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# Deep Learning-Based DDoS-Attack Detection for Cyber-Physical System Over 5G Network

## A Deep Learning Approach to 5G Cybersecurity

Department: School of Cyber Security and Digital Forensics

Program: M.Tech Artificial Intelligence & Data Science

(Specialization in Cyber Security Specialization)

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Subject Code: CTMTAIDS SII P1

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CYBER SECURITY  
INTELLIGENCE

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# Paper Highlights



- **Authors:** B. Hussain, Q. Du, B. Sun, Z. Han
- **Journal:** IEEE Transactions on Industrial Informatics, Feb 2021
- **Goal:** Efficient DDoS detection using Deep Learning (CNN & BLSTM)
- **Key Contribution:** Real-time detection for CPS in 5G
- **DOI:** [10.1109/TII.2020.2974520](https://doi.org/10.1109/TII.2020.2974520)



# Problem Statement:

## DDoS Detection in 5G Network Slices



- 5G connects billions of devices, increasing the attack surface.
  - Network slicing creates isolated yet interdependent segments.
  - Real-time applications need ultra-fast protection.
  - Traditional DDoS detection is too slow and rule-based.
  - Encrypted and dynamic traffic bypasses static filters.
  - CNNs offer fast, adaptive, lightweight, and accurate detection.
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# Objectives

- Design and implement a **CNN-based model** to **detect DDoS attacks** in **5G networks**.
- Focus on **protecting network slices** in **Cyber–Physical Systems (CPS)**.
- Train a **binary classifier** to distinguish between benign and **malicious traffic**.
- Achieve high accuracy, precision, and **low false positives**.
- Integrate the solution into **Software Defined Networking (SDN)** and **Network Function Virtualization (NFV)** environments.

