

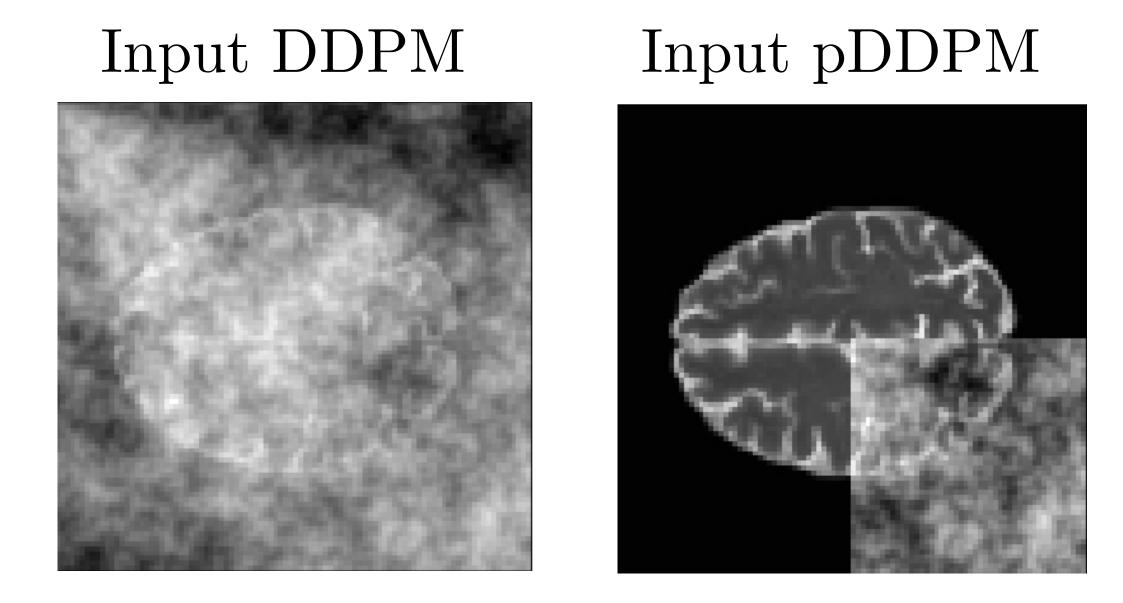
# Patched Diffusion Models for Unsupervised Anomaly Detection in Brain MRI

Finn Behrendt\* Debayan Bhattacharya\* Julia Krüger<sup>†</sup> Roland Opfer<sup>†</sup> Alexander Schlaefer\* \* Institute of Medical Technology and Intelligent Systems, Hamburg University of Technology, Germany <sup>†</sup>Jung Diagnostics GmbH, Hamburg, Germany

