

### Organizer: Angelos Amditis (ICCS) Moderator: Anastasia Bolovinou (ICCS)

Speaker: Abhinav Valada, Professor (University of Freiburg Speaker: Antti Kangasrääsiö, Head of Research, Sensible 4 Speaker: Eren Aksoy Professor, Halmstad University





### Intro by Moderator







### Part A: **Speakers short** pitches (up to 30' total)



Part B: Roundtable discussion and QnAs from the audience (up to 20' total)







### Perception based on Radar, Camera, LiDAR Scene understanding (2D/3D visual + audio) Safety of the intended functionality => Fusion, Redundancy by design **EVENTS**

till recently...

more recently...









Funded by the **European Union** 

"EVENTS project is funded by the European Union, under grant agreement No 101069614. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or European Commission. Neither the European Union nor the granting authority can be held responsible for them."



### Replaying real-world recordings

Sim

Sim

Realworld





Game Changer

# Intro by moderator (2/4)

### Perception testing Open Road Trials, Closed Test track trials, Chamber trials Virtual testing and synthetic data augmentation New Benchmark Datasets (raw data, annotated objects, annotated scenarios)







\*Source: Multi-weather city: Adverse weather stacking for autonomous driving, ICCV, Oxf Brookes Uni



ORGANISED



HOSTED BY:









Autonomous vehicle Other vehicle

Vehicle traffic light Other traffic light

- 1 Incoming cycle lane
- 2 Incoming lane
- 3 Incoming lane
- 4 Outgoing lane
- 5 Outgoing cycle lane
- 6 Vehicle lane
- 7 Incoming lane

forAutonomous Driving, PAMI, 2022





e Game Changer

### **Object and events detection and their representations:**

### 2D-bounding box, 3D-bounding box segmented instances, pixel-based Lanes, Driving free space actions, events

Source: AD PerDevKit: An Autonomous Driving Perception Development Kit using CARLA simulator and ROS















**3D** 







# Intro by moderator (4/4) Our panel participants



Anastasia Bolovinou Research engineer, Institute of Communication and Computer Systems (ICCS)





Antti Kangasrääsiö Head of Research at Sensible 4, a Finnish all-weather autonomous driving startup



Eren Erdal Aksoy Professor (Associate) at Halmstad University Germany



Abhinav Valada Assistant Professor and Director of the Robot Learning Lab at the University of Freiburg, Germany.









![](_page_6_Picture_1.jpeg)

![](_page_6_Picture_2.jpeg)

### Semantics-aware Multi-modal Domain Translation: Going from LiDAR to Camera

![](_page_6_Picture_4.jpeg)

### An industry view: Dealing with environment change in all-weather perception for autonomous driving.

![](_page_6_Picture_6.jpeg)

# regions.

![](_page_6_Picture_8.jpeg)

![](_page_6_Picture_9.jpeg)

![](_page_6_Picture_10.jpeg)

![](_page_6_Picture_11.jpeg)

### Training an amodal panoptic segmentation module able to understand occluded

![](_page_6_Picture_13.jpeg)

![](_page_6_Picture_14.jpeg)

![](_page_6_Picture_15.jpeg)

![](_page_7_Picture_0.jpeg)

# Eren Erdal Aksoy

Associate Professor Halmstad University Sweden

![](_page_7_Picture_4.jpeg)

# **Uncertainty-aware Semantic Perception with LiDAR-only Data**

![](_page_8_Picture_1.jpeg)

![](_page_8_Picture_2.jpeg)

![](_page_8_Picture_3.jpeg)

![](_page_8_Picture_4.jpeg)

### Interpretability makes clear what the system "knows" while uncertainty awareness reveals what the system does not "know."

![](_page_8_Picture_6.jpeg)

![](_page_8_Picture_7.jpeg)

Cortinhal et al., "Salsanext: Fast, Uncertainty-aware Semantic Segmentation of LiDAR Point Clouds for Autonomous Driving," in International Symposium on Visual Computing, 2020.

![](_page_8_Picture_9.jpeg)

![](_page_8_Picture_10.jpeg)

![](_page_8_Picture_11.jpeg)

![](_page_8_Picture_12.jpeg)

![](_page_8_Picture_13.jpeg)

Game Changer

![](_page_9_Picture_1.jpeg)

![](_page_9_Picture_2.jpeg)

![](_page_9_Picture_3.jpeg)

![](_page_9_Picture_4.jpeg)

![](_page_9_Figure_6.jpeg)

IJCAI - AI4AD 2021

![](_page_9_Picture_11.jpeg)

The Game Changer.

