

# ***UAV Velocity Prediction***

## ***Using Audio data***

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# AGENDA

01

## Introduction

Background  
Motivation

02

## Methodology

Dataset  
Models

03

## Result

MFCC Result  
Accuracy result

04

## Conclusion

Summary

# Introduction

Background  
Motivation

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# Problem Statement

**Drone crashes into Russian oil refinery in possible attack [1]**

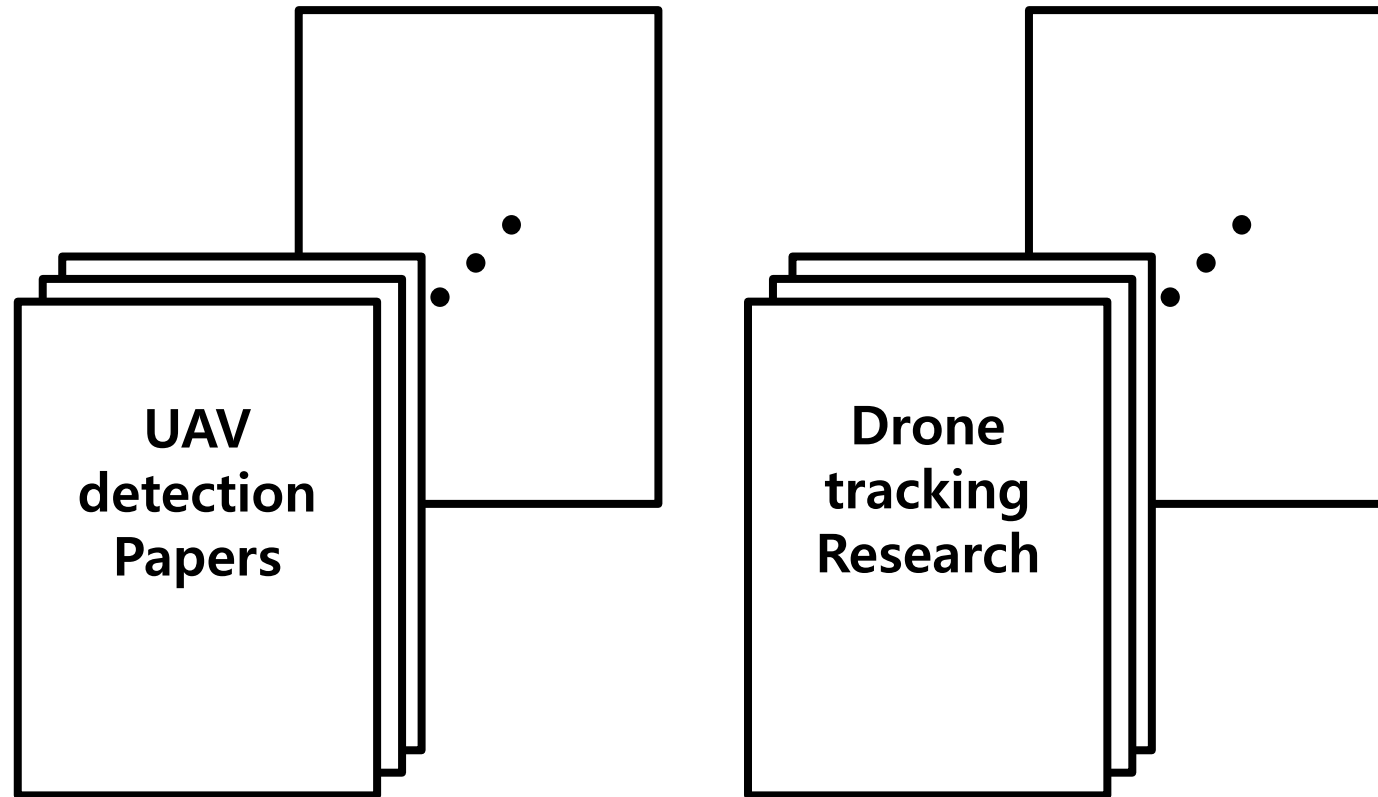


**FBI says PA electricity station likely 'target' of drone incident [2]**



# *Problem Statement*

**How can we protect ourselves from the malicious UAVs?**



# *Motivation*

Our goal is

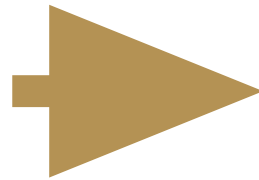


Our goal is

UAV Velocity Prediction  
Using Audio data

## How to Experiment

The U.S. FAA set  
UAV speed limit  
**100mph.**



**100mph over speed**  
**UAV** is  
a high probability that  
**malicious UAV**



# Motivation

Why Audio data? [3], [4], [5]



