```
import praw
import pandas as pd
import nltk
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
from sklearn.ensemble import RandomForestClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import classification report
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
from sklearn.model selection import train test split
import joblib
import warnings
WARNING:tensorflow:From C:\Users\Hp\anaconda3\lib\site-packages\keras\
src\losses.py:2976: The name tf.losses.sparse softmax cross entropy is
deprecated. Please use
tf.compat.v1.losses.sparse_softmax_cross entropy instead.
warnings.filterwarnings("ignore")
nltk.download("stopwords")
nltk.download("wordnet")
[nltk data] Downloading package stopwords to
[nltk data]
                C:\Users\Hp\AppData\Roaming\nltk data...
[nltk data]
              Package stopwords is already up-to-date!
[nltk_data] Downloading package wordnet to
[nltk data]
                C:\Users\Hp\AppData\Roaming\nltk data...
[nltk data]
              Package wordnet is already up-to-date!
True
# Authentication Function
def authenticate reddit(client id, client secret, user agent):
    try:
        reddit = praw.Reddit(
            client id=client id,
            client secret=client secret,
            user_agent=user agent,
        print("Authenticated successfully!")
        return reddit
    except Exception as e:
        print("Authentication failed:", e)
        return None
# Data Scraping
def scrape reddit data(reddit, subreddit name, keyword=None,
limit=100):
```

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try:
        subreddit = reddit.subreddit(subreddit name)
        posts = subreddit.search(keyword, limit=limit) if keyword else
subreddit.hot(limit=limit)
        data = []
        for post in posts:
            post.comments.replace more(limit=0)
            comments = " ".join([comment.body for comment in
post.comments.list()])
            data.append({
                "Title": post.title,
                "Comments": comments,
                "Score": post.score,
                "Sentiment Text": post.title + " " + comments,
                "Created At": pd.to datetime(post.created utc,
unit="s"),
            })
        return pd.DataFrame(data)
    except Exception as e:
        print("Error during scraping:", e)
        return pd.DataFrame()
# Text Preprocessing
def preprocess text(df):
    stop words = set(stopwords.words("english"))
    lemmatizer = WordNetLemmatizer()
    df["Cleaned Text"] = df["Sentiment Text"].str.lower()
    df["Cleaned Text"] = df["Cleaned Text"].apply(
        lambda x: " ".join([lemmatizer.lemmatize(word) for word in
x.split() if word not in stop words])
    return df
# Sentiment Analysis
def analyze sentiment(df):
    analyzer = SentimentIntensityAnalyzer()
    df["Sentiment Score"] = df["Cleaned Text"].apply(lambda x:
analyzer.polarity scores(x)["compound"])
    return df
# Feature Engineering
def add features(df):
    df["Hour"] = df["Created At"].dt.hour
    df["Day"] = df["Created At"].dt.day_name()
    df["Comment Length"] = df["Comments"].str.len()
    return df
```

```
# Model Training
def train hybrid model(df):
    \# Define target (upward movement if sentiment score > 0.2)
    df["Stock Movement"] = (df["Sentiment Score"] > 0.2).astype(int)
    # Features and target
    features = ["Sentiment Score", "Score", "Comment Length"]
    X = df[features]
    y = df["Stock Movement"]
    # Split and normalize data
    X_train, X_test, y_train, y_test = train_test_split(X, y,
test size=0.2, random state=42)
    \overline{\text{scaler}} = \text{Standard} \overline{\text{Scaler}}
    X train = scaler.fit transform(X train)
    X test = scaler.transform(X test)
    # Random Forest for feature importance
    rf model = RandomForestClassifier()
    rf model.fit(X train, y train)
    print("Random Forest Report:")
    print(classification report(y test, rf model.predict(X test)))
    # LSTM for sequential data
    X train lstm = X train.reshape((X train.shape[0], 1,
X train.shape[1]))
    X_test_lstm = X_test.reshape((X_test.shape[0], 1,
X test.shape[1]))
    lstm model = Sequential([
        LSTM(50, input shape=(X train lstm.shape[1],
X train lstm.shape[2])),
        Dense(1, activation="sigmoid"),
    lstm model.compile(optimizer="adam", loss="binary crossentropy",
metrics=["accuracy"])
    lstm model.fit(X_train_lstm, y_train, epochs=10, batch_size=16,
verbose=1)
    print("LSTM Evaluation:")
    lstm_model.evaluate(X_test_lstm, y_test)
    return rf model, lstm model, scaler
# Saving Models
def save_models(rf_model, lstm_model, scaler):
    joblib.dump(rf_model, "random_forest_model.pkl")
    joblib.dump(scaler, "scaler.pkl")
    lstm_model.save("lstm_model.h5")
    print("Models saved successfully.")
```

```
def main():
    # Reddit API credentials
    client id = "q9C9 B3Km7S9L4rofTLUvw"
    client secret = "fy7yX-94w8usjBQaxspmrbCb dHD5A"
    user agent = "AdditionalCoast770"
    # Authenticating with Reddit
    reddit = authenticate reddit(client id, client secret, user agent)
    # Scrape data from Reddit
    data = scrape reddit data(reddit, "StockMarket", keyword="SAIL",
limit=100)
    if not data.empty:
        # Preprocess text data
        data = preprocess text(data)
        # Perform sentiment analysis
        data = analyze sentiment(data)
        # Add features for prediction
        data = add features(data)
        # Train hybrid model
        rf model, lstm model, scaler = train hybrid model(data)
        # Save trained models and scaler
        save models(rf model, lstm model, scaler)
if name == " main ":
    main()
Authenticated successfully!
Random Forest Report:
              precision recall f1-score
                                              support
                                                    2
           0
                   1.00
                             1.00
                                       1.00
           1
                   1.00
                             1.00
                                       1.00
                                                   12
                                                   14
    accuracy
                                       1.00
                             1.00
                                       1.00
                                                   14
   macro avq
                   1.00
weighted avg
                   1.00
                             1.00
                                       1.00
                                                   14
WARNING:tensorflow:From C:\Users\Hp\anaconda3\lib\site-packages\keras\
src\layers\rnn\lstm.py:148: The name
tf.executing eagerly outside functions is deprecated. Please use
tf.compat.vl.executing eagerly outside functions instead.
WARNING:tensorflow:From C:\Users\Hp\anaconda3\lib\site-packages\keras\
src\optimizers\ init .py:309: The name tf.train.Optimizer is
deprecated. Please use tf.compat.vl.train.Optimizer instead.
```

WARNING:tensorflow:From C:\Users\Hp\anaconda3\lib\site-packages\keras\

Epoch 1/10

```
src\utils\tf utils.py:492: The name tf.ragged.RaggedTensorValue is
deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.
WARNING:tensorflow:From C:\Users\Hp\anaconda3\lib\site-packages\keras\
src\engine\base layer utils.py:384: The name
tf.executing eagerly outside functions is deprecated. Please use
tf.compat.vl.executing eagerly outside functions instead.
accuracy: 0.8113
Epoch 2/10
accuracy: 0.8491
Epoch 3/10
accuracy: 0.9811
Epoch 4/10
accuracy: 1.0000
Epoch 5/10
accuracy: 1.0000
Epoch 6/10
accuracy: 1.0000
Epoch 7/10
accuracy: 1.0000
Epoch 8/10
accuracy: 1.0000
Epoch 9/10
4/4 [======== ] - 0s 5ms/step - loss: 0.6259 -
accuracy: 1.0000
Epoch 10/10
accuracy: 1.0000
LSTM Evaluation:
accuracy: 1.0000
Models saved successfully.
```