



CAPSTONE PROJECT

PROJECT TITLE

**AI-Based Spam Email
Detection System using
Microsoft Azure Machine
Learning**

PRESENTED BY

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OUTLINE:

- **Problem Statement**
- **Proposed System/Solution**
- **System Development Approach**
- **Algorithm & Deployment**
- **Result (Output Image)**
- **Conclusion**
- **Future Scope**
- **References**

PROBLEM STATEMENT:

Email users receive many unwanted spam emails daily. These emails can contain malicious links, fraud messages, and promotional content. Manually identifying spam emails is inefficient and time-consuming.

PROPOSED SOLUTION:

- The proposed system uses Artificial Intelligence and Machine Learning to automatically classify emails as Spam or Not Spam. The system is trained using labeled email data and deployed on Microsoft Azure Machine Learning for real-time prediction.
- **Data Collection:**
 - Collected a dataset of emails labeled as Spam or Not Spam.
 - Ensured diversity by including emails from different sources and categories.
- **Data Preprocessing:**
 - Cleaned and preprocessed email text to remove noise, symbols, and irrelevant content.
 - Converted text into a suitable format for machine learning (feature extraction).
- **Machine Learning Algorithm:**
 - Trained a supervised learning algorithm (e.g., Logistic Regression) for binary classification.
 - The model learns patterns in email text to accurately predict Spam or Not Spam.
- **Deployment:**
 - Deployed the trained model as a real-time endpoint on Microsoft Azure Machine Learning.
 - Users can send email text to the endpoint and get instant predictions.
- **Evaluation:**
 - Evaluated model performance using metrics like accuracy and confusion matrix.
 - Fine-tuned the model for better prediction and reduced false positives.
 - Result: AI model accurately classifies emails as Spam or Not Spam.

SYSTEM APPROACH:

SYSTEM DEVELOPMENT APPROACH (TECHNOLOGY USED)

Microsoft Azure Portal:

Used Azure portal to manage and monitor machine learning resources. Enabled cloud-based storage, computation, and model deployment.

Azure Machine Learning Studio :

Designed and trained the ML model using the drag-and-drop interface.

Facilitated model experimentation, versioning, and testing easily.

Dataset (CSV File):

Uploaded labeled email dataset (Spam/Not Spam) for model training.

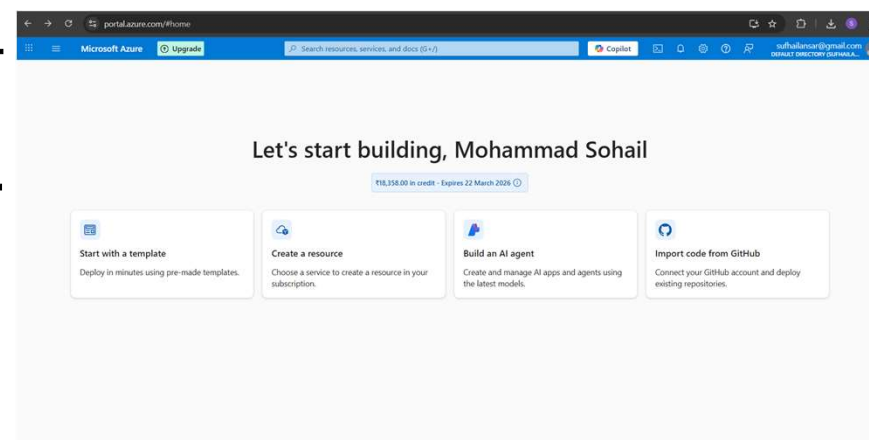
Ensured dataset is clean and ready for preprocessing.

Two-Class Classification Algorithm:

Used supervised algorithm Logistic Regression for binary classification.

