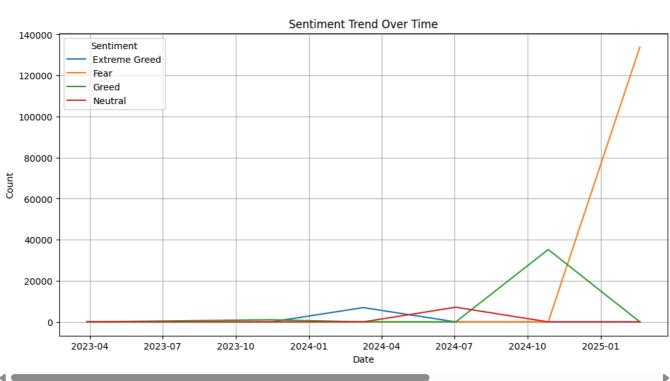
```
# 1. Import Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from datetime import datetime
import plotly.express as px
trades df = pd.read csv("/content/historical data.csv")
sentiment_df = pd.read_csv("/content/fear_greed_index.csv")
print(trades df.columns)
print(sentiment_df.columns)
Index(['Account', 'Coin', 'Execution Price', 'Size Tokens', 'Size USD', 'Side', 'Timestamp IST', 'Start Position', 'Direction', 'Closed PnL', 'Transaction Hash', 'Order ID', 'Crossed', 'Fee', 'Trade ID',
            'Timestamp'],
           dtvpe='object')
     Index(['timestamp', 'value', 'classification', 'date'], dtype='object')
trades_df['Timestamp'] = pd.to_datetime(trades_df['Timestamp'])
sentiment_df['timestamp'] = pd.to_datetime(sentiment_df['timestamp'])
trades_df['date_only'] = trades_df['Timestamp'].dt.date
sentiment_df['date_only'] = sentiment_df['timestamp'].dt.date
sentiment_df = sentiment_df.rename(columns={'classification': 'Sentiment'})
# Merge on date
merged_df = pd.merge(trades_df, sentiment_df[['date_only', 'Sentiment']], on='date_only', how='inner')
print(trades_df.info(memory_usage='deep'))
print(sentiment_df.info(memory_usage='deep'))
<<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 211224 entries, 0 to 211223
     Data columns (total 16 columns):
                         Non-Null Count Dtype
      # Column
     ---
          -----
                             _____
                            211224 non-null object
          Account
          Coin
                            211224 non-null object
          Execution Price 211224 non-null float64
Size Tokens 211224 non-null float64
          Size USD
                            211224 non-null float64
                            211224 non-null object
          Side
          Timestamp IST 211224 non-null object Start Position 211224 non-null float64
                            211224 non-null object
      8
          Direction
                            211224 non-null float64
          Closed Pnl
      10 Transaction Hash 211224 non-null object
      11 Order ID 211224 non-null bool
                             211224 non-null int64
                            211224 non-null
                                              float64
      13 Fee
      14 Trade ID
                            211224 non-null float64
      15 Timestamp
                             211224 non-null float64
     dtypes: bool(1), float64(8), int64(1), object(6)
     memory usage: 111.9 MB
     None
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 2644 entries, 0 to 2643
     Data columns (total 4 columns):
      # Column
                       Non-Null Count Dtype
          timestamp
                          2644 non-null
          value
                           2644 non-null
                                           int64
          classification 2644 non-null
                                           object
                           2644 non-null object
          date
     dtypes: int64(2), object(2)
     memory usage: 380.6 KB
     None
import pandas as pd
# Step 1: Convert timestamps safely
trades_df['Timestamp'] = pd.to_datetime(trades_df['Timestamp'], unit='s', errors='coerce')
sentiment_df['timestamp'] = pd.to_datetime(sentiment_df['timestamp'], unit='s', errors='coerce')
```

```
# Drop rows where conversion failed
trades df.dropna(subset=['Timestamp'], inplace=True)
sentiment_df.dropna(subset=['timestamp'], inplace=True)
# Step 2: Extract only date (not full datetime)
trades_df['date_only'] = trades_df['Timestamp'].dt.date
sentiment_df['date_only'] = sentiment_df['timestamp'].dt.date
# Step 3: Clean and deduplicate sentiment data
sentiment_df = sentiment_df.rename(columns={'classification': 'Sentiment'})
sentiment_df = sentiment_df[['date_only', 'Sentiment']].drop_duplicates()
# Step 4: Merge safely
merged_df = pd.merge(trades_df, sentiment_df, on='date_only', how='inner')
# Step 5: Preview result
print(merged_df.head())
print(merged_df['Sentiment'].value_counts())
→ Empty DataFrame
     Columns: [Account, Coin, Execution Price, Size Tokens, Size USD, Side, Timestamp IST, Start Position, Direction, Closed PnL, Transac
     Index: []
     Series([], Name: count, dtype: int64)
print(trades_df['Timestamp'].min(), trades_df['Timestamp'].max())
print(sentiment df['timestamp'].min(), sentiment df['timestamp'].max())
    1680000000000.0 1750000000000.0
     1517463000 1746163800
#break the process into cleaning steps for both DataFrames:
# 1. Clean trades df
# Convert 'Timestamp' to numeric (handle scientific notation like 1.73E+12)
trades_df['Timestamp'] = pd.to_numeric(trades_df['Timestamp'], errors='coerce')
# Remove invalid timestamps: Keep only Unix timestamps between 2000 and 2030
valid_start = pd.Timestamp("2000-01-01").timestamp() * 1000 # in milliseconds
valid_end = pd.Timestamp("2030-12-31").timestamp() * 1000
trades_df = trades_df[(trades_df['Timestamp'] >= valid_start) &
                      (trades df['Timestamp'] <= valid end)]</pre>
# Convert to datetime from milliseconds
trades\_df['Timestamp'] = pd.to\_datetime(trades\_df['Timestamp'], \ unit='ms', \ errors='coerce')
# Drop rows where conversion failed
trades_df = trades_df.dropna(subset=['Timestamp'])
# Create clean date column
trades_df['date_only'] = trades_df['Timestamp'].dt.date
#Clean sentiment df:
# Convert 'timestamp' to numeric
sentiment_df['timestamp'] = pd.to_numeric(sentiment_df['timestamp'], errors='coerce')
# Remove invalid timestamps
valid_start = pd.Timestamp("2000-01-01").timestamp()
valid end = pd.Timestamp("2030-12-31").timestamp()
sentiment_df = sentiment_df[(sentiment_df['timestamp'] >= valid_start) &
                            (sentiment_df['timestamp'] <= valid_end)]</pre>
# Convert to datetime from seconds
sentiment_df['timestamp'] = pd.to_datetime(sentiment_df['timestamp'], unit='s', errors='coerce')
# Drop rows where conversion failed
sentiment_df = sentiment_df.dropna(subset=['timestamp'])
# Create date column and rename for merge
sentiment_df['date_only'] = sentiment_df['timestamp'].dt.date
sentiment df = sentiment df.rename(columns={'classification': 'Sentiment'})
```