# Introduction - DDoS

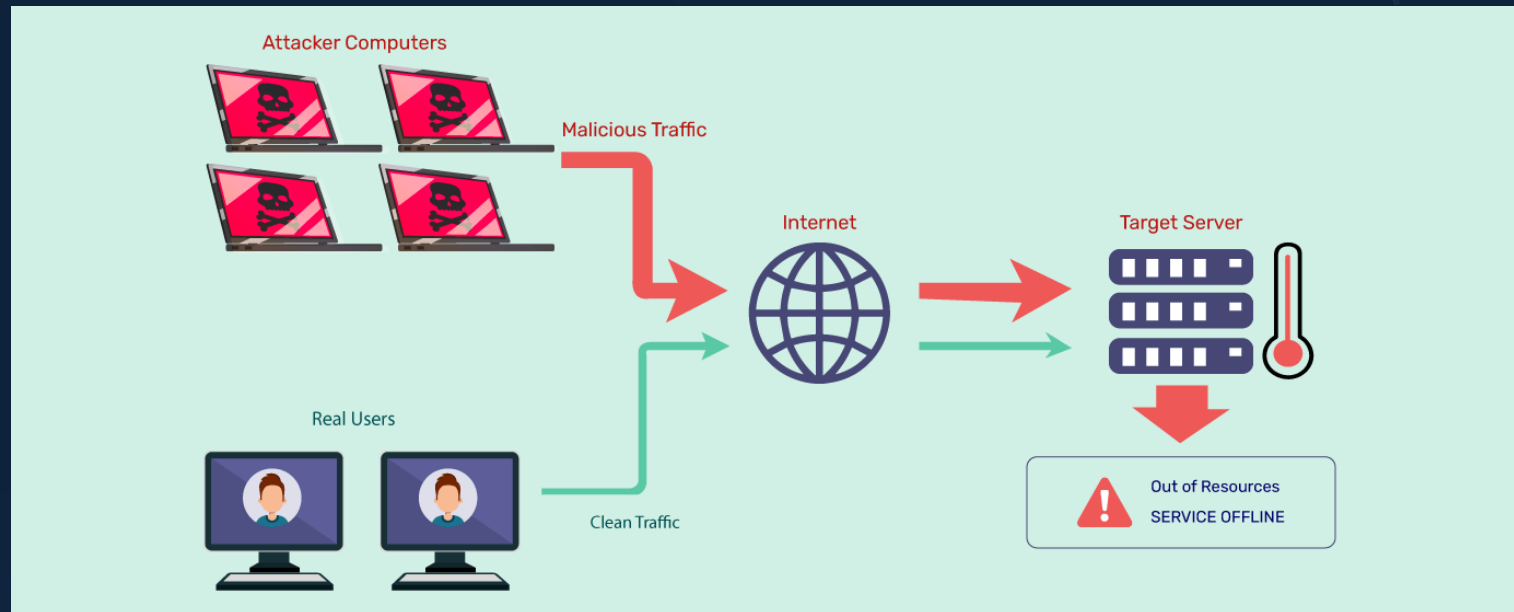


## What is a DDoS Attack?

- DDoS (Distributed Denial-of-Service) attacks **overwhelm a system or network** with massive traffic.
- Aim to make a **service or network resource unavailable** to legitimate users.
- Widely used in **cyber warfare, botnets, and extortion campaigns**.

## Real-World Consequences:

- Service Outages
- Financial Losses
- Data Breach Risk
- Customer Trust Damage
- Degraded Performance



# Introduction – 5G



## Why is 5G Vulnerable?

- 5G offers **ultra-reliable, low-latency communication** and **massive device connectivity**.
- **Network slicing** enables customized virtual networks—but also increases the **attack surface**.
- The real-time nature of 5G requires **instant threat detection and response**.

## Why 5G ?

- More Speed, More Devices, More Risk
- Massive Attack Surface
- Network Slicing Makes It Worse
- Edge Computing & AI Need Real-Time Security



