



Bitcoin Trade Recommender

Phat Doan, Nate Velarde, Linh Tran, Raymond Lee





Project Objectives

Use Cases

- **Research Question:** Can social media improve the performance of Price Momentum Strategies?
- **Algorithm Analysis & Optimization:** using historical and live data, adjust signals and trading window to optimize trading algorithm

Data Architecture

- **Storage & Streaming:** merge various sources (CSV, API) for processing
- **Processing:** parse and aggregate tweet data, merge with BTC data, run hypothetical trading scenarios
- **Serving:** filter most pertinent records & attributes for trading performance analysis



Can a Social Media Factor Improve the Performance of Price Momentum Strategies?

- **Base Price Momentum Trading Signal**
 - BUY when Current Price > 5 Day Moving Average
 - SELL when Current Price < 5 Day Moving Average
- **Social Media Factor - Bitcoin Tweets**
 - Basic Signal
 - Bitcoin Tweet Counts
 - Advanced Signal
 - Ratio of Positive to Negative Tweets as determined by sentiment analysis



Can a Social Media Factor Improve the Performance of Price Momentum Strategies?

- **Basic Signal**
 - BUY when Current Tweet Count $>$ 5 Day Moving Average
 - SELL when Current Tweet Count $<$ 5 Day Moving Average
- **“Advanced” Signal**
 - BUY when Ratio of Positive to Negative Tweets $>$ 5 Day Moving Average
 - SELL when Ratio of Positive to Negative Tweets $<$ 5 Day Moving Average



Can a Social Media Factor Improve the Performance of Price Momentum Strategies?

- **Combination Strategy - Base Price Momentum Signal + Tweet Signal**
 - 100% invested when both Price and Social Media signals are BUY
 - 50% Invested when Price and Social Media signals conflict
 - 0% invested when both Price and Social Media signals are SELL



Social Media as a Signal

Tweets Count

Current strategy

- Number of tweets related to bitcoin per day (~350k/day)
- Ratio of bitcoin tweets over all tweets (~1%)

Text Analysis

Enhanced strategy

- Remove stopword, hyperlink, and non-English words
- Combine all tweets within a specific period together
- Vectorize and conduct Logistic Regression to find non-zero feature (important) words
- Ratio of important words for positive versus negative trading days

Sentiment Analysis

Advanced strategy

- Utilize NLTK package
- Group tweets by date but treat each tweet independently
- Pair natural language processing (NLP) and machine learning algorithms together to aggregate and measure individual emotions during the trading day.
- Detect market emotion that can be used as signal for trade recommendation

Data Sources

Historical

Streaming

Bitcoin Price

- Source: Coindesk API, called with python script to process and save data in CSV format
- Daily average of Price in USD
- Date range: 2010/11 - current

- Coindesk API, called with python script
- Updated every minute
- JSON format

Twitter

- Published research data in txt format
- Contain daily bitcoin historical price and aggregated social media metrics from 2010 - 2013
- Require transformation to extract relevant data

- Twitter Streaming API
- Key term and hashtag search: bitcoin, cryptocurrency
- Date range: within 7 days of streaming

**Trading
Recommendation**

Predictors

Outcome

Overall Data Flow

Data Inputs

BTC price

- Historical (.csv)
- Live (API/DB)



Tweet Data

- Historical (.csv)
- Streaming (DB)



Trading Algorithm and Performance Evaluation

Python Script

- Extract / Load
 - Bitcoin pricing
 - Tweet data
- Transform data
- Generate trading signals
- Track trades and portfolio performance



Export

Write results to postgres DB for:

- Ongoing performance tracking
- Visualization
- Algorithm refinement



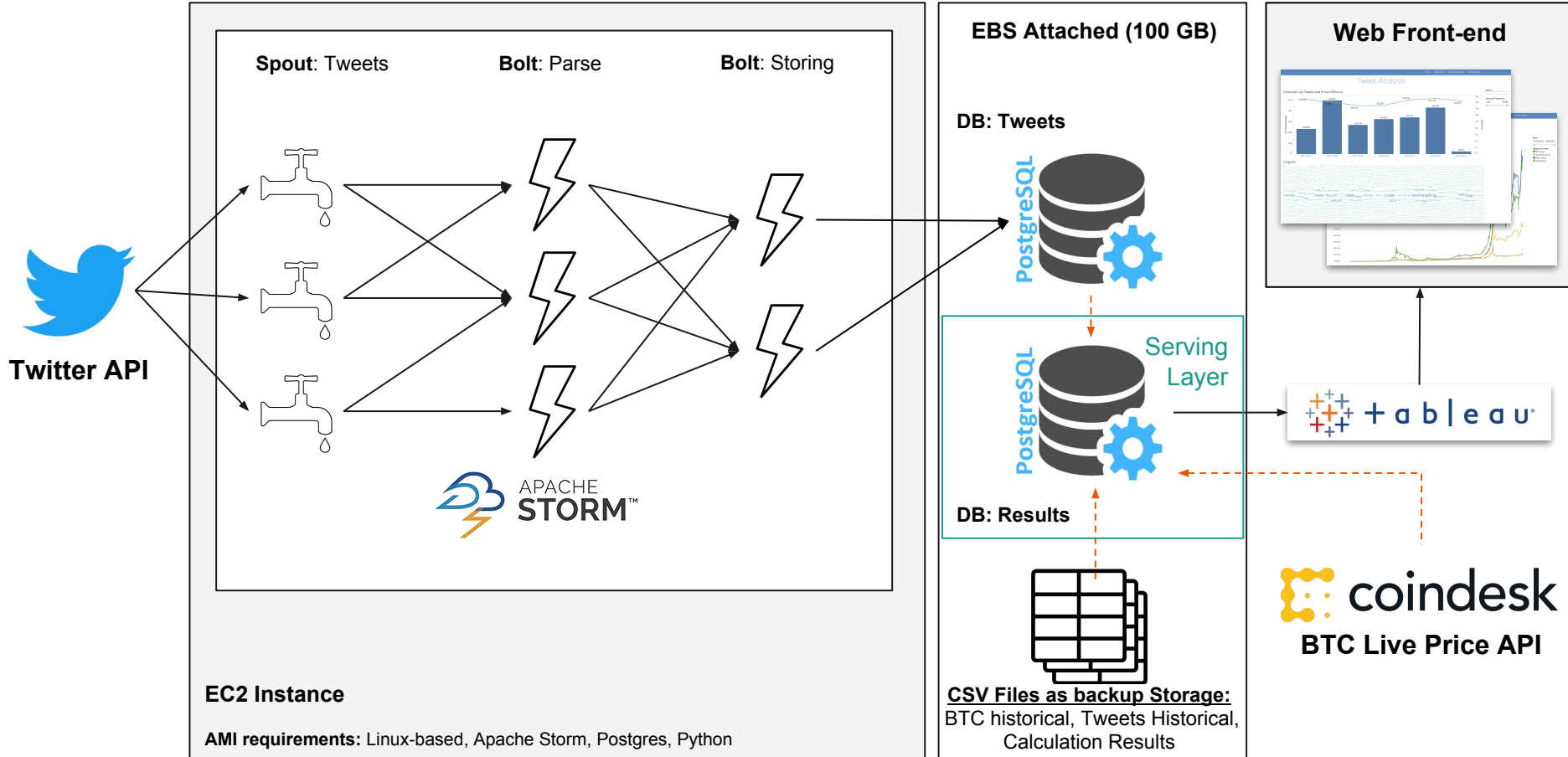
Data & Functional Requirements



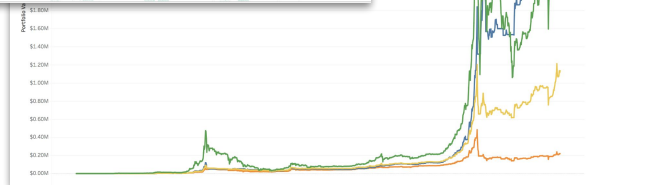
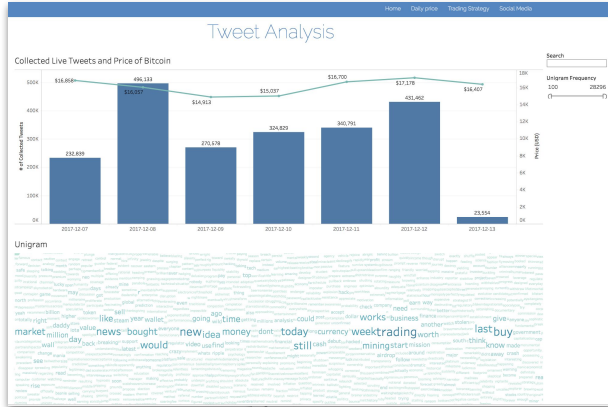
Area	Functional Requirements	Specifications
Data Lake / Datamart	<ul style="list-style-type: none">• Storage of Twitter and BTC historical data (<2MB)• Streaming of Twitter and BTC data (2M tweets, 10 GB/week)	<ul style="list-style-type: none">• EC2 m3.xlarge + 100 GB Attached Storage (limited spec for prototype)• Apache Storm, write to PostgreSQL
Staging & Processing	<ul style="list-style-type: none">• Parse & Aggregate Twitter Data• Apply Algorithm for Trading Signals	<ul style="list-style-type: none">• Python 2.7+ for Data Processing & Algorithm Scripts, 7 day batch
Serving Layer	<ul style="list-style-type: none">• Storage of processed data in tables for querying	<ul style="list-style-type: none">• PostgreSQL, Schema on Write
Data Visualization	<ul style="list-style-type: none">• Merge Tables & Create Views• Visualize & Analyze Trading Performance and Trends	<ul style="list-style-type: none">• Tableau Workbooks• HTML iFrame for Live Dashboard