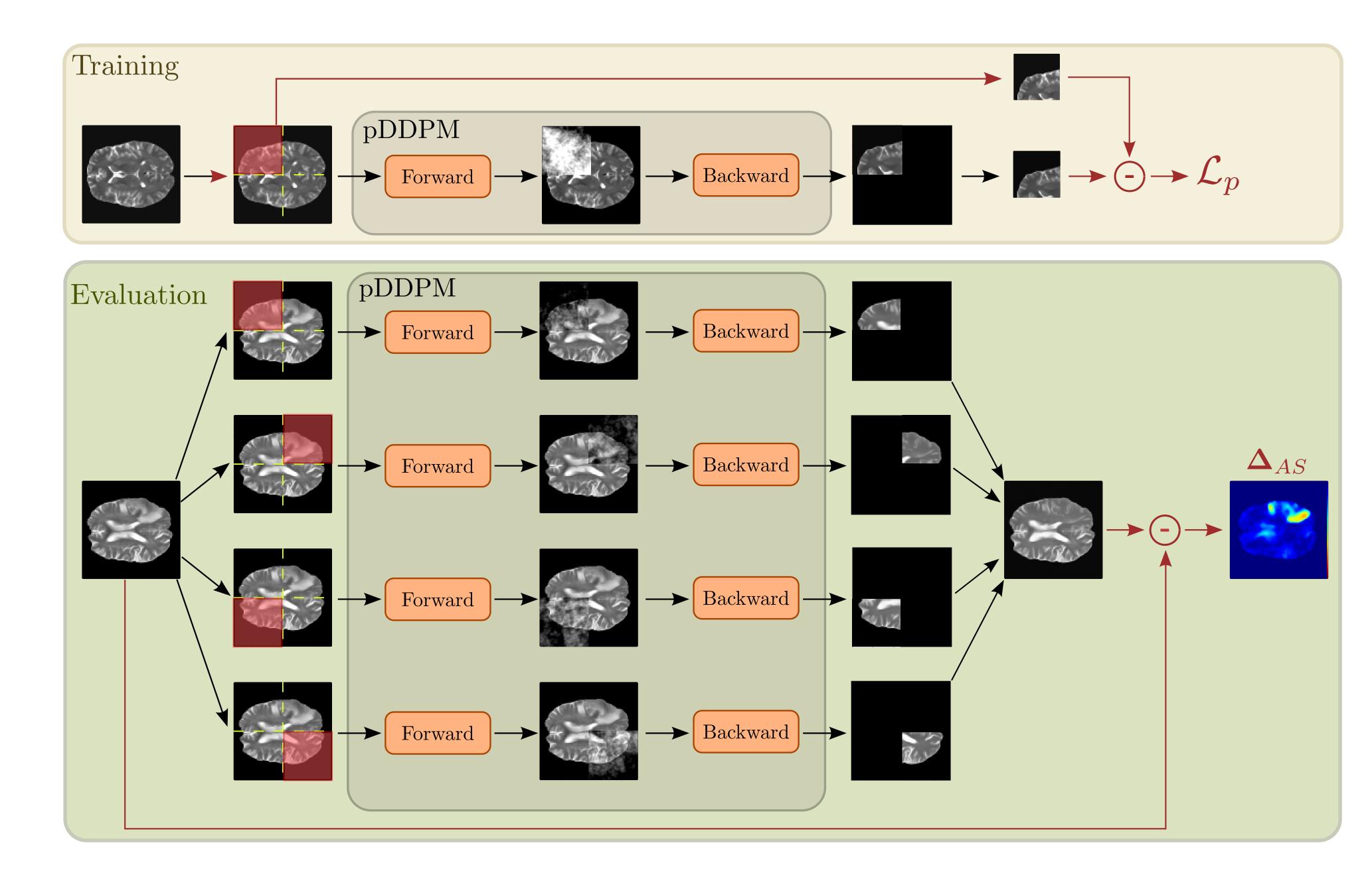
#### Introduction

- Diffusion models are well suited for unsupervised anomaly detection [3].
- However, with increasing noise during the forward process, detailed reconstructions become challenging.
- We propose a patch-based DDPM (pDDPM), where only a part of the input is noised and the patch surroundings provide context during reconstruction.

#### Which is easier to denoise?



# Methods



- In our approach, we apply the forward and backward process of DDPMs in a patched fashion.
- During the forward process, i.e., the noising step we add noise to the input image only at a given patch position.
- In the backward process, the partly noised image is fed to the denoising network to estimate the patch reconstruction.
- During evaluation, we slide the noise patch over the image, subsequently perform the diffusion process and stitch the individual patch reconstructions to obtain the full image.

#### Results

Table 1: Anomaly segmentation performance. DICE

## Conclusion

• pDDPMs allow to incorporate context information about individual

scores are reported in percent.

**MSLUB** Model BraTS21 AE [1]  $7.10{\pm}0.68$  $32.87 \pm 1.25$ VAE [1]  $31.11 \pm 1.50$  $6.89 \pm 0.09$ **DAE** [2]  $37.05 \pm 1.42$  $3.56 \pm 0.91$ DDPM [3]  $40.67 \pm 1.21$  $6.42{\pm}1.60$  $49.00 \pm 0.84 \ 10.35 \pm 0.69$ pDDPM

brain structure and appearance when estimating brain anatomy.

Thereby, pDDPMs outperform both their non-patched counterparts and various baseline methods considering the UAD performance.

## References

[1] Baur, Christoph et al.: Autoencoders for unsupervised anomaly segmentation in brain mr images: a comparative study. Medical Image Analysis, pp. 101952, 2021.

[2] Kascenas, Antanas et al.: *Denoising autoencoders for unsupervised anomaly detection in brain mri.* MIDL/PMLR, 2022.

[3] Wyatt, Julian et al: Anoddpm: Anomaly detection with denoising diffusion probabilistic models using simplex noise. CVPR, pp. 650-656, 2022.