

Bridging the Syn-to-Real Gap in Microorganism Detection Using Blended Synthetic Data

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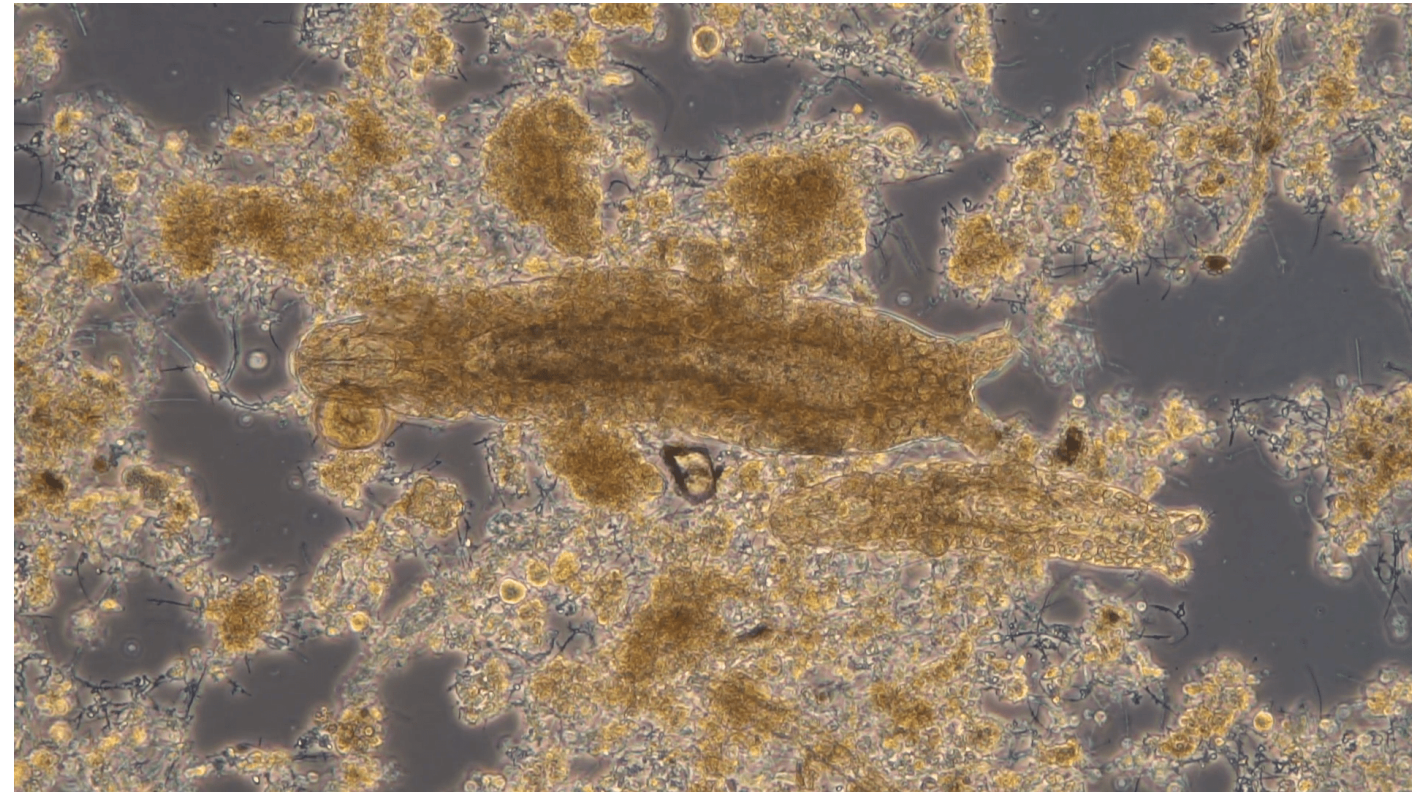
Monitoring Microorganisms

The number of different types of microorganisms is an indicator for the quality of the sewage plant process.

