

# Quantitative Research Internship Assessment Report

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**Position:** Quantitative Research Intern

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## Overview

This report presents a comprehensive analysis of temporary market impact for three listed tickers — **CRVV**, **FROG**, and **SOUN** — using trade-level data. The primary goals are:

- To **model short-term market impact** from aggressive trade activity.
- To **formulate a mathematical execution strategy** that minimizes transaction costs.

The project leverages robust data processing, statistical modeling, and optimization techniques that reflect real-world trading microstructure and align with industry practices in execution algorithms.

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## 1. Dataset Description

Each dataset contains trade-by-trade records with the following fields:

- **timestamp**: UTC time of trade execution
- **price**: Trade execution price
- **size**: Volume of shares traded
- **side**: Indicates trade direction ( **BUY** or **SELL** )

All datasets were sampled at **millisecond-level granularity** and span a full trading day. For analysis, data was resampled into **1-minute buckets**.

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## 2. Data Processing Pipeline

Steps followed per ticker:

1. **Parsed** CSV files using **pandas**; converted **timestamp** to datetime.
2. **Computed signed volume** using:

```
df['signed_volume'] = df['size'] * df['side'].map({'BUY': 1, 'SELL': -1})
```

3. **Resampled data** at 1-minute intervals to compute:
4. **mean\_price**: Volume-weighted average price (VWAP)
5. **total\_signed\_volume**
6. **log\_return**: Log of price ratio between consecutive minutes

All NaN values were forward-filled or dropped based on context.

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### 3. Market Impact Modeling

#### Hypothesis

Temporary market impact is linearly related to the signed trading volume within short time intervals.

#### Regression Model

We use an **ordinary least squares (OLS)** regression:

$$\Delta P_t = \alpha + \beta \cdot V_t^{\text{signed}} + \epsilon_t$$

Where:

- $\Delta P_t$  : Log price change (return)
- $V_t^{\text{signed}}$  : Signed volume
- $\beta$  : Market impact coefficient

#### Model Performance

Ticker	$\beta$ Coefficient	R-squared	Interpretation
CRWV	+0.0014	0.36	Moderate sensitivity to signed volume
FROG	+0.0023	0.42	High price sensitivity (strong impact)
SOUN	+0.0019	0.38	Balanced price-volume relationship

#### Plots

- **Scatter plots** show minute-wise price change vs. signed volume.
- **Fitted regression lines** validate the linear relationship visually.
- Plots available in the `/plots` directory and embedded in the README.

### 4. Optimal Execution Strategy

#### Objective

Design an execution schedule that **minimizes transaction cost** while completing a large order (e.g., 100,000 shares) over a time horizon.

#### Cost Function

The cost model incorporates both **market impact** and **risk aversion**:

$$C = \sum_{t=1}^T \left[ \beta \cdot v_t^2 + \lambda (v_t - v_{t-1})^2 \right]$$

Where:

- $v_t$  : Volume traded in interval  $t$
- $\beta$  : Temporary impact coefficient from regression
- $\lambda$  : Risk aversion coefficient
- $T$  : Number of execution intervals

## Optimization Technique

Used `cvxpy` to solve the **quadratic optimization** problem:

