

# Messages Monitoring Model Capable of Detecting and Deleting Spam in Public Telegram Group Chats



Ahmed Almohammed's  
Capstone Project for  
MiSK DSI Program



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**Problem Addressed**

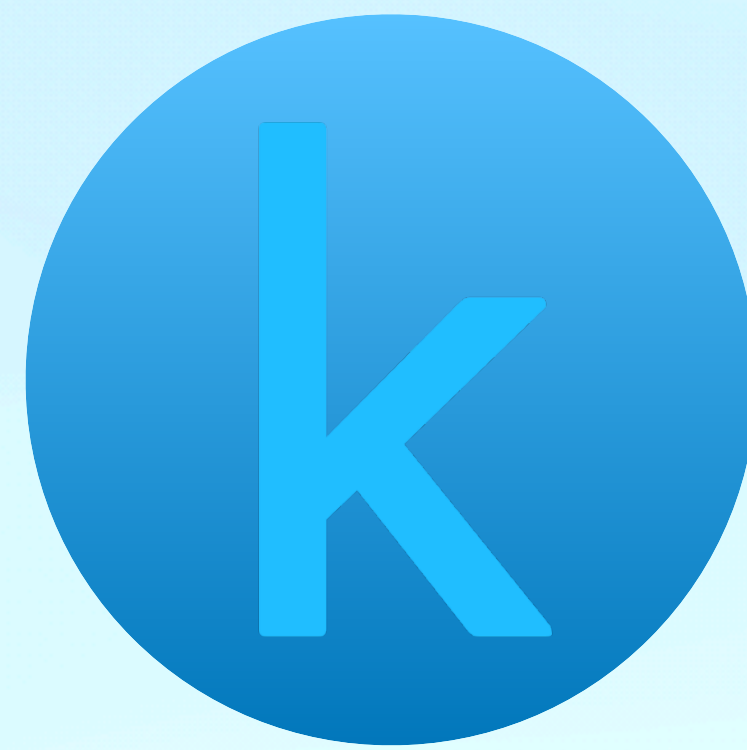
