

Aligning Chandra X-ray spectra and scientific literature

Nicolò Pincioli



POLITECNICO
MILANO 1863

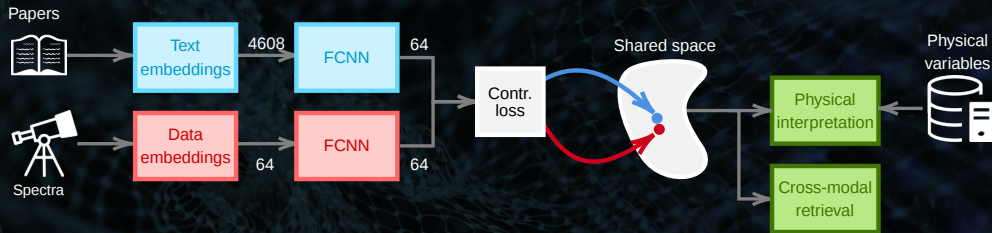
Challenges

- ▶ Rare source classes lack sufficient **labeled data** $\implies$ unsupervised and self-supervised approaches

Challenges

- ▶ Rare source classes lack sufficient **labeled data** $\implies$ unsupervised and self-supervised approaches
- ▶ **Scientific literature** describes astrophysical sources in natural language $\implies$ automated cross-modal alignment between text and observations

Enhancing X-ray spectra with contrastive learning

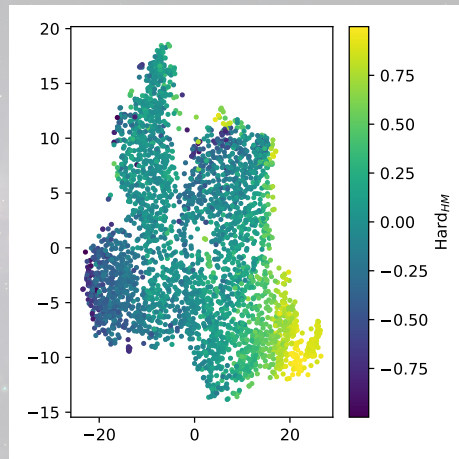


- ▶ Aligns text summaries (from literature) with X-ray spectra using InfoNCE loss
- ▶ 11,447 text-spectra pairs (1,719 test set pairs)
- ▶ The goal is to obtain a shared representational space that optimizes cross-modal retrieval

Compressing X-ray spectra with representation learning

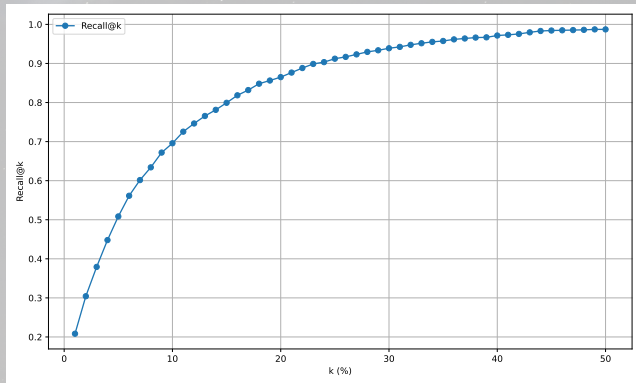
Physical explainability via regression

- ▶ Symbolic regression directly correlates embedding vectors with combinations of measurable fluxes
- ▶ Latent space accurately predicts critical Hardness Ratio parameters
- ▶ Confirms that autoencoder features are correlated with physical quantities, not noise



Enhancing X-ray spectra with contrastive learning

Contrastive retrieval

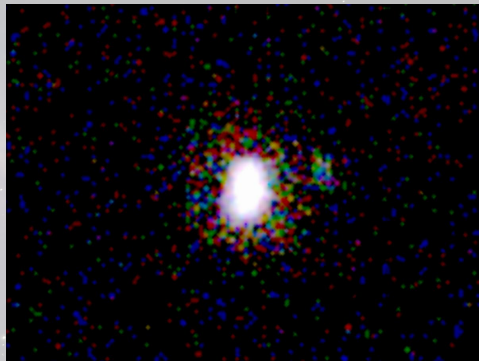


- ▶ 20% Recall@1% when pre-initialized with autoencoder data (90% Recall@23%)
- ▶ Correct textual descriptions retrieved from thousands of scientific abstracts
- ▶ Scientific texts enhance spectral data (97% reduction in latent space size)

Enhancing X-ray spectra with contrastive learning

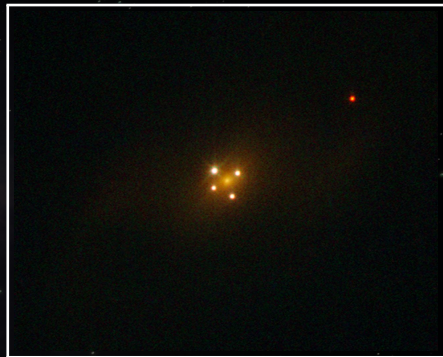
Outlier detection

- ▶ Isolation Forest on joint text/spectra embeddings identifies outliers
- ▶ Successfully flagged a known Pulsating Ultraluminous X-ray candidate
- ▶ Provides a scalable mechanism for discovering novel phenomena in petabyte surveys



Strengths

- ▶ Representation learning can transform raw data into compressed catalogs encoding physical information
- ▶ Using contrastive learning results in 97% dimensionality reduction, ensuring computational feasibility on large datasets



Quadruply-lensed quasar (ESA/Hubble)

Limitations

- ▶ LLM-generated summaries not available for rare sources $\implies$ contrastive learning may provide descriptions of similar objects
- ▶ Contrastive learning limited to two modalities $\implies$ apply to additional modalities (e.g., images)
- ▶ Lack of systematic outliers analysis $\implies$ measuring deviations from *average* sources for each class



Cygnus A (Chandra)

Thank you for your attention

AIMLSystems 2026

Explainable AI for Astrophysics

9 OCTOBER
LECCO, LAKE COMO, ITALY

Registration deadline 6 September (early bird, 80€)
to assist (speakers already invited)

