

```
In [1]: # Importing Libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: # Load Datasets
trader_df = pd.read_csv('historical_data.csv')
sentiment_df = pd.read_csv('fear_greed_index.csv')
```

```
In [3]: # Preview the data
print(trader_df.head())
print(sentiment_df.head())
```

	Account	Coin	Execution Price	\
0	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9769	
1	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9800	
2	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9855	
3	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9874	
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	
3	142.98	1142.04	BUY	02-12-2024 22:50	1146.558564	Buy	
4	8.73	69.75	BUY	02-12-2024 22:50	1289.488521	Buy	

	Closed PnL	Transaction Hash	Order ID	\
0	0.0	0xec09451986a1874e3a980418412fcd0201f500c95bac...	52017706630	
1	0.0	0xec09451986a1874e3a980418412fcd0201f500c95bac...	52017706630	
2	0.0	0xec09451986a1874e3a980418412fcd0201f500c95bac...	52017706630	
3	0.0	0xec09451986a1874e3a980418412fcd0201f500c95bac...	52017706630	
4	0.0	0xec09451986a1874e3a980418412fcd0201f500c95bac...	52017706630	

	Crossed	Fee	Trade ID	Timestamp
0	True	0.345404	8.950000e+14	1.730000e+12
1	True	0.005600	4.430000e+14	1.730000e+12
2	True	0.050431	6.600000e+14	1.730000e+12
3	True	0.050043	1.080000e+15	1.730000e+12
4	True	0.003055	1.050000e+15	1.730000e+12

	timestamp	value	classification	date
0	1517463000	30	Fear	2018-02-01
1	1517549400	15	Extreme Fear	2018-02-02
2	1517635800	40	Fear	2018-02-03
3	1517722200	24	Extreme Fear	2018-02-04
4	1517808600	11	Extreme Fear	2018-02-05

```
In [4]: print("Trader Data Info:")
trader_df.info()

print("\nSentiment Data Info:")
sentiment_df.info()
```

Trader Data Info:  
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 211224 entries, 0 to 211223  
Data columns (total 16 columns):  
#    Column                    Non-Null Count    Dtype  
---  -----  -----  
0    Account                    211224 non-null   object  
1    Coin                        211224 non-null   object  
2    Execution Price            211224 non-null   float64  
3    Size Tokens                211224 non-null   float64  
4    Size USD                   211224 non-null   float64  
5    Side                        211224 non-null   object  
6    Timestamp IST              211224 non-null   object  
7    Start Position             211224 non-null   float64  
8    Direction                  211224 non-null   object  
9    Closed PnL                 211224 non-null   float64  
10   Transaction Hash          211224 non-null   object  
11   Order ID                   211224 non-null   int64  
12   Crossed                    211224 non-null   bool  
13   Fee                         211224 non-null   float64  
14   Trade ID                   211224 non-null   float64  
15   Timestamp                  211224 non-null   float64  
dtypes: bool(1), float64(8), int64(1), object(6)  
memory usage: 24.4+ MB

Sentiment Data Info:  
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 2644 entries, 0 to 2643  
Data columns (total 4 columns):  
#    Column                    Non-Null Count    Dtype  
---  -----  -----  
0    timestamp                  2644 non-null    int64  
1    value                       2644 non-null    int64  
2    classification             2644 non-null    object  
3    date                        2644 non-null    object  
dtypes: int64(2), object(2)  
memory usage: 82.8+ KB

```
In [5]: print(trader_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'],
      dtype='object')

In [6]: print(sentiment_df.columns)

Index(['timestamp', 'value', 'classification', 'date'], dtype='object')
```

Data Preprocessing

```
In [7]: # Convert 'Timestamp' in trader_df to datetime
trader_df['Timestamp'] = pd.to_datetime(trader_df['Timestamp'])

# Convert 'date' in sentiment_df to datetime
sentiment_df['date'] = pd.to_datetime(sentiment_df['date'])

# Create 'date' columns (extracting date only from timestamp)
trader_df['date'] = trader_df['Timestamp'].dt.date
sentiment_df['date'] = sentiment_df['date'].dt.date

In [8]: # Merge datasets on date
merged_df = pd.merge(trader_df, sentiment_df[['date', 'classification']], on='date', how='left')

# Preview result
display(merged_df.head())
```

