



Computer vision and machine learning for automated driving

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Engineering Center Cluj
Our Innovation your Future

Bosch in Cluj-Napoca

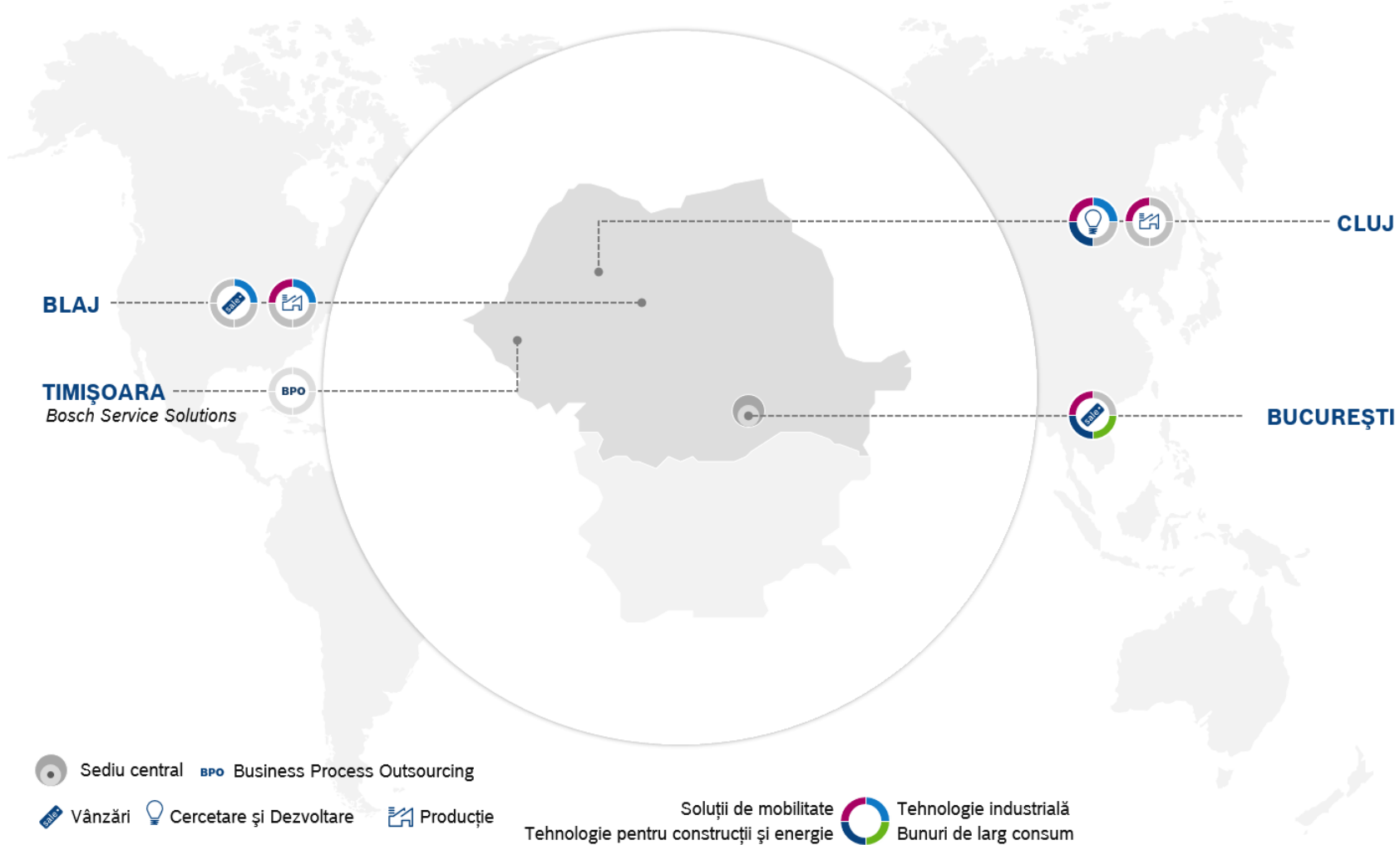


Engineering Center Cluj
@offices in the city center,
Cluj-Napoca (The Office and
Someșului St. 14)
SOFTWARE PROJECTS

Engineering Center Cluj
@Jucu
HARDWARE & MECHANICS +
RELIABILITY ENGINEERING
PROJECTS

