## AGATA Week 2024

# **Liverpool R&D: (ML/AI) Status**

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### Presentation Overview

### **AGATA signal response**

- **Parametric vs Optimisation based PSA**
- Recap of SIMPLEX PSA algorithm
- Initial results & issues
- Masking within AGAPRO
- Dual-Segment PSA
- Crystal-wide Optimisation
- Intelligent Preprocessing



### **AGATA at LNL (22/10/23)**





- High-Fold signals are separated and processed individually when possible.
- Conventionally this is achieved using segment windows, renormalisation & recursive subtraction.
- $\blacktriangleright$  This only works on well-separable signals, typically a few segments apart.



## Multiple Interactions Within the Same Segment

These can look near-identical to Fold-1 signals and will regularly fool parametric, PSA & ML methods.

- The underlying signals are heavily entangled and cannot be solved individually.
- They have massive implications for tracking (e.g. a [60°,30°] scatter pair looking like a 90° scatter).



Within AGATA Pulse Shape Analysis is used to quickly compare experimental signals  $(M)$  against simulated signals  $(S)$  using a Figure of Merit (FoM).

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$$
FoM = \sum_{i=1}^{37} \left( \sum_{j=1}^{121} (|M(j)_i - S(j)_i|)^2 \right)
$$

The position of the closest match in the basis is then used for the predicted interaction position and given to the GRT.

Current PSA utilises a 2-part grid search (AGS), initial optimisation of a coarse grid followed by a fine-grid search around the local optima.

This work focusses on solving 2-interactions within the same segment for AGATA.

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### Parametric Vs Non-Parametric PSA

Parametric PSA concerns features that can be easily derived  $(T<sub>10-90</sub>$ , ICA)

- These methods are easy to program, simple to apply to ASICs & FPGAs
- They are typically surjective (many positions have the same value).
	- Insufficient for use in  $\gamma$ -ray localisation & tracking.
- Apply poorly to multi-interaction PSA.

Non-parametric PSA concerns features that are more complex:

- Components of the signal (e.g. from PCA/SVD or AI)
- Can be used to form regression predictors (e.g. in CZT)
	- Inaccurate but fast.
- Often used in optimisation  $(kNN)$  for localisation:
	- Accurate but slow.



### Non-parametric Responses – Coax Detectors

- Monolithic Coaxial detectors show some signal variation with position, most signals form a continuous response manifold in  $\mathcal{R}_n$ .
- This can be observed with covariance matrices.
- Alternatively, we can use manifold learning.





### AGATA Response

- Segmentation within AGATA provides a wealth of information within the signal response.
- Detector response goes from temporal to spatiotemporal.
- Segment responses concentrate in submanifolds.
- Strong radial and azimuthal dependence observed.
- $\blacktriangleright$  Transient signals provide additional fidelity  $(x, y, z)$ .
- $\blacktriangleright$  These effects allow for  $\sim$  3 interaction PSA within a segment.
	- $\triangleright$  Upper limit of  $\sim$  9 interactions across a crystal.



### Graph Structures & Accelerated PSA



As part of my work I investigated the use of graph structures to accelerate PSA, they have several marked benefits:

- ▶ Data-driven : no parametric methods (e.g. T<sub>90</sub>) needed.
- Self-organised ∴ no rigid grid structure required.
- ► Hierarchical : adaptive resolution.

In my work I profiled several methods for accelerated PSA:

- $\blacktriangleright$  Space Partitioning (exact  $kNN$ ):  $kD$ -Tree, M-Trees, etc.
- $\triangleright$  Coarse Indexing (approximate  $kNN$ ): HNSWLib, FAISS.

Recommendations from this work are being implemented.

My current work is focussed on multiple-interactions.



### Multiple-Interaction PSA

Multiple interaction events make up a significant fraction of the data collected by AGATA.

Around 30% of events in AGATA are comprised of events where multiple interactions occur in the same segment.

- $\blacktriangleright$  This will vary as a function of y-ray energy
- In these cases, the PSA tends to fit a weighted average of the positions.
- Preference to predict barycentre at segment centres.

Evaluation of multi-interaction events using conventional PSA is not feasible.

Initial implementation of GRETA algorithm was performed but a flaw was found in their approach that is currently irreconcilable.

Currently working with the GRETA team to see if it's fixable.





### Simplex De-mixing in High Dimensions

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A new convention was developed to evaluate multi-interaction signals using distance geometry and simplex de-mixing.

$$
\blacktriangleright \overrightarrow{Op_1} \cdot \overrightarrow{qO} = \overrightarrow{Op_2} \cdot \overrightarrow{qO} = \overrightarrow{p_1p_2} \cdot \overrightarrow{qO} = 0
$$

 $\blacktriangleright$  ∠0 $p_1p_2$  & ∠0 $p_2p_1$  < 90°

**This convention allows for direct fitting of multi-interaction signals.** 

Also works recursively, allowing for 9-interaction fits.