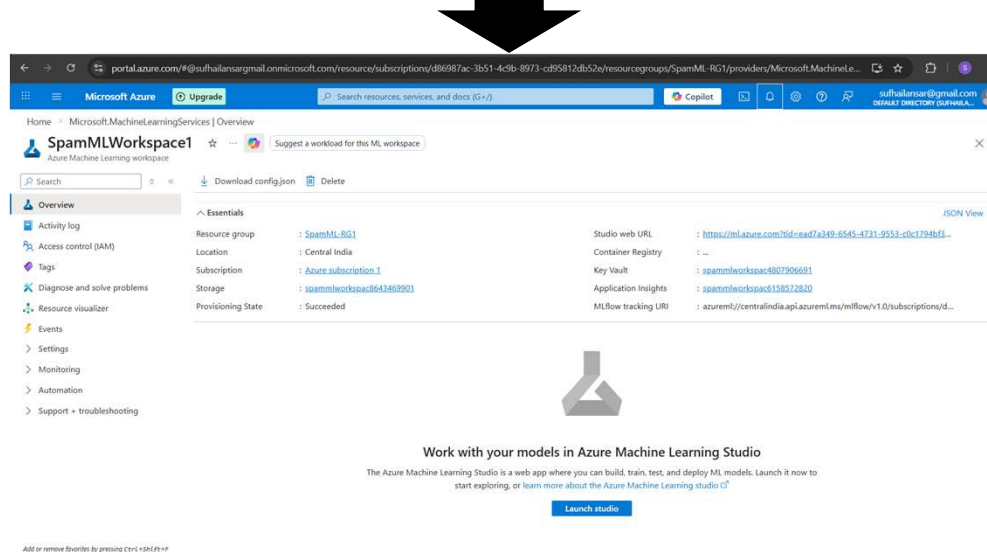
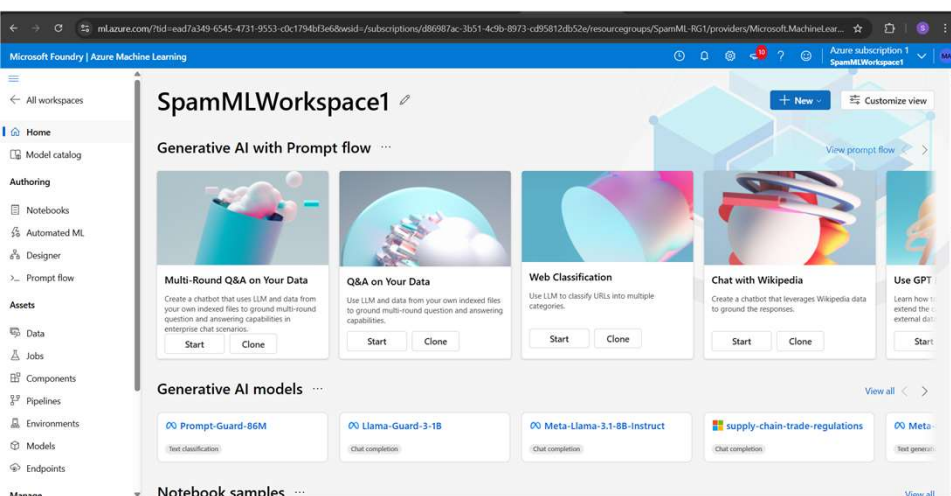
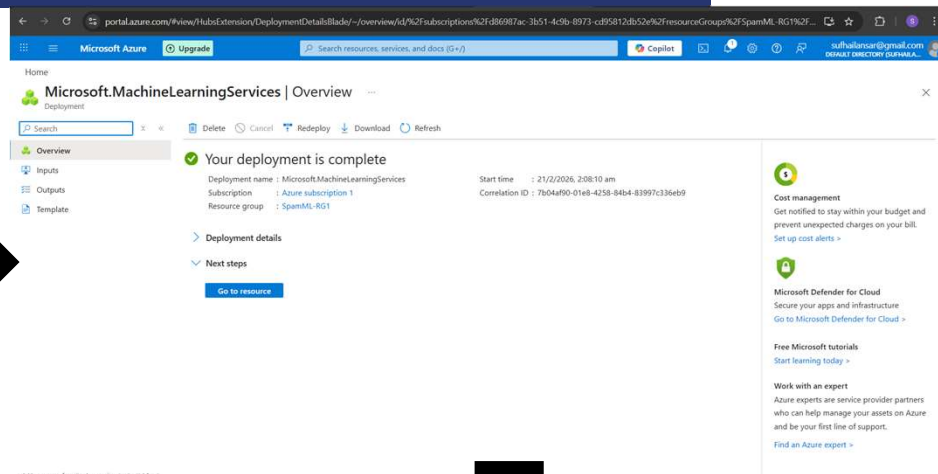
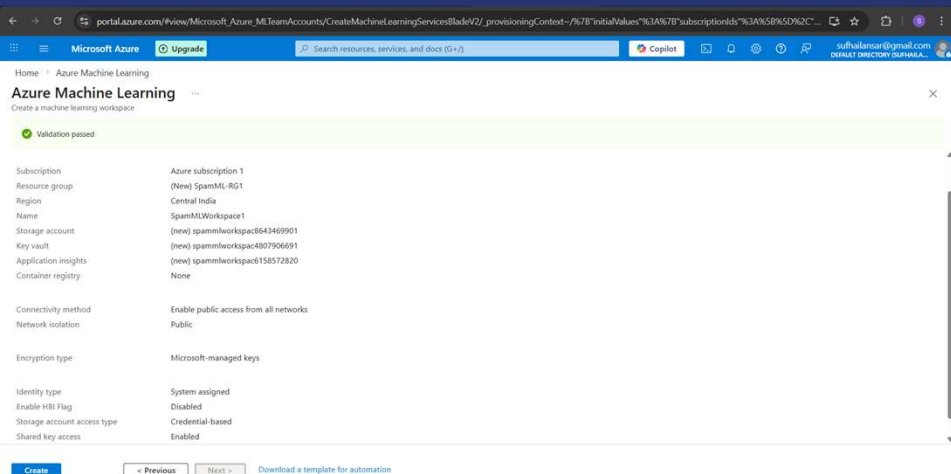
Model predicts email as Spam or Not Spam based on text patterns.