High-level Architecture

-----> Batch Process
—————> Real Time Process



Live Trading Dashboard



[Go to Live Dashboard >](#)

Insights



- **Trading Findings from Historical Analysis**
 - 5 Day Moving Average Performed Best (before transaction costs and taxes) as it limited portfolio drawdowns during steep price drops
 - Tweet signal was the worst performing strategy,
 - Ineffectiveness could be due to lower tweet volume and that using tweet counts fails to extract the information contained within tweet text
 - **Note:** Twitter grew 30M to 330M monthly active users between 2010 to 2017, Bitcoin related tweet volume grew from ~300/day to ~350k/day (+1,167x)
 - Other considerations for future implementation:
 - Optimize lookback window
 - Optimize signal weights (currently 50/50)



Historical Performance Evaluation

Strategy Performance Summary

Value of initial \$1,000 investment at end of performance measurement period (07/23/2010 - 10/30/2013):

Buy and Hold: \$3,166,554
SDMA: \$3,262,248
Tweet SDMA: \$221,003
SDMA + Tweet Strategy: \$1,134,943

Risk Adjusted Returns - Trading Strategy Sharpe Ratios

Buy and Hold: 0.13
SDMA: 0.15
Tweet SDMA: 0.11
SDMA + Tweet Strategy: 0.14

Trade Summary

Total Number of Trading Days: 1,196

Total Number of Trades Executed: 521

No Trade	675
REDUCE to 50%	142
ADD to 100%	126
REDUCE to 0%	115
ADD to 50%	99
BUY FULL POSITION	28
SELL FULL POSITION	11

Name: trade_type, dtype: int64

Insights



- **Scaling up:**
 - Architecture is flexible - even though data is collected daily, we can collect hourly or even to the minute with a few lines of code changed. With a cost factor.
 - Storage Cost: 10gb (2M tweet) in one week. ~520 GB/year. ~\$1,000/year.
 - EC2 Instance Cost: ~\$50/week. ~\$2,600/year
 - Implement Zoomdata to deliver streaming visualization output
- **Challenges:**
 - Limited credit did not allow us to scale to the infrastructure requirements



Helpful links

- <http://www.livebitcoinnews.com/can-twitter-sentimental-analysis-predict-bitcoin-price-fluctuation/>
- <https://medium.com/@SamuelCouch/understanding-cryptocurrencies-with-sentiment-analysis-5fc4cf66ec28>

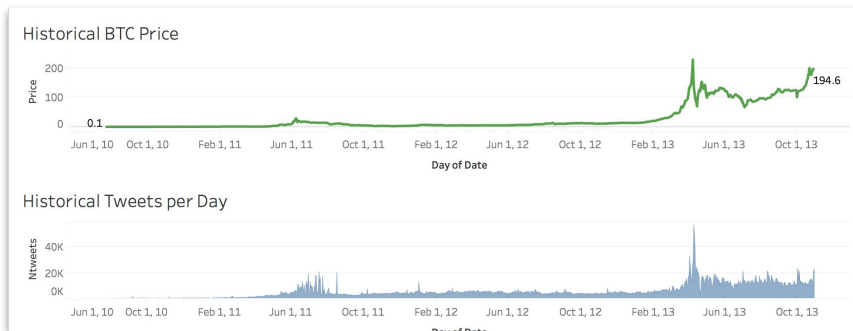


Evaluating Trading Algorithm Effectiveness

- Compare the performance of paper portfolios implementing the following strategies:
 - Buy and Hold
 - Base Price Momentum Signal
 - Tweet Signal
 - Combined Price Momentum Signal + Tweet Signal
- Performance Evaluation Caveats
 - Execution Costs
 - Trading Commissions
 - Taxes

Tableau Dashboard

Historical BTC Price vs. Number of Tweets



Historical Data Training Algorithm