- **Overarching Goal:**
 - Activated Sludge Monitoring
 - Automated using YOLO object detection
- **Challenge:**
 - Limited Data
 - Homogeneous Data

The Domain Gap: Muddy vs. Clear

- **Idea:** Expand dataset with public data
- **Challenge:** Different domains
- **Generalisation:** Models struggle with real-world sewage data (our dataset) vs. clean public datasets [3]



Tardigrades in a muddy environment (e.g. in sewage plants).
Source: own image



Tardigrade in a clear environment. Source: [2]

Our Solution: Synthetic Data Augmentation

- Generate synthetic data for our 'muddy' domain
- Method: **Cut-Paste** (overlying objects on varied backgrounds)
- Challenge: **Syn-to-Real Gap** (unrealistic composites, pixel artifacts)

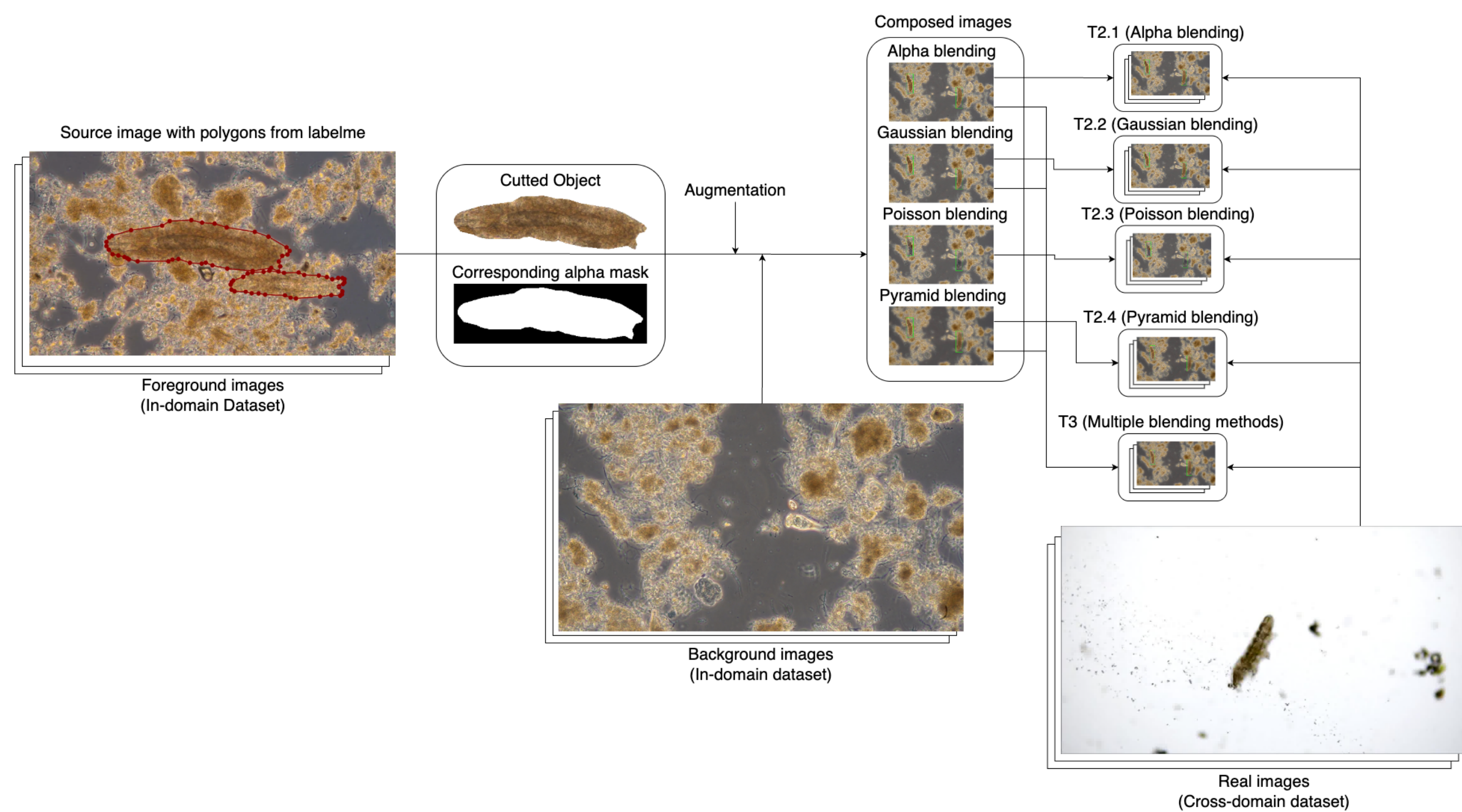
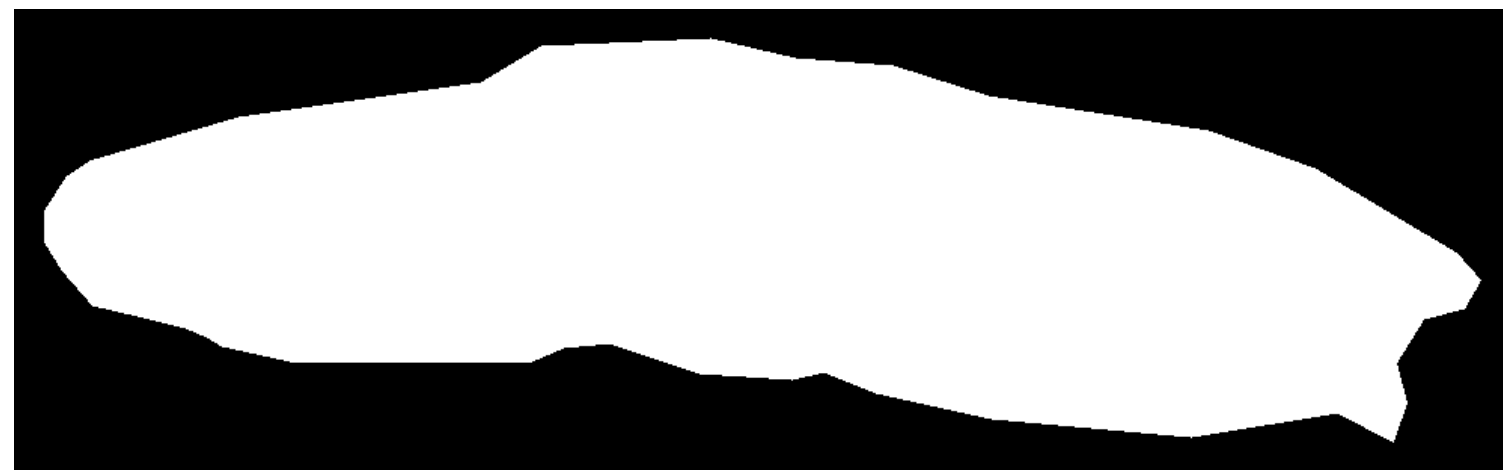