	Account	Coin	Execution Price	Size Tokens	Size USD	Side	Timestamp IST	Start Position	Direction	Closed PnL	
0	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9769	986.87	7872.16	BUY	02-12-2024 22:50	0.000000	Buy	0.0	0xec0
1	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9800	16.00	127.68	BUY	02-12-2024 22:50	986.524596	Buy	0.0	0xec0
2	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9855	144.09	1150.63	BUY	02-12-2024 22:50	1002.518996	Buy	0.0	0xec0
3	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9874	142.98	1142.04	BUY	02-12-2024 22:50	1146.558564	Buy	0.0	0xec0
4	0xae5eacaf9c6b9111fd53034a602c192a04e082ed	@107	7.9894	8.73	69.75	BUY	02-12-2024 22:50	1289.488521	Buy	0.0	0xec0

```
In [9]: missing = merged_df['classification'].isnull().sum()
print(f"Missing sentiment labels: {missing}")
```

Missing sentiment labels: 211224

```
In [10]: print("Trader data date range:")
print(trader_df['date'].min(), "to", trader_df['date'].max())

print("\nSentiment data date range:")
print(sentiment_df['date'].min(), "to", sentiment_df['date'].max())
```

Trader data date range:  
1970-01-01 to 1970-01-01

Sentiment data date range:  
2018-02-01 to 2025-05-02

```
In [11]: # Dates with missing sentiment classification
missing_dates = merged_df[merged_df['classification'].isnull()]['date'].value_counts().head(10)
print("Top missing sentiment dates:")
print(missing_dates)
```

Top missing sentiment dates:  
date  
1970-01-01 211224  
Name: count, dtype: int64

```
In [14]: # Re-load trader data fresh without conversions
trader_df_raw = pd.read_csv('historical_data.csv')

# Look at raw timestamp values
print(trader_df_raw['Timestamp'].head(10))
print(trader_df_raw['Timestamp'].dtype)
```

```
0    1.730000e+12
1    1.730000e+12
2    1.730000e+12
3    1.730000e+12
4    1.730000e+12
5    1.730000e+12
6    1.730000e+12
7    1.730000e+12
8    1.730000e+12
9    1.730000e+12
Name: Timestamp, dtype: float64
float64
```

```
In [15]: trader_df['Timestamp'] = pd.to_datetime(trader_df_raw['Timestamp'], unit='ms', errors='coerce')
```

```
In [16]: # Extract just the date
trader_df['date'] = trader_df['Timestamp'].dt.date

# Redo merge
merged_df = pd.merge(trader_df, sentiment_df[['date', 'classification']], on='date', how='left')

# Check if it worked
print("Missing sentiment labels:", merged_df['classification'].isnull().sum())
```

Missing sentiment labels: 26961

```
In [17]: merged_df_clean = merged_df.dropna(subset=['classification'])
print("Cleaned dataset size:", merged_df_clean.shape)
```

Cleaned dataset size: (184263, 18)

