

# Sucker punch makes you richer: The threat of rational miners

Runchao Han\*  
Monash University and  
CSIRO-Data61  
runchao.han@monash.edu

Zhimei Sui\*  
East China Normal University  
zhimeisui@gmail.com

Jiangshan Yu  
Monash University  
jiangshan.yu@monash.edu

Joseph Liu  
Monash University  
joseph.liu@monash.edu

Shiping Chen  
CSIRO-Data61  
shiping.chen@data61.csiro.au

## ABSTRACT

Honest majority is a vital assumption of Bitcoin-style blockchains. However, recent 51% attacks render this assumption unrealistic in practice. In this paper, we analyse two possible 51% attacks launched by a rational miner, who is profit driven. The first attack considers a rational miner, who moves his mining power from a stronger blockchain to a weaker blockchain to launch 51% attacks, provided that (1) the mining power is compatible in both blockchains, and (2) the transferred mining power dominates the weaker blockchain. We say a blockchain is stronger if the total mining power of this blockchain is higher than other (weaker) blockchains. The second attack considers a rational miner, who rents cloud mining power to launch 51% attacks. The former attack is new, and we name it *mining power migration attack*; the latter is called *cloud mining attack*, and it was initially covered by the bribery attack (FC' 16).

We formalise the two attacks by using Markov Decision Process. We then test the feasibility of launching both attacks on leading blockchains in the wild by using our model. We show that both attacks are feasible and profitable. For example, our result shows that with 12.5% mining power of Bitcoin, a rational miner can gain approximately 6% (\$18,946.5 USD) extra profit than honest mining, by launching mining power migration attack to double spend a transaction of 3000 BCH (equivalent to \$378,930) on BitcoinCash. We also investigate the 51% attack on Ethereum Classic happened in Jan. 2019, by applying our model into this attack to provide some insights to understand more about it.

## 1 INTRODUCTION

**Key assumption.** Honest majority has been a key assumption to guarantee the security of all proof-of-work (PoW) based blockchains. It assumes that no single attacker is able to control more than 50% of mining power in the entire network. If this assumption does not hold, then the so called 51% attack would be possible, and an attacker is able to rewrite the blockchain and spend a coin more than once [1].

It is obvious that the security guarantee is relative rather than absolute, as the difficulty to break the assumption is directly related to the total mining power in the system. So, the more the mining power in the system, the more difficult to control 51% mining power of the entire system, and so the higher the security guarantee will be. To encourage more miners to join the system to increase the mining power (so the security guarantee of the system), Bitcoin [1] introduced an incentive mechanism. In particular, miners invest

mining power, such as hashing power, to solve a crypto puzzle as the proof of their work to create a block in the blockchain. As return on their contribution to the mining power, miners who successfully created a block will get some reward. For example, with Bitcoin, the mining reward includes a pre-defined number of bitcoins and transaction fees of all transactions contained in the block. In this way, miners are encouraged to contribute their mining power to make the system more secure. This incentive concept has been employed by almost all major blockchains.

**Fact v.s. Fiction.** As mentioned, the honest majority assumption is relative to the mining power in the system. If there exists only one blockchain in the entire world, then all potential miners will contribute their mining power into this blockchain. In this case, controlling 51% mining power could be extremely difficult. However, if there is a competing blockchain of the same mining algorithm, then the total available mining power for this algorithm in the world will be split into two communities, one for each blockchain. This reduces the difficulty of controlling 51% mining power in both of the blockchains.

Currently, there are over 2,000 cryptocurrencies<sup>1</sup>, and many of them share the same mining algorithm. So that no single blockchain can take full advantage of the proof-of-work mechanism for their security.

In 2016, bribery attacks [2] (a.k.a. cloud mining attack) considered the possibility of renting cloud mining power to gain a majority of mining power in Bitcoin-style blockchains. In fact, since 2018, many successful 51% attacks have been identified, as shown in Table 1. These attacks result in the loss of more than \$23 million US dollars. One main stream suspicion [3, 4] is that the attacks are equipped with cloud mining power from e.g. Nicehash[5].

**Table 1: Successful 51% Attacks in 2018-2019 [6]**

DATE	Coin	Loss
April 4, 2018	Verge (XVG)	~\$1,100,000.
May 14, 2018	MonacoCoin (Mona)	~\$90,000.
May 22, 2018	XVG (Verge)	~\$1,750,000.
May 29, 2018	Bitcoin Gold (BTG)	~\$18,000,000.
June 2, 2018	ZenCash (ZEN)	~\$550,000.
June 4, 2018	Litecoin Cash (LCC)	Unknown
September 8, 2018	FLO Blockchain (FLO)	~\$27,500.
November 8, 2018	Aurum Coin (AU)	~\$500,000.
December 2, 2018	Vertcoin (VTC)	~\$100,000.
January 7, 2019	Ethereum Classic (ETC)	~\$1,100,000.

**Miners are rational.** In 2005, years before the birth of Bitcoin, BAR (Byzantine, Altruistic, Rational) model [7] was considered in

\*Both authors contributed equally to the paper

<sup>1</sup><https://coinmarketcap.com/all/views/all/>

the context of fault tolerance for cooperative services. In this model, a participant can be Byzantine, altruistic, or rational.

A Byzantine (a.k.a. malicious) participant would try everything to break the security of the system with any cost; an altruistic (a.k.a. honest) participant will always follow the system specification; and a rational attacker would only depart from the protocol specification and being Byzantine if s/he can gain more profit.

As the nature of the incentive mechanism in blockchains is to encourage miners joining the system with reward, so it is reasonable to assume that most miners are rational, rather than Byzantine or altruistic.

**Mining power migration attack.** Given the fact that (1) multiple blockchains share the same mining algorithm; (2) total mining power in different blockchains could be very different; and (3) miners are rational, we suspect the possibility of a new attack, which we call mining power migration attack.

In this attack, we consider two blockchains ( $BC_1$  and  $BC_2$ ) of the same type of mining algorithms, i.e. mining power in  $BC_1$  is compatible with  $BC_2$ . A rational miner of  $BC_1$  controls a considerable portion (much less than 50%) of mining power. However, if moving (a part of) the rational miner's mining power to  $BC_2$ , then the miner may control a majority of mining power in  $BC_2$ .

This may motivate this miner to move a sufficient portion of his mining power from  $BC_1$  to  $BC_2$  to launch 51% attacks, if this would give him some extra revenue.

**Contribution.** We aim to understand the security guarantee of Bitcoin-style blockchains, in the presence of rational miners. In particular, we provide a study on two possible "Sucker Punch Attacks" (SPA), where the attacker is rational and is willing to gain extra profit by "hit-and-run" the weaker blockchain  $BC_2$ , through our identified mining power migration attack, or the previously proposed cloud mining attack. Our contributions are summarised as follows:

- we identify a new way to launch 51% attack, called mining power migration attack;
- we propose SPA-MDP to formalise the two "Sucker Punch Attacks" (SPA) using Markov Decision Process (MDP);
- we implement the model in Python, and evaluated the impact of different parameters; the model provides an estimated cost of an attack, and the relative revenue of the attack. This allows us to generate recommendations on how to setup the parameters, including the maximum value of transactions and the number of confirmations, for a given blockchain;
- we provided a feasibility study on the two attacks, by testing several selected blockchains in the wild in our model. The result includes estimated cost of successful attacks on different blockchains, and the potential net revenue gained from the attack;
- we investigate the 51% attack on Ethereum Classic (ETC) in Jan. 2019, by applying our model into this attack to provide some insights on this attack.

**Key results** Our result shows that both attacks are feasible. In fact, on some blockchains it is very profitable to do so. For example, we show in Section 5 that with 12.5% mining power of Bitcoin (BTC), a rational miner can gain approximately 6% (\$18,946.5 USD) extra profit (than honest mining) by double-spending a transaction

of 3000 BCH (equivalent to \$378,930) on BitcoinCash. This required mining power is in fact not difficult to achieve. At the time of writing (10/May/2019), the top two mining pools in BTC, i.e. BTC.com with 15.5% mining power and AntPool with 14.1% mining power, already achieves this requirement.

