```
In [1]: # Import necessary libraries
        import subprocess
        import sys
        import argparse
        import json
        import os
        import gc
        import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        import PyPDF2
        import re
        import nltk
        import emoji
        from nltk.sentiment.vader import SentimentIntensityAnalyzer
        from nltk.tokenize import sent_tokenize, word_tokenize
        from nltk.stem import WordNetLemmatizer
        from nltk.corpus import stopwords
        from sklearn.feature extraction.text import TfidfVectorizer
        from sklearn.ensemble import AdaBoostClassifier
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import accuracy score, classification report, confusion matrix
        from imblearn.over_sampling import SMOTE
        from imblearn.under sampling import RandomUnderSampler
        from imblearn.pipeline import Pipeline
        import logging
        import traceback
        import numpy as np
        from sklearn.svm import SVC
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.naive_bayes import MultinomialNB
        from sklearn.linear_model import LogisticRegression
        from sklearn.neural network import MLPClassifier
        from sklearn.model_selection import cross_val_score
```

```
In [2]: # Download necessary NLTK data
  nltk.download('punkt')
  nltk.download('wordnet')
  nltk.download('stopwords')
  nltk.download('vader_lexicon')
  nltk.download('omw-1.4')
```

```
[nltk_data] Downloading package punkt to
[nltk_data]
               C:\Users\ELITEBOOK\AppData\Roaming\nltk_data...
[nltk_data] Package punkt is already up-to-date!
[nltk_data] Downloading package wordnet to
               C:\Users\ELITEBOOK\AppData\Roaming\nltk_data...
[nltk_data]
[nltk_data]
             Package wordnet is already up-to-date!
[nltk_data] Downloading package stopwords to
[nltk data]
               C:\Users\ELITEBOOK\AppData\Roaming\nltk data...
[nltk_data] Package stopwords is already up-to-date!
[nltk_data] Downloading package vader_lexicon to
[nltk_data]
               C:\Users\ELITEBOOK\AppData\Roaming\nltk_data...
[nltk_data] Package vader_lexicon is already up-to-date!
[nltk_data] Downloading package omw-1.4 to
               C:\Users\ELITEBOOK\AppData\Roaming\nltk_data...
[nltk_data]
[nltk data] Package omw-1.4 is already up-to-date!
```

