

GANomaly: Semi-Supervised Anomaly Detection via Adversarial Training

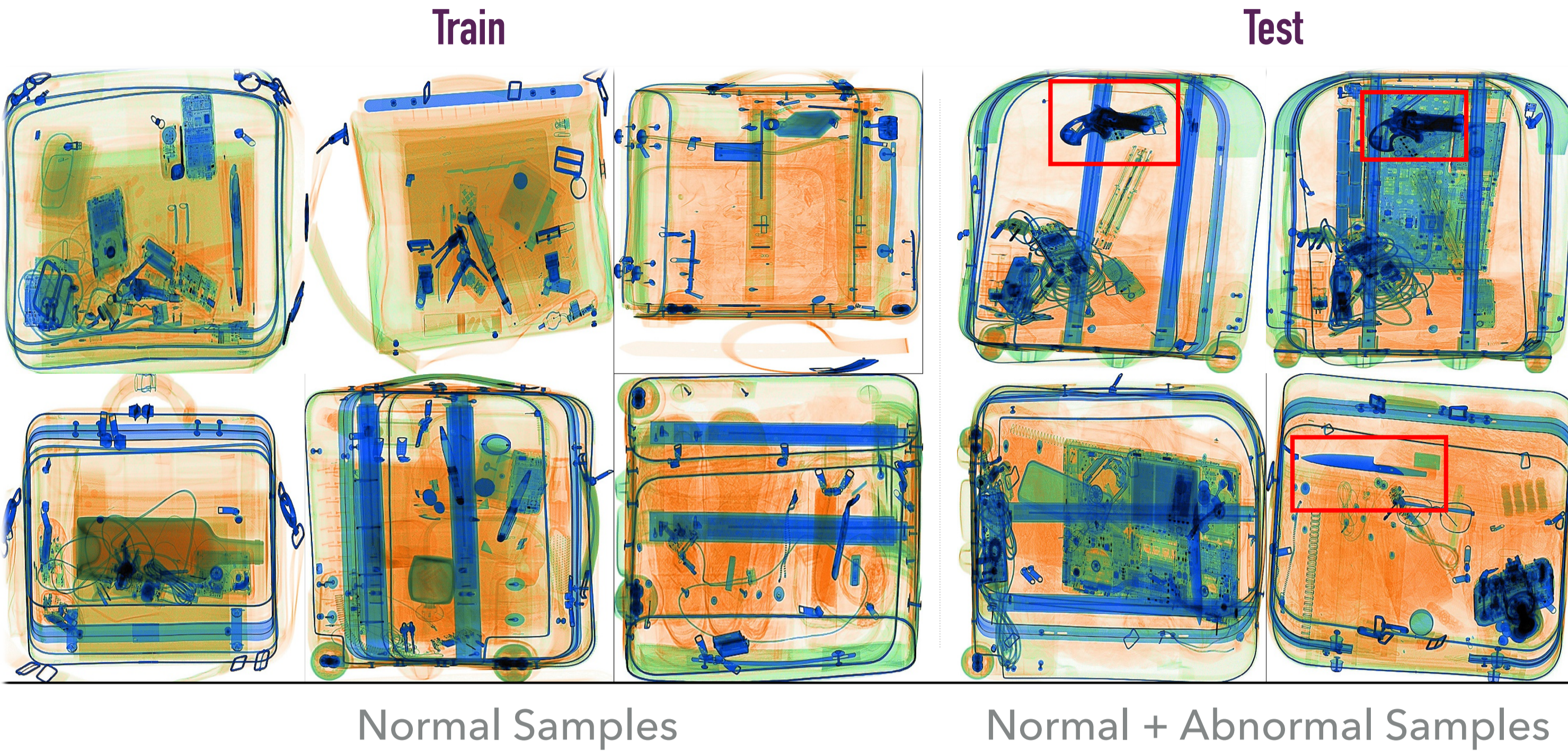
Samet Akcay, Amir Atapour-Abarghouei and Toby Breckon

