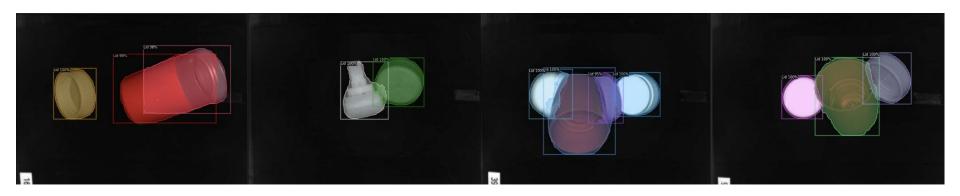


#### Anomaly Detection for the Singulation of Plastic Wastes in Polymer Recycling

Group 5: Charlotte Goos, Aleksandr Eismont, Dmitrii Seletkov

Supervisors: Constantin Seibold, Simon Reiß

Final Presentation, 25.07.2022



### Agenda



**Problem Setting** 

Copy & Paste Data Augmentation

Classification

**Instance Segmentation** 

Summary



# **Problem Setting**

#### Task



- Setting: plastic lids for recycling
- Task: object counting only one lid allowed
- Given: small unlabeled dataset
  - 2751 images, 0-5 lids in trays

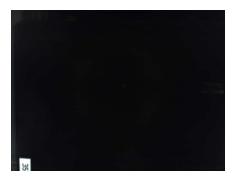




Recycling Production Line

#### **Dataset: Images**





Label: 0



Label: 1



Label: 2



Label: 3



Label: 4

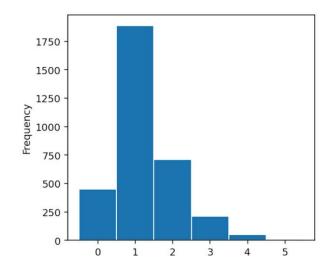


Label: 5

#### **Dataset: Labeling**



- Image-level labels for dataset
- Problems:
  - Small amount of data
  - Class Imbalance
  - Hard cases:
    - Different Colors 25.7%
    - Overlapping 22.5%
    - Transparent 16.1%
    - Inside 6.9%
    - Dark Color 5.3%
    - One Color 3.9%
    - Open Lid 0.6%
    - Edge 0.5%

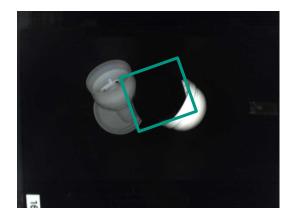


Frequency of Labels 0-5 in Original Dataset

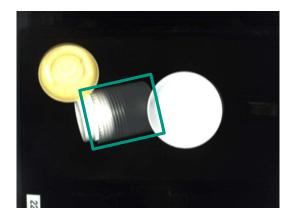
#### **Dataset: Analysis**



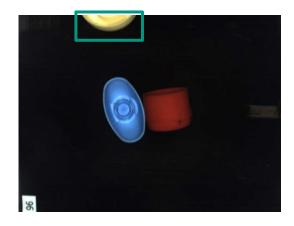
- Focus on: overlapping, edge, dark, transparent
- Target and tags stratified train/test split



Overlapping and Dark Label: 4



Overlapping and Transparent Label: 4



Edge Label: 3



# **Copy & Paste Data Augmentation**

#### **Copy & Paste Data Augmentation - Approach**



- Inspired by [Dwibedi17, Ghiasi20]
- Synthetic images of label 0: add gaussian noise
- Synthetic images of label 1-5:
  - Background: empty tray
  - Objects: from images of label 1
  - Positions of object based on heatmap
- Copy & Paste
  - Binary masks of lid
  - Lid with black background
  - Bitwise\_or of background and mask
  - Add lid with black background