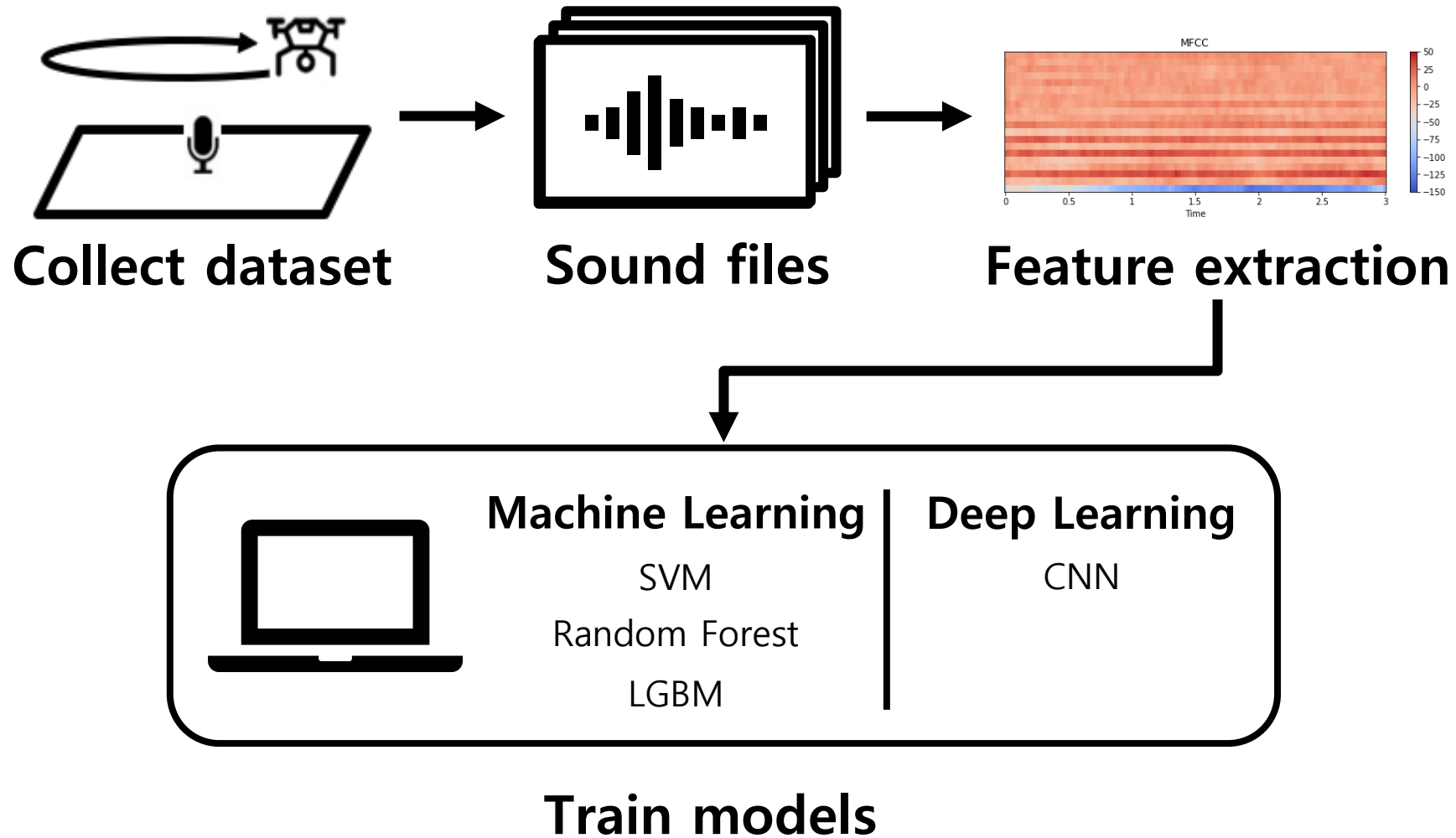
- Audio data obtain relatively results at **less cost** than other methods.
- Even with noise limitations, it provides **good results** for distinguishing the drone's sound.

