

# Label Assistant: A Workflow for Assisted Data Annotation in Image Segmentation Tasks

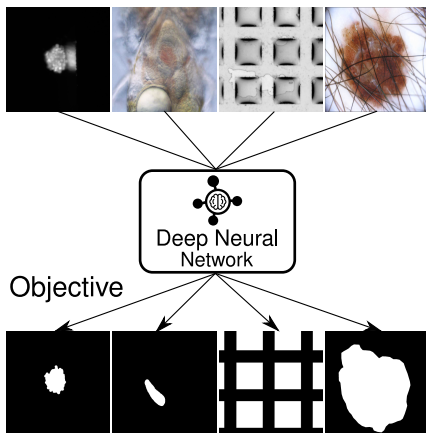
31<sup>st</sup> Workshop Computational Intelligence, Berlin

M. P. Schilling, L. Rettenberger, F. Münke, H. Cui, A. A. Popova, P. A. Levkin, R. Mikut, M. Reischl | 26<sup>th</sup> November 2021

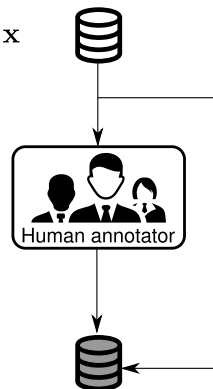
# Outline

- 1 Introduction
- 2 Workflow
  - Selector
  - Pre-Assistance
  - Post-Assistance
- 3 Implementation
- 4 Datasets
- 5 Results
- 6 Conclusion and Outlook

# Introduction - Labeling

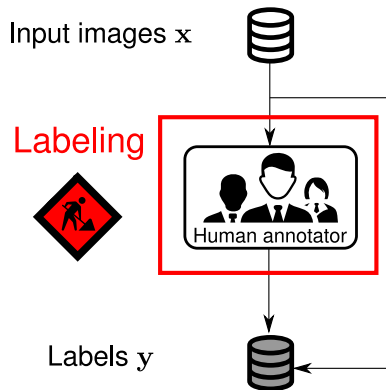
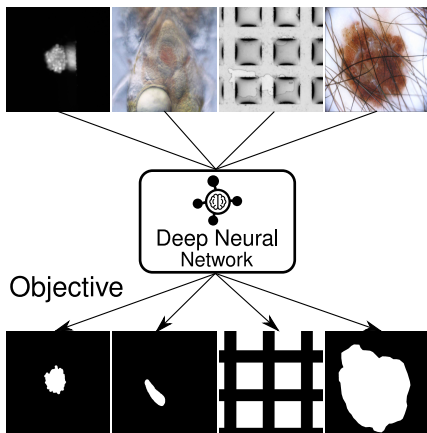


Input images  $x$

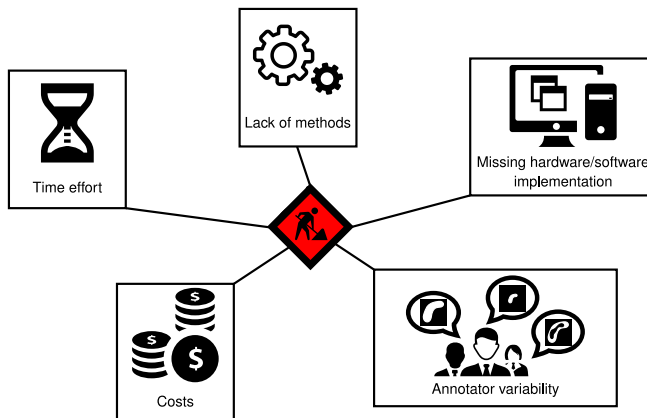


Labels  $y$

# Introduction - Labeling

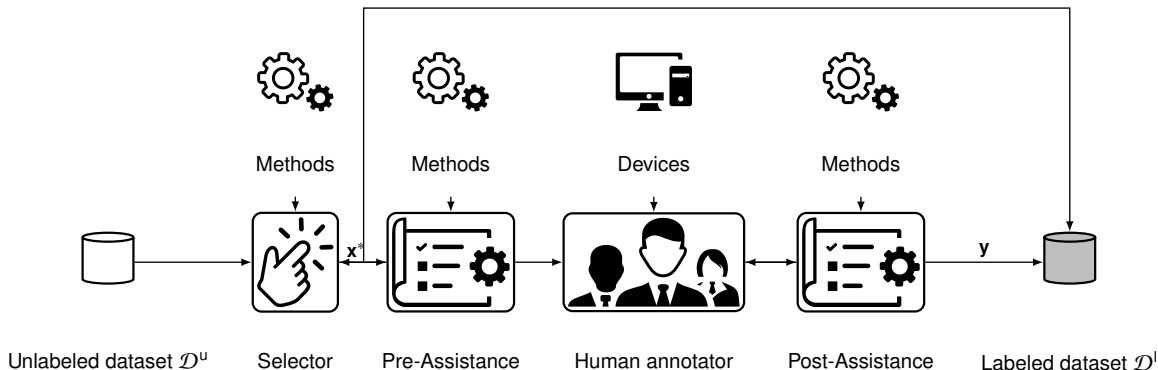


# Introduction - Challenges



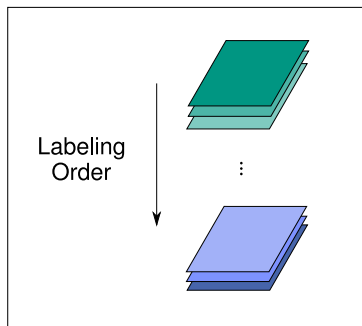
→ 🎯 **Objective:** Workflow for assistance to enhance labeling process [1, 2]