Binary Mask of Object

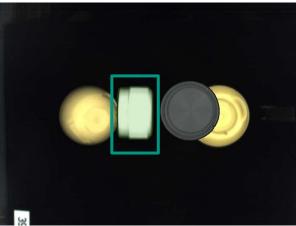


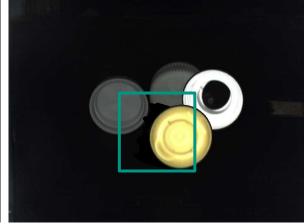
Object on black background

# **Copy & Paste Data Augmentation - Tags**





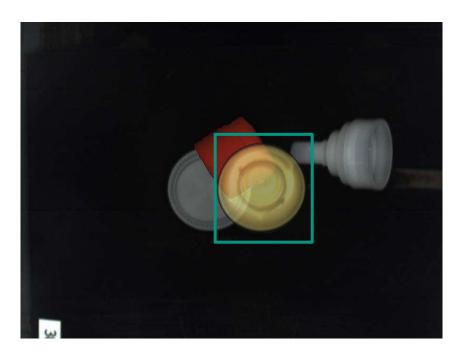


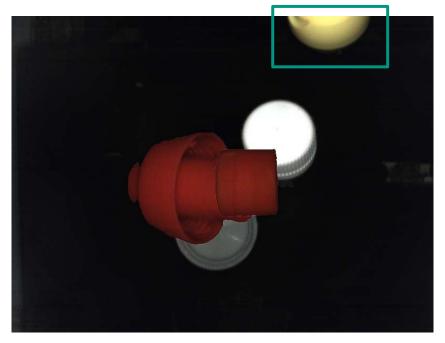


Rotate Color Dark

## **Copy & Paste Data Augmentation - Tags**







Transparent Edge

#### **Copy & Paste Data Augmentation - Annotation**



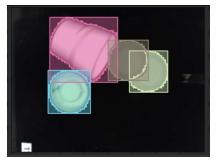
Binary mask of each lid in synthetic image







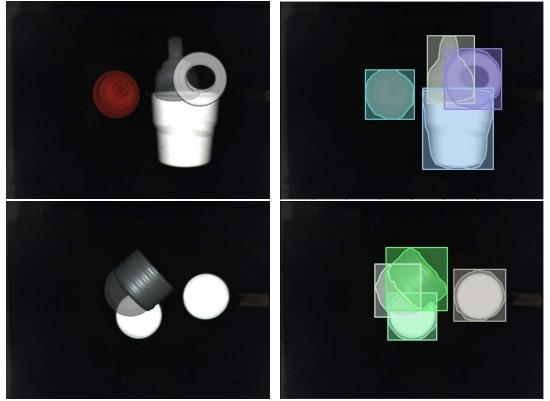






### **Copy & Paste Data Augmentation - Annotation**





Synthesized Images and Their COCO Annotations

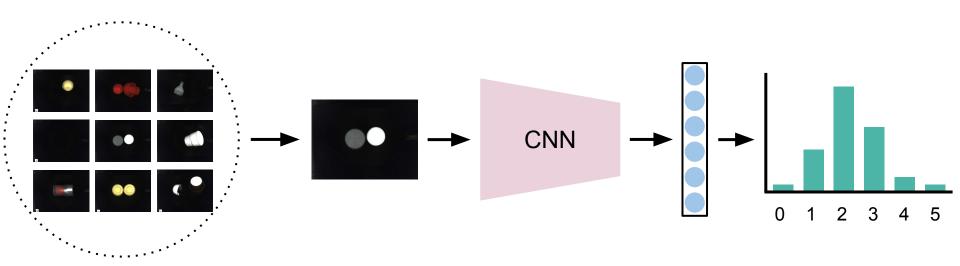


#### Classification

#### Classification: Idea



Approach: Salient Object Subitizing (SOS) [Zhang15, Zhang16]



#### **Classification: Settings**



#### Implementation Set-Up:

- ResNet18
- Cross Entropy Loss with Weights
- Stratified Batch Sampling
- Stochastic Gradient Descent with Learning Rate 0.001
- Augmentation Random Horizontal Flip

#### Metric:

• Decision Accuracy 
$$= \frac{Accepted\ Trays\ +\ Rejected\ Trays}{All\ Trays}$$

#### **Classification: Results**



Name	Accuracy	Decision Accuracy
SOS	0.66 (±0.02)	0.71 (±0.02)
GoogleNet	0.90 (±0.01)	0.95 (±0.00)
ResNet18	0.92 (±0.01)	0.97 (±0.00)
ResNet18_weights	0.93 (±0.01)	0.97 (±0.00)
ResNet18_synthetic	0.95 (±0.01)	0.98 (±0.00)

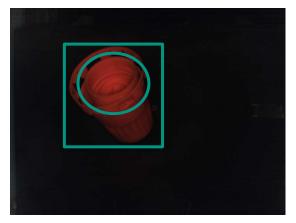
**Evaluation of Classification Approaches** 