In addition, we simulate and analyse the 51% attack on Ethereum Classic (ETC) in Jan. 2019 [8] in order to quantify the incentive and investigate the attacking strategy of the attacker. By using our model, we found that the attacker's net revenue from the attack is close to the value (\$100,000) that the attacker gave back to Gate.io. Moreover, we describe the attacker's behaviors during the attack, and these actions are proved to be smart strategies to maximize and stabilize his revenue.

We also show that our model can be used to adjust different parameters of a blockchain to eliminate the incentive of a rational miner. In particular, we use the 51% attack on ETC as an example, and show that if we go back to Jan 2019, and set the maximum value of a transaction to 9,000 ETC (approximately \$38,340), then the attacker will not get any net revenue. If an exchange wants to allow higher value transactions, e.g. 52,800 ETC (approximately \$224,928) which is the largest value of transactions being double spent in this attack, then our model recommends that the number of confirmation should be increased from 12 to 18, to eliminate the incentive of a rational miner.

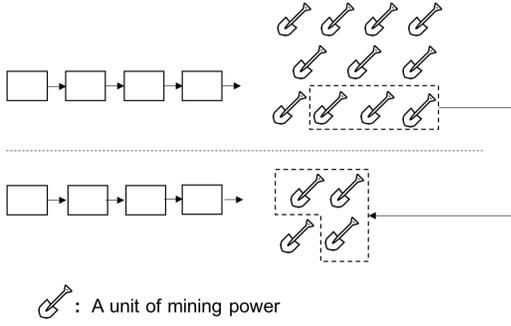
**Paper organisation.** This paper is structured as follows. Section 2 provides background knowledge on the Bitcoin-style blockchains. Section 3 presents our system model and the formalisation of the two sucker punch attacks (SPA) in Markov Decision Process (MDP). In section 4, we implement the SPA-MDP in Python, and evaluate the impact of different parameters on the two attacks. In section 5, we evaluate the security of selected blockchains based on our model, and investigate the recent 51% attack on EthereumClassic in January 2019. In Section 6, we discuss potential remedies based on the observations we derived from our model, and using ETC as an example to show how to use our model to provide recommendations to a given blockchain. We provide related work in Section 7 and conclude in Section 8.

## 2 BACKGROUND

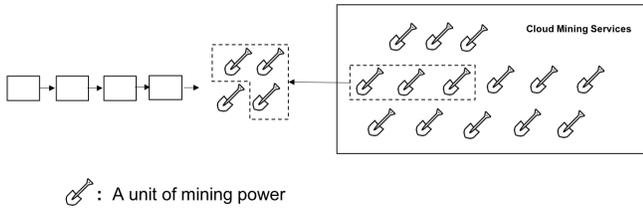
Proof-of-work (PoW) based consensus was first introduced by Bitcoin [1], to allow distributed nodes in the system agreeing on the same global state.

In Bitcoin, all transactions are recorded as a chain of blocks. Anyone can create a block of transactions, and append it into the Bitcoin blockchain. To create a block, one needs to solve a crypto puzzle which is computational hard. In particular, one needs to find a random nonce to make the hash value of the block smaller than a target value. The target value is dynamically adjusted by the system every 2016 blocks. This guarantees that new blocks are created every ten minutes on average.

Different miners may create conflict valid blocks, and this creates a fork in the system. To resolve a fork, miners only accept the longest branch. However, a branch that is currently longer may be over taken by the other branch, and all transactions in the currently longer chain will be deemed invalid. This creates an opportunity for



**Figure 1: A high-ranked blockchain miner can migrate his mining power to a low-ranked blockchain, and he is able to launch a 51% attack on low-ranked blockchain.**



**Figure 2: Cloud mining services provide miners extra mining power, so that an adversary can increase his mining power quickly and launch a 51% attack.**

the attacker to spend a coin in a branch, and later on creates another longer branch to erase this transaction. This is called “double spending attack”.

Intuitively, to launch a double spending attack, an attacker should have enough mining power to create a branch that is longer than the current one. This requires the attacker to control a majority of mining power in the network. An attacker with a majority of mining power to double spend is known as “51% attacks”.

Currently, all PoW-based blockchains have to assume that the majority of mining power is honest. To enforce this assumption, Bitcoin-style systems incentivise more miners (so more mining power) to join the system through a reward system. Each time a block is created, the creator will obtain some reward in two parts. The first part is a pre-defined number of coins; and the second part is transaction fees carried by all transactions in the block. However, as mentioned previously, recent 51% attacks render the assumption on honest majority unrealistic.

### 3 FORMALIZATION

This section presents our system model, the two aforementioned attacks, and our formalisation of the attacks in Markov Decision Process.

#### 3.1 System model

We consider a world of more than one blockchain, and some of the blockchains share the same type of mining algorithms. For simplicity, we define two blockchains,  $BC_1$  and  $BC_2$ , sharing the

**Table 2: Notations of Parameters**

Notations	definition
$D_1$	The difficulty of $BC_1$
$D_2$	The difficulty of $BC_2$
$d$	The fraction of $BC_1$ 's difficulty towards $BC_2$ 's difficulty ( $d = \frac{D_1}{D_2}$ )
$H_{a,1}, H_{h,1}$	The honest and adversarial mining power on $BC_1$ , respectively
$H_{a,2}, H_{h,2}$	The honest and adversarial mining power on $BC_2$ , respectively
$H_a, H_h$	The total honest and adversarial mining power, respectively ( $H_a = H_{a,1} + H_{a,2}, H_h = H_{h,1} + H_{h,2}$ )
$h_1$	The fraction of the adversarial mining power towards the honest mining power on $BC_1$ ( $h_1 = \frac{H_{a,1}}{H_{h,1}}$ )
$h_2$	The fraction of the adversarial mining power towards the honest mining power on $BC_2$ ( $h_2 = \frac{H_{a,2}}{H_{h,2}}$ )
$R_1$	The mining reward of a block of $BC_1$
$R_2$	The mining reward of a block of $BC_2$
$r$	The fraction of $BC_1$ 's mining reward of a block towards $BC_2$ 's ( $r = \frac{R_1}{R_2}$ )
$v_{tx}$	The amount of the attacking transactions
$\gamma$	The propagation parameter of the adversary
$pr$	Renting price of a hash algorithm when mining <sup>2</sup>
$\beta$	The fraction of migrated mining power by the adversary

same mining algorithm. We call the blockchain with more mining power a *stronger blockchain*, and another one *weaker blockchain*. Since they share the same mining algorithm, the mining difficulty of a stronger blockchain is higher than the weaker one, as they need to create the same number of blocks within the same time period. In this paper,  $BC_1$  is the stronger blockchain, and  $BC_2$  is the weaker one. We also consider rentable cloud mining power from organizations such as NiceHash, and the cloud mining power is compatible with the used mining algorithm.

We consider a rational miner, who is profit driven. When considering mining power migration attack, we assume that the rational miner already has a considerable portion (less than 50%) of mining power on  $BC_1$ . When considering cloud mining power attack, we assume that the attacker has some initial investment to rent sufficient cloud mining power. Later in Section 5, we will analyse the soundness of such an environment. Figure 1 and Figure 2 present examples of the two attacks.

For the simplicity of analysis, we also assume that the non-attacking mining power in the two blockchains are constant during the attack. That is, during the time period of the attack, only the attacker is able to move its mining power back and forth between the two blockchains, or to put his rented mining power into the  $BC_2$ . This assumption is reasonable as most of the time the mining power is fairly stable at least within a small time period.

More formally, we define the system environment as follows. A summary of notions is available in Table 2. Let  $BC_1$  and  $BC_2$  be the stronger blockchain and the weaker blockchain in Section 3.1, where  $BC_2$  is the attacker's target. Let  $pr$  be the price of renting a unit of mining power (e.g. hash per second) for a time unit. Let  $D_1$  and  $D_2$  be the difficulties,  $R_1$  and  $R_2$  be the mining rewards of  $BC_1$  and  $BC_2$ , respectively. We have  $D_1 > D_2$  and  $R_1 > R_2$ . We define

the fractions of the difficulties and mining rewards as:  $d = \frac{D_1}{D_2}$ ,  $r = \frac{R_1}{R_2}$ .