Out[2]: True

```
In [3]: logging.basicConfig(level=logging.INFO)
        logger = logging.getLogger(__name__)
        # Function to extract sentences from PDFs using PyPDF2
        def read pdf sentences(file path):
            sentences = []
            try:
                with open(file_path, "rb") as file:
                    reader = PyPDF2.PdfReader(file)
                    for page in reader.pages:
                         text = page.extract_text()
                         if text:
                             sentences.extend(sent_tokenize(text))
            except Exception as e:
                print(f"Error reading PDF file {file path}: {str(e)}")
            return sentences
        def extract_and_merge(pdf_path, csv_path):
                print(f"Attempting to read PDF files from: {pdf path}")
                if not os.path.exists(pdf path):
                     raise FileNotFoundError(f"PDF directory not found: {pdf_path}")
                pdf_files = [os.path.join(pdf_path, file) for file in os.listdir(pdf_path) if
                print(f"Found {len(pdf_files)} PDF files")
                pdf_sentences = []
                for file in pdf files:
                     pdf sentences.extend(read pdf sentences(file))
                print(f"Extracted {len(pdf sentences)} sentences from PDF files")
                pdf_df = pd.DataFrame({'content': pdf_sentences})
                print(f"Attempting to read CSV file: {csv_path}")
                if not os.path.exists(csv_path):
                     raise FileNotFoundError(f"CSV file not found: {csv_path}")
                 news_data = pd.read_csv(csv_path, encoding='latin1')
                 content_column = next((col for col in news_data.columns if col.lower().strip()
                if content column is None:
                     raise KeyError(f"No 'content' column found in the CSV file: {csv_path}")
```

```
news_data_paragraphs = []
    for content in news_data[content_column].dropna():
        paragraphs = content.split('\n\n')
        news_data_paragraphs.extend(paragraphs)
    print(f"Extracted {len(news_data_paragraphs)} paragraphs from CSV file")
    news_df = pd.DataFrame({'content': news_data_paragraphs})
    merged_data = pd.concat([pdf_df, news_df], ignore_index=True)
    print(f"Merged data shape: {merged_data.shape}")
    return merged_data
except Exception as e:
    print(f"Error in extract and merge: {str(e)}")
    print(f"Current working directory: {os.getcwd()}")
   print(f"Contents of current directory: {os.listdir('.')}")
    if os.path.exists(pdf_path):
        print(f"Contents of PDF directory: {os.listdir(pdf path)}")
    raise
```

```
In [4]: # Assign Sentiment Analyzer Score
        sid = SentimentIntensityAnalyzer()
        def assign sentiment scores(text):
            scores = sid.polarity_scores(text)
            return scores['compound']
        def assign scores(data):
            data['sentiment'] = data['content'].apply(assign sentiment scores)
            return data
        # Function to assign direction and new_direction based on sentiment scores
        def assign directions(data):
            data['direction'] = data['sentiment'].apply(lambda x: 'bearish' if x < 0.0 else (</pre>
            data['new\_direction'] = data['sentiment'].apply(lambda x: 2 if x < 0.0 else (1 if
            return data
        # Function to preprocess individual text
        def preprocess_text(text):
            lemmatizer = WordNetLemmatizer()
            stop_words = set(stopwords.words('english'))
            # Lowercase the text
            text = text.lower()
            # Remove emojis
            text = emoji.replace emoji(text, '')
            # Remove emoticons (this is a basic implementation, might need refinement)
            text = re.sub(r'[:;=]-?[()DPp]', '', text)
            # Remove punctuation and numbers
            text = re.sub(r'[^\w\s]', '', text)
            text = re.sub(r'\d+', '', text)
            # Remove extra spaces
            text = re.sub(r'\s+', ' ', text).strip()
```

```
# Tokenize
   tokens = word tokenize(text)
   # Remove stop words and Lemmatize
   tokens = [lemmatizer.lemmatize(word) for word in tokens if word not in stop_words]
   try:
       tokens = [lemmatizer.lemmatize(word) for word in tokens if word not in stop_wd
   except LookupError:
       # If Lemmatization fails, just use the original tokens
        tokens = [word for word in tokens if word not in stop_words]
   return ' '.join(tokens)
# Function to preprocess the entire DataFrame
def preprocess data(df):
   df cleaned = df.copy()
   df_cleaned['content'] = df_cleaned['content'].apply(preprocess_text)
   return df cleaned
# Count the number of bearish, bullish, and neutral sentiments
def sentiment_counts(data):
   return data['direction'].value counts()
```

```
In [5]: # Prepare Dataset Function
        def prepare_dataset(data, sample_frac=0.1, random_state=42):
            print("Preparing dataset...")
            data = data.sample(frac=sample_frac, random_state=random_state).reset_index(drop=1
            X = data['content']
            y = data['new_direction']
            # TF-IDF Vectorization
            vectorizer = TfidfVectorizer(max features=5000)
            X = vectorizer.fit transform(X)
            # Split the data
            X train, X test, y train, y test = train test split(X, y, test size=0.2, random st
            # Define resampling strategy
            over = SMOTE(sampling_strategy='auto', random_state=random_state)
            under = RandomUnderSampler(sampling_strategy='auto', random_state=random_state)
            # Create a pipeline with SMOTE and RandomUnderSampler
            resampling = Pipeline([('over', over), ('under', under)])
            # Apply resampling
            X train resampled, y train resampled = resampling.fit resample(X train, y train)
