

## 2 Unsupervised AD

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### 2.1 Feature-Embedding-based Methods

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#### 2.1.1 Teacher-Student

- Contextual Affinity Distillation for Image Anomaly Detection [\[WACV 2024\]](#)
- Revisiting Reverse Distillation for Anomaly Detection [\[CVPR 2023\]](#) [\[code\]](#)
- Uninformed students: Student-teacher anomaly detection with discriminative latent embeddings [\[CVPR 2020\]](#)
- Multiresolution knowledge distillation for anomaly detection [\[CVPR 2021\]](#)
- Glancing at the Patch: Anomaly Localization With Global and Local Feature Comparison [\[CVPR 2021\]](#)
- Reconstruction Student with Attention for Student-Teacher Pyramid Matching [\[2021\]](#)
- Student-Teacher Feature Pyramid Matching for Anomaly Detection [\[2021\]](#)[\[code\]](#)
- PFM and PEFM for Image Anomaly Detection and Segmentation [\[CASE 2022\]](#) [\[TII 2022\]](#)[\[code\]](#)
- Reconstructed Student-Teacher and Discriminative Networks for Anomaly Detection [\[2022\]](#)
- Anomaly Detection via Reverse Distillation from One-Class Embedding [\[CVPR 2022\]](#)[\[code\]](#)
- Asymmetric Student-Teacher Networks for Industrial Anomaly Detection [\[WACV 2022\]](#)[\[code\]](#)
- Informative knowledge distillation for image anomaly segmentation [\[2022\]](#)[\[code\]](#)
- Remembering Normality: Memory-guided Knowledge Distillation for Unsupervised Anomaly Detection [\[ICCV 2023\]](#)
- A Discrepancy Aware Framework for Robust Anomaly Detection [\[2023\]](#)[\[code\]](#)
- Enhanced multi-scale features mutual mapping fusion based on reverse knowledge distillation for industrial anomaly detection and localization [\[TBD 2024\]](#)
- AEKD: Unsupervised auto-encoder knowledge distillation for industrial anomaly detection [\[JMS 2024\]](#)
- Masked feature regeneration based asymmetric student-teacher network for anomaly detection [\[Multimedia Tools and Applications 2024\]](#)

#### 2.1.2 One-Class Classification (OCC)

- Patch svdd: Patch-level svdd for anomaly detection and segmentation [\[ACCV 2020\]](#)
- Anomaly detection using improved deep SVDD model with data structure preservation [\[2021\]](#)

- A Semantic-Enhanced Method Based On Deep SVDD for Pixel-Wise Anomaly Detection [\[2021\]](#)
- MOCCA: Multilayer One-Class Classification for Anomaly Detection [\[2021\]](#)
- Defect Detection of Metal Nuts Applying Convolutional Neural Networks [\[2021\]](#)
- Panda: Adapting pretrained features for anomaly detection and segmentation [\[2021\]](#)
- Mean-shifted contrastive loss for anomaly detection [\[2021\]](#)
- Learning and Evaluating Representations for Deep One-Class Classification [\[2020\]](#)
- Self-supervised learning for anomaly detection with dynamic local augmentation [\[2021\]](#)
- Contrastive Predictive Coding for Anomaly Detection [\[2021\]](#)
- Cutpaste: Self-supervised learning for anomaly detection and localization [\[ICCV 2021\]](#)[\[unofficial code\]](#)
- Consistent estimation of the max-flow problem: Towards unsupervised image segmentation [\[2020\]](#)
- MemSeg: A semi-supervised method for image surface defect detection using differences and commonalities [\[2022\]](#)[\[unofficial code\]](#)
- SimpleNet: A Simple Network for Image Anomaly Detection and Localization [\[CVPR 2023\]](#)[\[code\]](#)
- End-to-End Augmentation Hyperparameter Tuning for Self-Supervised Anomaly Detection [\[2023\]](#)
- Anomaly Detection under Distribution Shift [\[ICCV 2023\]](#)[\[code\]](#)
- Learning Transferable Representations for Image Anomaly Localization Using Dense Pretraining [\[WACV 2024\]](#)[\[code\]](#)