For *mining power migration attack*, we define mining-related parameters for two blockchains. Let  $H_{h,1}, H_{h,2}$  be the honest mining power, and  $H_{a,1}, h_{a,2}$  be the adversarial mining power of  $BC_1$  and  $BC_2$ , respectively. Let  $H_1 = H_{h,1} + H_{a,1}$  and  $H_2 = H_{h,2} + H_{a,2}$  be the total mining powers on  $BC_1$  and  $BC_2$ , respectively. Note that  $H_1 > H_2$  according to the security model. Let  $h_1 = \frac{H_a}{H_{h,1}}$  and  $h_2 = \frac{H_a}{H_{h,2}}$  be the fractions of the adversarial mining power towards the honest mining power on  $BC_1$  and  $BC_2$ . Let  $\beta = \frac{H_{a,2}}{H_a}$  be the fraction of migrated mining power by the adversary.

For *cloud mining attack*, since the mining power is not coming from another blockchain, we only consider the target blockchain  $BC_2$ . Let  $H_{h,2}$  be the honest mining power, and  $H_a$  be the rentable mining power from the cloud mining service. Let  $h_2 = \frac{H_a}{H_{h,2}}$  be the fraction of rented mining power out of rentable mining power. Let  $\beta = \frac{H_{a,2}}{H_a}$  be the fraction of rented mining power towards the rentable mining power. Let  $pr$  be the price of renting mining power (in  $\$/h/s$ ).

Let  $\gamma \in [0, 1]$  be the propagation parameter of the adversary, i.e. the connectivity of the adversary within the network. When the adversary and the honest miners release blocks simultaneously,  $\gamma$  equals to the fraction of the network that agrees on the adversary's block.

### 3.2 Formalization

We design SPA-MDP, a Markov Decision Process (MDP) to formalize the two sucker punch attack (SPA) processes. It takes blockchain parameters as input, and produces the optimal relative revenue of the adversary. In detail, our MDP model is a four-element tuple  $M := (S, A, P, R)$  where:

- $S$  is the state space containing all possible states of an adversary.
- $A$  is the action space containing all possible actions performed by an adversary.
- $P$  is the stochastic transition matrix presenting the probabilities of all state transitions.
- $R$  is the reward matrix presenting the rewards of all state transitions.

We detail each element of  $M$  below, and present an overview on the state transitions and reward matrices of SPA-MDP in Table 3.

**3.2.1 The State Space  $S$ .** The state space  $S$  is defined as a tuple  $S := (l_h, l_a, \beta, fork)$ , where  $l_h$  and  $l_a$  are the length of the honest and the adversarial blockchain, respectively;  $\beta$  is the portion of mining power at  $BC_2$  allocated by the adversary; and  $fork$  represents the state of the adversarial branch.

For simplicity, we choose  $\beta$  from (0.0, 0.2, 0.4, 0.6, 0.8, 1.0). The state  $fork$  of the adversarial branch has three values, defined as follows:

- irrelevant ( $fork = ir$ ) means the adversarial branch is published and confirmed in network. This indicates a successful attack.

- relevant ( $fork = r$ ) means the adversarial branch is published but the honest chain is confirmed by the network. This indicates that the attack is unsuccessful at present. (Note that the adversary can keep trying and may succeed in the future.)
- private ( $fork = p$ ) means the adversarial branch is private and only the adversary is mining on it. This indicates that an attack is in process.

**3.2.2 The Action Space  $A$ .** An adversary can perform the following actions:

- **ADOPT** The adversary accepts the branch of the honest network and discards his own adversarial branch, which means the adversary aborts his attack.
- **OVERRIDE** The adversary publishes his adversarial branch (which is longer than the honest one). Consequently, the honest branch is overridden, and the payment transaction from the adversary is successfully reverted.
- **MATCH** The adversary publishes his fork with the same length as the honest blockchain. This action has three variants  $MATCH$ ,  $MATCH\_INC$ , and  $MATCH\_DEC$ , where  $\_INC$  and  $\_DEC$  represent the increase and decrease of malicious mining power ratio  $\beta$ , respectively.
- **WAIT** The adversary keeps mining on his own fork. This action can be performed in two scenarios. One is  $l_h < nConfirm$ , indicating that the merchant is still waiting for the payment confirmation. Another one is that after the adversary publishes his blockchain by  $MATCH$ ,  $l_a \leq l_h$  still holds. In addition, there are three types in this action  $WAIT$ ,  $WAIT\_INC$ ,  $WAIT\_DEC$ , which represent the adversarial mining power adjustment, similar to that of **MATCH**.

**3.2.3 The State Transition Matrix  $P$ .** The State Transition Matrix  $P$  is defined as  $S \times A \times S' : Prob(S \times A \Rightarrow S')$ , where the participant in the state  $S$  does the action  $A$  to transit his state to  $S'$  with the probability  $Pr(S \times A \Rightarrow S')$ . In the double-spending context, the state transition is invoked by a new block (during **WAIT**), a blockchain branch selection (by **OVERRIDE** or **MATCH**) or quitting the attack (by **ADOPT**).

In MDP, an action can trigger a state  $S$  to transit to another state  $S'$  with some probability. Note that the resulting state  $S'$  can have multiple possibilities  $S'_1, S'_2, \dots, S'_n$ , and  $\sum_{i=1}^n Pr(S \times A \rightarrow S'_i) = 1$ .

The state transition matrix  $P$  (Table 3) describes all state transition possibilities.  $P$  is a 3-dimensional matrix ( $S \times A \times S'$ ), where  $S$  is the current state,  $S'$  is the resulting state, and  $A$  is the action triggering this state transition. The value of this matrix is the possibility of the corresponding state transition.

*Without Eclipsed Mining Power ( $A = WAIT[\_INC\_DEC]$ ).* When  $A = WAIT[\_INC\_DEC]$ , the adversary is mining his private adversarial branch alone. The next state update is triggered by a newly created block either by the honest network or by the adversary, with a probability directly associated to their mining power on  $BC_2$ , namely  $H_{a,2} = \beta H_a$  of the adversary and  $H_{h,2}$  of the honest network.

Therefore, the probability that the adversary gets the next block ( $l_a \rightarrow l_a + 1$ ) is

