

# Spoofing Detection in Digital Asset Centralized Exchanges

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# Outline

- 1 Introduction
- 2 Our Methodology
- 3 Our Results
- 4 Next Steps

# Introduction

# The Why ?, The What ?

## Why Detect Manipulation

- Core to Sun Zu Lab's mission of transparency.
- Enhancement of clients' alpha in three ways.

## What we did

- 1 Implemented a function to warn investors of manipulating risk.
- 2 Pinned Orders suspected of being manipulative

# Limit Order Book (LOB) & Conventions

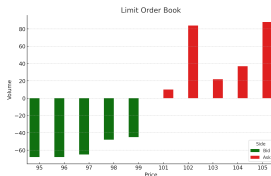
## Type of orders

### (1) Limit Orders :

- Bid Orders : The market-maker proposes to buy a given amount at a given price. (We will always use the maker convention)
- Ask Order : The maker proposes to sell.

### (2) Market Orders :

- Sell Trade : The order is an aggressive bid order, so the trade happens at Ask. The market-maker sells.
- Buy Trade : The order is an aggressive ask order, so the trade happens at Bid.



## Types of spoofing and our focus

- ① Spoofeer Maker : Places an order at best bid (resp. ask) and wishes to get executed fast. He places a manipulating order on the ask (resp. bid) side. He modifies the imbalance, and increase the execution probability of his initial order.
- ② Spoofeer Taker : Wants to initiate a market order at bid (resp. ask) at the lowest (resp. highest) possible price. He places a manipulating order deep in the book, that results in the appearance of new limit order on the same side that he can execute. He then cancels his order.
- We focused on the the spoofeer maker : manipulation in this case is much more profitable as the seller (resp. buyer) is able to avoid both crossing the spread and being applied taker fees.
- In this presentation, we will focus on the maker spoofeer willing to sell (i.e they manipulate at bid)

## Spoofers' behaviour

- The spoofer keeps his manipulating order until the next down best price movement : the market didn't react as expected, and fears the execution of his manipulating order.
- The spoofer puts a small order size at ask, so his fill ratio for this order is in  $\{0,1\}$

# Detection Algorithm



# Notations

- $\phi_a^{\delta, Q}$  : Fill ratio on ask side of an order placed  $\delta$  bps from the best, and of size  $Q$
- $W^{\text{maker}}$  : Wealth resulting of the sell of one ATS of a token, by spoofing
- $W^{\text{post-wait}}$  : Wealth resulting of the sell of one ATS of a token, with post-and-wait strategy.
- $\forall t \geq 0, p_t^b, p_t^a$  : Best Bid Price, Best Ask Price
- $\forall t \geq 0, \Delta p_t^b := p_t^b - p_0^b$
- $h := \inf(t > 0, \Delta_t^b < 0)$
- $f_b^+, f_a^+$  : maker fees on bid and ask side.
- $f_b^-, f_a^-$  : taker fees on bid and ask side
- $\mathbb{E}_x[.], \mathbb{P}_x$  : Expectancy and Probability parametrized by imbalance depth and value.
- $\mathbb{E}_{x^-}[.], \mathbb{P}_{x^-}$  : At the previous tick

# The Equations

- 1 We compare the expected wealth resulting from the spoofing strategy with the one resulting from a post-and-wait.
- 2 Spoofing Wealth :

$$\begin{aligned}\mathbb{E}_x[W^{\text{maker}}] = & \mathbb{E}_x[\phi_b^{\delta, Q}] Q[-f_b^+(p_0^b - \delta) + f_a^- p_0^b] \\ & + \mathbb{E}_x[\phi_b^{\delta, Q} \Delta p_h^b] Q f_a^- \\ & + \mathbb{P}_x(\phi_a^{0,1} = 0) f_a^-(p_0^b + \mathbb{E}_x[\Delta p_h^b \mid \phi_a^{0,1} = 0]) \\ & + \mathbb{P}_x(\phi_a^{0,1} = 1) f_a^+ p_0^a\end{aligned}$$

- 3 Post-and-Wait Wealth :

$$\begin{aligned}\mathbb{E}_{x^-}[W^{\text{post-wait}}] = & \mathbb{P}_{x^-}(\phi_h^{0,1} = 0) f_a^-(p_0^b + \mathbb{E}_{x^-}[\Delta p_h^b \mid \phi_a^{0,1} = 0]) \\ & + \mathbb{P}_{x^-}(\phi_a^{0,1} = 1) f_a^+ p_0^a\end{aligned}$$

- 4 Spoofer's excess wealth function :

$$W = \mathbb{E}_x[W^{\text{maker}}] - \mathbb{E}_{x^-}[W^{\text{post-wait}}]$$

## Calculating ATS

For each pair and venue, we compute over a day of data the average trade size. It will be used as parameter in the calculation of other functions.