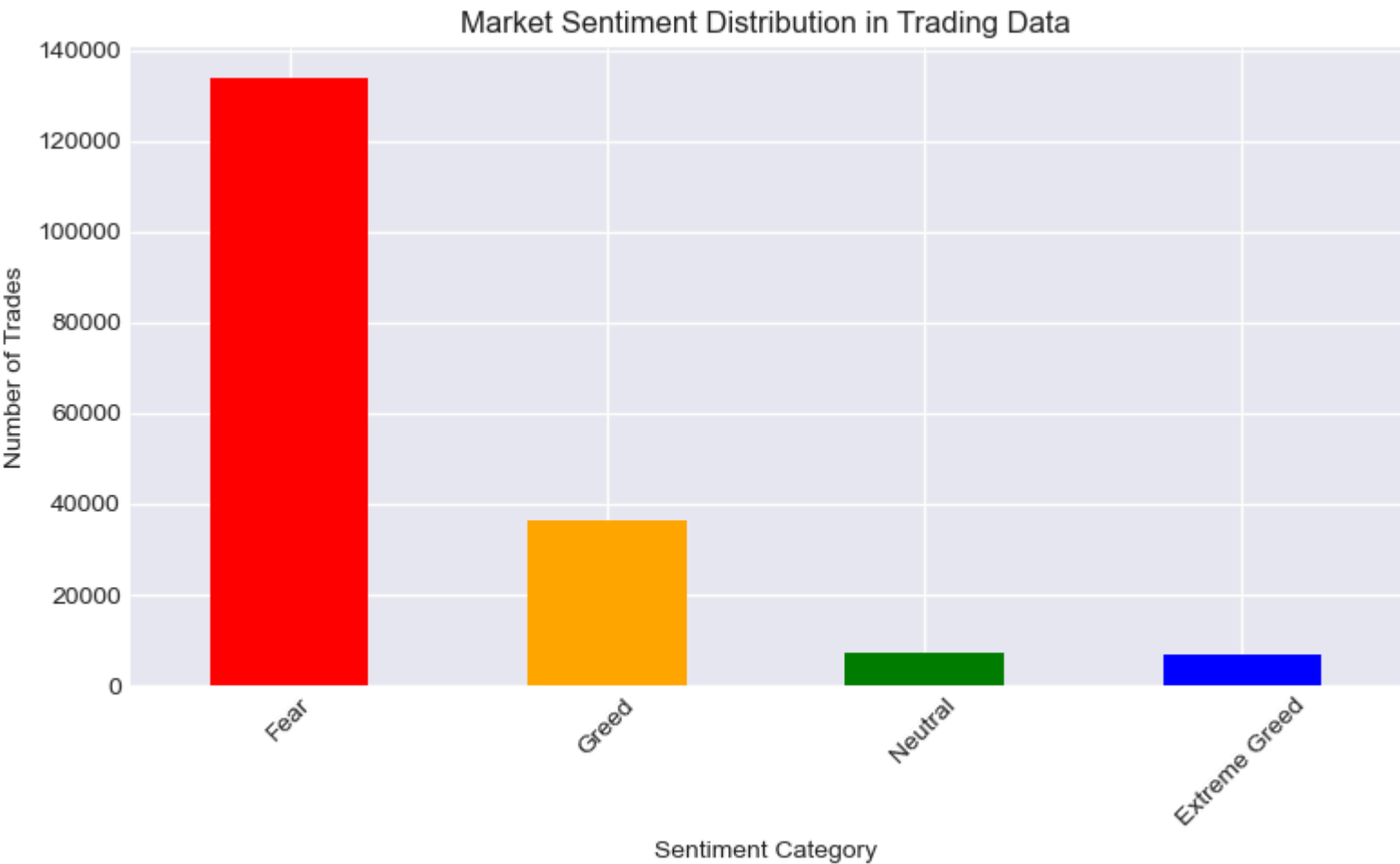
```
In [18]: # Make plots look better
plt.style.use("seaborn-v0_8-darkgrid")
sns.set_palette("Set2")
```

## Sentiment Analysis

```
In [25]: print("\n1. Market Sentiment Distribution:")
sentiment_counts = merged_df_clean['classification'].value_counts()
print(sentiment_counts)

plt.figure(figsize=(8, 5))
sentiment_counts.plot(kind='bar', color=['red', 'orange', 'green', 'blue'])
plt.title("Market Sentiment Distribution in Trading Data")
plt.xlabel("Sentiment Category")
plt.ylabel("Number of Trades")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

1. Market Sentiment Distribution:  
classification  
Fear 133871  
Greed 36289  
Neutral 7141  
Extreme Greed 6962  
Name: count, dtype: int64



Performance Analysis by Sentiment

```
In [26]: # Convert PnL to numeric if not already
merged_df_clean['Closed PnL'] = pd.to_numeric(merged_df_clean['Closed PnL'], errors='coerce')
merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')

