



AFRL

AIR FORCE RESEARCH LABORATORY BRIEFING OVERVIEW

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Topological Multimodal Sensor Data Analytics for Target
Recognition and Information Exploitation in Contested
Environments

Information Directorate (RI) 08AUG2023

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AFRL – 2023 - 1778



Agenda

- Technical Background
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 - Topological Data Analysis (**TDA**)
 - Persistent Homology (**PH**)/Persistence Diagrams (**PD**)
 - Metrics of PDs
- The TDA AI/ML Pipeline (US Patent Pending)
- TDA AI/ML Pipeline Acoustic Modality Results
- TDA AI/ML Pipeline Electro-Optical (**EO**) Modality Results
- TDA AI/ML Pipeline Infrared (**IR**) Modality Results
- Future Directions

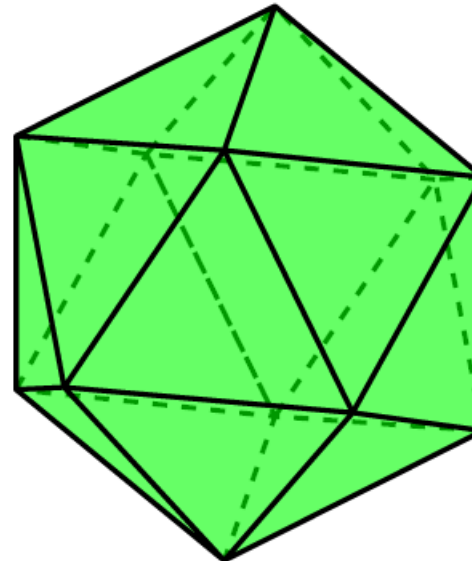
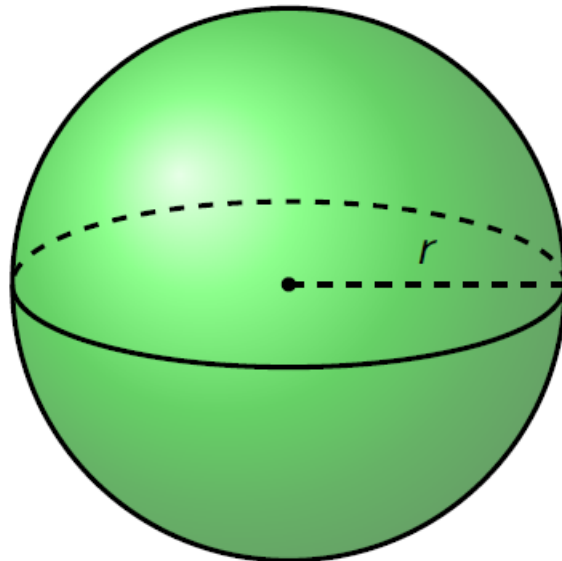


Technical Background (TB)



(TB1) Algebraic Topology

- **Topology** is the mathematical study of abstract spaces (set theoretical) and the continuous transformations (or deformations) between them.
- As its name implies, **algebraic topology** provides information about a topological space through its algebraic characteristics.
- **Homology** is a mature compression methodology in algebraic topology that can measure/record important topological features of a space such as connected components, holes, and voids.