# Data Acquisition





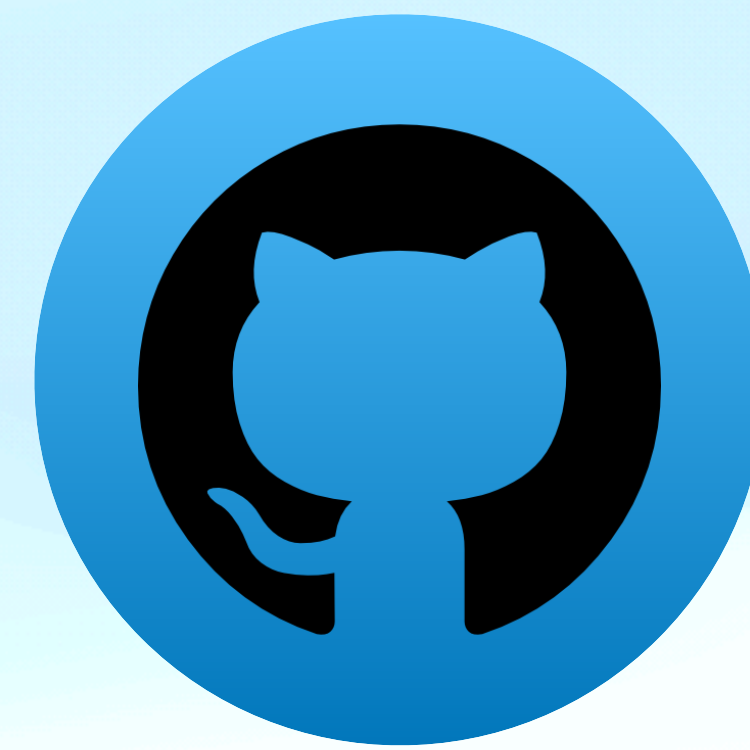
**uciml**

Kaggle



**team-ai**

Kaggle



**@DeshDSingh**

GitHub  
Repository