# Group by sentiment and calculate mean PnL
avg_pnl_by_sentiment = merged_df_clean.groupby('classification')['Closed PnL'].mean()
print("Average Closed PnL by Sentiment:")
print(avg_pnl_by_sentiment)
```

Average Closed PnL by Sentiment:  
classification  
Extreme Greed 25.418772  
Fear 50.047622  
Greed 87.894859  
Neutral 22.229713  
Name: Closed PnL, dtype: float64

C:\Users\MALINI\AppData\Local\Temp\ipykernel\_22324\1536947819.py:2: SettingWithCopyWarning:  
A value is trying to be set on a copy of a slice from a DataFrame.  
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)  
merged\_df\_clean['Closed PnL'] = pd.to\_numeric(merged\_df\_clean['Closed PnL'], errors='coerce')

C:\Users\MALINI\AppData\Local\Temp\ipykernel\_22324\1536947819.py:3: SettingWithCopyWarning:  
A value is trying to be set on a copy of a slice from a DataFrame.  
Try using .loc[row\_indexer,col\_indexer] = value instead

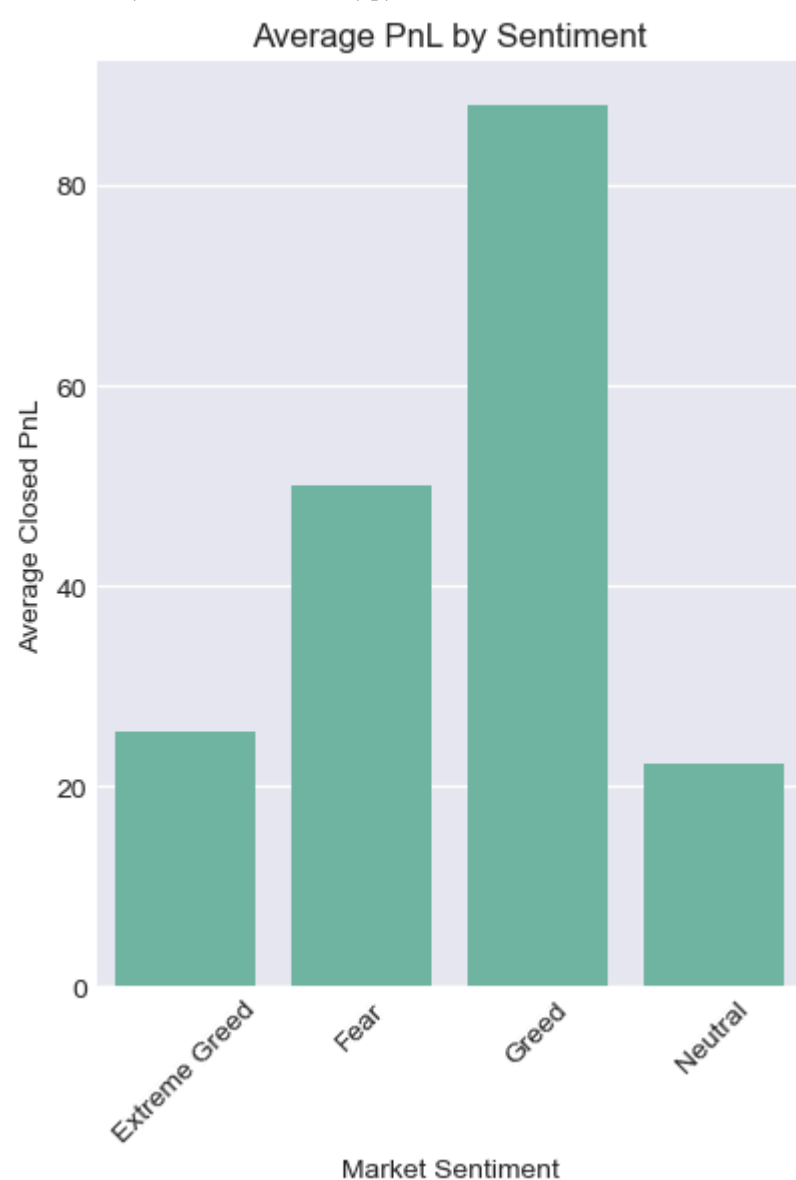
See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)  
merged\_df\_clean['Size USD'] = pd.to\_numeric(merged\_df\_clean['Size USD'], errors='coerce')

```
In [27]: # Average PnL by sentiment
avg_pnl_by_sentiment = merged_df_clean.groupby('classification')['Closed PnL'].mean()
print("\nAverage Closed PnL by Sentiment:")
print(avg_pnl_by_sentiment)