#### **Classification: Misclassified Images**

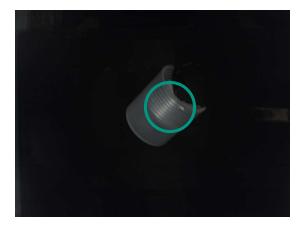




Predicted Label: 1
Correct Label: 2



Predicted Label: 1 Correct Label: 2



Predicted Label: 2 Correct Label: 3



# **Instance Segmentation**

#### **Instance Segmentation**



- Train: fully synthesized data
- Test:
  - Separate fully synthesized test dataset for COCO evaluation
  - The same real-world test dataset used for classification
- Model:
  - Mask R-CNN (ResNet50 + FPN)
  - SGD, 5000 iterations, annealing LR 0.001 with steps (3000, 4000)

Name	AP	AP50	AP75
Seg_all	80.25	96.85	88.33

COCO Evaluation of Segmentation Model on Separate Synthesized Test Dataset

### **Segmentation: Ablation of Augmentation**



Name	Rotate	Color	Edge	Dark	Transparent	Accuracy	Decision Accuracy
Seg_no	X	Х	Х	X	X	0.83	0.89
Seg_edge	+	+	+	X	X	0.83	0.89
Seg_edge_dark	+	+	+	+	X	0.85	0.90
Seg_all	+	+	+	+	+	0.86	0.91

Ablation Study of Different Augmentations for Segmentation Model on Real-world Test Dataset

### Segmentation: Ablation of Augmentation





Confusion Matrix of Seg\_no Model (without Any Transformation)



Confusion Matrix of Seg\_all Model (with All Transformations)

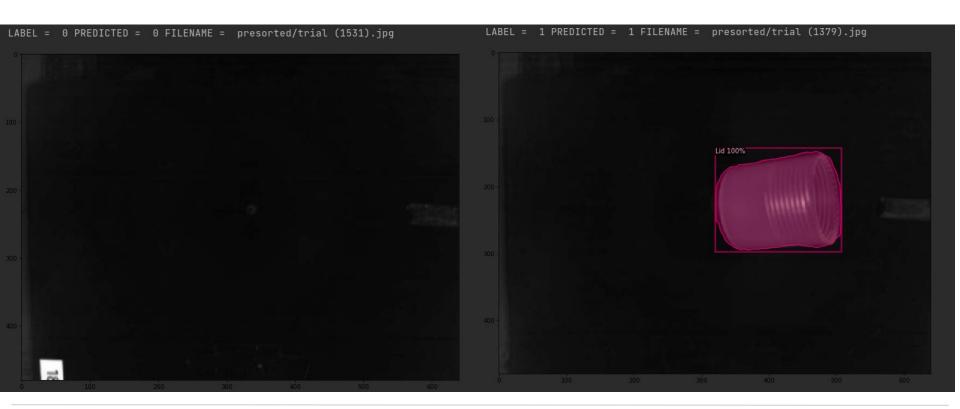
### Segmentation vs. Classification



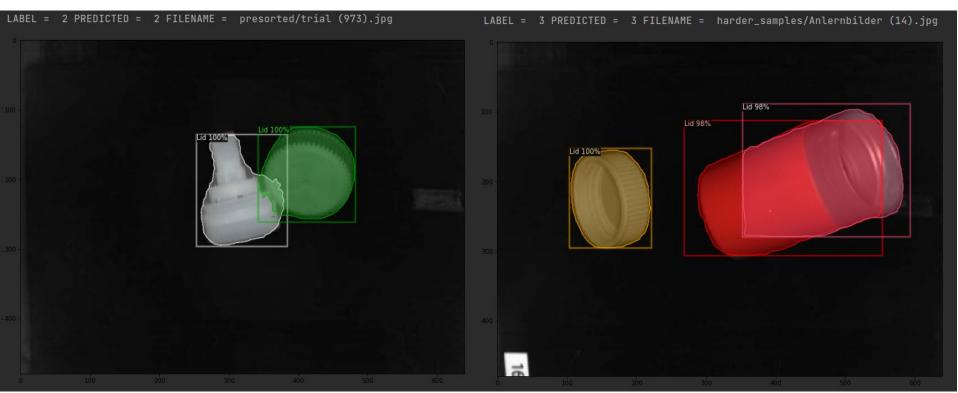
Name	Label 0	Label 1	Label 2	Label 3	Label 4	Label 5	Accuracy	Decision Accuracy
Seg_all	1.00	0.94	0.74	0.44	0.20	0.00	0.86	0.91
Classification_best	1.00	0.99	0.87	0.70	0.70	0.50	0.95	0.98

