

Deinterleaving RADAR pulses

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- 1 Context
- 2 Proposed Approach
- 3 Application case
- 4 Conclusion and Next objectives

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Industrial thesis (CIFRE)

- Phd student, 3rd year
- The Signals and System Laboratory of CentraleSupélec
 - ▶ Thesis director: Frédéric PASCAL
 - ▶ Co-supervisor: Gilles CHARDON
- AVANTIX (ATOS Group)



AVANTIX (ATOS Group)

- French company specialized in Electronic Warfare
- Designs critical high-tech systems to provide homeland protection and military forces
- Provides strategic information to military platforms



The subject

- Use and development of machine learning tools for passive RADAR intelligence to identify RADARs present in the environment.
- Improvement of the RADAR deinterlacing chain:
 - ① Separation and grouping of pulses
 - ② Identification of the RADAR present or updating of the database

Work organisation

- Full time in the laboratory (Avantix is based in Aix-en-Provence)
- Weekly meeting with the company via teams
- Half-yearly meeting with the company, the supervisors and the DGA

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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

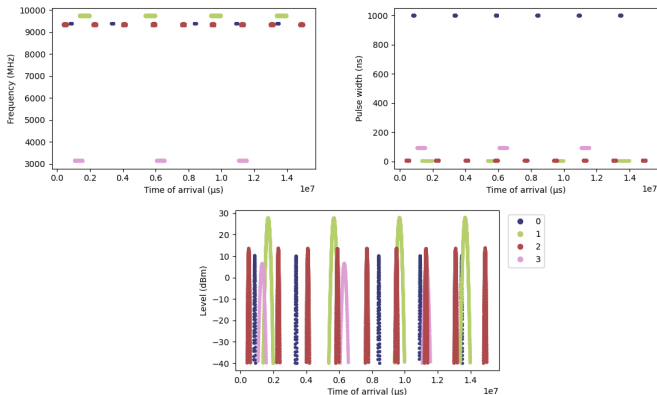


Figure: Example of a simple signal containing 4 RADARs.

Implementing a two-step strategy:

① **Deinterlacing pulses:**

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

② **RADAR Identification:** Extraction of group characteristics and comparisons to a reference database to identify the transmitter

Clustering with HDBSCAN

Methodology:

- 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

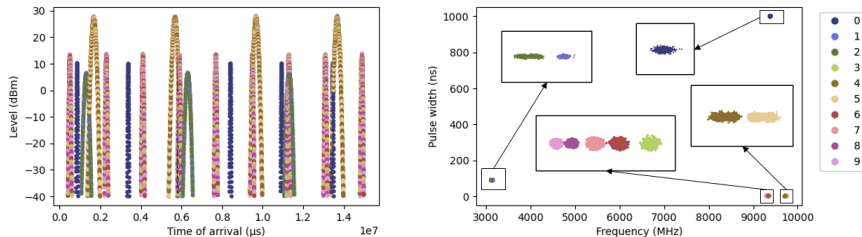


Figure: HDBSCAN Output. Colors represent the different clusters.

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

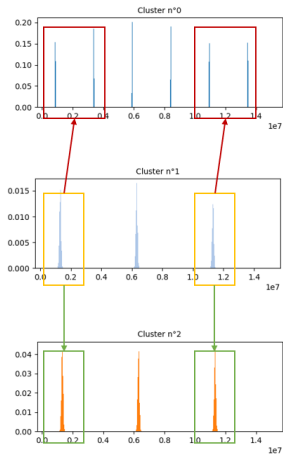


Figure: Histograms of the previous clusters.

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_m (transport cost)

P_{nm} = quantity of mass taken from x_n to y_m (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

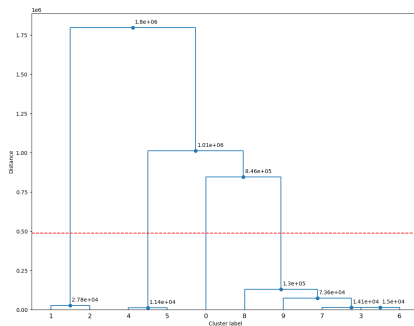


Figure: Cluster aggregations with Optimal Transport Distances.

