



# Tech Saksham

## Capstone Project Report

### ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING FUNDAMENTALS

## EMAIL SPAM DETECTION

### KINGSTON ENGINEERING COLLEGE

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## **ABSTRACT**

Nowadays, a big part of people rely on available email or messages sent by the stranger. The possibility that anybody can leave an email or a message provides a golden opportunity for spammers to write spam message about our different interests .Spam fills inbox with number of ridiculous emails . Degrades our internet speed to a great extent .Steals useful information like our details on our contact list. Identifying these spammers and also the spam content can be a hot topic of research and laborious tasks. Email spam is an operation to send messages in bulk by mail .Since the expense of the spam is borne mostly by the recipient ,it is effectively postage due advertising. Spam email is a kind of commercial advertising which is economically viable because email could be a very cost effective medium for sender .With this proposed model the specified message can be stated as spam or not using Bayes' theorem and Naive Bayes' Classifier and Also IP addresses of the sender are often detected .

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# CHAPTER 1

## INTRODUCTION

Today, Spam has become a major problem in communication over internet. It has been accounted that around 55% of all emails are reported as spam and the number has been growing steadily. Spam which is also known as unsolicited bulk email has led to the increasing use of email as email provides the perfect ways to send the unwanted advertisement or junk newsgroup posting at no cost for the sender. This chances has been extensively exploited by irresponsible organizations and resulting to clutter the mail boxes of millions of people all around the world. Spam has been a major concern given the offensive content of messages, spam is a waste of time. End user is at risk of deleting legitimate mail by mistake. Moreover, spam also impacted the economical which led some countries to adopt legislation. Text classification is used to determine the path of incoming mail/message either into inbox or straight to spam folder. It is the process of assigning categories to text according to its content. It is used to organized, structures and categorize text. It can be done either manually or automatically.

Machine learning automatically classifies the text in a much faster way than manual technique. Machine learning uses pre-labelled text to learn the different associations between pieces of text and it output. It used feature extraction to transform each text to numerical representation in form of vector which represents the frequency of word in predefined dictionary. Text classification is important to structure the unstructured and messy nature of text such as documents and spam messages in a cost-effective way. Machine learning can make more accurate precisions in real-time and help to improve the manual slow process to much better and faster analysing big data. It is important especially to a company to analyse text data, help inform business decisions and even automate business processes. In this project, machine learning techniques are used to detect the spam message of a mail. Machine learning is where computers can learn to do something 10 without the need to explicitly program them for the task. It uses data and produce a program to perform a task such as classification. Compared to knowledge engineering, machine learning techniques require messages that have been successfully pre-classified. The pre-classified messages make the training dataset which will be used to fit the learning algorithm to the model in machine learning studio. A combination of algorithms are used to learn the classification rules from messages. These algorithms are used for classification of objects of different classes. These algorithms are provided with pre labelled data and an unknown text. After learning from the prelabelled data each of these algorithms predict which class the unknown text may belong to and the category predicted by majority is considered as final.

### 1.1 Problem Statement

Unwanted e-mails irritating internet connection

Critical e-mail message are missed and delayed

Millions of compromised computers

It occupies more space in the cloud

Identity theft

Spam can crash mail servers and fill up hard drives

## **1.2 Proposed Solution**

In this system, to solve the problem of spam, the spam classification system is created to identify spam and nonspam. Since spammers may send spam messages many times, it is difficult to identify it every time manually. So we will be using some of the strategies in our proposed system to detect the spam. The proposed solution not only identifies the spam word but also identifies the IP address of the system through which the spam message is sent so that next time when the spam message is sent from the same system our proposed system directly identifies it as blacklisted based on the IP address. In the proposed model, the web application is done using dot net and spam detection is done using machine learning. The web application consists of following modules:

## **1.3 Feature**

Email spam detection relies on a variety of features extracted from email data to distinguish between spam and legitimate messages. These features serve as input variables for machine learning algorithms and statistical models used in spam detection systems. Here are some common features used in email spam detection:

1. **Sender Information:** Characteristics of the email sender, including the sender's email address, domain reputation, sender's IP address, and authentication status (e.g., SPF, DKIM, DMARC). Anomalies or inconsistencies in sender information can indicate potential spam.
2. **Content Analysis:** Analysis of the textual content of the email, including subject line, body text, and embedded links. Features extracted from content analysis may include:
  - Presence of spam-related keywords or phrases (e.g., "free," "discount," "limited time offer").
  - Frequency of certain words or phrases.
  - Use of HTML or rich text formatting.