# Workflow



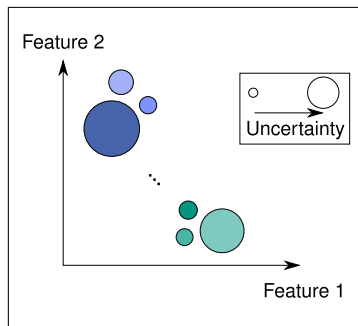
# Selector

Status Quo

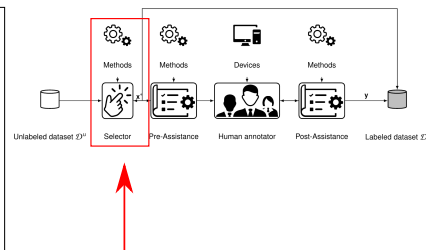


→ Sequential sampling can be sub-optimal

Selector (Deep Active Learning)

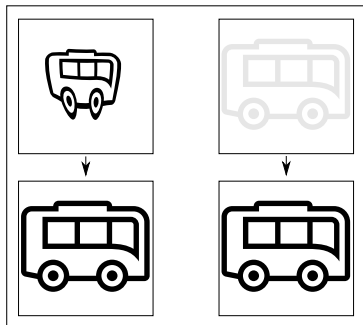


→ Use selection strategy (uncertainty, heterogeneity, ...)



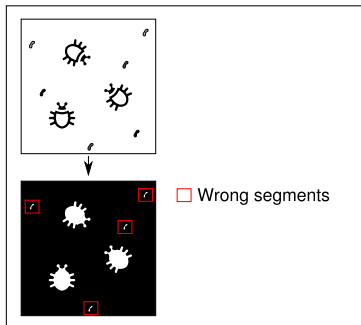
# Pre-Assistance

## Image Pre-Processing

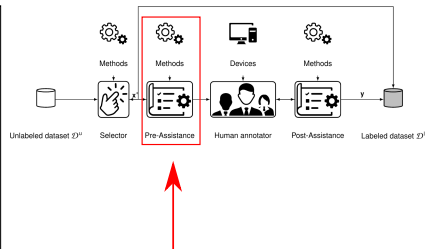


→ Prepare image to simply labeling for annotators

## Pre-Labeling



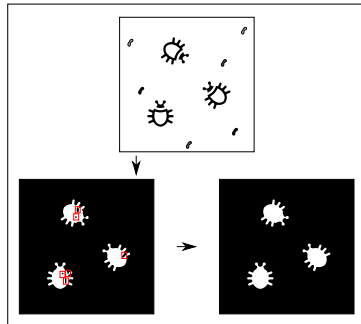
→ Pre-Labels to reduce labeling effort





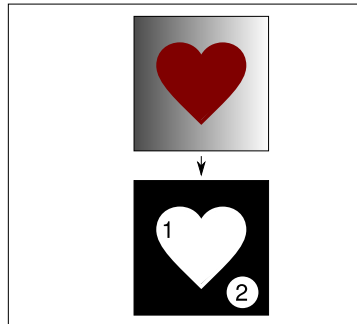
# Post-Assistance

## Post-Processing

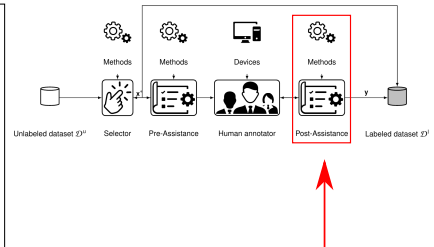


→ Post-Processing to improve label quality

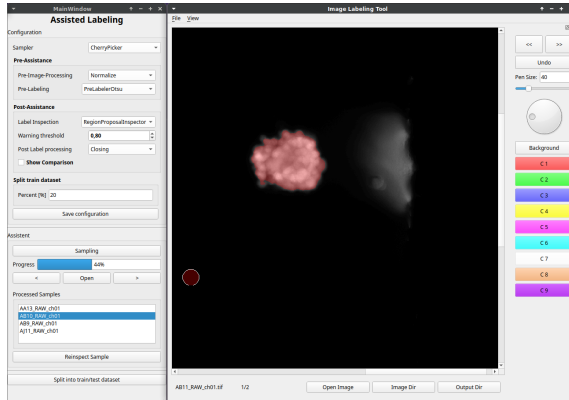
## Label Inspection



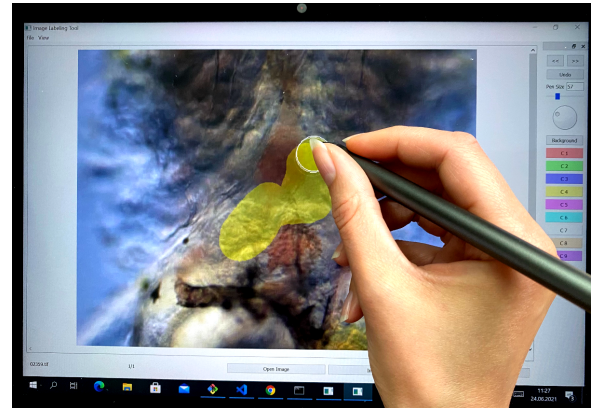
→ User warning based on defined criteria to enhance label quality



