# MultiOrg: A Multi-Rater Organoid Detection Dataset

# NeurIPS 2024: Datasets & Benchmarks

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# **Motivation & Background**

What are organoids?



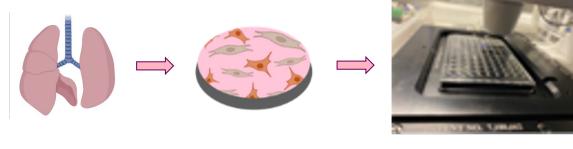
#### Organoids are used for:

- Disease understanding
- Drug development
- Personalised medicine

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# **How Organoids are made**

Growing tiny organs in the lab

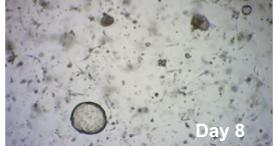




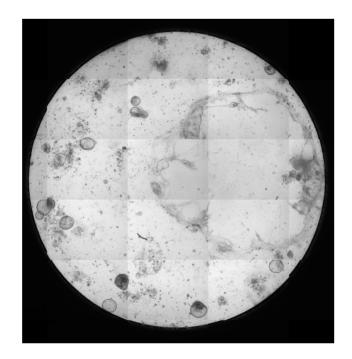
Organoid culture

Image acquisition

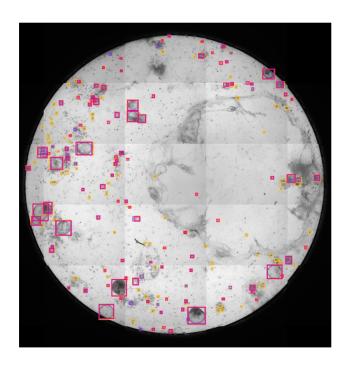




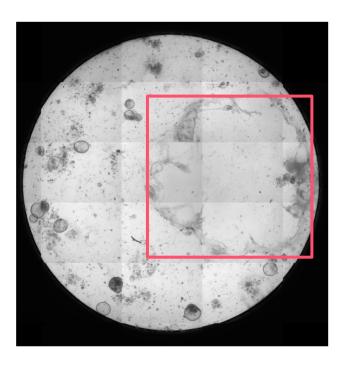




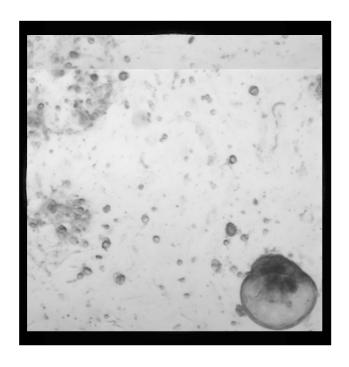
Labeling uncertainty and noise



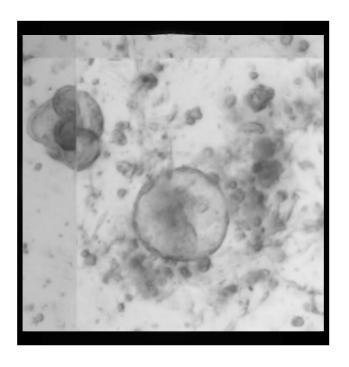
Time-consuming manual annotations



- Time-consuming manual annotations
- Noise due to artifacts, dead cells and debris

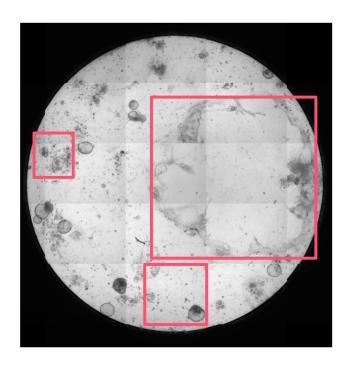


- Time-consuming manual annotations
- Noise due to artifacts, dead cells and debris
- Large range of organoid sizes



- Time-consuming manual annotations
- Noise due to artifacts, dead cells and debris
- Large range of organoid sizes
- Organoid clusters

Labeling uncertainty and noise



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- Noise due to artifacts, dead cells and debris
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- Organoid clusters

No datasets in organoid detection addressing label uncertainty!