### 2.1.3 Distribution-Map

- Anomaly Detection in Nanofibrous Materials by CNN-Based Self-Similarity [\[Sensors 2018\]](#)
- A Multi-Scale A Contrario method for Unsupervised Image Anomaly Detection [\[2021\]](#)
- Modeling the distribution of normal data in pre-trained deep features for anomaly detection [\[2021\]](#)
- Transfer Learning Gaussian Anomaly Detection by Fine-Tuning Representations [\[2021\]](#)
- PEDENet: Image anomaly localization via patch embedding and density estimation [\[2022\]](#)
- Unsupervised image anomaly detection and segmentation based on pre-trained feature mapping [\[2022\]](#)
- Position Encoding Enhanced Feature Mapping for Image Anomaly Detection [\[2022\]](#)[\[code\]](#)
- Focus your distribution: Coarse-to-fine non-contrastive learning for anomaly detection and localization [\[ICME 2022\]](#)
- Anomaly Detection of Defect using Energy of Point Pattern Features within Random Finite Set Framework [\[2021\]](#)[\[code\]](#)

- Fastflow: Unsupervised anomaly detection and localization via 2d normalizing flows [\[2021\]](#) [\[unofficial code\]](#)
- Same same but differnet: Semi-supervised defect detection with normalizing flows [\[WACV 2021\]](#) [\[code\]](#)
- Fully convolutional cross-scale-flows for image-based defect detection [\[WACV 2022\]](#)[\[code\]](#)
- Cflow-ad: Real-time unsupervised anomaly detection with localization via conditional normalizing flows [\[WACV 2022\]](#)[\[code\]](#)
- CAINNFlow: Convolutional block Attention modules and Invertible Neural Networks Flow for anomaly detection and localization tasks [\[2022\]](#)
- AltUB: Alternating Training Method to Update Base Distribution of Normalizing Flow for Anomaly Detection [\[2022\]](#)
- Collaborative Discrepancy Optimization for Reliable Image Anomaly Localization [\[TII 2023\]](#)[\[code\]](#)
- PyramidFlow: High-Resolution Defect Contrastive Localization using Pyramid Normalizing Flow [\[CVPR 2023\]](#)[\[code\]](#)
- Attention Modules Improve Image-Level Anomaly Detection for Industrial Inspection: A DifferNet Case Study [\[WACV 2024\]](#)
- Fascinating Supervisory Signals and Where to Find Them: Deep Anomaly Detection with Scale Learning [\[ICML 2023\]](#)
- FRAnomaly: flow-based rapid anomaly detection from images [\[Applied Intelligence 2024\]](#)
- Image alignment-based patch distribution framework for anomaly detection [\[ICCVDM 2024\]](#)

## 2.1.4 Memory Bank

- ReConPatch: Contrastive Patch Representation Learning for Industrial Anomaly Detection [\[WACV 2024\]](#)

+ Sub-image anomaly detection with deep pyramid correspondences [\[\[2020\]\]](#)(<https://arxiv.org/pdf/2005.02357.pdf>) + Semi-orthogonal embedding for efficient unsupervised anomaly segmentation [\[\[2021\]\]](#)(<https://arxiv.org/pdf/2105.14737.pdf>) + Anomaly Detection Via Self-Organizing Map [\[\[2021\]\]](#) (<http://arxiv.org/pdf/2107.09903>) + PaDiM: A Patch Distribution Modeling Framework for Anomaly Detection and Localization [\[\[ICPR 2021\]\]](#)([https://link.springer.com/chapter/10.1007/978-3-030-68799-1\\_35](https://link.springer.com/chapter/10.1007/978-3-030-68799-1_35))[\[\[unofficial code\]\]](#)(<https://github.com/xiahaifeng1995/PaDiM-Anomaly-Detection-Localization-master>) + Industrial Image Anomaly Localization Based on Gaussian Clustering of Pretrained Feature [\[\[2021\]\]](#)(<https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=9479740>) + Towards total recall in industrial anomaly detection[\[\[CVPR 2022\]\]](#)(<http://arxiv.org/pdf/2106.08265>)[\[\[code\]\]](#)(<https://github.com/amazon-science/patchcore-inspection>) + CFA: Coupled-Hypersphere-Based Feature Adaptation for Target-Oriented Anomaly Localization[\[\[2022\]\]](#) (<https://arxiv.org/pdf/2206.04325.pdf>)[\[\[code\]\]](#)(<https://github.com/sungwool/>