            print(f"Dataset prepared with train size: {X_train_resampled.shape[0]} and test si
            return X_train_resampled, X_test, y_train_resampled, y_test, vectorizer
```

```
import sklearn
def train_and_evaluate_multiple_models(X_train, X_test, y_train, y_test):
    print("Training and evaluating multiple models...")

models = {
        'AdaBoost': AdaBoostClassifier(estimator=DecisionTreeClassifier(max_depth=3),
        'SVM': SVC(kernel='rbf', random_state=42),
```

```
'Random Forest': RandomForestClassifier(n estimators=100, random state=42),
        'Naive Bayes': MultinomialNB(),
        'Logistic Regression': LogisticRegression(random_state=42),
        'Neural Network': MLPClassifier(hidden_layer_sizes=(100,), max_iter=500, randc
    }
    results = {}
    try:
        for name, model in models.items():
            print(f"Training and evaluating {name}...")
            # Train the model
            model.fit(X_train, y_train)
            # Make predictions
            y pred = model.predict(X test)
            # Calculate accuracy
            accuracy = accuracy_score(y_test, y_pred)
            # Generate classification report
            report = classification_report(y_test, y_pred, target_names=['bullish', 'r
            report df = pd.DataFrame(report).transpose()
            # Generate confusion matrix
            cm = confusion_matrix(y_test, y_pred)
            # Perform cross-validation
            cv_scores = cross_val_score(model, X_train, y_train, cv=5)
            results[name] = {
                'model': model,
                'accuracy': accuracy,
                'report': report_df,
                'confusion matrix': cm,
                'cv_scores': cv_scores
            }
            print(f"{name} - Accuracy: {accuracy}, Cross-validation mean score: {np.me
        return results
    except Exception as e:
        print(f"An error occurred in train_and_evaluate_multiple_models: {str(e)}")
        print(f"Error details: {traceback.format_exc()}")
        return None
def create comprehensive report multiple models(company name, results):
```

```
In [7]: def create_comprehensive_report_multiple_models(company_name, results):
    report_data = {'Company': company_name}

for model_name, model_results in results.items():
    report_data[f'{model_name} Accuracy'] = model_results['accuracy']
    report_data[f'{model_name} CV Mean Score'] = np.mean(model_results['cv_scores' report_data[f'{model_name} CV Std Score'] = np.std(model_results['cv_scores'])
    report_data[f'{model_name} Confusion Matrix'] = model_results['confusion_matri

    for class_name in ['bullish', 'neutral', 'bearish']:
        if class_name in model_results['report'].index:
```

```
report_data[f'{model_name} Precision ({class_name})'] = model_results[
    report_data[f'{model_name} Recall ({class_name})'] = model_results['re
    report_data[f'{model_name} F1-Score ({class_name})'] = model_results['
    return pd.DataFrame([report_data])
```

```
In [8]:
        def main(company_name, pdf_path, csv_path):
            try:
                 logger.info(f"Processing {company_name}...")
                # Load and preprocess data
                 raw_data = extract_and_merge(pdf_path, csv_path)
                 data_with_sentiment = assign_scores(raw_data)
                 data_with_directions = assign_directions(data_with_sentiment)
                 cleaned data = preprocess data(data with directions)
                 # Display sentiment counts
                 counts = sentiment_counts(cleaned_data)
                 logger.info(f"{company_name} Sentiment Counts:")
                 logger.info(counts)
                # Prepare dataset
                X_train, X_test, y_train, y_test, vectorizer = prepare_dataset(cleaned_data)
                 # Train and evaluate multiple models
                results = train_and_evaluate_multiple_models(X_train, X_test, y_train, y_test)
                 if results is None:
                    logger.error(f"Training and evaluation failed for {company name}")
                    return None
                 # Create comprehensive report
                 comprehensive_report = create_comprehensive_report_multiple_models(company_nam
                 return comprehensive_report
            except Exception as e:
                 logger.error(f"Error processing {company_name}: {str(e)}")
                 logger.error(traceback.format_exc())
                 return None
```

```
In [9]: if __name__ == "__main__":
             # Define paths for each company
             companies = {
                 'Lloyds': {
                     'pdf_path': 'data/lloyds',
                     'csv_path': 'data/lloyds/lloyds_news.csv'
                 },
                  'IAG': {
                     'pdf path': 'data/iag',
                     'csv_path': 'data/iag/iag_news.csv'
                 },
                 'Vodafone': {
                     'pdf_path': 'data/vodafone',
                      'csv_path': 'data/vodafone/vodafone_news.csv'
                 }
             }
             all reports = []
```

```
for company name, paths in companies.items():
        try:
            logger.info(f"Starting processing for {company_name}")
            company_report = main(company_name, paths['pdf_path'], paths['csv_path'])
            if company report is not None:
                all_reports.append(company_report)
        except Exception as e:
            logger.error(f"Failed to process {company_name}: {str(e)}")
    # Combine all reports into a single DataFrame
    if all_reports:
        combined_report = pd.concat(all_reports, ignore_index=True)
        combined report to csv('comprehensive classification report adaboost.csv', inc
        logger.info("Comprehensive classification report for all companies saved to CS
