

TS-Rep: Self-supervised time series representation learning for robot sensor data



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Highlights

TS-Rep learns a rich representation from the multimodal time series sensor data in robotics.

TS-Rep

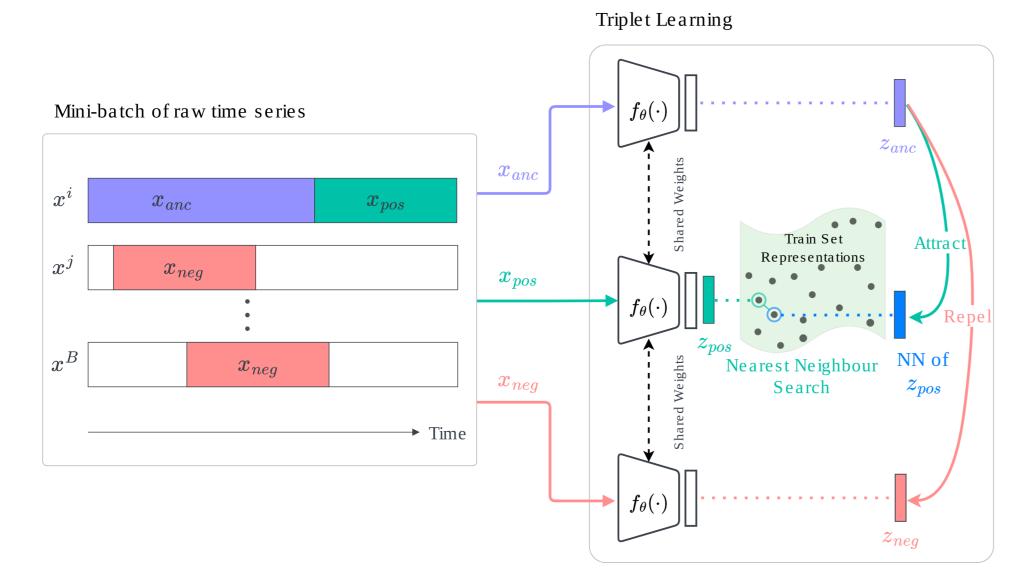
- learned representations have instances from the same class grouped together i.e., clusterability
- outperforms or performs closed to existing methods on anomaly detection and classification tasks
- on terrain classification outperforms unsupervised methods and performs closed to the supervised appraoches.
- is on average fastest method to train and ~8 times faster than T-Loss, on which our approach is based on.

Methodology

We present TS-Rep, a representation learning method that uses,

- Causal Dilated CNN Encoder
- Triplet loss
- Time-based negative sampling and,
- Positive sampling using nearest neighbour

Method Overview



TS-Rep overview: Anchor (x_{anc}) , Positive (x_{pos}) and Negatives (x_{neg}) sampling (left). Triplet loss brings z_{anc} closer to the nearest neighbour of z_{pos} while pushing apart z_{neg} (right).

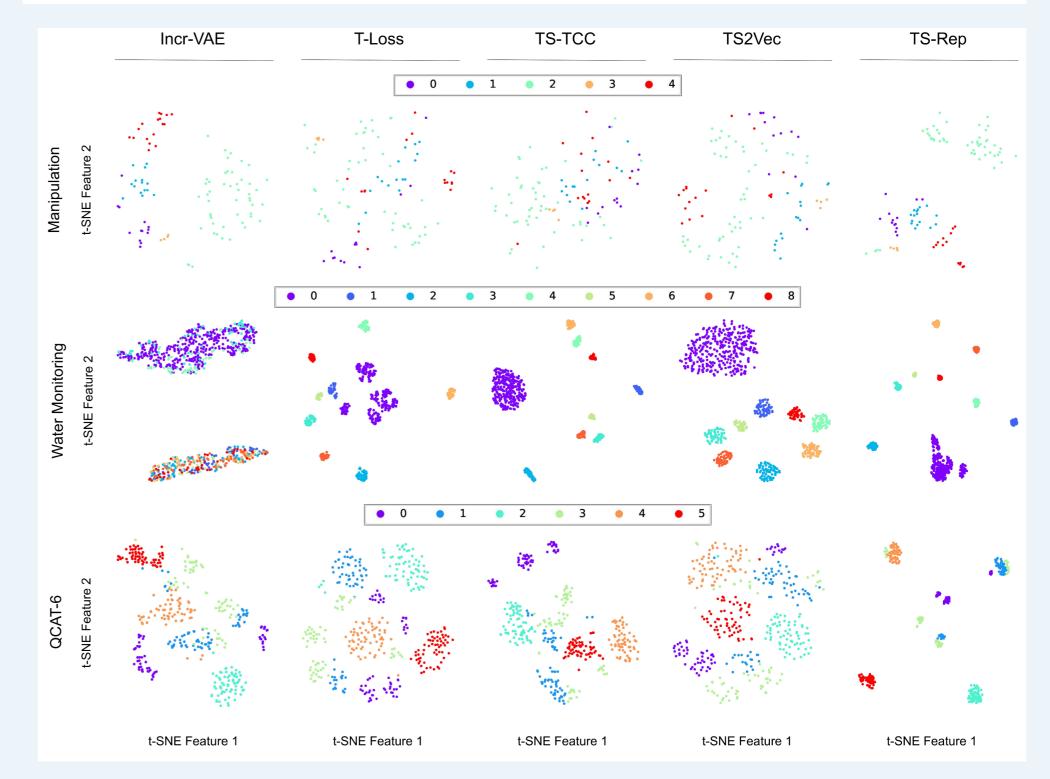
Triplet Loss

$$\mathcal{L}_{triplet} = \frac{1}{B} \sum_{i=1}^{B} \left[-\log \left(\sigma \left((z_{anc}^{i})^{T} NN(z_{pos}^{i}, S) \right) \right) - \sum_{j=1; j \neq i}^{B} \log \left(\sigma \left(-(z_{anc}^{i})^{T} (z_{neg}^{i}) \right) \right) \right]$$