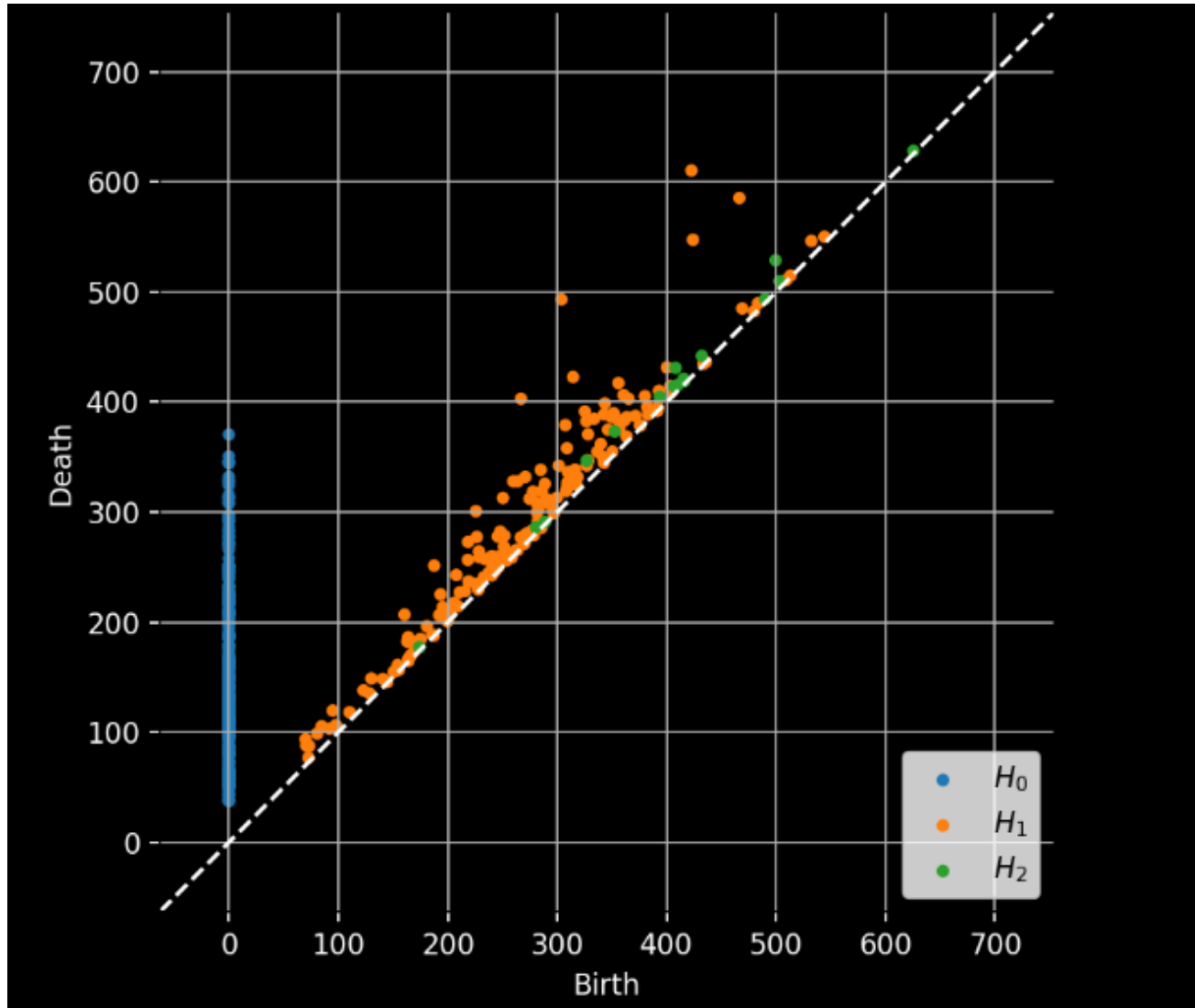


(TB2) Topological Data Analysis (TDA)

- **Topological Data Analysis (TDA)** is a rigorous algorithmically efficient synthesis of applied combinatorial and algebraic topology along with computational geometry motivated by the idea that topology and geometry expose qualitative and sometimes quantitative global features of a data set which are stable under small perturbations.
 - **Persistent Homology (PH)** is a rigorous mathematical compression scheme containing coherent algorithms which encode the developing homology for families of filtered topological spaces indexed by a set of real numbers.
 - **Persistence Diagrams (PD)**: PH gives us a means to monitor and record changes in a filtration by identifying its top'l features (e.g., connected components, holes, and voids) as intervals or ordered pairs (b, d) representing lifespan. The collection of these (b, d) pairs by PH dimension is called a PD of that filtration.



(TB3) An Example of a Persistence Diagram (PD)



- Each sample of acoustic (time-series), EO, or IR data produces a distinct PD
- H_0 or the 0th dimensional persistent homology records the lifespan of connected components
- H_1 or the 1st dimensional persistent homology records the lifespan of 1D holes
- H_2 or the 2nd dimensional persistent homology records the lifespan of 2D holes or voids