Comparison of the Best Segmentation (Seg\_all) and Best Classification Approach (Classification\_best)

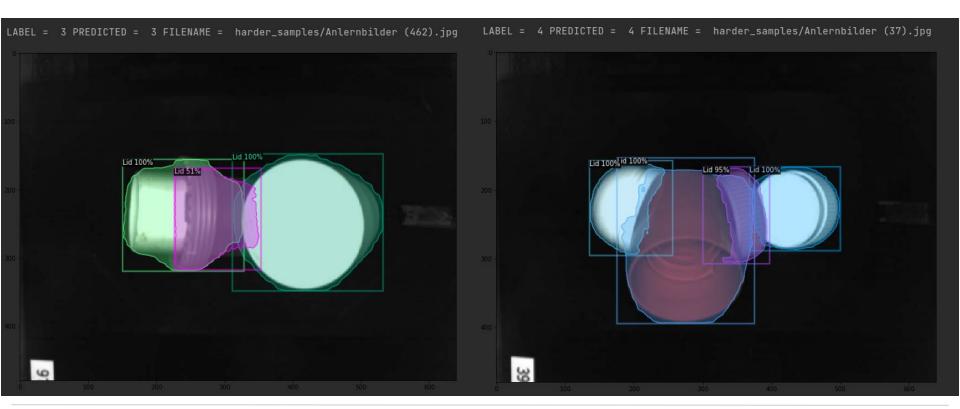






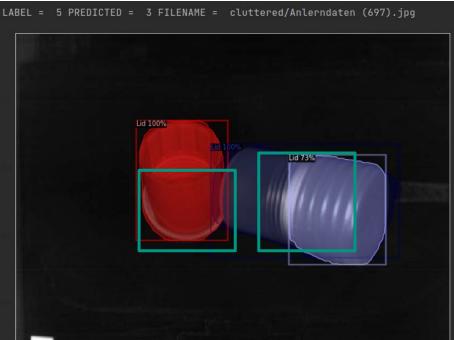














# **Summary**

#### **Summary**



- Unlabeled small dataset
- Copy & Paste Data Augmentation
- Classification: best performance
  - Model Tuning
  - Synthesized Data Improves Performance
- Instance Segmentation: poorer performance
  - But: interpretability
  - Synthesized data improves performance
- Future works:
  - Mixed approach: combine classification and segmentation
  - Improve for more complex object contours



# Thank you for your attention

#### References



- [Dwibedi17] Debidatta Dwibedi, Ishan Misra, and Martial Hebert. "Cut, Paste and Learn: Surprisingly Easy Synthesis for Instance Detection". In: CoRR abs/1708.01642 (2017). arXiv: 1708.01642. url: <a href="http://arxiv.org/abs/1708.01642">http://arxiv.org/abs/1708.01642</a>.
- [Ghiasi20] Golnaz Ghiasi et al. "Simple Copy-Paste is a Strong Data Augmentation Method for Instance Segmentation". In: CoRR abs/2012.07177 (2020). arXiv: 2012.07177. Url: <a href="https://arxiv.org/abs/2012.07177">https://arxiv.org/abs/2012.07177</a>.
- [Zhang15] Jianming Zhang, Shugao Ma, Mehrnoosh Sameki, Stan Sclaroff, Margrit Betke, Zhe Lin, Xiaohui Shen, Brian Price and Radomír Měch. "Salient Object Subitizing." In: Proc. IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2015.
- [Zhang16] Jianming Zhang, Shugao Ma, Mehrnoosh Sameki, Stan Sclaroff, Margrit Betke, Zhe Lin, Xiaohui Shen, Brian Price and Radomír Měch. "Salient Object Subitizing." Journal version under review, 2016. Url: <a href="https://arxiv.org/abs/1607.07525">https://arxiv.org/abs/1607.07525</a>.
- The code of our project: <a href="https://github.com/yayapa/AnomaliesRecycling">https://github.com/yayapa/AnomaliesRecycling</a>



