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# **Robot Learning Seminar WS 2023**

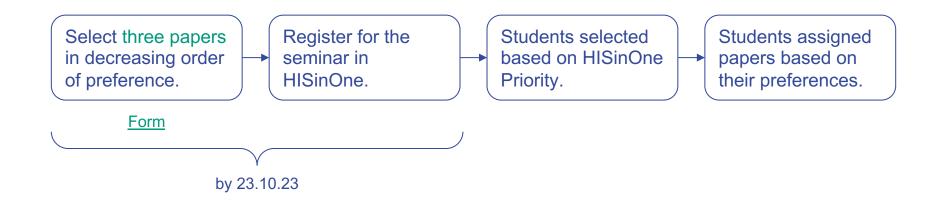
Julia Hindel

Robot Learning Lab

20 October 2023



### **Enrollment Procedure**



https://rl.uni-freiburg.de/teaching/ws23/seminar-robot-learning

### **Important Dates**

Event	Date
Lecture 1	20/10/2023
Paper selection + registration	23/10/2023
Place allocation	26/10/2023
Paper allocation	30/10/2023
Supervisor Meeting	12/2023
Lecture 2	12/01/2024
Seminar Presentation	09/02/2024
Paper Summary	23/02/2024

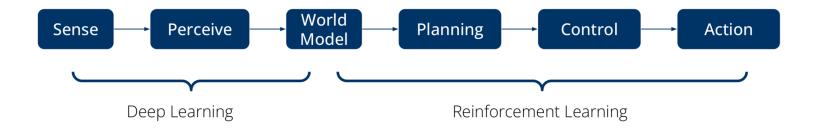
### **Evaluation**

Evaluation	Due Date
Seminar Presentation	09/02/2024
Paper Summary	23/02/2024

- Presentation: At most 20 minutes
- Summary: At most 10 pages excluding bibliography
- Final grade: Presentation + Summary + Seminar participation

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### **Autonomous Robotics**



Can we learn certain parts of this pipeline?

### **Research Areas**

#### Perception

- Recognition
- Depth estimation
- Motion estimation

#### **State Estimation**

- Tracking & Prediction
- SLAM
- Registration

#### **Motion Planning**

- Hierarchical learning
- Reinforcement learning
- Learning from demonstration



#### **Mobile Manipulation**

- · Whole-body motion
- Long-horizon reasoning
- Planning for sensing

#### **Human-Robot Interaction**

- · Socially-compliant behavior
- Human-robot collaboration
- Behavior adaptation & safety

#### **Learning Fundamentals**

- Self-supervised learning
- Continual & Interactive learning
- Multimodal & Multitask learning

#### **Responsible Robotics**

- Fairness
- Explainability & Privacy
- Practical ethics

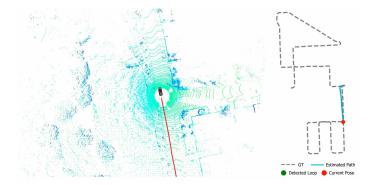
### **Many Seminal Works**



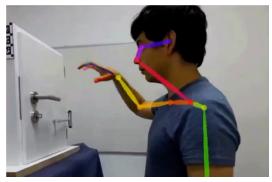
Scene Understanding



Motion Planning



### Simultaneous Localization and Mapping



Learning from Demonstration

### **Robotic Perception - Mobility**

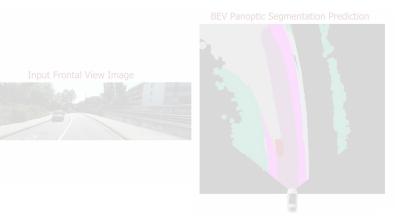
Mohan, Valada: RA-L'22 Mohan, Valada: CVPR'22

Amodal Panoptic Segmentation



Gosala, Valada: RA-L'22 Gosala, Petek, Drews, Burgard, Valada: CVPR'23

### Bird's-Eye-View Panoptic Maps



Enabling robots to perceive objects as a whole regardless of partial occlusion

Predicting panoptic HD maps from monocular frontal view images

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### **Robotic Perception - Mobility**

