Efficiency in the Bitcoin Market

Zichao Yang

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Motivation

The differences between bitcoin market and stock market:

- Bitcoin market is thinner than stock market.
- Bitcoin market does not have a real close or open time.
- There are hundreds of bitcoin exchanges operating in the world.
- Algorithm plays a role in deciding the pattern of bitcoin price.

In this paper, we focus on the last two points.

Motivation

What does this paper do?

We take the technical aspect of bitcoin into consideration and study how the transaction fee functions differently in the bitcoin market.

We argue that there are two layers of efficiency in the bitcoin market:

- Efficiency on exchange level.
- Efficiency on the whole market level.

These two types of efficiency do not always move in the same direction.

Literature review

(1) Studies about the predictability of bitcoin return:

- People's attention and bitcoin return: results are inconclusive. (Kristoufek, 2015; Ciaian et al., 2016; Urquhart, 2018; Aalborg et al., 2018; Dastgir et al., 2019)
- Political or economic uncertainty and bitcoin return: results are inconclusive. (Dyhrberg, 2016; Ciaian et al., 2016; Bouri et al., 2017; Demir et al., 2018; Al-Khazali et al., 2018)
- Trade volume and bitcoin return: results are inconclusive. (Balcilar et al., 2017; Bouri et al., 2018; El Alaoui et al., 2018; Kjærland et al., 2018)

Literature review

(2) Studies testing weak form efficiency:

The results are inconclusive. Bitcoin is the most efficient one when compared with alter-coins, and efficiency is positively related with liquidity. (Urquhart, 2016; Nadarajah and Chu, 2017; Zhang et al., 2018; Bouri et al., 2018)

(3) Price across exchanges

- Brandvold et al. (2015) find that, in the period from April 2013 to February 2014, big bitcoin exchanges provide more information on price.
- Giudici and Abu-Hashish (2018) find that bitcoin price is highly interrelated among different exchanges.

Blocks and the blockchain

Blocks are defined as storage units which contain confirmed transaction data.

Blockchain is a sequence of blocks, where each block contains a piece of information about the location of its previous block.

0d2946bfex28bfcx8bfcx8bfcx8bfcx8bfcx8bfcx8bfcx8bfcx						
3L4b2HAeukqElEwDyJwzQs3p25hiF5Adnh (0.26226892 BTC - Output)		bc1 bc1	qhfv57d4c3dz255q8hinkc7ehnwd9wcq5pcg0 q836d3kcmhc25w6p7p6qdf2thxksj4rvxzhs0k6	ux - (Unspent) Ə - (Spent)	0.00410706 BTC 0.19419986 BTC	
					1 Confirmations	0.19830692 BTC
Summary				Inputs and Outputs		
Size	246 (bytes)			Total Input	0.26226692 BTC	
Weight	654			Total Output	0.19830692 BTC	
Received Time	2019-04-03 19:38:21			Fees	0.06396 BTC	
Lock Time	Block: 570071			Fee per byte	26,000 sat/B	
Included In Blocks	570072 (2019-04-03 19:49:25 + 11 minutes)			Fee per weight unit	9,779.817 sat/WU	
Confirmations	1			Estimated BTC Transacted	0.00410706 BTC	
Visualize	View Tree Chart			Scripts	Hide scripts & coint	ase

Transaction View information about a bitcoin transaction

Mining and Proof-of-Work

The structure of block header:

Field	Description
Version	A version number to track software/protocol upgrades
Previous Block Hash	A reference to the hash of the previous block in the chain
Merkle Root	A hash of the root of the merkle tree of this block's transactions
Timestamp	The approximate creation time of this block
Target	The Proof-of-Work algorithm target for this block
Nonce	A counter used for the Proof-of-Work algorithm

This table is adopted from Table 10-3 in Antonopoulos (2017).

Block header hash: Hash Function

0000000000000000000 a e 165831 e 934 ff 763 a e 46a 2a 6c 172 b 3 f 1 b 60 a 8 c e 26 f 160 a 8 c e 26 f 1

Mining: a process of hashing the block header repeatedly. In each time, the miner change the value of nonce a little bit until the block header hash satisfies the target set by the **Proof-of-Work** algorithm.

More on Proof-of-Work

- Difficult to find the nonce, but easy to verify the finding
- Makes the blockchain almost immutable (with mining competition)
- Auto-adjustable, generates blocks, on average, every 10 minutes
- Adjustment not symmetric See graph

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Exchange trade volume



On-chain transaction volume



On-chain transaction volume



The procedure of transaction confirmation



Transaction fee

Transaction fees are left by users, then collected by miners.

When miners build their blocks, they select the transactions which have the highest reward-size ratio (RSR):

$$\mathsf{RSR} = \frac{\mathsf{transaction \ fee \ paid \ in \ this \ transaction}}{\mathsf{the \ size \ of \ this \ transaction \ data}}$$

Users make transaction decisions based on the ratio of transaction value to transaction fee (TFR):

$$\mathsf{TFR} = \frac{\mathsf{transaction fee paid in this transaction}}{\mathsf{the value of this transaction}}$$

Facts

Mempool size and transaction fee per transaction



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Number of transactions and transaction fee per transaction

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Facts

Tx fee per tx and total tx volume



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Exchange-level data

Exchange-level data comes from CryptoCompare. It includes the bitcoin close price in the seven largest exchanges (Bitfinex, Bitstamp, Coinbase, Gemini, Kraken, Lakebtc and Quoine) and market trade volume data from 2015-10-08 to 2019-03-03.

