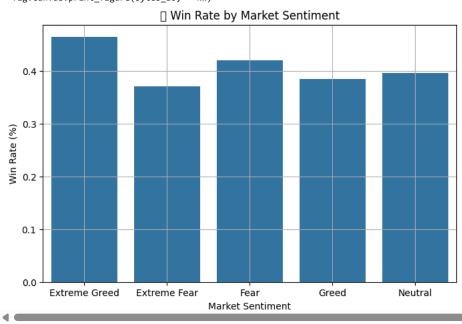
```
from google.colab import drive
drive.mount('/content/drive')
→ Mounted at /content/drive
import os
folder_path = '/content/drive/MyDrive'
files = os.listdir(folder_path)
print(" Files in MyDrive:")
for f in files:
   print(f)
→ Files in MyDrive:
     Classroom
     Yashika Wedding
     Untitled document.gdoc
     Admit Card
     54214244897286_54214254266869.jpg
     Rishali Resume.docx
     assignment_strategy stimulate (1).gdoc
     Cover_letter (1).pdf
     Cover_letter.pdf
     Rishali_2025 .pdf
     Rishali_25.pdf
     updated_file.py
     DS ASSIGNMENT
     Colab Notebooks
     Copy of historical_data.csv
import os
folder_path = '/content/drive/MyDrive/DS_ASSIGNMENT'
files = os.listdir(folder_path)
print(" | Files inside DS_ASSIGNMENT folder:")
for f in files:
   print(f)
    Files inside DS_ASSIGNMENT folder:
     historical_data.csv
     fear_greed_index.csv
import os
folder_path = '/content/drive/MyDrive/DS_ASSIGNMENT'
files = os.listdir(folder_path)
print(" Files inside DS_ASSIGNMENT folder:")
for f in files:
   print(f)
    Files inside DS_ASSIGNMENT folder:
     historical_data.csv
     fear_greed_index.csv
import pandas as pd
# File paths
trader_path = '/content/drive/MyDrive/DS_ASSIGNMENT/historical_data.csv'
sentiment_path = '/content/drive/MyDrive/DS_ASSIGNMENT/fear_greed_index.csv'
# Load CSVs
trader_df = pd.read_csv(trader_path)
sentiment_df = pd.read_csv(sentiment_path)
# Preview data
print("  Trader Data:")
print(trader_df.head())
nnint/"\n = Continent Data."\
```

```
print( \n = Sentiment Data: )
print(sentiment_df.head())
```

```
→ Trader Data:
                                          Account Coin Execution Price
     0 0xae5eacaf9c6b9111fd53034a602c192a04e082ed
                                                   @107
                                                                  7.9769
       0xae5eacaf9c6b9111fd53034a602c192a04e082ed
                                                                  7,9800
                                                   @107
       0xae5eacaf9c6b9111fd53034a602c192a04e082ed
                                                   @107
                                                                  7.9855
       0xae5eacaf9c6b9111fd53034a602c192a04e082ed
                                                                  7,9874
                                                   @107
     4 0xae5eacaf9c6b9111fd53034a602c192a04e082ed @107
                                                                  7.9894
        Size Tokens Size USD Side
                                      Timestamp IST Start Position Direction
     0
             986.87
                     7872.16 BUY
                                   02-12-2024 22:50
                                                           0.000000
                                                                          Buy
     1
             16.00
                      127.68
                              BUY
                                   02-12-2024 22:50
                                                          986.524596
                                                                          Buy
     2
             144.09
                     1150.63 BUY
                                   02-12-2024 22:50
                                                        1002.518996
                                                                          Buy
            142.98
                     1142.04 BUY
                                   02-12-2024 22:50
                                                        1146.558564
     3
                                                                          Buy
     4
              8.73
                        69.75 BUY
                                   02-12-2024 22:50
                                                        1289.488521
                                                                          Buy
        Closed PnL
                                                                         Order ID \
                                                     Transaction Hash
              0.0 0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630
     9
                   0xec09451986a1874e3a980418412fcd0201f500c95bac...
                                                                      52017706630
     1
                                                                      52017706630
                   0xec09451986a1874e3a980418412fcd0201f500c95bac...
     2
              0.0
                   0xec09451986a1874e3a980418412fcd0201f500c95bac...
     3
                                                                      52017706630
              0.0 0xec09451986a1874e3a980418412fcd0201f500c95bac... 52017706630
     4
       Crossed
                              Trade ID
                                           Timestamp
                     Fee
     0
           True 0.345404 8.950000e+14 1.730000e+12
                0.005600
                          4.430000e+14 1.730000e+12
     2
          True 0.050431 6.600000e+14 1.730000e+12
          True 0.050043 1.080000e+15 1.730000e+12
     3
          True 0.003055 1.050000e+15 1.730000e+12
       Sentiment Data:
         timestamp value classification
                                               date
       1517463000
                                         2018-02-01
                      30
                                    Fear
       1517549400
                      15
                           Extreme Fear
                                         2018-02-02
                                         2018-02-03
       1517635800
                      40
                                   Fear
     3
       1517722200
                      24
                           Extreme Fear 2018-02-04
       1517808600
                           Extreme Fear
                                         2018-02-05
# Convert sentiment 'date' column to datetime
sentiment df['date'] = pd.to datetime(sentiment df['date']).dt.date
# Convert trader 'Timestamp IST' to date
trader_df['Timestamp IST'] = pd.to_datetime(trader_df['Timestamp IST'], format='%d-%m-%Y %H:%M')
trader_df['date'] = trader_df['Timestamp IST'].dt.date
# Merge both DataFrames on 'date'
merged df = pd.merge(trader df, sentiment df[['date', 'classification']], on='date', how='left')
# Rename for clarity
merged df.rename(columns={'classification': 'Market Sentiment'}, inplace=True)
# Show merged data
merged_df[['date', 'Side', 'Size USD', 'Closed PnL', 'Market Sentiment']].head()
₹
              date Side Size USD Closed PnL Market Sentiment
      0 2024-12-02 BUY
                           7872.16
                                           0.0
                                                  Extreme Greed
      1 2024-12-02 BUY
                            127.68
                                           0.0
                                                  Extreme Greed
      2 2024-12-02 BUY
                           1150.63
                                           0.0
                                                  Extreme Greed
      3 2024-12-02 BUY
                           1142.04
                                           0.0
                                                  Extreme Greed
       2024-12-02 BUY
                             69 75
                                           0.0
                                                  Extreme Greed
merged_df.columns = merged_df.columns.str.strip().str.lower().str.replace(' ', '_')
# Check missing sentiment data
missing_sentiment = merged_df['market_sentiment'].isnull().sum()
print(f"Rows with missing sentiment: {missing_sentiment}")
```

