Table 2: Anomaly detection results under image-level AUROC metric on MVTec-AD [4].

	Texture					Object										
	Carp.	Grid	Leat.	Tile	Wood	Bott.	Cable	Caps.	Haze.	Meta.	Pill	Screw	Toot.	Tran.	Zipp.	Mean
GANomaly [2] SCADN [36] ARNet [14] SPADE [10] KDAD [30] PSVDD [37] TS [5]	50.4 70.6 - 79.3 98.6	98.3 88.3 78.0 90.3	65.9 86.2 95.1 76.7	$79.2 \\ 73.5 \\ - \\ 91.6 \\ 92.9$	96.8 92.3 - 94.3 94.6	95.7 94.1 - 99.4 92.0	85.6 83.2 - 89.2 90.9	76.5 68.1 80.5 94.0	83.3 85.5 - 98.4 86.1	62.4 66.7 73.6 81.3	81.4 78.6 - 82.7 97.8	83.1 100 - 83.3 100	98.1 100 - 92.2 91.5	86.3 84.3 - 85.6 96.5	74.5 84.6 87.6 - 93.2 97.9 95.8	81.8 83.9 85.5 87.7 92.1
ADTR(ours) ADTR+(ours)																

Table 3: Anomaly detection results under image-level AUROC metric on CIFAR-10 [18].

	Airplane	Automobile	Bird	Cat	Deer	Dog	Frog	Horse	Ship	Truck	Mean
VAE [3]	63.4	44.2	64.0	49.7	74.3	51.5	74.5	52.7	67.4	41.6	58.3
KDE [7]	65.8	52.0	65.7	49.7	72.7	49.6	75.8	56.4	68.0	54.0	61.0
AnoGAN [31]	67.1	54.7	52.9	54.5	65.1	60.3	58.5	62.5	75.8	66.5	61.8
LSA [1]	73.5	58.0	69.0	54.2	76.1	54.6	75.1	53.5	71.7	54.8	64.1
DSVDD [28]	61.7	65.9	50.8	59.1	60.9	65.7	67.7	67.3	75.9	73.1	64.8
OCGAN [26]	75.7	53.1	64.0	62.0	72.3	62.0	72.3	57.5	82.0	55.4	65.7
GradCon [19]	76.0	59.8	64.8	58.6	73.3	60.3	68.4	56.7	78.4	67.8	66.4
Loc-Glo [34]	79.1	70.3	67.5	56.1	73.9	63.8	73.2	67.4	81.4	72.2	70.5
TS [5]	78.9	84.9	73.4	74.8	85.1	79.3	89.2	83.0	86.2	84.8	82.0
GT[15]	76.2	84.8	77.1	73.2	82.8	84.8	82.0	88.7	89.5	83.4	82.3
KDAD [30]	90.5	90.4	80.0	77.0	86.7	91.4	89.0	86.8	91.5	88.9	87.2
ADTR(ours)	94.1	97.4	92.3	89.0	93.2	94.4	97.4	95.8	96.3	96.7	94.7
ADTR+(ours)	96.2	98.0	94.5	91.7	95.1	95.6	98.0	97.1	98.0	96.9	96.1

anomaly-available case, the performance of ADTR+ is further improved by 1.4% with the help of external irrelevant dataset, reflecting the effectiveness of the designed image-level loss function, \mathcal{L}_{ima} .

4.4 Ablation Study

Extensive ablation studies with pixel-level AUROC metric are conducted on anomaly localization task of MVTec-AD [4].

Attention and auxiliary query embedding. As shown in Tab. 4a, a CNN revised from ResNet [17] is firstly included as the baseline of the reconstruction model. (1) The replacement of the attention layer is a concatenation followed by projection. If we remove the attention layer (w/o Attn) from the transformer, the performance shows no obvious superiority to CNN. (2) Without the auxiliary query embedding (w/o Query), meaning that only the encoder embedding is input to the decoder, the performance is even worse than CNN. (3) Equipped with both attention and auxiliary query embedding (Attn+Query), transformer stably outperforms CNN by 2.8%. This proves our assertion in Sec. 3.2 that the auxiliary query embedding in attention layer helps prevent transformer from reconstructing anomalies well.