# **Appendix**

#### **Classification: Confusion Matrix**





## **Tags**



different colors



transparent





one color

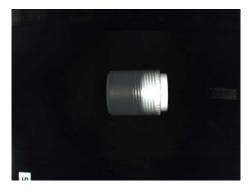


dark color

## **Tags**



open lid



inside





edge

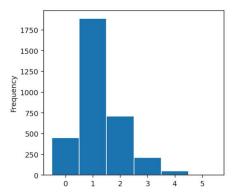


overlapping

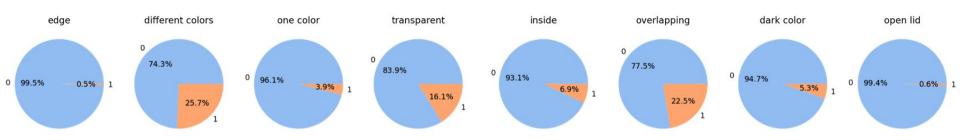
#### **Dataset: Labeling and Analysis**



- Image-level labels for dataset
- Problems:
  - Small amount of data
  - Class Imbalance
  - Hard cases analysis with tags



#### Frequency of Labels 0-5 in Original Dataset



Amount of Data Corresponding to Hard Cases Marked with Tags

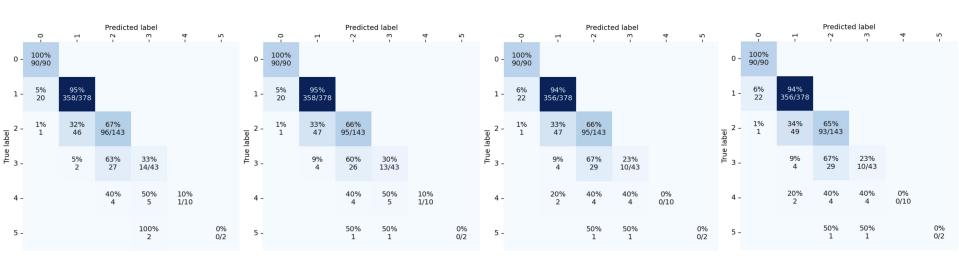
### **Segmentation: Ablation of Augmentation**



Name	Rotate	Color	Edge	Dark	Trans parent	Label 0	Label 1	Label 2	Label 3	Label 4	Label 5	Accuracy	Decision Accuracy
Seg_no	x	х	Х	х	х	1.00	0.94	0.64	0.28	0.20	0.00	0.83	0.89
Seg_edge	+	+	+	х	х	1.00	0.94	0.66	0.26	0.10	0.00	0.83	0.89
Seg_edge _dark	+	+	+	+	х	1.00	0.95	0.69	0.35	0.30	0.00	0.85	0.90
Seg_all	+	+	+	+	+	1.00	0.94	0.74	0.44	0.20	0.00	0.86	0.91

#### Segmentation: Ablation Of Test Threshold





Threshold: 0.5 Accuracy: 0.84

Decision Accuracy: 0.9

Threshold: 0.7 Accuracy: 0.84

Decision Accuracy: 0.89

Threshold: 0.9 Accuracy: 0.83

Decision Accuracy: 0.89

Threshold: 0.95

Accuracy: 0.82

Decision Accuracy: 0.88

#### Class Activation Map: GradCAM

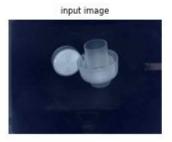


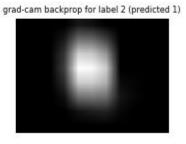
data/test/1/trial (704).jpg

input image

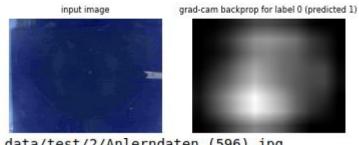
grad-cam backprop for label 1 (predicted 1)

data/test/2/Anlerndaten (618).jpg

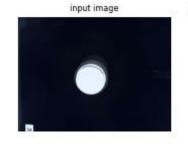


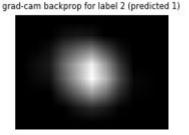


data/test/0/trial (1307).jpg



data/test/2/Anlerndaten (596).jpg





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