- Presence of misspellings, unusual characters, or obfuscation techniques.
3. Metadata Analysis: Examination of metadata associated with the email, such as timestamp, message ID, and header information. Metadata features may include:
    - Time of day the email was sent.
    - Geolocation of the sender's IP address.
    - Number of recipients.
    - Email client or software used to send the email.
  4. Structural Analysis: Analysis of the structural characteristics of the email, including:
    - Number of recipients (to, cc, bcc).
    - Presence of attachments or embedded media files.
    - MIME type of attachments.
    - HTML code analysis for suspicious elements (e.g., hidden text, invisible links).
  5. URL Analysis: Examination of URLs contained within the email, including:
    - URL length and format.
    - Domain reputation of linked websites.
    - Presence of URL redirects or URL shortening services.
    - Blacklisted or suspicious domains.
  6. Header Analysis: Inspection of email headers for anomalies or signs of spoofing, including:
    - Consistency between the "From" header and the sender's domain.
    - Presence of additional headers indicating email routing or forwarding.
    - Use of email authentication mechanisms (e.g., SPF, DKIM, DMARC).
  7. Behavioral Analysis: Analysis of user behavior and interaction patterns with emails, such as:
    - User engagement metrics (e.g., open rate, click-through rate).
    - Frequency of marking emails as spam or moving them to spam folders.
    - Analysis of historical email interactions and user preferences.
  8. Machine Learning-Based Features: Derived features generated through machine learning algorithms, such as:
    - Predicted probability scores from spam detection models.
    - Feature importance scores indicating the contribution of each feature to the classification decision.
    - Clustering or grouping of emails based on similarity in feature space.

#### **1.4 Advantages**

- Protection Against Malicious Activities:
- Enhanced Productivity:
- Improved User Experience
- Protection Against Offensive Content
- Reduced Risk of Security Breaches
- Preservation of Network Bandwidth
- Compliance with Regulations:
- Cost Savings

#### **1.5 Scope**

It provides sensitivity to the client and adapts well to the

Future spam techniques

It considers a complete message instead of single words with

Respect to its organization

It increases security and control

It reduces IT administration costs

It also reduce Network Resource costs

#### **1.3 Future work**

The future work of email spam detection will likely focus on addressing emerging challenges and leveraging advanced technologies to improve detection accuracy, efficiency, and user experience. Here are some potential areas of future research and development

- Deep Learning Techniques:
- Unsupervised Learning Approaches
- Multi-Modal Analysis
- Contextual Analysis
- Adversarial Defense Mechanisms
- Privacy-Preserving Techniques
- Real-Time Feedback Loops:
- Cross-Platform Integration
- Explainable AI (XAI)
- User-Centric Design



## **CHAPTER 2**

### **SERVICES AND TOOLS REQUIRED**

#### **2.1 Services Used**

- Email spam detection typically involves the utilization of various services, both standalone and integrated within larger email security solutions. Here are some key services commonly used for email spam detection
- Email Authentication Service
- URL and Domain Reputation Services
- Threat Intelligence Feeds
- Machine Learning and AI Services
- Anomaly Detection Services
- Reporting and Feedback Mechanisms
- Cloud-Based Spam Filtering Services
- Managed Security Services
- Anti-Spam Filtering Services

## **2.2 Tools and Software used**

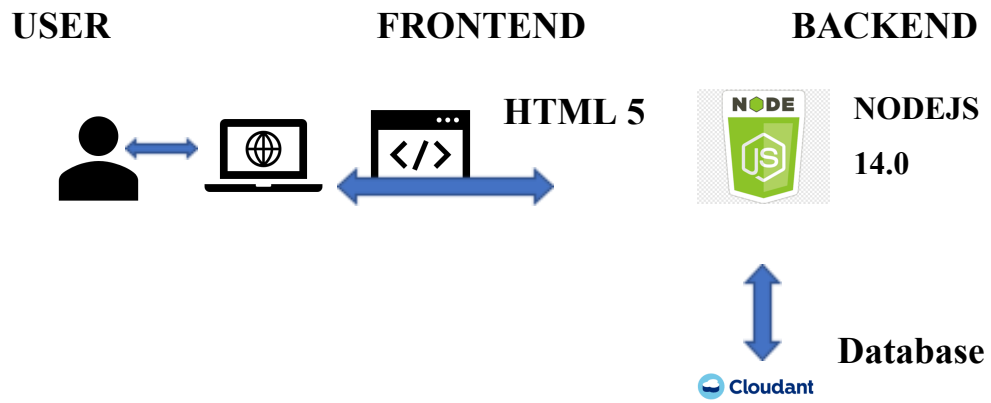
### **Tools and software used of email spam detection**

