

Implementation and evaluation of a real-time capable approach to sensorbased sorting using CNNs



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# Problem

#### Sorting system





#### **Problem**

conventional algorithms often are...

- engineered to fit a specific sorting task
- overgrown with configuration parameters
- limited to primitive features
- Convolutional Neural Networks (CNNs) have proven their effectiveness in many computer vision tasks
  - development of a CNN-based approach
  - evaluation on a real sorting system and sorting results



### **Materials for evaluation**



#### Peanuts & Hibiscus tea



Construction and Demolition Waste (CDW) with high amounts of dust



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# Approach

Approach



#### semantic segmentation



#### valve activation



pneumatic valves



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#### **Architecture**



- based on U-Net architecture (Ronneberger et al., 2015)
- padded convolutions
- reduction of number of featuremaps by factor of 4
- for CDW: downsampling and upsampling layers on input / output



# Training

#### **Synthetic data generation** Extraction of objects





#### **Synthetic data generation** Extraction of objects





#### **Synthetic data generation** Additional dust images





























# Inference

#### Inference Overview





#### Inference

time



i + 1

i + 2





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### Inference

time





i + 2





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# Results

### **Results** Classification result peanuts & hibiscus tea



image



result (semantic segmentation)



## **Results** Sorting quality: peanuts & hibiscus tea





## **Results** Sorting quality: peanuts & hibiscus tea

	False Positive Rate	False Negative Rate
Conventional	19 %	4 %
CNN	2 %	5 %

#### Materials:

- 400 g hibiscus tea
- 20 g peanut



## **Results Sorting quality: peanuts & hibiscus tea**

	conventional		CNN	
	hibiscus	peanut	hibiscus	peanut
accepted	81.3%	3.5%	97.8%	5.0%
rejected	18.8%	96.5%	2.3%	95.0%

#### Sorting quality in % of true material class.

- 400 g hibiscus tea
- 20 g peanut



## **Results** Classification result construction and demolition waste



image



result (semantic segmentation)



## **Results** Sorting quality: construction and demolition waste

	False Positive Rate	False Negative Rate
Conventional	14 %	47 %
CNN	4 %	10 %

#### Materials:

- 2 kg concrete
- 3 kg other (brick and aerated concrete)
- 100 g dust



## **Results Sorting quality: construction and demolition waste**

	conventional		CNN	
	other	concrete	other	concrete
accepted	86.2%	42.4%	96.3%	9.5%
rejected	13.9%	52.9%	2.1%	90.0%
lost	0.0%	4.8%	1.5%	0.5%

#### Sorting quality in % of true material class.

- 2 kg concrete
- 3 kg other (brick and aerated concrete)
- 100 g dust



#### **Realtime capability** Example: construction and demolition waste

- 30 ms max. allowed latency
- 10 ms for line buffering (64 + 2x32 lines)
- 4 ms for all GPU calculations (NVIDIA RTX 4070 GPU)
- throughput > 13,333 lines (4096 pixels) per second
  - Iimited by camera



## Conclusion

#### Conclusion

- conventional algorithms are limited in quality
- direct mapping of semantic segmentation result to valves
- training with synthetically generated training data
- CNN achieves increased sorting quality
- real-time capable



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ocm-conference.com