- Exhaustive combinations are required to find the best solution:
	- ▶ ~1,100,000 combinations for 2 interactions.
	- ▶ ~550,000,000 combinations for 3 interactions.
	- $\blacktriangleright$  ~10<sup>30</sup> combinations for 9 interactions.
- $\triangleright$  Not feasible to do on a CPU exhaustively real-time.
	- Somewhat possible on a GPU.
	- How about using graph acceleration?



### Hypergraph-accelerated PSA

- ▶ 2-interaction solutions exist as edges that connect between nodes of two orthogonal trees.
- Lower levels have more nodes ∴ more edges.
- **Just like how nodes can have child nodes edges can have child** edges.
	- In the validity of an edge applies recursively.
		- If an edge is invalid, all child edges are also invalid.
- $\blacktriangleright$  This allows for us to cull large sections of the graph.



### Hypergraph-accelerated PSA

Two hypergraph-accelerated methods have been developed:

- $\blacktriangleright$  Hyper-M-Tree
	- Recursive covering (Exact  $k$ NS search).
- Hyper-Hierarchical New Small World Graphs.
	- Greedy simplex traversal (Approximate  $k$ NS search).

These overcome conventional limitations of the existing method:

- Mative support for PSA uncertainty
- ▶ Self-learned hierarchy.
- Support for dynamic resolution (FEM) bases.



### General PSA Update



SIMPLEX and GRETINA PSA codes have been implemented within AGAPRO;

- **Exhaustive & AGS approaches available.**
- Rudimentary support for uncertainty propagation.
- Accelerated using SIMD instructions.

### **A flaw was found in the precomputation step W.R.T masking of distances**

- Investigating the remedy uncovered other issues unrelated to the new algorithm.
- ▶ Currently working on a more comprehensive PSA pipeline.
	- ▶ Better handling of single vs multi-interaction events & Recursive Subtraction & Time-shifting.
	- Integrates a more intelligent preprocessing using ML.
	- Will have a crystal-wide optimiser for the full superpulse.

### Assessing PSA performance

A dataset was extracted from E680 at GANIL, fission  $^{98}$ Zr nucleus with a strong y-ray transition at 1.2 MeV.

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- Fragments were emitted with velocities of  $\beta \approx 10$ , making the resolution of such transition strongly dependant on the quality of the Doppler correction.
	- Doppler correction only applies to first interaction.
- ▶ PSA also will heavily influence tracking accuracy.
	- Should influence peak prominence.

### **Important note:**

- The spectral performance of AGATA is the combined effect of PSA, GRT & simulated bases accuracy.
	- GRT has yet to be tuned for the new PSA.
	- **Bases are being recalculated at higher fidelity.**
	- **This is all with the current flaws, I am working on fixes.**
	- **Performance will likely change in the future.**

### First Hit Predictions – Front Faces **15**





### First Hit Predictions – Back Faces **16**





### Interaction Positions

Observed number of interactions changes drastically with the new algorithm

Prevalence of single-interaction signals is reduced significantly, average track length increased.

Previously 44% of tracks were P.E single-site Now only 17% are P.E single-site





### Dynamic Masking in AGAPRO

Whenever a signal is processed in AGAPRO a window is created to allow PSA to process events independently.

- $\blacktriangleright$  This occurs roughly 2/3 of the time.
- Comes in two forms hmask & lmask (& nmask):
	- hmask is pre-event, depends on the hit segment.
		- $\triangleright$  neighbours =  $-2$
		- ▶ 36 static masks that are well-known.
	- I mask is live, depends on the other events in the crystal.
		- Removes other net charge signals.
		- Could remove neighbour transients, currently disabled.
		- Around 3500 unique masks
- Because lmask changes for each event the precomputed distances are directly affected.
	- **This is problematic, there are 35 million of them.**





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### Memory Constraints



Precomputed distances are a key part of the multi-interaction algorithms, vastly outnumber novel distances.

- Segment 6 has ~2,000 novel distances, ~2,000,000 precomputed.
- Distance computation is expensive, 1-2,000 float operations each time.
- $\blacktriangleright$  Having to re-compute  $d_c$  for each event is not feasible to do real-time.

Thankfully  $L2^2$  is commutative, can be split into sub-distances:

- $\blacktriangleright$   $d_c$  can be precomputed for each segment, summed once the lmask is determined
	- Requires significant memory increase of  $\sim$ 36x
- Could be optimised using Boolean Covering to find prime implicants using Quine-McCluskey algorithm.
- $\blacktriangleright$  d<sub>c</sub> can be precomputed for hmask and its components
	- ▶ Calculate hmask distance and then subtract lmask components
		- **Slightly better memory efficiency, only 8x previous version**
- Current RAM utilisation is 6 GB/crystal, could be reduced to ~2.5 GB

### Dual-Segment PSA

Occurrence of hits that influence lmask is around 63%, vast majority occur in neighbouring segments.