# Introduction - CPS

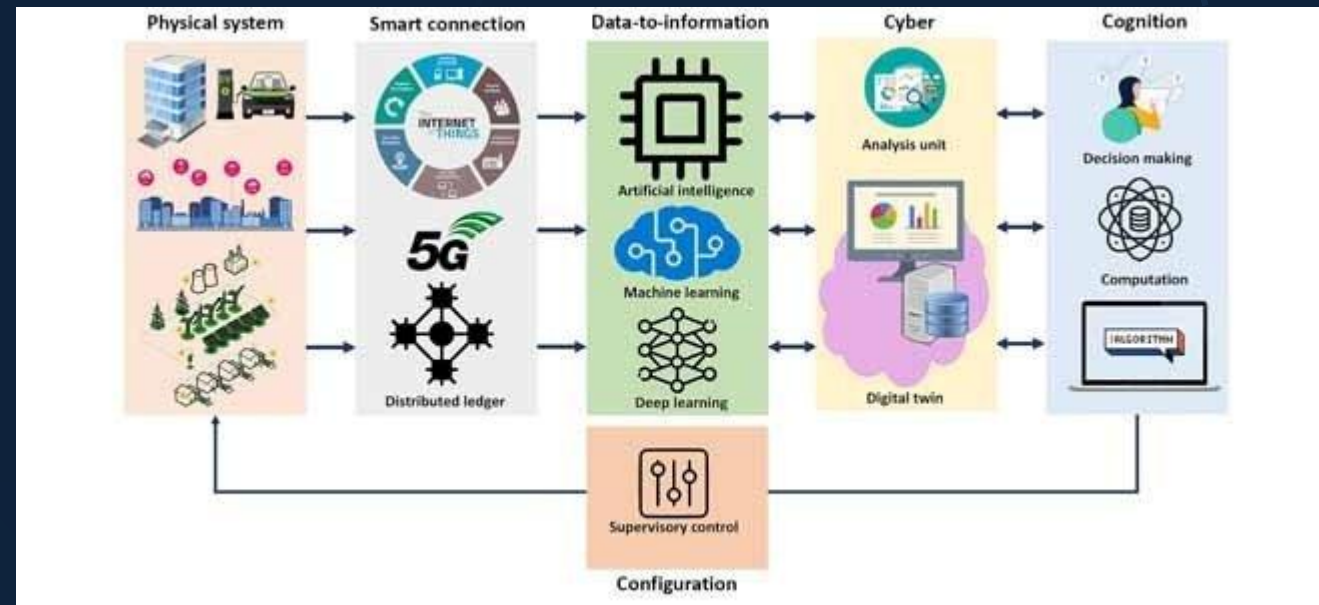


## What is Cyber Physical System?

- Integrations of **computation, networking, and physical processes..**
- 5G enables CPS to operate in real time with ultra-low latency and high reliability.
- Common in smart grids, healthcare, industrial IoT, autonomous vehicles, and critical infrastructure..

## Importance of CPS?

- Critical Infrastructure Backbone
- Healthcare & Life-Critical Systems
- Autonomous & Smart Transportation
- Real-Time Operation Requirements
- Target of Cyber Attacks





# Introduction - CNN

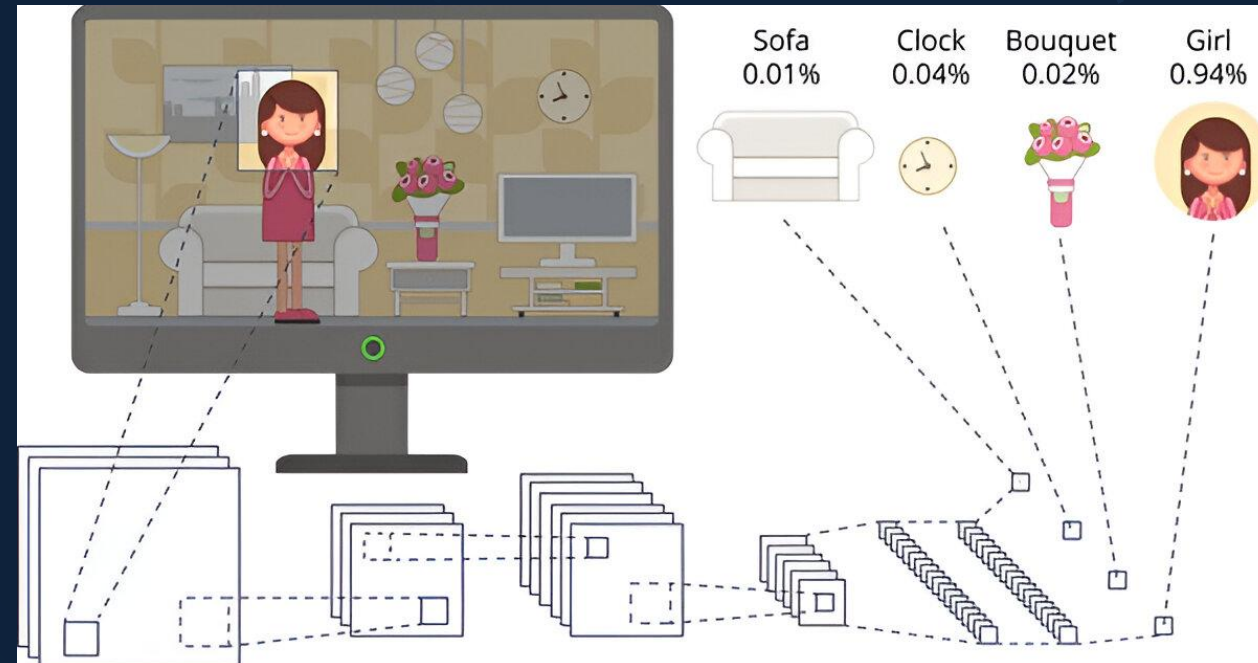


## 🧠 Why Use Deep Learning (CNN)?

- Traditional ML techniques struggle with **dynamic patterns** in 5G data.
- CNNs can automatically extract **spatial and temporal features** from network traffic.
- Achieve **higher accuracy, lower false positives**, and better **generalization**.

