







Artificial Intelligence Natural Language Processing (NLP)

2025

Represent By

Team Members:

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- SHATHA KAMAL

Group Number: 2

Project Number: 3





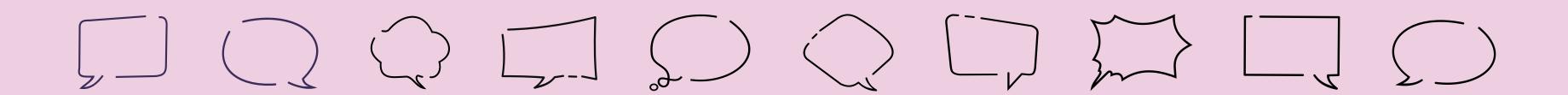


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Natural Language Processing (NLP) is a branch of AI that helps computers understand and analyze human language. It is used in applications like chatbots, sentiment analysis, and text classification.

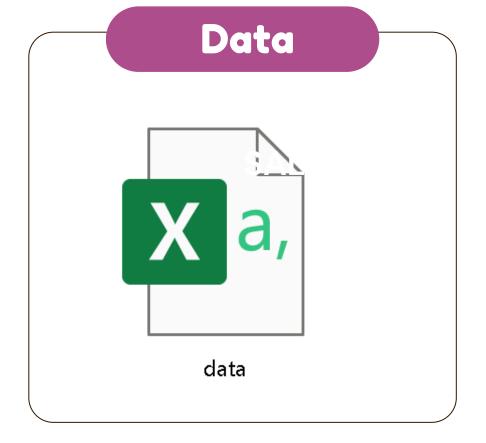


Main Objective of the Project

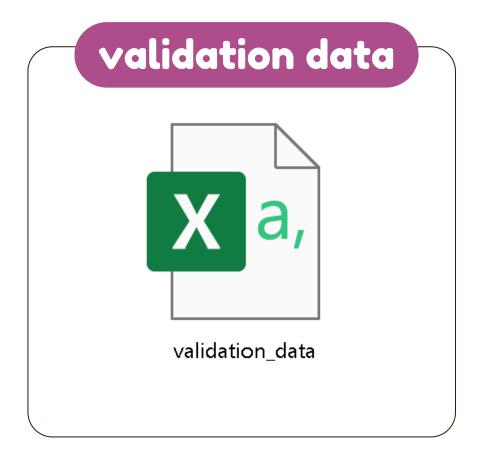
This project involves detecting fake news articles by classifying them into "real" or "fake" categories using machine learning.

Understand Dataset

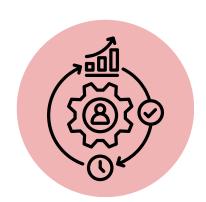








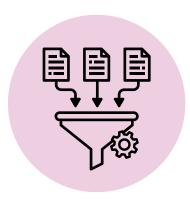
Data Preparation & Pre-processing



1

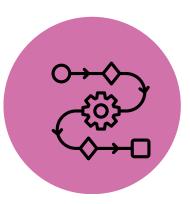
Drop unnecessary columns

Drop unnecessary columns
data = data[['title', 'text', 'label']]
data.dropna(inplace=True)



Text cleaning function

```
# Text cleaning function
def preprocess_text(text):
    #Function to clean and preprocess text
    text = text.lower() # Convert to lowercase
    text = re.sub(r'\d+', '', text) # Remove numbers
    # Remove punctuation
    text = text.translate
    (str.maketrans('', '', string.punctuation))
    # Remove extra spaces
    text = re.sub(r'\s+', '', text).strip()
    return text
```



3

Features and target variable

```
# Features and target variable
X = data['title'] + " " + data['text']
y = data['label']
```

Vectorization Using TF-IDF

Convert text to numerical features

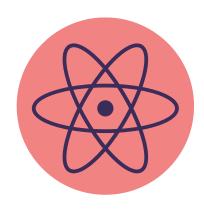
```
vectorizer = TfidfVectorizer(max_features=5000)

X_train_tfidf = vectorizer.fit_transform(X_train)

X_test_tfidf = vectorizer.transform(X_test)
```