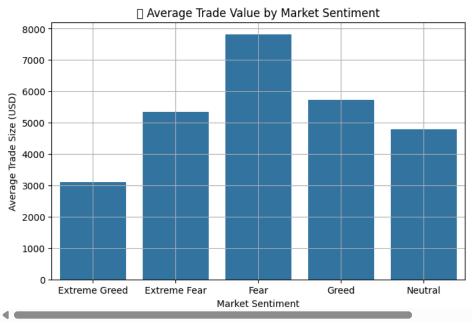
```
→ Rows with missing sentiment: 6
# Drop rows with missing sentiment
merged_df = merged_df.dropna(subset=['market_sentiment'])
# Confirm it's clean now
print("  Final rows:", len(merged_df))
    ☑ Final rows: 211218
# Create profit flag: 1 if profit, 0 if loss
merged\_df['is\_profit'] = merged\_df['closed\_pnl'].apply(lambda x: 1 if x > 0 else 0)
# Rename for consistency
merged_df.rename(columns={'size_usd': 'trade_value_usd'}, inplace=True)
# Quick summary by sentiment
summary = merged_df.groupby('market_sentiment')[['is_profit', 'trade_value_usd']].mean()
print(" | Summary:")
print(summary)
→ II Summary:
                       is_profit trade_value_usd
     market_sentiment
                        0.370607
                                      5349.731843
     Extreme Fear
     Extreme Greed
                        0.464943
                                      3112.251565
                        0.420768
                                      7816.109931
     Fear
                        0.384828
                                      5736.884375
     Greed
                        0.396991
                                      4782.732661
     Neutral
import seaborn as sns
import matplotlib.pyplot as plt
# Plot: Win rate (is_profit) across sentiment
plt.figure(figsize=(8, 5))
sns.barplot(data=merged_df, x='market_sentiment', y='is_profit', estimator='mean', errorbar=None)
plt.title(' Win Rate by Market Sentiment')
plt.ylabel('Win Rate (%)')
plt.xlabel('Market Sentiment')
plt.grid(True)
```

/usr/local/lib/python3.11/dist-packages/IPython/core/pylabtools.py:151: UserWarning: Glyph 128200 (\N{CHART WITH UPWARDS TREND}) missing fig.canvas.print_figure(bytes_io, **kw)



plt.show()

/usr/local/lib/python3.11/dist-packages/IPython/core/pylabtools.py:151: UserWarning: Glyph 128176 (\N{MONEY BAG}) missing from font(s) D fig.canvas.print_figure(bytes_io, **kw)



/usr/local/lib/python3.11/dist-packages/IPython/core/pylabtools.py:151: UserWarning: Glyph 128315 (\N{DOWN-POINTING RED TRIANGLE}) missi fig.canvas.print_figure(bytes_io, **kw)