![](_page_10_Picture_1.jpeg)

![](_page_10_Picture_2.jpeg)

![](_page_10_Picture_3.jpeg)

![](_page_10_Picture_4.jpeg)

IJCAI - AI4AD 2021

![](_page_10_Picture_9.jpeg)

Game Changer.

![](_page_11_Picture_0.jpeg)

![](_page_11_Picture_1.jpeg)

![](_page_11_Picture_2.jpeg)

![](_page_11_Picture_3.jpeg)

![](_page_11_Picture_4.jpeg)

![](_page_11_Figure_7.jpeg)

![](_page_11_Picture_8.jpeg)

![](_page_11_Picture_9.jpeg)

Cortinhal et al., "Depth- and Semantics-aware Multi-modal Domain Translation" in arXiv, 2023.

![](_page_11_Picture_13.jpeg)

![](_page_11_Picture_14.jpeg)

![](_page_11_Picture_15.jpeg)

![](_page_11_Picture_16.jpeg)

![](_page_11_Picture_17.jpeg)

![](_page_11_Picture_18.jpeg)

![](_page_11_Picture_19.jpeg)

The Game Changer.

![](_page_12_Picture_0.jpeg)

![](_page_12_Picture_1.jpeg)

ORGANISED BY: \*\*\*\* ERTICO ITS EUROPE

![](_page_12_Picture_3.jpeg)

![](_page_12_Picture_4.jpeg)

![](_page_12_Picture_6.jpeg)

![](_page_12_Picture_7.jpeg)

on Augmented LiDAR

![](_page_12_Picture_9.jpeg)

![](_page_12_Picture_10.jpeg)

# ROADVIEW **Robust Automated Driving in Extreme Weather** HORIZON-CL5-2021-D6-01-01: More powerful and reliable on-board perception and decisionmaking technologies addressing complex environmental conditions (Innovation Action) Is partners, ~10 M Euro, TRL 6-7, 2022-2026

![](_page_13_Picture_1.jpeg)

![](_page_13_Picture_2.jpeg)

FORD OTOSAN

![](_page_13_Picture_4.jpeg)

![](_page_13_Picture_5.jpeg)

![](_page_13_Picture_6.jpeg)

![](_page_13_Picture_7.jpeg)

![](_page_13_Picture_8.jpeg)

![](_page_13_Picture_9.jpeg)

![](_page_13_Picture_10.jpeg)

Vti LAPIN AMK

![](_page_13_Picture_13.jpeg)

Funded by the European Union Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or the European Climate, Infrastructure and Environment Executive Agency (CINEA). Neither the European Union nor the granting authority can be held responsible for them.

![](_page_13_Picture_16.jpeg)

![](_page_13_Picture_17.jpeg)

he Game Changer

![](_page_14_Picture_0.jpeg)

![](_page_14_Picture_1.jpeg)

![](_page_14_Picture_2.jpeg)

### **Robust Automated Driving in Extreme Weather** De-noising LiDAR Point Clouds for Object Detection in Extreme Weather

![](_page_14_Picture_5.jpeg)

![](_page_14_Picture_7.jpeg)

Funded by the European Union

Funded by the European Union. Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or the European Climate, Infrastructure and Environment Executive Agency (CINEA). Neither the European Union nor the granting authority can be held responsible for them.

### **Snowy LiDAR Point Cloud**

	mAP	
	Easy	Med
PointPillars (Baseline)	83.4	76.0
Augmented LiDAR	85.0	77.6

### **Filtered LiDAR Point Cloud**

![](_page_14_Figure_14.jpeg)

![](_page_14_Picture_15.jpeg)

![](_page_14_Picture_16.jpeg)

he Game Changer.

# ITS: The Game Changer.

The second s

![](_page_15_Picture_1.jpeg)

![](_page_15_Picture_2.jpeg)

### EUROPEAN CONGRESS LISBON, PORTUGAL 22-24 MAY 2023

![](_page_15_Picture_5.jpeg)

![](_page_16_Picture_0.jpeg)

![](_page_16_Picture_1.jpeg)

![](_page_16_Picture_3.jpeg)

### An industry view: Dealing with environment change in all-weather perception for autonomous driving.

![](_page_16_Picture_5.jpeg)

![](_page_16_Picture_7.jpeg)

![](_page_16_Picture_8.jpeg)

![](_page_16_Picture_9.jpeg)

![](_page_16_Picture_10.jpeg)

![](_page_16_Picture_13.jpeg)

![](_page_16_Picture_15.jpeg)

![](_page_16_Picture_16.jpeg)

**ITS European Congress 2023 Lisbon** Dr. Antti Kangasrääsiö, Head of Research Sensible 4, Finland

![](_page_17_Picture_2.jpeg)

![](_page_17_Picture_3.jpeg)

![](_page_17_Picture_4.jpeg)

# An Industry View: Dealing with Environment Change

![](_page_17_Picture_6.jpeg)

Funded by the European Union

# Our mission is to enable autonomous driving everywhere

People, businesses and society should have the benefits of autonomous driving everywhere and everytime. We make that happen. Our unique software will take autonomous vehicles to places where others can't operate.