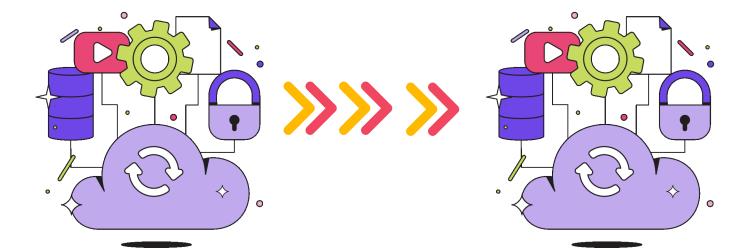
Model Training

Split The Dataset Into Training & Test Sets

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
print(f"Training Data: {len(X_train)} samples")
print(f"Test Data: {len(X_test)} samples")
```

- F or
- A nd
- N or
- B Ut
- o r
- Y et
- SO

Models Training



1 KNN

2 Naive Bayes 3
Logistic
Regression

4 XGBoost

S Random Forest

KNeighborsClassifi eighborsClassifier(

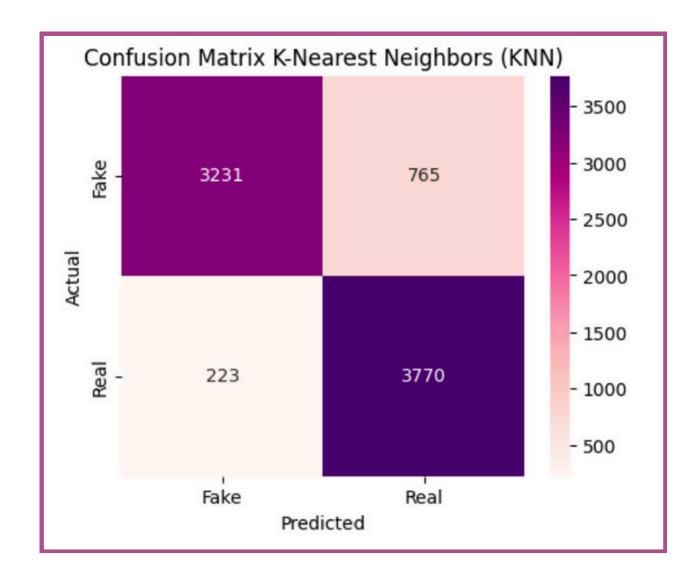
• MultinomialNB MultinomialNB() LogisticRegression()

XGBClassifier
booster=None, call
