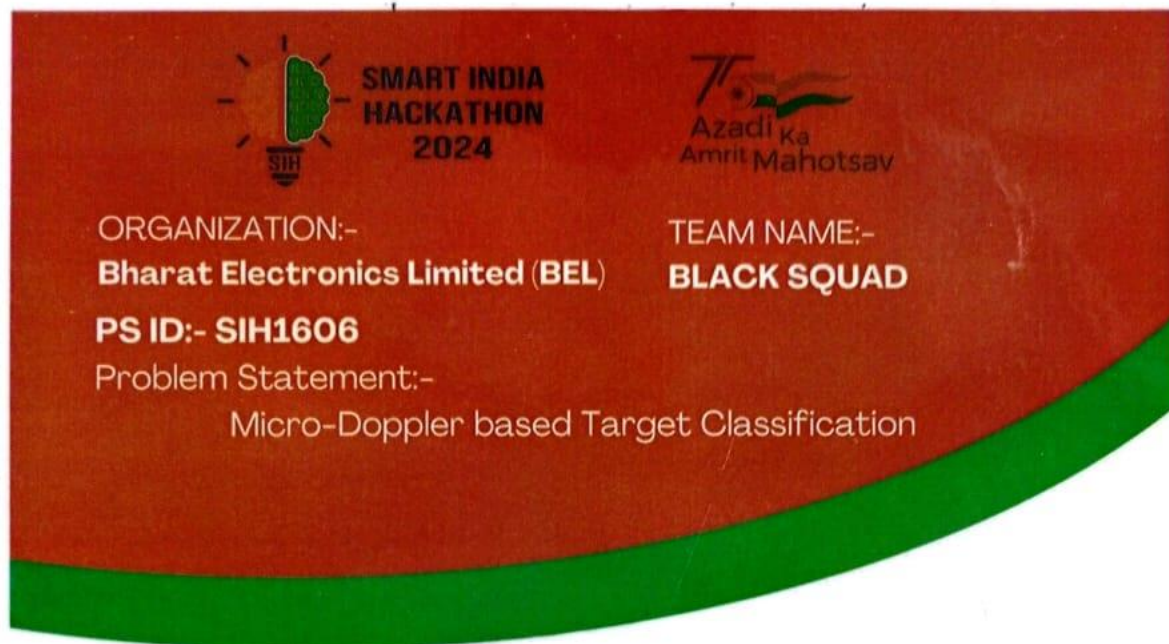


## Project Banner



### OBJECTIVE

To design and implement a radar-based system capable of classifying drones and birds using micro-Doppler signature analysis. The system aims to enhance situational awareness by reliably differentiating between drones and birds, leveraging the unique time-frequency features of their motion patterns for improved security and surveillance applications.

### PROPOSED SOLUTION

To design and implement a radar-based system capable of classifying drones and birds using micro-Doppler signature analysis. The system aims to enhance situational awareness by reliably differentiating between drones and birds, leveraging the unique time-frequency features of their motion patterns for improved security and surveillance applications.

### OUTCOME

The outcome is a system that classifies objects as either birds or drones using radar signals within milliseconds.

### BENEFACTOR

The beneficiaries are security agencies, surveillance operators, and airspace regulators.

### ASSISTED BY

**Mr. Rajesh Joseph Abraham,**  
Associate Professor,  
Indian Institute of Space Science  
and Technology  
**Mr. Harish Kumar,**  
Lead Data Scientist,  
NTT DATA Services



**Sri Eshwar**  
**College of Engineering**  
an autonomous institution  
affiliated to Anna University, Chennai



## MENTORING REVIEW - 1

Round - 1 Mentoring:

Time : 12:00 PM

Venkat Nagabhushanam Jetty : Multi-class Target Classification

Faculty coordinator/evaluator : Computational expensiveness?

Solution : we have trained in CPU,  
(Already provided) it requires CPU not any  
higher end devices.

Alumni Coordinator : Positive looking.

Work to be done :

→ Multiclass target classification.

---

Time : 7:14 PM

Round 1 - Evaluation

Work done :

→ For Multi-class target classification, we have done upto "Extract Spectrum" from the time domain data.

→ PM's input : Ideology from him about distance, direction of the target.