CFA\_for\_anomaly\_localization) + FAPM: Fast Adaptive Patch Memory for Real-time Industrial Anomaly Detection [[2022]](<https://arxiv.org/pdf/2211.07381.pdf>) + N-pad: Neighboring Pixel-based Industrial Anomaly Detection [[2022]](<https://arxiv.org/pdf/2210.08768.pdf>) + Multi-scale patch-based representation learning for image anomaly detection and segmentation [[2022]]([https://openaccess.thecvf.com/content/WACV2022/papers/Tsai\\_Multi-Scale\\_Patch-Based\\_Representation\\_Learning\\_for\\_Image\\_Anomaly\\_Detection\\_and\\_Segmentation\\_WACV\\_2022\\_paper.pdf](https://openaccess.thecvf.com/content/WACV2022/papers/Tsai_Multi-Scale_Patch-Based_Representation_Learning_for_Image_Anomaly_Detection_and_Segmentation_WACV_2022_paper.pdf)) + SPot-the-Difference Self-supervised Pre-training for Anomaly Detection and Segmentation [[ECCV 2022]](<https://arxiv.org/pdf/2207.14315.pdf>) + Diversity-Measurable Anomaly Detection [[CVPR 2023]](<https://arxiv.org/abs/2303.05047>) + SelfFormaly: Towards Task-Agnostic Unified Anomaly Detection [[2023]](<https://arxiv.org/abs/2307.12540>) + REB: Reducing Biases in Representation for Industrial Anomaly Detection [[2023]](<https://arxiv.org/abs/2308.12577>)[[code]](<https://github.com/ShuailYU/REB>) + PNI : Industrial Anomaly Detection using Position and Neighborhood Information [[ICCV 2023]]([https://openaccess.thecvf.com/content/ICCV2023/papers/Bae\\_PNI\\_Industrial\\_Anomaly\\_Detection\\_using\\_Position\\_and\\_Neighborhood\\_Information\\_ICCV\\_2023\\_paper.pdf](https://openaccess.thecvf.com/content/ICCV2023/papers/Bae_PNI_Industrial_Anomaly_Detection_using_Position_and_Neighborhood_Information_ICCV_2023_paper.pdf))[[code]]([https://github.com/wogur110/PNI\\_Anomaly\\_Detection](https://github.com/wogur110/PNI_Anomaly_Detection)) + Inter-Realization Channels: Unsupervised Anomaly Detection Beyond One-Class Classification [[ICCV 2023]]([https://openaccess.thecvf.com/content/ICCV2023/papers/McIntosh\\_Inter-Realization\\_Channels\\_Unsupervised\\_Anomaly\\_Detection\\_Beyond\\_One-Class\\_Classification\\_ICCV\\_2023\\_paper.pdf](https://openaccess.thecvf.com/content/ICCV2023/papers/McIntosh_Inter-Realization_Channels_Unsupervised_Anomaly_Detection_Beyond_One-Class_Classification_ICCV_2023_paper.pdf))[[code]](<https://github.com/DeclanMcIntosh/InReaCh>) + Grid-Based Continuous Normal Representation for Anomaly Detection [[2024]](<https://arxiv.org/abs/2402.18293>)[[code]](<https://github.com/tae-mo/GRAD>) + PointCore: Efficient Unsupervised Point Cloud Anomaly Detector Using Local-Global Features [[2024]](<https://arxiv.org/abs/2403.01804>) + DMAD: Dual Memory Bank for Real-World Anomaly Detection [[2024]](<https://arxiv.org/abs/2403.12362>)

