

Pre-processing for Anomaly Detection on Linear Accelerator

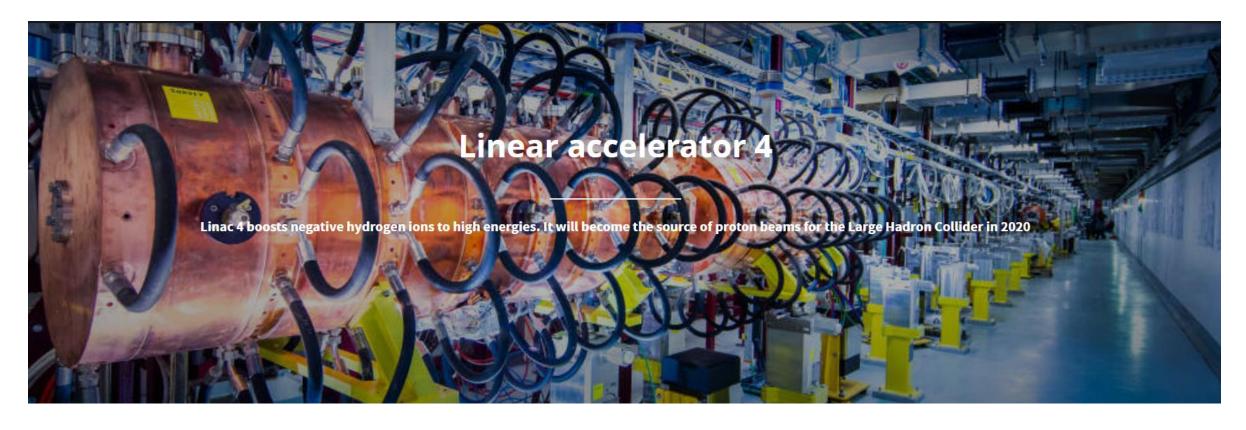
Martin Molan

Dr. Yann Donon

Dr. Alberto Di Meglio

Introduction

Anomaly detection for linear accelerators (LINACs, LINAC4 H- source for LHC)





Approaches to anomaly detection

Semi-supervised approaches:

Anomalies are rare events

Autoencoders

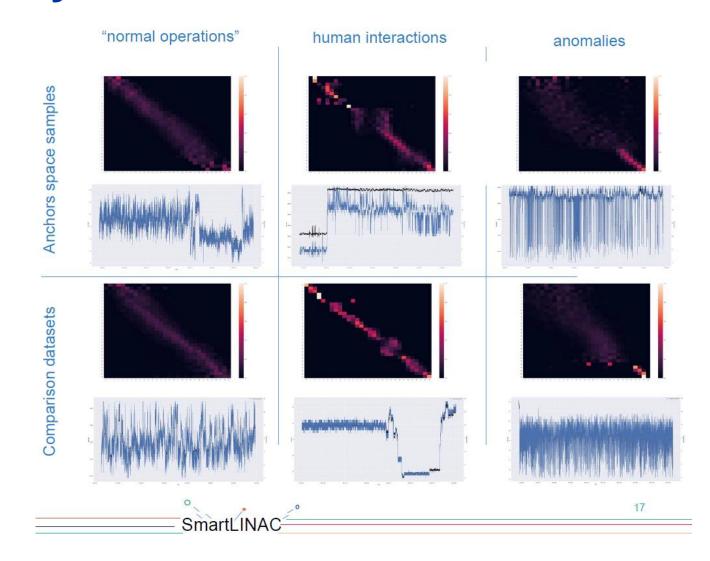
Reproduction error

Hidden Markov models

Probability of sequence

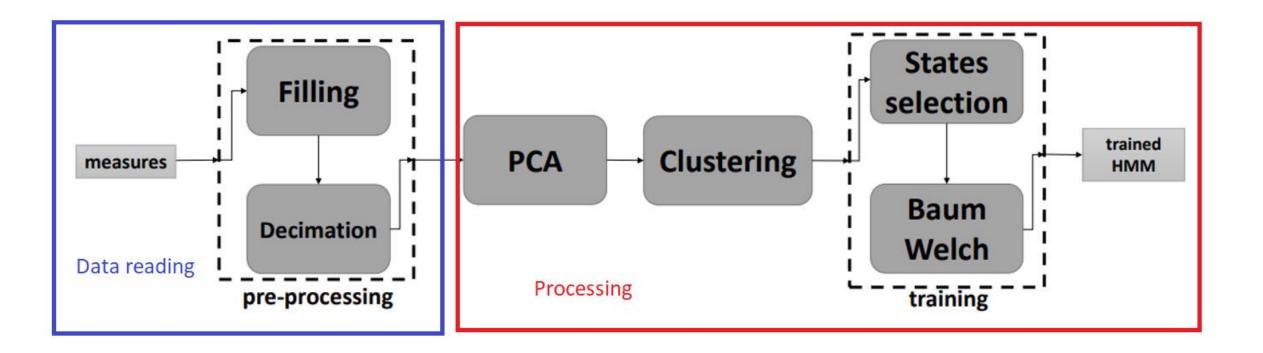
Proposed method

Stochastic matrix as input for CNN





Estimating confusion matrix





Parallelization

Dask (running on Swan notebook)

