

Deinterleaving RADAR pulses

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PhD students' day

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Industrial thesis (CIFRE)

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- Phd student, 3rd year



AVANTIX (ATOS Group)

- French company specialized in Electronic Warfare
- Designs critical high-tech systems to provide homeland protection and military forces
- Radar processing chain
 - ▶ Data processing based on the interception of a signal by a listening system
 - ▶ Provides strategic information to military platforms



The subject

- Use and development of machine learning tools for passive RADAR intelligence
- Improvement of the RADAR deinterlacing chain
- Passive RADAR(s):
 - ▶ Emit no electromagnetic waves
 - ▶ Only receivers
 - ▶ Detect, track and identify RADAR(s)

Radar Deinterlacing process:

- ➊ From a signal, separation and grouping of pulses
- ➋ Extraction of characteristics
- ➌ Comparison of groups to Radar database which contains all the characteristics of the RADARS
- ➍ Identification of the RADAR present or updating of the database

Content

- ① Data description
- ② Proposed Approach
- ③ Application case
- ④ Conclusion

Content

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- 2 Proposed Approach
- 3 Application case
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Data description

- Work from simulated labeled data
- Each dataset is composed of RADAR pulses, described by only 4 features:
 - ▶ Frequency (F)
 - ▶ Pulse Width (PW)
 - ▶ Level (L)
 - ▶ Time of Arrival (TOA)

Data description

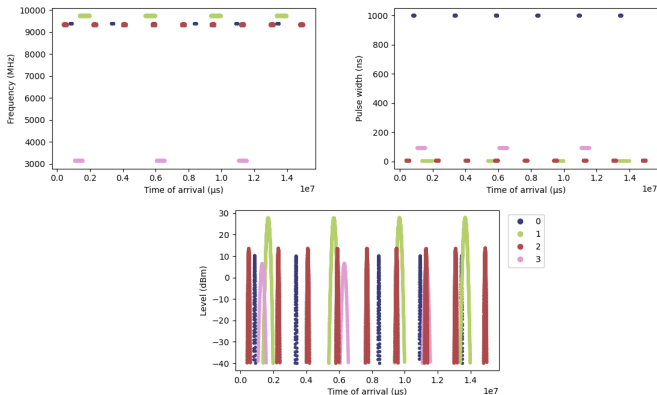


Figure: Example of a simple signal containing 4 RADARs.

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Implementing a two-step strategy for signal deinterlacing:

- ❶ **Clustering:** Cluster each observation of this signal under the constraint of not merging different radars in one cluster
- ❷ **Cluster fusion:** Hierarchical agglomerative clustering combined with optimal transport distances

Step 1: Clustering with HDBSCAN

- Application of clustering algorithm on two features: Frequency and Pulse Width
- Why HDBSCAN ?
 - ▶ Unsupervised algorithm that allows us to search for clusters of varying densities
 - ▶ Few parameters to optimize
- Over-estimation of the number of clusters
 - ▶ Allows to receive RADARs that do not transmit continuously
 - ▶ Each cluster must not contain the observations of different RADARs

Output of HDBSCAN

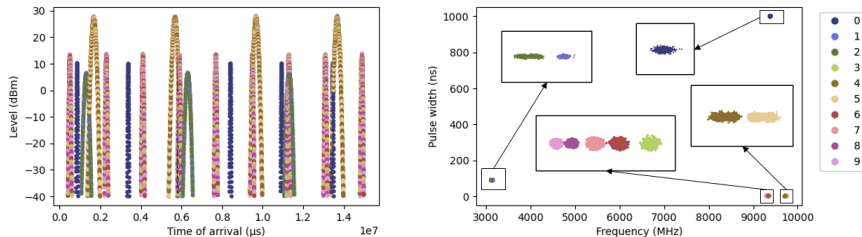


Figure: HDBSCAN Output. Colors represent the different clusters.

Step 2: Cluster fusion

Methodology:

- ➊ Dimensionality reduction: each cluster is represented by a histogram of its TOA and LEVEL
- ➋ Compute distances between cluster histograms using optimal transport
- ➌ Aggregation of the 2 closest clusters i.e., the 2 clusters with the lowest transport costs
- ➍ Computation of new costs between old clusters and the new ones
- ➎ Repeat steps 2 to 4 until a single cluster is obtained

Histograms

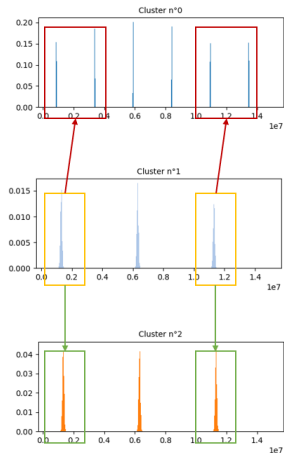


Figure: Histograms of the previous clusters.

