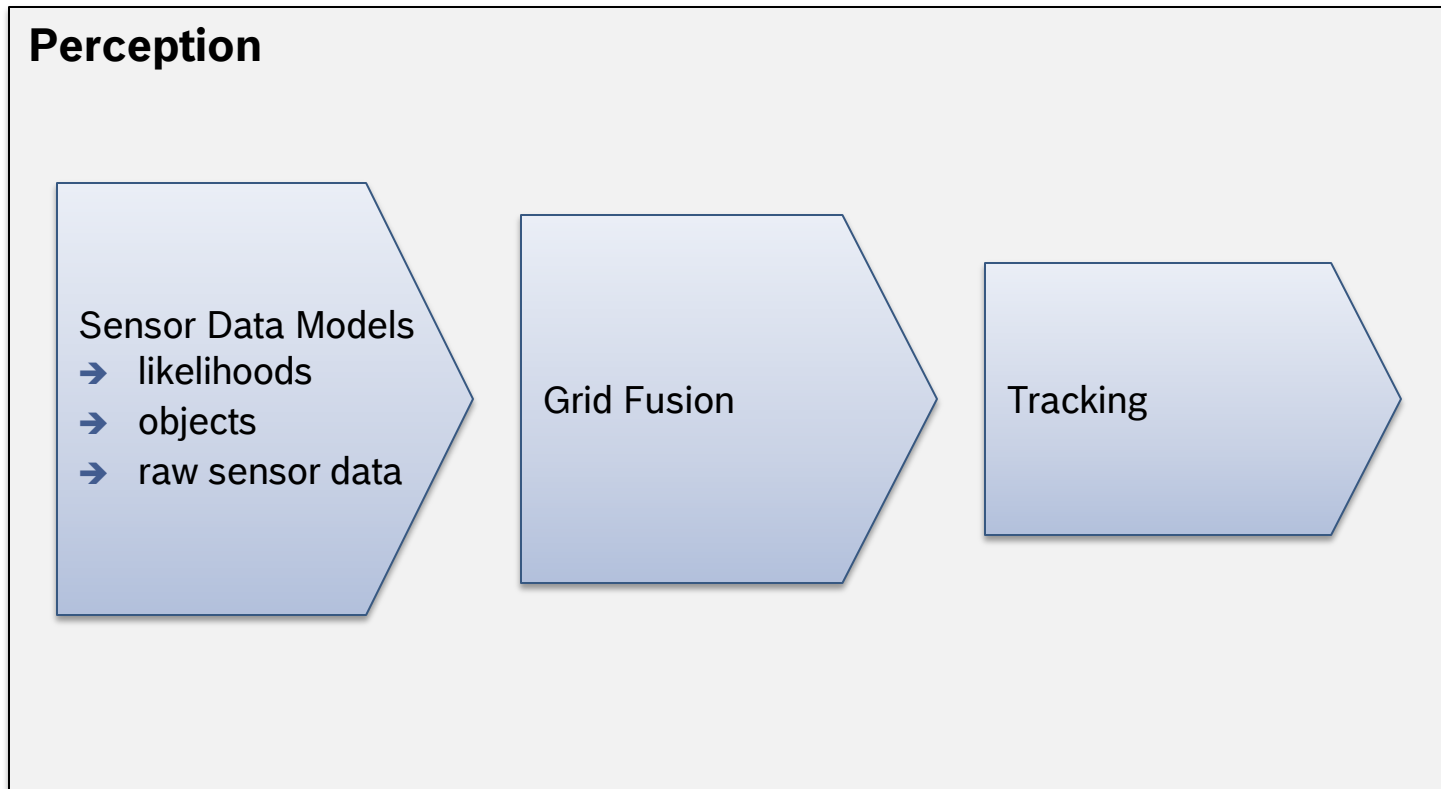
Probabilistic fusion of all information into a single surround model