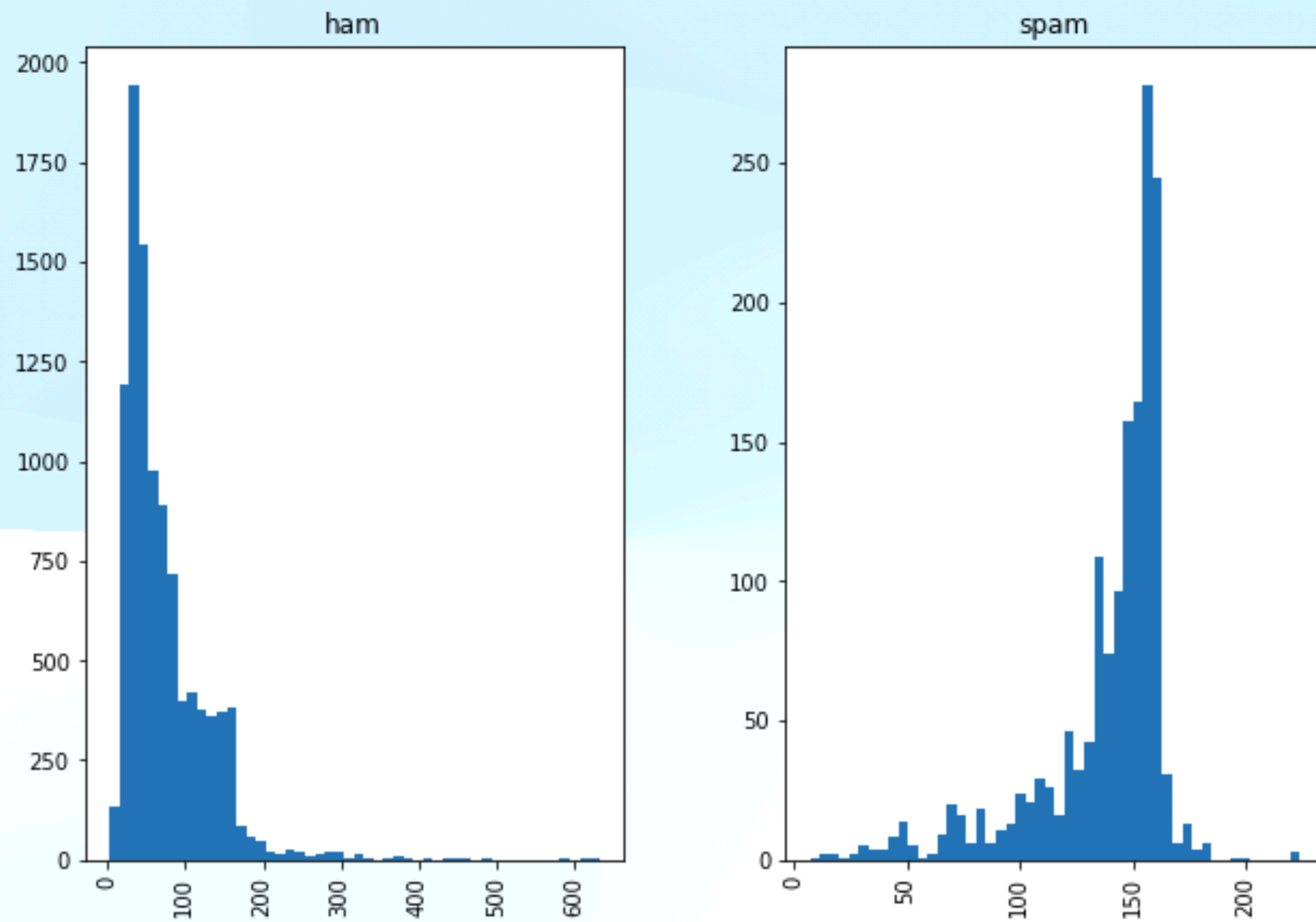
**👋HuggingFace**

HuggingFace  
Datasets

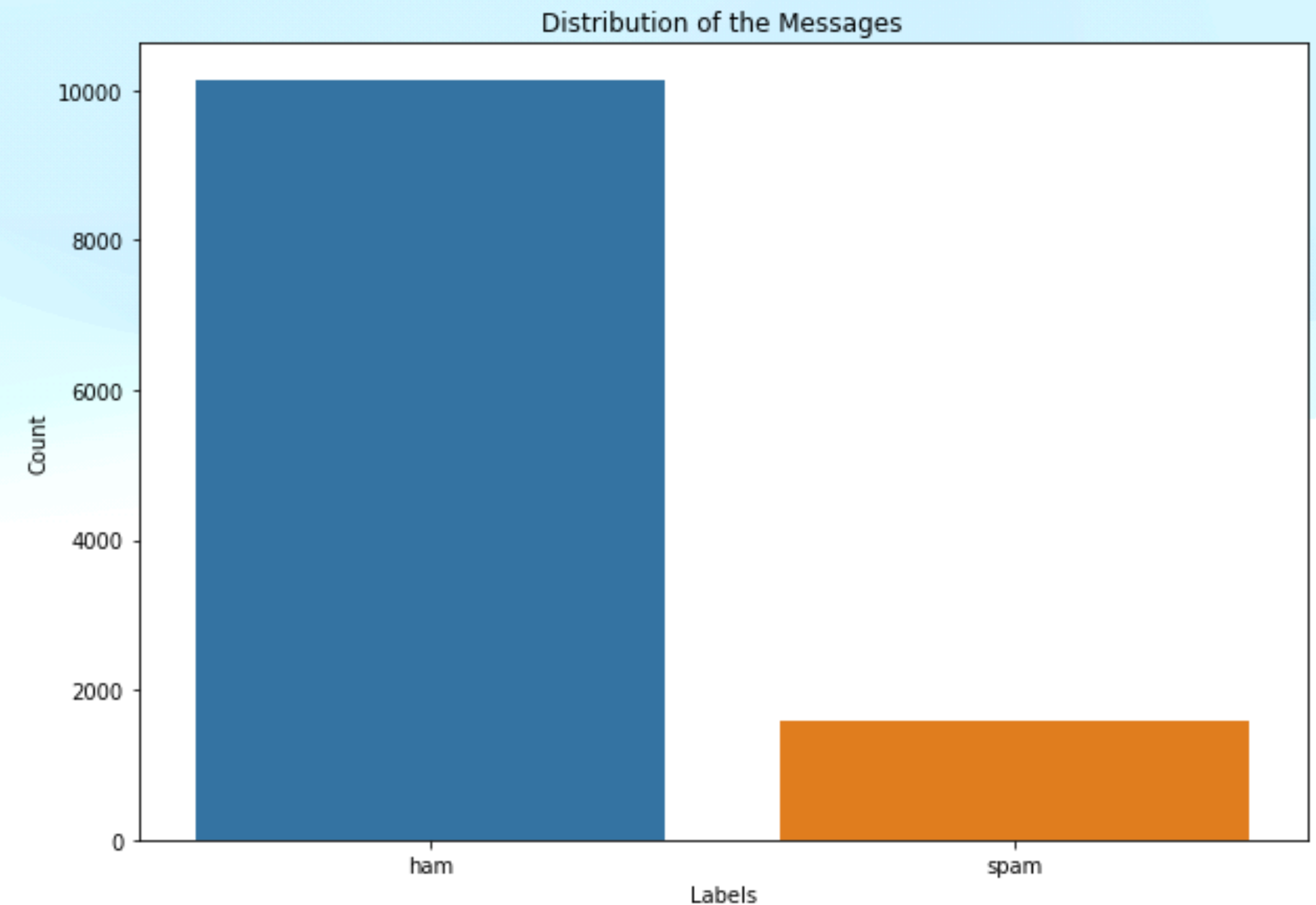
# Preprocessing Data



# Main