## 🧠 What is CNN?






A Convolutional Neural Network (CNN) is a type of deep learning model designed to automatically extract patterns from data originally for images, but now used widely for network traffic, time-series, and anomaly detection too.







# Key Components of CNN:

Layer	Function
 <b>Convolution Layer</b>	Extracts features using filters (kernels) from the input data.
 <b>Activation (ReLU)</b>	Introduces non-linearity (e.g., detects complex attack patterns).
 <b>Pooling Layer</b>	Reduces feature size, improves performance, and focuses on key info.
 <b>Flatten Layer</b>	Converts feature maps into a 1D vector for prediction.
 <b>Fully Connected Layer</b>	Final decision-making (e.g., DDoS = 0, Benign = 1)

# CNN Architecture



## 1. Input Layer:

- Data: Features extracted from network traffic (e.g., flow duration, packet lengths, protocols).
- Shape: (number of samples, number of features, 1)

## 2. Convolution Layers:

- Apply filters (kernels) to extract spatial features from the input.
- Purpose: Detect patterns like traffic spikes or irregularities, potentially indicating DDoS attacks.

## 3. Activation Function (ReLU):

- Introduces non-linearity to detect complex patterns that simpler models may miss.

## 4. Pooling Layer (Optional):

- Reduces spatial dimensions, keeping only essential features.
- Benefit: Improves performance and reduces overfitting.

# CNN Architecture

## 5. Flattening Layer:

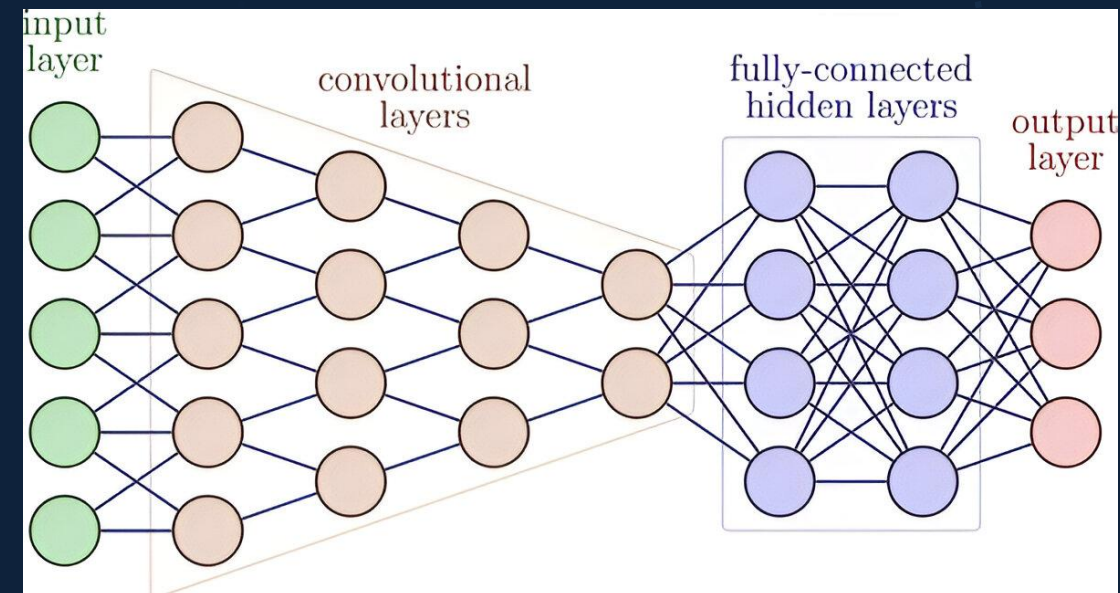
- Converts 2D features (from convolution and pooling) into a 1D vector for classification.

