CapX

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1.) Data collection and importing necessary libraries

```
import tweepy
import re
import pandas as pd
import numpy as np
!pip install praw
import praw
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
import nltk
nltk.download('punkt')
nltk.download('stopwords')
# Reddit API credentials
reddit = praw.Reddit(client_id='CF5EOmlCOrpL7qbc3GhxmA',
           client_secret='zRq-FDzbq0xQ4gXGzZ_EmpdppbqxOg',
           user_agent='StockSentimentAnalysis')
subreddit = reddit.subreddit('wallstreetbets')
posts = subreddit.search('GME', limit=1000)
```

Store post data in DataFrame

posts_data = [[post.title, post.selftext, post.score, post.created_utc] for post in
posts]

df_posts = pd.DataFrame(posts_data, columns=['Title', 'Body', 'Score',
'Created'])

df_posts.to_csv('/content/stocks_posts.csv', index=False)

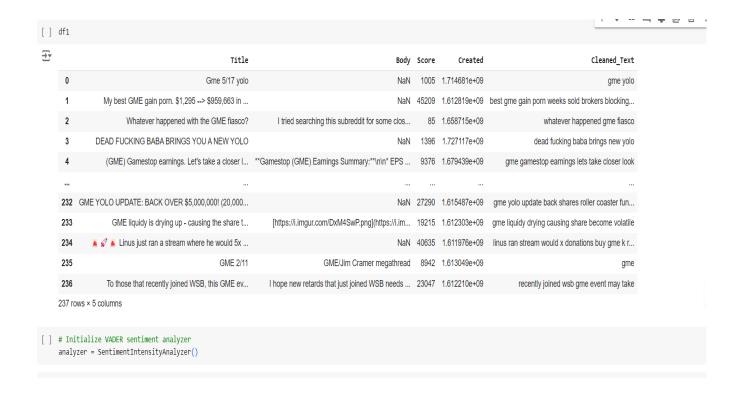
df = pd.read_csv('/content/stocks_posts.csv')

	Title	Body	Score	Created
0	Gme 5/17 yolo	NaN	1005	1.714681e+09
1	My best GME gain porn. \$1,295> \$959,663 in	NaN	45209	1.612819e+09
2	Whatever happened with the GME fiasco?	I tried searching this subreddit for some clos	85	1.658715e+09
3	DEAD FUCKING BABA BRINGS YOU A NEW YOLO	NaN	1396	1.727117e+09
4	(GME) Gamestop earnings. Let's take a closer I	**Gamestop (GME) Earnings Summary:**\n\n* EPS	9376	1.679439e+09
232	GME YOLO UPDATE: BACK OVER \$5,000,000! (20,000	NaN	27290	1.615487e+09
233	GME liquidy is drying up - causing the share t	[https://i.imgur.com/DxM4SwP.png](https://i.im	19215	1.612303e+09
234	🌋 🚀 🌋 Linus just ran a stream where he would 5x	NaN	40635	1.611976e+09
235	GME 2/11	GME/Jim Cramer megathread	8942	1.613049e+09
236	To those that recently joined WSB, this GME ev	I hope new retards that just joined WSB needs	23047	1.612210e+09

237 rows × 4 columns

Function to clean text
def clean_text(text):

```
text = re.sub(r'http\S+', '', text) # Remove URLs
  text = re.sub(r'@\w+', '', text) # Remove mentions
  text = re.sub(r'#', ", text) # Remove hashtags
  text = re.sub(r'[^A-Za-z\s]', '', text) # Remove special characters
  text = text.lower() # Convert to lowercase
  text = ' '.join([word for word in word tokenize(text) if word not in
stopwords.words('english')])
  return text
# Apply cleaning function to the dataset
df['Cleaned Text'] = df['Title'].apply(clean text)
df.to_csv('/content/cleaned_stock_tweets.csv', index=False)
!pip install vaderSentiment
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
df1=pd.read csv('/content/cleaned stock tweets.csv')
```



Function to get sentiment scores

```
def get_sentiment(text):
    score = analyzer.polarity_scores(text)
    return score['compound']
```

Apply sentiment analysis

df['Sentiment_Score'] = df['Cleaned_Text'].apply(get_sentiment)

Label sentiment as Positive, Negative, or Neutral

 $df['Sentiment_Label'] = df['Sentiment_Score'].apply(lambda x: 'Positive' if x > 0 else ('Negative' if x < 0 else 'Neutral'))$

Save results

df.to_csv('sentiment_stock_tweets.csv', index=False)

from sklearn.feature_extraction.text import CountVectorizer from sklearn.decomposition import LatentDirichletAllocation