![](_page_18_Picture_2.jpeg)

Key customers

![](_page_18_Picture_5.jpeg)

![](_page_18_Picture_6.jpeg)

# All-weather Perception The environment is always changing

Changes happen both suddenly and over a long time Weather around the vehicle: rain, slush, snow, fog Accumulation of ice and snow over the winter Seasonal changes in trees and other roadside plants Roadwork and new buildings

Perception systems need to adapt to these challenges Measurement noise due to particles in the air Maps of the environment lose precision over time

![](_page_19_Picture_5.jpeg)

![](_page_19_Picture_6.jpeg)

# Robust Object Detection Always aware of the surroundings

### Smart filtering

Remove noise from the sensor signals Focus detection on where it matters Efficient high-speed filtering methods

### Safe Object Detection

Identify persistent parts of an object Robust to noise in the sensor signal Robust to partial occlusion of the object Low-latency updates

![](_page_20_Picture_5.jpeg)

![](_page_20_Picture_7.jpeg)

![](_page_20_Picture_32.jpeg)

# Robust Localization Never losing our way

### Probabilistic mapping

High-definition 3D maps of the environment
 Enables efficient probabilistic localization

### **Robust Localization**

Probability distribution over possible locations
 Efficient high-frequency updates

### Benefits

. Tolerates large changes in the environment . Tolerates high measurement noise

![](_page_21_Picture_7.jpeg)

![](_page_21_Picture_8.jpeg)

# **Research Collaboration** Defining the methods of the future

![](_page_22_Picture_1.jpeg)

Robust automated driving in extreme weather conditions Visibility estimation Friction estimation Safe control Safe navigation

We are always open for new collaboration!

![](_page_22_Picture_4.jpeg)

![](_page_22_Picture_5.jpeg)

# Scene analysis Robust perception Robust localization

Robust perception and decision making for autonomous driving Robust motion planning

![](_page_22_Picture_8.jpeg)

![](_page_22_Picture_9.jpeg)

![](_page_22_Picture_10.jpeg)

![](_page_22_Picture_11.jpeg)

Funded by the European Union

![](_page_22_Picture_13.jpeg)

### **Robust & automated mapping** and localization methods Autonomous mapping HD maps

Map updating

Map quality assessment

![](_page_23_Picture_0.jpeg)

![](_page_23_Picture_1.jpeg)

![](_page_23_Picture_2.jpeg)

![](_page_23_Picture_3.jpeg)

![](_page_24_Picture_0.jpeg)

![](_page_24_Picture_1.jpeg)

![](_page_24_Picture_2.jpeg)

![](_page_24_Picture_4.jpeg)

![](_page_24_Picture_6.jpeg)

# regions.

![](_page_24_Picture_8.jpeg)

![](_page_24_Picture_9.jpeg)

![](_page_24_Picture_10.jpeg)

![](_page_24_Picture_11.jpeg)

### Training an amodal panoptic segmentation module able to understand occluded

![](_page_24_Picture_15.jpeg)

![](_page_24_Picture_17.jpeg)

![](_page_24_Picture_18.jpeg)

### universitätfreiburg

### Prof. Dr. Abhinav Valada

Robot Learning Lab

03 May 2023

![](_page_25_Picture_5.jpeg)

# **Amodal Panoptic Segmentation**

![](_page_25_Picture_8.jpeg)

![](_page_25_Picture_9.jpeg)

![](_page_25_Picture_10.jpeg)

![](_page_25_Picture_11.jpeg)

![](_page_25_Picture_12.jpeg)

### What Is Panoptic Segmentation?

# Pixel-level semantic segmentation of stuff classes and instance segmentation of thing classes

![](_page_26_Picture_2.jpeg)

![](_page_26_Picture_3.jpeg)

![](_page_26_Picture_4.jpeg)

### How do we go beyond?

![](_page_26_Picture_6.jpeg)

### **Cityscapes Panoptic Quality**

![](_page_26_Picture_9.jpeg)

![](_page_26_Picture_10.jpeg)

