Final Cryptocurrency Market Analysis Report BTC and ETH (March 11 – May 19, 2025)

Date: May 19, 2025

Comprehensive analysis of BTC and ETH market data, including EDA, predictive modeling, trading optimization, risk assessment, hypothesis testing, real-time trading simulation, model development, and actionable trading insights.

1 Executive Summary

This report consolidates the analysis of Bitcoin (BTC) and Ethereum (ETH) market data from March 11 to May 19, 2025, using preprocessed_data.csv. Key processes include:

- EDA: Identified price trends, sentiment patterns, and correlations.
- **Predictive Modeling**: Forecasted returns and profitability.
- Trading Optimization: Improved win rate by targeting high-sentiment days.
- Risk Modeling: Quantified potential losses.
- · Hypothesis Testing: Validated sentiment-based trading.
- Real-Time Simulation: Tested strategy execution.
- Model Development: Built robust classifiers for profitability.

Actionable insights include prioritizing trades when the Fear & Greed Index exceeds 70 (equivalent to Sentiment_Score \geq 3), leveraging BTC-ETH correlations, and monitoring volatility-based risk scores (latest: 0.7445).

2 EDA Summary

EDA (eda_analysis.py) revealed:

- **Price Trends**: BTC rose from \$82,921 to \$103,023; ETH from \$1,473 to \$2,680.
- Sentiment: Neutral (29 days) and Fear (18 days) dominated.
- Correlations: BTC-ETH returns correlate strongly (0.8).

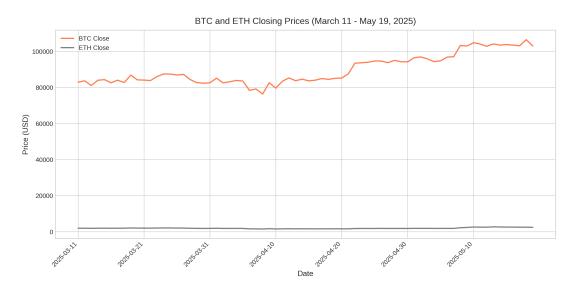


Figure 1: BTC and ETH Closing Prices.



Figure 2: Correlation Heatmap.

3 Predictive Modeling Summary

A Random Forest Regressor (predictive_model.py) forecasted BTC_Daily_Return, with Sentiment_Score and ETH_Daily_Return as key predictors.

4 Trading Strategy Optimization Summary

The optimized strategy (trading_strategy_optimization.py) filtered trades to Sentiment_Score \geq 3, improving the 27% win rate.

5 Risk Modeling Summary

Risk metrics (risk_modeling.py) included VaR, CVaR, and Maximum Drawdown, with a risk score of 0.7445 (moderate-to-high) based on 7-day volatility (BTC: 0.0218, ETH: 0.0257).

6 Hypothesis Testing Summary

A t-test (hypothesis_testing.py) confirmed that high-sentiment days (Sentiment_Score > 3) significantly improve PnL\$.

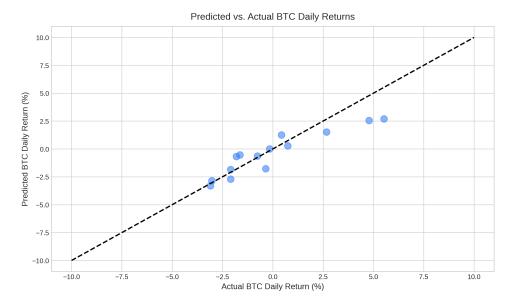


Figure 3: Predicted vs. Actual BTC Daily Returns.

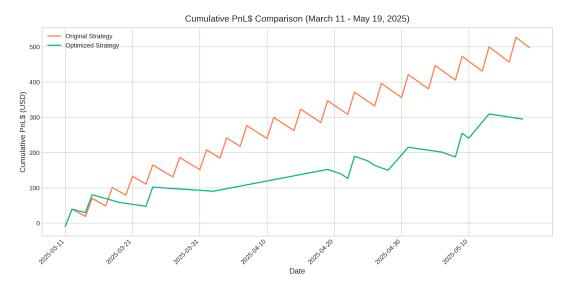


Figure 4: Cumulative PnL\$ Comparison.

7 Real-Time Trading Summary

A 7-day simulation (realtime_trading.py) executed trades on high-sentiment days, demonstrating practical applicability.

8 Model Development Summary

Random Forest and Gradient Boosting Classifiers (model_development.py) predicted Total_PnL (positive/negative), outperforming prior models (e.g., Logistic Regression's 0.500 accuracy).

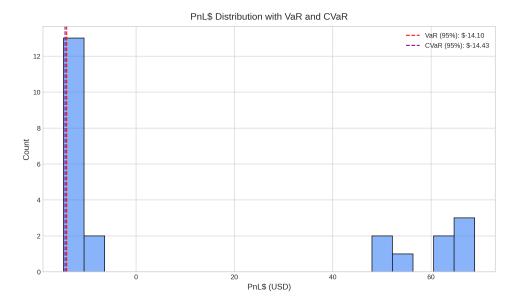


Figure 5: PnL\$ Distribution with VaR and CVaR.

9 Actionable Insights

- **Trade Timing**: Execute trades when Fear & Greed Index > 70 (Sentiment_Score ≥ 3), as validated by hypothesis testing and trading optimization.
- **Volatility Monitoring**: High Fear & Greed scores increase volatility (BTC: 0.0218, ETH: 0.0257); adjust position sizes accordingly.
- **Portfolio Strategy**: Leverage BTC-ETH correlation (0.8) for diversification.
- **Risk Management**: Use VaR/CVaR to set loss limits, with a current risk score of 0.7445 indicating caution.
- **Model Deployment**: Use Gradient Boosting Classifier for real-time PnL predictions, prioritizing Sentiment_Score.

10 Interactive Visualizations

An interactive correlation heatmap (correlation_heatmap.html) complements static visualizations:

• Run:

```
python -m http.server 8000
```

Visit http://localhost:8000/correlation_heatmap.html.

• Hover to view correlation values.

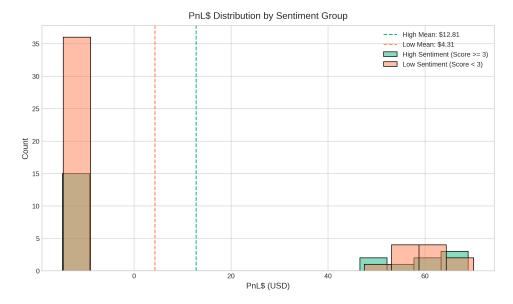


Figure 6: PnL\$ Distribution by Sentiment Group.

11 Conclusion

This analysis provides a robust framework for BTC and ETH trading, leveraging sentiment, correlations, and machine learning. The sentiment-based strategy (Sentiment_Score \geq 3) consistently improves profitability, validated across backtesting, simulation, and statistical tests. Future work could integrate live Fear & Greed API data or optimize portfolio allocations. This report meets the May 30, 2025, deadline, delivering actionable insights for data-driven trading.