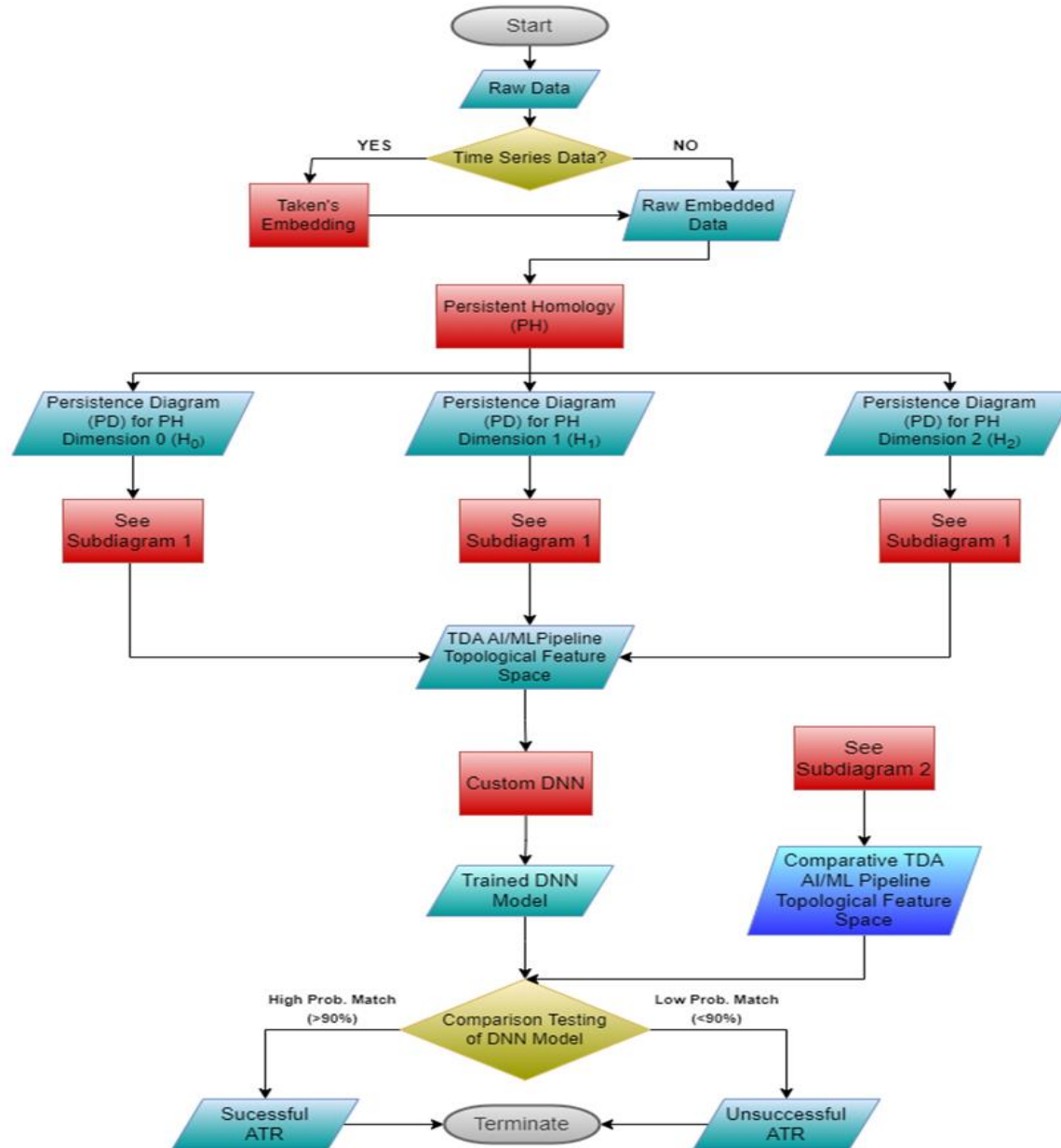
(TB4) The 9 Topological Metrics Used in the TDA AI/ML Feature Space

- Persistent Entropy (**PE**)
 - Number of 'Off Diagonal' Points in a PD (**NoP**)
 - Bottleneck Distance Amplitude (**Btl**)
 - q -Wasserstein Distance Amplitude (**Wass**)
 - Persistence Landscape Amplitude (**PL**)
 - Persistence Image Amplitude (**PI**)
 - Betti Curve Amplitude (**Bet**)
 - Persistence Silhouette Amplitude (**Sil**)
 - Persistence Heat Kernel Amplitude (**Heat**)
- In this study there are a total of 27 topological features (H_0, H_1, H_2 in each metric) in the TDA AI/ML feature space as real scalars.