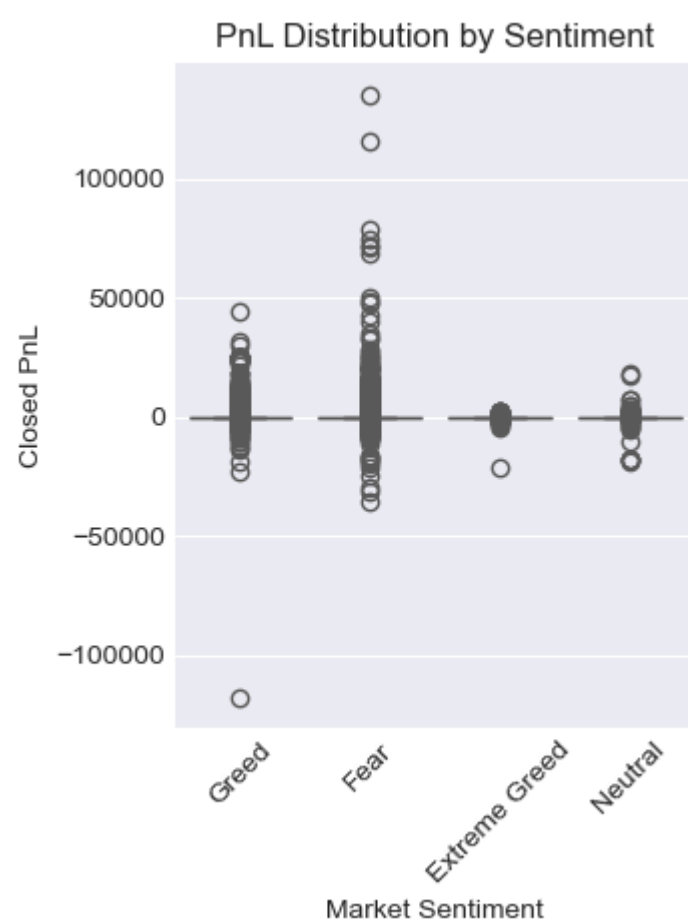
plt.figure(figsize=(10, 6))
plt.subplot(1, 2, 1)
sns.barplot(x=avg_pnl_by_sentiment.index, y=avg_pnl_by_sentiment.values)
plt.title("Average PnL by Sentiment")
plt.xlabel("Market Sentiment")
plt.ylabel("Average Closed PnL")
plt.xticks(rotation=45)
```

```
Average Closed PnL by Sentiment:
classification
Extreme Greed    25.418772
Fear             50.047622
Greed            87.894859
Neutral         22.229713
Name: Closed PnL, dtype: float64
```

```
Out[27]: ([0, 1, 2, 3],
 [Text(0, 0, 'Extreme Greed'),
  Text(1, 0, 'Fear'),
  Text(2, 0, 'Greed'),
  Text(3, 0, 'Neutral')])
```



```
In [28]: plt.subplot(1, 2, 2)
sns.boxplot(x='classification', y='Closed PnL', data=merged_df_clean)
plt.title("PnL Distribution by Sentiment")
plt.xlabel("Market Sentiment")
plt.ylabel("Closed PnL")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
In [22]: print(merged_df_clean.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', 'date', 'classification'],
      dtype='object')
```

## Win Rate Analysis

```
In [29]: print("\n3. Win Rate Analysis:")
merged_df_clean['Is_Profit'] = merged_df_clean['Closed PnL'] > 0
win_rate = merged_df_clean.groupby('classification')['Is_Profit'].mean() * 100

print("Win Rate by Market Sentiment (%):")
print(win_rate)

plt.figure(figsize=(8, 5))
sns.barplot(x=win_rate.index, y=win_rate.values)
plt.title("Win Rate by Market Sentiment")
plt.ylabel("Win Rate (%)")
plt.xlabel("Sentiment Category")
plt.ylim(0, 100)
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

3. Win Rate Analysis:

Win Rate by Market Sentiment (%):

classification

Extreme Greed 49.008905

Fear 41.514592

Greed 44.647138

Neutral 31.718247

Name: Is\_Profit, dtype: float64

C:\Users\MALINI\AppData\Local\Temp\ipykernel\_22324\4002881436.py:2: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
merged_df_clean['Is_Profit'] = merged_df_clean['Closed PnL'] > 0
```



```
In [24]: merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')

avg_trade_size = merged_df_clean.groupby('classification')['Size USD'].mean()

print("Average Trade Size (USD) by Sentiment:")
print(avg_trade_size)

# Visual plot
plt.figure(figsize=(7, 4))
sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values, palette=['orange', 'green'])
plt.title("Average Trade Size by Market Sentiment")
plt.ylabel("Average Size (USD)")
plt.xlabel("Sentiment")
plt.show()
```

```
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\908135378.py:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