## Calculating relevant depths

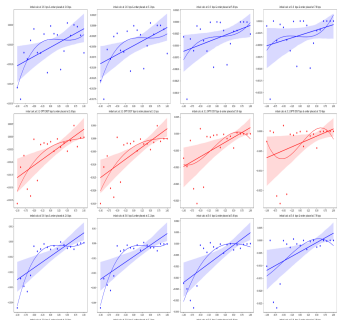
- For each venue and symbol, the book depth is different.
- We compute over a day the distances so that the cumulative bid volume over this distance amounts to a certain dollar value, between \$10,000 and \$1,000,000.

## Calculating the helper functions

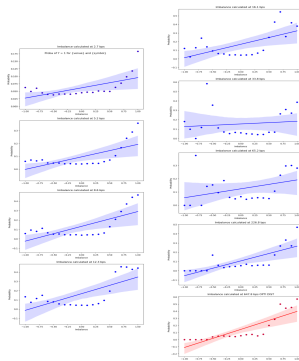
- We compute  $\mathbb{E}_x[\phi_b^{\delta, Q}]$ , etc for values of  $\delta$ ,  $Q$  (multiple of ATS), and values of  $x$  : The expectancy is a function of the order book imbalance and depth at which it is calculated.
- We compute and plot the functions with depth and imbalance values as arguments, and select the best depths with the following heuristic (given for fill-ratio as an example)
  - 1 For each depth, calculate the spearman correlation of imbalances and fill ratios. Map this correlation to the correct subset : Zone 1 :  $[0, 0.33]$ , ..., Zone 3 :  $[0.66, 1]$ . If only one depth corresponds to the best zone, take it. If tie :
  - 2 For each depth that ties, calculate the pearson correlation, and do the same thing. If tie :
  - 3 Select the biggest depth.
  - 4 Perform a linear regression to parametrize the model.

# Methodology (3)

Impact of depths on precalculated data



$\mathbb{E}[\phi_b^{\delta, Q}]$   
ETH-USD, gemini



$\mathbb{P}(\phi_a^{0,1} = 1)$   
XRP-USD, kraken

# Methodology (4)

## Running the spoofing suspicion algorithm

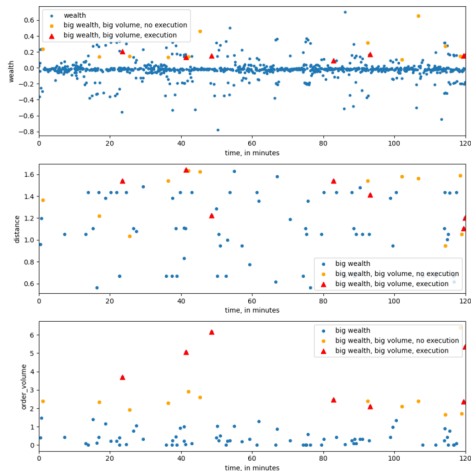


Figure: 2 hours of bitstamp, BTC-USD

# Post Analysis

Gain of execution probability, when Wealth is in top 10%.

Symbol & Venue	Exec Proba Gain
Kraken BTC-USD	(2%)
Bitstamp BTC-USDT	82%
Bitstamp SHIB-USD	0.17 %
Bitstamp XRP-USD	(22%)
Gemini BTC-USD	6 %
Gemini SHIB-USD	15 %

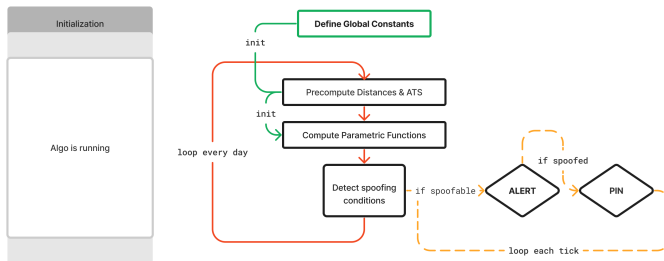
## What it tells us

The relevance of the wealth function is established, but varies in the time, from an exchange to another, from a symbol to another.

# Live Detection



# Algorithm Lifecycle



Live Spoofing Detection Algo Design  
Pattern

## What's next ?

- Detect Taker spoofer.
- Launch production.
- Think about a visual interface.

Thank you for your attention