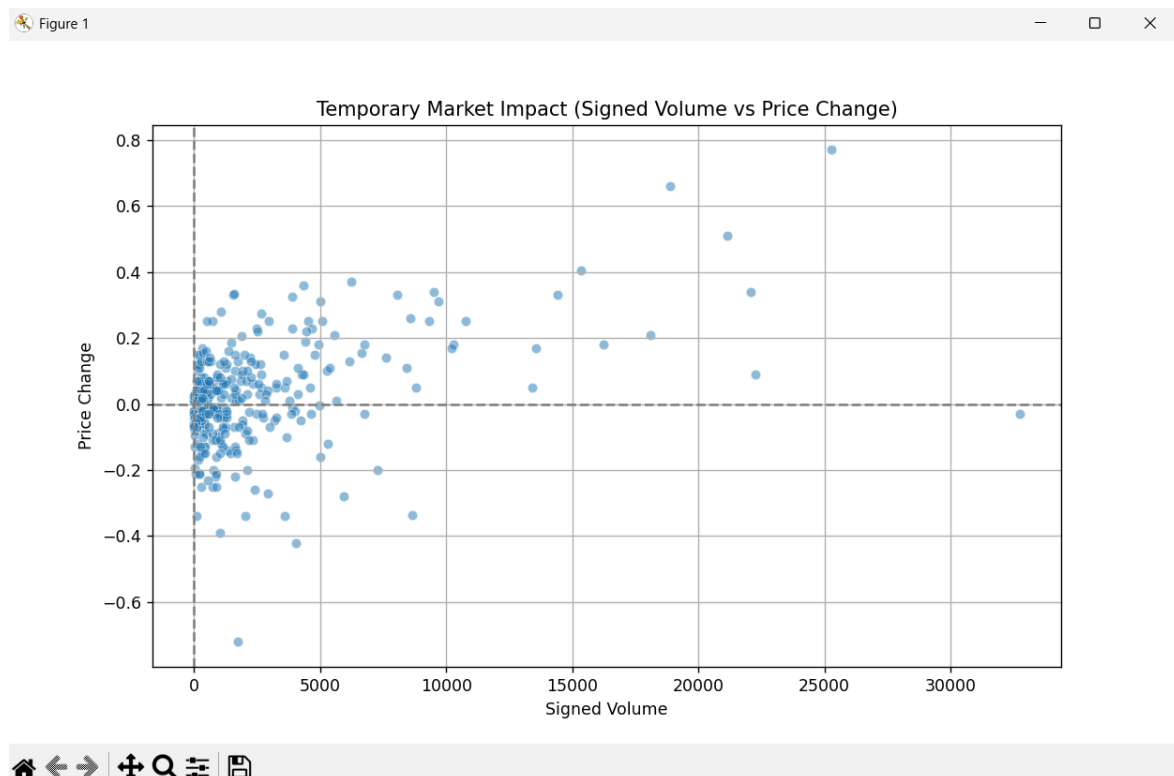
- Equality constraint:  $\sum v_t = V_{total}$
- Non-negativity constraint:  $v_t \geq 0$