Bosch Plant
@Jucu
PRODUCTION

Bosch Romania



Computer vision in today's intelligent cars

Four business sectors at Bosch

Automotive is the main business @Bosch

Mobility solutions



Industrial Technology



Energy and Building Technology



Consumer Goods



Engineering Center Cluj



SOFTWARE ENGINEERING



**RELIABILITY ENGINEERING
& VALIDATION**



**HARDWARE & MECHANICAL
ENGINEERING**



Ultrasonic Systems



Video Systems



Engine Control Unit



Light Electric Mobility



Radar systems



Electric Power Steering



Industry 4.0 Sensor Solutions

Computer vision in today's intelligent cars

Future Mobility - Electrified, automated and connected



costs hybrid e-motor
eBike power electronics

electrified

plug-in eScooter range
fun-to-drive battery
charging infrastructure

legislation driver assistance
emergency braking autopilot

automated

highway-pilot sensors
redundancy electric steering
valet parking

electronic horizon
smartphone integration

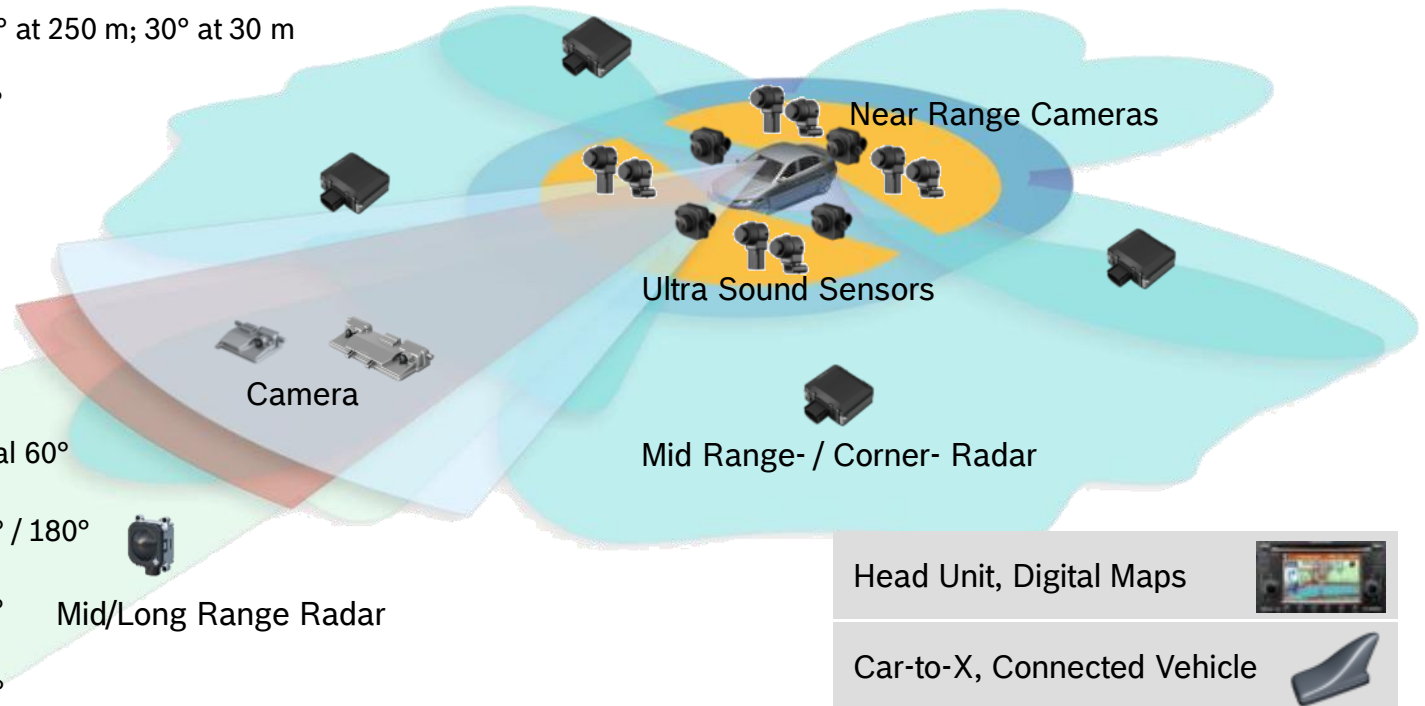
connected

eCall cloud
services fleet management
car2car augmented reality

Computer vision in today's intelligent cars

Bosch sensors portfolio

- Long-range radar**
Detection range: ~250 m | Field of View: horizontal 12° at 250 m; 30° at 30 m
- Night vision camera**
Detection range: ~150 m | Field of view: horizontal 32°
- Mid-range radar front**
Detection range: ~160 m |
Field of View: horizontal 12° at 160 m; 90° at 25 m
- Multi purpose camera / stereo video camera**
Detection range: ~150 m (for objects) |
Field of View: horizontal 50°
- Ultrasonic sensor**
Detection range: ~2.5 m / 5 m | Field of View: horizontal 60°
- Near range camera**
Detection range: ~15 m | Field of View: horizontal 130° / 180°
- Multi-camera system**
Detection range: ~15 m | Field of View: horizontal 360°
- Mid-range radar rear**
Detection range: ~80 m | Field of View: horizontal 150°



Computer vision in today's intelligent cars

Driver assistance functions portfolio

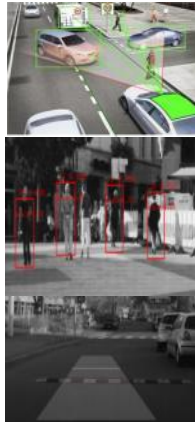
Light & sight at night functions

- **High Beam Control**
- **Adaptive Headlight Control**
- Masked Continuous High Beam
- Night Vision with Pedestrian Detection



Object / Surface functions

- **Adaptive Cruise Control**
- **Emergency braking on cars**
- **Emergency braking on pedestrians**
- Maneuver Assistance
- Distance information & warning
- Cross Traffic Alert
- Forward Collision Warning
- Speed Bump Assist
- **Adaptive Suspension**
- Height clearance assist



Road Sign functions

- Speed Limits
- Stop Signs
- Give Way Signs
- Additional signs (Pictograms)
- Fusion with digital map
- general signs (triangular, rectangular)



Lane functions

- Lane Departure Warning
- **Lane Keeping Assist**
- Lane Change Assist / Blind Spot
- **Construction Zone Assist**
- **Narrow Passage Assist**
- Traffic Jam Assist
- Evasion Assist
- Driver Drowsiness Detection



Computer vision in today's intelligent cars

Vision of automated driving

Light & sight at night functions

- High Beam Control
- Adaptive Headlight
- Masked Continuous
- Night Vision with Pedestrian Detection

Object / Surface

- Adaptive Cruise Control
- Emergency Lane Change
- Emergency Steering Assist
- Maneuver Assistance
- Distance Information
- Cross Traffic Alert
- Forward Collision Warning
- Speed Bump Assist
- Adaptive Suspension
- Height clearance assist

Road Sign functions

- Speed Limits
- Stop Signs
- Give Way Signs
- Road signs (Pictograms)
- Map



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Levels of automation

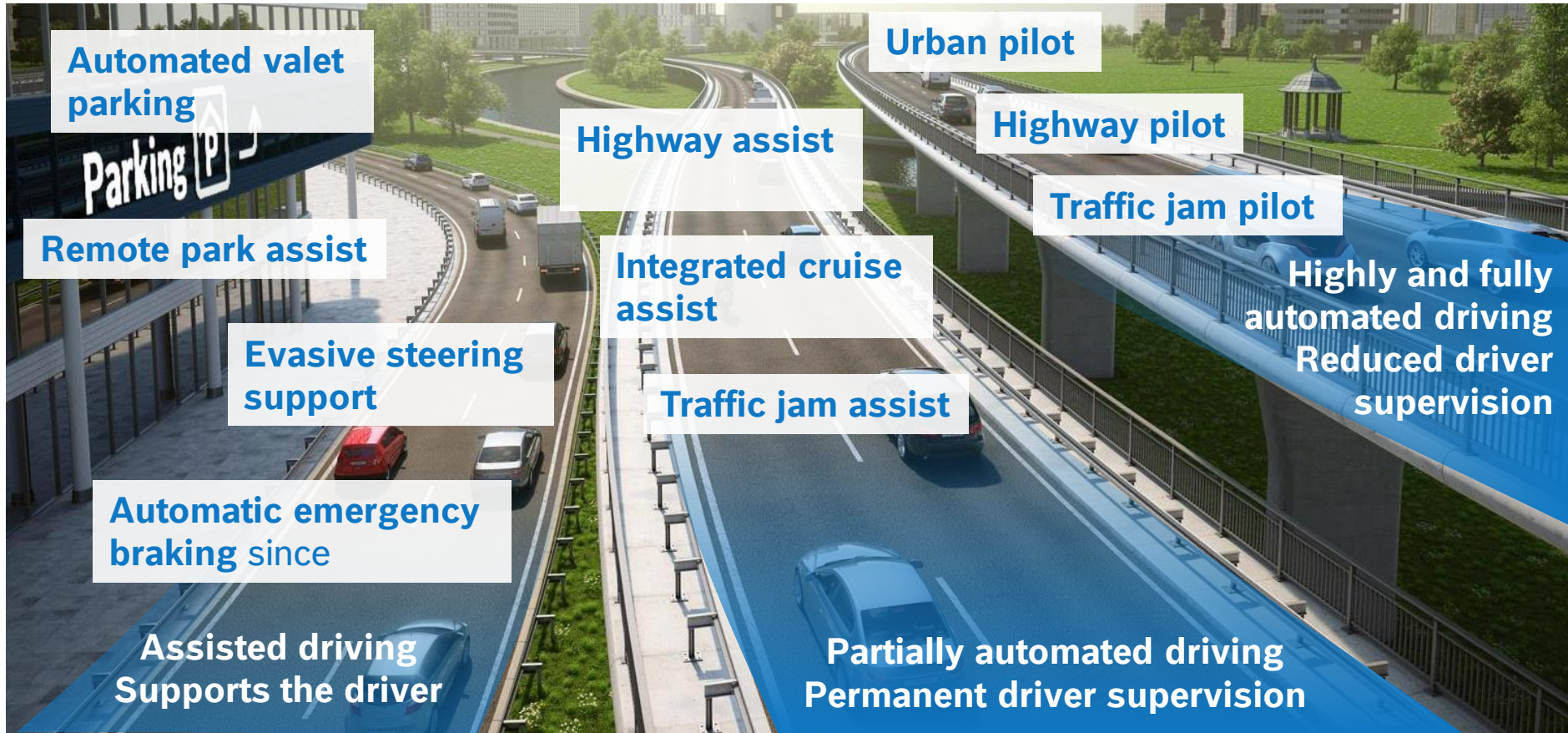
The five stages of autonomy



Sources: Evercore ISI, SAE International

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Roadmap Highly Automated Driving Functions



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Heavy rain example



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Technologies for automated driving

Surround Sensors



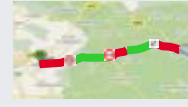
are highly robust in all use cases; fusion

Driver Monitoring



Mandatory for partly automatic

Online Map Data



is precise & up to date every moment

Perception and Localization

leads to unambiguous & comprehensive 360° environment model



Reasoning and Decision Making

allows for correct decisions, even in highly dynamic situations & at incomplete information



Motion Control

works safe, fast & precise in all dimensions



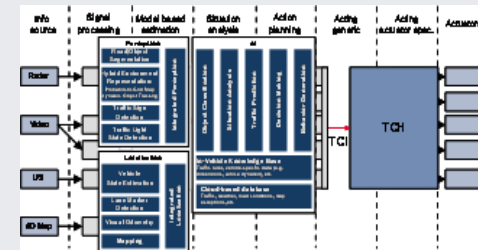
Functional Safety

guarantees high standard @ reasonable effort



Architecture

supports safety, performance & cost targets



Bosch has all necessary key technologies available and is getting them ready for market entry

Computer vision in today's intelligent cars

What are we doing in Cluj?

