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Sri Adichunchanagiri Shikshana Trust ®



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An Autonomous Institution, Affiliated to Visvesvaraya Technological University, Belagavi,

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## Department of Computer Science & Engineering



### Project Phase-2 Presentation on “Deep Learning Framework to detect Cyber- Attacks in IoT Network”

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# ABSTRACT

- \* The Internet of Things(IoT) is a network of interrelated devices that connect and exchange data with other IoT devices and the cloud.
- \* These devices represent a vast attack surface for cyber-attacks.
- \* With the increase in the IoT devices being connected to the internet the chances of the cyber-attacks have increased .
- \* Hence the KDD network dataset was used to train the models to detect and classify the attacks.

# INTRODUCTION

- \* The Internet of Things (IoT) is a paradigm that refers to the interconnected network of physical devices and other objects embedded with sensors.
- \* The connection of IoT devices to the internet has lead to the lack of security as the systems get attacked and the data is lost.
- \* The ‘Generative Deep Learning to Detect Cyber attacks for the IoT-23 Dataset’ paper has been referred.
- \* The deep learning methods are effectively used in the detection of the attacks in the IoT devices.

# LITERATURE SURVEY

Slno.	AUTHOR	METHOD	ATTACKS ADDRESSED	DATASET	RESULT
1	Shire	Angle-Based Outlier Detection , Principal Component Analysis, Minimum Covariance Determinant , ARIES GAN.	Man In The Middle attacks,Ping Distributed DoS attacks,	Modbus/TCP network flows,DNP3 network flows, operational data	The accuracy of the proposed mechanism reaches 91%.
2	Saharkhizan	Modbus IoT environments, RNN Long-Short-Term-Memory, MENSA model.	TCP SYN DoS attacks, Modbus query flood attacks.	Simoes, CICFlowMeter	Compared to the existing anomaly-based IDS, MENSA addresses efficiently the FP.
3	Shafi	Virtual Honeypot Devices, Markov models,BlackBox IDS	DoS attack,privacy,eavesdropping,backdoor attacks.	IoT testbed	Detects both known and unknown attacks with high detection rates & low false-positive alerts

# LITERATURE SURVEY Contd..

4	Shire	DL-based detection method, State Vector Estimator, DL-Based Identification , MENSA	False Data Injection, DoS attacks.	Modbus/TCP,DNP3	MENSA achieves the best performance either for detecting operational anomalies or discriminating the Modbus/TCP and DNP3 cyberattacks
5	Imtiaz Ullah	Transfer learning, RNN, CNN	Replay, Flooding and DoS attacks	BoT-IoT, IoT Network Intrusion, MQTT-IoT-IDS2020,	Proposed binary and multiclass classification models showed high accuracy, precision, recall, and F1 score

# PROBLEM STATEMENT



- \* The aim of our project is to design a framework to detect cyber attacks using Deep-Learning models.

# OBJECTIVES

- ❖ To implement an efficient network intrusion detection system.
- ❖ To implement a multi-class classification to identify different types of attacks.
- ❖ To design an intelligent intrusion detection and prevention system.
- ❖ To compare various ML algorithms to detect intrusion detection system.