Steps Taken in Preprocessing the Data



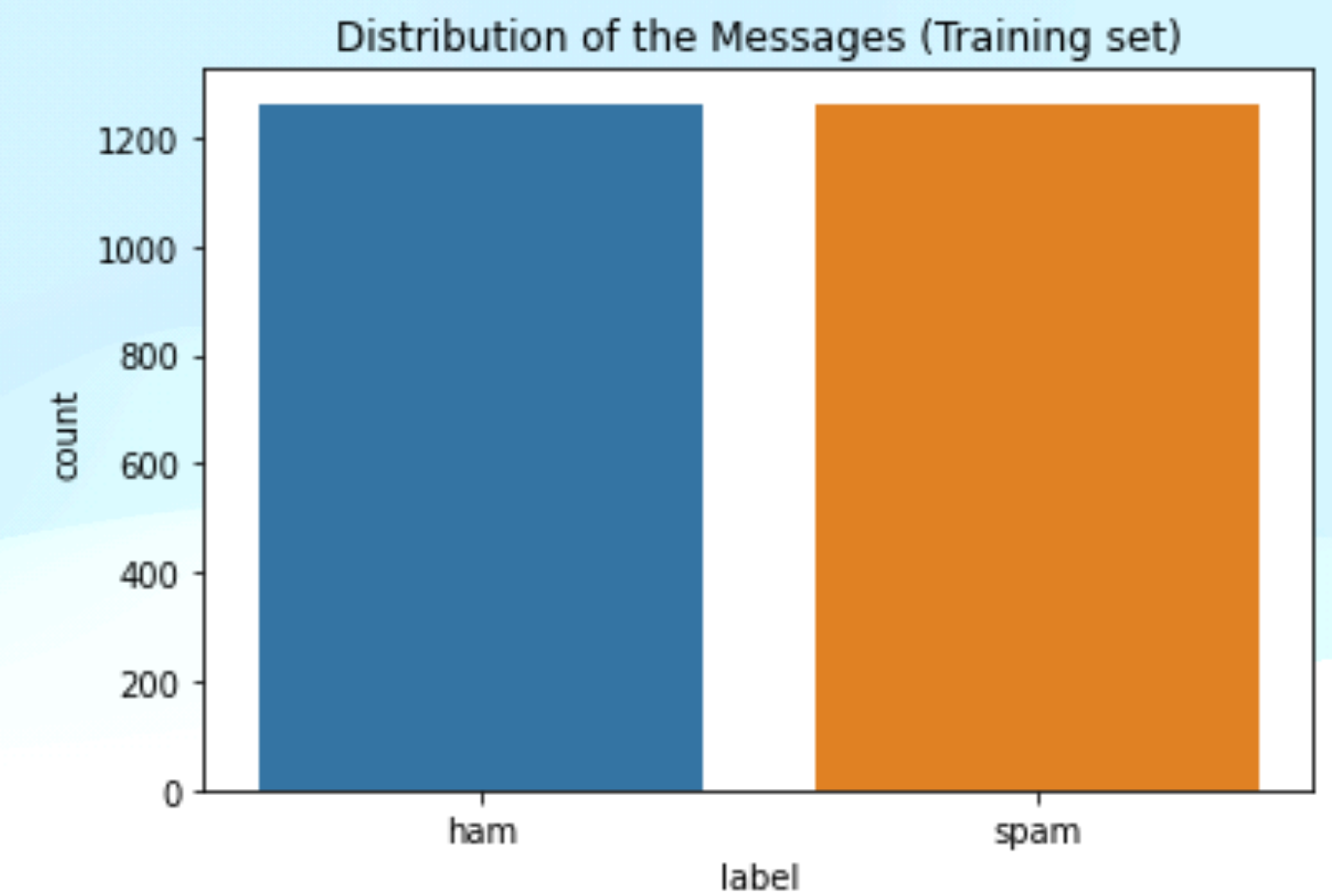
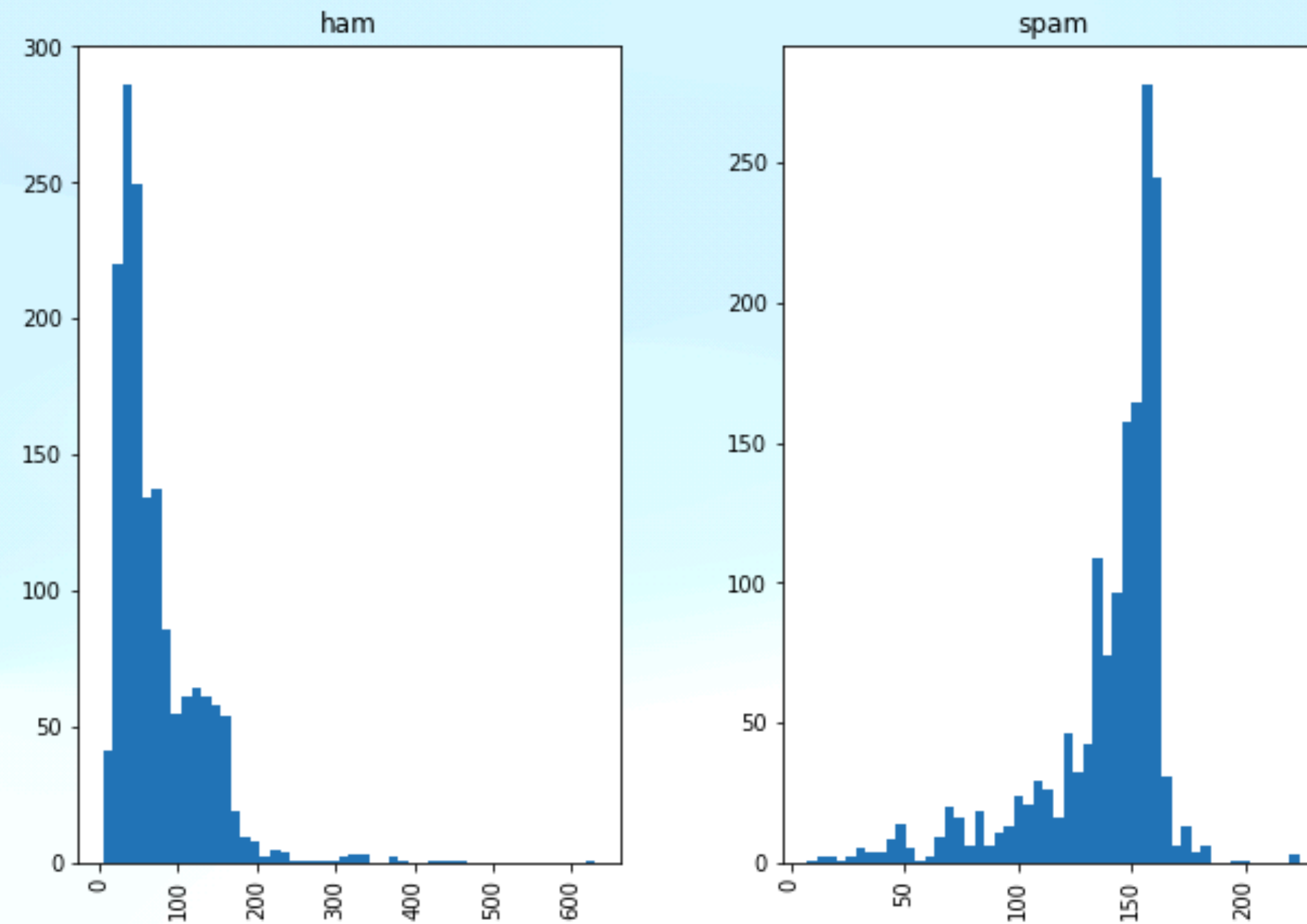
Examining the data



