



# **SMS AND EMAIL SPAM DETECTION SYSTEM**

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# Problem Statement

MILLIONS OF SPAM SMS/EMAILS ARE SENT DAILY, CAUSING CONFUSION FOR USERS. MANY PEOPLE CANNOT DISTINGUISH BETWEEN LEGITIMATE MESSAGES (E.G., BANK ALERTS) AND SPAM (E.G., FAKE LOTTERY SCAMS). FINANCIAL FRAUD, PHISHING ATTACKS, AND PRIVACY BREACHES OCCUR DUE TO DECEPTIVE SPAM.

# PROBLEM STATEMENT

Millions of spam SMS/emails are sent daily, causing confusion for users.

Many people cannot distinguish between legitimate messages (e.g., bank alerts) and spam (e.g., fake lottery scams)

Financial fraud, phishing attacks, and privacy breaches occur due to deceptive spam.

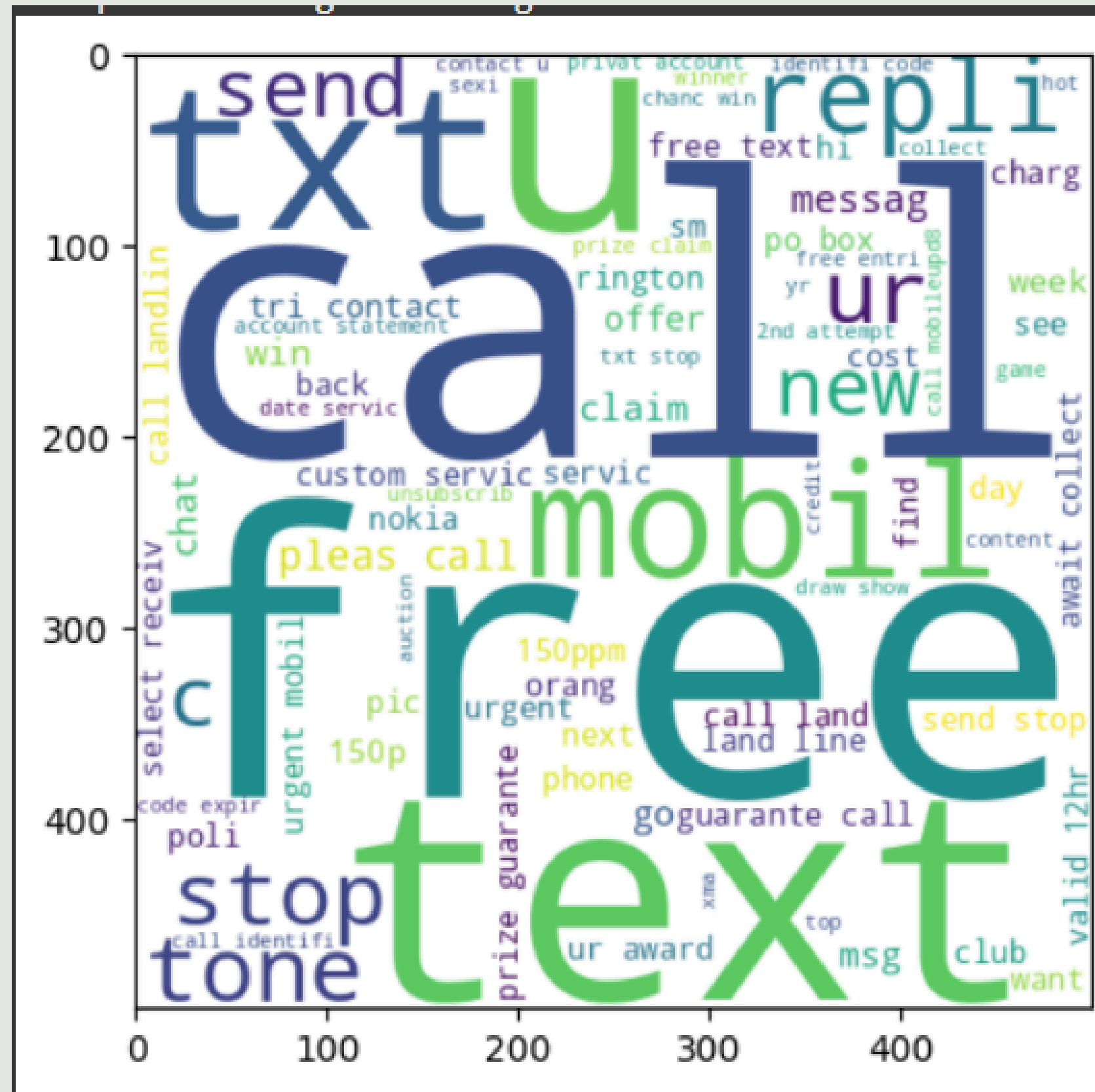
# MACHINE LEARNING MODELS

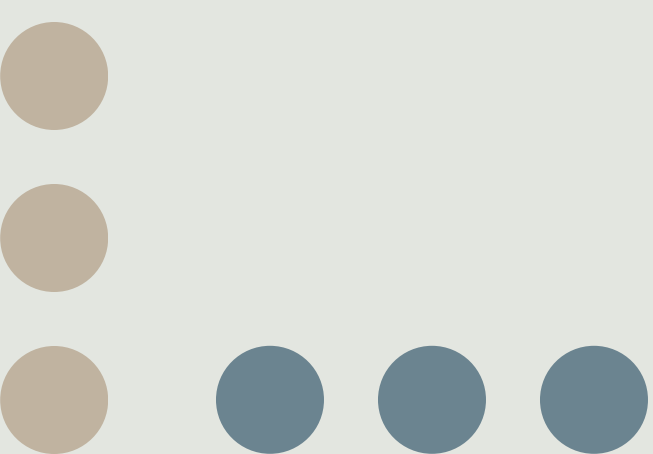
- Classifiers Used
- Naive Bayes
- Logistic Regression
- K-Nearest Neighbors
- Decision Trees
- Random Forest
- Support Vector Machine

