## Risk of Bitcoin Addresses to be Identified from Features of Output Addresses

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### Cryptocurrency and Anonymity of Bitcoin

Outflow incident of NEM

□Do not arrest the criminal

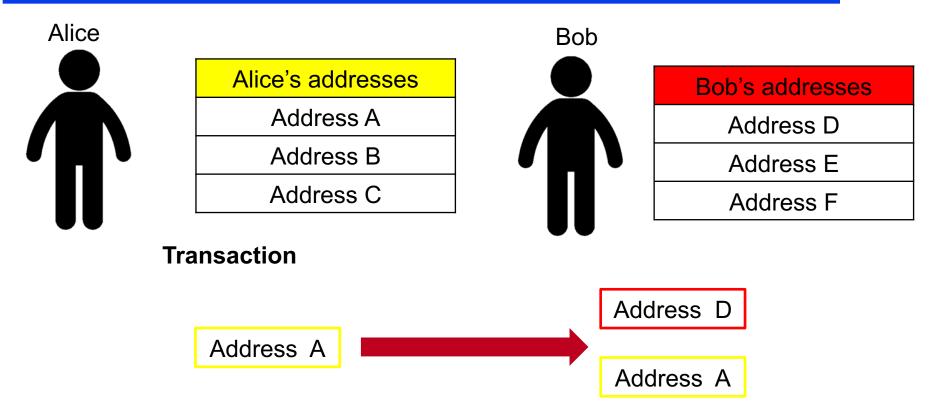
□Difficult to trace the stolen NEM



#### Coincheck: World's biggest ever digital currency 'theft'

https://www.bbc.co.uk/news/world-asia-42845505

### Bitcoin user and address



- Addresses are pseudonyms
- User are not be identified from an address(anonymity)

#### **Previous studies & Objectives**

Several studies have been done for the deanonymization of bitcoin addresses

□ Combined addresses managed by the same user[Sarah,2013]

□Revealed a target user's time zone by analyzing the time distribution of transactions[Dupont,2015]

How much anonymous Bitcoin address is?

#### Problem

We have no ground truth

□Nobody knows who has which address

□It is impossible to identify the owner from an address

### **Our Solutions**

Collected the ground truth in two ways

- 1. Addresses that have been published via website Bitcointalk
- 2. Addresses that have been specified by the coinbase

& Summary Name: Posts: Activity:	macbook-air macbook-air 324 324	Addr	Name	Locatio n				
Merit: Position: Date Registered: Last Active: ICQ:	250 Sr. Member May 30, 2011, 01:02:02 AM September 02, 2017, 08:29:08 AM		macbook- air	China				
AIM: MSN: YIM: Email: Website: Current Statue Bitcoin	hidden F2Pool IC Office 1KFHE7w8BhaENAswwryaoccDb6qcT6DbYY	1DNNERMT5MMusfYnCBfcKCBjBKZWB C5Lg2	BitHits	None				
address: Gender: Age: Location: Local Time: Trust:	Male N/A China February 05, 2018, 02:20:59 PM <b>0</b> : -0 / +0	1Anduck6bsXBXH7fPHzePJSXdC9AEsR mt4	Anduck	None				

Profile page in Bitcointalk

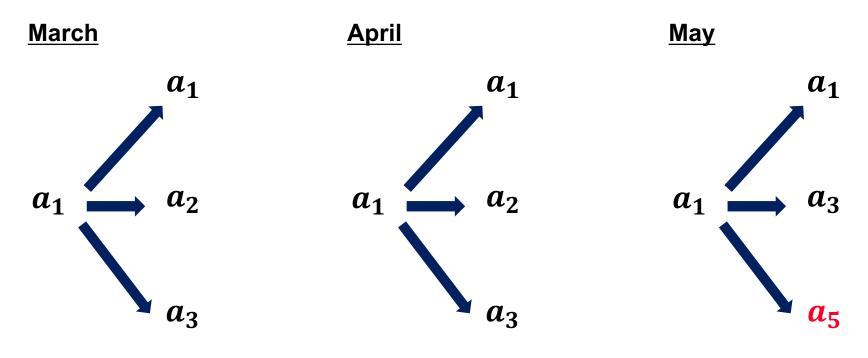
#### What is coinbase?