# REQUIREMENTS ANALYSIS

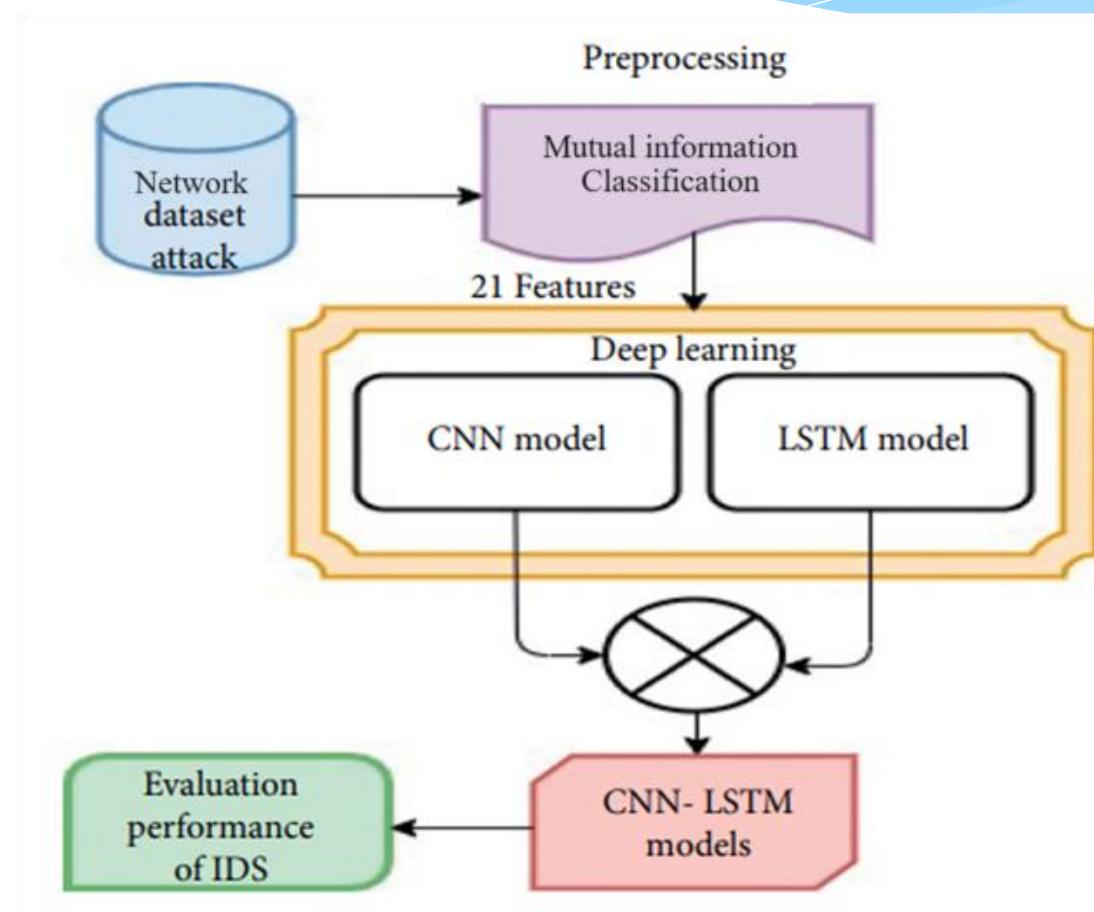
## Hardware Requirements

- \* Processor Type : Intel CORE i5
- \* Speed : 2.4 GHz
- \* RAM : 8GB
- \* Hard Disk HDD : 80GB