=None, colsample b

RandomForestClassif: domForestClassifier

1 KNN

Confusion matrix



Accuracy Calculation

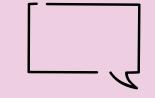
Accuracy K-Nearest Neighbors (KNN): 87.63%





















Confusion matrix
2

| The state of the

Confusion Matrix Naive Bayes

Confusion matrix

Accuracy Naive Bayes: 93.85%

Predicted

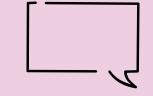


Naive Bayes

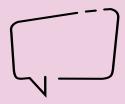












Fake



Real



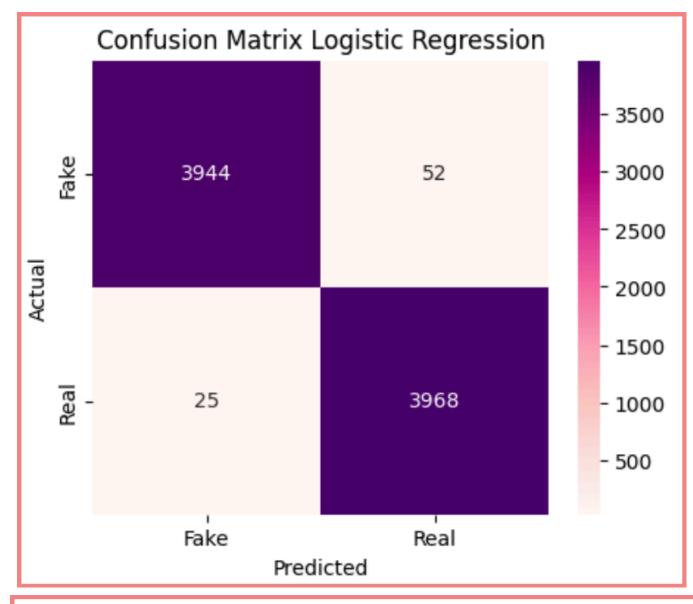


3

Logistic Regression

Confusion matrix

Accuracy Calculation



