Assignment Report - Machine Learning Laboratory

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Problem Statement:

A Spam message is any unsolicited digital communication message that is sent in bulk, and is often malicious.. A Ham message is any message that is not Spam. In other words, they are legitimate messages. With the rise in frequency and complexity of spam attacks, it is becoming increasingly difficult for users to identify and protect themselves from these attacks. It is essential to have a solution to automatically classify SMS messages as either spam or ham to enhance the security and privacy of users.

Dataset description:

The dataset¹ for this project has been taken from Kaggle.com which is an online community that lets users collaborate and publish datasets, use notebooks and compete in challenges. The dataset available at Kaggle.com titled **SMS Spam Collection Dataset** comprises a corpus that has been collected from free or free for research purposes sources at the Internet. The collection of SMS messages have been tagged as spam or legitimate.

- 425 SMS spam messages from Grumbletext, a UK forum where users make public claims about SMS spam. The messages have been manually extracted from the website where users often don't include the very message that they are reporting and hence, the process of identifying the right messages for the dataset is challenging and time consuming as it requires close examination of many pages.
- 3,375 SMS ham messages (a subset of a dataset of about 10,000 messages) of the NUS SMS Corpus², a dataset of ham messages collected for research purposes. This corpus was collected by Tao Chen and Min-Yen Kan³. This data was gathered for research purposes at the Department of Computer Science at the National University of Singapore. The messages mostly came from Singaporeans, with a significant portion being students at the university. Participants were informed that their contributions would be publicly available before volunteering to provide the message.

- 450 SMS ham messages from Caroline Tag's PhD Thesis⁴.
- 1,002 SMS ham messages and 322 SMS spam messages from SMS Spam Corpus v.0.1 Big⁵.

The dataset is contained in a CSV file - **spam.csv** (503.66 kB). This file contains one message per line and each line has two columns - v1 and v2.

- v1 This column contains the label that classifies the message as ham or spam.
- v2 This column contains the raw text or message

Statistics of dataset:

- Ham 87%
- Spam 13%
- No. of unique values = 5169

Tools to be used

- Coding Environment: Google Colab/ VSCode
- Language: Python

• Libraries:

- Scikit-learn It is an open source library for Python that is built on top of NumPy, SciPy, and Matplotlib. It provides a user-friendly interface for performing tasks like classification, regression, clustering, and dimensionality reduction. It is favored for its ease of use and performance in MAchine Learning applications.
- Pandas It is an open source data manipulation and analysis library for Python
 that provides flexible data structures like dataframes and series, to make working
 with relational data easier. It's used to load data from sources like CSV, SQL, and
 is useful in working with large datasets
- Numpy It is an open source library for Python that is used for numerical computing and performs fast array-oriented operations.
- Matplotlib It is an open source data visualization library for Python that creates static, animated and interactive visualizations for a wide range of plots like bar charts, histograms, scatter plots, and more.

Challenges Faced

- A large, diverse dataset is required for our project for training as spam messages can have very different content from one another, which can be challenging to accurately identify all types of spam.
- Some spam messages try to evade spam filters by using words and phrases that resemble or are used in ham messages.
- To make sure the class distribution is not imbalanced, i.e., the number of spam messages are much smaller than the number of ham messages.

Contribution of Team Members

This table contains the work assigned to each member.

Roll Number	Name	Contributions
20z201	Aadhithyashri	Data Visualization
20z214	Dhanush	Model Creation
20z234	Nashita V	Data Preprocessing
20z250	Shivani Sri S	Model Creation (Alternative)
20z261	Vivekanand	Data Visualization

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