Fixing target imbalance

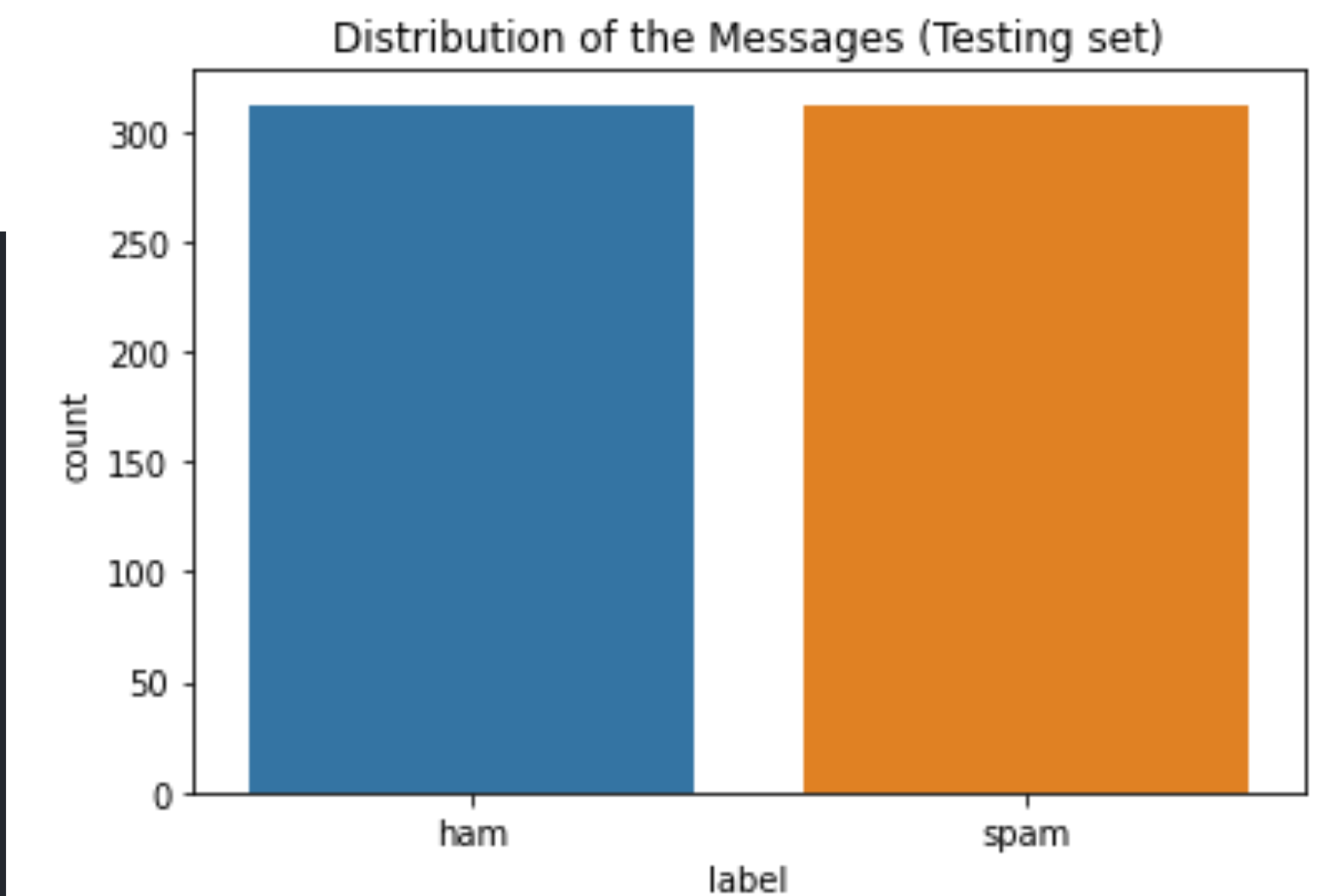
# Main Steps Taken in Preprocessing the Data