## 2.1.5 Vision Language AD

- Random Word Data Augmentation with CLIP for Zero-Shot Anomaly Detection [BMVC 2023]
- AnomalyCLIP: Object-agnostic Prompt Learning for Zero-shot Anomaly Detection [ICLR 2024] [code]
- WinCLIP: Zero-/Few-Shot Anomaly Classification and Segmentation [CVPR 2023]
- ClipSAM: CLIP and SAM Collaboration for Zero-Shot Anomaly Segmentation [2023]
- CLIP-AD: A Language-Guided Staged Dual-Path Model for Zero-shot Anomaly Detection [2023]
- AnoVL: Adapting Vision-Language Models for Unified Zero-shot Anomaly Localization [2023] [code]
- AnomalyGPT: Detecting Industrial Anomalies using Large Vision-Language Models [AAAI 2024] [code][project page]
- Anomaly Detection by Adapting a pre-trained Vision Language Model [2024]

- Customizing Visual-Language Foundation Models for Multi-modal Anomaly Detection and Reasoning [\[2024\]](#)[\[code\]](#)

## 2.2 Reconstruction-Based Methods

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### 2.2.1 Autoencoder (AE)

- Improving unsupervised defect segmentation by applying structural similarity to autoencoders [\[2018\]](#)
- Automatic Fabric Defect Detection with a Multi-Scale Convolutional Denoising Autoencoder Network Model [\[Sensors 2018\]](#)
- An Unsupervised-Learning-Based Approach for Automated Defect Inspection on Textured Surfaces [\[TIM 2018\]](#)
- Unsupervised anomaly detection using style distillation [\[2020\]](#)
- Unsupervised two-stage anomaly detection [\[2021\]](#)
- Dfr: Deep feature reconstruction for unsupervised anomaly segmentation [\[Neurocomputing 2020\]](#)
- Unsupervised anomaly segmentation via multilevel image reconstruction and adaptive attention-level transition [\[2021\]](#)
- Encoding structure-texture relation with p-net for anomaly detection in retinal images [\[2020\]](#)
- Improved anomaly detection by training an autoencoder with skip connections on images corrupted with stain-shaped noise [\[2021\]](#)
- Unsupervised anomaly detection for surface defects with dual-siamese network [\[2022\]](#)
- Divide-and-assemble: Learning block-wise memory for unsupervised anomaly detection [\[ICCV 2021\]](#)
- Reconstruction from edge image combined with color and gradient difference for industrial surface anomaly detection [\[2022\]](#)[\[code\]](#)
- Spatial Contrastive Learning for Anomaly Detection and Localization [\[2022\]](#)
- Superpixel masking and inpainting for self-supervised anomaly detection [\[BMVC 2020\]](#)
- Iterative image inpainting with structural similarity mask for anomaly detection [\[2020\]](#)
- Self-Supervised Masking for Unsupervised Anomaly Detection and Localization [\[2022\]](#)
- Reconstruction by inpainting for visual anomaly detection [\[PR 2021\]](#)
- Draem-a discriminatively trained reconstruction embedding for surface anomaly detection [\[ICCV 2021\]](#)[\[code\]](#)
- DSR: A dual subspace re-projection network for surface anomaly detection [\[ECCV 2022\]](#)[\[code\]](#)