- Main responsibilities
 - **Software development** and pre-development for mono and stereo video sensors
 - **Computer vision**, image processing and **machine learning** algorithms development for driver assistance and automated driving



**stereo
camera**



**mono
camera**

Computer vision in today's intelligent cars

What is computer vision?

- ▶ Understand images content algorithmically
- ▶ 3D perception (measurements) from images
 - ▶ Mono cameras
 - ▶ Stereo camera
- ▶ **AI course brings you the basics knowledge for working in this field**



Computer vision in today's intelligent cars

Image representation

► Robustly detect elements in the traffic:

- Cars
- Pedestrians
- Cyclists
- Guardrail
- Lanes
- Poles
- Road signs

etc



Bi-dimensional array of values between 0 (black) and 255 (white) for 8bits per pixel

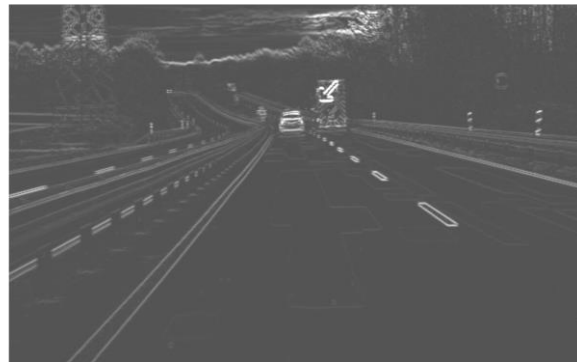
120	120	121	122	70	40
60	50	40	41	7	8
100	120	121	122	1	0
200	120	200	122	12	14
200	220	225	250	30	40

Computer vision in today's intelligent cars

Example – starting point for lane detection

► Partial derivative the math formula: $\frac{\delta F(x, y)}{\delta x} = \lim_{\epsilon \rightarrow 0} \frac{F(x + \epsilon, y) - F(x, y)}{\epsilon}$
 Concept since 1600 – Newton, Leibnitz

→ Partial derivative – for discrete values: $\frac{\delta F(x, y)}{\delta x} \approx \frac{F(x + 1, y) - F(x, y)}{1}$



-1	0	+1
-2	0	+2
-1	0	+1

Convolution with Sobel filter to get horizontal and vertical edge features

+1	+2	+1
0	0	0
-1	-2	-1

→ Lane detection can be done by clustering gradient points together based on magnitude & orientation

Computer vision in today's intelligent cars

Perception capabilities of a mono-video system

→ Example 2D information from one camera

- Traffic signs (machine learning):
 - Speed Limits
 - Stop Signs
 - Give Way Signs
 - Additional signs (Pictograms)
 - General signs (triangular, rectangular)
- Lane markings & road segmentation
- Vehicle lights / traffic lights
- Object classification (machine learning)