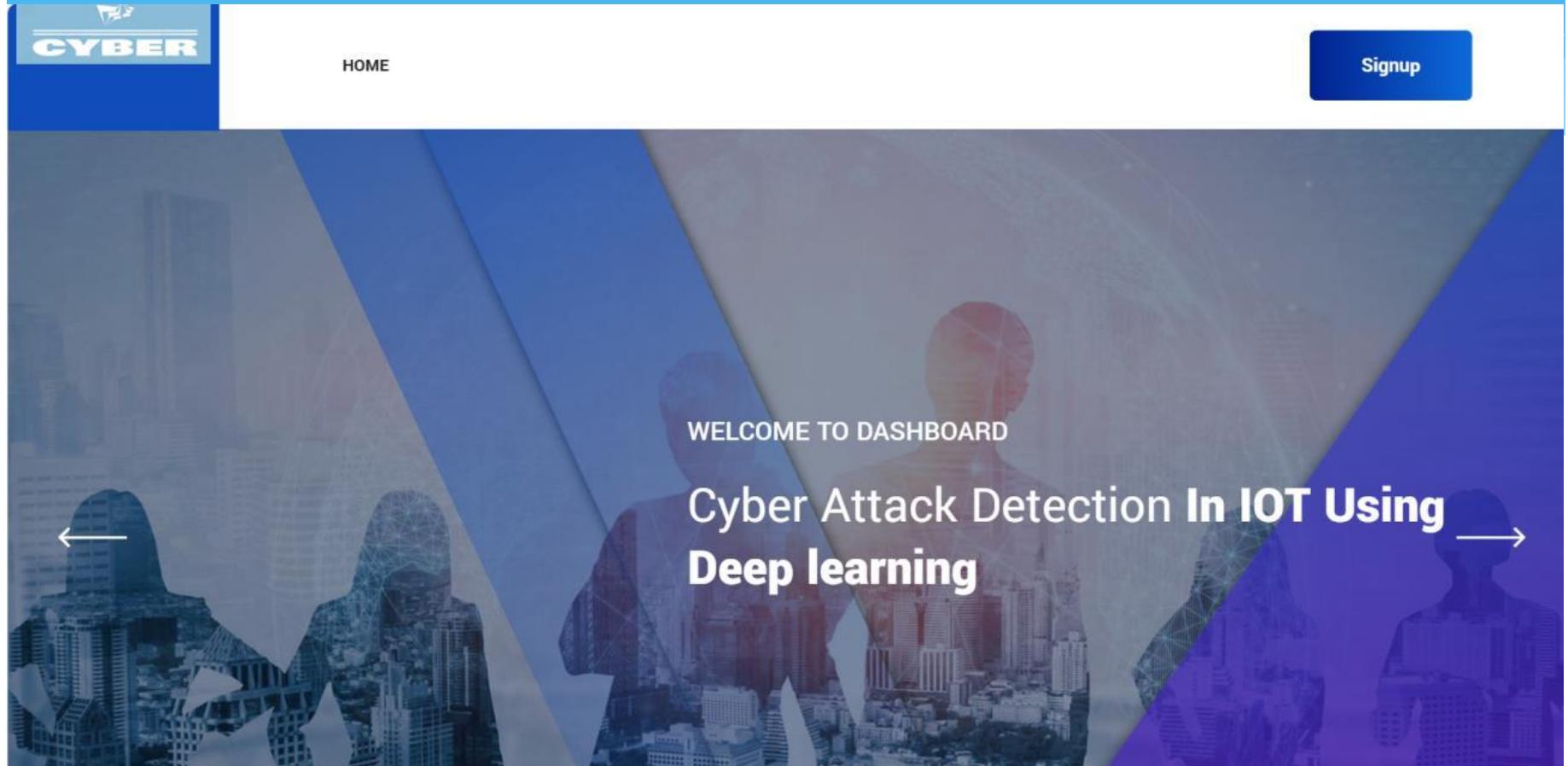
## Software Requirements

- \* Operating System : Windows 64-bit
- \* Technology : Python
- \* IDE : PythonIDLE
- \* Tools : Anaconda
- \* Python Version : Python 3.6

# IMPLEMENTATION



# RESULTS



Home Page



## ADD ACCOUNT?

 Robert

 .....

LOGIN

Register here! [Sign Up](#)



**Login Page**



-1.69431

dst\_host\_same\_srv\_rate

0.599396

dst\_host\_diff\_srv\_rate

-0.28287

dst\_host\_same\_src\_port\_rate

-1.02208

dst\_host\_srv\_diff\_host\_rate

-0.15863

Predict

## Prediction Page



HOME NOTEBOOK ▾ ABOUT

Signout



Result: **Attack is Detected and its DOS Attack!**



**Result Page showing name of attack detected**

	<b>ML Model</b>	<b>Accuracy</b>	<b>Precision</b>	<b>Recall</b>	<b>F1-Score</b>
0	CNN	0.989	0.994	0.989	0.991
1	RNN	0.793	1.000	0.793	0.885
2	RBM	0.991	0.993	0.991	0.992
3	DNN	0.994	0.994	0.994	0.994
4	CNN LSTM	1.000	0.993	0.990	0.992

## Results of KDDCUP

# CONCLUSION

- \* This project has shown that for a limited set of attacks and IoT devices, it is possible to use generative deep learning methods like mutual information classification and CNN-LSTM to classify attacks with high accuracy.
- \* We have also used a recent dataset called KDD network attack dataset and implemented baseline models, mutual information classification and CNN-LSTM.
- \* We also tried randomizing the test set in a way that we can inject new information and the model was able to consider it as an anomaly.
- \* Our results show that the LSTM based models are more effective at identifying attacks and classifying them.

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# **THANK YOU**