# The MultiOrg dataset

#### **Overview**

400+ 2D microscopy images

Study Type	Normal		Macros		Combined			
	# Images	# Organoids	# Images	# Organoids	# Images	# Organoids		
Train set								
train <sub>A</sub>	181	30,710	15	2,669	196	33,379		
train <sub>B</sub>	135	20,263	25	1,781	160	22.044		
Total	316	50,973	40	4,450	356	55,423		
Test set								
$test^0_A$	8	1,145	14	1,865	22	3,010		
$test_B^{\vec{0}}$	20	3,020	13	1,493	33	4,513		
Total (Label set test <sup>0</sup> )	28	4,165	27	3,358	55	7,523		

# The MultiOrg dataset

#### Overview

- 400+ 2D microscopy images
- 60,000+ organoid annotations

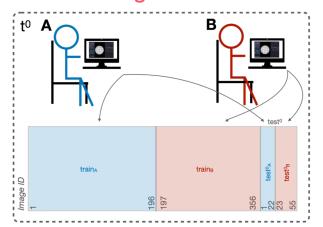
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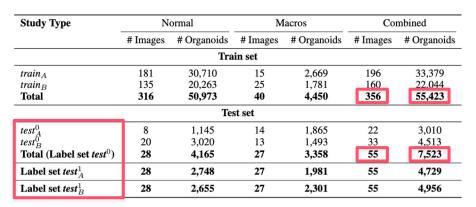
# The MultiOrg dataset

#### Overview

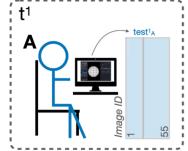
- 400+ 2D microscopy images
- 60,000+ organoid annotations
- Three expert labels on the test set

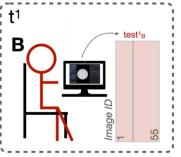
#### Multi-rater setting







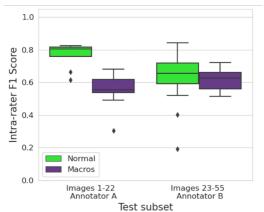


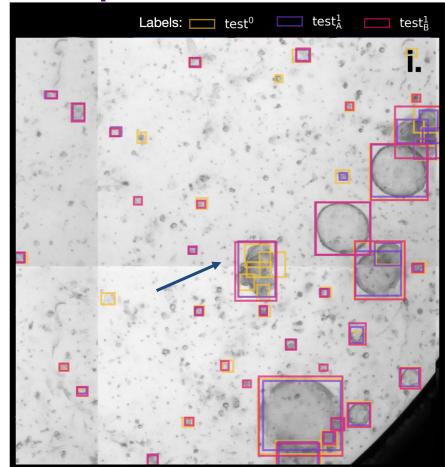


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Multi-rater setting at two distinct timepoints

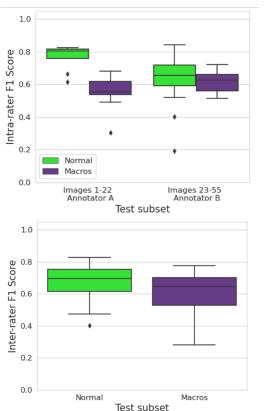
Allows for intra- and inter-rater analyses

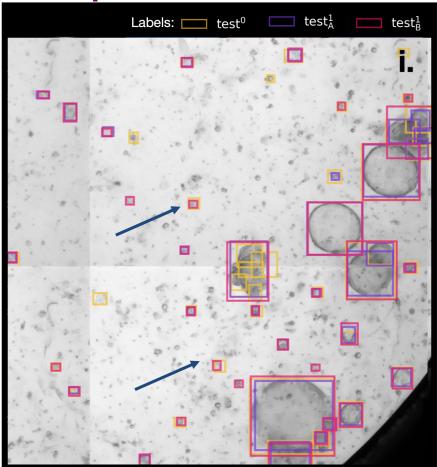




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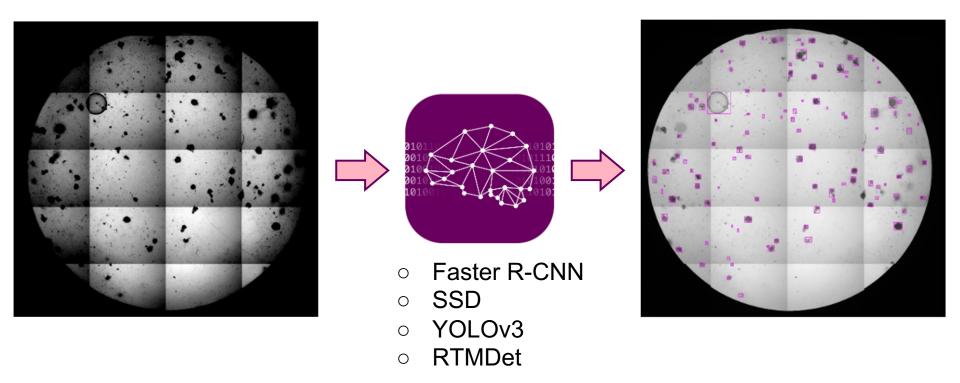