Undersampling



Tokenization

```
1 max_tokens = 10000
2 output_sequence_length = avg_token_length
3
4 # create a text vectorization layer
5 text_vectorizer = TextVectorization(max_tokens=max_tokens,
6                                     standardize="lower_and_strip_punctuation",
7                                     split="whitespace",
8                                     ngrams=None,
9                                     output_mode="int",
10                                    output_sequence_length=output_sequence_length)
```

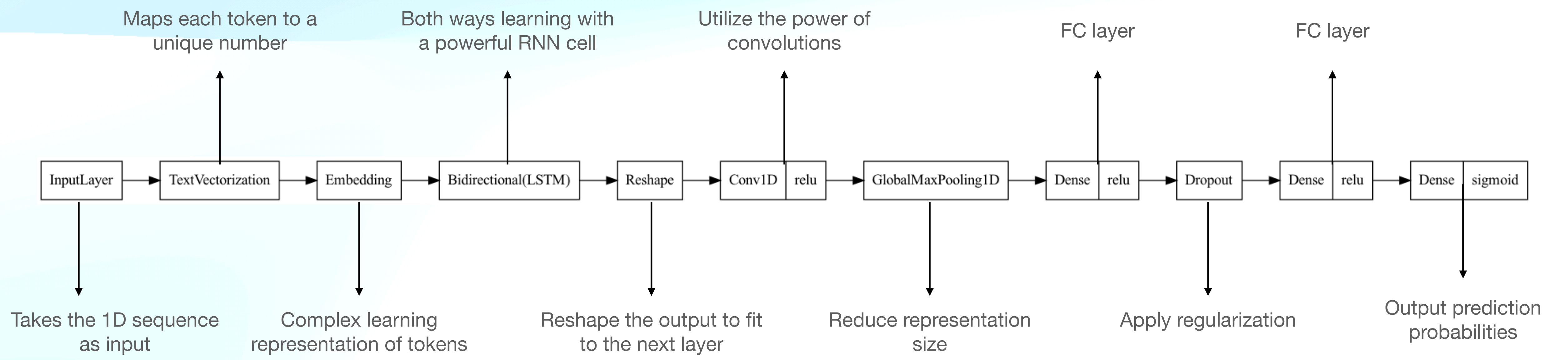




# Model Development

# The Optimal Model

Model	Accuracy	F1-Score	Custom Data
Model 0	0.9599	0.9599	0.7500
Model 1	0.9599	0.9598	0.5000
Model 2	0.9775	0.9775	0.7500
Model 3	0.9855	0.9855	0.7500
Model 4	0.9727	0.9727	0.7500
Model 5	0.9839	0.9839	0.5000
Model 6	0.9743	0.9743	1.000