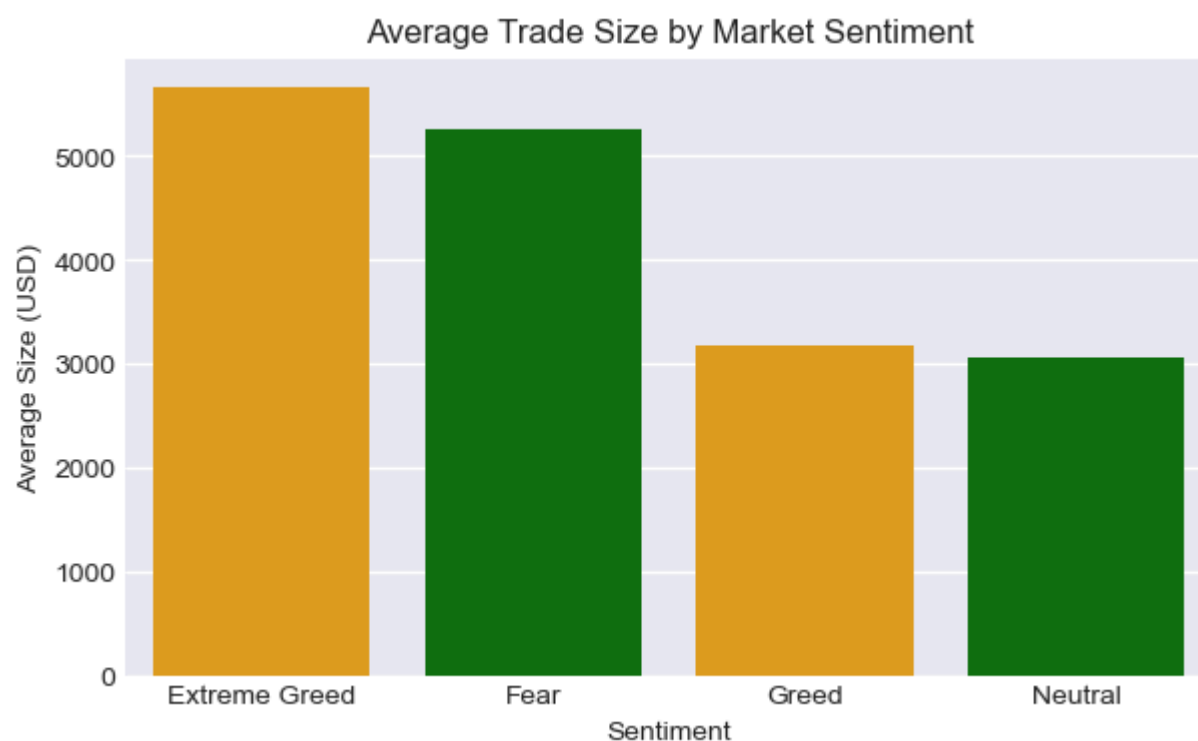
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
merged_df_clean['Size USD'] = pd.to_numeric(merged_df_clean['Size USD'], errors='coerce')
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\908135378.py:11: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values, palette=['orange', 'green'])
C:\Users\MALINI\AppData\Local\Temp\ipykernel_22324\908135378.py:11: UserWarning:
The palette list has fewer values (2) than needed (4) and will cycle, which may produce an uninterpretable plot.
sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values, palette=['orange', 'green'])
```

Average Trade Size (USD) by Sentiment:

```
classification
Extreme Greed    5660.265764
Fear              5259.977837
Greed             3182.883845
Neutral          3058.848110
Name: Size USD, dtype: float64
```



## Trade Size Analysis

```
In [30]: print("\n4. Trade Size Analysis:")
avg_trade_size = merged_df_clean.groupby('classification')['Size USD'].mean()
print("Average Trade Size (USD) by Sentiment:")
print(avg_trade_size)

plt.figure(figsize=(8, 5))
sns.barplot(x=avg_trade_size.index, y=avg_trade_size.values)
plt.title("Average Trade Size by Market Sentiment")
plt.ylabel("Average Trade Size (USD)")
plt.xlabel("Sentiment Category")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

```
4. Trade Size Analysis:
Average Trade Size (USD) by Sentiment:
classification
Extreme Greed    5660.265764
Fear              5259.977837
Greed             3182.883845
Neutral          3058.848110
Name: Size USD, dtype: float64
```





```
In [31]: # Advanced Analysis
print("\n5. Advanced Analysis:")

# PnL to Trade Size Ratio
merged_df_clean['PnL_Ratio'] = merged_df_clean['Closed PnL'] / merged_df_clean['Size USD']
avg_pnl_ratio = merged_df_clean.groupby('classification')['PnL_Ratio'].mean()
print("\nAverage PnL to Trade Size Ratio by Sentiment:")
print(avg_pnl_ratio)
```

