

# Improved Domain Generalization via Disentangled Multi-task Learning in Unsupervised Anomalous Sound Detection

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Cambridge, Massachusetts, USA

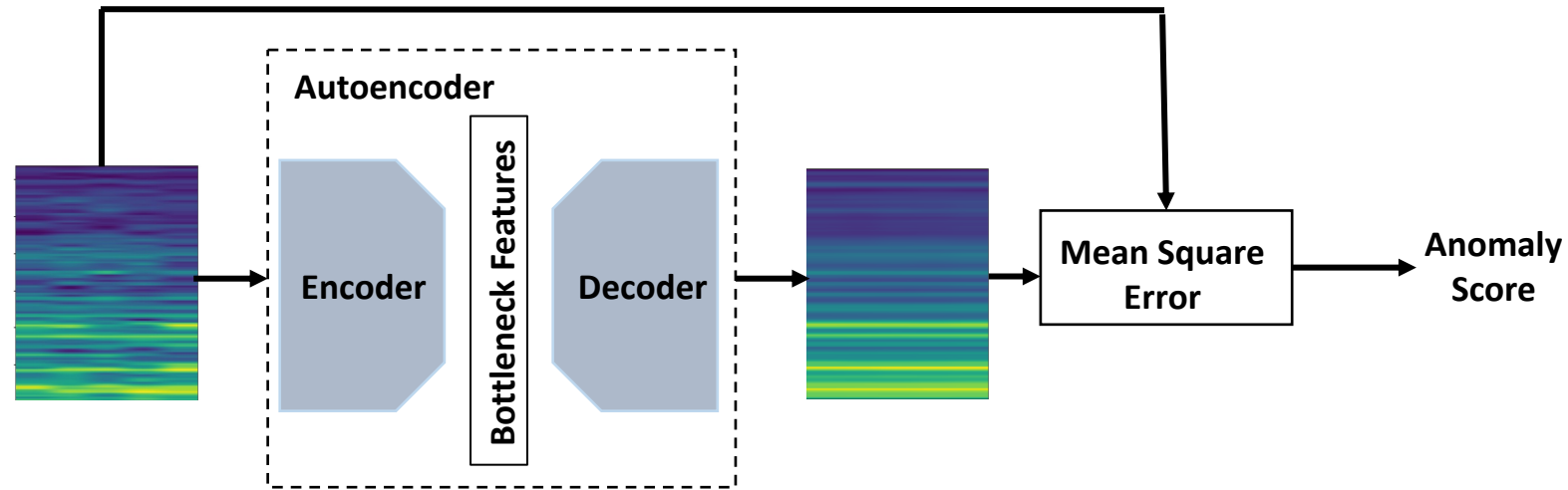
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## DCASE Challenge 2022 Task 2

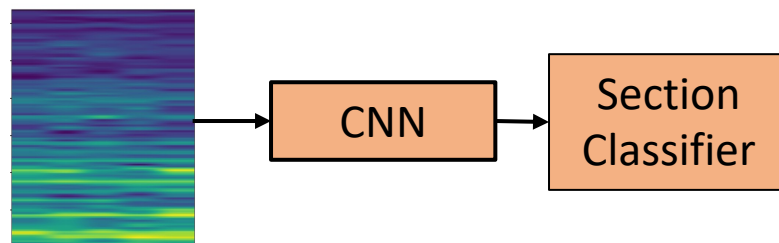
- Task title: Unsupervised Anomalous Sound Detection for Machine Condition Monitoring Applying Domain Generalization Techniques
- Train only on normal operation data (Unsupervised)
- Domain generalization:
  - Machine's physical parameters
  - Environmental conditions
  - Maintenance
  - Recording method
- Motivation: AI-based factory automation
- Our system ranked 5<sup>th</sup> out of 32 teams in the challenge