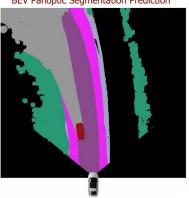


Gosala. Valada: RA-L'22 Gosala, Petek, Drews, Burgard, Valada: CVPR'23

### Bird's-Eye-View Panoptic Maps







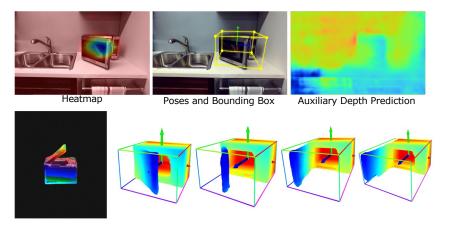
Predicting panoptic HD maps from monocular frontal view images

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### **Robotic Perception - Manipulation**

Heppert, et al.: CVPR'23

#### Single-Shot Reconstruction



Category-independent reconstruction and pose estimation of articulated objects from latent codes

Von Hartz, et al.: RA-L'23

### Bayesian Keypoints for Manipulation



Learning scale-invariant compact representation for mobile manipulation

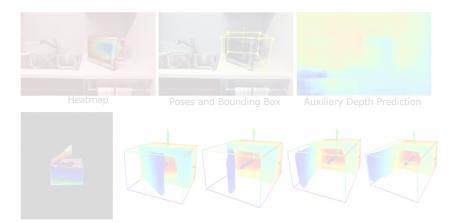
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### **Robotic Perception - Manipulation**

Heppert, et al.: CVPR'23

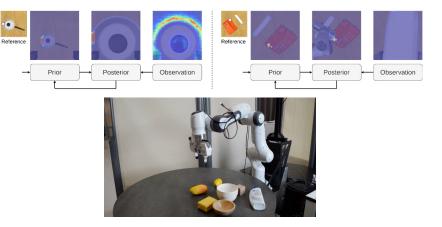
#### Von Hartz, et al.: RA-L'23

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Category-independent reconstruction and pose estimation of articulated objects from latent codes

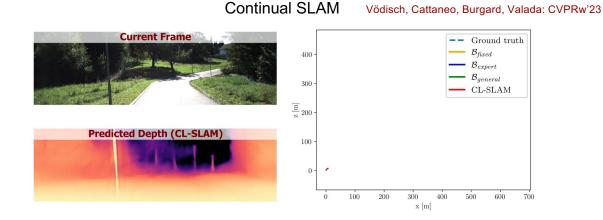
### Bayesian Keypoints for Manipulation



# Learning scale-invariant compact representation for mobile manipulation

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### **Mapping and Localization**



Continual Depth Estimation and Segmentation

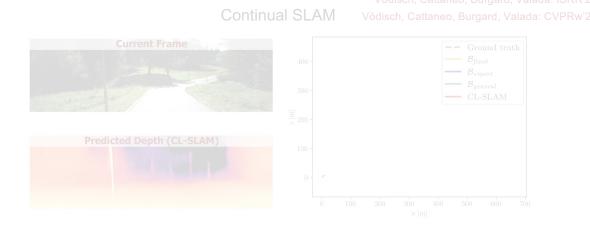
ödisch, Petek, Burgard, Valada: RSS'23



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Vödisch, Cattaneo, Burgard, Valada: ISRR'22

### **Mapping and Localization**



#### Continual Depth Estimation and Segmentation

Vödisch, Petek, Burgard, Valada: RSS'23



### **Mobile Manipulation**

Honerkamp, Welschehold, Valada: RA-L'21 Honerkamp, Welschehold, Valada: T-RO'23

### Neural Navigation for Mobile Manipulation

Schmalstieg, Honerkamp, Welschehold, Valada : under review Schmalstieg, Honerkamp, Welschehold, Valada : ISRR'22

Long-Horizon Object Search



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# **Seminar Topics**

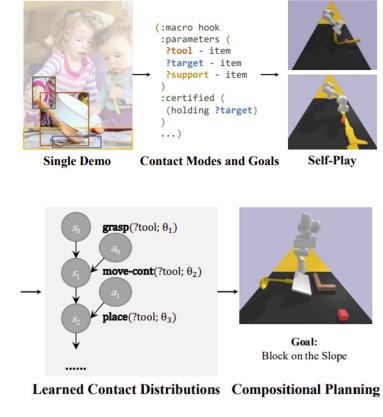


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### **Learning Reusable Manipulation Strategies**