Inputs from Mentors:

→ Model for classification

→ Business prototype / End product (technical)

→ Scalogram images of human walk, human run

→ Architecture

→ Feature extraction / Distance, direction

## MENTORING REVIEW - 2

Time: 9:23 AM

Round 2 - Mentoring

Work done:

- Model documentation
- Completed & developed a efficient model
- Build architecture for end product
- Direction & Distance.

Time: 10.30 am

Round 2 - Evaluation

- Completed & developed a efficient model
- Completed the UI Part of the application
- Connected the ESP8266 with the model.

Work to be done:

- Notification feature.
- Chatbot Integration.

Time: 5:00 pm

work done:

- completed notification feature
- Completed chatbot integration
- Project completed.



# MODEL DOCUMENTATION

## Model Documentation

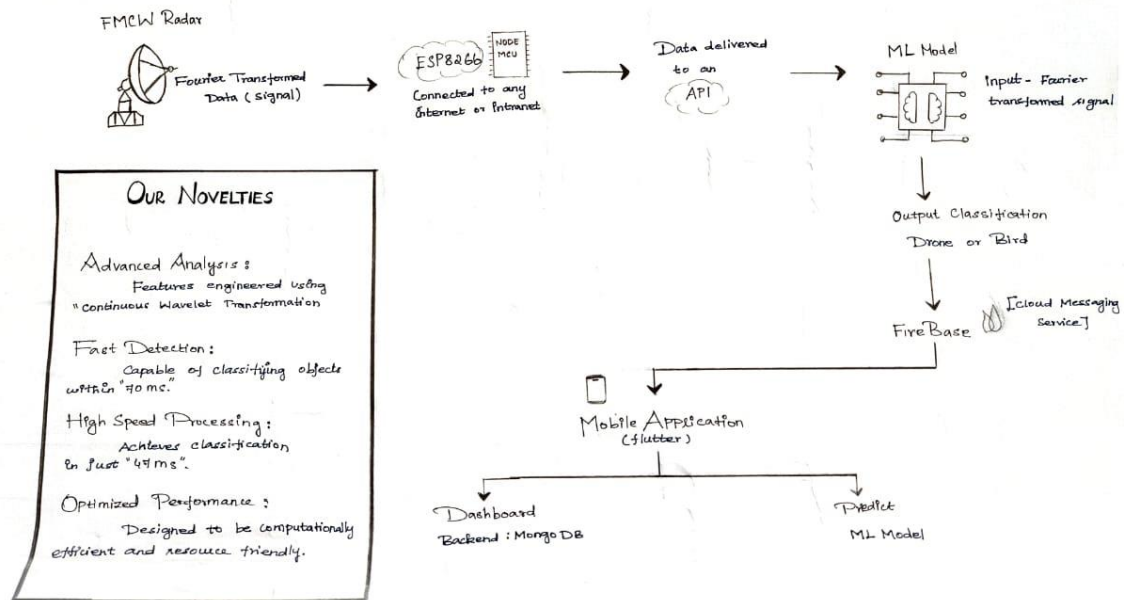
S. No	Model Name	Input	Transformation technique	Accuracy
1.	Serial Network	Spectrogram	STFT	85.36%
2.	ResNet50	Spectrogram	STFT	66.67%
3.	LSTM-ALRO	Spectrogram	STFT	91.30%
4.	Random Forest	PCA-transformed, flattened spectrogram images	STFT	92.00%
5.	Gated Recurrent Unit	Spectrogram	STFT	66.67%
6.	Shallow CNN	SNR normalized FFT matrix	FFT	68%
7.	Resnet 50	Entire scan segment → scalogram	CWT	85%
8.	Resnet 50	Center ranges cell → scalogram	CWT	76.92% Biased overfitting
9.	Resnet 50 (callback func)	Center range cells → scalogram	CWT	99.9% Biased over drone.
10.	Resnet 50 (stratified k-fold)	Entire scan → scalogram	CWT	100% Biased over bird.

## Conclusion : (feature extracted models)

11. Random forest	features	CWT	91.88 %
12. Adaboost	features	CWT	85.81 %
13. XGBoost	features	CWT	91.47 %
14. SVM	features	CWT	63.14 %
15. Decision Tree	features	CWT	89.04 %

## Architecture of End Product

### Architecture Of End Product.



# Architecture of Model