# Methodology

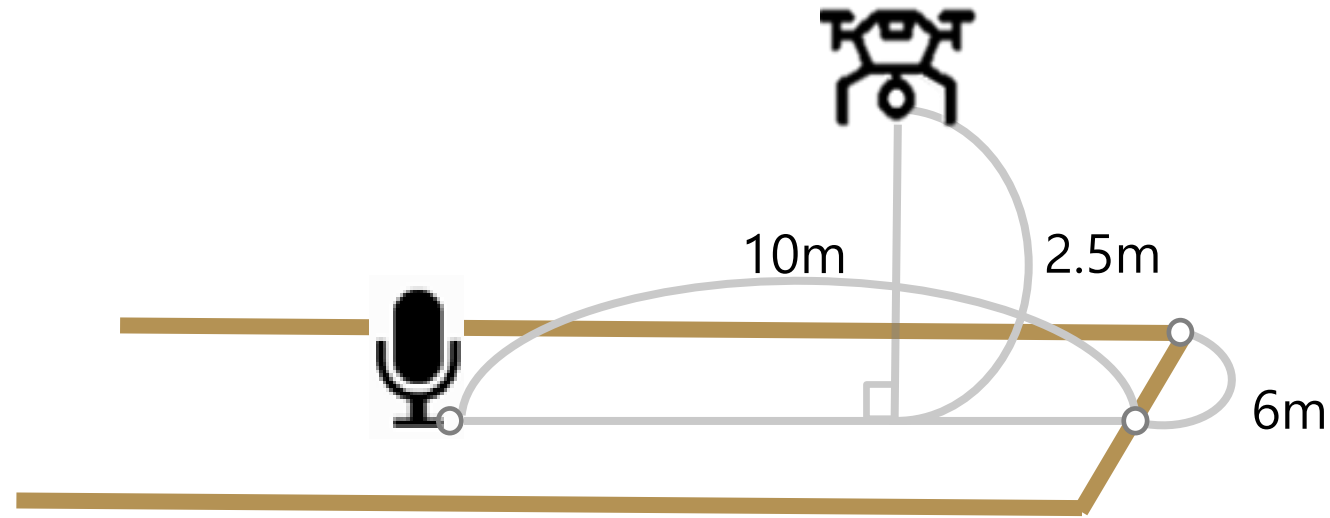
Dataset  
Models

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# Overview

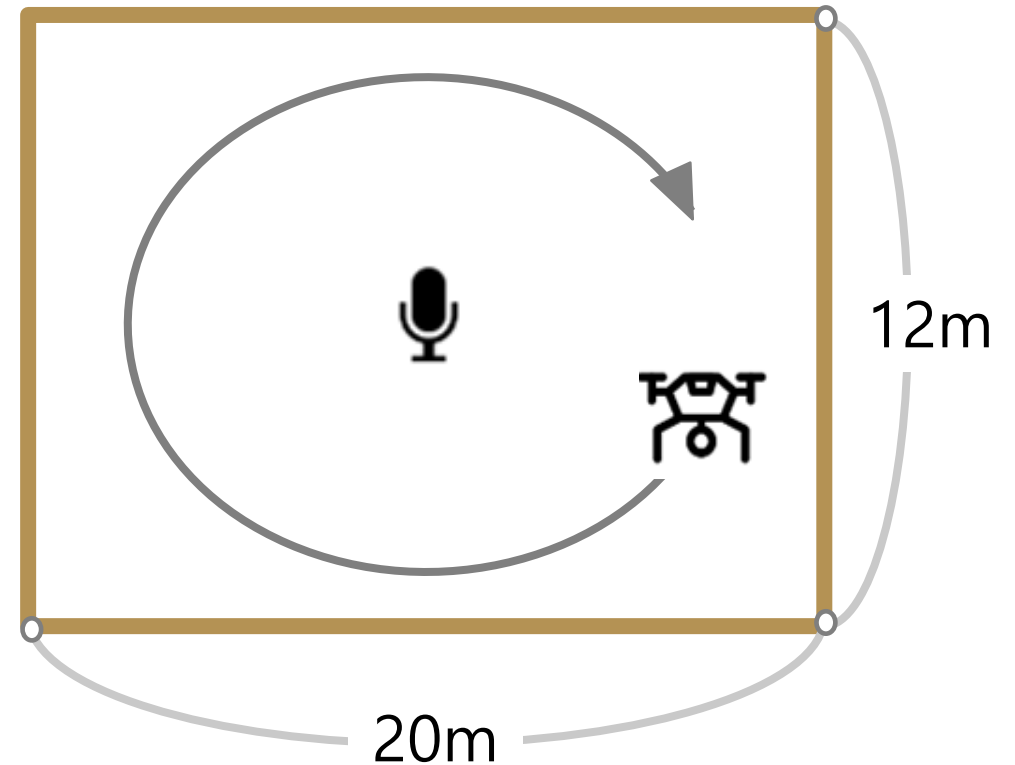
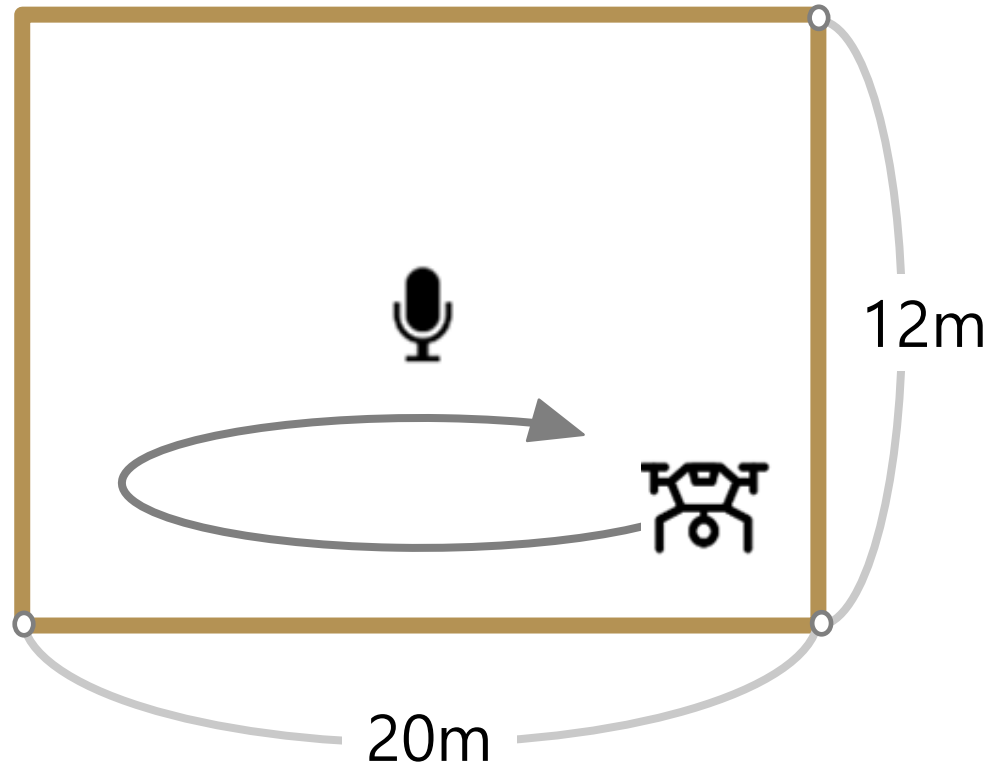