Supervisor: Adrian Röfer

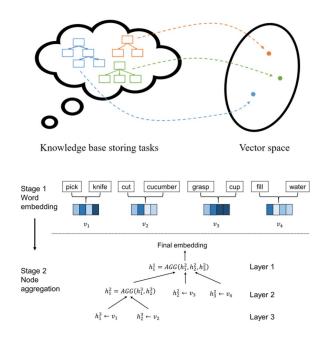
- Learning of tool usage from a single demonstration
- Extract high-level task structure from contact switches
- Learn samplers to fill the gaps in the structure
- Simulated evaluation



### Behavior-Tree Embeddings for Robot Task-Level Knowledge

Supervisor: Adrian Röfer

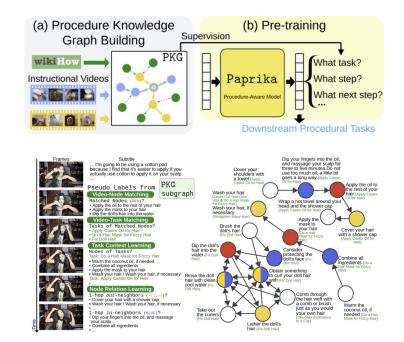
- Behavior Trees are a powerful mechanism to structure complex long-horizon behaviors
- Cannot be used in learning due to lack of fixedsize representation
- Learning of vector representation of BTs based on word-embeddings
- Statistical evaluation of knowledge-transfer, tasksimilarity, and relatedness of subtasks



### **Procedure-Aware Pretraining for Instructional Video Understanding**

### Supervisor: Adrian Röfer

- Trying to generate procedural graphs from instructional videos
- Matching descriptions from WikiHow with videos
- Representation improves video labeling
- Structural representation can be used to relate tasks and track task progress

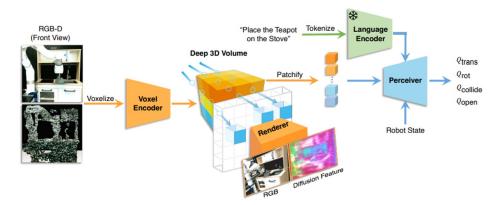


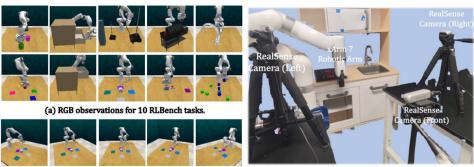
# **GNFactor: Multi-Task Real Robot Learning with Generalizable Neural**

### **Feature Fields**

Supervisor: Nick Heppert

- Multi-Task imitation learning on a voxel grid
- Distill pre-trained semantic features from 2D foundation models into a NeRF
- Use PerAct to train policy
  - Embed language
- Simulation and real world evaluation





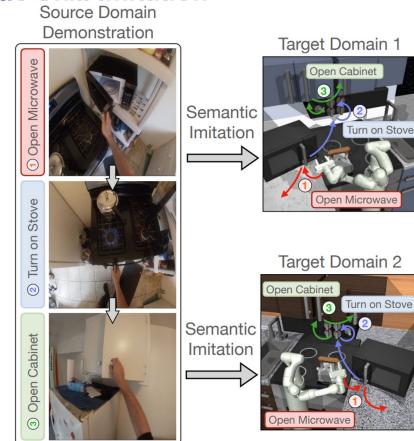
(b) Sampled views for GNF training in simulation.

(c) Real robot setup.

# **Cross-Domain Transfer via Semantic Skill Imitation**

Supervisor: Nick Heppert

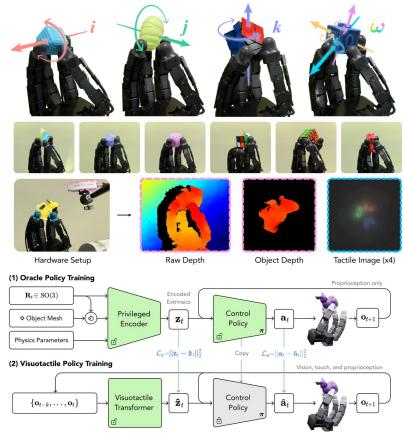
- Classically: imitate low-level actions
- Here: high-level semantic skills
- Use prior experience to learn semantic "skills"
- Enables cross-domain transfer → learning from different sources



### **General In-Hand Object Rotation with Vision and Touch**

Supervisor: Nick Heppert

- Rotate object around axis
- Multi-modality: vision and touch
- Underlying technique: rapid domain adaptation
- learn encoding + policy based on full information
- learn to encode with observation



### M2T2: Multi-Task Masked-Transformer for Object-centric Pick and Place

Supervisor: Jan Ole von Hartz

- Transformer model, predicting object grasp and placement from raw point cloud.
- Zero-shot Sim2Real transfer.
- Can be language-conditioned.



