



Radioisotope Identification with List-Mode Gamma Ray Data

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Introduction

- List-mode data is comprised of energy-time pairs.
- Full gamma ray spectral analysis is the industry standard.
- Current analysis ignores the time domain, has long dwell times, and struggles with sources containing similarly-spaced spectral peaks.

Research Question:

Can the time domain be leveraged to aid in the identification of radioisotopes?

Datasets

- 3 datasets were created, all with a stationary detector and stationary sources
- Events within the first 100s to 1200s interval were used

Descriptions:

July 2022:

- Sources: Cs-137, Y-88, Ba-133, U-232, Ba-133+U-232

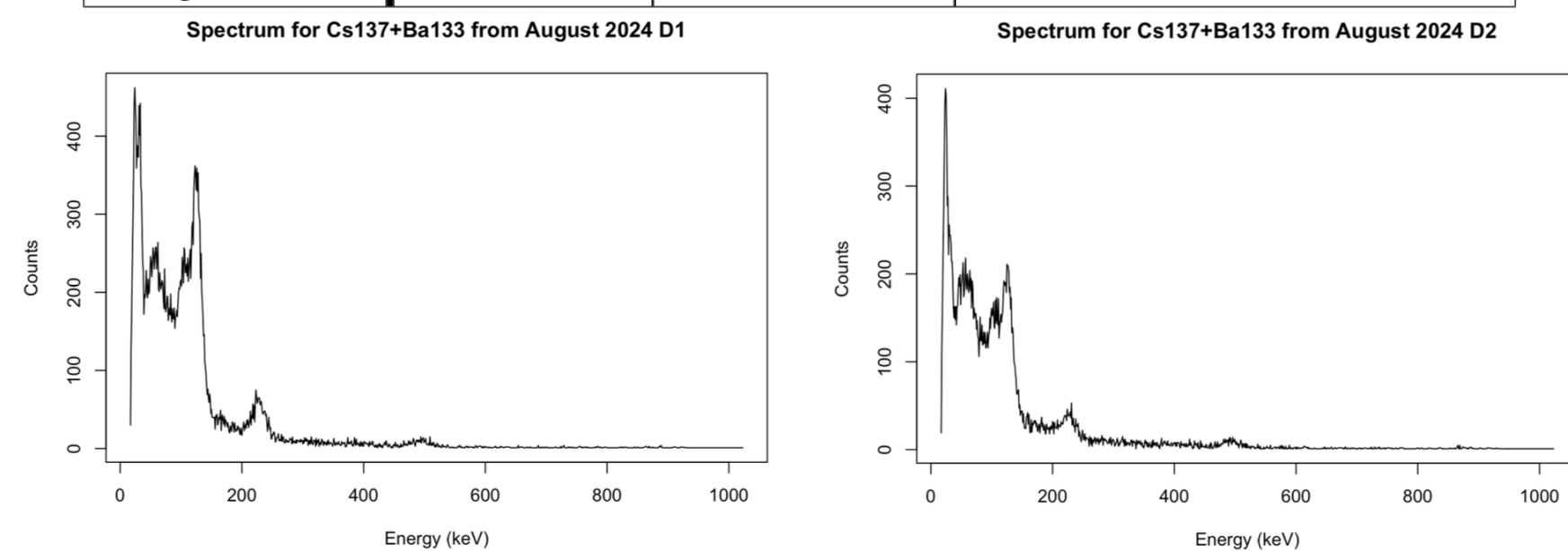
April 2024:

- Sources: Ba-133, U-232, Ba-133+U-232

August 2024:

- Sources: Cs-137, Y-88, Ba-133, Cs-137+Ba-133
- The gross count rate is roughly the same across sources.
- Two sets of distances were used

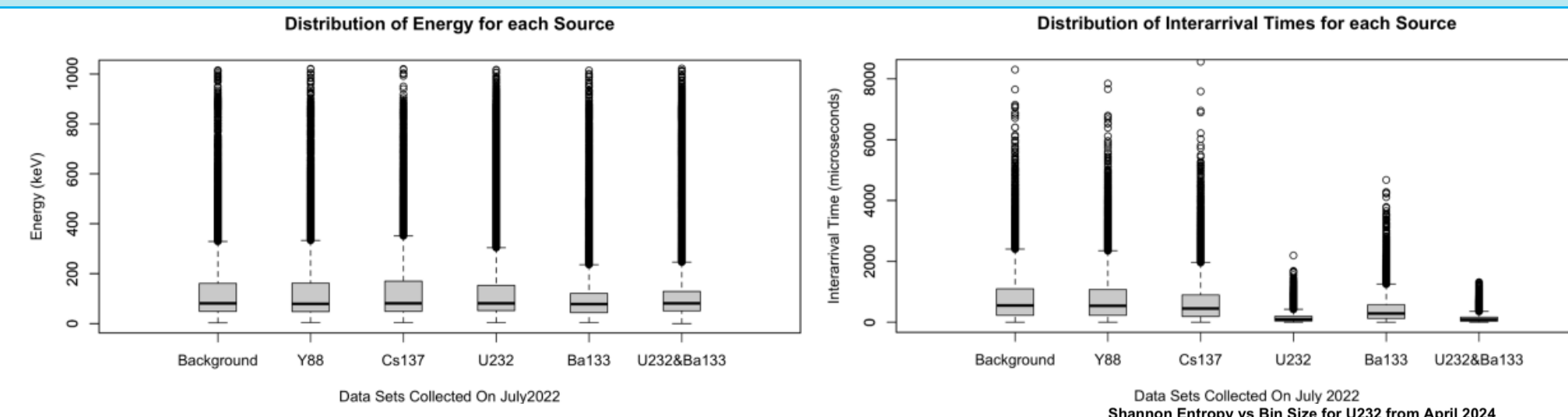
Source	Distance 1 (D1)		Distance 2 (D2)	
	Distance	Gross Count Rate	Distance	Gross Count Rate
Cs-137	10ft 2in	1598	15ft 2.5in	1181
Y-88	7ft 1.5in	1738	11ft 4in	1230
Ba-133	21ft 6in	1827	29ft	1381
Cs-137 & Ba-133	22ft 9in	1849	31ft	1350
Background		751		