Count Vectorizer to find mentions of specific stocks

```
vectorizer = CountVectorizer(stop_words='english', ngram_range=(1, 1))
X = vectorizer.fit transform(df['Cleaned Text'])
```

Identify words related to stocks

```
stock_words = ['AAPL', 'GOOGL', 'TSLA', 'GME'] # Add stock symbols here
mention_matrix = X[:, [vectorizer.vocabulary_.get(w.lower()) for w in
stock words if w.lower() in vectorizer.vocabulary ]]
```

Get top words per topic

```
def print_top_words(model, feature_names, n_top_words):
    for topic_idx, topic in enumerate(model.components_):
        print(f"Topic #{topic_idx}: ", " ".join([feature_names[i] for i in topic.argsort()[:-n_top_words - 1:-1]]))
```

print_top_words(lda, vectorizer.get_feature_names_out(), 10)

import matplotlib.pyplot as plt from scipy.stats import pearsonr

Load sentiment data and stock price data

```
sentiment_df = pd.read_csv('/content/sentiment_stock_tweets.csv')
```

stock_price_df = pd.read_csv('/content/stocks_posts.csv') # Assuming you
have stock price data

Merge data on time or another relevant feature

merged_df = pd.merge(sentiment_df, stock_price_df, on='Created')

print(stock_price_df.columns)

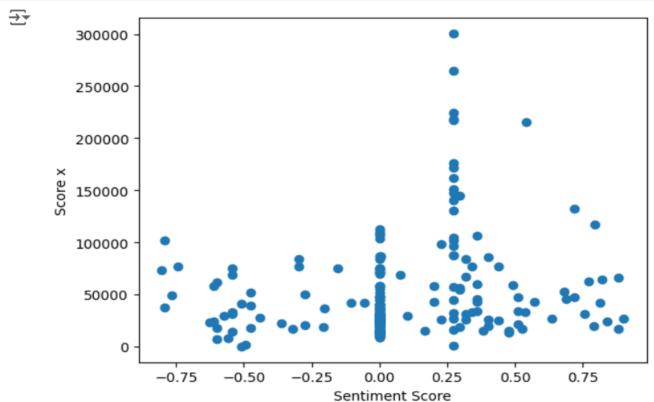
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corr, _ = pearsonr(merged_df['Sentiment_Score'], merged_df['Score_x'])
print(f'Correlation between sentiment score and score x: {corr}')

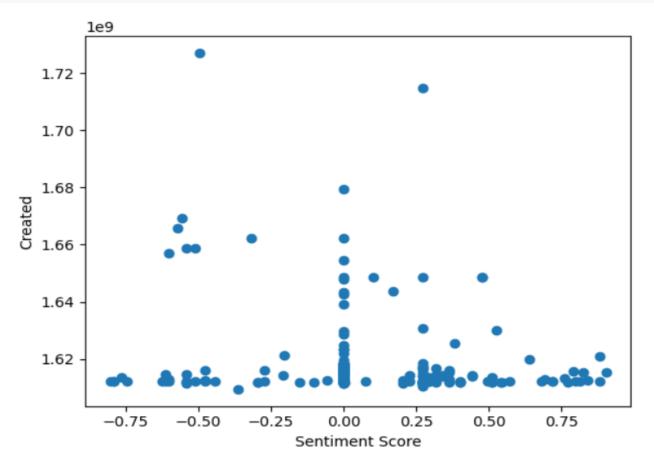
corr1, _ = pearsonr(merged_df['Sentiment_Score'], merged_df['Created'])

print(f'Correlation between sentiment score and Created time: {corr1}')

```
[ ] # Plot sentiment vs stock price change
    plt.scatter(merged_df['Sentiment_Score'], merged_df['Score_x'])
    plt.xlabel('Sentiment Score')
    plt.ylabel('Score x')
    plt.show()
```



```
# Plot sentiment vs created
plt.scatter(merged_df['Sentiment_Score'], merged_df['Created'])
plt.xlabel('Sentiment Score')
plt.ylabel('Created')
plt.show()
```



from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression from sklearn.metrics import mean_squared_error

```
# Prepare features (Sentiment Score) and target (Stock Price Change)
X = merged_df[['Sentiment_Score']]
y = merged_df['Score_x']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)
```

Build Linear Regression model

```
model = LinearRegression()
model.fit(X_train, y_train)
```

Predict and evaluate

```
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print(f'Mean Squared Error: {mse}')
```

output: Mean Squared Error: 2314710322.6136622