### Semantic-SAM Segment and Recognize Anything at Any Granularity

Supervisor: Jan Ole von Hartz

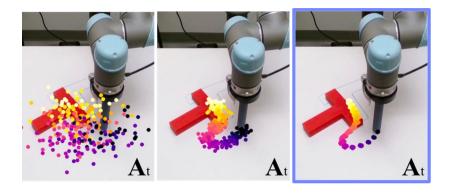
• Segmentation Model: instance, semantic, panoptic and part, multi-level.



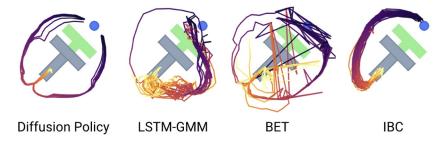
# **Diffusion Policy: Visuomotor Policy Learning via Action Diffusion**

Supervisor: Jan Ole von Hartz





• Diffusion. Policy. Exciting!

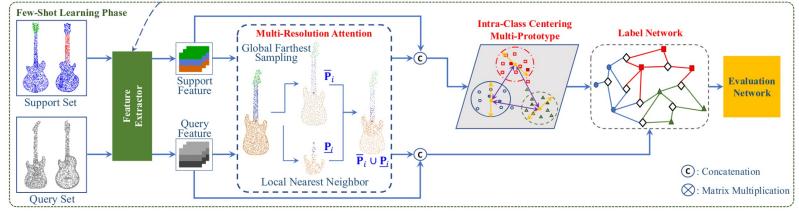


# Few-Shot Point Cloud Semantic Segmentation via Contrastive Self-

### **Supervision and Multi-Resolution Attention**

Supervisor: Julia Hindel

- Few-shot learning: training with limited labeled samples (support set) with the help of unlabeled data (query set)
- Self-supervised pre-training of feature extractor.
- Class prototype extraction and label propagation.

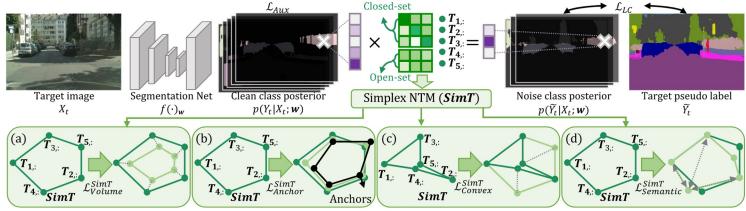


# Handling Open-Set Noise and Novel Target Recognition in Domain

### **Adaptive Semantic Segmentation**

Supervisor: Julia Hindel

- Open-set domain adaptation: target domain with unknown classes, black-box source model.
- Self-training with pseudo-labels
- Learn noise transition matrix (NTM) with constraints

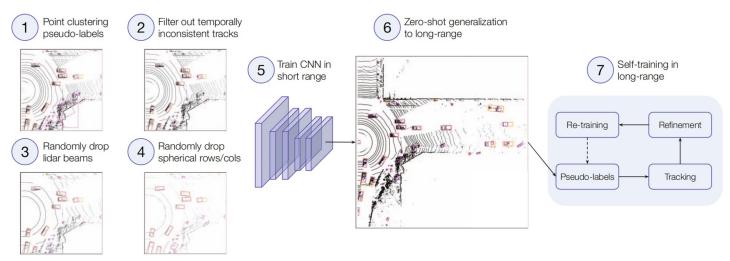


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## **Towards Unsupervised Object Detection from LiDAR Point Clouds**

Supervisor: Julia Hindel

- Near-range clustering, augmentation and temporal consistency
- Self-training on selected pseudo-labels



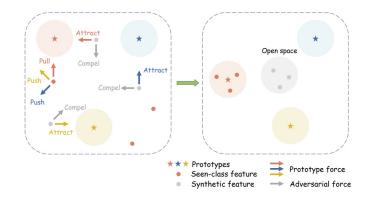
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# **Open-set Semantic Segmentation for Point Clouds via Adversarial**

# **Prototype Framework**

Supervisor: Rohit Mohan

- Open set: Detect new classes but maintain performance on seen-class points.
- Feature extraction, prototypical constraint, feature adversarial modules
- The modules do the following:
  - Learn prototype for seen classes
  - Implicit distribution estimation of unseen-classes
  - Synthesize unseen-class features
- Adversarial mechanism to improve discriminability