- Even without removing neighbours of coincident events the lmask is significantly reduced.
	- Some PSA is run on a single segment, not great.
	- **Presence of transients is not accounted for, will affect PSA.** 
		- Removing the neighbour segments altogether is worse.
- ▶ 2-interaction PSA is not explicitly limited to a single segment.

▶ Can be modified to run on the full crystal.

- ▶ Significant memory requirements again, ~32x larger
- ▶ Can be modified to run on overlapping windows.

▶ Slightly better, 8-12x larger.







- Within AGAPRO PSA the superpulse is separated into windows, and processed at the segment level.
	- Masking minimises contributions from other interactions but is not perfect.
		- No way to account for transients on net charge signal.
		- **Fitting errors affect subsequent fits.**
	- A global fit across multiple clusters isn't feasible.
		- Mathematically possible, but time and memory consuming.
- $\blacktriangleright$  Formalism for a generalised solution for *n*-interactions was developed.
- This allows for a crystal-wide solution to be found for the full superpulse, not just each window.
- The best 1 & 2 interaction solutions for each segment can be passed to a global fitter.
	- A global solution is then found from the reduced set.
	- Allows for crystal-wide re-optimisation of energies, even within the same segment.

### Intelligent PSA Preprocessing

- ▶ 2-interaction PSA is a time-consuming process.
- Assessment of 1 vs 2 interactions is performed after PSA search via fallback vetoes:
	- Interaction separation.
	- **Energy fraction & absolute energy.**
	- $\rightarrow \chi^2$  reduction.
- It'd be useful to skip the 2-interaction PSA if it's unlikely to give a good result.
- Can we infer what the expected  $\chi^2$  , energy fractions & separation are before PSA?
	- If so, then can we only run 2-interaction on events that will show an improvement?
	- ▶ Can this information provide a better starting guess for PSA?

#### Self-Organising Maps are a graph-ANN that use competitive learning to fit maps to an underlying manifold of data.

- I The map is initialised and slowly iterates to form a topologically constrained approximation of the data.
- Nodes in the graph then learn from their immediate surroundings via inference.
- **Often used for clustering and feature detection.**
- I've adapted them for polylinear regression.







## **SOMs for Pseudo-Parametric PSA 25**

- SOMs can be augmented with existing analytical PSA techniques to perform regression.
- Map is converted into a simplicial complex.
- This allows for a poly-linear segmentation of the response space.
- Deviations from the map spline can be considered noise.
	- Significant deviation could be considered multi-interaction.
- Analytical code is easy to accelerate, map is easily tuneable.
	- Multithreaded capable of running at 3 MHz.
	- Can be implemented on FPGAs, Digitisers.
- **Should work for multi-input regression:** 
	- **Estimate both regression parameters and energy fractions.**
	- Maths is very long, requires 4-interaction PSA.





**AGATA Response (UMAP Embedding)**

### Performance on Coaxials





- Further refinement using experimental data.
- Inference then applied to learn radial position.
	- $\triangleright$  Should be able to learn t<sub>0</sub> offset & neutron damage.
- **Initial experimental results look promising:** 
	- $\blacktriangleright$  Worse performance than  $k$ NN, better than parametric.
	- **Limitations in interpolation, will be fixed with GSOMs.**

Implementation of FPGA-native code ongoing.



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ML-Assisted Interaction Veto

- In the number of interactions can be inferred by estimating the  $\chi^2$  for approximations of the underlying manifold.
- ▶ 1-interaction SOM approximation requires determining best simplex interpolation in the map.
	- Essentially conventional 2-interaction PSA.
- ▶ 2-interaction SOM approximation requires determining the best interpolation of 2 interpolations.
	- **Essentially conventional 4-interaction PSA.** 
		- Solving of a 5-simplex, (hyper-hyper tetrahedron).
		- **Moderately difficult.**
- If the  $\chi^2$  for the 2-interaction approximation is significantly better we can assume the full search will do the same.
- If the  $\chi^2$  for the 2-interaction approximation fails the vetoes we can assume that the full search will fail to find a good solution either.

**Signal Manifold SOM Nodes Experimental Query 1-int Approximation 2-int Approximation**





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**Signal Manifold SOM Nodes Experimental Query 1-int Approximation 2-int Approximation**





### Intelligent Search Structures

- $\blacktriangleright$  Initial testing of the veto looks promising but needs proper implementation in AGAPRO.
- **Initial predictions of the energy fractions & indices** are ignored, could be factored into further PSA.
- Estimation of  $t_0$  offset would reduce iterative fitting.
- Ideally should inform predictions of further PSA.
	- Could be used to build better search structure.





# **Thanks for Listening Any Questions?**

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