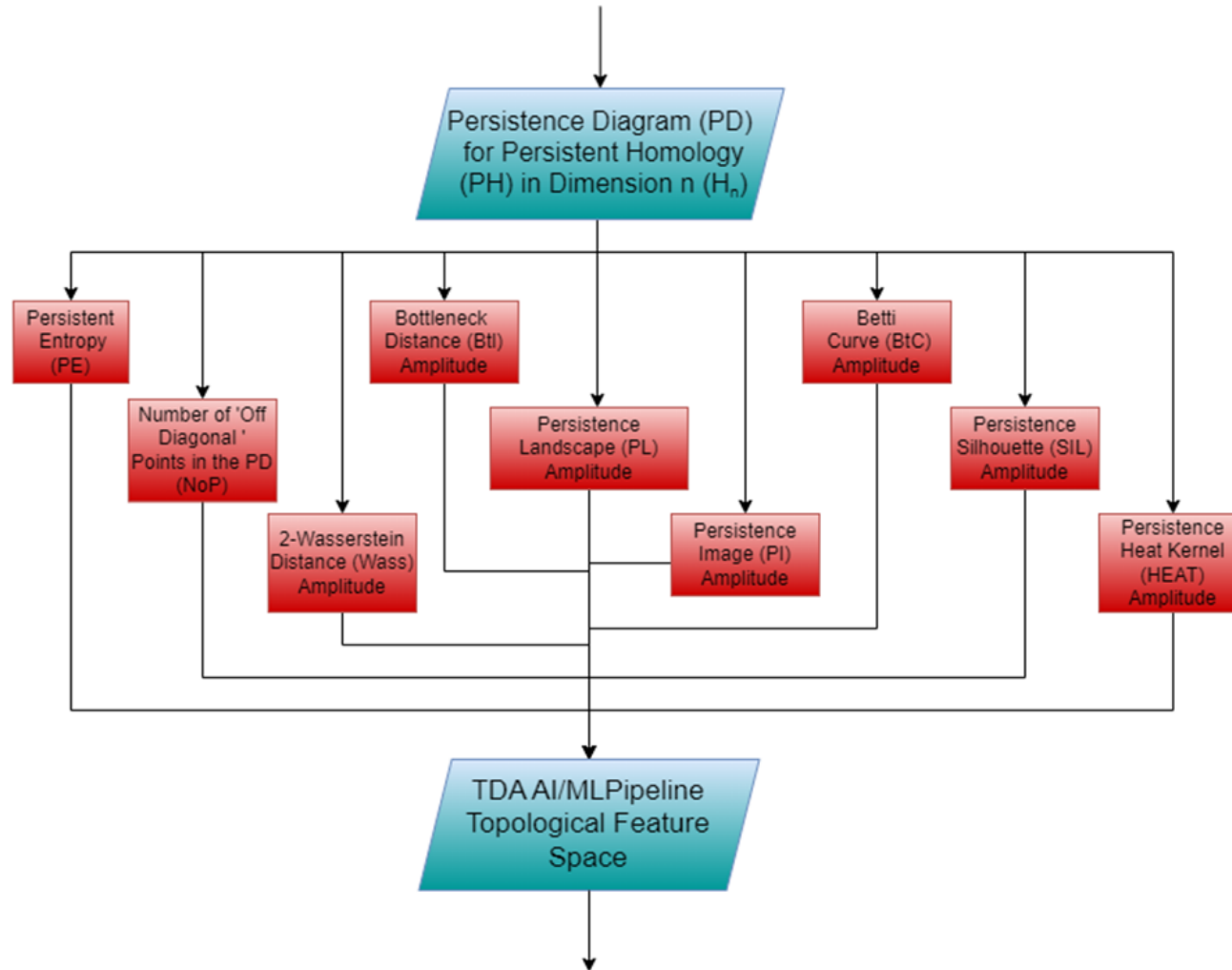
The TDA AI/ML Pipeline Architecture

The Primary TDA AI/ML Workflow Diagram (US Patent Pending)



- Time Series Data = acoustic, seismic, p-RF, radar
- DNN = supervised multi-classification deep neural network
- PH dimension 0 (H_0) records and tracks connected components in the filtration
- PH dimension 1 (H_1) records and tracks 1 dimensional holes in the filtration
- PH dimension 2 (H_2) records and tracks voids (2D holes) in the filtration
- ATR = automatic target recognition

Subdiagram 1 of the Primary TDA AI/ML Workflow Diagram



- The data in each PH dimension (H_0, H_1, H_2) is processed through Subdiagram 1.
- These individual results are then collected in the 27 (9 metrics of 3 dimensions each) dimensional TDA AI/ML pipeline topological feature space.

