





Ham vs Spam Emails

Project 4 Group 2

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 **Inbox** 9,350 Starred Snoozed Important Sent Drafts Categories **Social** 8,470 **Updates** 8,982 Forums **Promotions** 60,288 More

INTRODUCTION

Problem:

- Receiving too many spam emails in your inbox
- Having to find a ham email in your spam/junk folders

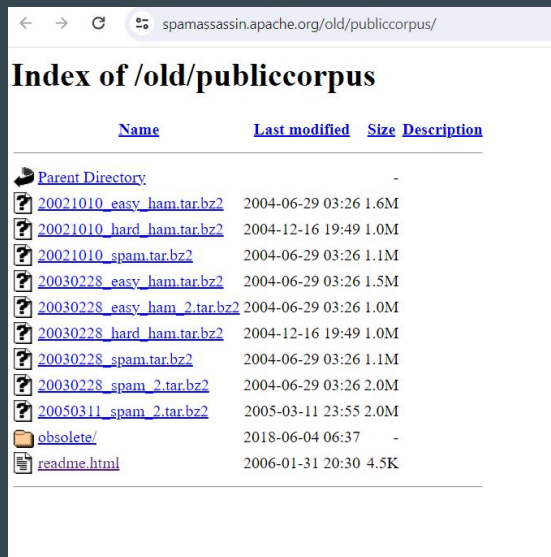
Question: How can we accurately sort spam and ham emails?

What are Spam and Ham emails?:

- **Spam:** unsolicited and unwanted junk email sent out in bulk to an indiscriminate
- **Ham:** non-spam email, “good/wanted” email

Data Source and Explanation of Data

- “Email Spam Dataset (Extended)” from Kaggle containing 9,000 files
- Original dataset came from SpamAssassin’s Old Public Corpus which had about 6,000 files
- Altogether, contains 6,951 ‘ham’ and 2,398 ‘spam’ raw email files.
- Includes data between 2002 and 2005.
- Spam emails marked by use of HTML, unusual HTML markup, colored text, and “spammish-sounding” phrases



The screenshot shows a web browser window with the URL `spamassassin.apache.org/old/publiccorpus/`. The page title is "Index of /old/publiccorpus". Below the title is a table with columns: Name, Last modified, Size, and Description. The table lists various tar.gz files, including "Parent Directory", "20021010_easy_ham.tar.bz2", "20021010_hard_ham.tar.bz2", "20021010_spam.tar.bz2", "20030228_easy_ham.tar.bz2", "20030228_easy_ham_2.tar.bz2", "20030228_hard_ham.tar.bz2", "20030228_spam.tar.bz2", "20030228_spam_2.tar.bz2", "20050311_spam_2.tar.bz2", "obsolete/", and "readme.html".

Name	Last modified	Size	Description
Parent Directory	-		
20021010_easy_ham.tar.bz2	2004-06-29 03:26	1.6M	
20021010_hard_ham.tar.bz2	2004-12-16 19:49	1.0M	
20021010_spam.tar.bz2	2004-06-29 03:26	1.1M	
20030228_easy_ham.tar.bz2	2004-06-29 03:26	1.5M	
20030228_easy_ham_2.tar.bz2	2004-06-29 03:26	1.0M	
20030228_hard_ham.tar.bz2	2004-12-16 19:49	1.0M	
20030228_spam.tar.bz2	2004-06-29 03:26	1.1M	
20030228_spam_2.tar.bz2	2004-06-29 03:26	2.0M	
20050311_spam_2.tar.bz2	2005-03-11 23:55	2.0M	
obsolete/	2018-06-04 06:37	-	
readme.html	2006-01-31 20:30	4.5K	