# Dataset



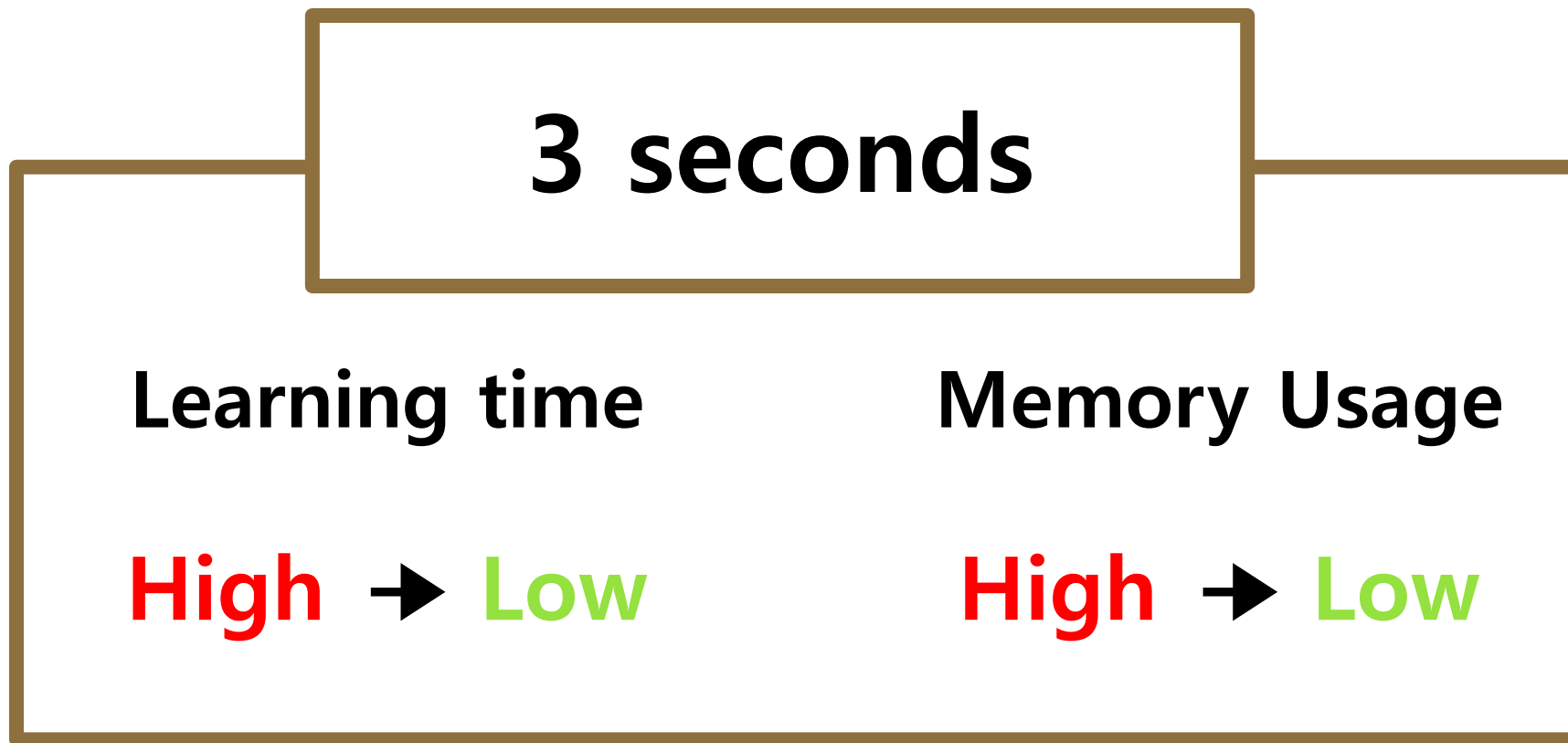
Microphone	Speed gun	Place
Dell XPS15 9570 SAMSUNG Ion 2020 NT950XCR-G58A	Bushnell Velocity Speed Gun (Accuracy: +/- 1 mph)	Indoor lab