Optimal Transport

- Transporting distribution to another has a cost:

$$C(P) = \sum_{n=1}^N \sum_{m=1}^M C_{nm} P_{nm} = \langle C, P \rangle \quad (1)$$

$C(P)$ = Total cost

C_{nm} = cost transporting unit mass from x_n to y_n (transport cost)

P_{nm} = quantity of mass taken from x_n to y_n (transport plan)

- The optimal transport plan is a simple minimization problem:

$$P^* = \operatorname{argmin}_{P \in \mathbb{R}_+^{N \times M}} \langle C, P \rangle \text{ such that } P \mathbf{1}_M = a, P^T \mathbf{1}_N = b^T. \quad (2)$$

Dendrogram

- Decision model based on several unsupervised metrics to cut the dendrogram:

- ▶ Silhouette score
- ▶ Calinski-Harabasz score
- ▶ Davies-Bouldin score

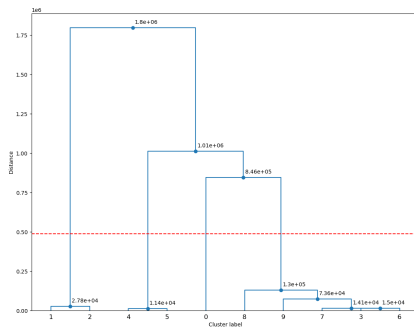


Figure: Dendrogram that presents in a simplified way the cluster aggregations with the distances of the optimal transport.

Final Grouping

- Perfect grouping
- Statistics:
 - ▶ Homogeneity = 1
 - ▶ Completeness = 1
 - ▶ Unclassified rate = 2%

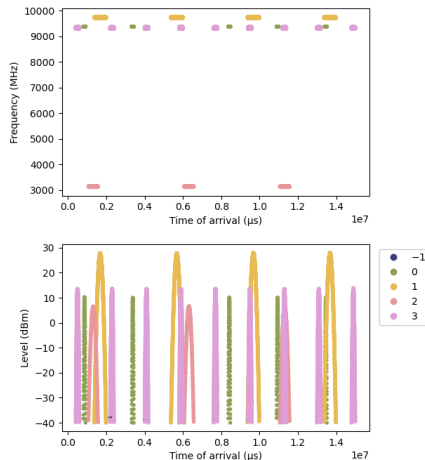


Figure: Final grouping

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Unlabeled data

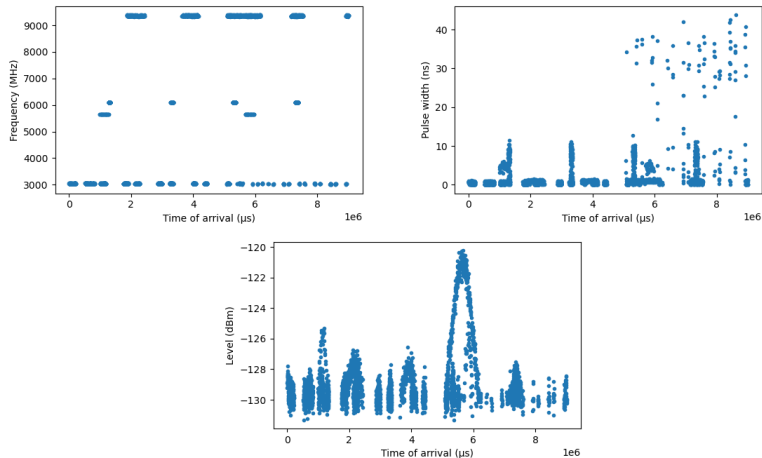


Figure: Unlabeled data with 5 RADARs.

Step 1: Clustering with HDBSCAN

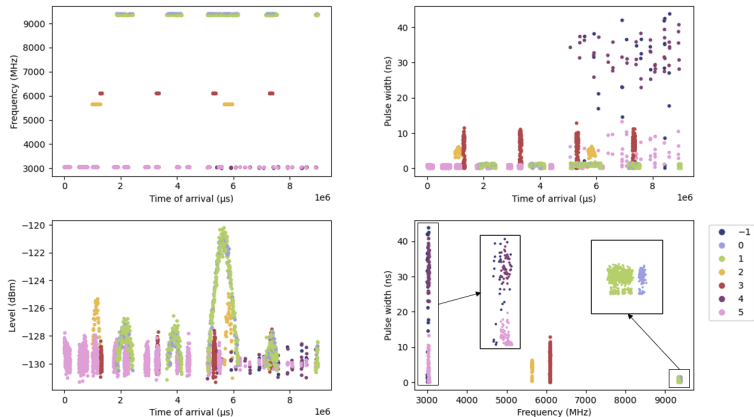


Figure: HDBSCAN Output.