Final Grouping

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

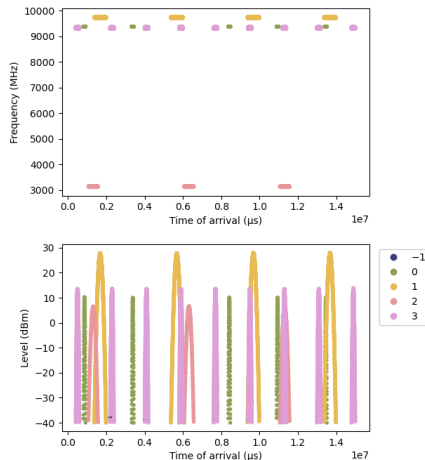


Figure: Final grouping

RADAR Identification

Methodology:

- 1 Construction of a measure that describes each RADAR belonging to a reference database based on its frequencies:

$$\mu_j = \sum_{n=1}^N \alpha_n \delta_{f_n} \quad (3)$$

N = the number of frequencies on which the RADAR transmits

α = the weight of the frequency

δ = dirac mass (with $\sum \alpha_n = 1$)

RADAR Identification

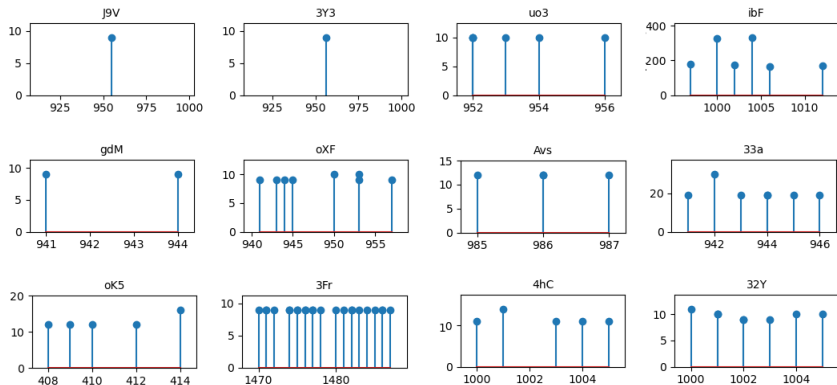


Figure: Referential Database representation. Example of the frequency representation of the RADAR present in the identification database. The Radar ID and value range have been changed.

RADAR Identification

- ② Construction of a measure that describes each cluster from the frequency:

$$\gamma = \frac{1}{N} \sum_{m=1}^N \delta_{f_m} \quad (4)$$

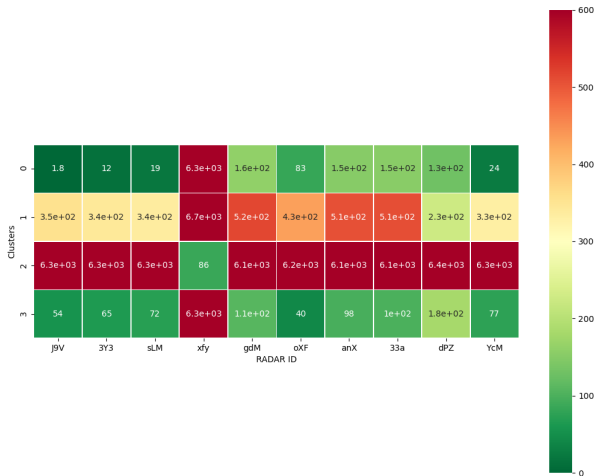
N = the number of pulses in the cluster

- ③ Identification of the transmitter:

$$j^* = \operatorname{argmin} d(\mu_j, \gamma) \quad (5)$$

Classification Output

- **Extract** of the cost matrix between the groups and the some waveforms.
- Red: High transport cost
- Green: Low transport cost



Classification Decision

Clusters ID	Rank 1		Rank 2		Rank 3	
	ID	Cost	ID	Cost	ID	Cost
0	eF2	1.62	J9V	1.83	sxh	6.02
1	BGF	1.58	d2B	1.54	smn	28.70
2	viY	1.43	xfy	85.75	oK5	98.75
3	D9V	2.14	36y	13.05	BF8	13.06

Table: Classification decision.