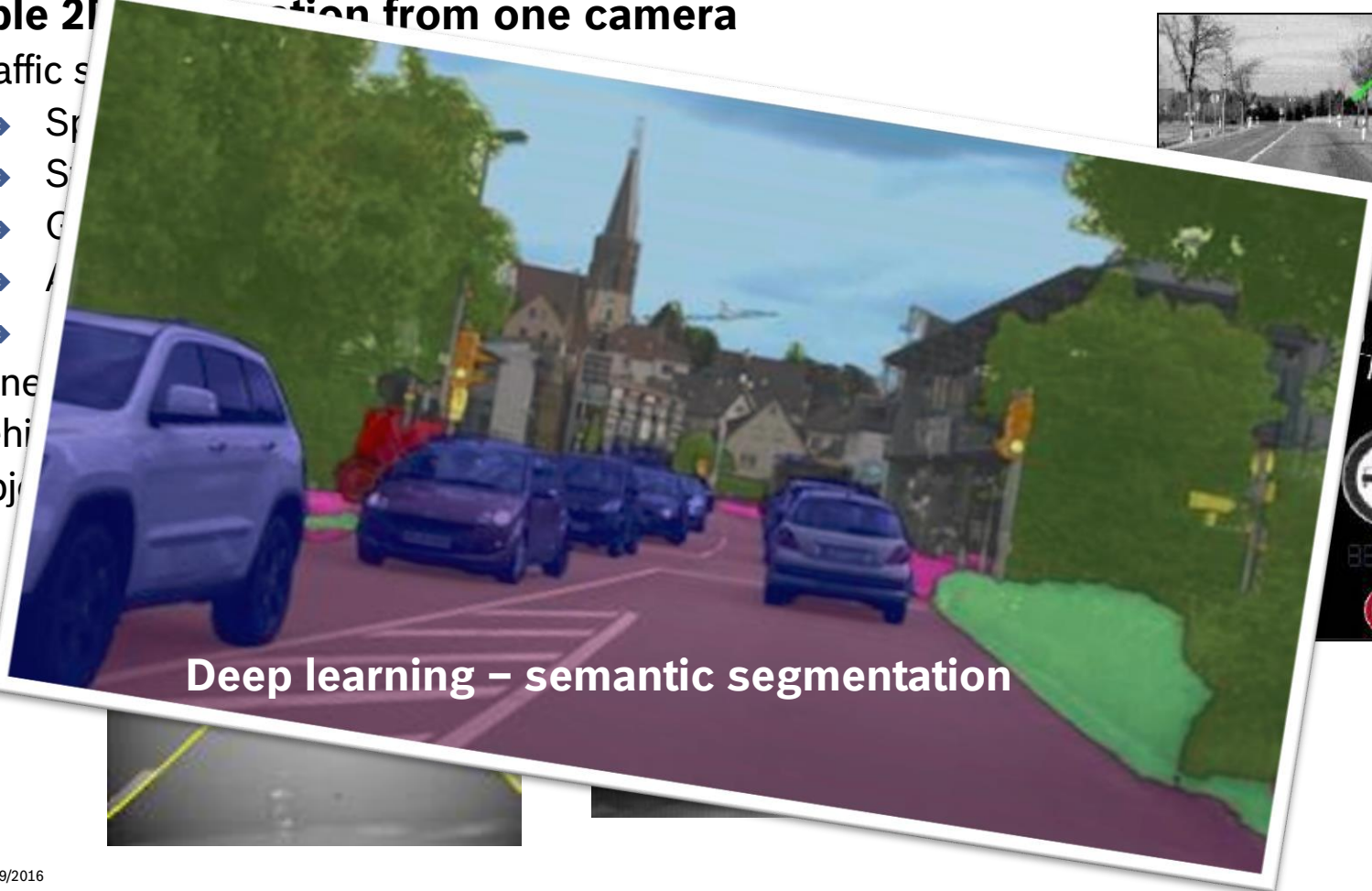


Computer vision in today's intelligent cars

Perception capabilities of a mono-video system

→ Example 2D perception from one camera

- Traffic signs
- Speed limit
- Stop
- Go
- Lane
- Vehicle
- Object



Computer vision in today's intelligent cars

Example 3D point cloud from images

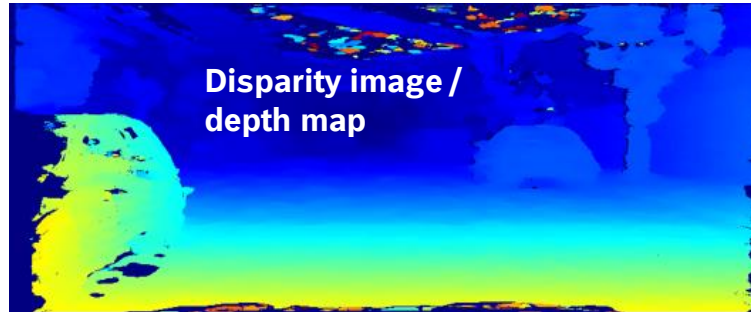


Legend

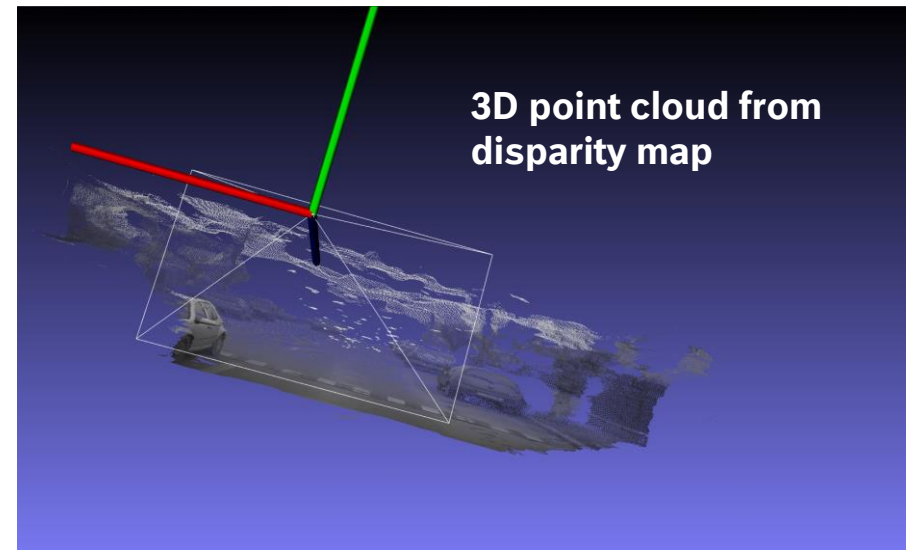
Far objects



Close objects



3D reconstruction



- ▶ Big advantage compared to other sensors: density of the data.
- ▶ Challenge: process this amount of relatively noisy data on low power embedded hardware.
- ▶ Highly optimized algorithms needed.

3D RECONSTRUCTION

General concepts

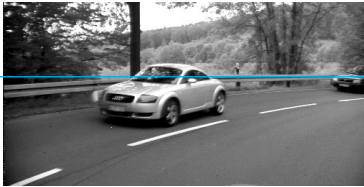
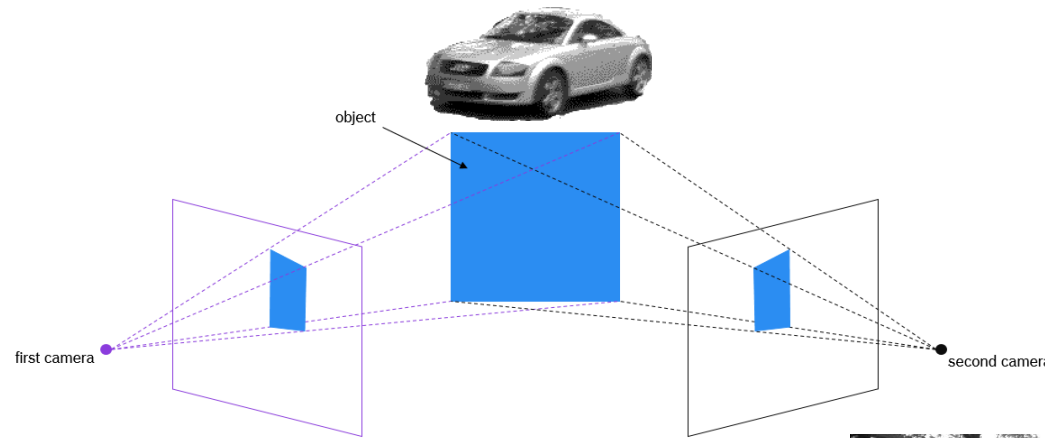


stereo
camera



mono
camera

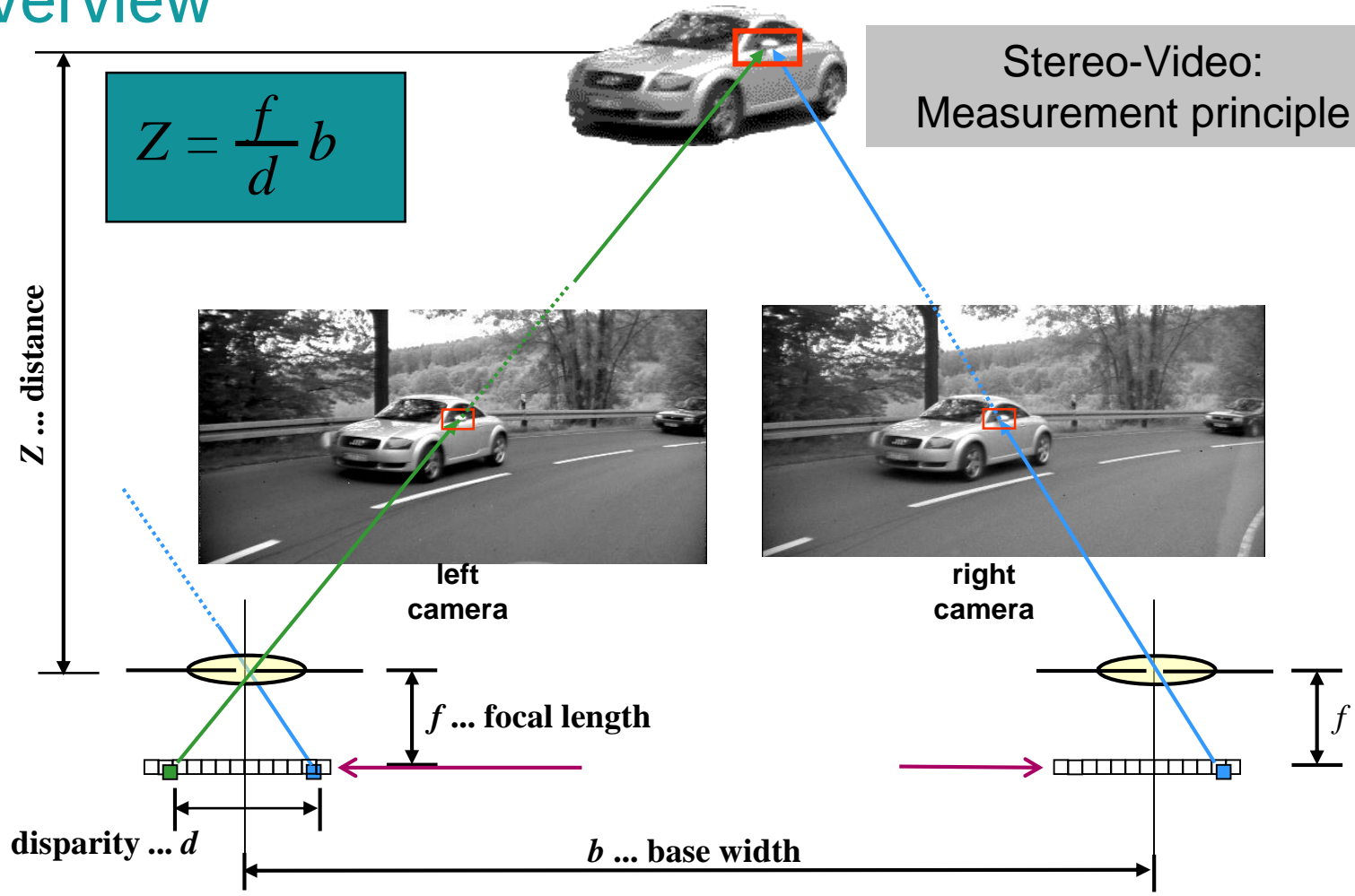
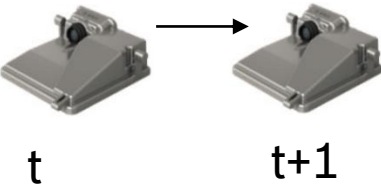
- 3D reconstruction – the process of creating the 3D shape and position of real objects from images
- in computer vision for automated driving
 - using the stereo system - two cameras from different positions, targeting the same scene
 - using the mono system - same camera, targeting the same scene at different points in time