Subdiagram 2 of the Primary TDA AI/ML Workflow Diagram

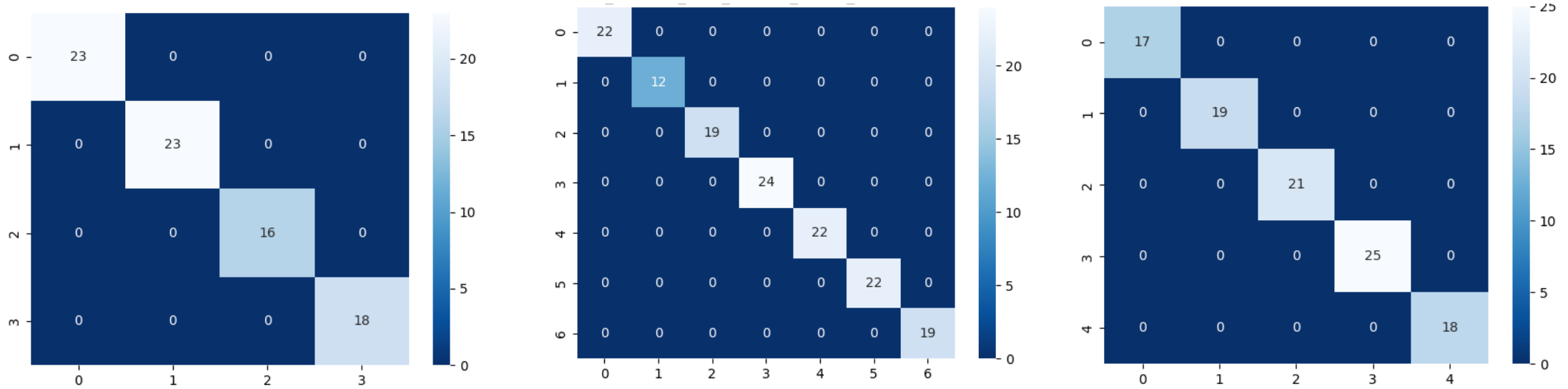


- The comparative data is derived from different timestamps (1-10 seconds away from the training data) in the relative environment scenarios from the training data.
- This comparative data from Subdiagram 2 is then ingested back into the primary TDA AI/ML workflow where a random test set is extracted.



TDA AI/ML Pipeline Acoustic Modality Results

TDA AI/ML Pipeline AFRL Acoustic Modality Results

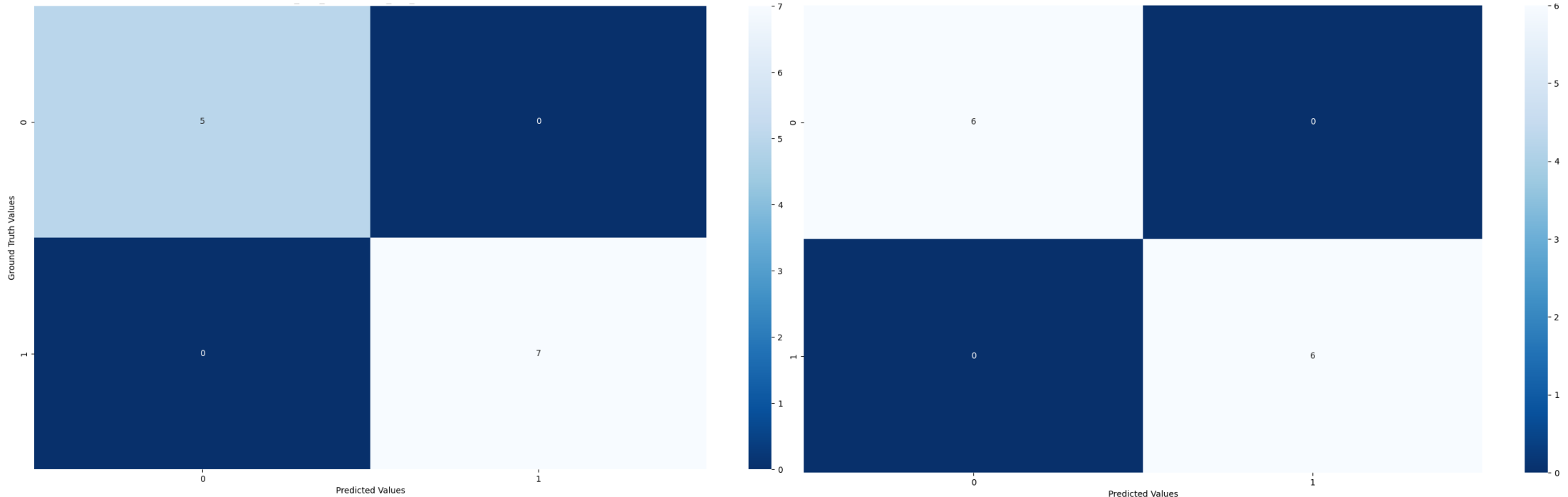


- (Leftmost) The confusion matrix visualizes the predictive ATR capability of the trained TDA AI/ML model for 4 SUAS.
- (Middle) The confusion matrix visualizes the predictive ATR capability of the trained TDA AI/ML model for 7 ground vehicles.
- (Rightmost) The confusion matrix visualizes the predictive ATR capability of the trained TDA AI/ML model for 5 ground personnel (dismounts).
- All three matrices indicate perfect accuracy, precision, and recall.



