

# Semi-Supervised Siamese Network for Identifying Bad Data in Medical Imaging Datasets

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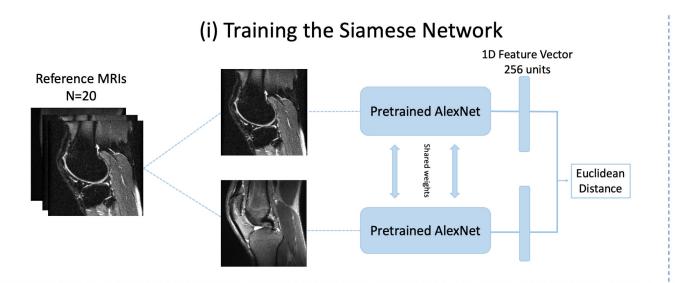


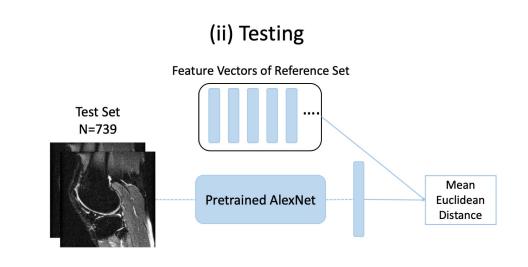




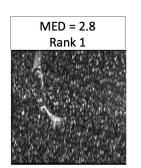
## **Proposed Method**

**Objective:** Develop a pre-processing technique to identify bad data that could harm the model's training performance in future analysis.

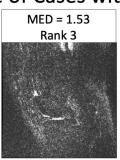


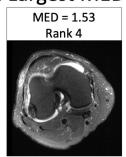


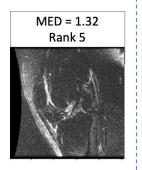
#### (iii) Mid-Slice of Cases with Largest MED











#### (iv) Additional Bad Data Examples

Example A
Class => Bad data
Siamese network, MED =>
0.98
Siamese network,
classification => Bad data
Isolation Forest,
classification => Bad data



Example B
Class => Bad data
Siamese network, MED => 1.2
Siamese network,
classification => Bad data
Isolation Forest, classification
=> Not bad data



### Model Performance

 Threshold chosen based on the largest Euclidean Distance between reference MRIs.

	AUC	Sensitivity	Specificity
Siamese Network (proposed)	0.989	100%	89%
Isolation Forest	0.802	71%	92%

### **Advantages**

- Achieves good performance.
- Identifies a wide variety of bad data.
- Requires only a fraction of the training data that previous methods require.
- Less tedious labelling process in comparison to other semi-supervised techniques.