Rectification for 3D
reconstruction => corresponding image
points are on the same scan-line

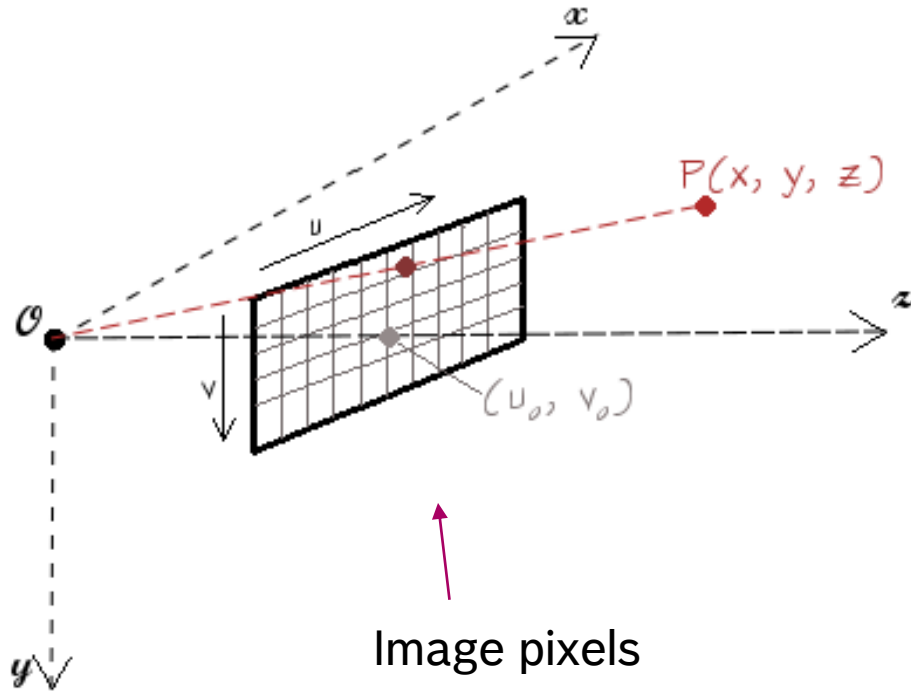
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Going 3D - overview



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Perspective projection model



O - camera centre

u_0, v_0 - principal point

u, v - image coordinates

f - $d(O, \text{principal point})$

dp_x, dp_y - size of pixels

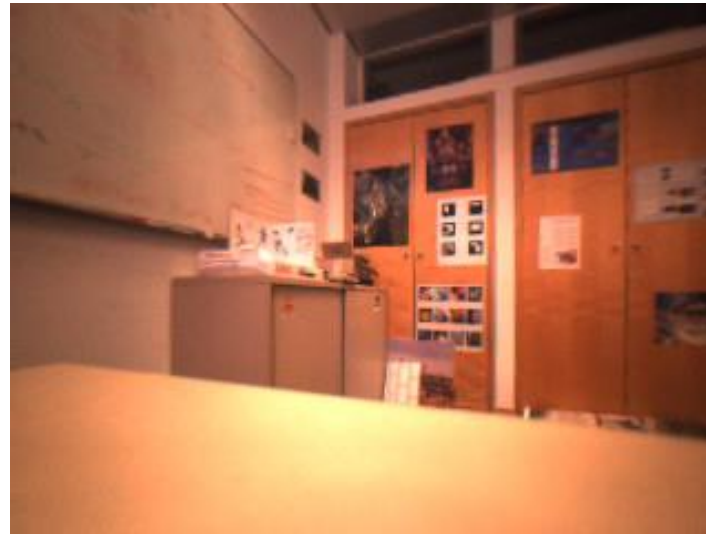
f_x, f_y - Focal length in pixels

Intrinsic calibration parameters needs to be determined by calibration:

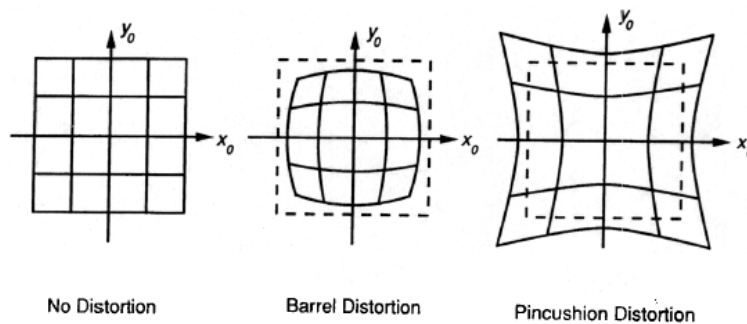
- Distortion parameters
- Principal point
- Focal length

Computer vision in today's intelligent cars

Lens distortion removal



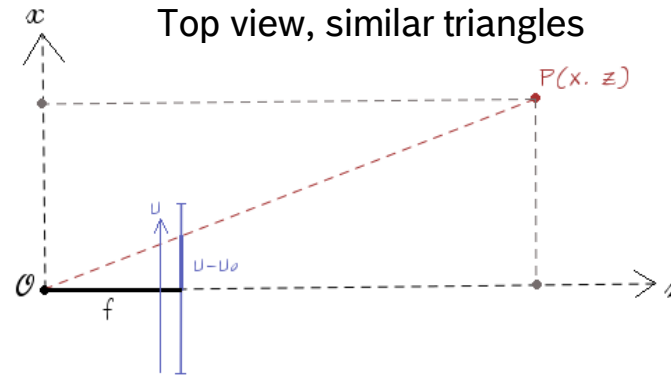
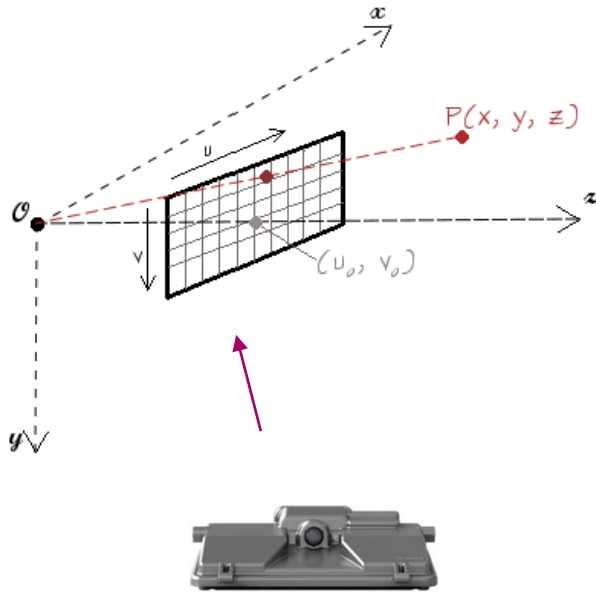
Dynamic Vision, T. Schon



- ▶ Basic idea for **offline calibration**: use a pattern with known geometry that can be detected in images and then warp the image such that the projection of the calibration pattern follows the pinhole model.
- ▶ **Online calibration** also performed in the series products.