- Natural Synthetic Anomalies for Self-supervised Anomaly Detection and Localization [\[ECCV 2022\]](#) [\[code\]](#)
- Self-Supervised Training with Autoencoders for Visual Anomaly Detection [\[2022\]](#)
- Self-supervised predictive convolutional attentive block for anomaly detection [\[CVPR 2022 oral\]](#) [\[code\]](#)
- Self-Supervised Masked Convolutional Transformer Block for Anomaly Detection [\[TPAMI 2022\]](#) [\[code\]](#)
- Iterative energy-based projection on a normal data manifold for anomaly localization [\[2019\]](#)
- Towards visually explaining variational autoencoders [\[2020\]](#)
- Deep generative model using unregularized score for anomaly detection with heterogeneous complexity [\[2020\]](#)
- Anomaly localization by modeling perceptual features [\[2020\]](#)
- Image anomaly detection using normal data only by latent space resampling [\[2020\]](#)
- Noise-to-Norm Reconstruction for Industrial Anomaly Detection and Localization [\[2023\]](#)
- Patch-wise Auto-Encoder for Visual Anomaly Detection [\[2023\]](#)
- FAIR: Frequency-aware Image Restoration for Industrial Visual Anomaly Detection [\[2023\]](#)[\[code coming soon\]](#)
- Template-guided Hierarchical Feature Restoration for Anomaly Detection [\[ICCV 2023\]](#)
- FastRecon: Few-shot Industrial Anomaly Detection via Fast Feature Reconstruction [\[ICCV 2023\]](#) [\[code coming soon\]](#)
- Produce Once, Utilize Twice for Anomaly Detection [\[2023\]](#)
- RealNet: A Feature Selection Network with Realistic Synthetic Anomaly for Anomaly Detection [\[CVPR 2024\]](#)[\[code\]](#)

## 2.2.2 Generative Adversarial Networks (GANs)

- Omni-frequency Channel-selection Representations for Unsupervised Anomaly Detection [\[TIP 2023\]](#)[\[code\]](#)
- Learning semantic context from normal samples for unsupervised anomaly detection [\[AAAI 2021\]](#)
- Anoseg: Anomaly segmentation network using self-supervised learning [\[2021\]](#)
- A Surface Defect Detection Method Based on Positive Samples [\[PRICAI 2018\]](#)
- Few-shot defect image generation via defect-aware feature manipulation [\[AAAI 2023\]](#)[\[code\]](#)

## 2.2.3 Transformer

- VT-ADL: A vision transformer network for image anomaly detection and localization [\[ISIE 2021\]](#)

- ADTR: Anomaly Detection Transformer with Feature Reconstruction [\[2022\]](#)
- AnoViT: Unsupervised Anomaly Detection and Localization With Vision Transformer-Based Encoder-Decoder [\[2022\]](#)
- HaloAE: An HaloNet based Local Transformer Auto-Encoder for Anomaly Detection and Localization [\[2022\]](#)
- Inpainting transformer for anomaly detection [\[ICIAP 2022\]](#)
- Masked Swin Transformer Unet for Industrial Anomaly Detection [\[2022\]](#)
- Masked Transformer for image Anomaly Localization [\[TII 2022\]](#)
- Focus the Discrepancy: Intra- and Inter-Correlation Learning for Image Anomaly Detection [\[ICCV 2023\]](#)[\[code\]](#)
- AMI-Net: Adaptive Mask Inpainting Network for Industrial Anomaly Detection and Localization [\[TASE 2024\]](#)

## 2.2.4 Diffusion Model

- AnoDDPM: Anomaly Detection With Denoising Diffusion Probabilistic Models Using Simplex Noise [\[CVPR Workshop 2022\]](#)
- Unsupervised Visual Defect Detection with Score-Based Generative Model [\[2022\]](#)
- DiffusionAD: Denoising Diffusion for Anomaly Detection [\[2023\]](#)[\[code\]](#)
- Anomaly Detection with Conditioned Denoising Diffusion Models [\[2023\]](#)
- Unsupervised Surface Anomaly Detection with Diffusion Probabilistic Model [\[ICCV 2023\]](#)
- Removing Anomalies as Noises for Industrial Defect Localization [\[ICCV 2023\]](#)
- TransFusion -- A Transparency-Based Diffusion Model for Anomaly Detection [\[2023\]](#)
- LafitE: Latent Diffusion Model with Feature Editing for Unsupervised Multi-class Anomaly Detection [\[2023\]](#)
- DiAD: A Diffusion-based Framework for Multi-class Anomaly Detection [\[AAAI 2024\]](#)[\[code\]](#)
- D3AD: Dynamic Denoising Diffusion Probabilistic Model for Anomaly Detection [\[2024\]](#)

## 2.2.5 Others

- Anomaly Detection using Score-based Perturbation Resilience [\[ICCV 2023\]](#)