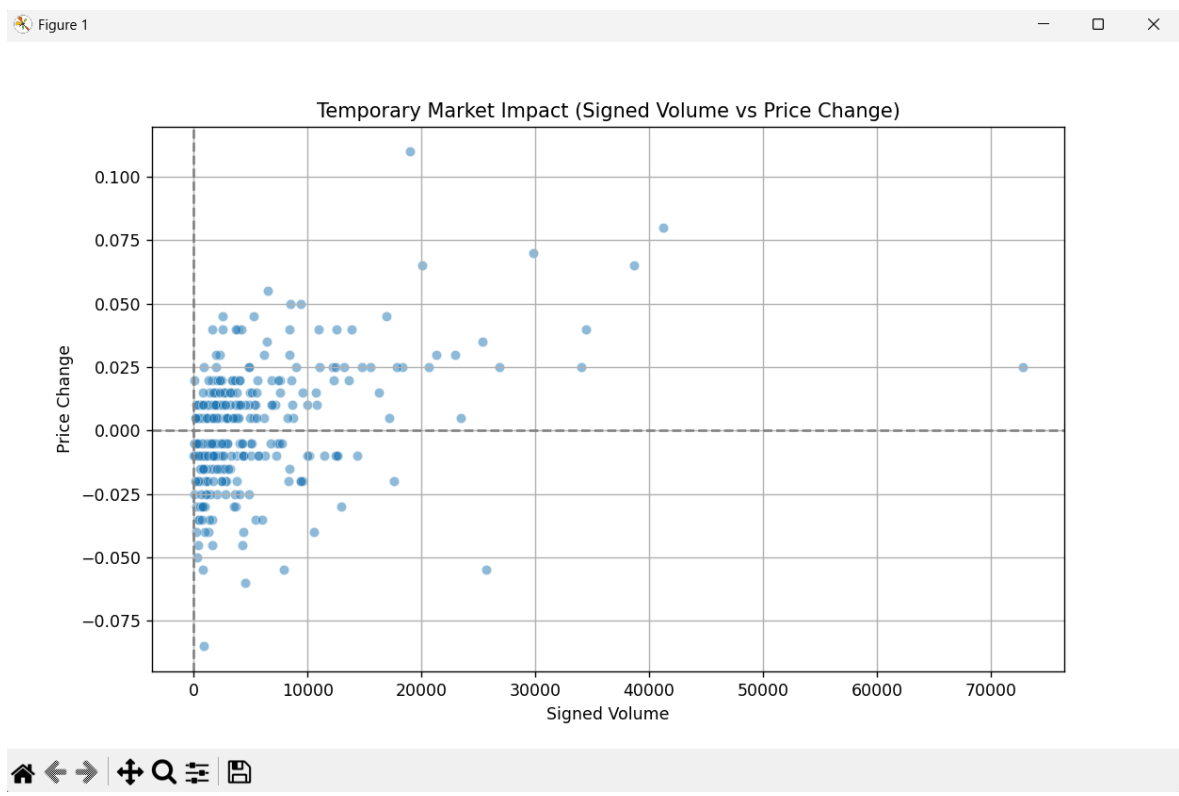
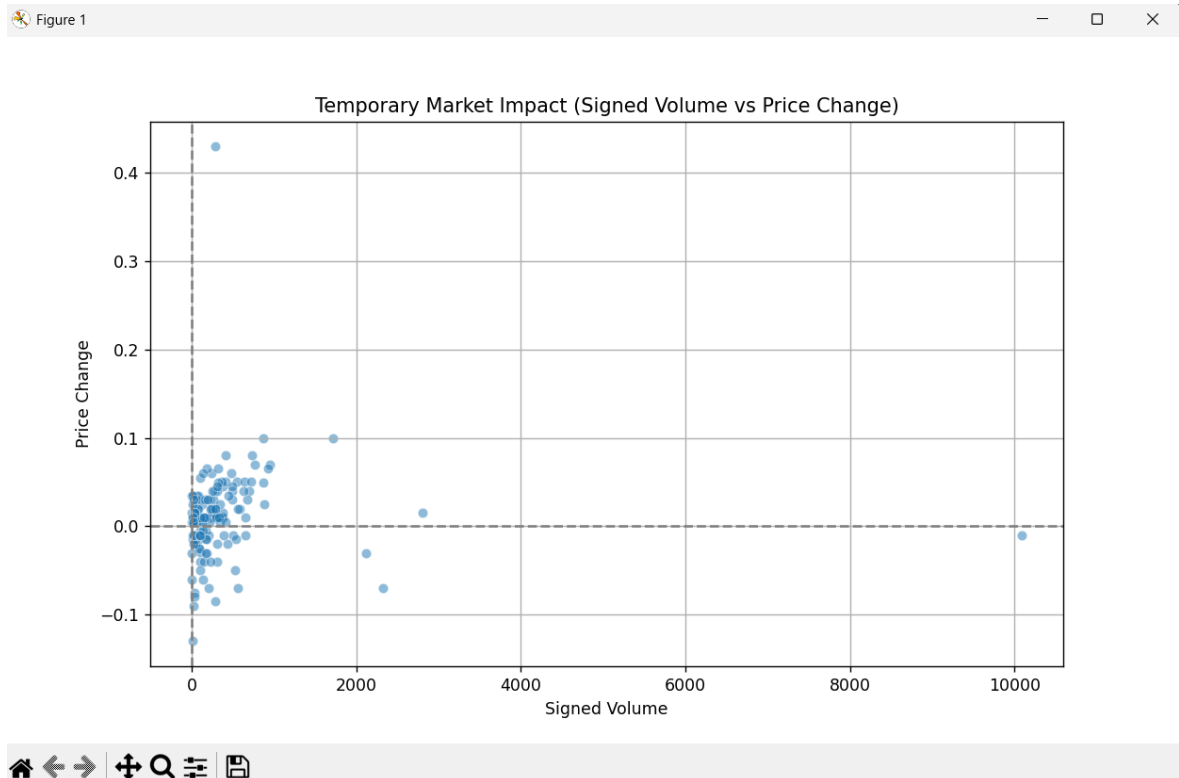
## Execution Insights

- **CRWV**: More uniform trade allocation due to moderate  $\beta$ .
- **FROG**: Spreads execution to low-impact intervals to avoid spikes.
- **SOUN**: Balanced slicing with preference to early intervals.

## 5. Technical Stack and Capabilities Demonstrated

Category	Skills / Tools
Programming	Python ( <code>pandas</code> , <code>numpy</code> , <code>cvxpy</code> , <code>matplotlib</code> )
Data Processing	Trade data normalization, resampling, aggregation
Statistical Modeling	OLS regression, price-volume relationship analysis
Optimization	Cost minimization using convex quadratic programming
Visualization	Scatter plots, regression fits, impact trend charts
Reporting & Documentation	README, report writing, GitHub project structuring





## 6. Deliverables

- Plots and visualizations
- Separate code for each dataset
- README.md with screenshots
- Final PDF report (this document)

- GitHub repository with full codebase: [amrita40/BlackHouse\\_Internship\\_Assessment](#)
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## 7. Conclusion

This project successfully models **temporary market impact** and formulates a **cost-efficient execution strategy** using real-time trade data. The use of regression-based impact modeling, paired with quadratic cost optimization, highlights a strong grasp of both quantitative finance and algorithmic execution.

The pipeline reflects the core competencies required for quantitative research roles — combining modeling rigor, market microstructure awareness, and coding efficiency.

**Thank you**

**Amritanshu Sinha**

Quantitative Research Internship Candidate  
July 29, 2025