    else:
        logger.warning("No reports were generated.")
INFO: __main__:Starting processing for Lloyds
INFO:__main__:Processing Lloyds...
Attempting to read PDF files from: data/lloyds
Found 20 PDF files
Extracted 66875 sentences from PDF files
Attempting to read CSV file: data/lloyds/lloyds_news.csv
Extracted 1834 paragraphs from CSV file
Merged data shape: (68709, 1)
INFO:__main__:Lloyds Sentiment Counts:
INFO: main :direction
neutral
          27490
bullish
           26909
bearish
          14310
Name: count, dtype: int64
Preparing dataset...
Dataset prepared with train size: 6618 and test size: 1375
Training and evaluating multiple models...
Training and evaluating AdaBoost...
```

```
C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble\_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(
C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble\ weight boosting.py:5
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27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(
AdaBoost - Accuracy: 0.7258181818181818, Cross-validation mean score: 0.7582316578863
839
Training and evaluating SVM...
SVM - Accuracy: 0.79345454545454545, Cross-validation mean score: 0.8396774016756753
Training and evaluating Random Forest...
Random Forest - Accuracy: 0.79781818181819, Cross-validation mean score: 0.84511912
18347024
Training and evaluating Naive Bayes...
Naive Bayes - Accuracy: 0.6567272727272727, Cross-validation mean score: 0.7064107482
Training and evaluating Logistic Regression...
Logistic Regression - Accuracy: 0.77454545454545, Cross-validation mean score: 0.81
17239040631358
Training and evaluating Neural Network...
INFO: __main__:Starting processing for IAG
INFO:__main__:Processing IAG...
Neural Network - Accuracy: 0.7629090909090909, Cross-validation mean score: 0.8188256
571510779
Attempting to read PDF files from: data/iag
Found 11 PDF files
Extracted 34291 sentences from PDF files
Attempting to read CSV file: data/iag/iag news.csv
Extracted 2037 paragraphs from CSV file
Merged data shape: (36328, 1)
INFO: __main__:IAG Sentiment Counts:
INFO: __main__:direction
neutral
         17607
bullish
          12229
bearish
           6492
Name: count, dtype: int64
C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble\_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(
```

```
fin_sent_adaboost-multiple_models
Preparing dataset...
Dataset prepared with train size: 4182 and test size: 727
Training and evaluating multiple models...
Training and evaluating AdaBoost...
C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble\ weight boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(
C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble\_weight_boosting.py:5
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C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble\ weight boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(
AdaBoost - Accuracy: 0.7042640990371389, Cross-validation mean score: 0.7400842036665
Training and evaluating SVM...
SVM - Accuracy: 0.7372764786795049, Cross-validation mean score: 0.8804462279844284
Training and evaluating Random Forest...
Random Forest - Accuracy: 0.781292984869326, Cross-validation mean score: 0.863952198
8418423
Training and evaluating Naive Bayes...
Naive Bayes - Accuracy: 0.657496561210454, Cross-validation mean score: 0.76136520839
Training and evaluating Logistic Regression...
Logistic Regression - Accuracy: 0.7359009628610729, Cross-validation mean score: 0.83
88385839149846
Training and evaluating Neural Network...
INFO: __main__:Starting processing for Vodafone
INFO:__main__:Processing Vodafone...
Neural Network - Accuracy: 0.71939477303989, Cross-validation mean score: 0.857009540
7956189
Attempting to read PDF files from: data/vodafone
Found 14 PDF files
Extracted 51164 sentences from PDF files
Attempting to read CSV file: data/vodafone/vodafone news.csv
Extracted 0 paragraphs from CSV file
Merged data shape: (51164, 1)
INFO:__main__:Vodafone Sentiment Counts:
INFO: __main__:direction
neutral 24998
bullish
          18868
bearish
           7298
Name: count, dtype: int64
Preparing dataset...
Dataset prepared with train size: 5721 and test size: 1024
```

Training and evaluating multiple models...

Training and evaluating AdaBoost...

C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(

C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
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 warnings.warn(

C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(

C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
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 warnings.warn(

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ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(

C:\Users\ELITEBOOK\anaconda3\Lib\site-packages\sklearn\ensemble_weight_boosting.py:5
27: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be remo
ved in 1.6. Use the SAMME algorithm to circumvent this warning.
 warnings.warn(

AdaBoost - Accuracy: 0.7734375, Cross-validation mean score: 0.7722460072678413 Training and evaluating SVM...

SVM - Accuracy: 0.802734375, Cross-validation mean score: 0.8739760894127707 Training and evaluating Random Forest...

Random Forest - Accuracy: 0.8359375, Cross-validation mean score: 0.8783462607261734 Training and evaluating Naive Bayes...

Naive Bayes - Accuracy: 0.6748046875, Cross-validation mean score: 0.7545897334106941 Training and evaluating Logistic Regression...

Logistic Regression - Accuracy: 0.7861328125, Cross-validation mean score: 0.83621843 22227991

Training and evaluating Neural Network...

INFO:__main__:Comprehensive classification report for all companies saved to CSV.
Neural Network - Accuracy: 0.7822265625, Cross-validation mean score: 0.8680312700400
036