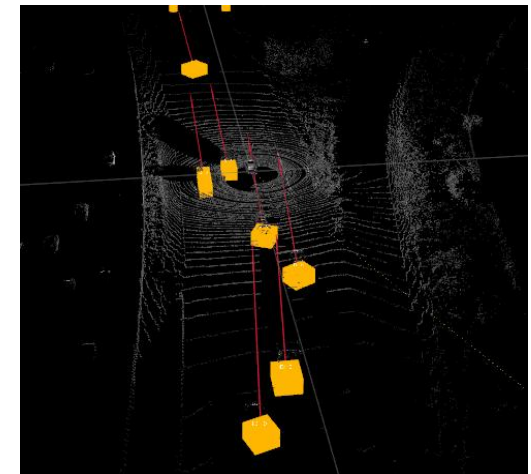
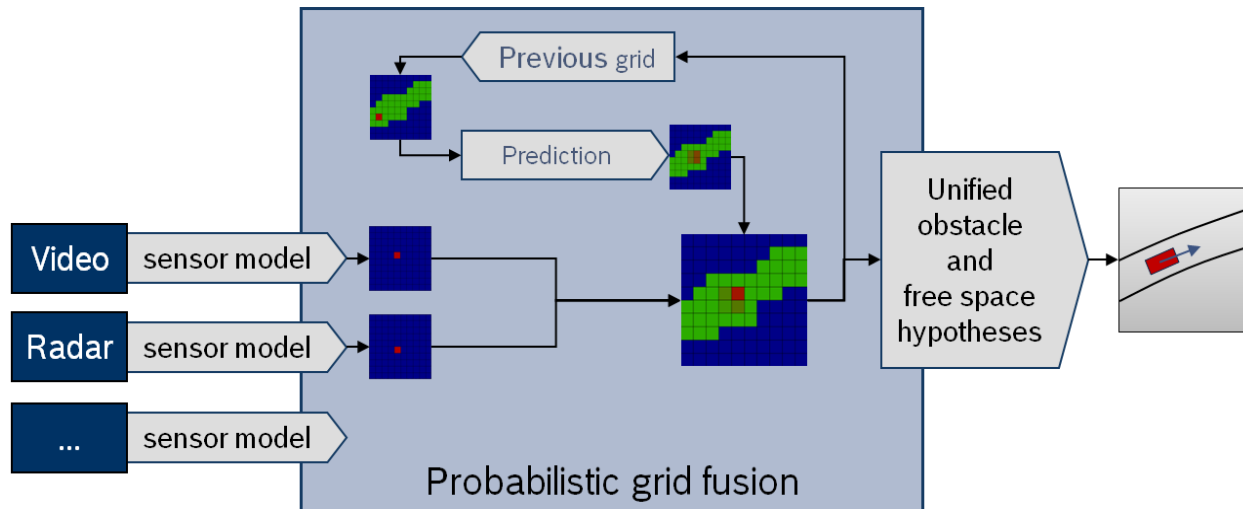
### Decision Making

Context-aware, probabilistic interpretation of fused environmental model from perception and situational data

# Perception Subsystem

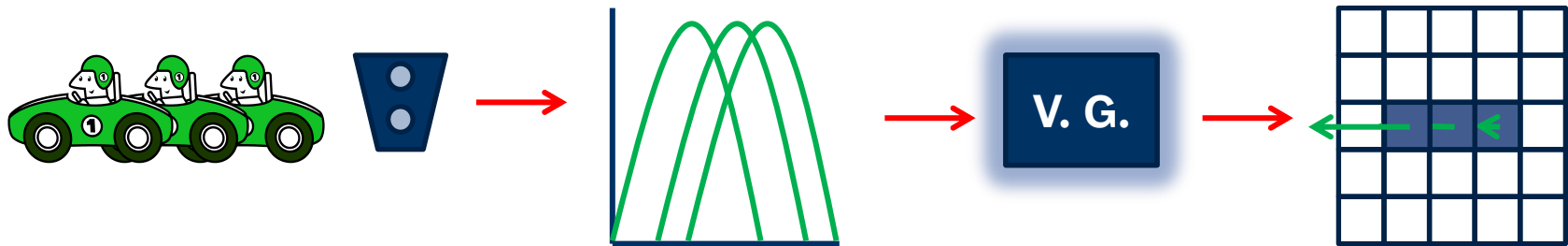


## Grid Fusion



- ➔ Grid based data fusion determines the occupancy probability of a cell by evaluating the current sensor reading and the history from past cycles