	Bitfinex	Bitstamp	Coinbase	Gemini	Kraken	Lakebtc	Quoine
	Unit: USD (dollar)						
Mean	3760.6	3749.2	3758.9	3755.4	3749.0	3973.3	3799.3
Median	2461.0	2502.0	2505.6	2497.6	2497.8	2598.3	2681.7
Maximum	19210.0	19187.8	19650.0	19500.0	19356.9	19315.7	19653.5
Minimum	243.6	243.4	243.4	243.9	242.7	240.6	244.5
Std. dev.	3781.3	3786.5	3808.0	3800.49	3783.6	3991.3	3871.8
Observations	1243	1243	1243	1243	1243	1243	1243

Table 2: Summary statistics of price

Price efficiency in a single exchange

We apply the autoregressive analysis to study the dynamics of bitcoin return. The order of lag terms is chosen based on AIC. The autoregressive model can be written as:

$$R_{i,t} = Const + A(L)R_{i,t} + \mu_t$$

where $R_{i,t}$ denotes bitcoin return from exchange *i* at time *t*. A(L) is lag operator and the order of lag terms is chosen based on AIC.

We combine the AR model with the rolling windows analysis (300-days window). Then we use an F-test to test the joint significance of all these lag terms in each regression. We count, in each period, how many exchanges have joint significant lag terms. There are seven exchanges, so the result in each period can range from 0 to 7.

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Plot results with the mean of exchange trade volume



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Price correlation among different exchanges

We still use bitcoin return data from the seven largest exchanges and conduct the test for Granger causality on the bitcoin return data for every two exchanges.

$$R_{i,t} = Const + A(L)R_{i,t} + B(L)R_{j,t} + \mu_t$$

where $R_{i,t}$ and $R_{j,t}$ denote bitcoin return from exchange *i* and *j* at time *t*, A(L) and B(L) are lag operators, which are chosen based on AIC.

We combine the Granger causality test with the rolling windows analysis. We calculate the number of correlated pairs (NCP) of exchanges by counting the number of exchanges, from which the return data can Granger cause the return of other exchanges in each period. In our dataset, there are seven different exchanges, so the maximum value of NCP is $7 \times 6 = 42$.

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Plot results with the inverse of CV of bitcoin price

The coefficient of variation (CV) of price can be defined as:

 $CV_t = rac{\text{Standard Deviation of the price in } t^{th} \text{ period}}{\text{Mean of the price in } t^{th} \text{ period}}$



Price dispersion and trade volume



Price correlation among different exchanges

Price dispersion and trade volume dispersion



Price dispersion and on-chain transaction volume

When on-chain trade volume is high, price dispersion becomes high and exchanges become more isolated with each other.



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Daily data

Daily blockchain-related data come from CoinMetrics. The date range is chosen from 2015-01-01 to 2018-01-01.

	on-chain tx volume/tx	tx fee/tx	bitcoin return
	log-form	log-form	log-form
Mean	1.032	-7.998	0.003
Median	1.022	-8.211	0.003
Maximum	2.124	-5.539	0.224
Minimum	0.250	-9.165	-0.236
Std. dev.	0.310	0.752	0.038
Observations	1096	1096	1096

Table 4: Summary statistics of blockchain-related data

Structural VAR

The structural VAR model is comprised of daily data on the log-form of bitcoin daily return (R_t) , the daily average on-chain transaction volume per transaction in BTC(V_t), the daily average transaction fee per transaction in BTC(F_t).

$$\begin{bmatrix} a_{11}^{0} & 0 & 0 \\ a_{21}^{0} & a_{22}^{0} & 0 \\ a_{31}^{0} & a_{32}^{0} & a_{33}^{0} \end{bmatrix} \begin{bmatrix} F_t \\ R_t \\ V_t \end{bmatrix} = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix} + \begin{bmatrix} a_{11}^{1} & a_{12}^{1} & a_{13}^{1} \\ a_{21}^{1} & a_{22}^{1} & a_{23}^{1} \\ a_{31}^{1} & a_{32}^{1} & a_{33}^{1} \end{bmatrix} \begin{bmatrix} F_{t-1} \\ R_{t-1} \\ V_{t-1} \end{bmatrix} + \dots + \begin{bmatrix} \epsilon_F \\ \epsilon_R \\ \epsilon_V \end{bmatrix}$$

Results



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Block-level data

We match minute-level trade volume and price data from Bitstamp with block-level total transaction fees and number of transactions data from BlockSci. Proxy Variable Explain

	Bitstamp tx volume/tx	tx fee/tx	bitcoin return
	log-form	log-form	log-form
Mean	-4.600	0.000	0.000
Median	-4.352	-0.003	0.000
Maximum	5.992	6.329	0.070
Minimum	-18.787	-6.290	-0.152
Std. dev.	2.037	0.411	0.004
Observations	150,785	150,785	150,785

Results



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Increase in total tx vol

Price dispersion increases

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Return increase

Increase in total tx vol

Price dispersion increases

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Increase in total tx vol



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Extensions

- Get more accurate block-level on-chain transaction volume data from BlockSci.
- Introduce structural breaks into our analysis.
- Use local projection method as a robustness check.

Future researches

- Cryptocurrency Confidence Survey
- A theoretical multi-exchanges model
- A network model using the data from BlockSci (block-level individual/clustered data)

BlockSci: how many clusters are involved in every block

Mt. Gox filed for bankruptcy at 9:30 am UTC, 2014-02-28.



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BlockSci: the value of transactions in every block

Mt. Gox filed for bankruptcy at 9:30 am UTC, 2014-02-28.



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Hash function

Hashing Algorithm



Plain Text: arbitrary length Hashed Text: certain length (like 32 bytes)

Mining and Proof-of-Work

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Supplemental materials

Difficulty and computing power (hash rate)



On-chain transaction volume VS. Exchange trade volume

We match daily on-chain transaction volume data from CoinMetrics with minute-level trade volume data from Bitstamp. The correlation is 0.886.



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