TDA AI/ML Pipeline EO Modality Results

TDA AI/ML Pipeline EO Modality Results for 2 SUASs

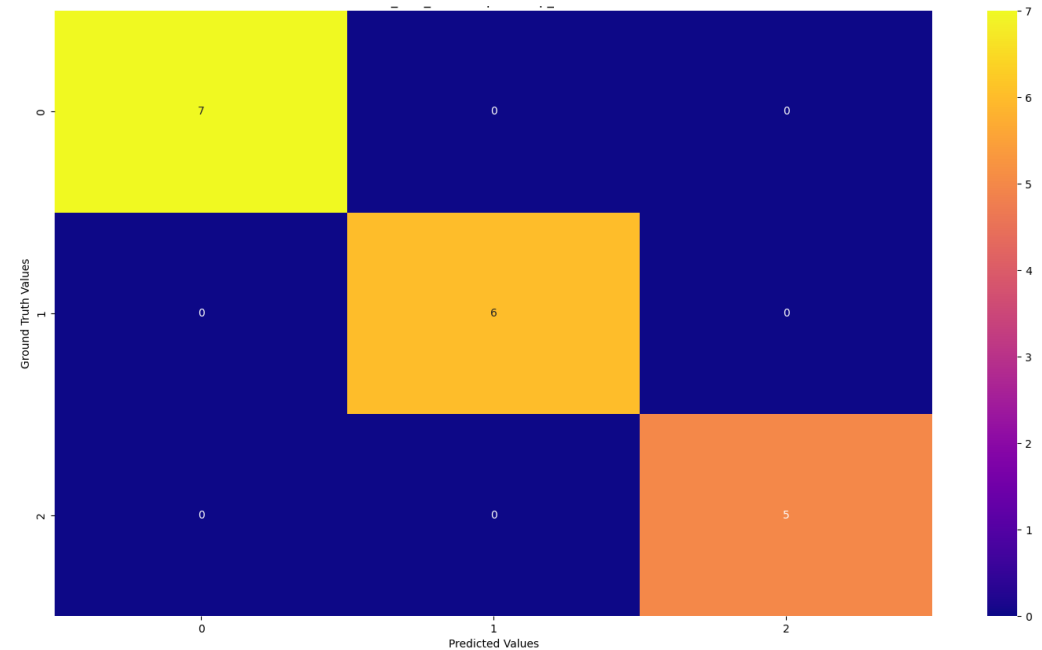
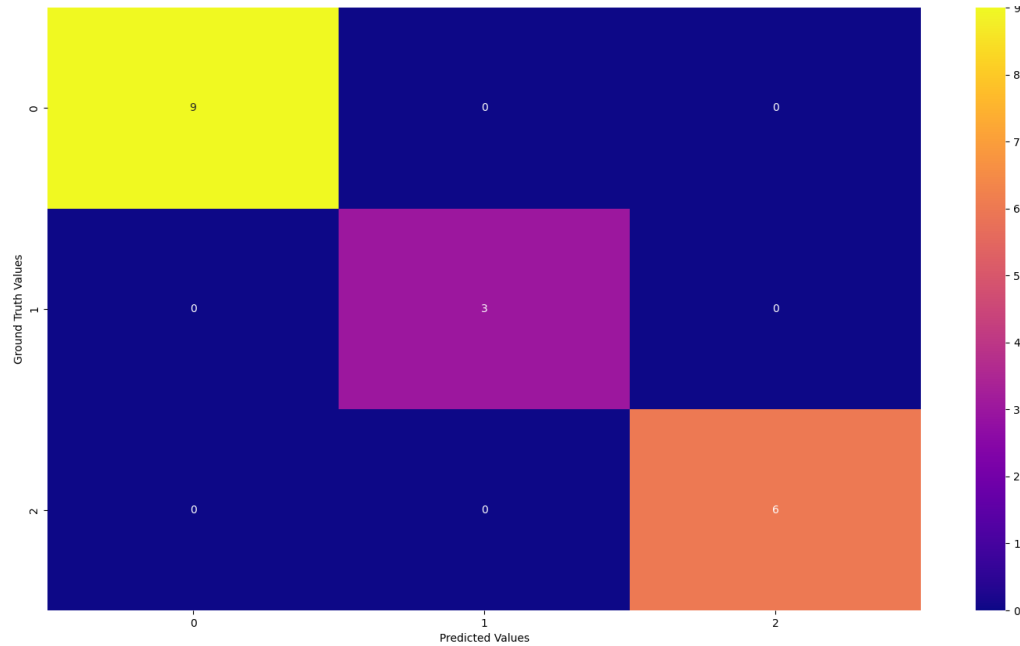