Data Cleaning and Preprocessing

1. Parse emails to just their text
2. BeautifulSoup to convert raw email files to plain text
3. Normalize text
 - a. Lowercasing
 - b. Removing punctuation
 - c. Stemming
4. Create a vector of word counts

```
#exapmle of stemming
text = "Hello, today I am going to London for performing and dancing"
stemmer = nltk.PorterStemmer()

for word in text.split():
    stemmed_word = stemmer.stem(word)
    print(stemmed_word, end=" ")
```

hello, today i am go to london for perform and danc

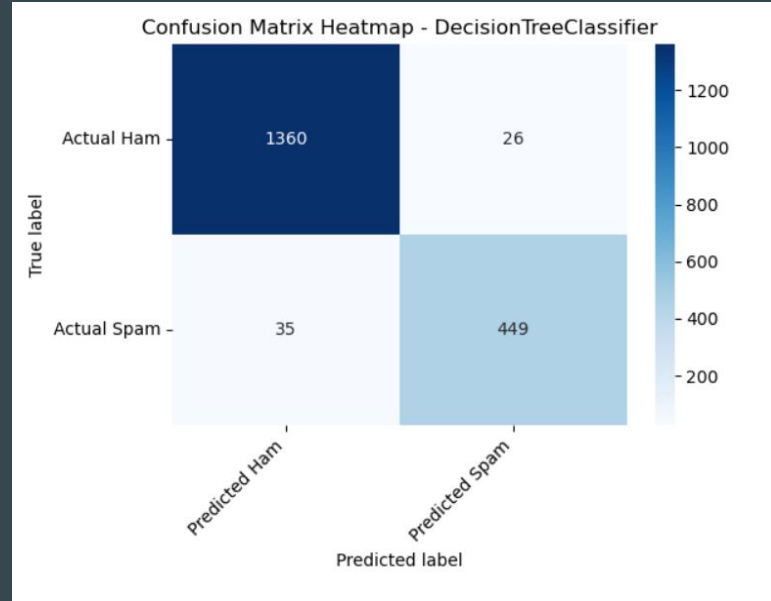
Models Used During Optimization Process

- Random Forest
- Decision Tree
- Logistic Regression

Decision Tree

"Tree-like model with nodes representing features and leaves representing outcomes. Used for both classification and regression."

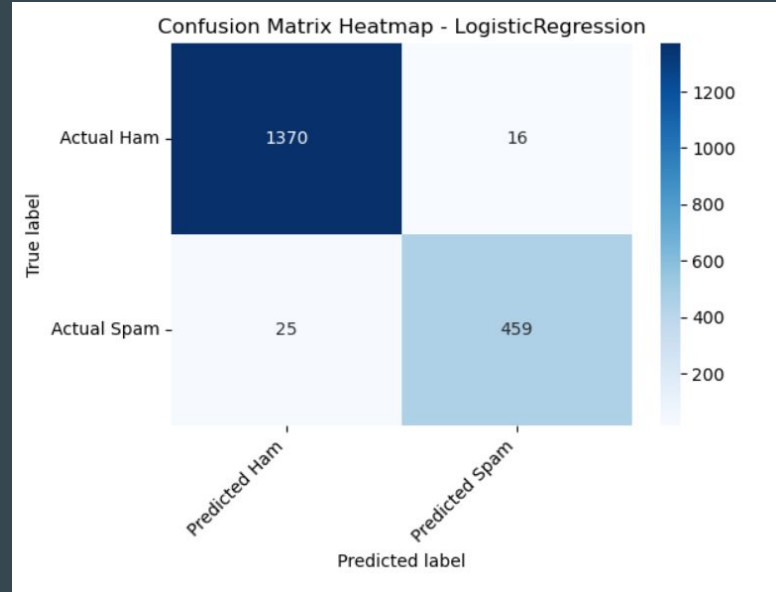
- 97% Accuracy
 - True Positive - 449
 - True Negative - 1360
 - False Positive (Spam) - 26
 - False Negative (Ham) - 35



Logistic Regression

"Linear model predicting probabilities for binary classification. Extends to multinomial logistic regression for multiple classes."