Abbildung 1: Procedure of the pipeline. Adapted from: [1]

Reduce the Syn-to-Real Gap

- **Blending:** Integrates objects seamlessly
- **Methods:**
 - Alpha blending
 - Gaussian blending
 - Poisson blending
 - Pyramid blending
 - Multi-Blending (Combination)
- Example of an Alpha Mask blended with Gaussian blending:



Alpha Mask

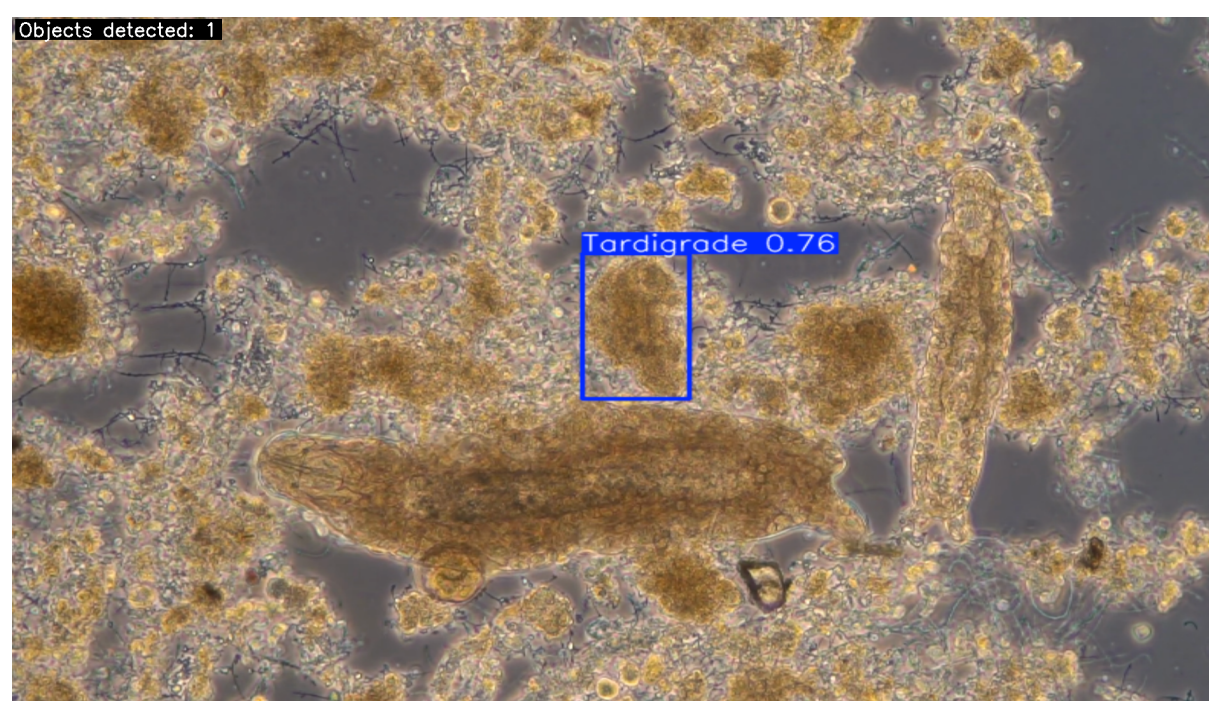


Blurred Alpha Mask with Gaussian blending

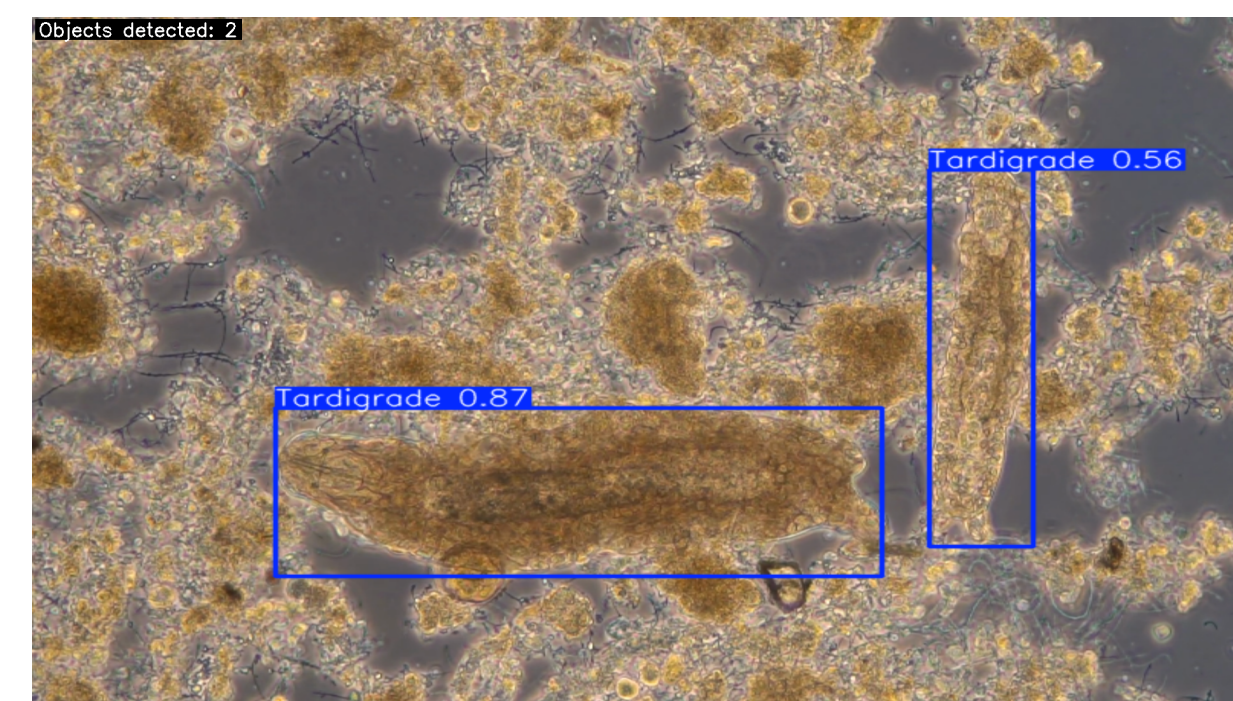
Evaluation & Datasets

- **Real Data:** Our In-Domain (**muddy**) & Public Cross-Domain (**clear**) datasets
- **Datasets for Evaluation:**
 - **T1: Baseline** (Real Data Only)
 - **T2: Single Blending** (Alpha, Gaussian, Poisson, Pyramid)
 - **T3: Multi-Blending** (Alpha + Gaussian + Pyramid)
- Varying **Synthetic Ratios:** 0%, 30%, 60%, 90%, 100%
- **Test Set:** 223 In-Domain images
- **Model:** YOLOv11m (25 Epochs)
- **Evaluation Metrics:**
 - mAP@[0.50:0.95] (Object detection performance)
 - FID & CMMD (Image Quality / Syn-Real Gap)

Improved Detection: Before vs. After



Prediction with a model trained on the T1 dataset
(Before)



Prediction with a model trained on the T4 dataset
(After)

Results

- **mAP Performance (YOLOv11m):**
 - Baseline (T1): **0.58 mAP**
 - Single Blending (T2): Up to **0.82 mAP** (Pyramid)
 - Poisson Blending: Underperformed significantly (0.27 mAP)
 - Multi-Blending (T3): **Best at 0.85 mAP**
 - **Optimal Ratio:** 60% Synthetic Data