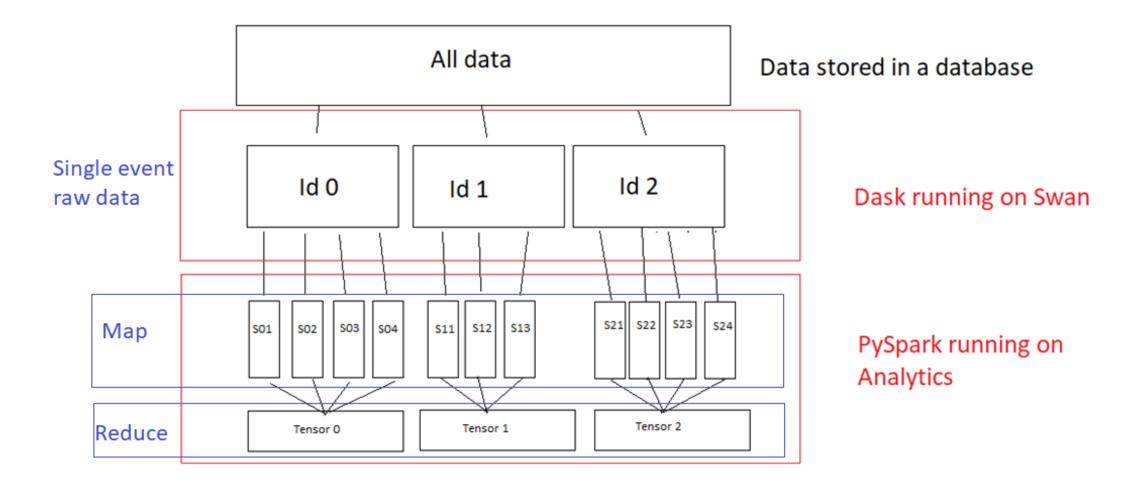
- Data reading
- Input: Id
- Output: (relevant) raw data

Spark (running on analytics)

- Map:
 - Gaussian filter
 - Binning
 - Stochastic matrix
- Reduce
 - Combine tensors

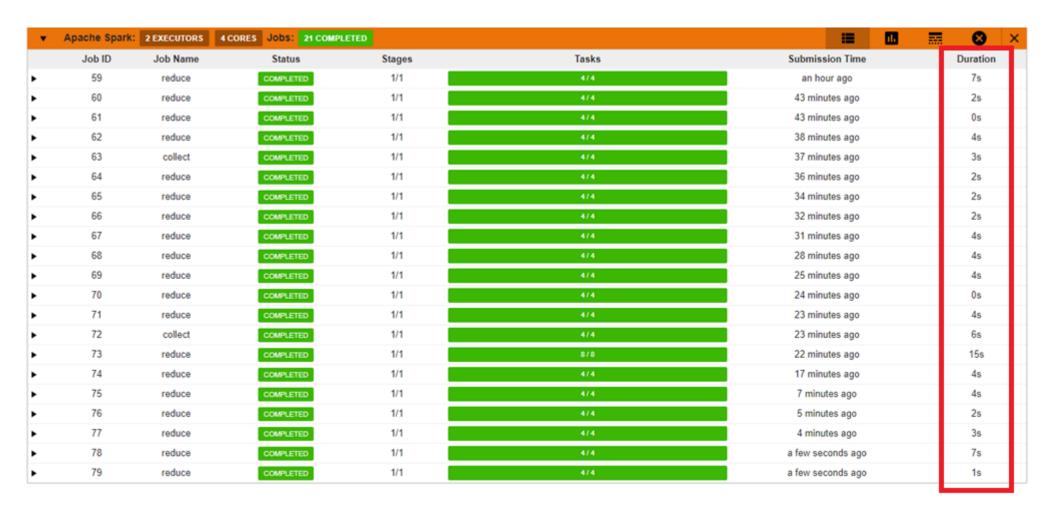
```
def submit_spark_job(context, local_bind_address = 3306, idSource = 27, increment = 5000, test = False):
try:
    results_l = get_raw_data_list(local_bind_address, idSource, increment, test)
except Exception as e:
    print(e)
    return []
rdd = context.parallelize(results_l)
rdd_calc = rdd.map(statisticalAnalysis_raw)
fr = rdd_calc.reduce(combine_tensor)
return fr
```

Approach to parallelization





Implementation and performance



Total read time: 1.6 h

Total map reduce time: 80s