**Table 3: State transitions and reward matrices of SPA-MDP. Notations are summarised in Table 2.**

State $\times$ Action	Resulting State	Probability	Reward				Condition
			$cost_{multichain}$	$cost_{cloudmining}$	blockreward	$v_{tx}$	
$(l_h, l_a, \beta, fork)$ , ADOPT	$(0, 0, \beta, ir)$	1	0	0	0	0	$l_h > l_a \geq nConfirm$
$(l_h, l_a, \beta, fork)$ , OVERRIDE	$(0, 0, \beta, ir)$	1	0	0	$l_a R_2$	$v_{tx}$	$l_a > l_h \geq nConfirm$
$(l_h, l_a, \beta, fork)$ , WAIT	$(l_h, l_a + 1, \beta, p)$	$\frac{\beta h_2}{\beta h_2 + 1}$	$\frac{-\beta h_2 R_1}{d(1+\beta h_2)}$	$\frac{-\beta h_2 D_2 p r}{1+\beta h_2}$	0	0	$l_h < nConfirm$
	$(l_h + 1, l_a, \beta, p)$	$\frac{1}{\beta h_2 + 1}$	$\frac{-\beta h_2 R_1}{d(1+\beta h_2)}$	$\frac{-\beta h_2 D_2 p r}{1+\beta h_2}$	0	0	$l_h < nConfirm$
$(l_h, l_a, \beta, fork)$ , WAIT_INC	$(l_h, l_a + 1, \beta + 0.2, p)$	$\frac{(\beta + 0.2) h_2}{(\beta + 0.2) h_2 + 1}$	$\frac{-(\beta + 0.2) h_2 R_1}{d(1+(\beta + 0.2) h_2)}$	$\frac{-(\beta + 0.2) h_2 D_2 p r}{1+(\beta + 0.2) h_2}$	0	0	$l_h < nConfirm$
	$(l_h + 1, l_a, \beta + 0.2, p)$	$\frac{1}{(\beta + 0.2) h_2 + 1}$	$\frac{-(\beta + 0.2) h_2 R_1}{d(1+(\beta + 0.2) h_2)}$	$\frac{-(\beta + 0.2) h_2 D_2 p r}{1+(\beta + 0.2) h_2}$	0	0	$l_h < nConfirm$
$(l_h, l_a, \beta, fork)$ , WAIT_DEC	$(l_h, l_a + 1, \beta - 0.2, p)$	$\frac{(\beta - 0.2) h_2}{(\beta - 0.2) h_2 + 1}$	$\frac{-(\beta - 0.2) h_2 R_1}{d(1+(\beta - 0.2) h_2)}$	$\frac{-(\beta - 0.2) h_2 D_2 p r}{1+(\beta - 0.2) h_2}$	0	0	$l_h < nConfirm$
	$(l_h + 1, l_a, \beta - 0.2, p)$	$\frac{1}{(\beta - 0.2) h_2 + 1}$	$\frac{-(\beta - 0.2) h_2 R_1}{d(1+(\beta - 0.2) h_2)}$	$\frac{-(\beta - 0.2) h_2 D_2 p r}{1+(\beta - 0.2) h_2}$	0	0	$l_h < nConfirm$
$(l_h, l_a, \beta, fork)$ , MATCH	$(l_h, l_a + 1, \beta, ir)$	$\frac{\beta h_2 + \gamma}{\beta h_2 + 1}$	$\frac{-\beta h_2 R_1}{d(1+\beta h_2)}$	$\frac{-\beta h_2 D_2 p r}{1+\beta h_2}$	$\frac{(l_a + 1) R_2 \beta h_2}{\beta h_2 + \gamma}$	$v_{tx}$	$l_h = l_a \geq nConfirm$
	$(l_h + 1, l_a, \beta, r)$	$\frac{1 - \gamma}{\beta h_2 + 1}$	$\frac{-\beta h_2 R_1}{d(1+\beta h_2)}$	$\frac{-\beta h_2 D_2 p r}{1+\beta h_2}$	0	0	$l_h = l_a \geq nConfirm$
$(l_h, l_a, \beta, fork)$ , MATCH_INC	$(l_h, l_a + 1, \beta + 0.2, ir)$	$\frac{(\beta + 0.2) h_2 + \gamma}{(\beta + 0.2) h_2 + 1}$	$\frac{-(\beta + 0.2) h_2 R_1}{d(1+(\beta + 0.2) h_2)}$	$\frac{-(\beta + 0.2) h_2 D_2 p r}{1+(\beta + 0.2) h_2}$	$\frac{(l_a + 1) R_2 (\beta + 0.2) h_2}{(\beta + 0.2) h_2 + \gamma}$	$v_{tx}$	$l_h = l_a \geq nConfirm$
	$(l_h + 1, l_a, \beta + 0.2, r)$	$\frac{1 - \gamma}{(\beta + 0.2) h_2 + 1}$	$\frac{-(\beta + 0.2) h_2 R_1}{d(1+(\beta + 0.2) h_2)}$	$\frac{-(\beta + 0.2) h_2 D_2 p r}{1+(\beta + 0.2) h_2}$	0	0	$l_h = l_a \geq nConfirm$
$(l_h, l_a, \beta, fork)$ , MATCH_DEC	$(l_h, l_a + 1, \beta - 0.2, ir)$	$\frac{(\beta - 0.2) h_2 + \gamma}{(\beta - 0.2) h_2 + 1}$	$\frac{-(\beta - 0.2) h_2 R_1}{d(1+(\beta - 0.2) h_2)}$	$\frac{-(\beta - 0.2) h_2 D_2 p r}{1+(\beta - 0.2) h_2}$	$\frac{(l_a + 1) R_2 (\beta - 0.2) h_2}{(\beta - 0.2) h_2 + \gamma}$	$v_{tx}$	$l_h = l_a \geq nConfirm$
	$(l_h + 1, l_a, \beta - 0.2, r)$	$\frac{1 - \gamma}{(\beta - 0.2) h_2 + 1}$	$\frac{-(\beta - 0.2) h_2 R_1}{d(1+(\beta - 0.2) h_2)}$	$\frac{-(\beta - 0.2) h_2 D_2 p r}{1+(\beta - 0.2) h_2}$	0	0	$l_h = l_a \geq nConfirm$

$$\begin{aligned}
 P(l_a \rightarrow l_a + 1) &= \frac{H_{a,2}}{H_{a,2} + H_{h,2}} \\
 &= \frac{\beta H_a}{\beta H_a + H_{h,2}} = \frac{\beta h_2}{\beta h_2 + 1}
 \end{aligned} \tag{1}$$

And vice versa for  $l_h \rightarrow l_h + 1$ .

$$P(l_h \rightarrow l_h + 1) = 1 - P(l_a \rightarrow l_a + 1) = \frac{1}{\beta h_2 + 1} \tag{2}$$

With Eclipsed Mining Power ( $A = \text{MATCH}[_\text{INC}]_\text{DEC}$ ). Besides the mining power owned by the adversary, the eclipsed mining power of  $\gamma H_{h,1}$  mines on the adversarial blockchain after a **MATCH** attempt. Therefore, the possibility of  $l_a \rightarrow l_a + 1$  becomes

$$\begin{aligned}
 P(l_a \rightarrow l_a + 1) &= \frac{H_{a,2} + H_{eclipsed}}{H_{a,2} + H_{h,2}} \\
 &= \frac{\beta H_a + \gamma H_{a,2}}{\beta H_a + H_{h,2}} = \frac{\beta h_2 + \gamma}{\beta h_2 + 1}
 \end{aligned} \tag{3}$$

And vice versa for  $l_h \rightarrow l_h + 1$ .

$$P(l_h \rightarrow l_h + 1) = 1 - P(l_a \rightarrow l_a + 1) = \frac{1 - \gamma}{\beta h_2 + 1} \tag{4}$$

**3.2.4 The Reward Matrix  $R$ .** The Reward Matrix  $R$  is defined as  $S \times A \times S' : Re(S \times A \Rightarrow S')$ , where the participant in the state  $S$  transits to a new state  $S'$  with the reward  $Re(S \times A \Rightarrow S')$ . The reward from a double-spending attack contains two parts: the block reward (including transaction fees) of the published longer blockchain and the double-spending transaction. To calculate the net reward, we also need to consider the cost of launching an attack. Here, we define the reward  $Re(S \times A \Rightarrow S')$  as a tuple  $(cost, blockreward, v_{tx})$  to fit into the MDP model, where  $cost$  represents the cost of launching a double-spending attack,  $blockreward$  represents the reward from the mined blocks including transaction fees, and  $v_{tx}$  represents the reward from the double-spent transaction.

With a state transition  $S \times A \rightarrow S'$ , an adversarial will get some reward, which can be of a positive or negative value. The state transition matrix  $R$  is a 3-dimension matrix ( $S \times A \times S'$ ), where  $S, A$  and  $S'$  are the same as  $P$ , but the value of this matrix is the reward of the corresponding state transition.

In our context, the reward consists of two parts, namely the block reward  $blockreward$  on  $BC_2$  and the reward gained from the double-spent transaction.