# Implementation



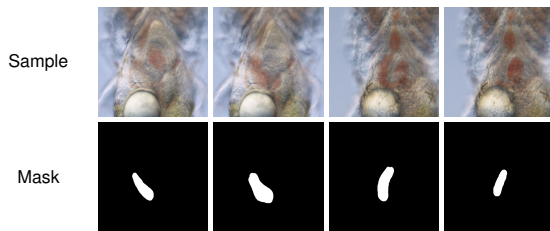
GUI [3]



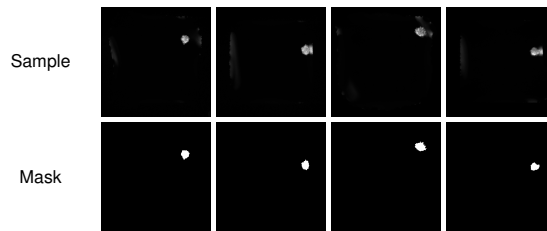
Support of different devices (tablet, laptop)/operating systems

Available soon (Git repository <https://git.scc.kit.edu/sc1357/kaida>)

# Datasets



Medaka [4]



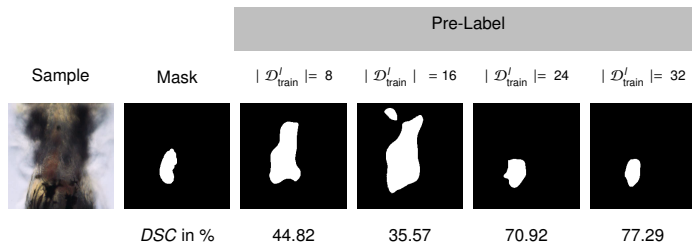
DMA Spheroid [5]

# Result excerpts - Medaka

- Performance comparison: Different selection strategies to obtain labeled subset in contrast to labeling complete dataset (baseline)

	Configurations			
	Sequential	Random	Sequence-aware <sup>1</sup>	Baseline
$ \mathcal{D}'_{\text{train}} $	32	32	32	400
$DSC_{\text{test}}$ in %	46.50	77.67	80.63	82.70

- Pre-Labeling via trained U-Net [6] on small dataset



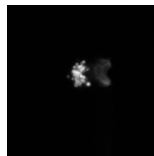
<sup>1</sup> Selection of one random sample per sequence

# Result excerpts - DMA Spheroid

- Image Pre-Processing

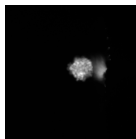


Raw



Pre-processed

- Pre-Labeling via Otsu thresholding



Sample



Mask



Pre-Label

80.18 %

*DSC* (Mask, Pre-Label)

# Conclusion and Outlook

## Conclusion

- Presentation of a generic workflow combining and extending various ideas of labeling enhancement
- Template for community usage in deep learning projects
- Software prototype which implements proposed workflow

## Outlook

- Extension of methods depicted in each assistance module
- Integration of other tasks (e.g. classification) in Label Assistant
- Open-source deployment of software prototype as `pip` package for community usage<sup>2</sup>

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<sup>2</sup>Git repository <https://git.scc.kit.edu/sc1357/kaida>

# Many thanks for your attention!



# References I

- [1] Davood Karimi et al. “Deep Learning with Noisy Labels: Exploring Techniques and Remedies in Medical Image Analysis”. In: *Medical Image Analysis* 65.5 (2020), p. 101759.
- [2] Niall O’ Mahony et al. “Deep Learning vs. Traditional Computer Vision”. In: *Advances in Computer Vision*. 2019, pp. 128–144.
- [3] Andreas Bartschat. *Image Labeling Tool*. Accessed: 2021-05-31, Available: <https://bitbucket.org/abartschat/imagelabelingtool>. 2019.
- [4] Mark Schutera et al. “Machine Learning Methods for Automated Quantification of Ventricular Dimensions”. In: *Zebrafish* 16.6 (2019), pp. 542–545.
- [5] Anna A. Popova et al. “Facile One Step Formation and Screening of Tumor Spheroids Using Droplet-Microarray Platform”. In: *Small* 15.25 (2019), p. 1901299.
- [6] O. Ronneberger, P.Fischer, and T. Brox. “U-Net: Convolutional Networks for Biomedical Image Segmentation”. In: *Medical Image Computing and Computer-Assisted Intervention (MICCAI)*. 2015, pp. 234–241.