# Why velocity grids?



A velocity grid representation provides a **probabilistic framework** for fusing **multiple sensors** with **different models**, while representing **uncertainty** and **avoiding data association**

## Occupancy Grids

→ Represent the map as a field of binary random variables corresponding to the occupancy

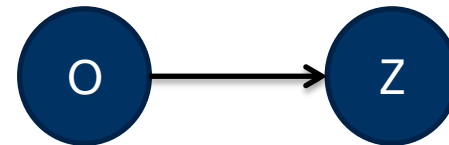
→ Assumptions :

- Static map
- Each cell is independent
- Robot location is known



$p(\text{occupancy})$

0.5	1.0	0.7
0.5	1.0	0.7
0.5	0.1	0.1

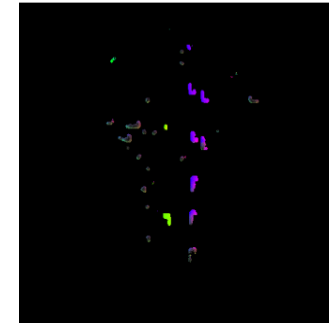


## Velocity Grids

→ Represent the map as a field of binary and discrete/continuous random variables corresponding to the occupancy and the velocity

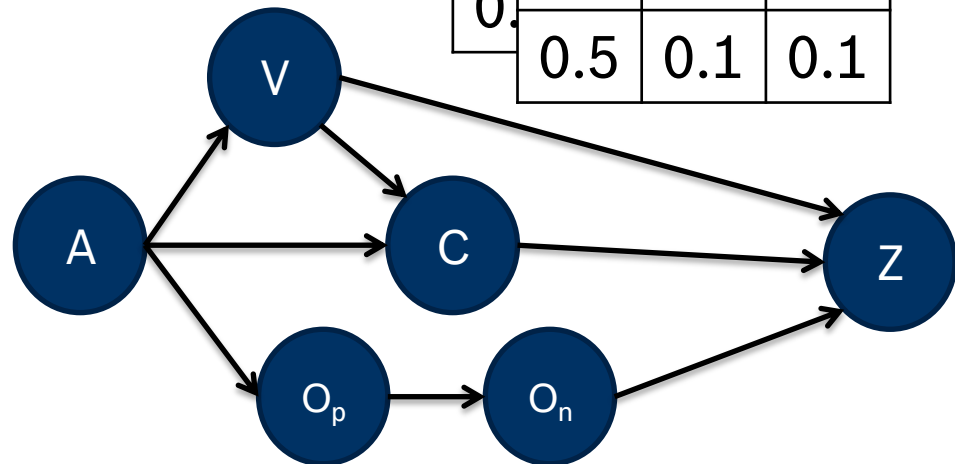
→ Assumptions :