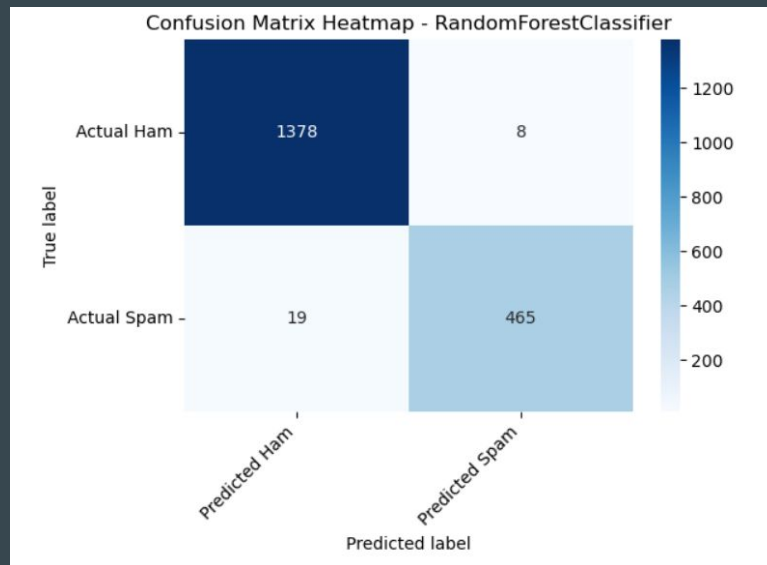
- Accuracy 98%
 - True Positive - 459
 - True Negative - 1370
 - False Positive (Spam) - 16
 - False Negative (Ham) - 25



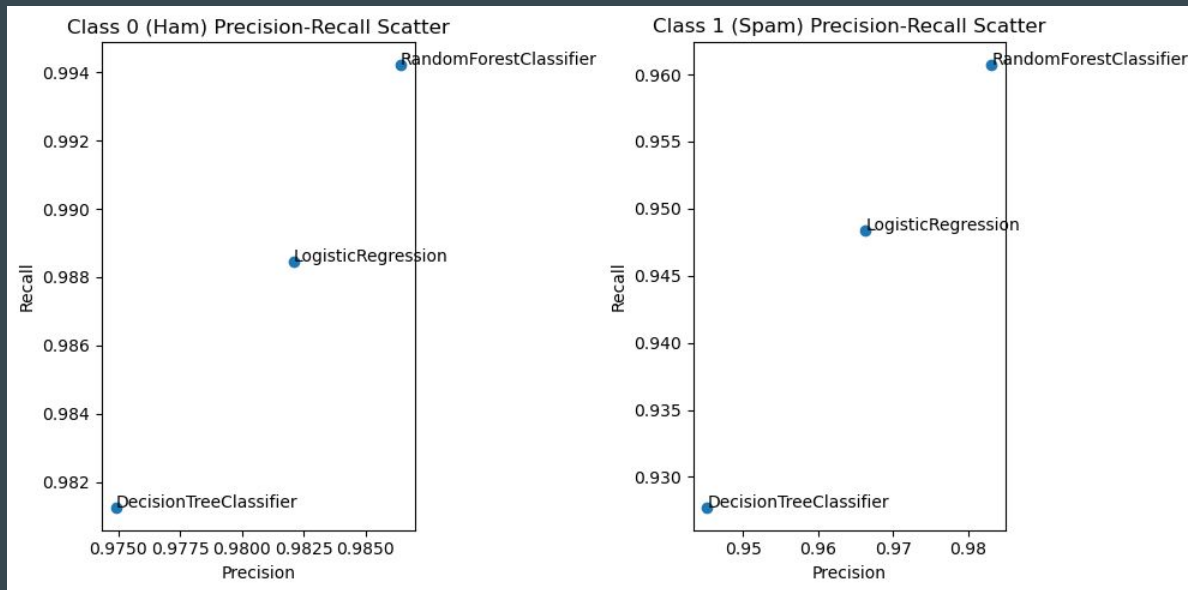
Random Forest Model

"Ensemble of decision trees. Each tree is built with randomness to improve generalization and reduce overfitting. Used for both classification and regression."