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Going 3D



z is unknown from single image

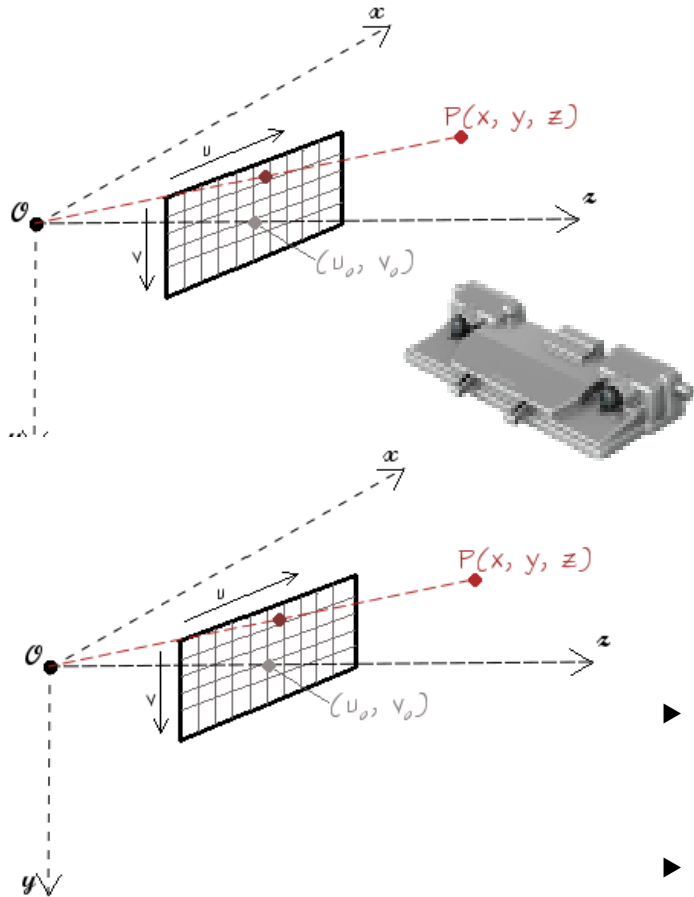
$$x = \frac{z * (u - u_0)}{f_x}$$

$$y = \frac{z * (v - v_0)}{f_y}$$

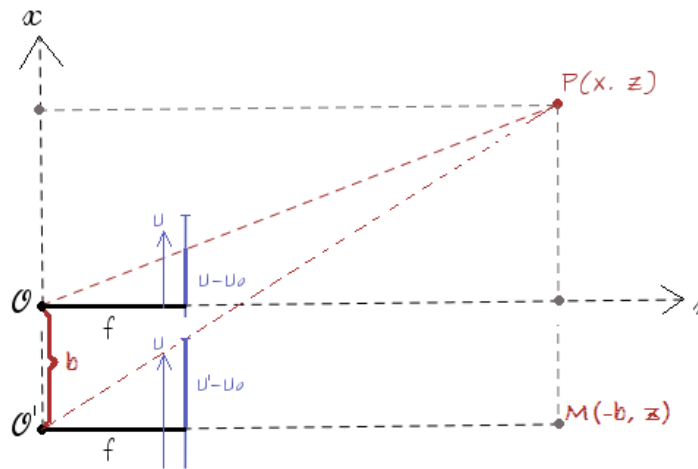
- Remark: 3D reconstruction can be achieved with lower accuracy from consecutive frames (structure from motion)

Computer vision in today's intelligent cars

Going 3D



Top view, similar triangles, 2 cameras



z is unknown from single image

$$x = \frac{z * (u - u_0)}{f_x}$$

$$y = \frac{z * (v - v_0)}{f_y}$$

$$z = \frac{b * f_x}{d}$$

- ▶ For precision:
 - ▶ Matching between left & right image are important.
 - ▶ Alignment between left & right camera is important
- ▶ Remark: 3D reconstruction can also be achieved with lower accuracy form consecutive frames od a mono camera with a similar geometric model (**structure from motion**)

Computer vision in today's intelligent cars

Example 3D point cloud from images

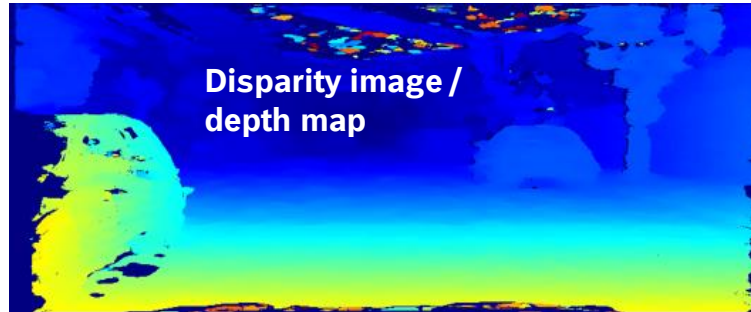


Legend

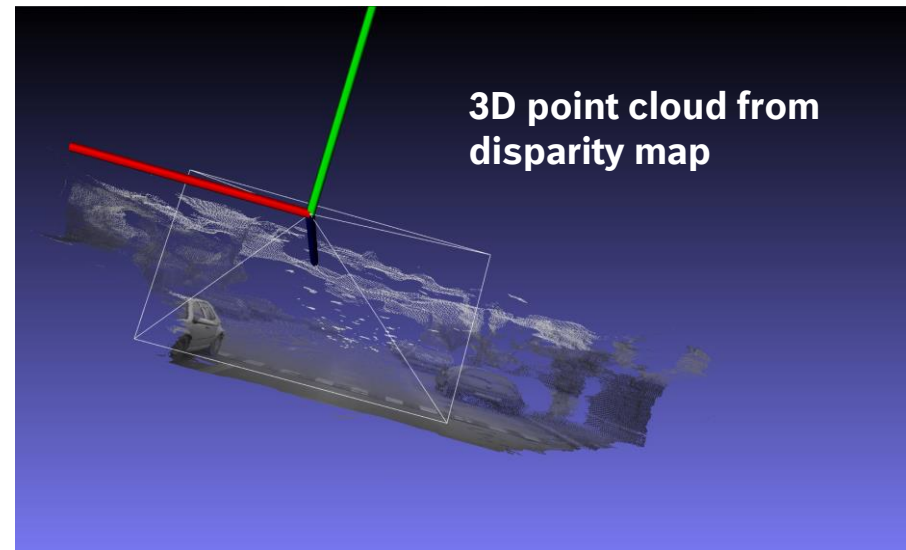
Far objects



Close objects



3D reconstruction



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- ▶ Challenge: process this amount of relatively noisy data on low power embedded hardware.
- ▶ Highly optimized algorithms needed.