- Dynamic map
- Cells are correlated
- Robot location is known



$p(\text{occupancy}=\{0,1\}, \text{velocity}=5)$

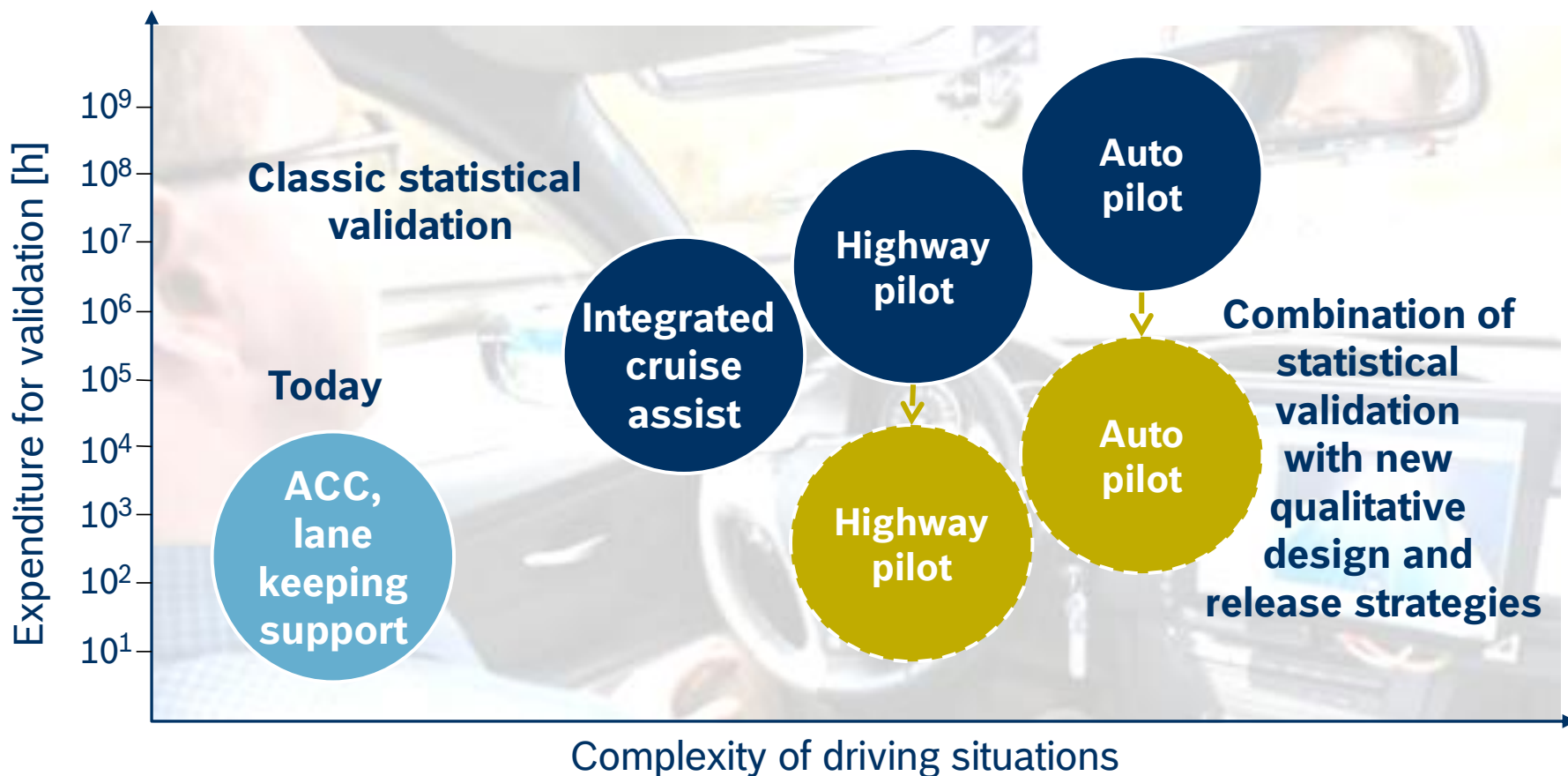
	0.5	1.0	0.7
0.5	0.5	0.0	0.1
0.0	0.5	0.0	0.7
0.0	0.5	0.1	0.1



# Software

- On-going **algorithm development**:
  - **Perception**:
    - High-speed traffic situations
    - Classification to support traffic prediction (e.g. indicator cue)
    - Map inconsistency detection (localization/planning)
  - **Decision making**:
    - Traffic prediction
    - Safe-stop (potentially high-dynamic maneuvers)
- **Validation** of system behavior
- On-going **system engineering**:
  - Addressing **scale** in object number, computational demands
  - Redundancy in computation: **system supervision**