#### Example of block

	ID	Input	Output	Remittance[10 <sup>-8</sup> ]	
Coinbase 💻	$\rightarrow Tx_1$	N/A	<i>a</i> <sub>2</sub>	250000000	
	$Tx_2$	a2	<i>a</i> <sub>4</sub>	900000	
	$Tx_3$	<i>a</i> <sub>3</sub>	$a_2, a_3$	6000000	<b>R</b> eward
	Tx <sub>4</sub>	$a_2, a_2, a_5$	<i>a</i> <sub>1</sub> , <i>a</i> <sub>2</sub>	11000000	
	$Tx_5$	a <sub>3</sub>	$a_1, a_2, a_3, a_5$	4000000	

#### Propose method(Jaccard re-identification)

• Note that an output addresses



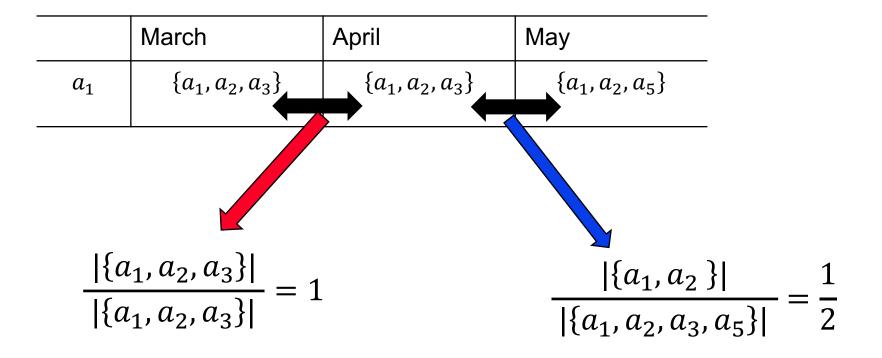
 $\{a_1, a_2, a_3\}$ 

 $\{a_1, a_2, a_3\}$ 

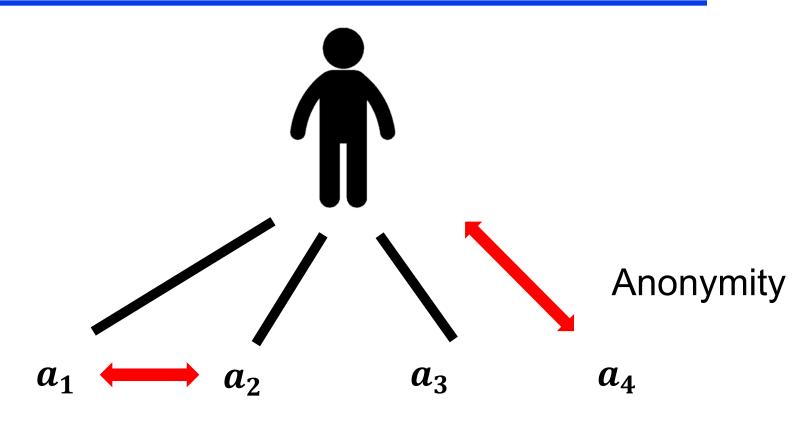
 $\{a_1, a_3, a_5\}$ 

#### Jaccard coefficient

$$J(A,B) = \frac{|A \cap B|}{|A \cup B|}$$
: Jaccard coefficient



### Anonymity and Unlinkability



Unlinkability

#### **Research questions**

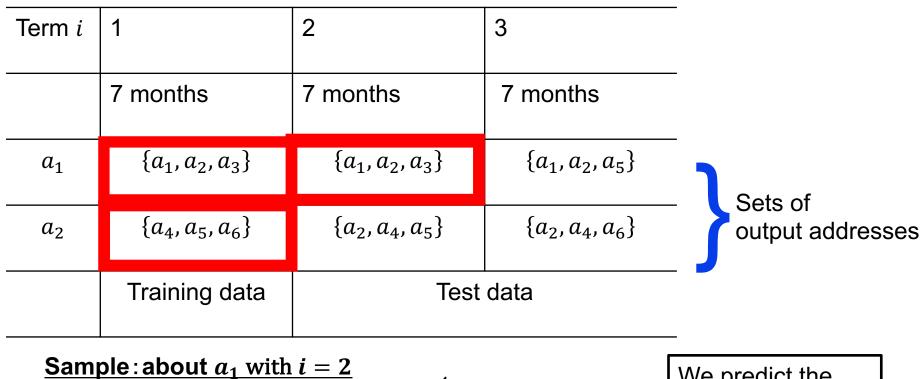
- Does the number of transactions for an address affect an unlinkability?
- Which is more unlinkable output address set or transaction time set [Dupont,2015]?
- 3. How often is address identified?

