# **Autoencoder Latent Representations and One-Dimensional Convolutional**

**Neural Networks for the Identification of Composite Radioisotopes** 

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Kazem Abdolpour, S. Farhad Masoudi and Atefeh Fathi

Department of Physics, K.N. Toosi University of Technology, P.O. Box 15875-4416, Tehran, 15418-49611, Iran

#### **Abstract:**

Reliable isotope identification from short, low-count  $\gamma$ -ray spectra is critical for security and environmental monitoring. However field spectra suffer from background and instrumental effects. We present two models for the classification of mixed isotopes: (1) Autoencoder and (ii) a one-dimensional convolutional neural network (1D-CNN) that learns spectral features directly from the 1024-channel input spectrum. Training uses simulated NaI(Tl) spectra spanning source-detector distances and counting times. Robustness is tested on held-out sets with mismatched conditions, including calibration shifts unseen during training. Results show both approaches achieve high accuracy on matched conditions, but the autoencoder-based model provides more stable evidence and better generalization under varying calibration, outperforming the 1D-CNN in disturbed settings.

#### Introduction

Accurate and rapid identification of radioisotopes from γ-ray spectra is essential for security, environmental monitoring, and emergency response [1]. Field spectra are low-count, and distorted by shielding, calibration drift, and gain variations, which severely degrade the performance of traditional peak identification and energy-library matching methods [1]. Data-driven models that learn from full spectra can exhibit more stable performance against overlapping peaks and calibration/gain drift [2].Deep learning approaches such as CNNs using gain/calibration augmentation and channel-attention mechanisms improve accuracy, but because the models are large, they require very substantial training data, making these methods complex to train [3,4].

### Methodology

Training and testing data are synthesized with simulation using a 2"×2" NaI(Tl) detector. We consider six isotopes (Co-60, Cs-137, I-131, Ba-133, Am-241, Tc-99m), eleven source detector distances (1-100 cm), and six acquisition times (30-180 s), producing mixed-nuclide and low-count conditions.

An example of simulated spectra with high and low counts is shown in Figure 1.

### **Results and Discussion**

Across increasing isotope counts and under peak broadening, the autoencoder sustains a more favorable precision-recall trade-off than the smoothed 1D-CNN, yielding higher F1 at greater mixture complexity. These results indicate that the autoencoder preserves discriminative structure that remains robust as peaks widen and

In Fig. 2, the smoothed 1D-CNN's recall degrades rapidly as additional isotopes are introduced—particularly within the ±10-20% window—leading to an early drop in F1. By contrast, the autoencoder's curves are flatter: precision increases with isotope count while recall declines more gradually. In Fig. 3, peak broadening harms both models, yet the smoothed 1D-CNN again shows a steeper decline in recall and F1 as mixtures grow more complex. The autoencoder maintains a more balanced precisionrecall.

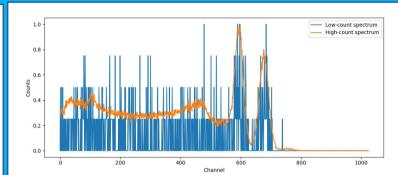


Figure 1 shows the simulated spectra with high and low counts.

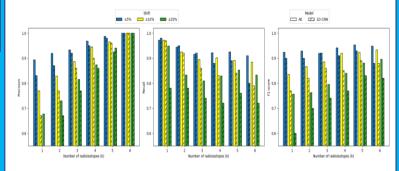


Figure 2. Performance under uniform calibration shifts

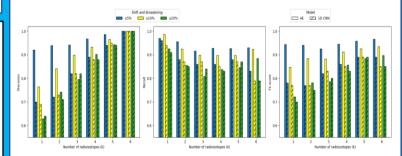


Figure 3. Performance under energy-resolution degradation.

## Conclusions

Based on the results, the autoencoder model learns meaningful structures and patterns which, under various conditions, deliver higher accuracy and stability than a one-dimensional convolutional neural network (1D-CNN). This efficiency and performance make it practical for field deployment in nuclear security, environmental monitoring, nonproliferation verification, and emergency response.

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