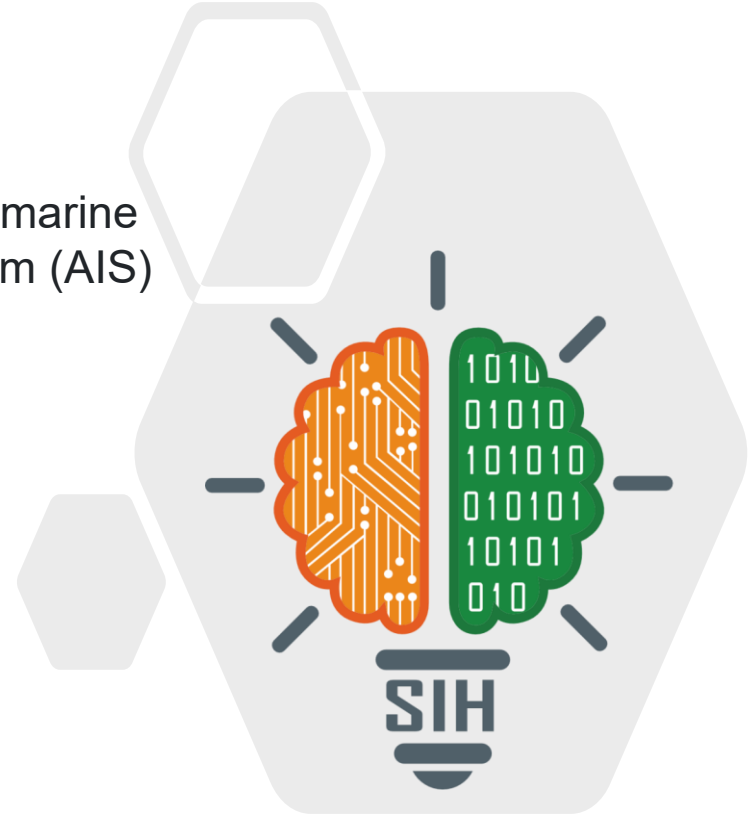


SMART INDIA HACKATHON 2024



- **Problem Statement ID** – SIH1655
- **Problem Statement Title**- Detecting oil spills at marine environment using Automatic Identification System (AIS) and satellite datasets.
- **Theme**- Smart Automation
- **PS Category**- Software
- **Team ID**- 19570
- **Team Name** - FrostSpire



Oil Spill Detection

IDEA/ SOLUTION

Implementing real-time oil spill detection in the marine environment using **Automatic Identification System (AIS)** and **satellite** datasets with the help of **machine learning**.

- **Machine Learning for Anomaly Detection:** Utilize DBSCAN clustering algorithm on AIS data to detect anomalies in vessel speed, course, and deviations, indicating potential distress or illegal activities.
- **Customized CNN for Oil Spill Detection:** Train a Convolutional Neural Network (CNN) on Synthetic Aperture Radar (SAR) images to accurately identify oil spills.
- **Real-Time Detection System:** Implement a real-time oil spill detection system by integrating AIS data and satellite datasets.

PROBLEM RESOLUTION

- **Early Anomaly Detection:** Monitor AIS data to detect unusual vessel behaviors indicating potential distress or oil leaks.
- **Accurate Spill Identification:** Use satellite imagery to confirm oil spills at the locations identified by AIS anomalies.

UNIQUE VALUE PROPOSITIONS (UVP)

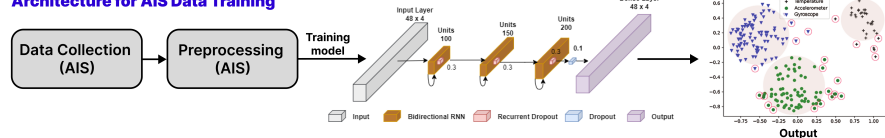
- **Comprehensive Detection Approach:** First-of-its-kind solution that fuses AIS data with satellite imagery using machine learning for oil spill detection.
- **Advanced Machine Learning Implementation:** Employs DBSCAN for anomaly detection in AIS data and a customized CNN for analyzing SAR images, enhancing detection precision.

TECHNOLOGIES USED

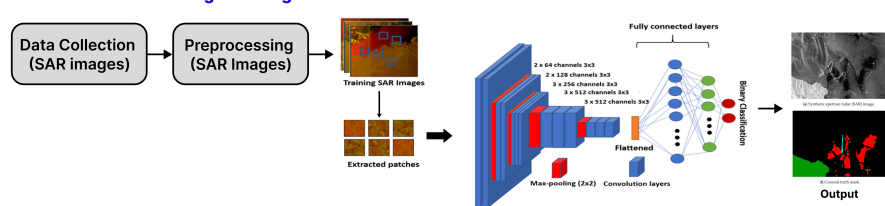
- **Programming Languages:** Python for data processing, machine learning, and image analysis.
- **Machine Learning Frameworks:** Scikit-learn for DBSCAN anomaly detection, TensorFlow/Keras for training the CNN model on SAR images.
- **Data Processing:** Pandas and NumPy for AIS data handling and preprocessing, OpenCV for image manipulation.
- **Frameworks & Tools:** Jupyter notebook, Google colab, SNAP tool for processing SAR datasets
- **Database:** MongoDB