Cloud Deployment: Deployed trained model as a real-time endpoint in Azure.



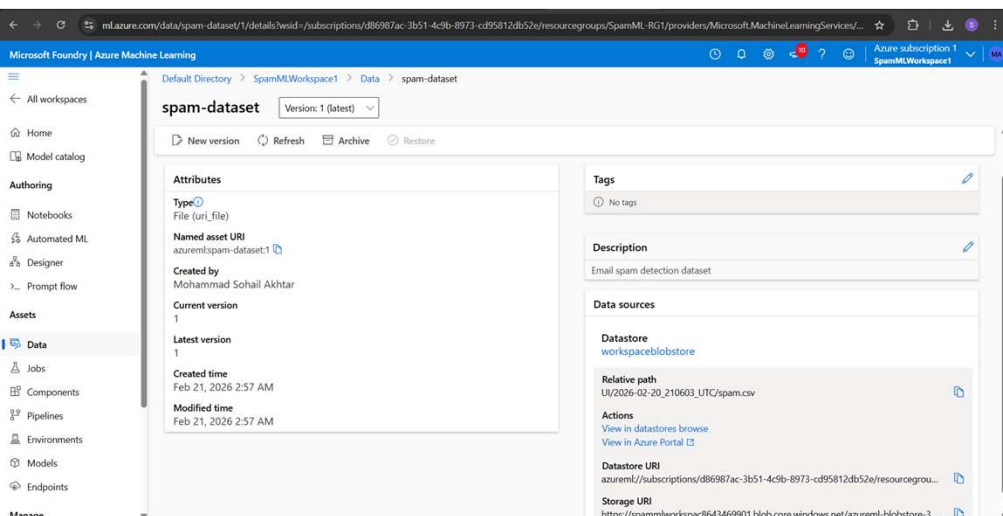
ALGORITHM & DEPLOYMENT:

- The project uses a Supervised Machine Learning algorithm such as Logistic Regression for binary classification. The model analyzes text patterns and predicts whether an email is Spam or Not Spam:
- **Algorithm Selection:**
 - Selected a supervised learning algorithm (Logistic Regression) for binary classification. Chosen due to its efficiency, simplicity, and good performance in text classification tasks.
- **Data Input:**
 - Input data consists of email text in CSV format with Spam/Not Spam labels. Text data is converted into numerical features using text vectorization techniques.
- **Training Process:**
 - The dataset is split into training and testing sets. The algorithm learns patterns from labeled data to classify emails accurately.
- **Prediction Process:**
 - New email text is sent to the trained model endpoint. The model analyzes patterns and predicts whether the email is Spam or Not Spam.