# Implementation Workflow



# INTRODUCTION THE SPAM WORD



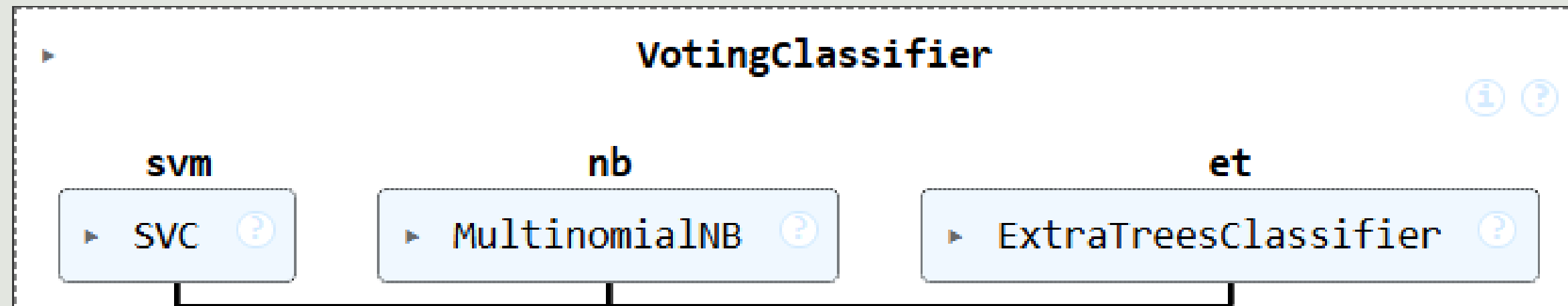
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# EXPERIMENTAL RESULTS WITH COUNT VECTORIZER AND SMOTE

|    | Algorithm | Accuracy | Precision |
|----|-----------|----------|-----------|
| 1  | KN        | 0.910853 | 1.000000  |
| 8  | ETC       | 0.972868 | 1.000000  |
| 5  | RF        | 0.967054 | 1.000000  |
| 4  | LR        | 0.974806 | 0.980198  |
| 6  | AdaBoost  | 0.936047 | 0.938462  |
| 10 | xgb       | 0.972868 | 0.935780  |
| 9  | GBDT      | 0.950581 | 0.928571  |
| 3  | DT        | 0.936047 | 0.913043  |
| 2  | NB        | 0.977713 | 0.909836  |
| 7  | BgC       | 0.955426 | 0.888889  |
| 0  | SVC       | 0.967054 | 0.886957  |
|    |           |          |           |



# VOTING CLASSIFIER



# RESULT FINAL

|    | Algorithm | Variable  | Value    | Accuracy_max_ft_3000 | Precision_max_ft_3000 |
|----|-----------|-----------|----------|----------------------|-----------------------|
| 0  | KN        | Accuracy  | 0.910853 | 0.910853             | 1.000000              |
| 1  | ETC       | Accuracy  | 0.972868 | 0.972868             | 1.000000              |
| 2  | RF        | Accuracy  | 0.967054 | 0.967054             | 1.000000              |
| 3  | LR        | Accuracy  | 0.974806 | 0.974806             | 0.980198              |
| 4  | AdaBoost  | Accuracy  | 0.936047 | 0.936047             | 0.938462              |
| 5  | xgb       | Accuracy  | 0.972868 | 0.972868             | 0.935780              |
| 6  | GBDT      | Accuracy  | 0.950581 | 0.950581             | 0.928571              |
| 7  | DT        | Accuracy  | 0.936047 | 0.936047             | 0.913043              |
| 8  | NB        | Accuracy  | 0.977713 | 0.977713             | 0.909836              |
| 9  | BgC       | Accuracy  | 0.955426 | 0.955426             | 0.888889              |
| 10 | SVC       | Accuracy  | 0.967054 | 0.967054             | 0.886957              |
| 11 | KN        | Precision | 1.000000 | 0.910853             | 1.000000              |
| 12 | ETC       | Precision | 1.000000 | 0.972868             | 1.000000              |
| 13 | RF        | Precision | 1.000000 | 0.967054             | 1.000000              |
| 14 | LR        | Precision | 0.980198 | 0.974806             | 0.980198              |
| 15 | AdaBoost  | Precision | 0.938462 | 0.936047             | 0.938462              |
| 16 | xgb       | Precision | 0.935780 | 0.972868             | 0.935780              |
| 17 | GBDT      | Precision | 0.928571 | 0.950581             | 0.928571              |

# CONCLUSION

- Using multiple vectorization methods and classifiers model achieved the highest accuracy in SMS spam detection.

## Limitations

- Experiments limited to a single UCI dataset which may not represent diverse SMS spam patterns globally.
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***THANK YOU***