- Correct identification
- Problem with mono frequency RADAR
- Difficulty to identify sub-modes
- Frequency not sufficient

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Unlabeled data

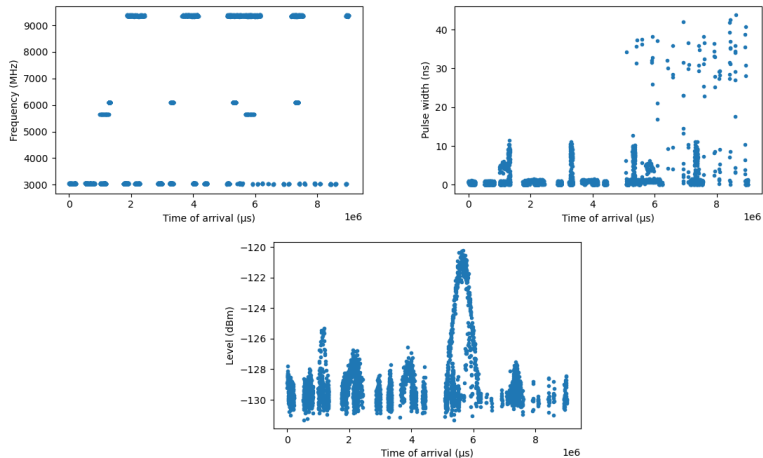


Figure: Unlabeled data with 5 RADARs.

Clustering with HDBSCAN

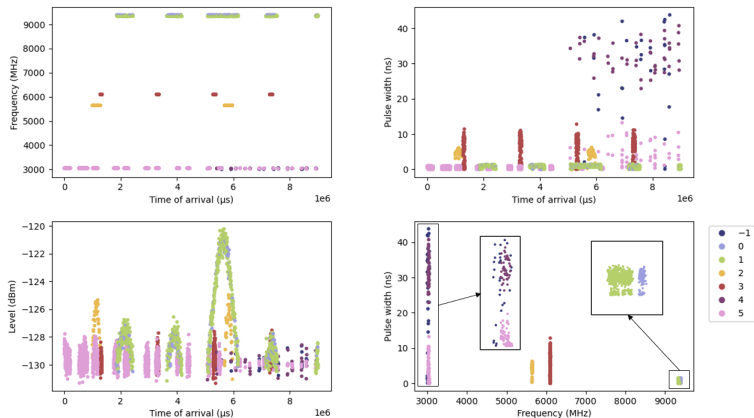


Figure: HDBSCAN Output. Colors represent the different clusters (including outliers' class).

Histograms

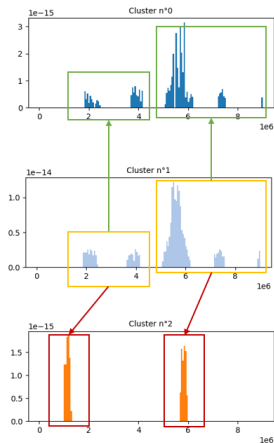


Figure: Histograms of the previous clusters.

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%

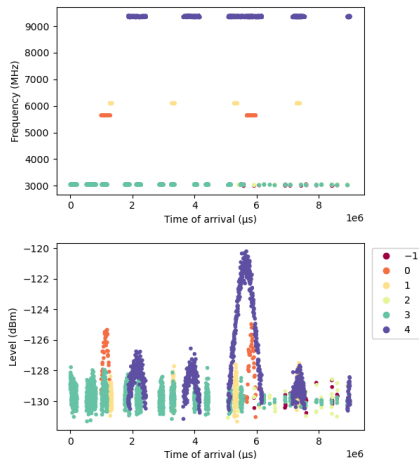


Figure: Final grouping.

Identification

Clusters	Rank 1		Rank 2		Rank 3	
	Id	Cost	Id	Cost	Id	Cost
0	32P	2.35	yfN	22.60	YuK	29.49
1	5fd	2.51	38M	30.84	32P	455.27
2	XAQ	7.23	xfy	10.45	oK5	14.04
3	xfy	2.72	XAQ	16.28	oK5	20.11
4	uo3	1.96	3Rs	1.99	Ykn	6.50

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Conclusions

- Works on moderately complex cases
- Identification phase gives encouraging results
- Communication:
 - ▶ Publication of a Conference paper: “Deinterleaving and Clustering unknown RADAR pulses” - Manon Mottier; Gilles Chardon; Frédéric Pascal
 - ▶ International Conference: IEEE RadarConf 2021, Atlanta, GA, USA
 - ▶ International Seminar: Atos Scientific Community

Next objectives

- Improve the proposed approach:
 - ▶ Deinterleaving Radar pulses:
 - ★ Define a metric to identify identical RADARs in Frequency and Pulse Width
 - ★ Improve model decision to cut the dendrogram
 - ▶ Identification:
 - ★ Add Pulse width to improve identification
 - ★ Define a distance between Groups
 - ★ Detection of new transmitter
- Publication of an article in a specialized journal

Thank you for your attention

For more information please contact:

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