# Gamma-Ray Spectrum Analysis and Classification 

A Machine Learning Perspective

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## Outline

- Overview
- Data
- Experiments
- Results
- Conclusion / Future Work


## Background

- HC Radiation Protection Bureau
- Radioactivity monitoring network
- Support CTBT
- Secure public events against possible radioactive threats, etc.


## HC Problem Statement

- Goal: monitor and detect radioactive threats
- Political gatherings, sporting events, etc.
- Strategy: Utilize human experts and ML algorithms
- Minimize human involvement


## HC Objective

"Compile a multi-categorical dataset of gamma-ray spectra for use in the development and testing of ML algorithms"


## An ML Perspective

- Few "real" isotopes measured, thus...
- How to build the model
- Multi-class, one-class, static, dynamic, etc.
- How to evaluate the model
- What metrics to use
- How certain can we be of the model


## NARNIA Data

- Collection:
- 18 GRS detectors at Vancouver 2010
- 1,024 int-valued channels
- One reading / station / minute - $\mid$ DS $\mid=43,000 \times 18$
- Bkgrnd + I, Tc, Th, Cs, Co


## NARNIA Data

## Bkgrnd <br> Rain Vs No-Rain



TC
Rain Vs Bkgrnd


## NARNIA Data

PCA Rain Vs No-Rain


## NARNIA Data

Naive Bayes Classification Results


## ML System

Hyp: Readings that are most dissimilar are of significant interest


## Data Processing



## An ML Perspective

- Rain separation: ML to identify heavy rain events
- Binary Vs OC learning
- Extra info Vs labelling
- Static Vs temporal classification
- Simplicity Vs leveraged context
- ocSVM, SVM, J48, NB, IBK


## Experiments

## Binary rain separation



## ML System

Hyp: Readings that are most dissimilar are of significant interest


## Experiments

- Algorithms applied
- Binary: SVM, J48, NB, MLP, IBK
- Performed well, but inappropriate
- One class: ocSVM, AA
- Pour results, slow training


## Experiments

Alternate Algorithms: based on the hypothesis of ranking and anomalies

Mahalonibis<br>Distance



## Experiments

## Alternate Algorithms... continued

Variance in Angle Spectrum

## Hypothesis

## Anomalous instances near the top



Progression of Mahalanobis Distance across ranks


## Results

| Isotope | Station 13 | Station 12 | Station 6 |
| :---: | :---: | :---: | :---: |
| Thallium | NA | NA | $0 / 2$ |
| Iodine | $1 / 5$ | $1 / 2$ | $1 / 2$ |
| Technicium | $1 / 3$ | $0 / 2$ | $1 / 7$ |
| Caesium | $8 / 15$ | NA | NA |
| Cobalt 50.1 | $19 / 304$ | $3 / 304$ | $0 / 304$ |
| Cobalt 50.2 | $19 / 314$ | $3 / 314$ | $0 / 314$ |
| Cobalt 75.1 | $2 / 307$ | $0 / 307$ | $0 / 307$ |
| Cobalt 75.2 | $1 / 320$ | $0 / 320$ | $0 / 320$ |
| Cobalt 100.1 | $0 / 305$ | $0 / 305$ | $0 / 305$ |
| Cobalt 100.2 | 0 | 0 | 0 |
| Cobalt 200.1 | 0 | 0 | 0 |
|  | Cohnlt 2002 | 0 | 0 |
| Cobalt 300.1 | 0 | 0 | 0 |
|  | Cobalt 300.1 | 0 | 0 |
|  | Cobalt 500.1 | 0 | 0 |
| Cobalt 500.2 | 0 | 0 | 0 |

## Results

1-Phase Vs 2-Phase System


## Conclusion

- Proposed a 2-phased ML system
- I - separate rain from non-rain
- II - rank instance according to rarity
- Significantly outperforms HC's system
( HC - no Co less than 200 strength


## Future Work

- Aim: minimize expert involvement
- i.e., reduce FPs
- Improved rain separation may hold key
- OC learning on positive class
- Expectation Maximization
- Semi-supervised
- Temporal analysis


## Future Work

## Other open questions...

- Threshold selection
- Where do we draw the line between normal and abnormal
- Individual isotope classification
- Classification of instances belonging to multiple classes
- Temporal analysis


## Thank you!

## Gamma-Ray Spectrum Analysis

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