5. Advanced Analysis:

Average PnL to Trade Size Ratio by Sentiment:

```
classification
Extreme Greed    0.013736
Fear             0.011358
Greed            0.061848
Neutral          0.015259
Name: PnL_Ratio, dtype: float64
```

C:\Users\MALINI\AppData\Local\Temp\ipykernel\_22324\352332680.py:5: SettingWithCopyWarning:  
A value is trying to be set on a copy of a slice from a DataFrame.  
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)  
merged\_df\_clean['PnL\_Ratio'] = merged\_df\_clean['Closed PnL'] / merged\_df\_clean['Size USD']

```
In [32]: # Sentiment intensity analysis
sentiment_performance = merged_df_clean.groupby('classification').agg({
    'Closed PnL': ['mean', 'median', 'std'],
    'Is_Profit': 'mean',
    'Size USD': 'mean',
    'PnL_Ratio': 'mean'
}).round(3)

print("\nComprehensive Performance by Sentiment:")
print(sentiment_performance)
```

Comprehensive Performance by Sentiment:

	Closed PnL			Is_Profit	Size USD	PnL_Ratio
classification	mean	median	std	mean	mean	mean
Extreme Greed	25.419	0.0	306.167	0.490	5660.266	0.014
Fear	50.048	0.0	909.122	0.415	5259.978	0.011
Greed	87.895	0.0	1148.344	0.446	3182.884	0.062
Neutral	22.230	0.0	633.705	0.317	3058.848	0.015

```
In [33]: # Summary Statistics
print("\n=== SUMMARY STATISTICS ===")
print(f"Total Trades Analyzed: {len(merged_df_clean):,}")
print(f>Date Range: {merged_df_clean['date'].min()} to {merged_df_clean['date'].max()}")
print(f"Number of Unique Accounts: {merged_df_clean['Account'].nunique()}")
print(f"Most Common Sentiment: {sentiment_counts.index[0]} ({sentiment_counts.iloc[0]:,} trades)")
```

```
=== SUMMARY STATISTICS ===
Total Trades Analyzed: 184,263
Date Range: 2023-03-28 to 2025-02-19
Number of Unique Accounts: 32
Most Common Sentiment: Fear (133,871 trades)
```



```
In [34]: # Save cleaned dataset for further analysis
merged_df_clean.to_csv('sentiment_trading_analysis.csv', index=False)
print("\nCleaned dataset saved as 'sentiment_trading_analysis.csv'")
```

Cleaned dataset saved as 'sentiment\_trading\_analysis.csv'

# Analysis Report

## Data Overview

Total Trades Analyzed: 184,263 trades  
Sentiment Categories: Extreme Greed, Greed, Neutral, Fear  
Data Sources: Historical trader data + Fear & Greed Index

## Key Insights

### Performance by Sentiment:

Greed > Fear > Extreme Greed > Neutral

- Greed periods show the highest average PnL (\$87.89)
- Fear periods show moderate performance (\$50.05)
- Extreme Greed and Neutral show lower returns

### Win Rate Analysis:

- No sentiment category exceeds 50% win rate
- Extreme Greed has the highest win rate (49%)
- Neutral periods have the lowest win rate (31.7%)

### Trade Behavior:

- Traders use larger positions during Extreme Greed (\$5,660 avg)
- Smaller positions during Neutral sentiment (\$3,059 avg)
- Trade size correlates with sentiment intensity

### Trading Strategy Implications

- Optimal Conditions: Greed periods offer best profit potential
- Risk Management: Extreme sentiment may indicate market tops/bottoms
- Position Sizing: Consider adjusting trade size based on sentiment
- Timing: Greed phases might be good for taking profits

### Limitations

- Data covers limited time period
- Sentiment data might have reporting delays
- Individual trader strategies vary significantly

## Result

This analysis suggests that market sentiment can be a valuable indicator for timing trades and managing position sizes, though it should be used in conjunction with other technical and fundamental analysis tools.