## 6. Fully Connected Layer:

- Neurons in this layer are connected to all the previous layer's neurons.
- **Purpose: Final classification (DDoS or benign traffic).**

## 7. Output Layer:

- Sigmoid activation function for binary classification:
  - 0 = Attack
  - 1 = Benign



# LSTM Architecture

An advanced Recurrent Neural Network (RNN) variant, excel at capturing long-term dependencies in sequential data, essential for detecting temporal DDoS patterns in 5G traffic.

- Key Components:

- Cell State:

- ✓ A persistent memory conduit that retains critical information across time steps, enabling long-term context retention.

- Forget Gate:

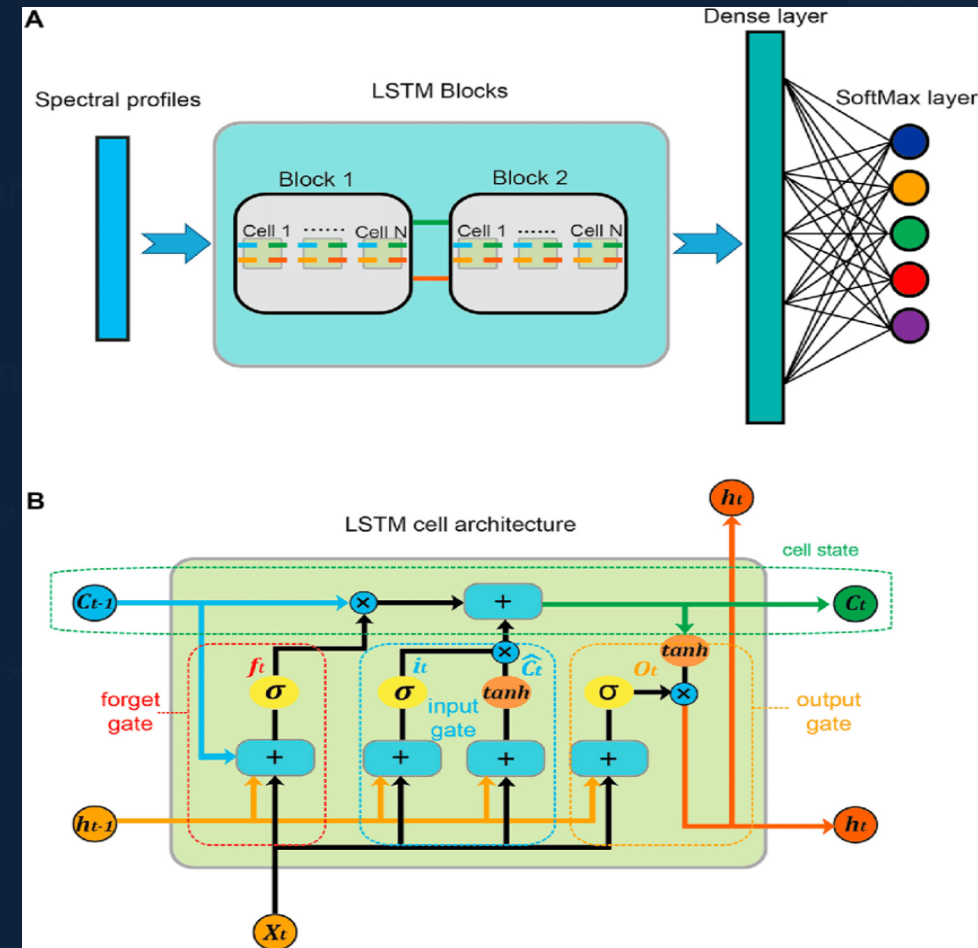
- ✓ Selectively discards irrelevant data using a sigmoid function ( $\sigma(W_f * [h_{t-1}, x_t] + b_f)$ ).

- Input Gate:

- ✓ Regulates new information addition via sigmoid ( $\sigma(W_i * [h_{t-1}, x_t] + b_i)$ ) and  $\tanh(W_C * [h_{t-1}, x_t] + b_C)$  activation.

- Output Gate:

- ✓ Filters the cell state to produce output using sigmoid ( $\sigma(W_o * [h_{t-1}, x_t] + b_o)$ ) and  $\tanh(C_t)$ .