Department of Computer Science, Durham University, UK



Motivation

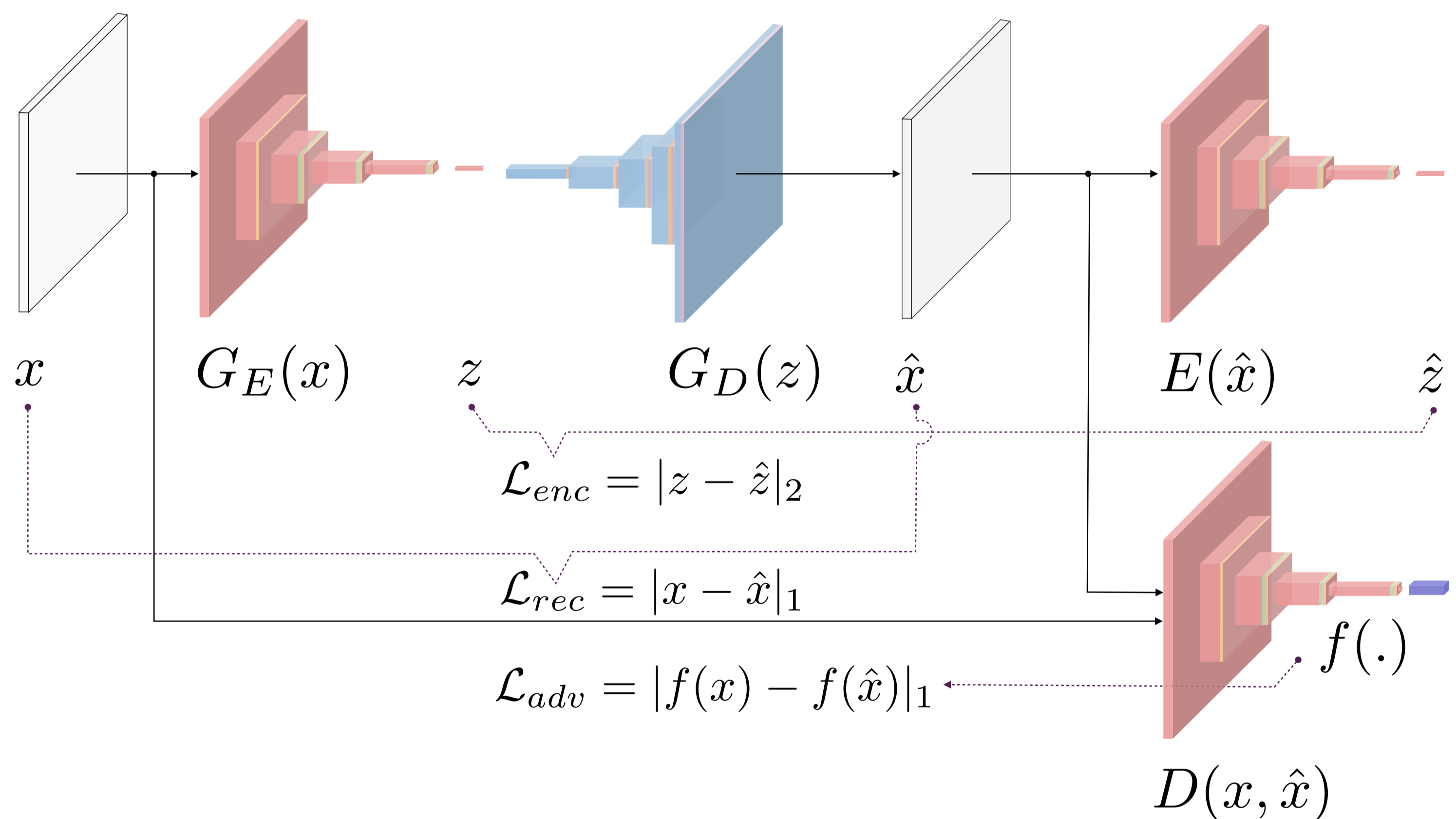
Datasets that are highly biased towards one class (normal) due to the insufficient sample size of the other class (abnormal) pose the task of anomaly detection - especially when abnormality collection is challenging.



Proposed Approach

(i) **Train a novel encoder-decoder-encoder network** to jointly capture the distribution of the normal data via the generation of high-dimensional image space and the inference of latent space.