Computer vision in today's intelligent cars

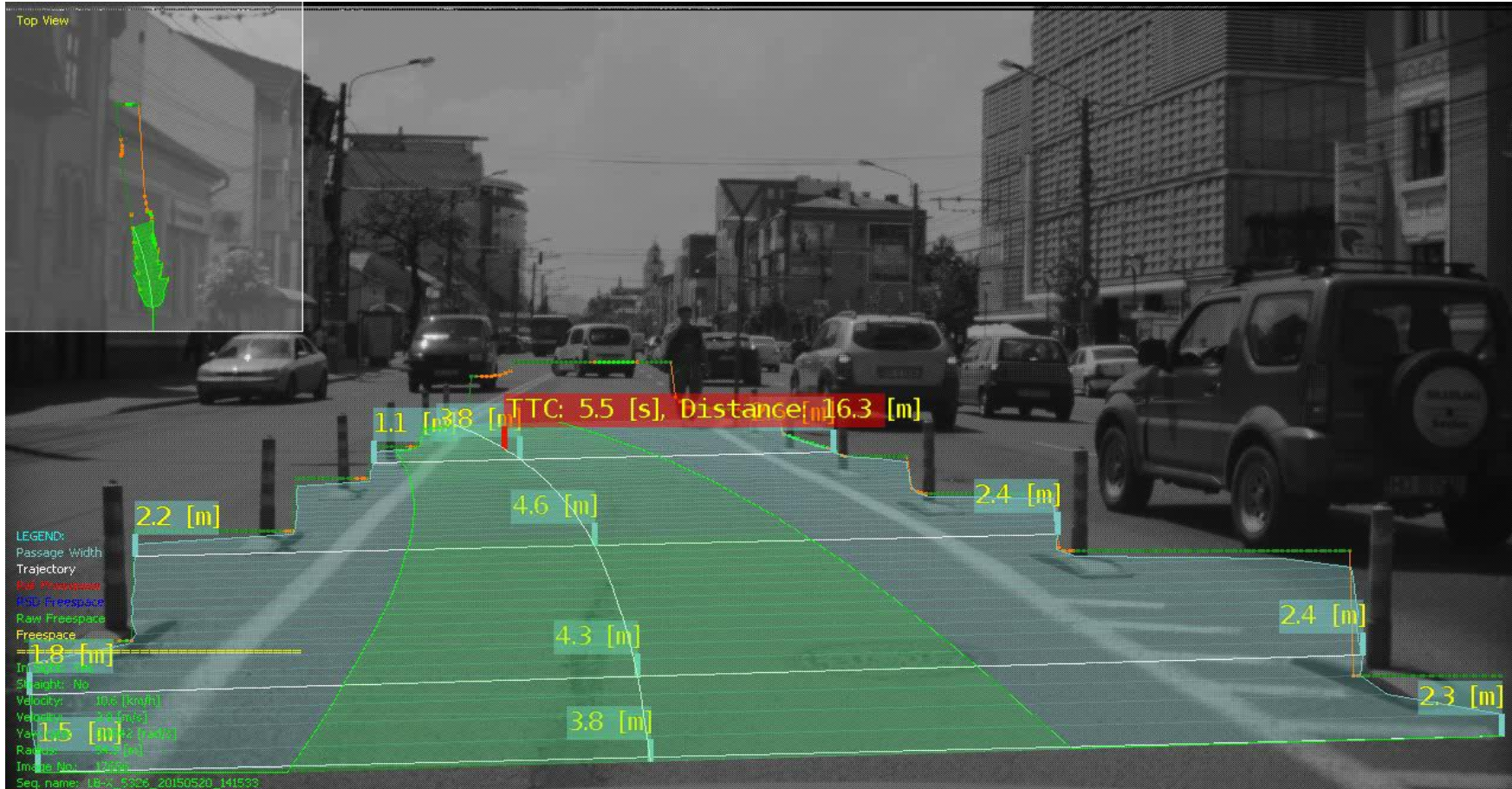
Perception capabilities of a stereo video system

- complete **3D environment information**
 - path topography (slopes, bumps, potholes, lane grooves)
 - height obstructions (bridges, trees)
 - path width (narrow passages, parking)
- complete **3D object information**
 - cars, trucks, cycles, pedestrians, ...
 - guard rails, poles, walls, trees, animals,...



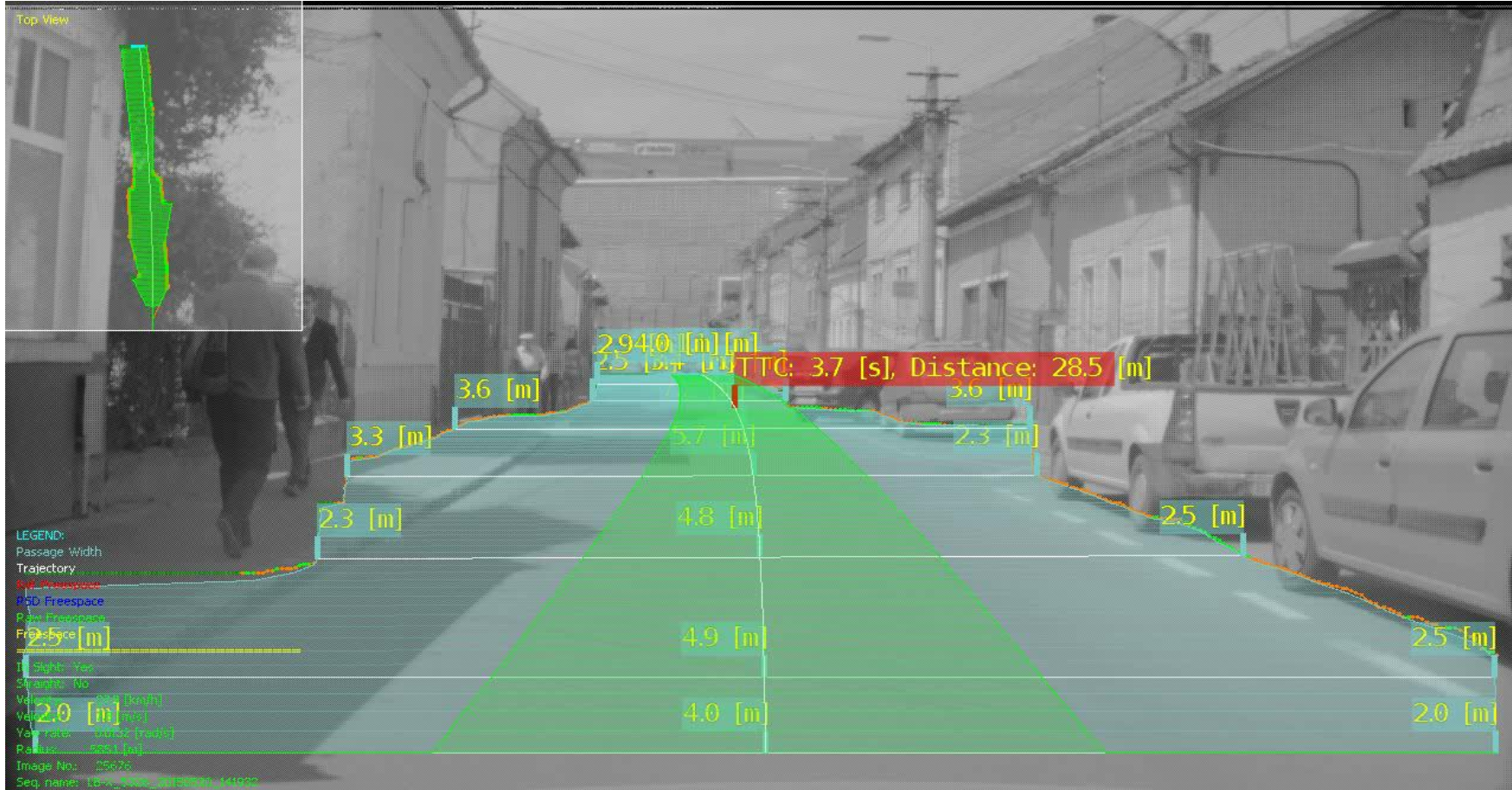
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Example free-space measurements