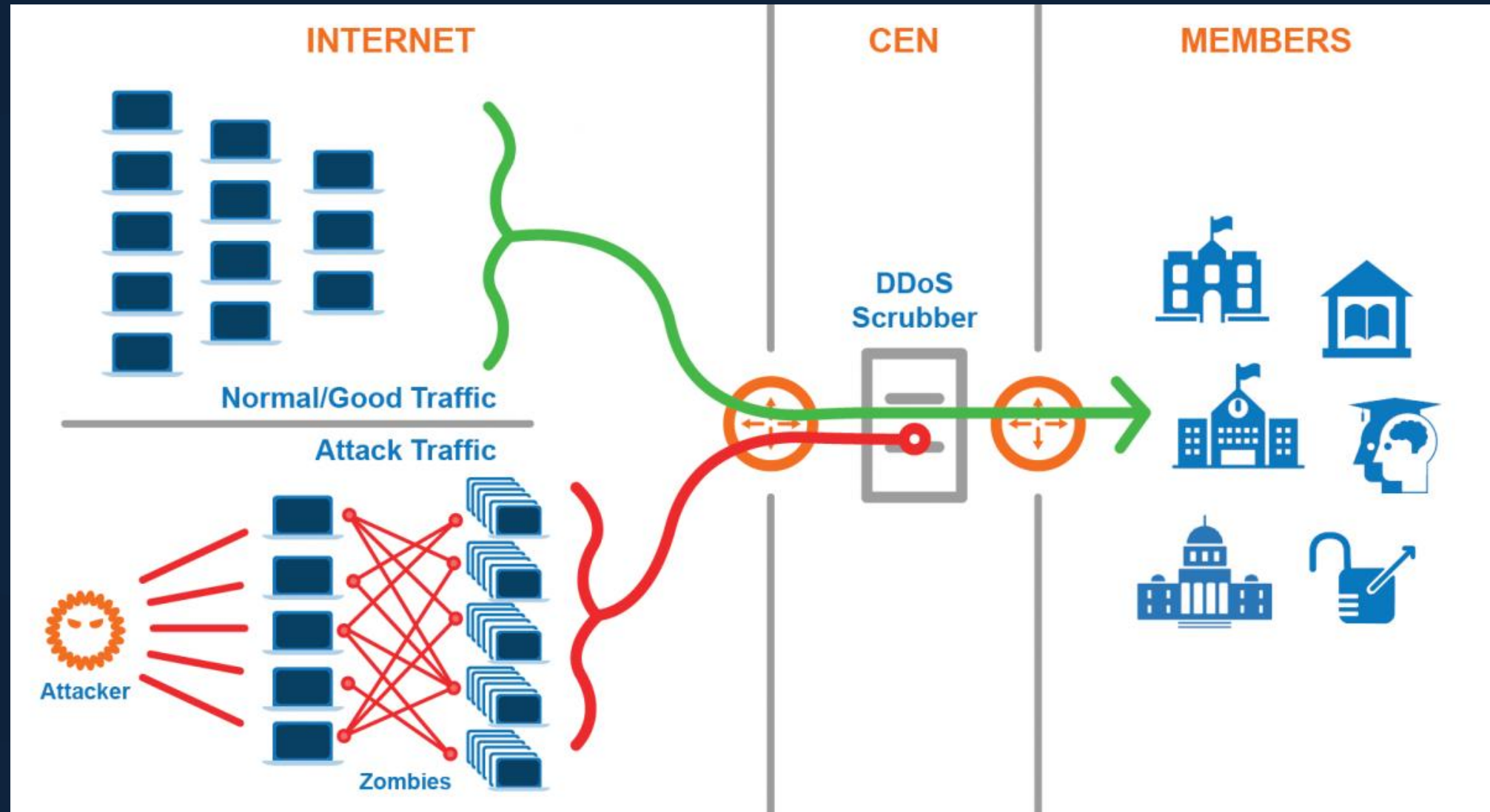
# System Architecture and Flowchart

- **Architecture:**
  - Input Layer: Processed 5G traffic features.
  - CNN Layers: Two layers with ReLU activation for feature extraction.
  - LSTM Layer: Sequential modeling of traffic anomalies.
  - Output Layer: Fully connected with softmax for classification.
- **Validation:** Rigorous cross-validation to ensure model robustness.

