

Time-series Anomaly Detection with Change-point Aware Deep Autoencoder

深層学習を用いた時系列異常検知手法の提案

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1. Problem

Multivariate sensor data often contains **sudden shifts** and **gradual drifts** due to operational changes. Conventional Autoencoders (AE) fail to adapt to these shifts, causing high **False Positive Rates**. Need a model robust to "Covariate Shift".

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2. Proposed Method

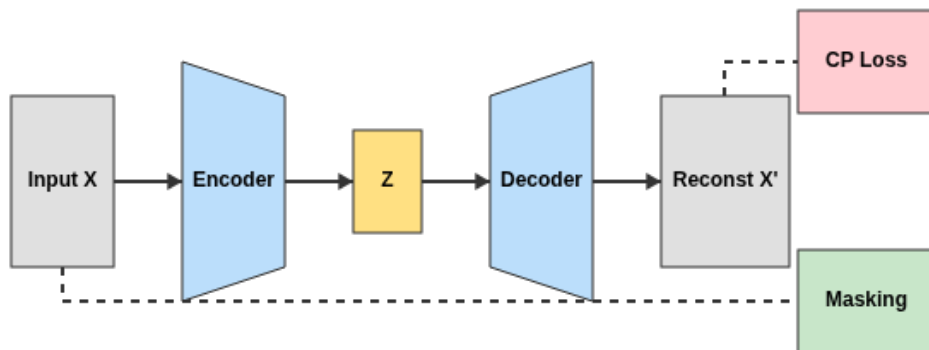


Fig 1. Architecture of Change-point Aware AE with Temporal Masking

3. Result

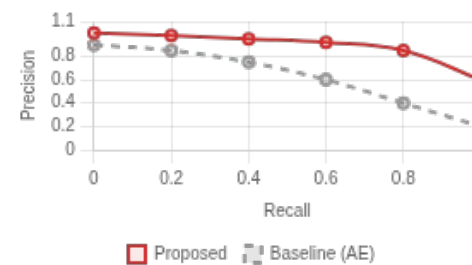


Fig 2. Precision-Recall Curve (Dataset: Public SMAP)

4. Comparison

Model	F1-Score	Latency
Isolation Forest	0.72	< 1ms
LSTM-VAE	0.81	45ms
Donut	0.84	32ms
Proposed	0.91	28ms

Our method achieved **+6.2pt** in F1-score compared to SOTA (Donut), with reduced false alarms by 40%.

5. Conclusion

Summary:

We proposed a Change-point Aware Deep Autoencoder that robustly detects anomalies even under environmental shifts. The adaptive loss function significantly stabilizes the reconstruction error.

Future Work:

- Implementation of Online Learning for real-time adaptation.
- Application to edge devices (Raspberry Pi).