#### Experimental method

- Divide transaction data into training data and test data
- 2. Predict the answer of the address of the test data by Jaccard re-identification
- 3. Calculate Recall, Precision and re-identification rate

Period	2012.09.22 – 2014.05.10(About 1.5 years)		
Address number	559		
Block number	200,001 – 300,000(100,000 blocks)		

#### Jaccard re-identification



$$\begin{array}{c} \text{nple: about } a_1 \text{ with } i = 2 \\ \{a_1, a_2, a_3\} \\ \{a_1, a_2, a_3\} \\ \{a_4, a_5, a_6\} \end{array} \begin{array}{c} 1 \\ \{a_1, a_2, a_3\} \\ \{a_1, a_2, a_3\} \end{array} \begin{array}{c} \text{We predict the} \\ \text{first one was sent} \\ \text{from the same} \\ \text{user to the} \\ \text{training set} \end{array} \right]_{12}$$

### Average recall, average precision and reidentification rate

Average recall *R* 

$$R = \frac{1}{n} \cdot \sum_{i=1}^{n} R_i$$

Average precision P

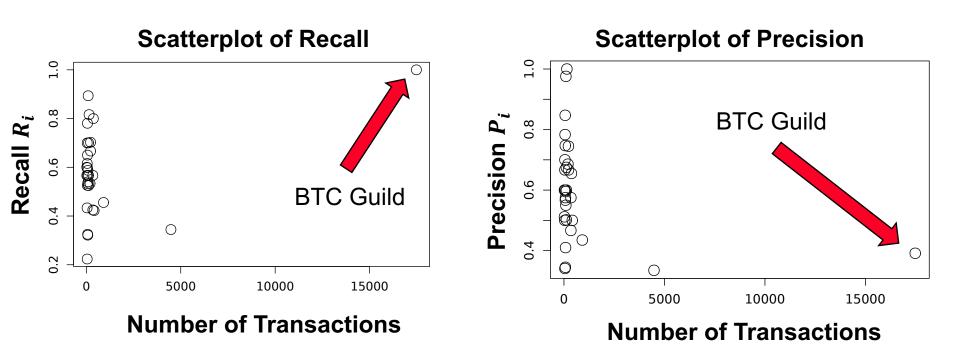
$$P = \frac{1}{n} \cdot \sum_{i=1}^{n} P_i$$