# System Architecture and Flowchart

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# How Model Prevents DDoS





# Model Training and Optimization

- **Configuration:**
  - **Optimizer:** Adam with adaptive learning rate.
  - **Loss Function:** Binary cross-entropy for two-class classification.
  - **Hyperparameters:** 50 epochs, batch size of 64, dropout rate of 0.2.
- **Regularization:** L2 normalization and dropout to prevent overfitting.


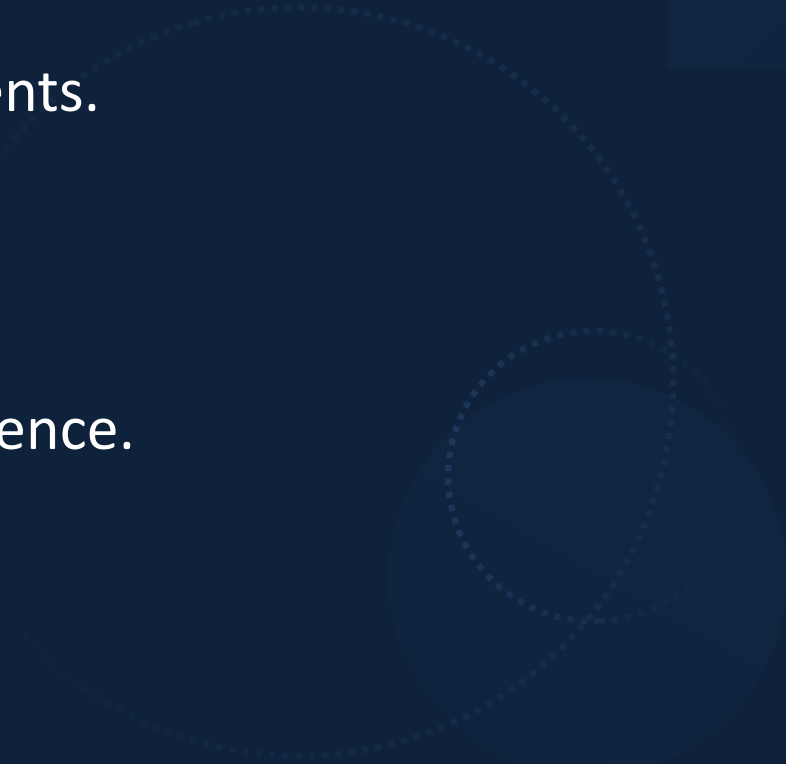
# Comparative Project Analysis

- **Project Title:** DDoS Attack Detection for 5G Network Slice using CNN
- **Source:** GitHub Repo - DDoS Attack Detection for 5G Network Slice using CNN <https://github.com/sajidkhan2067/DDoSAttackDetectionUsingCNN>
- **Objective:** Develop a CNN-based solution for DDoS detection in 5G network slices, achieving >99% accuracy.
- **Dataset:** Custom 10-million-row dataset, accessible via [IEEE DataPort \(DoS/DDoS Attack Dataset for 5G Network Slicing\)](#).



# Strengths and Innovations



- **Real-Time Capability:** Sub-millisecond detection latency, optimized for 5G.
  - **Scalability:** Handles terabyte-scale traffic in CPS environments.
  - **Adaptability:** Evolves with emerging attack signatures.
  - **Cyber Security Impact:** Fortifies critical infrastructure resilience.
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# Limitations and Challenges

- **Data Requirements:** Reliant on large, high-quality labeled datasets.
- **Computational Overhead:** Intensive GPU resources for training and inference.
- **Scope Limitation:** Optimized for specific DDoS variants; broader testing needed.
- **Network Specificity:** Performance unvalidated in non-5G contexts.