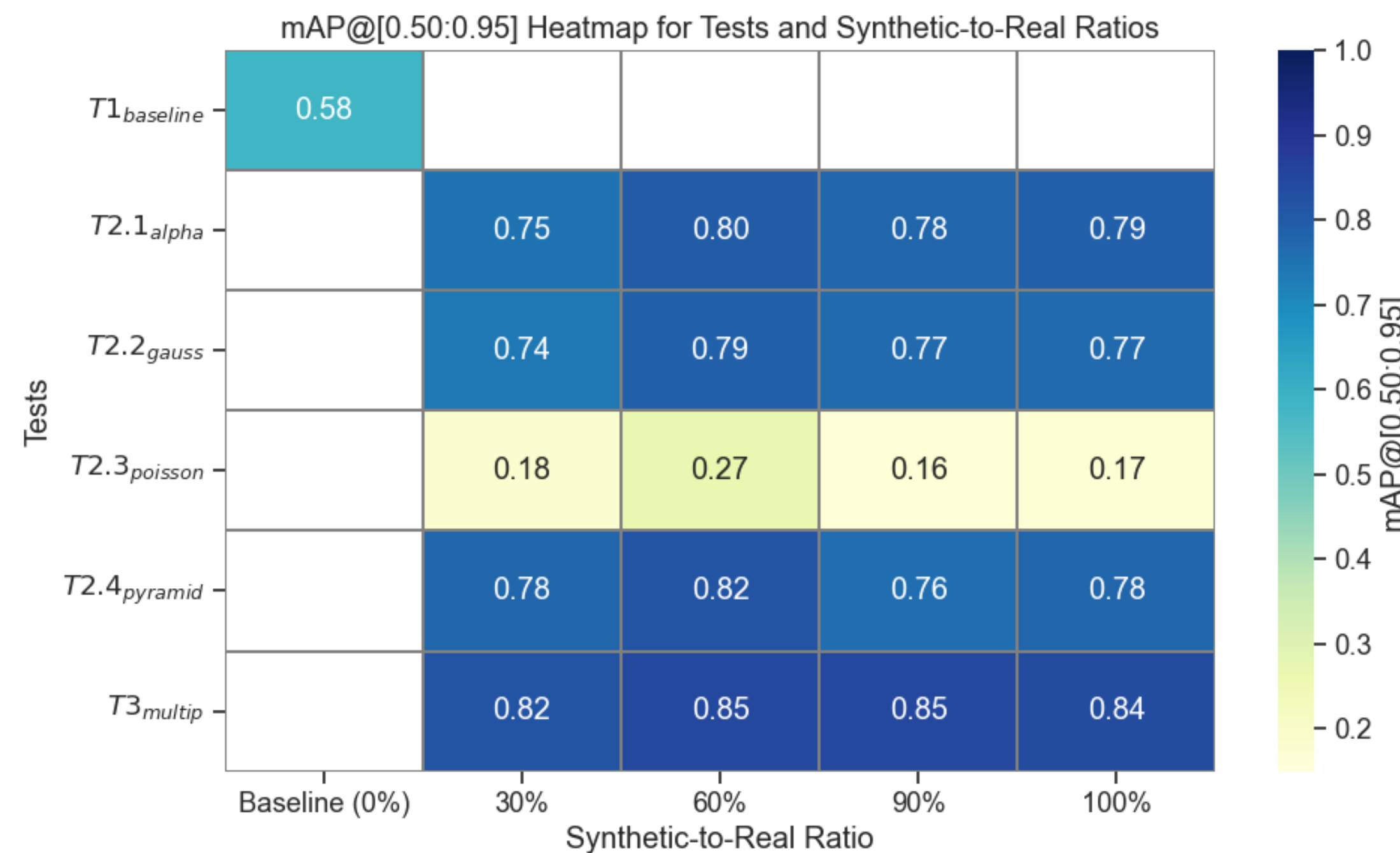


Abbildung 2: Heatmap of different models trained with different syn-to-real ratios and different blending methods

- **Image Quality (FID & CMMD):**
 - Lower scores = more realistic images
 - T1 (Baseline): Highest scores (least realistic).
 - **T3 (Multi-Blending): Lowest scores (most realistic)**
 - Correlates with mAP (smaller gap = better detection).