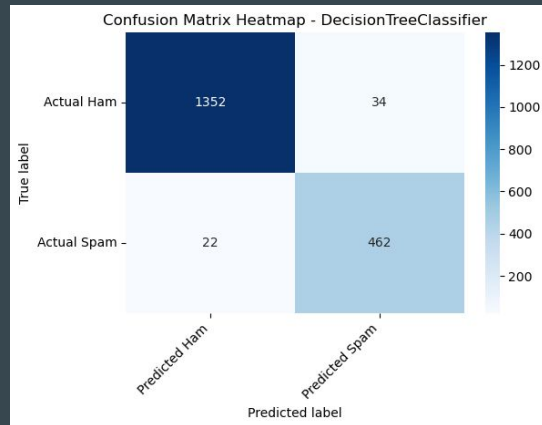
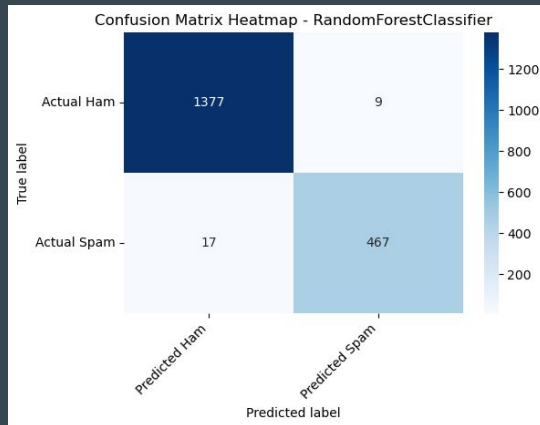
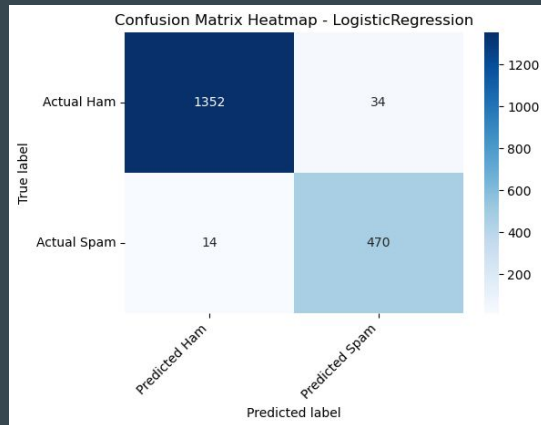
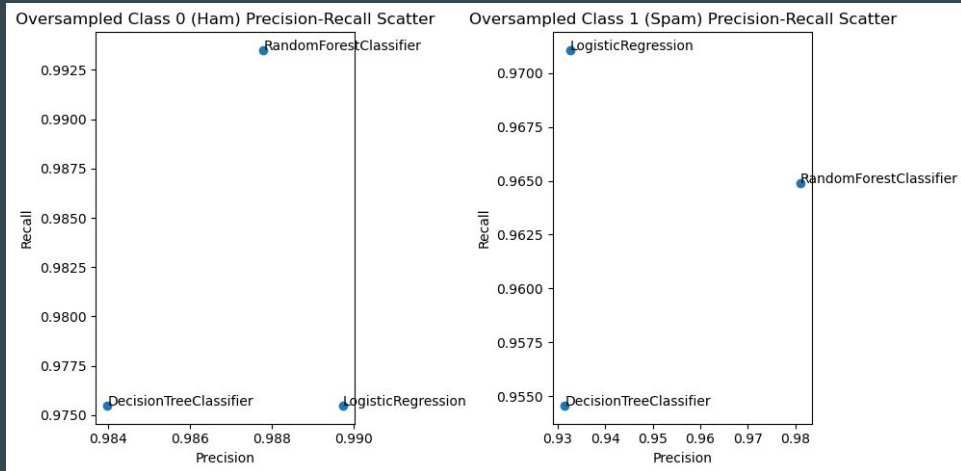
- Optimization - reducing false positives and false negatives.
- 99% Accuracy
 - True Positives - 465
 - True Negatives - 1378
 - False Positives (Spam) - 8
 - False Negatives (Ham) - 19