Histograms

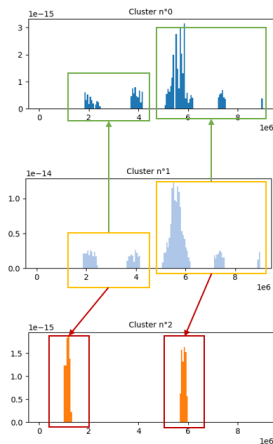


Figure: Histograms of the previous clusters.

Final Grouping

- Pulses of the 2 transmitters around 3GHz are mixed
- Hdbscan can not differentiate them at step 1
- Statistics:
 - ▶ Homogeneity = 0,94
 - ▶ Completeness = 1
 - ▶ Unclassified rate = 0,9%

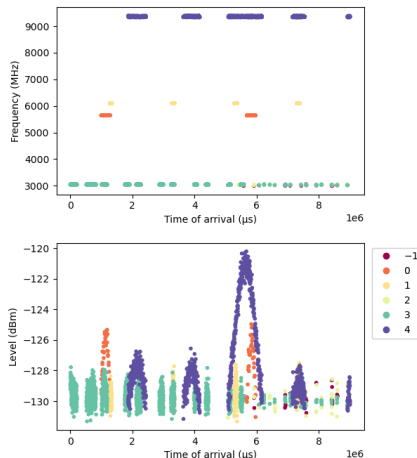


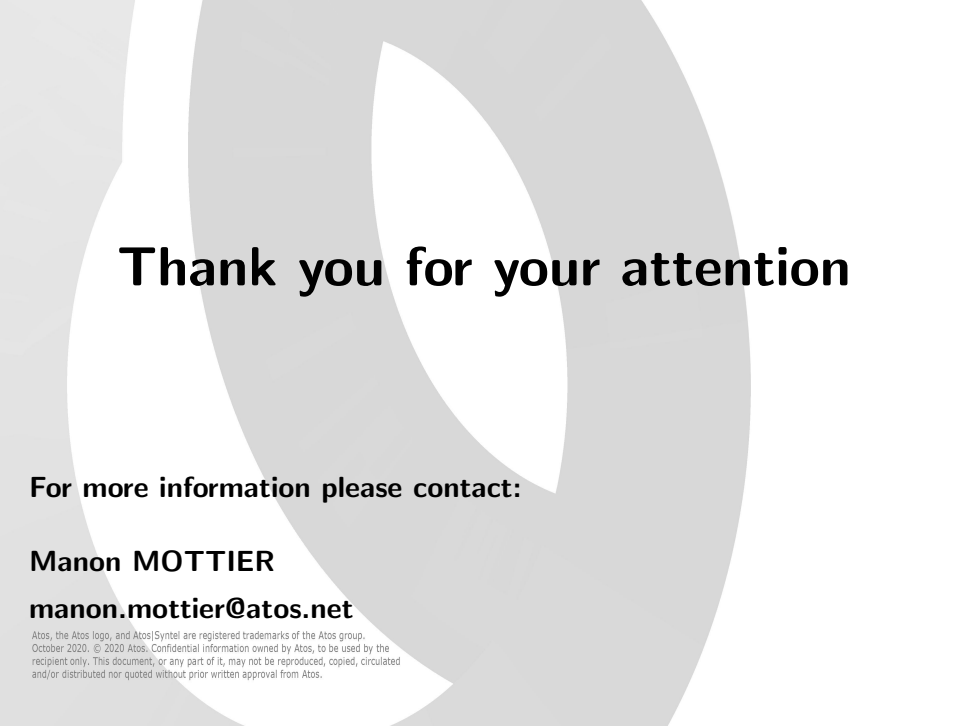
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Conclusions and Next objectives

- Works on moderately complex cases
- Publication of a Conference paper: “Deinterleaving and Clustering unknown RADAR pulses” - Manon Mottier; Gilles Chardon; Frédéric Pascal
- Next objectives:
 - ▶ Improvement of the proposed approach
 - ★ Define a metric to identify identical RADARs in Frequency and Pulse Width
 - ★ Improve model decision to cut the dendrogram
 - ▶ Classification and RADAR identification from a reference database



Thank you for your attention

For more information please contact:

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