Experiments

Clusterability

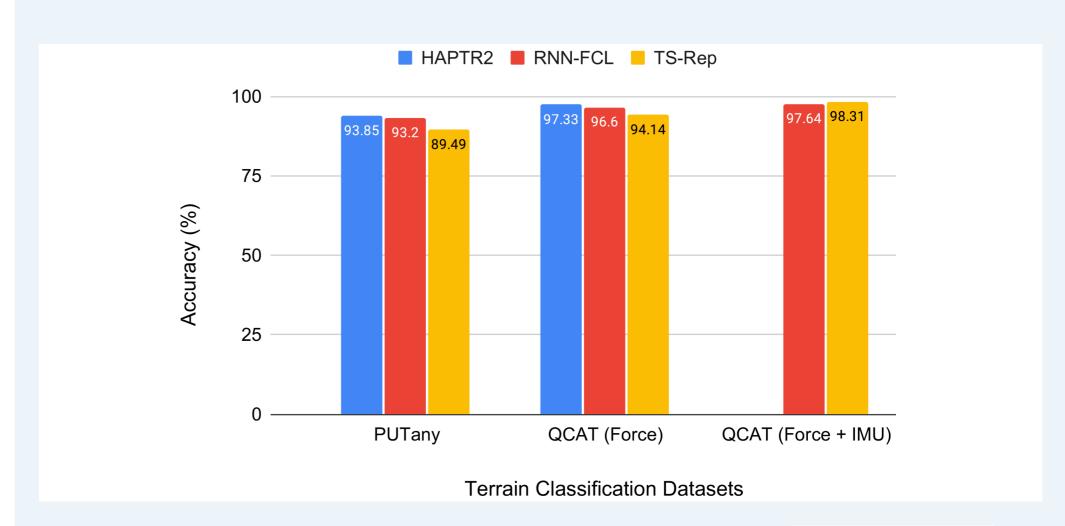
| | Manipulation | | | Water Monitoring Robot | | | QCAT-6 | | |
|----------|--------------|--------------|-------|------------------------|--------------|------|--------|--------------|-------|
| | NMI ↑ | Silhouette ↑ | DBI ↓ | NMI ↑ | Silhouette ↑ | DBI↓ | NMI ↑ | Silhouette ↑ | DBI ↓ |
| Incr-VAE | 0.70 | 0.26 | 1.39 | 0.19 | 0.22 | 1.44 | 0.65 | 0.16 | 2.14 |
| T-Loss | 0.45 | 0.11 | 2.34 | 0.84 | 0.25 | 1.72 | 0.72 | 0.11 | 2.75 |
| TS-TCC | 0.20 | 0.17 | 1.70 | 0.97 | 0.67 | 0.83 | 0.65 | 0.15 | 2.01 |
| TS2Vec | 0.34 | 0.11 | 2.14 | 0.80 | 0.11 | 2.27 | 0.48 | 0.07 | 2.75 |
| TS-Rep | 0.65 | 0.31 | 1.26 | 1.00 | 0.70 | 0.50 | 0.80 | 0.41 | 1.10 |



Anomaly Detection

| | Manipulation | | | Water Monitoring Robot | | | QCAT-6 | | |
|----------|--------------|---------|--|------------------------|---------|--|--------|---------|--|
| | F1 ↑ | AUROC ↑ | $_{95\%\text{-TPR}}^{\text{FPR-}}\downarrow$ | F1 ↑ | AUROC ↑ | $^{\mathrm{FPR-}}_{95\%\text{-TPR}}\downarrow$ | F1 ↑ | AUROC ↑ | $^{\mathrm{FPR-}}_{95\%\text{-TPR}}\downarrow$ |
| Incr-VAE | 0.83 | 0.99 | 0.02 | 0.33 | 0.80 | 0.56 | 0.99 | 1.00 | 0.01 |
| T-Loss | 0.77 | 0.90 | 0.37 | 1.00 | 1.00 | 0.00 | 1.00 | 1.00 | 0.00 |
| TS-TCC | 0.39 | 0.58 | 0.69 | 1.00 | 1.00 | 0.00 | 0.92 | 0.97 | 0.10 |
| TS2Vec | 0.76 | 0.75 | 0.50 | 0.99 | 1.00 | 0.00 | 0.96 | 0.98 | 0.09 |
| TS-Rep | 0.81 | 0.85 | 0.33 | 1.00 | 1.00 | 0.00 | 0.98 | 1.00 | 0.01 |

Terrain Classification



For more information please visit our project page at

project page at pratiksomaiya.in/projects/TS-Rep Contact: imprs@pratiksomaiya.in