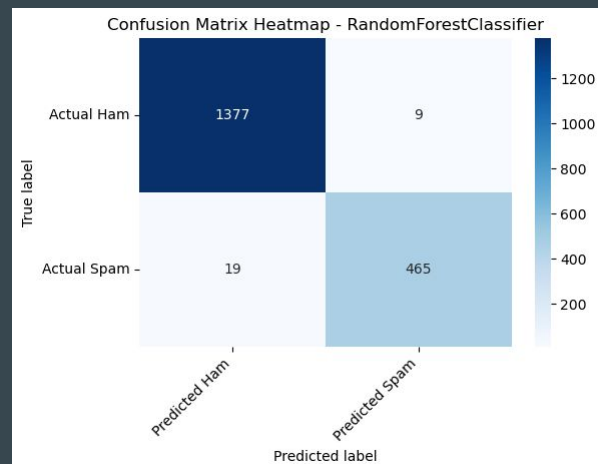
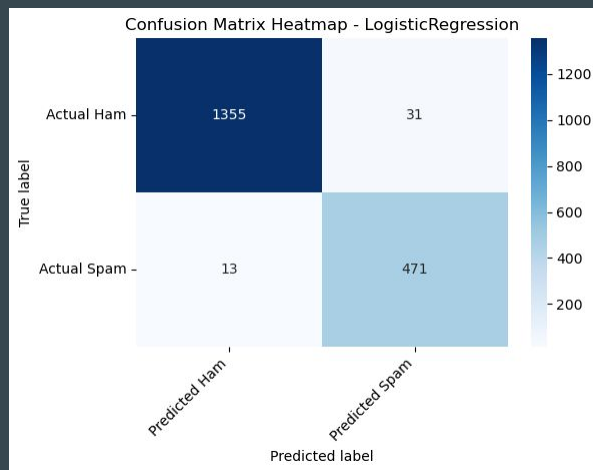
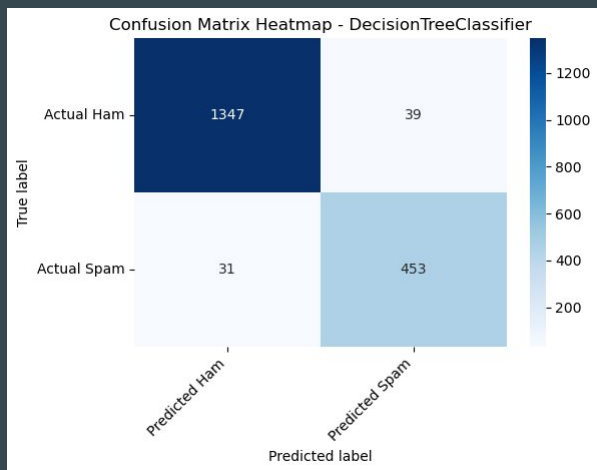
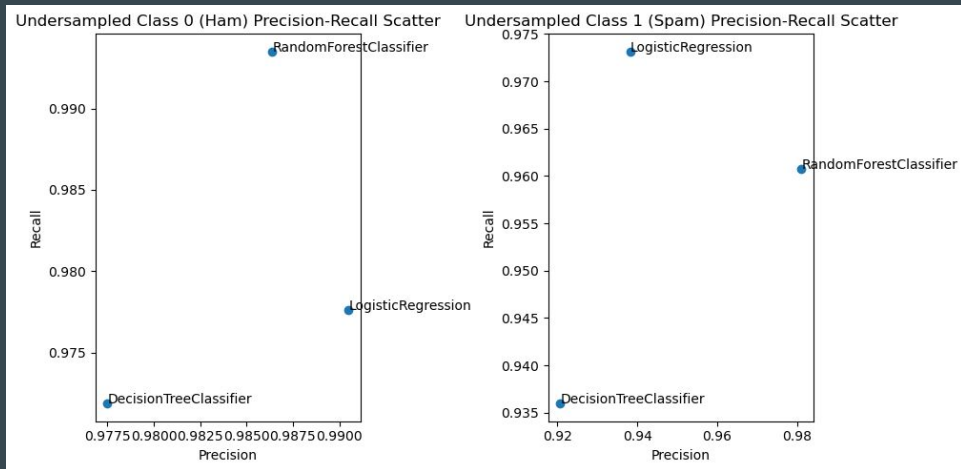
Comparison