- (Left) The confusion matrix for TDA AI/ML EO trained model ATR predictive testing on 2 SUAS RE scenario at the first comparative data timestamp.
- (Right) The confusion matrix for TDA AI/ML EO trained model ATR predictive testing on 2 SUAS RE scenario at the second comparative data timestamp.
- Both matrices indicate perfect accuracy, precision, and recall.



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TDA AI/ML Pipeline EO Modality Results for 3 Ground Vehicles



- (Left) The confusion matrix for TDA AI/ML EO trained model ATR predictive testing on 3 ground vehicle RE scenario at the first comparative data timestamp.
- (Right) The confusion matrix for TDA AI/ML EO trained model ATR predictive testing on 3 ground vehicle RE scenario at the second comparative data timestamp.
- Both matrices indicate perfect accuracy, precision, and recall.

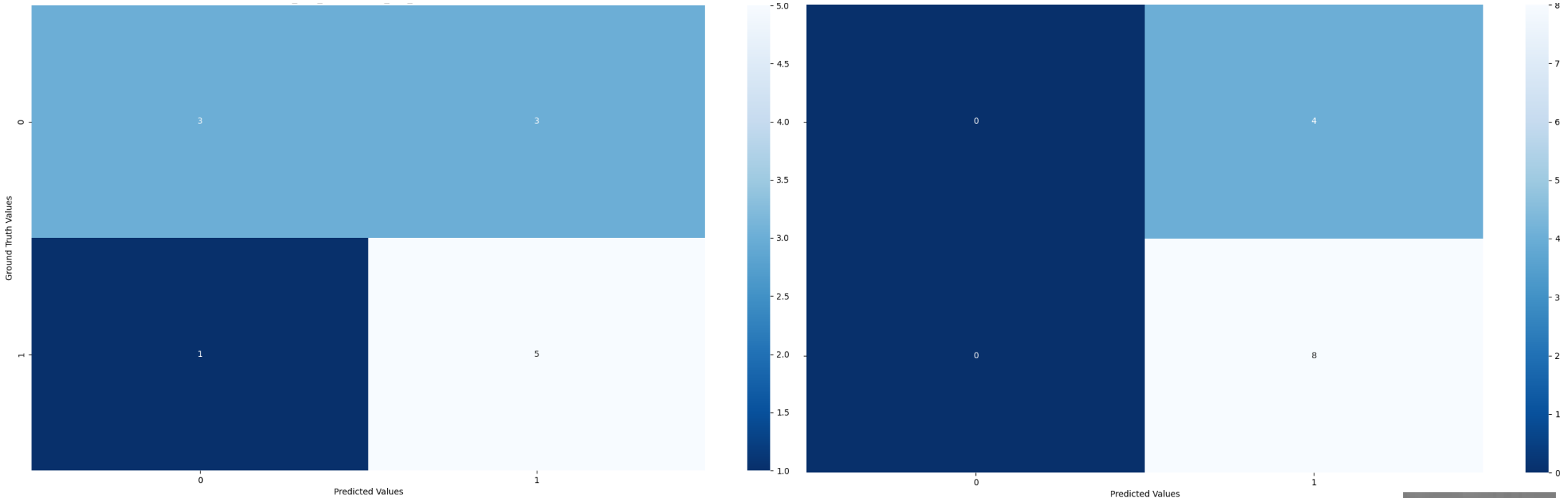


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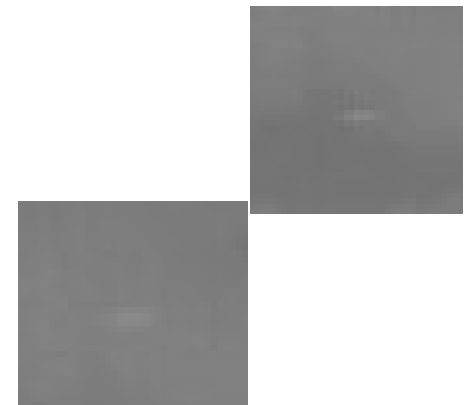