Exploratory Data Analysis

Correlation metrics:

- Pearson Correlation: X
- Spearman's rho: X
- Autocorrelation: X
- Cross-Correlation: X
- X = no correlation found

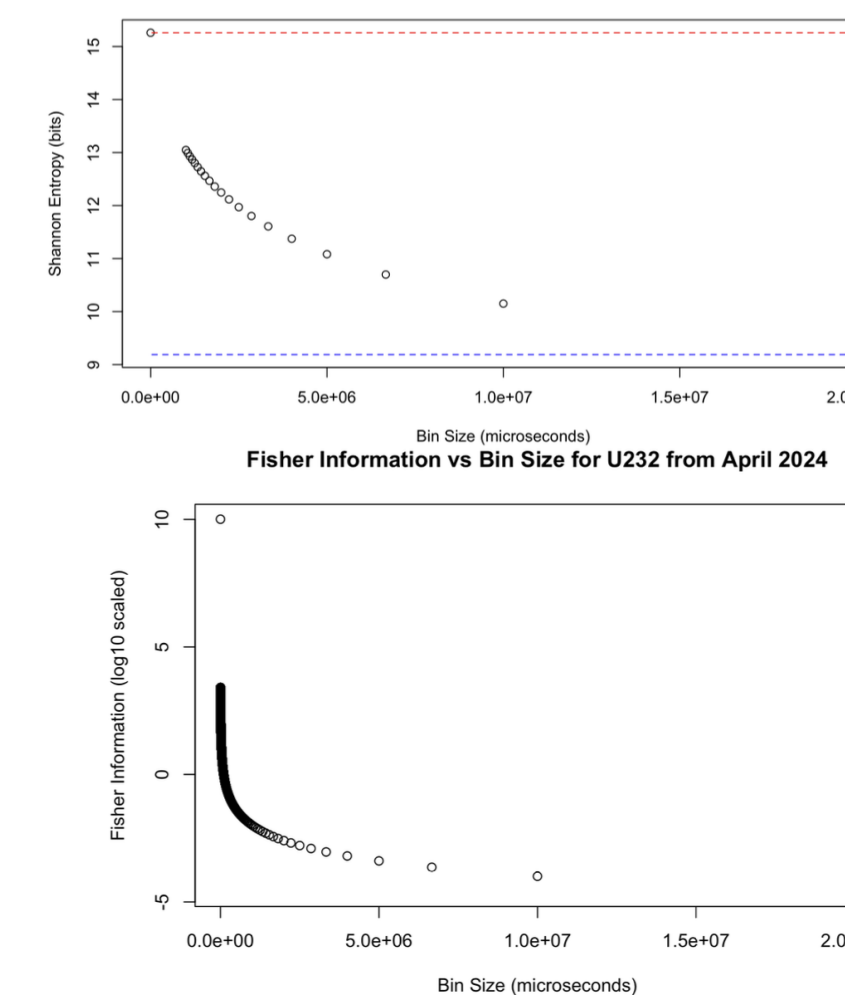


Box & Whisker Plots:

- Energy: Differences across sources
- Interarrival time: Differences across sources

Information Metrics:

- Shannon Entropy: Decreases with increasing bin size
- Fisher Information: Decreases with increasing bin size



Probabilistic Classifier

First Model:

- Assume: Radioactive decay is a Poisson process
- Assume: K isotopes + 1 background
- Assume: unique rates of decay per isotope so that $\lambda_0 \neq \dots \neq \lambda_k$
- Assume: independence between time and energy
- Assume: Mixture distribution of energies
- Event counts = $\sum_{i=0}^K N_t^i \sim \text{Poisson}(\sum_{i=0}^K \lambda_i t)$
- Interarrival times $\sim \text{Exp}(\sum_{k=0}^K \lambda_k)$
- Energy $\rightarrow p_E(e) = \sum_{i=0}^K \pi_i p_{E_i}(e)$, $p_{E_0}(e) = \sum_{b=1}^B p_b \frac{e^{-\alpha_b} \alpha_b^e}{e!}$
- Histogram density approximator for $p_E(e)$ is $p_{E_i}^*$
- Estimated probability density of source i : $\hat{p}_{E_i}(e) = \frac{p_{E_i}^*(e)}{\sum_{e \in E} p_{E_i}^*(e)}$
- Mixing proportions: $\hat{\pi} = \underset{\pi}{\text{argmin}} (\hat{p}_{E|K=k}(e) - \sum_{i=0}^K \pi_i \hat{p}_{E_i}(e))^2$
- Recursive MLEs of λ_k
- Classification based on max log-likelihood of new sequence