Re-identification rate

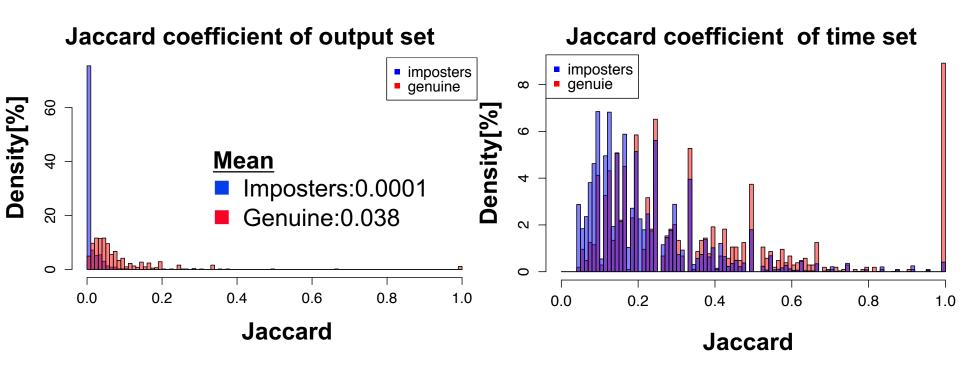
$$F = \frac{2 \cdot R_i \cdot P_i}{R_i + P_i}$$

An address can be re-identified if the F-measure is more than 0.5

# Result1 : Recall and precision by number of transactions in address

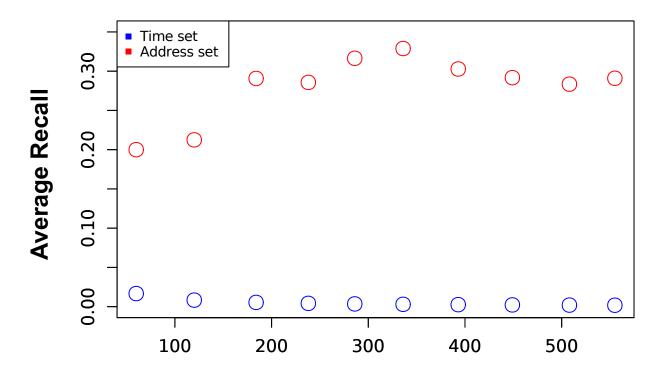


## Result2 : Comparison of output address set and time set



- The blue one is distributed within very small value
- The sets of transaction time are widely distributed

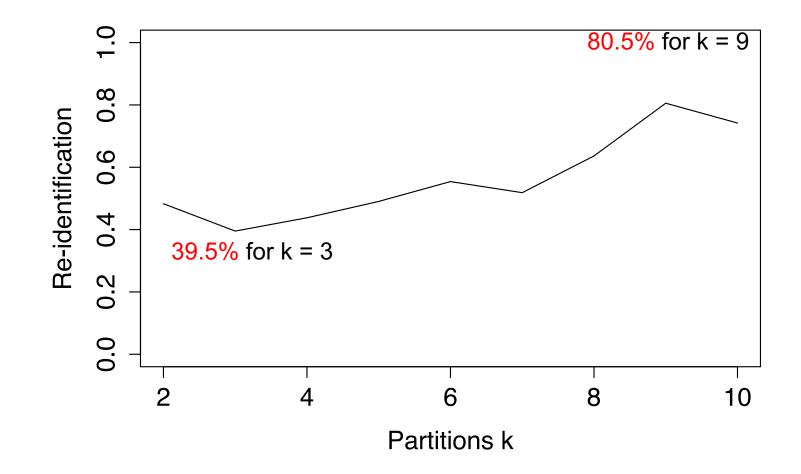
# Result3 : The average recall with respect to the number of addresses



#### Number of Addresses

• The average recall of the output address set is independent from the number of addresses.

### Result4 : Distribution of re-identification rate with respect to the number of partitions k



#### **Research questions**

 Does the number of transactions for an address affect a n unlinkability?

 $\rightarrow No$ 

 Which is more unlinkable, output address set or transaction time set [Dupont,2015]?

 $\rightarrow$  output address set

3. How often is address identified?

→80.5%

#### Conclusions

- An unlinkability is not affected by number of transactions
- Output address set affected an unlinkability by 30% more than time set
- 80.5% of addresses can be identified from the Jaccard coefficient between subsets of output addresses