### How Do Humans Perceive Objects?

# Amodal Perception: Humans perceive objects as a whole regardless of partial occlusion Connects our perception of the world to its cognitive understanding Autonomous Systems: Still limited to modal perception

![](_page_27_Picture_2.jpeg)

![](_page_27_Picture_3.jpeg)

![](_page_27_Picture_4.jpeg)

![](_page_27_Picture_5.jpeg)

![](_page_27_Picture_6.jpeg)

![](_page_27_Picture_7.jpeg)

![](_page_28_Picture_0.jpeg)

### We bridge this gap by formulating the novel amodal panoptic segmentation task

![](_page_28_Picture_2.jpeg)

![](_page_28_Picture_3.jpeg)

![](_page_28_Picture_4.jpeg)

![](_page_28_Picture_5.jpeg)

![](_page_28_Picture_6.jpeg)

![](_page_28_Picture_8.jpeg)

# R. Mohan, A. Valada: CVPR'22

![](_page_28_Picture_10.jpeg)

![](_page_28_Picture_11.jpeg)

### What Is Amodal Panoptic Segmentation?

# Simultaneously predict pixel-wise semantic segmentation labels of visible regions of *stuff* classes such as buildings, road, sidewalk, etc

![](_page_29_Picture_2.jpeg)

![](_page_29_Picture_3.jpeg)

![](_page_29_Picture_4.jpeg)

![](_page_29_Picture_5.jpeg)

![](_page_29_Picture_6.jpeg)

![](_page_29_Picture_7.jpeg)

![](_page_29_Picture_8.jpeg)

![](_page_29_Picture_9.jpeg)

### What Is Amodal Panoptic Segmentation?

### and instance segmentation labels of both the visible and occluded regions of thing classes such as people, car, truck etc.

![](_page_30_Picture_2.jpeg)

![](_page_30_Picture_3.jpeg)

![](_page_30_Picture_4.jpeg)

![](_page_30_Picture_5.jpeg)

![](_page_30_Picture_6.jpeg)

![](_page_30_Picture_7.jpeg)

![](_page_30_Picture_8.jpeg)

![](_page_30_Picture_9.jpeg)

![](_page_30_Picture_10.jpeg)

![](_page_30_Picture_11.jpeg)

# Why Amodal Panoptic Segmentation?

# Perceiving the entire structure of traffic participants at all times, regardless of occlusions, will minimize the risk of accidents and extend the capabilities of autonomous robots

![](_page_31_Picture_2.jpeg)

![](_page_31_Picture_3.jpeg)

![](_page_31_Picture_4.jpeg)

![](_page_31_Picture_5.jpeg)

![](_page_31_Picture_6.jpeg)

![](_page_31_Picture_7.jpeg)

![](_page_31_Picture_8.jpeg)

![](_page_31_Picture_9.jpeg)

### **Datasets and Metrics**

### We extend two challenging urban automated driving datasets with amodal panoptic annotations

![](_page_32_Picture_2.jpeg)

![](_page_32_Picture_3.jpeg)

![](_page_32_Picture_4.jpeg)

![](_page_32_Picture_5.jpeg)

![](_page_32_Picture_6.jpeg)

![](_page_32_Picture_7.jpeg)

### BDD100K-APS 3,000 annotations 10 stuff classes, 6 thing classes

![](_page_32_Picture_9.jpeg)

KITTI-360-APS 61,168 annotations

# http://amodal-panoptic.cs.uni-freiburg.de

# 10 stuff classes, 7 thing classes

![](_page_32_Picture_14.jpeg)

![](_page_32_Picture_15.jpeg)

### **Two Novel Amodal Panoptic Segmentation Methods**

### We propose two methods Proposal-Based: APSNet R. Mohan, A. Valada: CVPR'22 Proposal-Free: PAPS R. Mohan, A. Valada: RA-L'22

![](_page_33_Figure_2.jpeg)

### Proposal-Based APSNet Architecture

![](_page_33_Picture_4.jpeg)

![](_page_33_Picture_5.jpeg)

![](_page_33_Figure_6.jpeg)

### Proposal-Free PAPS Architecture

![](_page_33_Picture_9.jpeg)

![](_page_33_Picture_10.jpeg)

![](_page_34_Picture_0.jpeg)

![](_page_34_Picture_1.jpeg)

![](_page_34_Picture_2.jpeg)

### **Amodal Panoptic Segmentation Results**

![](_page_34_Picture_4.jpeg)

![](_page_34_Picture_5.jpeg)

![](_page_34_Picture_6.jpeg)

