



Figure 6: Visualization of anomaly localization on MVTec-AD [3]. The columns from left to right present anomalous images, the corresponding ground truths, our pixel-level predictions based on KL divergence, and the visualization of the final anomaly map on the original image. Our denoising diffusion models produce boundary-aware pixel-level anomaly scores to improve the quality of final anomaly scores.

adopt EfficientNet [30] as our feature extractor, where the dimension of features is relatively small. We select the intermediate feature map of stride (2, 4, 8, 16) with dimension (24, 32, 56, 160), resize to  $64 \times 64$  and concatenate them together into a tensor  $\mathbf{f} \in \mathbb{R}^{272 \times 64 \times 64}$ . We conduct experiments to verify the resolution of the concatenated features; see Tab. 2. Increasing resolution from  $16 \times 16$  to  $64 \times 64$  increases the PRO by 2.5%. Combining EfficientNet features of stride (2, 4, 8, 16) improves the PRO by 3.5%.

**Unified model.** Since our denoising diffusion model is capable of modeling complex real industrial data distribution, we conduct experiments to use a single model for anomaly localization for all categories of MVTec-AD, see Tab. 3. The performance of DRAEM [40] and PaDiM [8] drops greatly for the unified setting, with more than 5% degradation in the PRO metric. In contrast, our model still achieves 93.0% in PRO, with less than 1.1% performance drop.

#### 4.5. Qualitative results of reconstruction.

We show in Fig. 7 that our denoising gradient process can smoothly transform an anomalous image into a normal one under the guidance of a pre-trained feature extractor. The left two columns are ground truth and input anomalous

Resize	Stride				AUROC	PRO
	16	8	4	2		
64	✓				94.2	90.2
	✓	✓			95.8	91.9
	✓	✓	✓		96.4	93.2
	✓	✓	✓	✓	<b>96.7</b>	<b>93.7</b>
32	✓	✓	✓	✓	96.6	92.4
16	✓	✓	✓	✓	95.5	91.2

Table 2: Ablation study of feature level reconstruction. Experiments are conducted with features from different layers of EfficientNet [30], resized to different resolutions.

Method	Base		Unified	
	AUROC	PRO	AUROC	PRO
PaDiM[8]	<b>97.8</b>	92.8	90.5	85.3
DRAEM[40]	97.3	93.0	89.4	82.2
UniAD[36]	96.6	-	<b>97.0</b>	91.1
Ours	96.7	<b>94.1</b>	96.0	<b>93.0</b>

Table 3: Comparison of a single unified model for anomaly localization of all the categories on MVTec-AD.