

# CA02 Strategy Report- Jerry and Nicholson

## 1. Goal of This Project

The goal is to build a model to classify emails into: Spam (1) and Not Spam (0) We use Naive Bayes to learn patterns from training emails, then test the model on testing emails and report accuracy.

## 2. Data Understanding

This dataset is not a CSV file. It is a folder-based dataset:

- train-mails/ = training emails
- test-mails/ = testing emails

Each file is one email (a text file).

### How labels are decided

There is no label column. The label is from the **file name**:

- If the file name starts with “**spmsg**”, it is **Spam (1)**
- Otherwise, it is **Not Spam (0)**

This rule is used in both training and testing.

## 3. Sample Code Strategy / Logic

### Step 1 — Build Word Dictionary

Read all emails → Split text into words → Remove: non-alphabet words and single-letter words → Count word frequency → Keep top 3000 most frequent words as features.

### Step 2 — Feature Extraction

Each email becomes one row → Each dictionary word becomes one column → Cell value = number of times the word appears in the email → This converts text into numeric data.

### Step 3 — Label Creation

Labels are created from file names → spmsg\* → Spam (1) → Others → Not Spam (0)

#### **Step 4 — Model Training**

Use training feature matrix and labels → Train Naive Bayes model → Model learns word patterns for spam and non-spam.

#### **Step 5 — Model Testing**

Convert test emails using the same dictionary → Predict spam or not spam → Calculate accuracy.

### **4. Weakness of Sample Code**

#### **Weakness 1 — Model Choice**

The sample code uses Gaussian Naive Bayes but our features are word counts, which are discrete. Gaussian NB assumes the data is continuous and follows a normal distribution. So it is not the best match for this problem. The better choice is using Multinomial Naive Bayes.

#### **Weakness 2 — Dictionary built only from training**

The slide says to use all emails (train + test) to build the dictionary. The sample code builds dictionary from training only.

#### **Weakness 3 — Only uses one line of email text**

In the sample feature extraction, it only uses a specific line (line index = 2). This may miss important information in the full email body. The Better choice is use the full email text.

#### **Weakness 4 — Slow feature extraction**

The sample code checks each word by looping over all 3000 dictionary words repeatedly, this is slow and not efficient. The better choice is to create a word-to-index map, so we can find the column quickly.

#### **Weakness 5 — File path splitting may break on Windows**

The sample code splits file path using /. On Windows it may not work correctly

because Windows uses \. Th better choice is to use `os.path.basename()` to safely get file name.

## **5. Improved Design Plan**

- Use all emails to build dictionary.
- Read full email content.
- Keep cleaning rules: alphabet-only words and remove single-letter words
- Keep top 3000 words.
- Use Multinomial Naive Bayes.
- Use word-to-index mapping for faster feature extraction.
- Use safe filename extraction (`os.path.basename`).
- Train model and report accuracy.