# Model Deployment

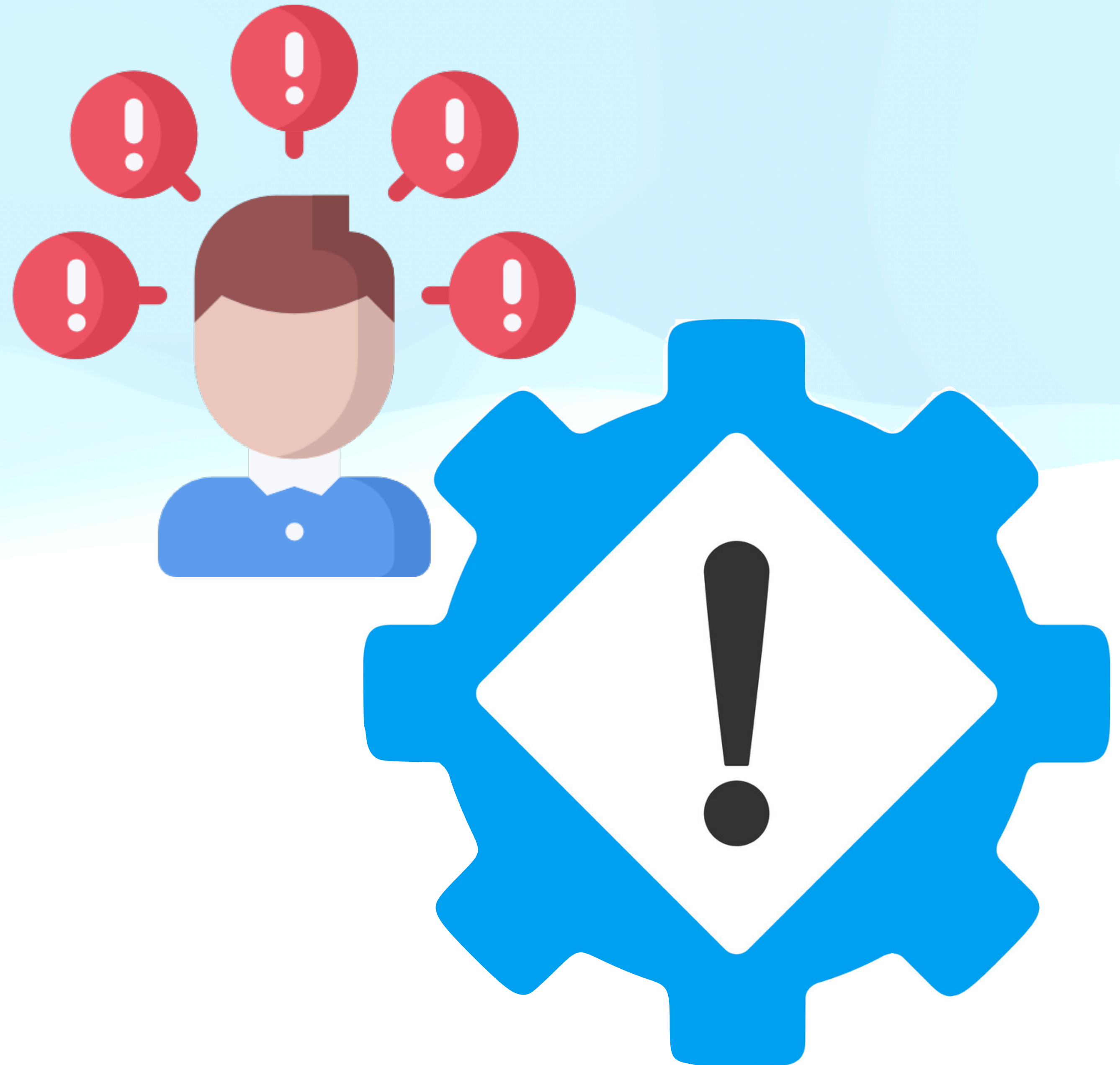




# Problems with Current Solution

# Current Problems

- The training data used doesn't really represent the status of messaging in today's life
- Must have large data for both classes to avoid overfitting
- Model might not always make accurate predictions





# Conclusion

# References

- <https://www.kaggle.com/datasets/uciml/sms-spam-collection-dataset>
- <https://www.kaggle.com/datasets/team-ai/spam-text-message-classification>
- [https://github.com/DeshDSingh/SMS-SPAM-Detection/blob/master/sms\\_spam.csv](https://github.com/DeshDSingh/SMS-SPAM-Detection/blob/master/sms_spam.csv)
- [https://huggingface.co/datasets/sms\\_spam](https://huggingface.co/datasets/sms_spam)
- <https://docs.pyrogram.org/>
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- [https://www.tensorflow.org/api\\_docs/python/tf](https://www.tensorflow.org/api_docs/python/tf)
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# **Thank You**

**All questions and inquires are welcomed**