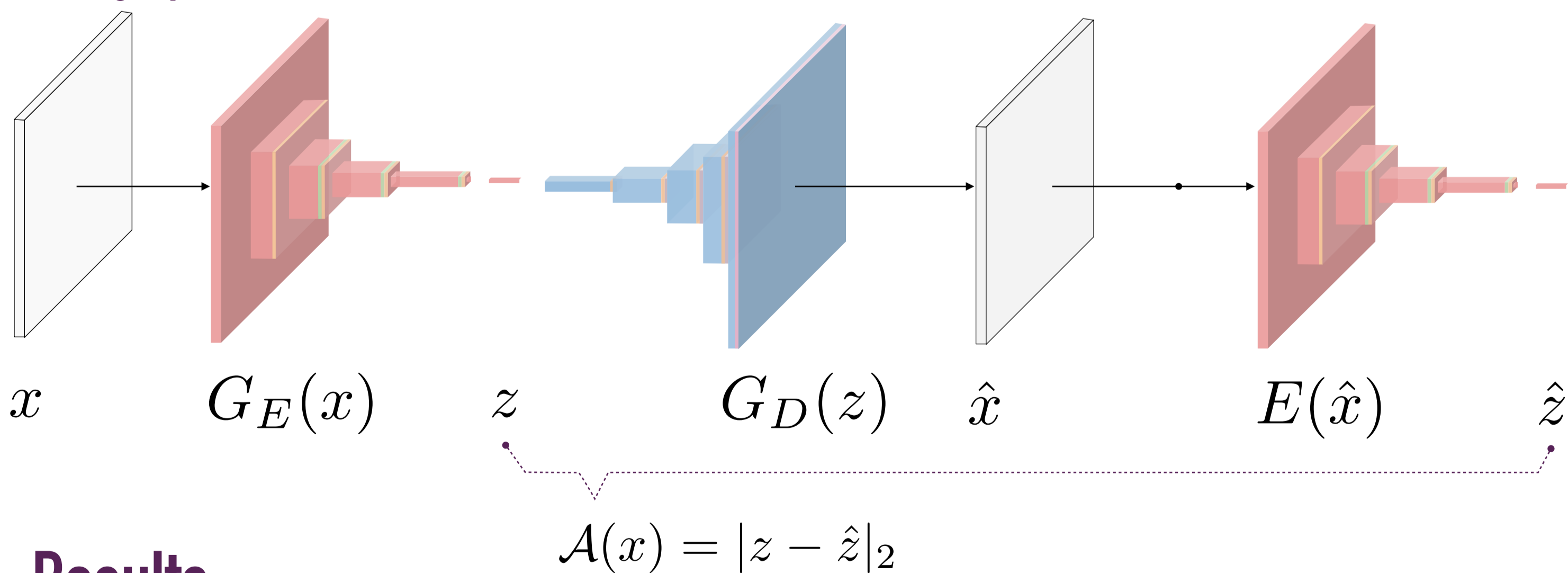
Training Pipeline



Legend: Input/Output (white), Conv (red), LeakyReLU (orange), BatchNorm (green), ConvTranspose (blue), ReLU (grey), Tanh (pink), Softmax (purple)

(ii) **Test anomalies within the latent space.** Anomaly detection is performed by computing the difference between latent vector representation of real and fake images.