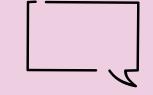
Accuracy Logistic Regression: 99.0362%







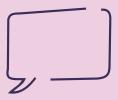










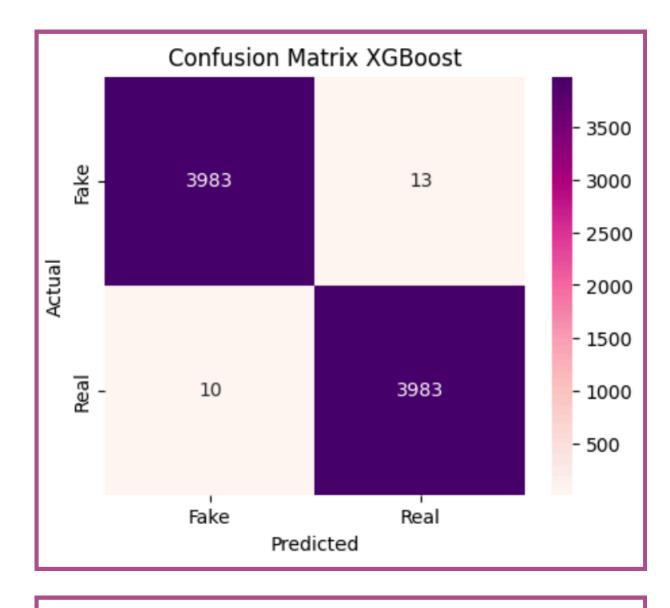




4 XGBoost

Confusion matrix

Accuracy Calculation



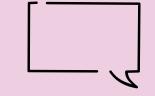
Accuracy XGBoost: 99.71%



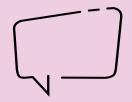














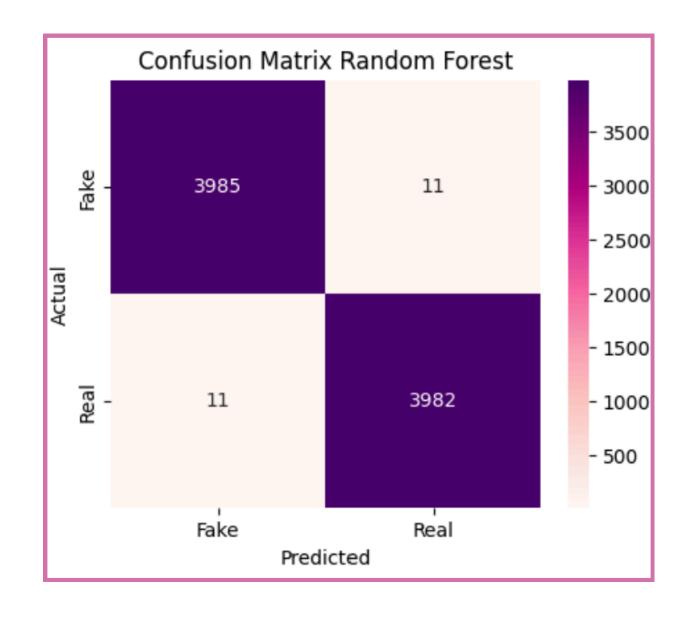




5
Random Forest

Confusion matrix

Accuracy Calculation



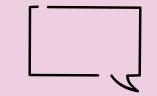
Accuracy Random Forest: 99.72%