## Methods for Collecting Drone Data



# Dataset

Change the length of the dataset



## How many collect Dataset?

Speed Type	Fast	Slow	Ratio
Train Data	776	639	80%
Validation Data	261	210	10%
Test Data	261	210	10%

# Dataset

A Feature Engineering Focused System for Acoustic UAV Detection [6]	
Feature	Accuracy average (SVM, GNB, KNN, NN)
Chroma_stft	0.878
Mel	0.831
<b>MFCC</b>	<b>0.994</b>
Contrast	0.854
tonnetz	0.731



# Models

**Machine  
Learning**

**SVM,  
Random Forest,  
LightGBM**



**Deep  
Learning**

**CNN**

## Layer Description

	Layer	In Channels	Kernel	Padding	Stride	Activate
Layer1	Conv1D	20	5	2	1	ReLU
Layer2	Conv1D	11	5	2	1	ReLU
Layer3	Conv1D	5	5	2	1	Sigmoid

# Result

## MFCC Result Result Graph

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# MFCC Result

Velocity	Feature 0	Feature 1	Feature 2	Feature 3	Feature 4
Slow	-48.16	-25.84	19.66	0.91	-17.52
Fast	-55.64	-11.42	25.14	-3.21	-23.56
(time: 3sec)	<b>Feature 5</b>	<b>Feature 6</b>	<b>Feature 7</b>	<b>Feature 8</b>	<b>Feature 9</b>
	21.85	-10.59	20.11	-15.26	9.50
	22.03	-14.79	14.18	-18.04	7.65
	<b>Feature 10</b>	<b>Feature 11</b>	<b>Feature 12</b>	<b>Feature 13</b>	<b>Feature 14</b>
	-0.63	-9.87	7.72	-1.24	-1.62
	-3.59	-6.54	1.83	-1.03	-5.80
	<b>Feature 15</b>	<b>Feature 16</b>	<b>Feature 17</b>	<b>Feature 18</b>	<b>Feature 19</b>
	0.52	-4.19	-6.14	-3.47	1.66
	-0.25	-5.22	-2.39	-3.67	0.12

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# Result Graph

## Machine Learning and Deep Learning Result

Model	Accuracy	Precision	Recall	F-1 Score
SVM	0.987	0.977	1.000	0.988
Random Forest	0.997	0.996	1.000	0.998
LGBM	0.995	0.992	1.000	0.996
CNN	1.000	1.000	1.000	1.000

# Result Graph

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# Conclusion

Summary  
Future Work

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# Summary

- **Why** we decided to do this project?
  - Drone Strike, Kamikaze attack
- How to **solve** the problem?
  - Predicting the velocity of the UAVs
- What does result **means**?
  - Possibility of prediction of UAV velocity

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***Thank you for listening***

**Q&A**