Testing Pipeline

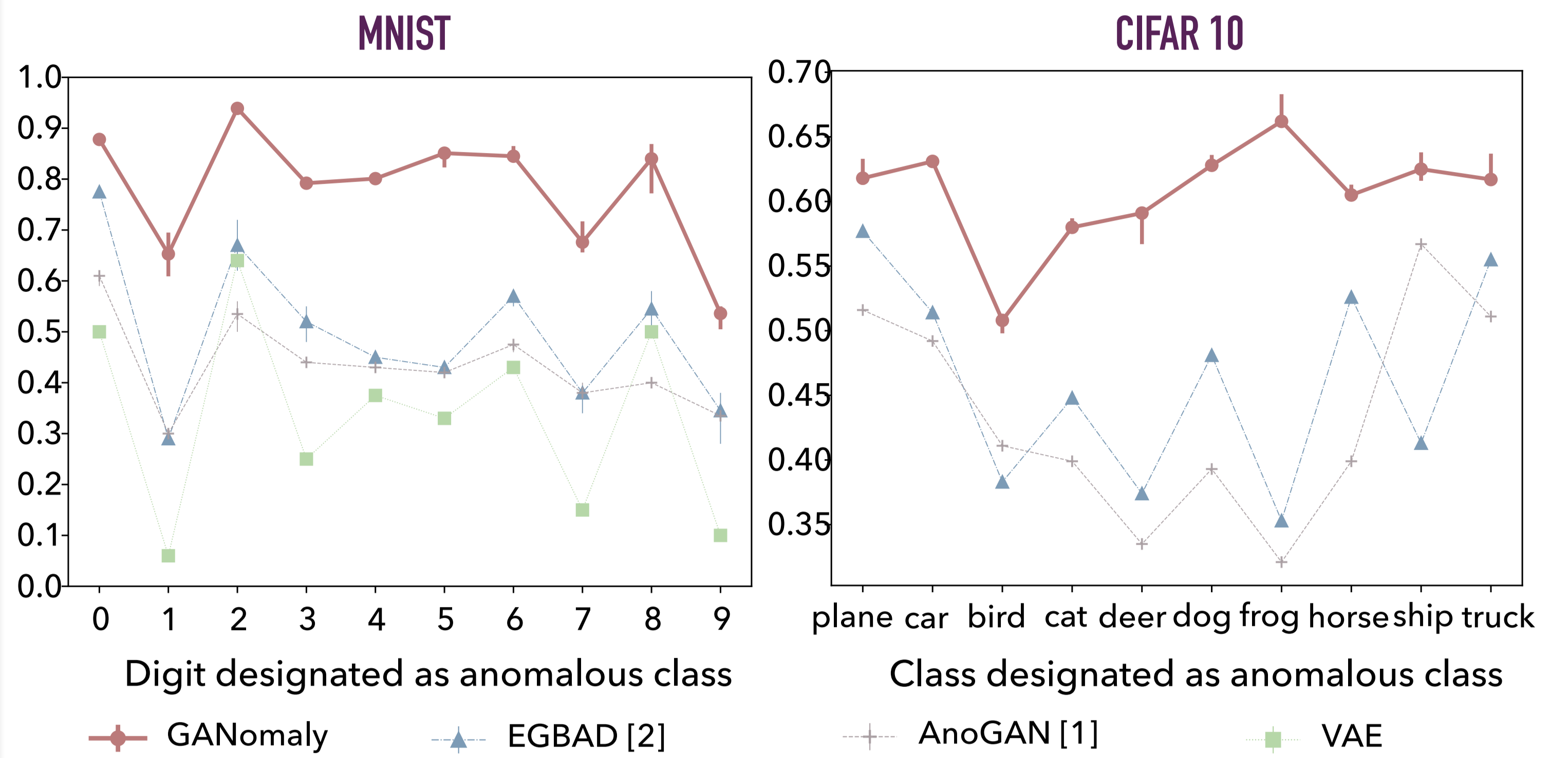


Results

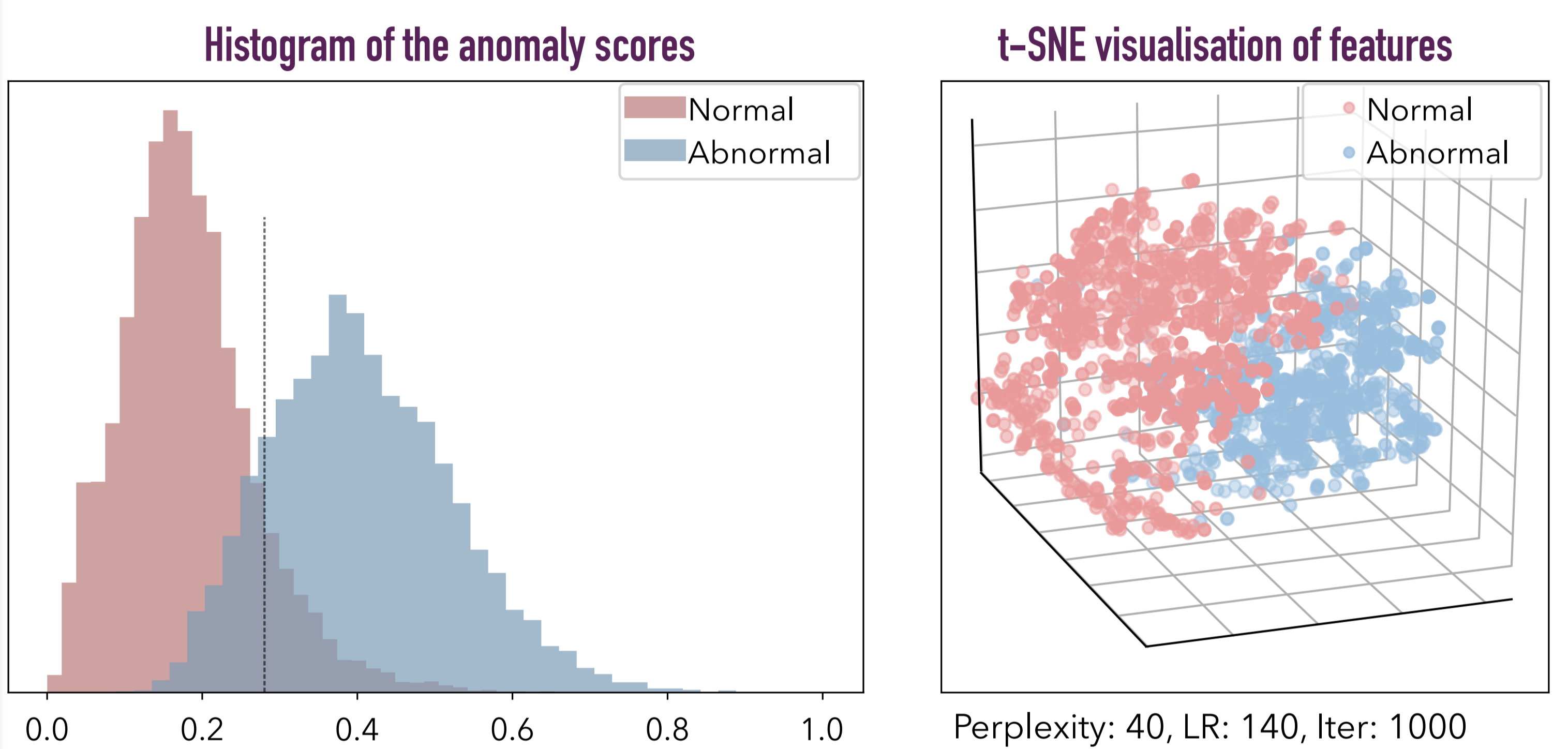
Our approach yields **quantitatively and computationally superior** performance in terms of run-time (ms) and abnormality detection (AUC).