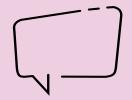


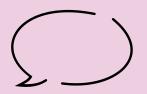








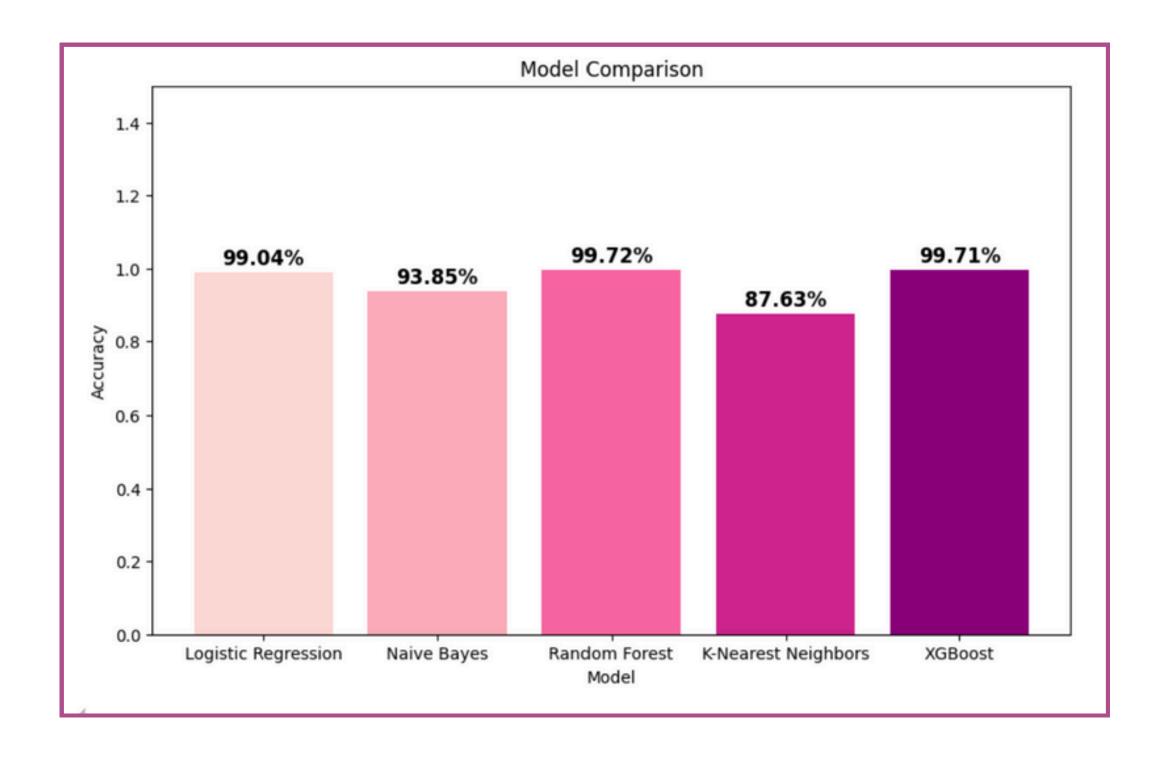








Models Comparison

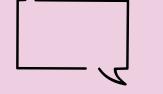




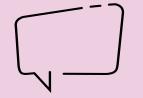










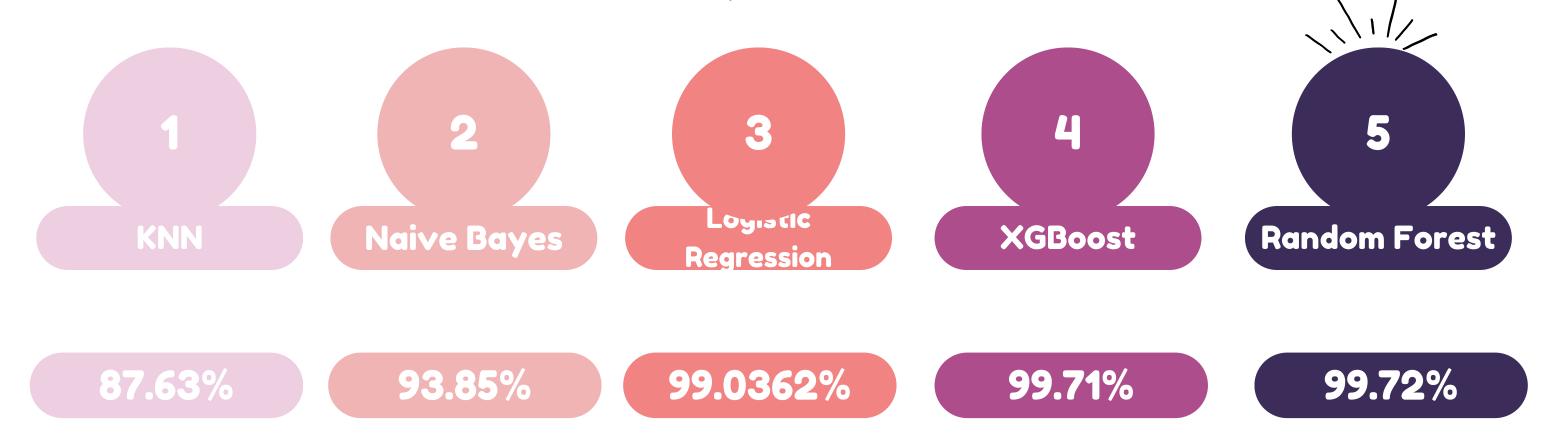


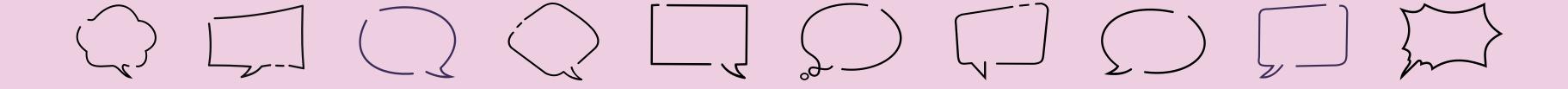


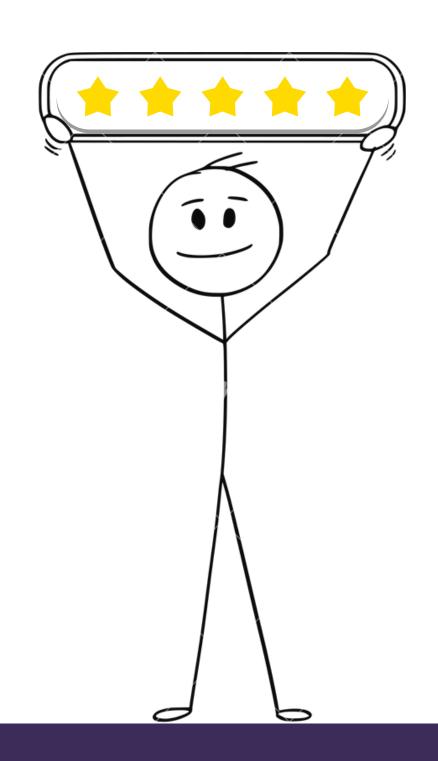




Accuracy Calculation







Best Model Selection



The best-performing model is: "Random Forest"



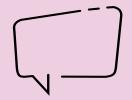


















predictions on validation data