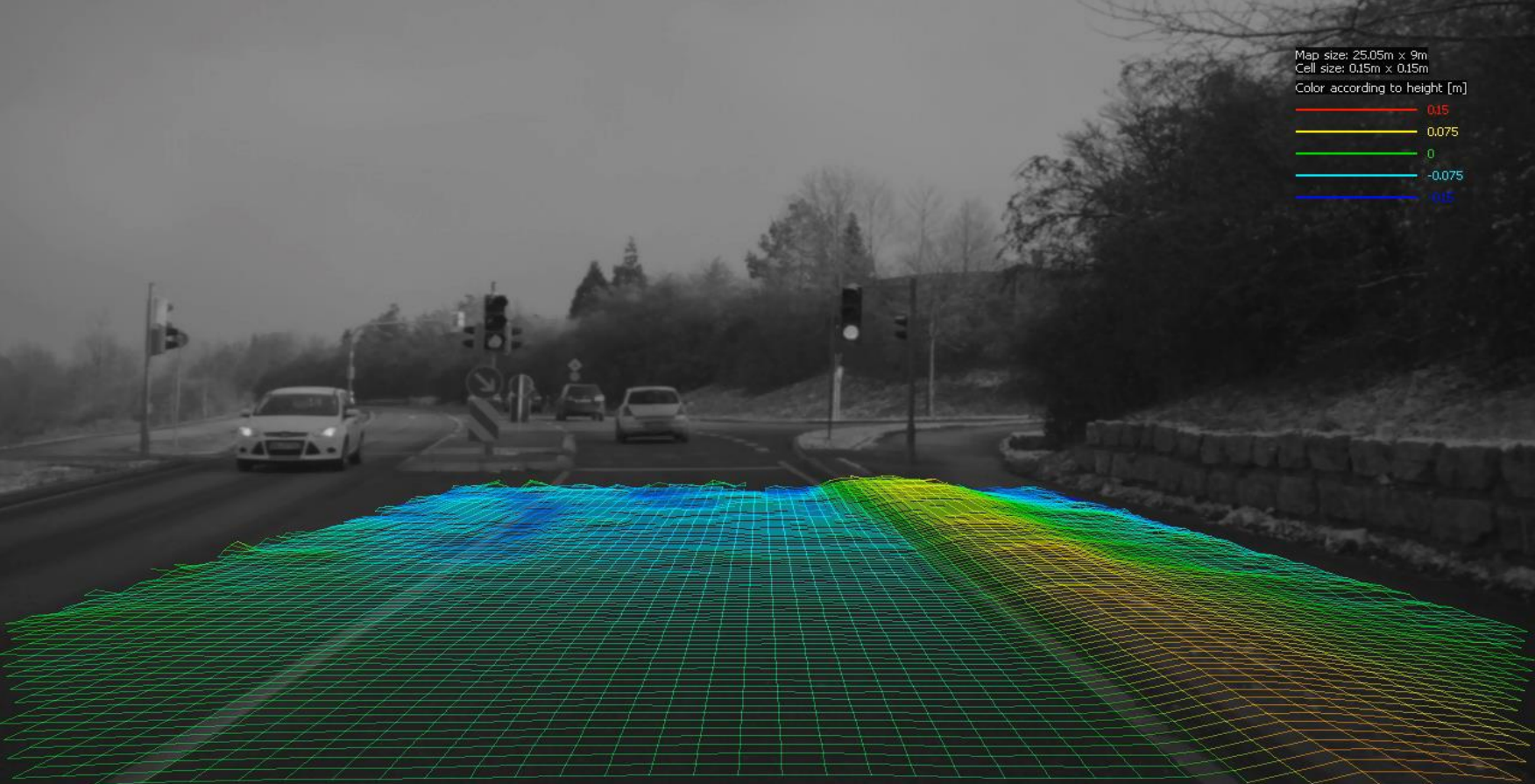
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Example free-space measurements on narrow roads



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Example surface topography measurements



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Height obstructions



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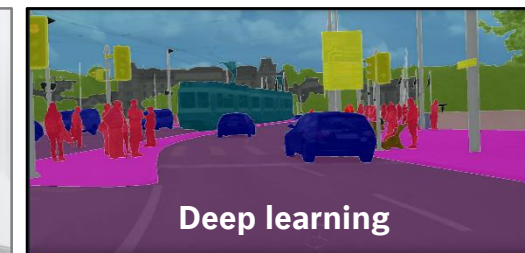
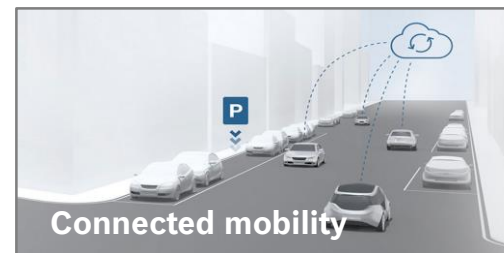
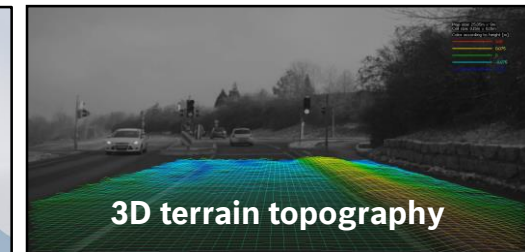
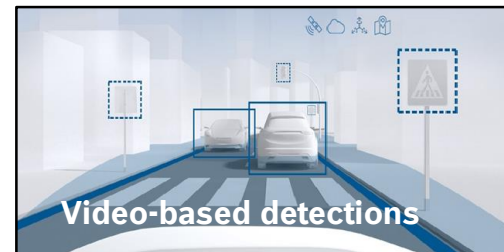
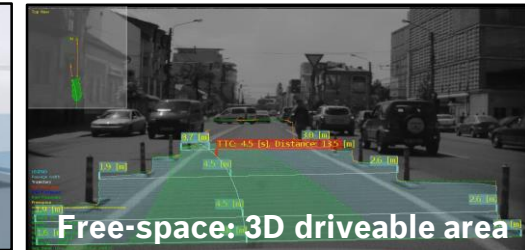
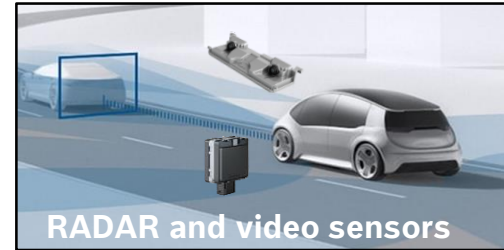
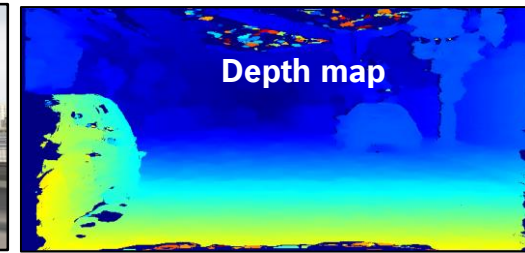
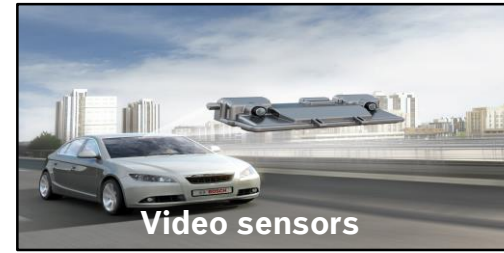
Student topics

► Computer vision topics for summer 2018

- (1) Algorithms development on automotive **parallel computing** architectures (like a GPU)
- (2) **Tracking and data fusion** (radar, video, LiDAR)
- (3) Algorithms for **3D data processing**:
 - All kind of objects detection from depth map (3D objects)
- (4) **Machine learning and deep learning**
- **Main technologies**: C/C++, python, Matlab, visualizations (ex: QT, OpenGL), Deep Learning (ex: TensorFlow)

► Connectivity & tools development

- **Full stack web** development (Java or C#/.Net, Angular, JavaScript, IoT)
- **Cloud Software** development (Spark, Hadoop, data analytics and visualization)





Thank you!