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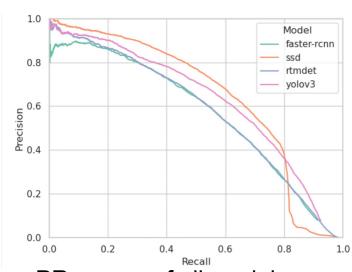
# MultiOrg: Model Benchmark

Object-detection models



# MultiOrg: Model Benchmark

### Object Detection Scores



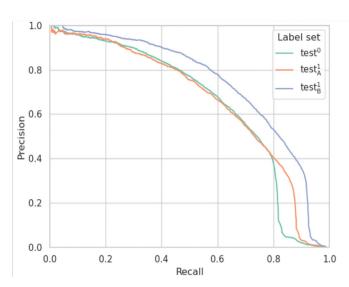
PR curves of all models on test<sup>0</sup>

Metric	Label set	Faster R-CNN	SSD	YOLOv3	RTMDet
Precision	$test^0$	0.23	0.61	0.73	0.64
	$test^1_A$	0.16	0.44	0.58	0.54
	$test_B^{\hat{1}}$	0.18	0.50	0.67	0.56
	mean	0.19	0.52	0.66	0.58
Recall	$test^0$	0.84	0.67	0.48	0.51
	$test^1_A$	0.92	0.78	0.62	0.69
	$test_B^{\hat{1}^*}$	0.97	0.83	0.67	0.68
	mean	0.91	0.76	0.59	0.63
F1-score	$test^0$	0.36	0.64	0.58	0.57
	$test^1_A$	0.27	0.57	0.60	0.61
	$test_B^1$	0.30	0.62	0.67	0.62
	mean	0.31	0.61	0.62	0.60
mAP@0.5IoU (%)	$test^0$	56.56	64.40	62.55	57.71
	$test^{\scriptscriptstyle 1}_A$	57.09	65.79	61.11	63.87
	$test^1_B$	68.36	<b>73.88</b>	70.25	63.23
	mean	60.67	68.09	64.64	61.60
mAP@0.75IoU (%)	$test^0$	17.48	21.81	19.15	22.56
	$test^1_A$	23.53	23.42	19.13	30.13
mAr &0.73100 (%)	$test_B^{\hat{1}^*}$	46.98	46.48	39.01	32.85
	mean	29.33	30.57	25.76	28.51

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# **MultiOrg: Model Benchmark**

## Object Detection Scores



PR curves of SSD model on all label sets

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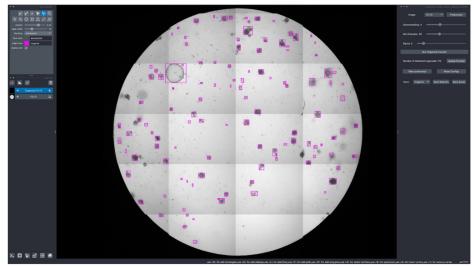
# MultiOrg: Key contributions

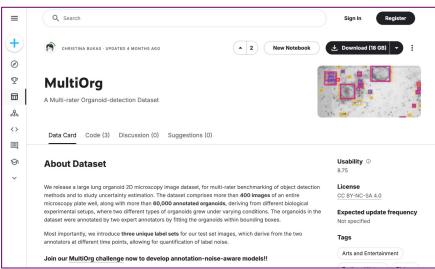
Open-source resources

# Napari plugin for the biologist community

# Dataset, benchmark & challenge







https://github.com/HelmholtzAl-Consultants-Munich/napari-organoid-counter

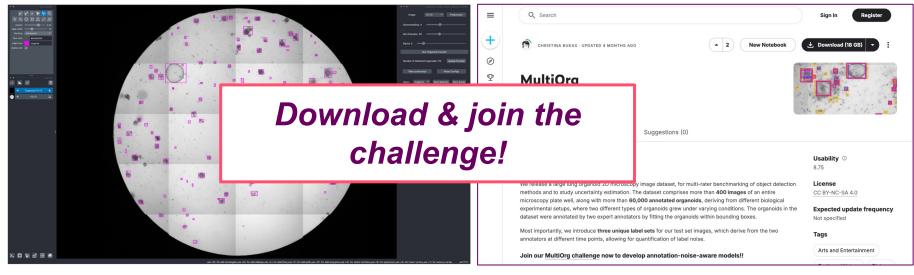
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Napari plugin for the biologist community

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