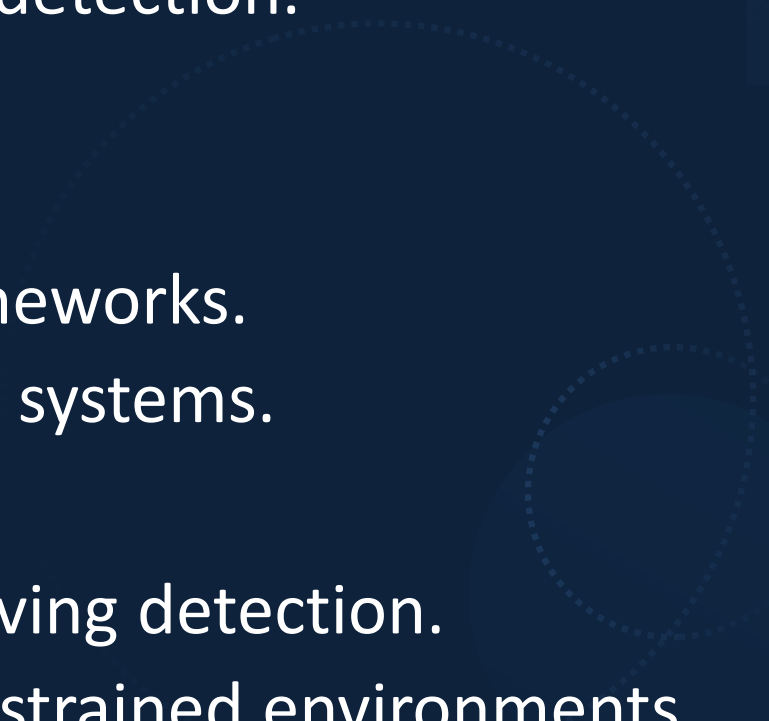
# Alignment with AML Curriculum

- **Core Topics:**
  - **Deep Learning:** CNN, LSTM (Unit 2: Advanced Neural Networks).
  - **Anomaly Detection:** Cyber security applications (Unit 3: ML in Security).
  - **Data Preprocessing:** Feature engineering (Unit 1: Data Science Foundations).
- **Competencies:**
  - Advanced model design and hyperparameter tuning.
  - Application of ML to solve real-world security challenges.



# Future Research Directions



- **Enhancements:**
    - Implement transfer learning for multi-attack detection.
    - Optimize for edge computing in CPS.
  - **Applications:**
    - Extend to IoT ecosystems and smart city frameworks.
    - Integrate with proactive intrusion prevention systems.
  - **Innovations:**
    - Explore federated learning for privacy-preserving detection.
    - Develop lightweight models for resource-constrained environments.
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# Conclusion

- **Summary:** The research paper and project showcase pioneering deep learning solutions for DDoS detection in 5G CPS, with accuracies of 98.7% and >99%, respectively.
- **Contribution:** Advances the field of cyber security for next-generation networks.
- **Takeaway:** Deep learning offers a transformative approach to securing critical infrastructures in the 5G era.



# References

- Hussain, B., Du, Q., Sun, B., & Han, Z. (2021). "Deep Learning-Based DDoS-Attack Detection for Cyber–Physical System Over 5G Network." *IEEE Transactions on Industrial Informatics*, 17(2), 860-870. DOI: [10.1109/TII.2020.2974520](https://doi.org/10.1109/TII.2020.2974520).
- AML Syllabus, National Forensic Sciences University, M.Tech AI & DS (Cyber Security), 2024-26.
- **GitHub Repository:** [DDoS Attack Detection for 5G Network Slice using CNN](https://github.com/sajidkhan2067/DDoSAttackDetectionUsingCNN), <https://github.com/sajidkhan2067/DDoSAttackDetectionUsingCNN>

THANK YOU