TDA AI/ML Pipeline IR Modality Results

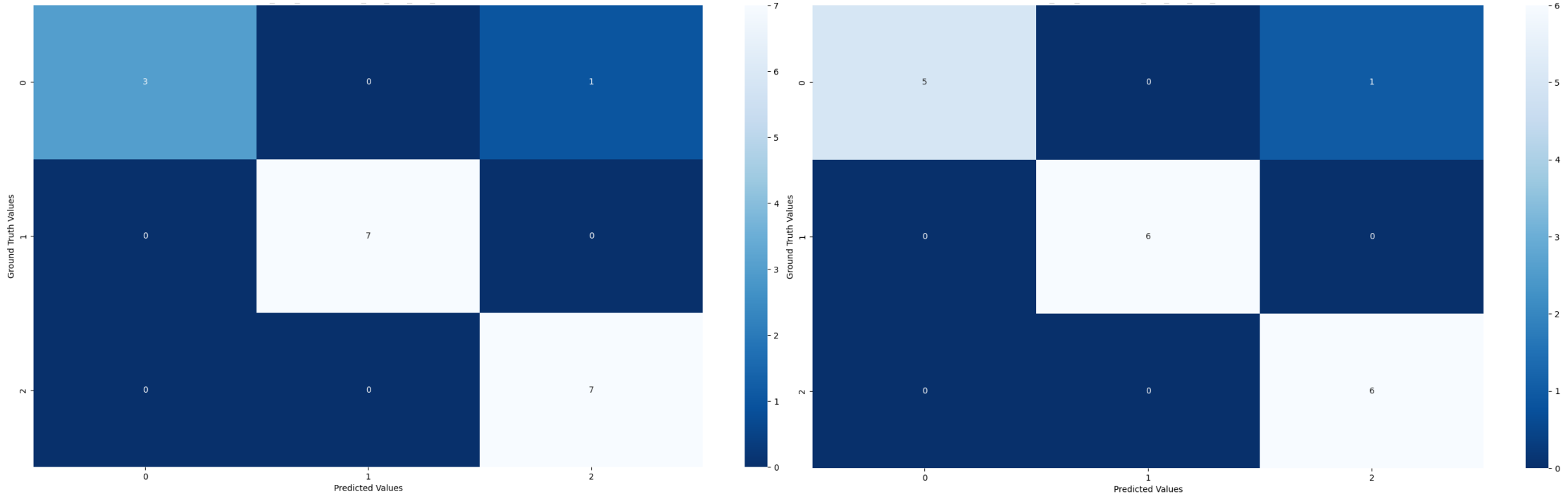
TDA AI/ML Pipeline IR Modality Results (2 SUASs)



- (Left) The confusion matrix for TDA AI/ML IR trained model ATR predictive testing on 2 SUAS RE scenario at the first comparative data timestamp.
- (Right) The confusion matrix for TDA AI/ML IR trained model ATR predictive testing on 2 SUAS RE scenario at the second comparative data timestamp.
- These matrices indicate accuracies of 66.7%, precision from 33.3-68.8%, and recall from 50-66.7% in identifying the two targets involved.



TDA AI/ML Pipeline IR Modality Results (3 Ground Vehicles)



- (Left) The confusion matrix for TDA AI/ML IR trained model ATR predictive testing on 3 ground vehicle RE scenario at the first comparative data timestamp.
- (Right) The confusion matrix for the TDA AI/ML IR trained model ATR predictive testing on 3 ground vehicle RE scenario at the second comparative data timestamp.
- The matrices indicate accuracies of 94.4%, precision from 95.2-95.8%, and recall from 91.7-94.4% in identifying the three targets involved.





Future Directions



Future Directions

- Data Fusion and Multimodal Sensor Data Applications for the TDA AI/ML Pipeline
- Intrinsic clustering capability for ATR of TDA AI/ML pipeline features and other associated TDA/PD metrics
- Generalizations of the TDA AI/ML pipeline using Sheaf Theory
- Intuitive user platform/dashboard designs of TDA AI/ML data for deployment in (near) real time SA analytics



Thank You!



Questions?
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