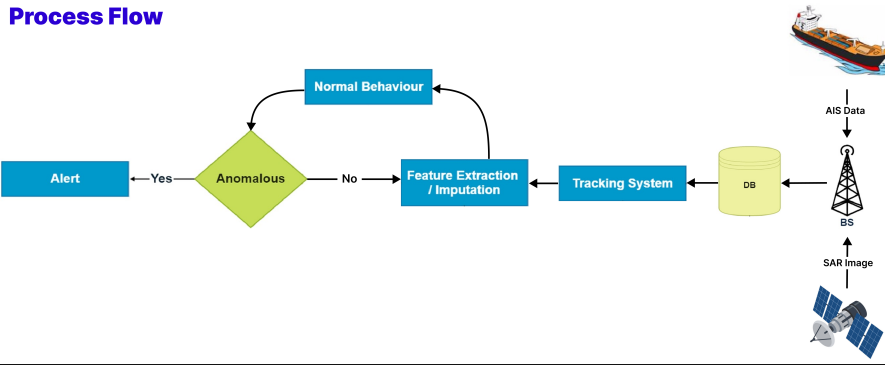
Architecture for AIS Data Training



Architecture for SAR Image Training



Process Flow



Feasibility And Viability

CHALLENGES AND RISKS:

1. Data Quality and Preprocessing:

- **Challenge:** AIS data may have missing, noisy, or incorrect values.
- **Risk:** Inaccurate data can lead to poor anomaly detection and model performance.

2. Model Accuracy and Generalization:

- **Challenge:** Risk of overfitting or underfitting, affecting anomaly detection in new data or different regions.
- **Risk:** Potential to miss anomalies or incorrectly classify normal behaviors.

3. Handling Complex Behaviors:

- **Challenge:** Vessel behavior is complex due to factors like weather, tides, and navigation rules.
- **Risk:** Misclassification due to not accounting for these contextual factors.

ANALYSIS OF THE FEASIBILITY OF THE IDEA

- **Proven Technology:** Machine learning algorithms (DBSCAN and CNN) have shown success in anomaly detection and image classification, making this approach technically feasible.
- **Data Availability:** Real-time AIS data is globally accessible, and satellite SAR imagery is available through various platforms, ensuring continuous data flow.

STRATEGIES FOR OVERCOMING CHALLENGES:

1. Efficient Data Handling:

- Use data streaming and real-time ingestion.
- Perform batch processing to manage computation load.

2. Model Optimization:

- Start with simpler models like Isolation Forest or DBSCAN.

POTENTIAL IMPACT ON THE TARGET AUDIENCE:

1. Maritime Authorities & Port Operators:

- **Impact:** Enhanced monitoring of vessel behaviors, improving safety and security. They can quickly identify deviations, suspicious speeds, or unusual patterns, allowing for faster interventions.

2. Shipping Companies:

- **Impact:** Improved fleet management through real-time vessel monitoring, detecting issues like unauthorized route deviations leading to optimized fuel usage and efficiency.

3. Environmental Agencies:

- **Impact:** Early detection of vessels in protected or restricted areas helps prevent environmental damage from illegal fishing, aiding the enforcement of environmental regulations.

4. Insurance Companies:

- **Impact:** Enables more accurate risk assessment by identifying behavior patterns linked to accidents or illegal activities, reducing fraudulent claims and enhancing vessel safety.

BENEFITS OF THE SOLUTION:

1. Social Benefits:

- Improves maritime security and public safety by detecting suspicious movements and preventing accidents.

2. Economic Benefits:

- Reduces costs for shipping companies through optimized routes and helps prevent illegal activities, minimizing economic losses.

3. Environmental Benefits:

- Protects marine ecosystems by preventing illegal activities and reducing pollution through adherence to approved routes.

4. Operational Benefits:

- Enhances fleet management with real-time tracking and allows proactive resolution of operational risks.

- Bui, N. A., Oh, Y., & Lee, I. (2024). Oil spill detection and classification through deep learning and tailored data augmentation. *International Journal of Applied Earth Observation and Geoinformation*, 129, 103845. <https://doi.org/10.1016/j.jag.2024.103845>
- Ribeiro, C.V., Paes, A. and de Oliveira, D., 2023. AIS-based maritime anomaly traffic detection: A review. *Expert Systems with Applications*, 231, p.120561.

- **Dataset Link**

AIS information: <https://marinecadastre.gov/accessais/>

(USA waters) <https://www.vesselfinder.com/>

(Indian Waters with subscription) <https://www.marinetraffic.com/en/ais/home/centerx:73.8/centery:13.7/zoom:8>
<https://www.aishub.net/>

Satellite datasets: <https://dataspace.copernicus.eu/> SNAP tool for processing SAR datasets
<https://step.esa.int/main>