Oversampling

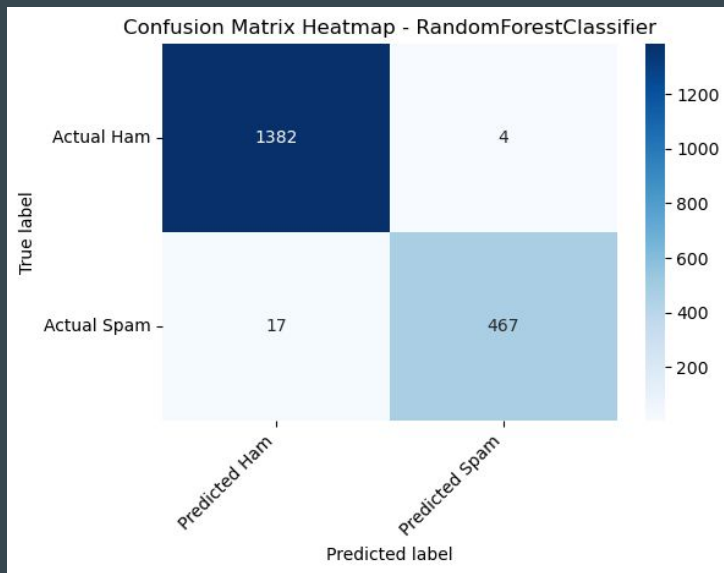


Undersampling

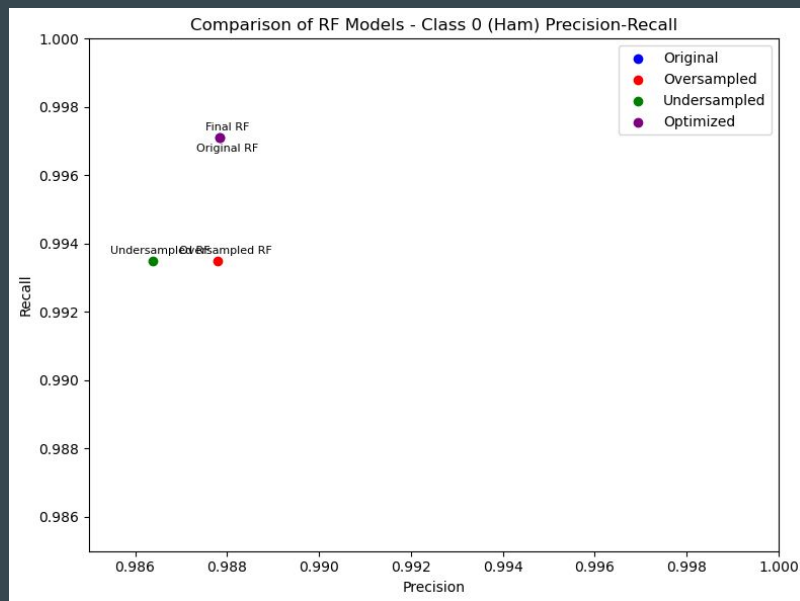


Hyperparameter Tuning

```
models = [  
    RandomForestClassifier(n_estimators=80, random_state=3301, max_depth=40,  
                           min_samples_split=25, bootstrap=False, ccp_alpha=0.0000008)  
]
```



Comparison of Random Forest Models



Possible Future Steps

- Compare the accuracy of the computer model we created for this data set to a dataset containing more recent emails to see how the accuracy changes.
 - Increase the number of most commonly used words in the word vector.
 - Add any words that we believe could be important in distinguishing 'ham' vs 'spam' emails.
-

Resources:

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Thank you!

Special thanks to Hunter, Sam, and Randy. Questions?