# Baseline Approaches for Unsupervised Anomalous Sound Detection

- Autoencoder

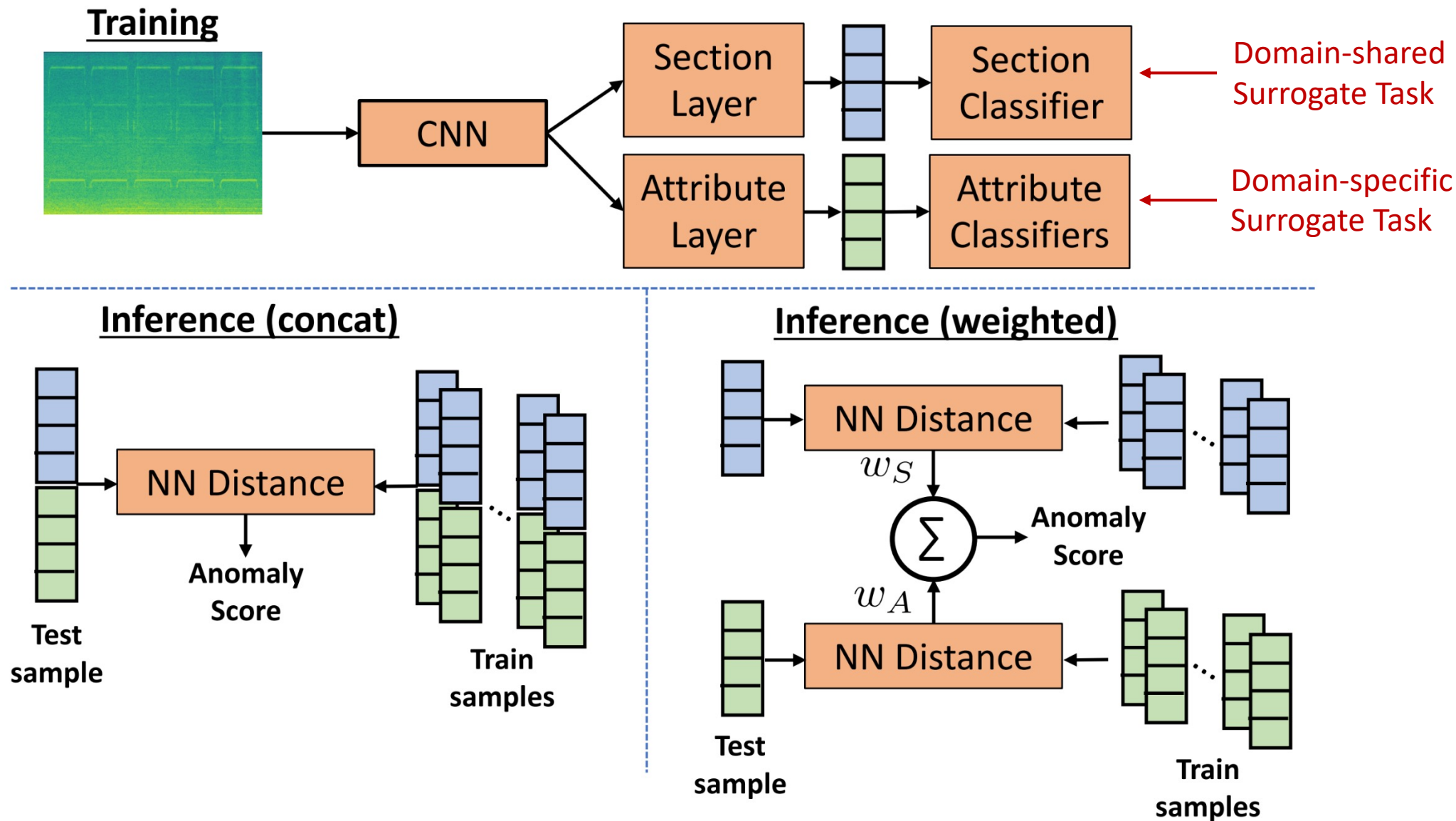


- Surrogate task classifier



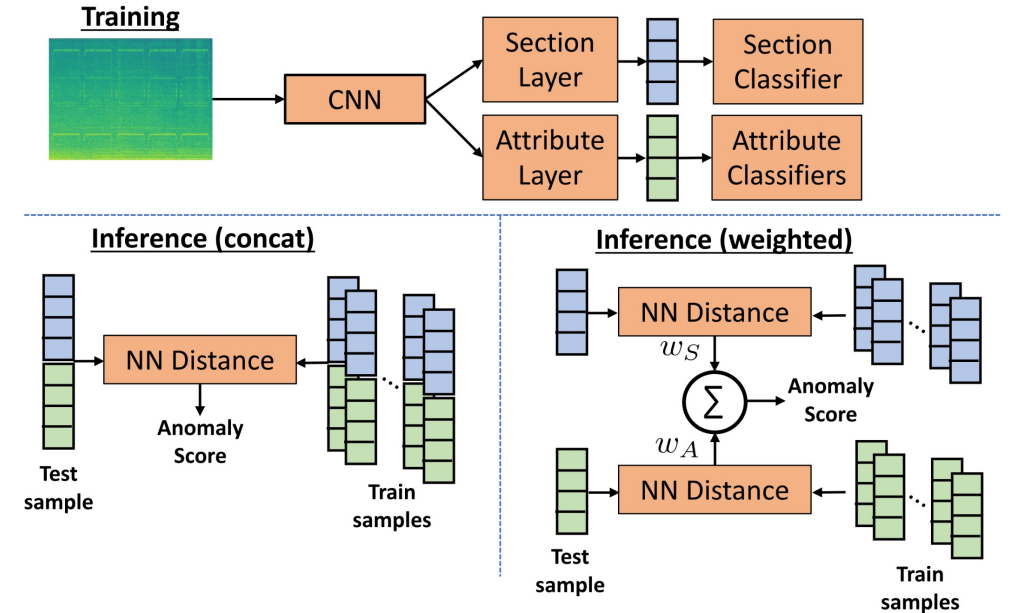
Post-processing by using confidence scores [Dohi et al., 2021] or Nearest Neighbor [Morita et al., 2021]

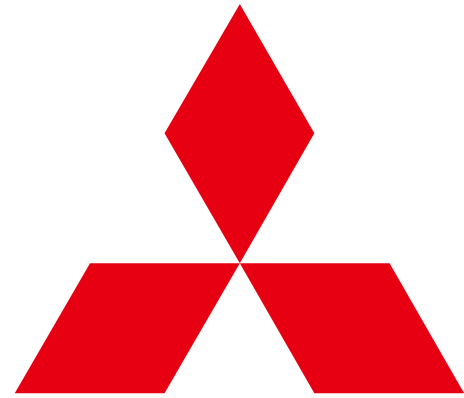
# Disentangled Anomaly Detector



# Benefits of Disentangled Embeddings

- Outperform “entangled” multi-task learning
- We can weight embedding dimensions for computing anomaly scores
  - Functions like a single system ensemble
- Explainability – embedding elements with high anomaly scores may indicate cause
  - Example: the embeddings trained to predict operating speed are far from normal training examples, so the operating speed is likely cause of anomaly





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