Extended α Model:

- Designed to handle varying background effects ("SNR")
- Maximum a posteriori estimate of α where $\lambda'_i = \alpha \lambda_i$

Classifier Analysis

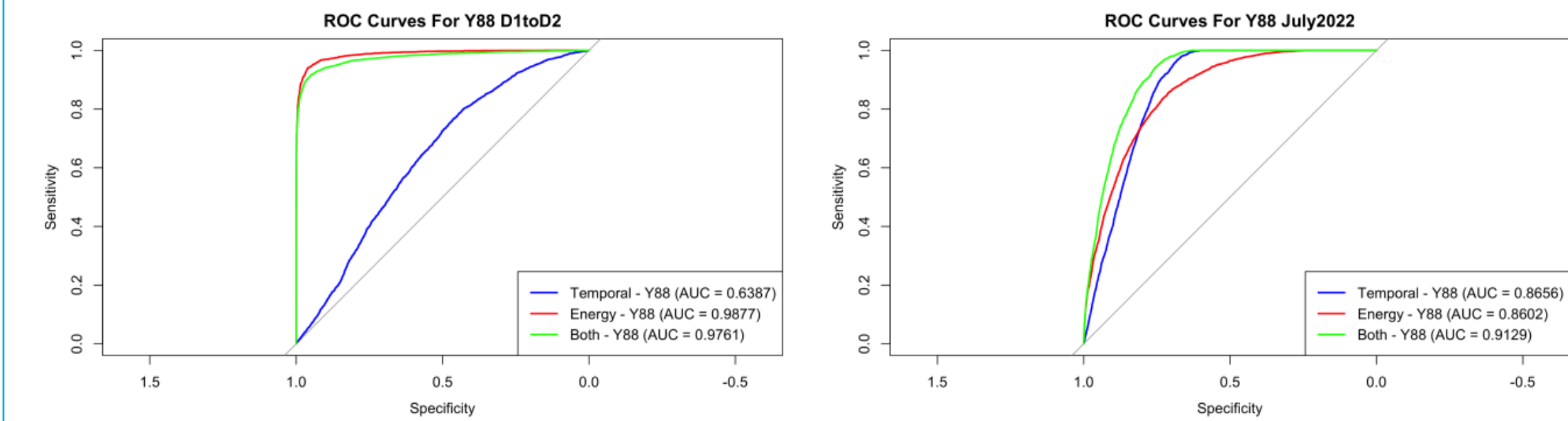
Accuracy:

- Better together: Energy & Time Models
- Accounting for "SNR" is beneficial

Experiment:	Accuracy Results		
	Time Model	Energy Model	Energy & Time Model
1: July 2022	0.7107	0.5498	0.6565
1- α : July 2022	0.7018	0.5498	0.6200
2: April 2024	0.8774	0.3890	0.6898
3: D1 to D1	0.4850	0.7368	0.8249
3- α : D1 to D1	0.3334	0.5436	0.8479
4: D1 to D2	0.1787	0.3864	0.5307
4- α : D1 to D2	0.3333	0.4782	0.7504
5: D2 to D1	0.2509	0.6636	0.7404
5- α : D2 to D1	0.3333	0.8799	0.9936
6: D2 to D2	0.5101	0.5476	0.7451
6- α : D2 to D2	0.3333	0.8372	0.9699
7: Mix to Mix	0.5037	0.6249	0.7488
7- α : Mix to Mix	0.3333	0.6442	0.9201

ROC Curves:

- Comparison shows problem with 1st Model
- Mostly better together



Multi-class AUC:

- Similar to ROC curves analysis

Experiment:	Multi-class AUC Results		
	Time Model	Energy Model	Energy & Time Model
1: July 2022	0.9454	0.9105	0.9644
2: April 2024	0.9742	0.4845	0.9774
3: D1 to D1	0.7905	0.9614	0.9772
4: D1 to D2	0.5676	0.9171	0.9077
5: D2 to D1	0.5934	0.9167	0.9146
6: D2 to D2	0.8197	0.9313	0.9599
7: Mix to Mix	0.8049	0.9462	0.9662

McNemar's Test:

- P-value analysis suggests that Energy & Time Models are better than Energy Models

Conclusion

- Linear relationships not found
- May be meaningful differences across the distributions of interarrival times across radiological sources.
- Created a new probabilistic classifier that allowed for the comparison of models that used temporal information, energy information, or both combined
- Using Energy & Time information leads to better performing models
- Further investigations are warranted