→

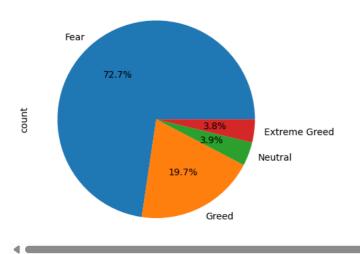
```
# Merge on cleaned 'date_only' column
merged_df = pd.merge(trades_df, sentiment_df[['date_only', 'Sentiment']], on='date_only', how='inner')
# Preview merged data
print("☑ Merged rows:", len(merged_df))
print(merged_df[['Timestamp', 'Sentiment']].head())
→ Merged rows: 184263
                 Timestamp Sentiment
     0 2024-10-27 03:33:20
     1 2024-10-27 03:33:20
                               Greed
     2 2024-10-27 03:33:20
                               Greed
     3 2024-10-27 03:33:20
                               Greed
     4 2024-10-27 03:33:20
                               Greed
# Count of each sentiment
print(merged_df['Sentiment'].value_counts())
→ Sentiment
                      133871
     Fear
     Greed
                       36289
     Neutral
                        7141
     Extreme Greed
                        6962
     Name: count, dtype: int64
# How performance varies by sentiment
avg_pnl_by_sentiment = merged_df.groupby('Sentiment')['Closed PnL'].mean()
print(avg_pnl_by_sentiment)
    Sentiment
                     25.418772
     Extreme Greed
     Fear
                      50.047622
     Greed
                      87.894859
     Neutral
                     22.229713
     Name: Closed PnL, dtype: float64
import matplotlib.pyplot as plt
# Daily sentiment counts
daily_sentiment = merged_df.groupby(['date_only', 'Sentiment']).size().unstack().fillna(0)
daily_sentiment.plot(kind='line', figsize=(12, 6))
plt.title("Sentiment Trend Over Time")
plt.xlabel("Date")
plt.ylabel("Count")
plt.grid()
plt.show()
```



merged_df.to_csv('merged_sentiment_trade_data.csv', index=False)