*Cost.* To calculate net revenue, we also need to consider a ‘‘negative reward’’, which is the cost of launching attacks. We use  $cost_{multichain}$  and  $cost_{cloudmining}$  to denote the cost of launching the mining power migration attack and the cloud mining attack, respectively. Compared to honest mining on  $BC_1$ , the cost  $cost_{multichain}$  of mining power migration is mainly from the loss of block rewards from  $BC_1$  due to the migrated mining power. Consequently, the cost can be computed as  $hashrate \cdot time \cdot difficulty \cdot R_1$ , which is the estimated mined block multiplies the block reward of  $BC_1$ . For **ADOPT** and **OVERRIDE** actions, the time of finishing a state transition is negligible. Meanwhile, for **WAIT**-style and **MATCH**-style actions, the state transition is triggered by mining a new block, so the time it takes is depending on the difficulty of mining a block. Therefore, the  $cost_{multichain}$  under **WAIT**-style and **MATCH**-style actions can be computed as follows:

$$\begin{aligned}
 cost_{multichain}(l_a \rightarrow l_a + 1) &= cost_{multichain}(l_h \rightarrow l_h + 1) \\
 &= hashrate \cdot R_1 \cdot time \cdot \frac{1}{difficulty} \\
 &= -\beta H_a \cdot R_1 \cdot \frac{D_2}{H_{h,2} + \beta H_a} \cdot \frac{1}{D_1} \\
 &= \frac{-\beta h_2 R_1}{d(1 + \beta h_2)}
 \end{aligned} \tag{5}$$

When conducting double-spending attacks by cloud mining, the cost is from renting the cloud mining power. The cloud mining

power is priced as  $\$(h/s)/s$ , which means the price of renting a mining power unit for a time unit. We denote the cloud mining power price as  $pr$ . Similar with the mining power migration, only **WAIT**-style and **MATCH**-style actions take a non-negligible time period. Therefore, either the cost of **ADOPT** or **OVERRIDE** is 0, and the cost of **WAIT**-style and **MATCH**-style actions is computed as

$$\begin{aligned}
 R_{BC_1}(l_a \rightarrow l_a + 1) &= R_{BC_1}(l_h \rightarrow l_h + 1) \\
 &= \text{hashrate} \cdot \text{price} \cdot \text{time} \\
 &= -\beta H_a \cdot Pr \cdot \frac{D_2}{H_{h,2} + \beta H_a} \quad (6) \\
 &= \frac{-\beta h_2 D_2 Pr}{1 + \beta h_2}
 \end{aligned}$$

*Block Reward on  $BC_2$ .* The adversary can get the block reward *blockreward* on  $BC_2$  only when his private adversarial branch is broadcasted and accepted by the honest network. Therefore, only **OVERRIDE** and the winning scenarios of **MATCH**-style actions have a positive reward, while *blockreward* = 0 under other scenarios.

When performing **OVERRIDE**, the adversarial blockchain of length  $l_a$  is directly accepted, so *blockreward* =  $l_a R_2$ . When performing **MATCH**-style actions, the adversary needs to get the next block so that his blockchain overrides the honest one. Therefore, *blockreward* =  $(l_a + 1)R_2$  holds for the winning scenarios.

*Reward from the Double-Spending Transaction  $v_{tx}$ .* Similar with *blockreward*, the adversary gets the double-spent money only when it successfully overrides the honest branch. Therefore,  $v_{tx}$  equals to the transaction amount for **OVERRIDE** and the winning scenarios of **MATCH**-style actions, while  $v_{tx} = 0$  for other scenarios.

## 4 MODEL EVALUATION

We implement our SPA-MDP in Python, and theoretically evaluate the impact of different parameters on the two attacks. We show the impact of different parameters on the relative revenue of an attacker. Later in the next section, we will make use of the real-world data to demonstrate the feasibility of these attacks.

### 4.1 Experimental methodology

*4.1.1 Experimental setting.* Our model is implemented in Python 2.7 and relies on the *pymdptoolbox* library [9], which is migrated from the Matlab MDP Toolbox. All experiments run on a MacBook Pro with a 2.2 GHz Intel Core i7 Processor, a 16 GB DDR4 RAM and 256 SSD storage disk.

*4.1.2 Markov decision process.* The proposed MDP model is infinite due to the unbounded  $l_a$  and  $l_h$ , so we convert the model to the finite MDP for the implementation. This is done by giving an upper bound *limit* for  $l_a$  and  $l_h$ . We apply the *ValueIteration* algorithm [10] with a discount value of 0.9 and an epsilon value of 0.1 in our MDP-based model.

### 4.2 Impacts of parameters

As presented in Table 4, we classify the parameters into five aspects, namely mining status related parameters, incentive related parameters, adversary network related parameters, the vigilance of

**Table 4: Value of parameters**

	Notation	Default	Permuted
Mining Status	$D_1$	100	N/A
	$D_2$	10	np.arange(5, 100, 5)
	$h_1$	0.1	N/A
	$h_2$	2.0	np.arange(1, 10, 1)
Incentive-Related Parameters	$R_1$	50	N/A
	$R_2$	5	np.arange(5, 50, 5)
	$v_{tx}$	100	np.arange(5, 100, 5)
Adversary Network	$\gamma$	0.3	np.arange(0.1, 1.0, 0.1)
the Vigilance of the Merchant	$nConfirm$	4	np.arange(1, 10, 1)
Mining Power Price	$pr$	2	np.arange(0.2, 4, 0.2)

the merchant related parameters, and mining power price related parameters. The mining status related parameters include mining difficulty ( $D_1$  and  $D_2$ ) and the ratio of adversarial mining power ( $h_1$  and  $h_2$ ). The incentive related parameters includes mining reward ( $R_1$  and  $R_2$ ) and the value  $v_{tx}$  of an adversarial transaction. The adversarial network related parameters include the propagation parameter  $\gamma$  of the adversary. The vigilance of the merchant related parameters include the number  $nConfirm$  of required block confirmations. The mining market related parameters include  $pr$ .

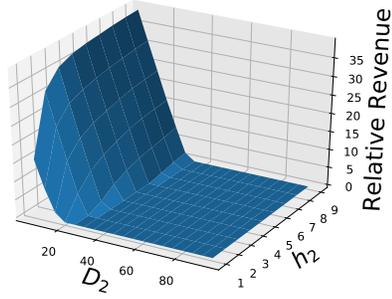
We evaluate the impact of each aspect on the relative revenue of an adversary in both types of attacks, by utilizing the *Control Varieties Method*. For aspects with multiple parameters (e.g. the mining status and the incentive), we correlate them by using a 3D surface. Since both attacks have common parameters  $D_2$ ,  $h_2$ ,  $R_2$ ,  $v_{tx}$ ,  $\gamma$  and  $nConfirm$ , we evaluate them on *mining power migration attack* only to avoid the repetition.

*4.2.1 Impact of mining status.* Figure 3a shows the impact of mining-related parameters on the relative revenue. We observe that the relative revenue increases monotonically with  $D_2$  decreasing and  $h_2$  increasing.

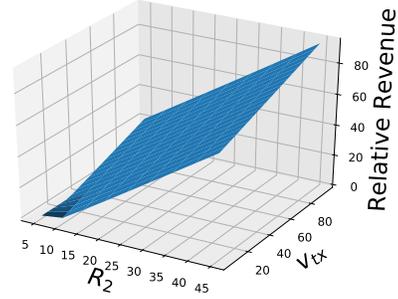
When  $D_2$  decreases, mining on  $BC_2$  will be easier, so that migrating to  $BC_2$  will be more profitable. This encourages both types of our attacks on  $BC_2$ . When  $h_2$  increases, launching both types of attacks on  $BC_2$  will be more possible to succeed, so the net revenue of attacks will increase. This also encourages 51% attacks on  $BC_2$ . Therefore, both decreasing  $D_2$  and increasing  $h_2$  incentivise 51% attacks on  $BC_2$ .

*4.2.2 Impact of incentive-related parameters.* Figure 3b shows the impact of incentive-related parameters on the relative revenue. We observe that increasing  $R_2$  and  $v_{tx}$  leads the adversary to profit more.

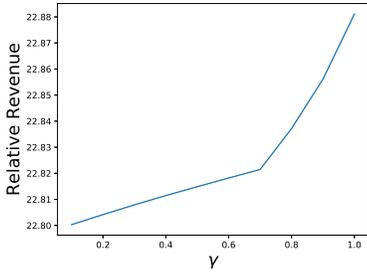
When  $R_2$  increases, mining  $BC_2$  will be more profitable, and 51% attacks on  $BC_2$  will also be more profitable. This encourages both types of 51% attacks on  $BC_2$ . The 51% attack generates  $v_{tx}$  out of thin air, so  $v_{tx}$  is the direct revenue of the 51% attack, and increasing  $v_{tx}$  directly increases the relative revenue. Therefore, both increasing  $R_2$  and  $v_{tx}$  incentivise 51% attacks on  $BC_2$ .