# **Upcoming: Exploiting Amodality for Various Tasks**

### Learning amodal panoptic segmentation with Depth prediction Optical flow Vehicle pose estimation Multi-object tracking

![](_page_35_Picture_2.jpeg)

![](_page_35_Picture_3.jpeg)

![](_page_35_Picture_4.jpeg)

![](_page_35_Picture_5.jpeg)

![](_page_35_Picture_6.jpeg)

![](_page_35_Picture_7.jpeg)

![](_page_35_Picture_9.jpeg)

![](_page_35_Picture_11.jpeg)

![](_page_35_Picture_12.jpeg)

![](_page_35_Picture_13.jpeg)

![](_page_35_Picture_14.jpeg)

# **Upcoming: AmodalSynthDrive Dataset**

![](_page_36_Picture_1.jpeg)

![](_page_36_Picture_2.jpeg)

![](_page_36_Picture_3.jpeg)

![](_page_36_Picture_4.jpeg)

![](_page_36_Picture_5.jpeg)

universitätfreiburg

### http://amodal-panoptic.cs.uni-freiburg.de

# Thank you for your attention!

SIS 60: Challenges of multi modal ML-based perception development & testing for automated driving applications

### Moderator: Anastasia Bolovinou (ICCS) Speaker: Abhinav Valada, Professor (University of Freiburg) Discussion Speaker: Antti Kangasrääsiö, Head of Research, Sensible 4 Time Speaker: Eren Aksoy Professor, Halmstad University

# The Game Changer.

![](_page_38_Picture_3.jpeg)

![](_page_38_Picture_5.jpeg)

### EUROPEAN CONGRESS LISBON, PORTUGAL 22-24 MAY 2023

![](_page_38_Picture_8.jpeg)

### **To Eren**

- understanding for AVs in contrast to indoor robotics?
- thoughts on these?

### **To Antti**

- or are these two problems fundamentally different?
- sensors' FoV is possible via V2X communications?
- next open road test is scheduled?

### **To Abhinav**

- 3.

ORGANISED ERTICO

![](_page_39_Picture_12.jpeg)

![](_page_39_Picture_13.jpeg)

![](_page_39_Picture_14.jpeg)

One big difference going from robotics to on-road vehicles is the speed. Based on your background on robots' perception how different is the task of semantic scene

In your latest publication you study LIDAR to RGB-D image translation based on semantic segmented scene produced by LIDAR point cloud processing and as pinpointed in the paper this could be extremely useful in solving conflicts between lidar and camera perception (similar to early fusion) but also it can be used to augment the two modalities with information from the other modality (similar to late fusion). In ASAM OSI they standardize the object level representation from different sensors. In a sense what you proposed could also mean that it is important to consider semantic segmented images as a mid-level representation in multi-modal settings. What are your

Based on your experience with vision-based scene understanding, which visual features are most helpful and which methods can be used to understand spatial relationships among objects in a scene? Pose seems as a good feature for action but not so often taken into account in scene understanding. Would that be a good candidate feature for revealing occluded scene objects identity?

ISO 22737 is the first standard on Low-Speed-ADS and in there we read that the vehicle also needs to be able to observe pedestrians and cyclists approaching the vehicle's path, even if they'd be partly covered. Occlusions could be due to other objects closer to the FoV or even due to environmental particles in the air like rain or snow. Do you think that perception algorithms trained to work well under rain or fog could work equally well for occluded objects due to other objects/road users' presence

Informed motion planning in complex urban intersections is challenging even in low speeds. Does Sensible 4 investigates solutions where perception beyond on-board

During testing under adverse weather conditions it is expected that deviations from what expected can occur in terms of vehicle behavior. How easy is to know, based on the logged data, whether this was due to a sensor fault, perception sw fault and what is your process to upgrade your vehicle SW/HW before the

How important do you think it is to be able to train ML/DL modules offline on synthetic datasets, considering also the fact that synthetic data can produce groundtruths that are 100% accurate? Have you tried to compare training only on synthetic dataset versus real-world? What is the progress in research community with designing metrics to quantify the ML/DL modules performance in an interpretable manner? Which expertise does that need? (is it only computer scientists/engineers or human experts too?) How close are we to see these panoptic/amodal segmentation modules running in real time within automotive embedded PCs/ECUs? Starting from the fact that your approach achieves SoA results in image semantic segmentation, have you tried to feed your perception module output to an AD controller to argue about the potential of amodal PS in better AD decision making?

![](_page_39_Picture_22.jpeg)

![](_page_39_Picture_23.jpeg)

![](_page_39_Picture_24.jpeg)