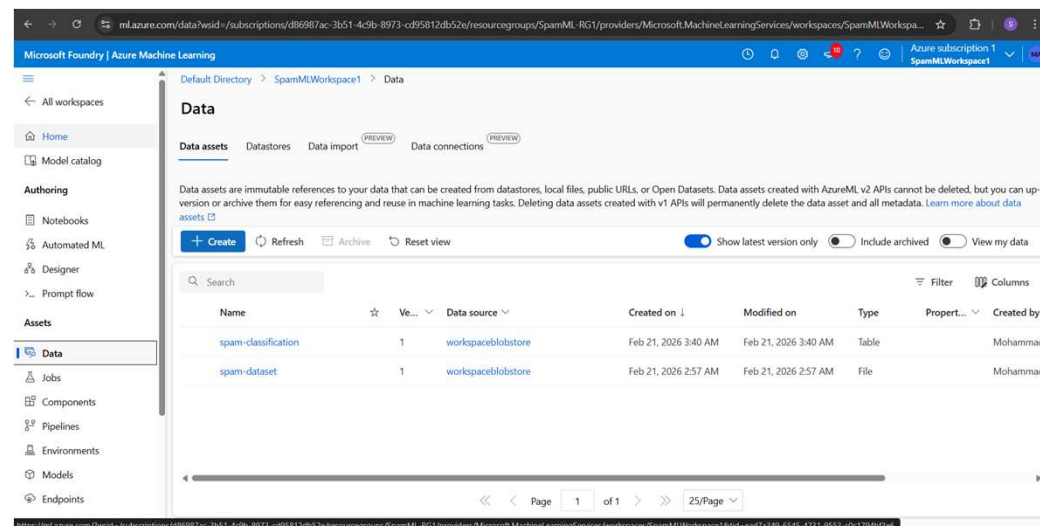


RESULT:

- The AI model successfully classifies emails as Spam or Not Spam with good accuracy.
- The system reduces manual effort and improves email security.
- The model achieved high accuracy with low false positives.



The screenshot shows the 'spam-dataset' details page in the Microsoft Foundry | Azure Machine Learning interface. The dataset is a File (url_file) with the named asset URI 'azureml-spam-dataset:1'. It was created by Mohammad Sohail Akhtar on Feb 21, 2026 2:57 AM. The current version is 1, and the latest version is also 1. The dataset is stored in the workspaceblobstore with the relative path 'URI/2026-02-20_210603_UTC/spam.csv'. The data source is 'workspaceblobstore' and the storage URI is 'https://spamlworkspace643469901.blob.core.windows.net/azureml-blobstore-3...'. The dataset is tagged with 'Email spam detection dataset'.



The screenshot shows the 'Data' section in the Microsoft Foundry | Azure Machine Learning interface. It displays a table of data assets with columns: Name, Version, Data source, Created on, Modified on, Type, and Created by. The table contains two rows: 'spam-classification' and 'spam-dataset'. Both assets are version 1 and are stored in the 'workspaceblobstore' data source. The 'spam-classification' asset was created on Feb 21, 2026 3:40 AM and is a Table type. The 'spam-dataset' asset was created on Feb 21, 2026 2:57 AM and is a File type. Both assets were created by Mohammad.

Name	Version	Data source	Created on	Modified on	Type	Created by
spam-classification	1	workspaceblobstore	Feb 21, 2026 3:40 AM	Feb 21, 2026 3:40 AM	Table	Mohammad
spam-dataset	1	workspaceblobstore	Feb 21, 2026 2:57 AM	Feb 21, 2026 2:57 AM	File	Mohammad

CONCLUSION:

This project demonstrates how Artificial Intelligence and Machine Learning can be effectively used to detect spam emails using Microsoft Azure Machine Learning. By training the model on labeled email data, the system is able to accurately classify emails as Spam or Not Spam. The cloud-based deployment ensures real-time prediction and scalability. Overall, the solution reduces manual effort, enhances email security, and showcases the practical implementation of ML in solving real-world problems.

FUTURE SCOPE:

In the future, this system can be integrated into real-time email platforms such as Gmail and Outlook to provide automatic spam filtering for users. The model can be further improved by using advanced techniques like Natural Language Processing (NLP) and Deep Learning algorithms to increase accuracy and reduce false predictions. Additionally, the system can be trained with larger and more diverse datasets to make it more robust, scalable, and efficient for large-scale applications.

REFERENCES:

This project was developed using resources and documentation from Microsoft Azure Machine Learning and Microsoft Learn for understanding cloud-based model training and deployment. The spam email dataset was obtained from Email/SMS Spam Collection Dataset [Kaggle], and machine learning concepts such as Logistic Regression were studied using official Scikit-learn documentation.

GitHub Link: <https://github.com/sufhailansar-bit/spam-email-classification-azure>

Thank You