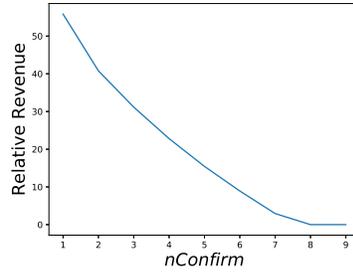
(a) Impact of  $h_2$  and  $D_2$ .



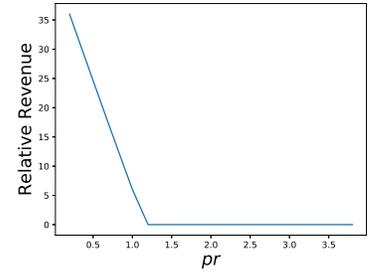
(b) Impact of  $v_{tx}$  and  $R_2$ .



(c) Impact of  $\gamma$ .



(d) Impact of  $nConfirm$ .



(e) Impact of  $pr$ .

Figure 3: Impacts of Parameters.

**4.2.3 Impact of adversary network.** Figure 3c shows the impact of  $\gamma$  on the relative revenue. In particular, we can see that the relative reward increases slightly with  $\gamma$  increasing. Interestingly, when the attacker's propagation parameter  $\gamma = 0.7$ , the curve slope increases.

According to our model,  $\gamma$  counts only when the adversary launches the *MATCH* action. When  $h_2 \geq 1$ , the adversary can always launch the 51% attack, regardless of the reward. Therefore, the *MATCH* action is an infrequent choice compared to *OVERRIDE*, so the influence of  $\gamma$  is negligible in our case.

The slope change is suspected to be when  $\beta H_a + \gamma H_{h,2} \geq (1 - \gamma)H_{h,2}$ . At that point, the allocated mining power from the adversary plus his eclipsed honest mining power outperforms the un-eclipsed honest power. Consequently, the adversary is confident to override the small blockchain by *MATCH* action.

**4.2.4 Impact of the vigilance of the merchant.** Figure 3d shows the impact of  $nConfirm$  on the relative revenue. We observe the relative revenue decreases monotonically with  $nConfirm$  increasing, and finally reaches 0.

More block confirmations require the adversary to keep mining secretly for a longer time. This leads to greater cost for launching the 51% attack through both types of attacks, and discourages 51% attacks on  $BC_2$ .

**4.2.5 Impact of mining power price.** The impact of the mining power price  $pr$  is shown in Figure 3e. We observe that the relative revenue decreases sharply with  $pr$  increasing, and finally reaches 0.

When the price of renting mining power is low, the related blockchains are vulnerable to the cloud mining attack as the attack cost is also low. Increasing  $pr$  leads to greater cost of launching 51% attack through renting cloud mining power, which will discourage this kind of 51% attacks on  $BC_2$ .

## 5 FEASIBILITY STUDY

In this section we evaluate our model with the real-world blockchains, including Bitcoin, BitcoinCash, Ethereum, EthereumChassic, Monero, and ByteCoin. In addition, we also revisit and use our model to explain the 51% attacks on EthereumChassic in January 2019 [11].

### 5.1 Testing blockchains in the wild

In order to evaluate the feasibility and profitability of both attacks, we apply our model to real-world blockchains. First, we evaluate the *mining power migration attack* on 3 pairs of top-ranked blockchains with the same hash algorithm, i.e., (1) Bitcoin (BTC) and BitcoinCash (BCH) with Sha256d, (2) Ethereum (ETH) and EthereumChassic (ETC) with Ethash, and (3) Monero(XMR) and ByteCoin (BCN) with CryptoNight. Second, we evaluate the *cloud mining attack* on 15 PoW blockchains chosen from the top 100 blockchains by their market caps [12]. Unfortunately, our result shows that both attacks are feasible and profitable on almost all selected blockchains.

**5.1.1 Mining Power Migration Attack on BTC/BCH, ETH/ETC and XMR/BCN.** We evaluate the profitability and feasibility of 3 pairs

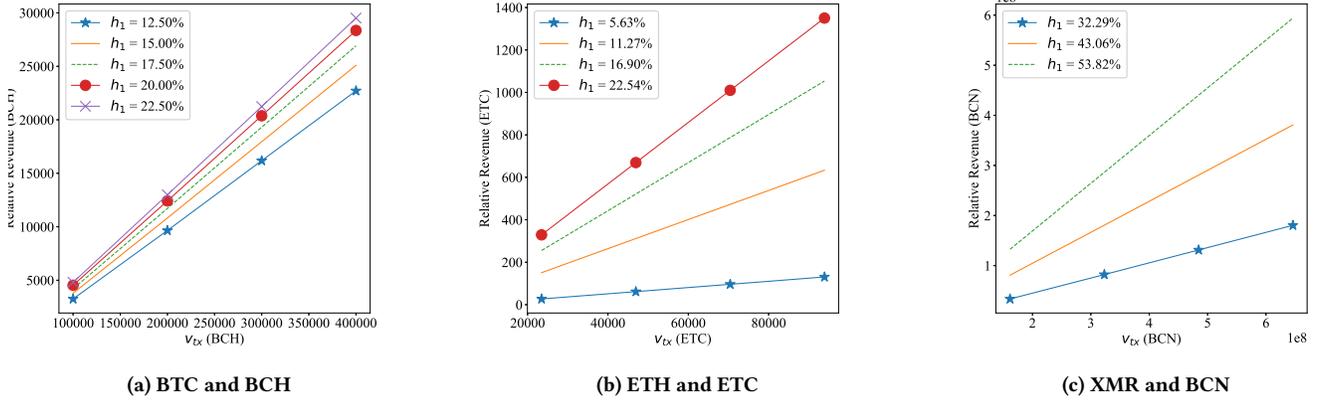


Figure 4: Testing real-world blockchains.

Table 5: Information of our selected blockchains

(a) BTC and BCH		
	BTC	BCH
Difficulty	6071846049920.0	199070336984
Price (USD)	3585.99	126.31
Algorithm	Sha256d	Sha256d
Hashrate(h/s)	39997.52E+15	1444.26E+15
Coins per Block	12.5	12.5

(b) ETH and ETC		
	ETH	ETC
Difficulty	1.91E+15	122025268093982
Price (USD)	118.53	4.26
Algorithm	Ethash	Ethash
Hashrate (h/s)	142.00E+12	8.62E+12
Coins per Block	2	4

(c) XMR and BCN		
	XMR	BCN
Difficulty	113361254717.0	40879087965
Price (USD)	43.64	0.000619
Algorithm	CryptoNight	CryptoNight
Hashrate (h/s)	9.29E+08	3.35E+08
Coins per Block	3.075	987.26

of top-ranked cryptocurrencies with the same hash algorithm: BTC/BCH, ETH/ETC, and XMR/BCN. Table 5 presents the blockchain data we use as the input of our model. The data was fetched on 05 March 2019 from <https://whattomine.com/> and <https://coinmarketcap.com>. By permuting the adversary mining power  $H_a$  and the transaction value  $v_{tx}$ , our experiments reveal their relationship with the relative revenue.

As shown in Figure 4, it is surprisingly easy and profitable for a miner of BTC/ETH to launch a 51% attack on BCH/ETC, but it is difficult and unprofitable for a XMR miner to attack BCN. In

detail, the requirement and the profitability of a 51% attack are summarized as follow:

- With approximately 12.5% mining power of BTC ( $5000E + 15h/s$ ), an adversary can gain 6% (150 BCH, or \$18,946.5 USD) extra profit (than honest mining) by double-spending a transaction of 3000 BCH (equivalent to \$378,930).
- With approximately 11.27% mining power of ETH ( $16E + 12h/s$ ), the adversary can gain 1.33% (600 ETC, or \$2,556 USD) extra profit by double-spending a transaction of 90000 ETC (equivalent to \$383,400).
- With approximately 43.05% mining power of XMR ( $4E+8h/s$ ), the adversary can gain 0.67% (1,000,000 BCN or \$619 USD) extra profit by double-spending a transaction of 600,000,000 BCN (equivalent to \$247,600).