```
# Make predictions using the best model
predictions = model.predict(X_validation_tfidf)
# Add the predictions to the validation dataframe
validation_data['label'] = predictions
```





















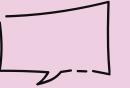
Save the result to a new CSV

```
# Save predictions to CSV
validation_data.to_csv('validation_predictions.csv', index=False)
print("Predictions saved to 'validation_predictions.csv'")
# Display the first few rows of the validation data with predictions
validation_data.head()
```

Predictions saved to 'validation_predictions.csv'

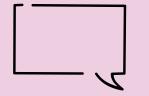
	label	title	text
0	1	UK's May 'receiving regular updates' on London	LONDON (Reuters) - British Prime Minister Ther
1	1	UK transport police leading investigation of L	LONDON (Reuters) - British counter-terrorism p
2	1	Pacific nations crack down on North Korean shi	WELLINGTON (Reuters) - South Pacific island na
3	1	Three suspected al Qaeda militants killed in Y	ADEN, Yemen (Reuters) - Three suspected al Qae
4	1	Chinese academics prod Beijing to consider Nor	BEIJING (Reuters) - Chinese academics are publ

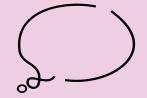














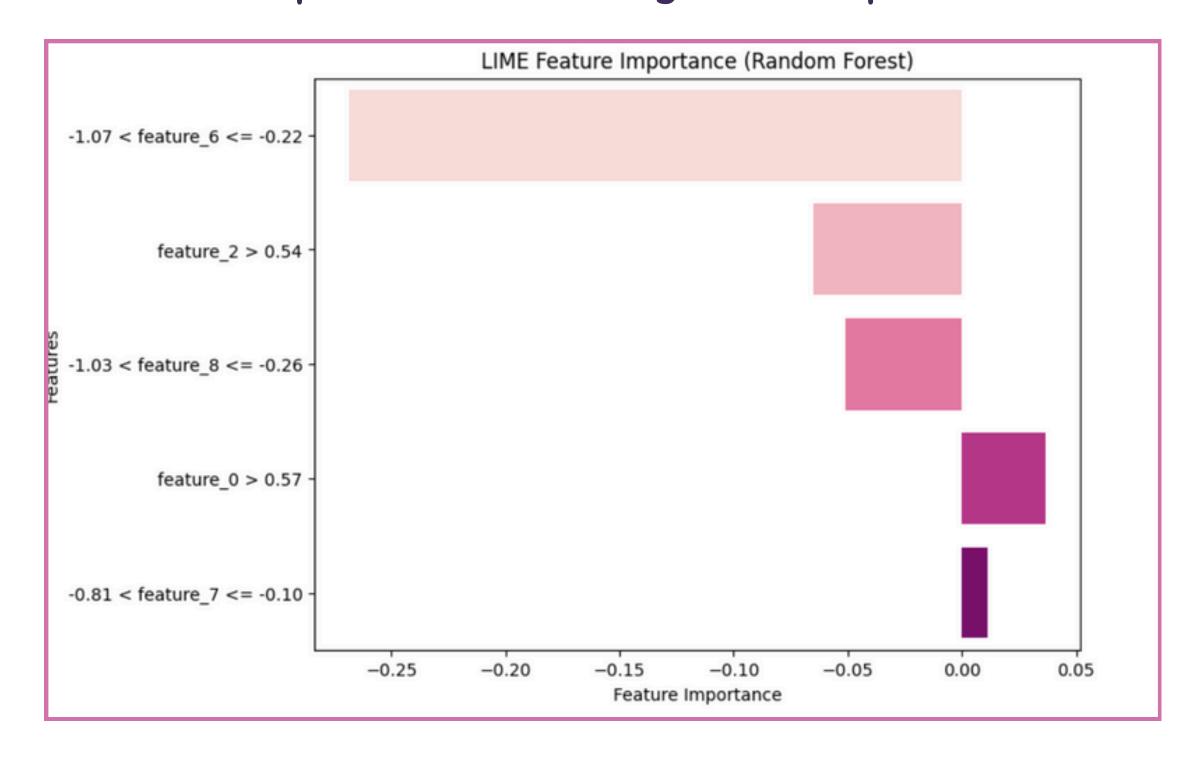






Explain Predictions with LIME

Local Interpretable Model-agnostic Explanations



Challenges

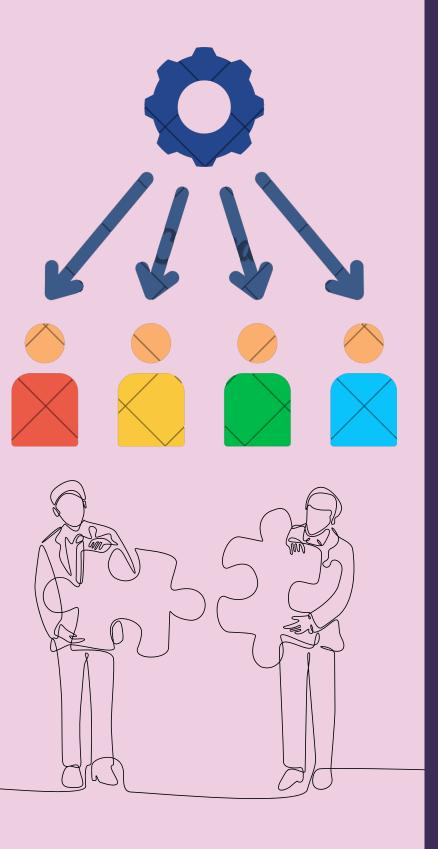
Solutions

Challenges

8

Solutions

1. Difficulty in selecting the best model	We tested multiple models and compared their accuracy to find the best-performing one.
2. Understanding model predictions	We used LIME (Local Interpretable Model-agnostic Explanations) to interpret model decisions.
3. Balancing workload among team members	We assigned tasks based on each member's expertise and skill set.
4. Time management and meeting deadlines	We utilized Slack to track progress and ensure efficient collaboration.



Organizing Labour Division in Our Team Strategies & Implementation

Ghala AlOtaibi

KNN - Logistic Regression - XGBoost

Hanan Alnbhani

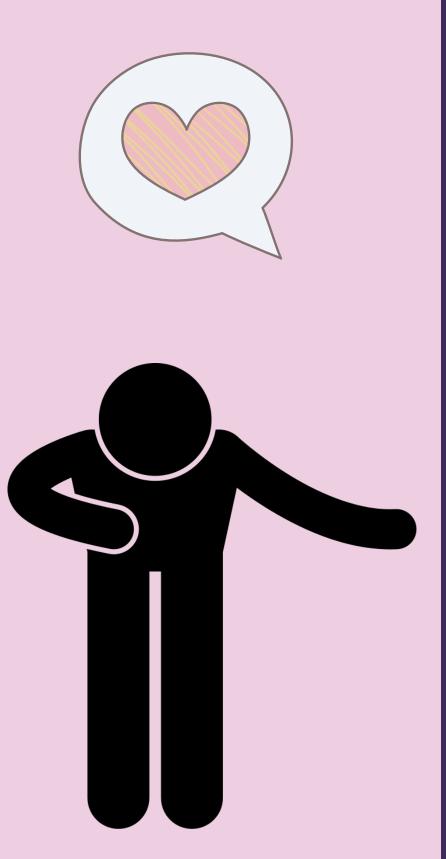
Presentation -LIME

Sarah Alqahtani

Random Forest

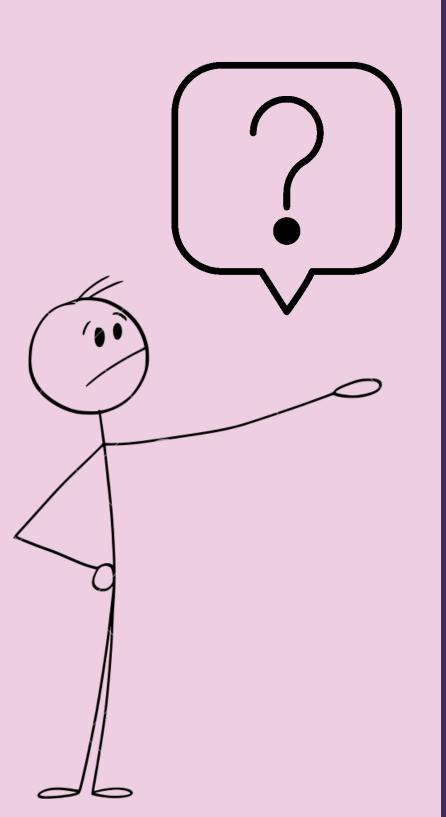
SHATHA KAMAL

Naive Bayes - Read Me



Thank you

For your kind listening



ANY tions?