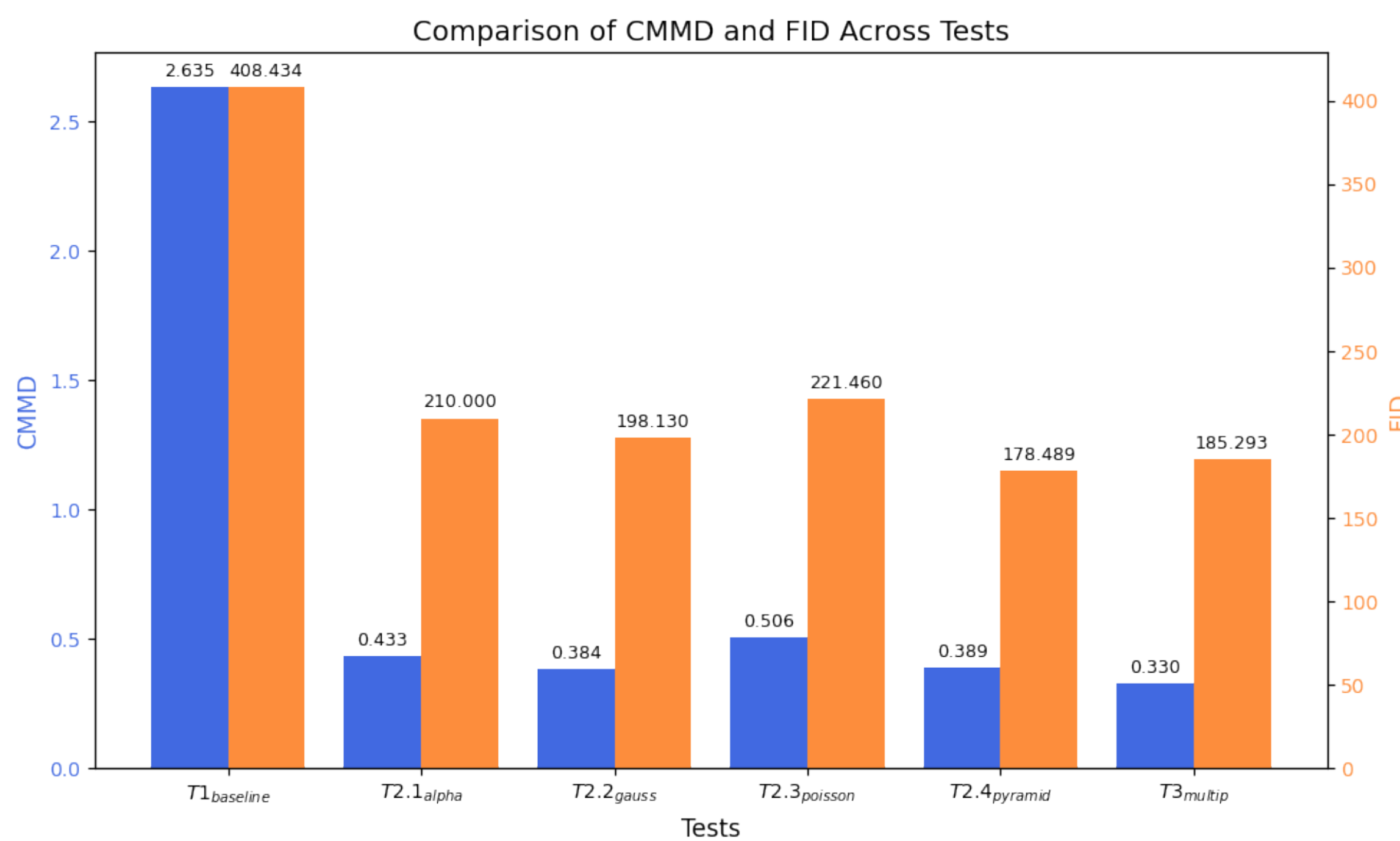


Abbildung 3: FID and CMMD, calculated on different test datasets in relation to the real dataset of the muddy environment

Conclusion & Future Directions

- **Synthetic data significantly boosts object detection in data-scarce domains**
- **Blending methods are crucial** for reducing the syn-real gap
- **Multi-blending excels**, yielding realistic images & highest mAP
- **Future Work:** Incorporate distractors (e.g., mud), noise, and explore more blending methods to enhance model resilience

References

- [1] Jonas Dirr u. a. "Cut-paste image generation for instance segmentation for robotic picking of industrial parts". In: *The International Journal of Advanced Manufacturing Technology* 130.1 (1. Jan. 2024), S. 191–201. DOI: 10.1007/s00170-023-12622-4.
- [2] Jasper John Jaso. *Tardigrade Dataset*. Roboflow. 2022. URL: <https://universe.roboflow.com/jasper-john-jaso/tardigrade> (besucht am 18.12.2024).
- [3] Luis Rangel DaCosta u. a. "A robust synthetic data generation framework for machine learning in high-resolution transmission electron microscopy (HRTEM)". In: *npj Computational Materials* 10.1 (29. Juli 2024). Publisher: Nature Publishing Group, S. 1–11. DOI: 10.1038/s41524-024-01336-0.