#Part 2: Exploratory Data Analysis (EDA)

merged_df['Sentiment'].value_counts().plot(kind='pie', autopct='%1.1f%'')



#Trader Performance by Sentiment

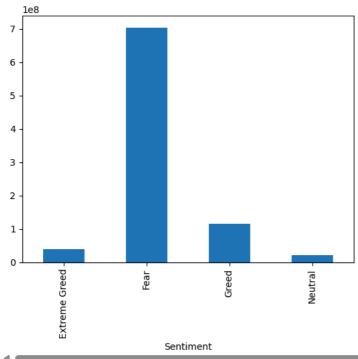
merged_df.groupby('Sentiment')['Closed PnL'].describe()

→		count	mean	std	min	25%	50%	75%	max	
	Sentiment									
	Extreme Greed	6962.0	25.418772	306.166937	-21524.40000	0.0	0.0	16.186514	2500.95750	
	Fear	133871.0	50.047622	909.121796	-35681.74723	0.0	0.0	6.042830	135329.09010	
	Greed	36289.0	87.894859	1148.343968	-117990.10410	0.0	0.0	9.239366	44223.45187	
	Neutral	7141.0	22.229713	633.704815	-18360.67007	0.0	0.0	1.950048	18282.19011	
	1									

#Total Trade Volume by Sentiment

merged_df.groupby('Sentiment')['Size USD'].sum().plot(kind='bar')

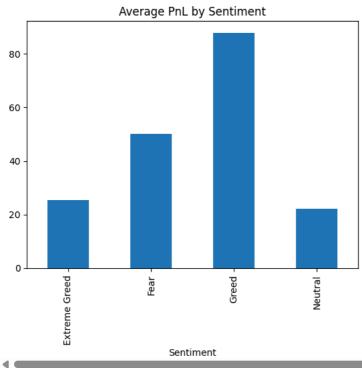




PnL vs Sentiment (Bar Plot)

merged_df.groupby('Sentiment')['Closed PnL'].mean().plot(kind='bar', title="Average PnL by Sentiment")





#Part 3: Visualizations

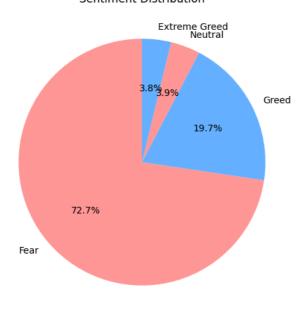
```
# Pie Chart: Sentiment Distribution
```

```
import matplotlib.pyplot as plt
```

```
sentiment_counts = merged_df['Sentiment'].value_counts()
plt.figure(figsize=(6, 6))
sentiment_counts.plot(kind='pie', autopct='%1.1f%%', startangle=90, colors=['#ff9999','#66b3ff'])
plt.title("Sentiment Distribution")
plt.ylabel("")
plt.show()
```

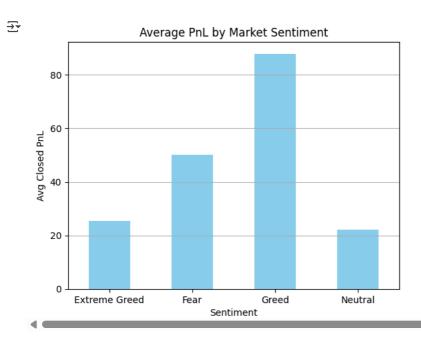


Sentiment Distribution



#Bar Chart: Average PnL by Sentiment

```
avg_pnl_by_sentiment = merged_df.groupby('Sentiment')['Closed PnL'].mean()
avg_pnl_by_sentiment.plot(kind='bar', color='skyblue')
plt.title('Average PnL by Market Sentiment')
plt.ylabel('Avg Closed PnL')
plt.xlabel('Sentiment')
plt.xticks(rotation=0)
plt.grid(axis='y')
plt.show()
```



```
#Time Series Plot: Sentiment Over Time vs Total PnL
```

```
merged_df['date_only'] = pd.to_datetime(merged_df['Timestamp']).dt.date
pnl_over_time = merged_df.groupby(['date_only', 'Sentiment'])['Closed PnL'].sum().unstack().fillna(0)
pnl_over_time.plot(figsize=(12, 5), marker='o')
plt.title('Total PnL Over Time by Sentiment')
plt.ylabel('Total Closed PnL')
plt.xlabel('Total Closed PnL')
plt.xlabel('Date')
plt.grid(True)
plt.legend(title='Sentiment')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