- Cisco Email Security
- Microsoft Exchange Online Protection (EOP)
- SpamTitan
- SpamAssassin
- MailScanner
- Barracuda Spam Firewall
- Proofpoint Email Protection
- Software used for email spam detect

## **CHAPTER 3**

# PROJECT ARCHITECTURE

## 3.1 Architecture



## CHAPTER 4

# PROJECT OUTCOME

The project outcome of an email spam detection endeavor can vary depending on the specific goals, scope, and requirements of the project. However, here are some potential project outcomes that can be achieved:

1. the project may be the successful development and implementation of a functional email spam detection system. This system would be capable of automatically classifying incoming emails as either spam or legitimate based on Development of a Functional Spam Detection System: The primary outcome of various features and criteria.
2. High Accuracy in Spam Detection: The project outcome may include achieving high levels of accuracy in spam detection, as measured by metrics such as precision, recall, F1-score, and accuracy. A well-performing spam detection system should minimize false positives (legitimate emails classified as spam) and false negatives (spam emails classified as legitimate).
3. Integration with Email Platforms: The spam detection system may be integrated into email servers, clients, or filtering gateways to provide real-time protection against spam. Integration with existing email platforms ensures seamless operation and user accessibility.
4. User-Friendly Interface: The project may result in the development of a user-friendly interface that allows users to manage spam filtering preferences, view spam detection results, and provide feedback on detected emails. A intuitive interface enhances user experience and engagement with the spam detection system.
5. Scalability and Efficiency: The spam detection system should be scalable and efficient, capable of handling large volumes of incoming emails without significant performance degradation. Optimized algorithms and data processing techniques contribute to scalability and efficiency.
6. Adaptability to New Threats: The outcome may include mechanisms for continuous monitoring and adaptation to new spamming techniques and emerging threats. The spam detection system should be able to dynamically adjust its algorithms and criteria to effectively detect and mitigate evolving spam campaigns.
7. Compliance with Regulations: If applicable, the project outcome may involve ensuring compliance with relevant regulations and standards governing email communications and data privacy. This may include adherence to regulations such as the CAN-SPAM Act or GDPR.
8. Documentation and Reporting: Comprehensive documentation and reporting on the project outcomes, including details of the spam detection system architecture, algorithms used, performance metrics achieved, and user feedback. Clear documentation facilitates knowledge transfer and future maintenance of the system.
9. Training and Support Materials: Creation of training materials and user guides to assist users in understanding and effectively utilizing the spam detection system. Providing ongoing support and training ensures optimal use and adoption of the system.
10. Evaluation and Validation: The project outcome may include thorough evaluation and validation of the spam detection system's performance through testing, validation, benchmarking against benchmark datasets or real-world email traffic. Validation ensures that the system meets the desired objectives and performance criteria.

## **CONCLUSION**

Email has been the most important medium of communication nowadays, through internet connectivity any message can be delivered to all over the world. More than 270 billion emails are exchanged daily, about 57% of these are just spam emails. Spam emails, also known as non-solicited, are undesired commercial or malicious emails, which affects or hacks personal information like bank ,related to money or anything that causes destruction to single individual or a corporation or a group of people. Besides advertising, these may contain links to phishing or malware hosting websites set up to steal confidential information. Spam is a serious issue that is not just annoying to the end-users but also financially damaging and a security risk. Hence this system is designed in such a way that it detects unsolicited and unwanted emails and prevents them hence helping in reducing the spam message which would be of great benefit to individuals as well as to the company .In the future this system can be implemented by using different algorithms and also more features can be added to the existing system.

## **FUTURE SCOPE**

- Enhanced Accuracy with AI and Machine Learning
- Behavioral Analysis and Contextual Understanding
- Multi-Modal Analysis
- Real-Time Threat Intelligence and Collaboration
- Privacy-Preserving Techniques
- Cross-Platform Integration
- Adaptive and Self-Learning Systems
- Explainable AI (XAI)

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## **CODE**

**Please Provide Code through Git Hub Repo Link**

**<https://github.com/aule-01eee/codersrsriram>**