

Crypto Market

There are 10+ liquid crypto exchanges (Binance, Coinbase, FTX, Huobi, OKX...) and hundreds of trading instruments. Cryptocurrency arbitrage strategies are based on low latency, which uses algorithmic trading to react to market events faster than the competition and increase profitability of trades.

Project Vision

We want to introduce a **real-time**, **low-latency** python-based trading system for **cryptocurrency arbitrage strategies**, including efficiently download historical data, record live data, train predictive model, and run model in real-time.



Project Achievements

- Build a crypto trading system that could be used to trade arbitrage trading strategies using Python.
- Connect to OKX crypto exchange based on REST/WebSocket API.
- Build functions such as data recording, signal/position calculation.
- Apply predictive models to trading strategies.
- Control the internal latency to the level of 7ms (from receiving data to sending orders).
- Extra functions: strategy monitor UI, database management, risk management, order management



Data Collection

Minute bar data

- BTC/BCH/ETH data 1/1/2020 4/22/2022, total 1,212,300 lines.
- Multi-threading with 5 threads improves the most with 1,568s.

Tick data

- BTC/BCH/ETH's spot, futures and swap contract data
 4/19/2022 4/22/2022, total 2,791,623 lines.
- An recorder function to register market data.
- Single thread implementation + SQLite database is used to store the live tick data.

time	ticker	open	high	low	close	volume
1/1/2020 00:00	BTC-USDT	7220	7220	7219.5	7219.5	2.7594
1/1/2020 00:01	BTC-USDT	7219.6	7222.9	7219.5	7222.9	2.841
1/1/2020 00:02	BTC-USDT	7223	7224.8	7222.9	7224.8	3.68631
1/1/2020 00:03	BTC-USDT	7224.8	7226.1	7224.8	7226	1.74307
1/1/2020 00:04	BTC-USDT	7226.1	7226.4	7225.7	7225.7	4.91684
1/1/2020 00:05	BTC-USDT	7225.8	7225.8	7225.3	7225.3	8.8941

Method	# of Threads/Processes	Time Cost (s)	
Theoretical optima		1212	
No speed up		3846	
Multi-threading	3	2628	
Multi-threading	5	1568	
Multi-processing	3	2811	
Multi-processing	5	1705	

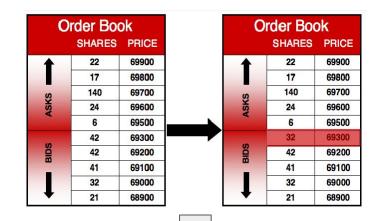


Features Selection

 Order book: buy and sell orders for a specific instrument organized by price level at a snapshot

$$Price\ Ratio\ = rac{Mid\ Price(Spot)}{Mid\ Price(Swap)}$$

$$Order \, Imbalance \, Ratio = \frac{(bid \, volume - ask \, volume)}{(bid \, volume + ask \, volume)}$$



Notation	Name			
x_1	Relative spread spot			
x_2	Relative spread swap			
x_3	Order imbalance ratio spot			
x_4	Order imbalance ratio swap			
x_5	$\Delta PriceRatio~(\%) for~the~past~10S$			
x_6	$\Delta PriceRatio~(\%) for~the~past~20S$			
x_7	$\Delta PriceRatio~(\%) for~the~past~60S$			
y	(30 Seconds)Change of future PriceRation			



Model Training Result & Optimization

Evaluation

- R2: total price movements explained by model
- IC: Information coefficient, shows how closely the analyst's financial forecasts match actual financial results; higher the better

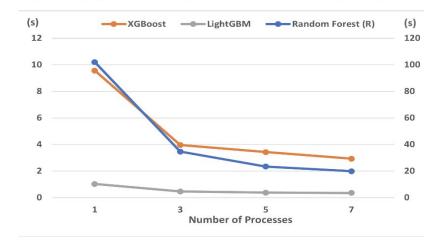
Optimization Methods Implemented

- Multi-threading
- Multi-processing

From 1 core to 7 cores:

- running speed of RF is increased by 413%
- running speed of XGBoost is increased by 226%

Model	# of Cores	R^2 Score	IC	Time (s)
Linear Reg	1	0.188	0.489	0.049
Ridge Reg	1	0.229	0.513	0.015
Random Forest	1	0.261	0.521	101.95
Random Forest	7	0.260	0.521	19.884
XGBoost	1	0.291	0.539	9.564
XGBoost	7	0.291	0.539	2.927
LGBM	1	0.132	0.386	1.022
LGBM	7	0.132	0.386	0.350





Trading System

Build system

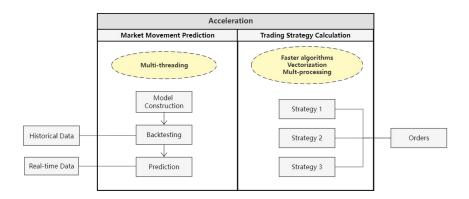
We build our trading system based on VeighNa, a python-based open-sourced trading system used to trade Chinese futures market.

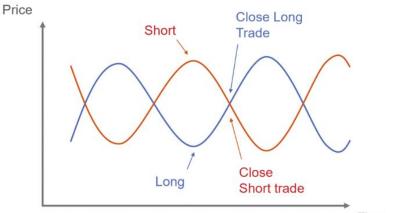
Add applications

We create our own applications including access to the OKX exchanges, data downloader, and trading strategy.

Load model

The predictive models are loaded in the trading system so we can predict the future price movements. Based on the prediction from the model (alpha), we can set a open limit parameter to decide when we will make the trades.







Trading System Latency

Latency Evaluation: Tick-to-Trade

Tick-to-Trade is the time interval between receiving a market tick data and processing the buy or sell order.

Implementation Plans

- (1) pandas DataFrame
- (2) NumPy array
- (3) Improved algorithms and built-in data structures including deque and hash-tables.

Model	Mean (ms)	Std (ms)
Plan 1: DataFrame	15.05	3.75
Plan 2: NumpyArray	10.33	3.07
Plan 3: Built-in data structures	7.05	2.06
with improved algorithms	7.05	2.00
- Update variables	0.02	0.01
- Model predict	7.19	2.04
- Strategy and Send orders	0.04	0.02

Limit Market	Advanced Limit	Stop	Trailing stop	Trigger	
Instrument	Order tim	e Side	Avg. fill Price	Filled Total	Filled Order Value
BTC/USDT >	2022-05-0	7 Buy	35,869.2 USDT	0.02004804 BTC	\$719.04
Isolated 5.00X	03:20:4	6 Buy	Market	1,144.4 USDT	\$1,144.29
BTCUSDT PERP >	2022-05-0	7	35,864.6 USDT	717.29 USDT	\$689.66
Cross 3.00X	03:20:4	4 Sell	Market	717.29 USDT	\$689.66
BTCUSDT PERP >	2022-05-0	7	35,811.3 USDT	716.22 USDT	\$689.66
Cross 3.00X	03:09:3	2 Buy	35,909.3 USDT	718.19 USDT	\$689.66
BTC/USDT >	2022-05-0	7	35,809 USDT	0.02 BTC	\$716.11
Isolated 5.00X	03:09:3	2 Sell	35,709.4 USDT	0.02 BTC	\$714.12
BTCUSDT PERP >	2022-05-0	7	35,807.3 USDT	716.14 USDT	\$689.66
Cross 3.00X	03:09:2	7 Sell	35,708,7 USDT	714.17 USDT	\$689.66



Trading System Demo