# Validation and release process – challenges

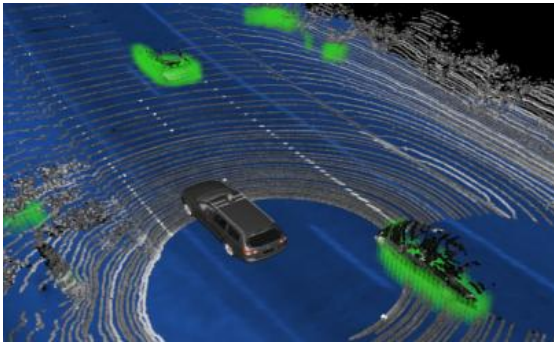


**Expenditure for validation will increase by a factor of  $10^6$  to  $10^7$**

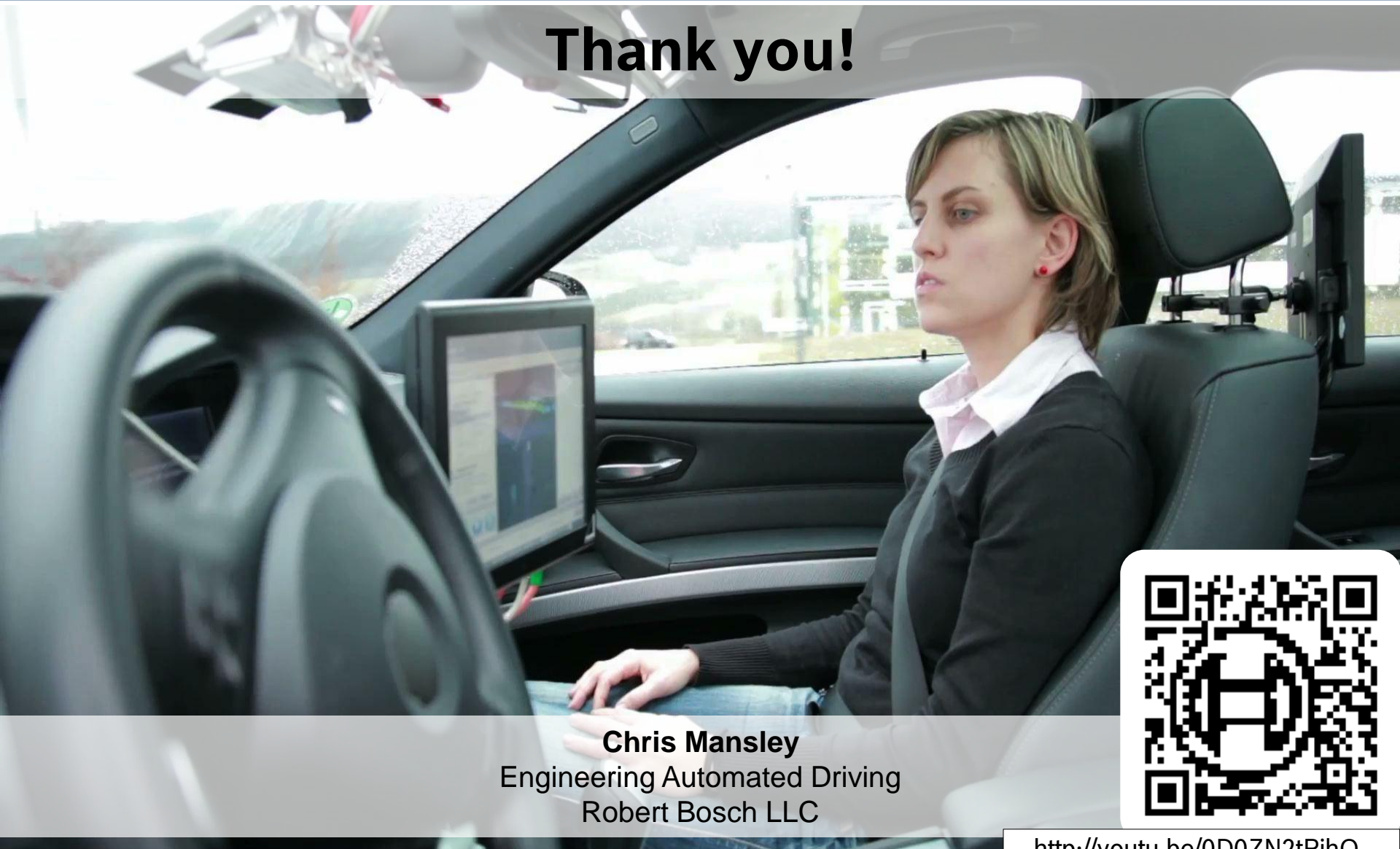
- Traditional statistical validation not suitable for higher degree of automation
- Highly automated systems require completely new release strategies

## Summary

- Automated driving functions will irreversibly change vehicle architecture (hardware, software) and system validation
- Technical and legal challenges still exist and need to be solved
  - Sensors, actuators, E/E architecture and driver monitoring
  - Algorithm development
- Stepwise implementation starting with Automated Highway Driving



# Thank you!



**Chris Mansley**  
Engineering Automated Driving  
Robert Bosch LLC



<http://youtu.be/0D0ZN2tPihQ>

Chassis Systems Control



**BOSCH**