The required mining power is not difficult to achieve. For example, the top three mining pools<sup>3</sup> in ETH are Ethermine (27.7%), Sparkpool (22.2%), f2pool2 (12.5%); and the top three mining pools in BTC are BTC.com (23.0%), AntPool (16.4%), and F2Pool (11.6%). In addition, the adversary may establish a powerful mining pool from scratch via the bribery attack [2].

Table 6<sup>4</sup> provides a summary on the mining power distribution of a selection of blockchains. In this table, the “portion” represents the ratio of a stronger blockchain over a weaker blockchain, where the stronger blockchain is the first row of each mining algorithm, and all other rows of the same mining algorithm are weaker chains. For a stronger chain, the “Top Miners” represents the percentage of mining power that the top mining pools control in the stronger chain. For weaker chains, the “Top Miners” show the ratio of mining power of a top miner over the total mining power of the weaker chain. For example, the top 1 mining pool in ETH controls 27.7% mining power, and this amount of mining power about 4.563 times of the total mining power in the entire ETC network.

However, an XMR miner does not profit much from the mining power migration attack. This is due to the fact that the total available mining power in Monero is only about 2.8 times of the mining

<sup>3</sup><https://www.etherchain.org/charts/topMiners>. Data collected on 19/Feb/2019.

<sup>4</sup>Data collected on 19/Feb/2019.

power in the BCN, although their market caps differ greatly. In comparison, the total available mining power in BTC is about 27.8 times of the total mining power in BCH; and the total available mining power in ETH is about 16.4 times of the total mining power in ETC.

**5.1.2 Cloud mining attack on top 15 PoW blockchains.** We evaluate the *cloud mining attack resistance* of 15 selected PoW blockchains from Top 100 blockchains by their market cap (Table 7). We fetched the blockchain data from Coinmarketcap [12] on 19 February 2019, and the renting price of mining power from NiceHash on 07 April 2019. We set  $v_{tx} = \$500,000$ , and  $h_2 = 2$ . That is, we assume the rentable mining power is twice of the honest mining power ( $h_2 = 2$ ) in these chains, and the double-spending transaction amount is \$500,000.  $nConfirm$  is the value recommended by the according community. We reveal the expected revenue based on these data, and the result is summarized in Table 7. The result shows that, unfortunately, all selected blockchains are vulnerable towards the *cloud mining attack*, for example:

- the attacker needs approximately \$2,000 to launch *cloud mining attack* on ETC for an hour, and the relative revenue will be \$33899 if successful;
- the attacker needs approximately \$2,600 to launch the attack on BCH for an hour, and the relative revenue will be \$117198 if successful;
- the attack needs approximately \$730 to launch the attack on Electroneum for one hour, and the relative revenue will be \$6222 if successful.

## 5.2 Investigating the 51% attack on ETC at Jan. 2019

Ethereum Classic (ETC) is a PoW-based blockchain distributed computing platform, and ETC uses the same hash algorithm with Ethereum (ETH). As shown in the previous section, ETC and ETH are vulnerable to our migration attacks, and ETC is also vulnerable to the cloud mining attack. In fact, a 51% attack happened to ETC on 7 Jan, 2019. NiceHash, a cloud mining service, is a suspected mining power source, though the actual source is still a mystery.

In this section, we apply our SPA-MDP to investigate the 51% attack on ETC. We estimate the net revenue of the attacker, describe the attack strategy adopted by the attacker, and compare the revenue between the mining power renting and the mining power migration.

**5.2.1 The attack details.** At the beginning of 2019, a 51% attack on ETC resulted in the loss of more than 1.1 million dollars [8]. The attack started from 0:40am UTC, Jan. 7th, 2019 and ended at 4:20am UTC, Jan. 7th, 2019 approximately, lasted 4 hours. The attacker launched a coin withdrawal transaction on the Gate.io exchange [13], then launched double-spending attacks multiple times. As a consequence, 12 transactions out of all attempts were successfully double-spent, listed in Table 8.

The source of the mining power for this attack remains uncertain due to the anonymity of miners. However, the NiceHash cloud mining platform is highly suspected [3, 4]. One day before the attack, an anonymous person rents all available Ethash (the hash algorithm used by ETH and ETC) mining power from NiceHash.

**5.2.2 Analysing the attacking strategy.** According to the actual attack happened, the attacker continuously increased the value of new transactions throughout the attack (except the last double spending of the first account). It is suspected that this behavior belongs to the strategies used by the attacker to maximize and stabilize his revenue, with the following reasons.

First, launching multiple small double-spending attempts can stabilize the expected revenue. The double-spending attack may fail in a limited time period, even if the adversary controls more than 50% of the computing power. Compared to a one-off attempt, the revenue will be more stable if dividing a transaction to multiple smaller transactions.

Second, this strategy may be used for avoiding the risk management system of the cryptocurrency exchanges. Most cryptocurrency exchanges run their own risk management system to combat the misbehaviors, like the fraudulent payments and the abnormal login attempts. A huge coin withdrawal transaction is highly possible to trigger the risk management system, while multiple small transactions would be overlooked. Meanwhile, a big transaction may lead to longer confirmation time, and a longer attack period is easier to be detected. Therefore, defeating the risk control system is naturally a part of the attacker’s strategy. According to the Gate.io report [8], the risk management system ignored transactions from the attacker, as the attack was decently prepared - they registered and real-name authenticated the account on Gate.io more than 3 months before the attack. Slowly increasing the transaction value is also highly suspected as an approach for reverse-engineering the threshold of invoking the risk management system.

In addition, we investigate the waiting time between each two attacks (quantified by using the number of blocks). The waiting time varies mostly from 67 blocks to 409 blocks. Interestingly, there are two much bigger gaps of more than 5000 blocks before the transactions 0xbba16320ec and 0xd592258715. The first gap is after the first attack, and the second gap is before the attacker changed to another account to send double-spending transactions. The first gap may be because the attacker was cautious when first launching the double-spending attack. The attacker launched a double-spending transaction of only 600 ETC coins, which is much smaller than his following transactions. After the first attack, the attacker waited for a long time to confirm that the attack is successful, then he started to increase the transaction value. The second gap may be because the attacker ran out of money in his first account 0x3ccc8f7415, and managed to change to another account 0x07ebd5b216. The last transaction 0xd592258715 sent by 0x3ccc8f7415 is right before the second gap. It’s value is 5000 ETC coins, which is much smaller than its previous transaction of 24500 ETC coins. After the transaction 0xd592258715, the attacker changed to his another account 0x07ebd5b216, which caused the time gap of 5480 blocks.

**5.2.3 Estimating the revenue of the attacker.** We use our model to estimate the revenue of the attacker. To analyse this attack, we collect relevant data of this attack for the time period of the attack on 07/01/2019, from coinmarketcap [12].

With Gate.io, the required number  $nConfirm$  of block confirmation is 12, which is also recommended by of ETH community and ETC community [15]. The price of ETC on that day was \$5.32, and the price of BTC was \$4061.47. The mining difficulty of ETC was

**Table 6: Summary of the blockchains sharing the same hash algorithm.**

Type	Mining Algorithm	Coin	Rank	Hashrate (h/s)	Portion	Top Miners		
						#1	#2	#3
ASIC-resistant	Ethash	Ethereum (ETH)	3	1.42E+14	N/A	27.7%	22.2%	12.5%
		EthereumClassic (ETC)	18	8.62E+12	1647.4%	456.3%	365.7%	205.9%
	CryptoNight	Monero (XMR)	14	9.29E+08	N/A	37%	26%	12%
		ByteCoin (BCN)	39	3.35E+08	277.3%	102.6%	72.1%	33.3%
	Equihash	Zcash (ZEC)	20	3.36E+09	N/A	33.4%	19.2%	17.8%
		BitcoinGold (BTG)	26	3.17E+06	111111.1%	37111.1%	21333.3%	19777.8%
		Komodo (KMD)	55	4.48E+07	7518.8%	2511.3%	1443.6%	1338.3%
Aion (AION)		84	7.22E+05	1000000.0%	334000.0%	192000.0%	178000.0%	
ASIC-friendly	Sha256d	Bitcoin (BTC)	1	4.00E+19	N/A	23%	16.4%	11.6%
		BitcoinCash (BCH)	4	1.44E+18	2777.8%	638.8%	455.6%	322.2%
	Scrypt	Dogecoin (DOGE)	23	3.76E+14	N/A	18.0%	16.0%	10.0%
		Litecoin (LTC)	8	2.77E+14	135.7%	24.4%	21.7%	13.6%
	X11	Dash (DASH)	15	2.32E+15	N/A	13.0%	11.0%	11.0%
		WaltonChain (WTC)	73	1.14E+15	203.5%	26.5%	22.4%	22.4%