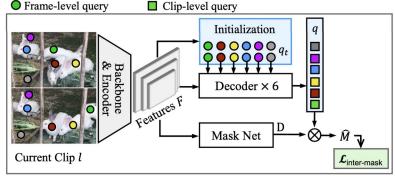


## **MDQE: Mining Discriminative Query Embeddings to Segment**

### **Occluded Instances on Challenging Videos**

Supervisor: Rohit Mohan

- Video instance segmentation with per-clip labels fail with crowded scenes
- Problem: Instance queries cannot encode discriminative embeddings -> poor query based performance
- Object Queries initialization with spatial contextual information and inter-frame object motion
   Frame-level query
   Clip-level query
- Inter-instance mask repulsion loss

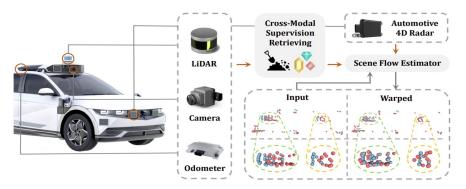


### Hidden Gems: 4D Radar Scene Flow Learning Using Cross-Modal

### **Supervision**

Supervisor: Rohit Mohan

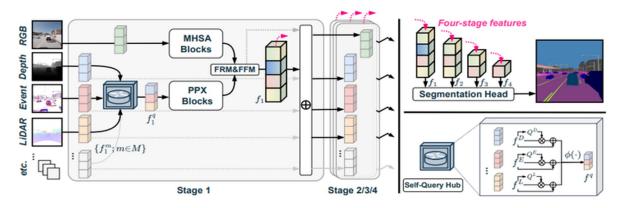
- Scene flow annotations are costly and requires tedious and intensive human labour -> Low-fidelity data further complicates procurement of annotations
- Self-supervised learning via cross-modal supervision utilizing co-located sensing
- Multitask model architecture with multiple cross-modal constrained losses



### **Delivering Arbitrary-Modal Semantic Segmentation**

Supervisor: Iana Zhura

- Main idea: extract effective information from any modality for superior semantic segmentation
- The RGB image is gradually processed by Multi-Head Self-Attention (MHSA) blocks.
- Feature maps from other modalities will be merged via the proposed Self-Query Hub.
- Features are fused and forwarded to decoder for segmentation prediction.

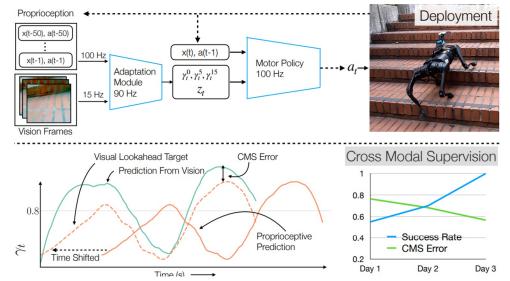


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### Learning Visual Locomotion with Cross-Modal Supervision

### Supervisor: Iana Zhura

- CMS uses time shifted
  proprioception to supervise vision.
- Policies takes an estimate of terrain in front and below the robot for visual adaptation to complex terrain
- Then, the current proprioceptive state and terrain information, is used to predict the target actions.

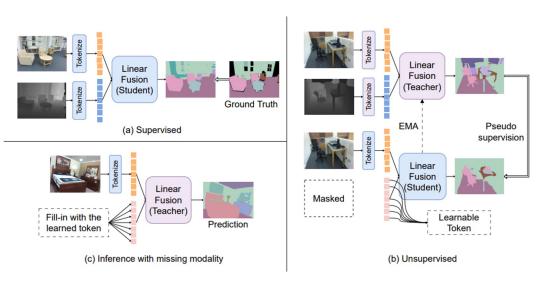


# **Missing Modality Robustness in Semi-Supervised Multi-Modal**

### **Semantic Segmentation**

Supervisor: Iana Zhura

- Robust to missing modality during training.
- Mask modality and replace with learnable tokens.
- Multi-modal mean (EMA) teacher generates a segmentation prediction which is used to supervise student.
- The learned token is used during inference if any modality is missing.



# Questions



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### **Announcement: Open positions**

- We have open positions for a hiwi, good opportunity to work on practical robotics and get to know the lab.
- We have multiple MSc project and thesis topics related to many directions of robot learning.

https://rl.uni-freiburg.de/open-positions