Runtime (ms)	MNIST	CIFAR	UBA	FFOB
AnoGAN [1]	7120	7120	7110	7223
EGBAD [2]	8.92	8.71	8.88	8.87
GANomaly	2.79	2.21	2.66	2.53

GANomaly achieves the **highest AUC** on public datasets – MNIST and CIFAR



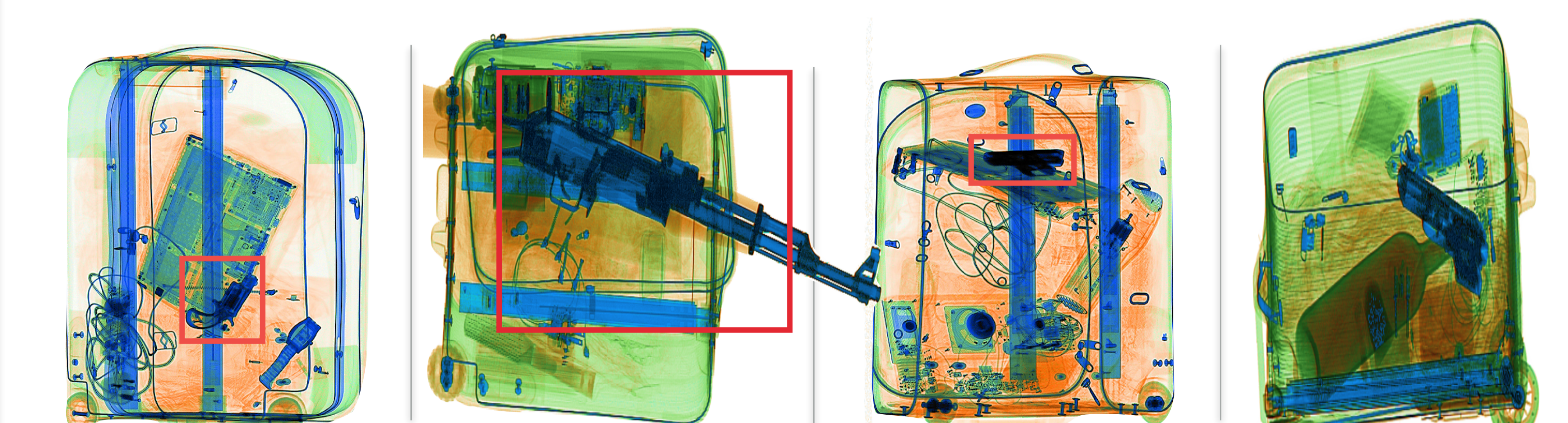
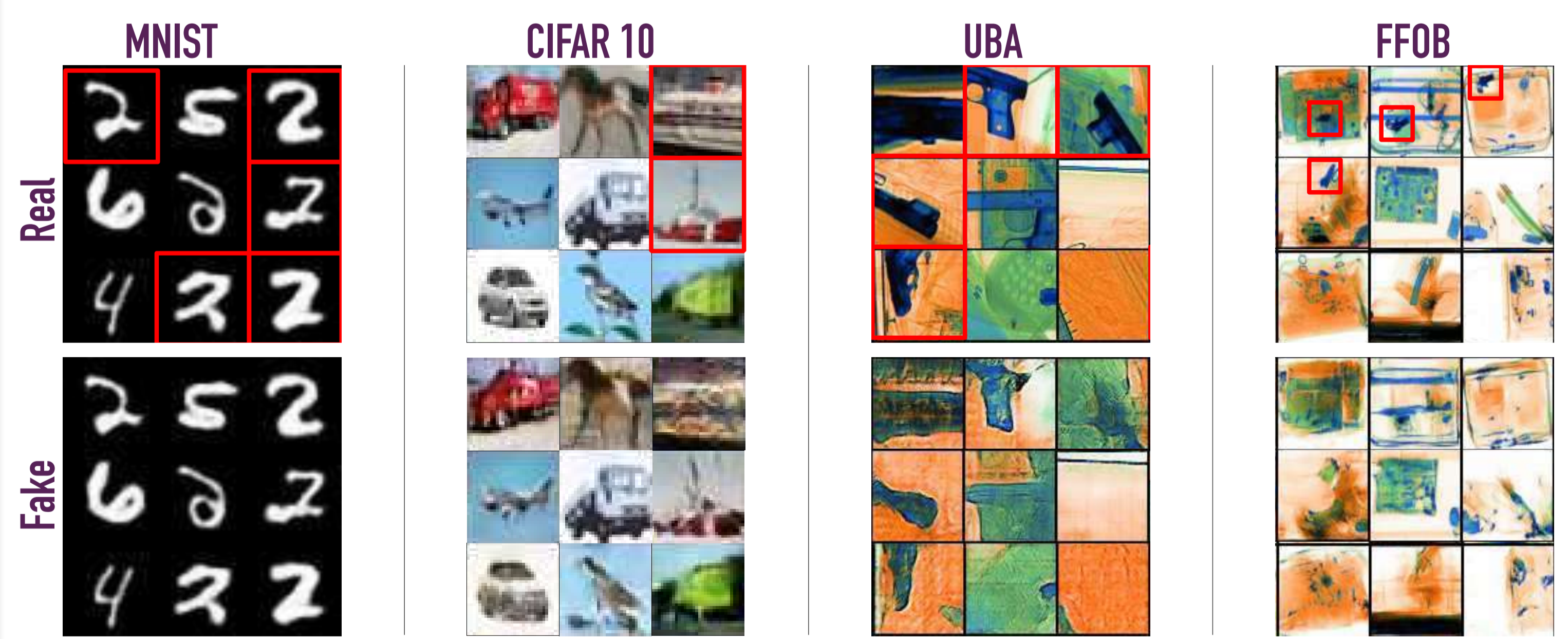
The model provides an **optimal feature space classification/separation.**



Our approach also **outperforms previous state-of-the-art** on challenging X-ray baggage security screening domain.

X-ray Datasets	UBA			FFOB
Model	gun	gun parts	knife	overall full weapon
AnoGAN [1]	0.598	0.511	0.599	0.569 0.703
EGBAD [2]	0.614	0.591	0.587	0.597 0.712
GANomaly	0.747	0.662	0.520	0.643 0.882

Our GAN approach generates normal samples but fails to produce abnormal ones - illustrating **effective capture of the normal class distribution.**



[1] Schlegl, T., Seeböck, P., Waldstein, S. M., Schmidt-Erfurth, U., & Langs, G. (2017). Unsupervised anomaly detection with generative adversarial networks to guide marker discovery. Lecture Notes in Computer Science, 10265 LNCS, 146-147.

[2] Zenati, H., Foo, C. S., Lecouat, B., Manek, G., & Chandrasekhar, V. R. (2018). Efficient GAN-Based Anomaly Detection. <http://arxiv.org/abs/1802.06222>

<https://github.com/samet-akcay/ganomaly>



Paper



Code