



||Jai Sri Gurudev||  
Sri Adichunchanagiri Shikshana Trust ®

# SJB INSTITUTE OF TECHNOLOGY

No. 67, BGS Health & Education City, Dr. Vishnuvardhan Road, Kengeri, Bengaluru - 560 060

Approved by AICTE - New Delhi.

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

Accredited by NAAC A+, Accredited by NBA. Certified by ISO 9001-2015

**Department of Computer Science & Engineering**



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

*Guide Name*

***Mrs. Manjula H S***

***Assistant Professor***

***Dept of CSE, SJBIT***

*Presented by :*

Nischitha B [1JB20CS073]

Pallavi R [1JB20CS076]

Preethi V [1JB20CS084]

Rakshith N Magi [1JB20CS092]

# TABLE OF CONTENTS

- \* Abstract
- \* Introduction
- \* Literature Survey
- \* Problem Statement
- \* Requirements Analysis
- \* Design/Flow
- \* Implementation
- \* Results
- \* Conclusion
- \* References

# 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.	Simoese, 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,DN P3	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

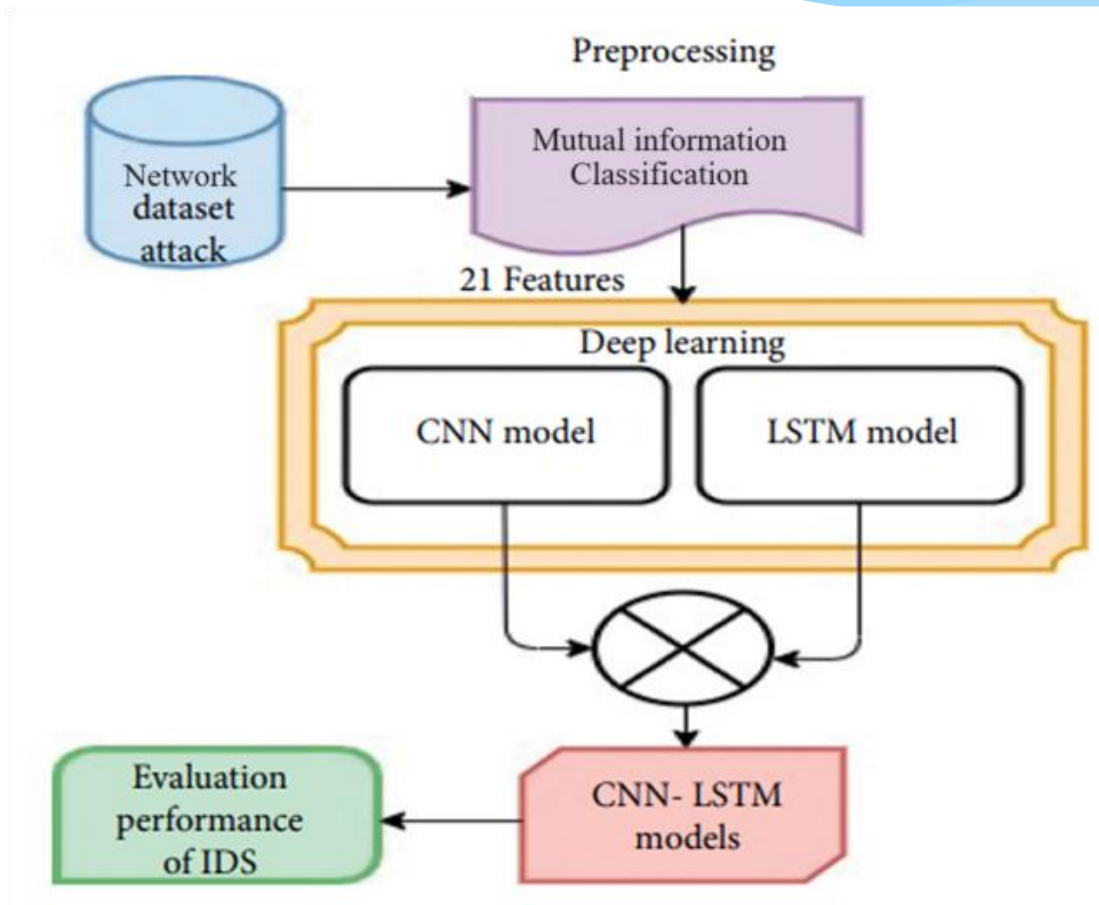
## Hardware Requirements

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

## Software Requirements

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

# IMPLEMENTATION



# RESULTS



HOME

Signup

WELCOME TO DASHBOARD

Cyber Attack Detection In IOT Using  
Deep learning

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>
<b>0</b>	CNN	0.989	0.994	0.989	0.991
<b>1</b>	RNN	0.793	1.000	0.793	0.885
<b>2</b>	RBM	0.991	0.993	0.991	0.992
<b>3</b>	DNN	0.994	0.994	0.994	0.994
<b>4</b>	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.



# REFERENCES

- [1] B. Alqahtani and B. AlNajrani, “A study of Internet of Things protocols and communication,” in Proc. 2nd Int. Conf. Comput. Inf. Sci. (ICCIS), Oct. 2020, pp. 1–6 doi: 10.1109/ICCIS49240.2020.9257652.
- [2] R. Shire, S. Shiaeles, K. Bendiab, B. Ghita, and N. Kolokotronis, “Malware squid: A novel IoT malware traffic analysis framework using convolutional neural network and binary visualization,” in Internet of Things, Smart Spaces, and Next Generation Networks and Systems. Cham, Switzerland:Springer, 2019, pp. 65–76.
- [3] C. Kolias, G. Kambourakis, A. Stavrou, and J. Voas, “DDoS in the IoT: Mirai and other Botnets,” Computer, vol. 50, no. 7, pp. 80–84, 2020, doi: 10.1109/MC.2020.201.

# REFERENCES Contd..

- [4] S. Kwon, H. Yoo, and T. Shon, “IEEE 1815.1-based power system security with bidirectional RNN-based network anomalous attack detection for cyber-physical system,” *IEEE Access*, vol. 8, pp.77572–77586, 2020.
- [5] F. Bonomi, R. Milito, J. Zhu, and S. Addepalli, “Fog computing and its role in the Internet of Things,” in *Proc. 1st Ed. MCC Workshop Mobile cloud Comput. (MCC)*, 2012, p. 13.
- [11] R. Ahmad and I. Alsmadi, “Machine learning approaches to IoT security: A systematic literature review,” *Internet Things*, vol. 14, Jun. 2021, Art. no. 100365, doi: 10.1016/j.iot.2021.100365.



**THANK YOU**