**Table 7: Expected revenue of 15 selected PoW blockchains.**

	Rank	Rent(\$/h/s)	Coin Price(\$)	Hashrate	Relative Reward(\$)
Bitcoin	1	2E-18	3585.99	4E+19	190940
Ethereum	3	1.36E-13	118.53	142E+14	32699
BitcoinCash	4	2E-18	126.31	1.44E+18	117198
Litecoin	8	3.34E-14	30.84	2.77E+14	115829
Monero	14	9.13E-11	43.64	9.29E+8	51199
Dash	15	3.53E-16	71.79	2.32E+15	114930
EthereumClassic	18	1.36E-13	4.26	8.62E+12	33899
Zcash	20	1.38E-08	54.77	3.36E+9	114751
Dogecoin	23	3.34E-14	0.002132	3.76E+14	114558
BitcoinGold	26	1.38E-08	11.93	3170000	51272
Siacoin	46	3.74E-17	0.002389	1.88E+15	114731
Komodo	55	1.38E-08	0.640292	4.48E+7	788
Electroneum	67	9.13E-11	0.006184	4.4E+9	6222
Ravencoin	94	3.36E-13	0.011905	5.9E+12	114623
Zcoin	99	2.79E-12	4.83	9.69E+10	114727

131.80E+12, and the ratio  $h_2$  of attacker’s mining power over the honest mining network was about 1.16, i.e., the attacker approximately controls 53.7% mining power during the attack. The reward of successfully mining a block is 4 ETC coins, and the price of renting Nicehash mining power on that day is 3.8290 BTC/TH/day. As there is no data on the attacker’s connectivity w.r.t. propagating his blocks, and the impact of  $\gamma$  is relatively small (as previously discussed), we assume that  $\gamma = 0.3$ .

We permute and mark the transaction values used by the attacker. We also plot the same curve in the mining power migration scenario to compare the profitability of two mining power sources. The result is shown in Figure 5.

The result shows that when the transaction value is over 5000 ETC, double-spending is more profitable than by honest mining. Having a transaction (or a set of transactions) of value over 5000

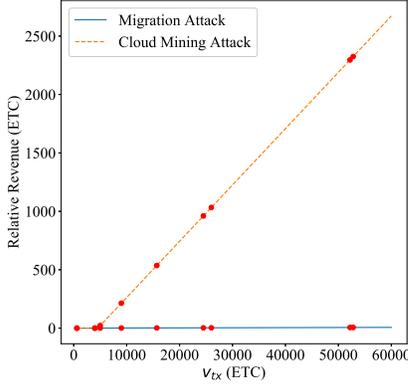
**Table 8: All double-spent transactions during the 51% attack on ETC [14]. In this attack, 12 transactions were double-spent from two accounts.**

Trans. ID (in short)	From	To	Amount (ETC)	Height	waiting time (#block)
0x1b47a700c0	0x3ccc8f7415	0xbbe1685921	600	7249357	-
<b>0xbba16320ec</b>	0x3ccc8f7415	0x2c9a81a120	4000	7254430	<b>5073</b>
0xb5e0748666	0x3ccc8f7415	0x882f944ece	5000	7254646	216
0xee31dff66	0x3ccc8f7415	0x882f944ece	9000	7255055	409
0xfe2da37fd9	0x3ccc8f7415	0x2c9a81a120	9000	7255212	157
0xa901fcf953	0x3ccc8f7415	0x2c9a81a120	15700	7255487	275
0xb9a30cee4f	0x3ccc8f7415	0x882f944ece	15700	7255554	67
0x9ae83e6fc4	0x3ccc8f7415	0x882f944ece	24500	7255669	115
0xaa50615e3	0x3ccc8f7415	0x53dfbb307	5000	7256012	343
<b>0xd592258715</b>	0x07ebd5b216	0xc4bcfee708	26000	7261492	<b>5480</b>
0x9a0e8275fc	0x07ebd5b216	0xc4bcfee708	52800	7261610	118
0x4db8884278	0x07ebd5b216	0xc4bcfee708	52200	7261684	74

Total: 219500 ETC

ETC (approximately 26,000 USD at the time of attack) should not be difficult for an attacker, so the incentive of launching double-spending attacks is very strong.

Moreover, our results give the estimated net revenue of the attacker: \$84773.40. It is approximate to \$100,000 - the value that attacker returned to Gate.io after the attack [16]. Summing all relative revenues of successful transactions, the total relative revenue of the attacker derived from our model is approximately 9000 ETC coins. Recall that the attacker controlled  $p = 53.7\%$  of ETC mining power. the probability  $P$  of a successful 51% attack is one minus the possibility of failing to attack. The failing scenario is that the adversary mines  $n < nConfirm$  blocks when the honest network has mined  $nConfirm$  blocks, where  $nConfirm$  is the number of



**Figure 5: Replay of the ETC Double-Spending Attack. The blue line denotes the performance of *Cloud Mining Attack*, we marked the amount of attack transactions on it in red. The orange line denotes the performance of *Mining Power Migration Attack* as the contrast experiment.**

confirmation blocks. Mining can be modeled as a binomial distribution model  $B(n + nConfirm, p)$ , where  $n + nConfirm$  blocks will be mined and the adversary mines the next block with the probability  $p$ . Therefore, given  $nConfirm = 12$ ,  $P$  is calculated as:

$$P = 1 - \sum_{i=0}^{12} C_{i+12-1}^i p^i (1-p)^{12} = 56.48\% \quad (7)$$

Our model produces the expected net revenue of a single attack, regardless whether it is successful or not. In our case, only successful attacks were observed, but the failed attacks also contribute to the theoretical expected net revenue. The successful attacks contribute to the expected net revenue of 9000 ETC coins, and their possibilities of success are 56.48%. Accordingly, the failed attacks contribute to the expected net revenue of  $\frac{9000}{56.48\%} (1 - 56.48\%) = 6934.85$  ETC coins. Therefore, the expected total net revenue is  $9000 + 6934.85 = 15934.85$  ETC coins, which is equivalent to \$84773.40 at the time of attack.

The expected net revenue based on our model is \$15226.6 less than the value returned by the attacker. This is because the success of an attack is probabilistic, and our model provides the mathematical expectation rather than the accurate value of the net revenue. According to the probability theory, when the number of attack attempts is small, the real value and the theoretical value will be biased.

Compared to the mining power migration, cloud mining is much more profitable. This means that for the ETH/ETC pair, renting mining power to attack ETC is much cheaper than migrating mining power from ETH. The reason may be the GPU friendliness of the ETH/ETC mining algorithm. ETH and ETC use Ethash [17] as the hash algorithm of PoW. Ethash is a memory-hard function, making it GPU-friendly while ASIC-resistant [18]. As GPU is not dedicated hardware, its mining power can be migrated to any blockchains. Therefore, renting mining power for ETH/ETH is much cheaper

compared to renting mining power with dedicated hardware such as ASICs.

## 6 DISCUSSION

This section discusses potential short term and long term methods to detect and prevent the two sucker punch attacks. We also show how to make use of our model to dynamically adjust some parameters to prevent potential sucker punch attacks.

### 6.1 Quick remedies

We first discuss several quick remedies for cryptocurrency exchanges to reduce the damage of 51% attacks. It consists of detecting potential attack attempts, and reacting upon detection through conventional risk management techniques.

**6.1.1 Detecting 51% attacks.** For the sucker punch attacks, the attacker needs to move a considerable amount of mining power from somewhere, such as the other blockchain or a cloud mining service.

This gives us an opportunity to detect the anomaly state where a “large” portion of mining power suddenly disappears. The threshold of “large” is blockchain specific according to the risk management rules. For example, a blockchain which cares less on such attacks can set the threshold to 100% of its current total mining power. That is, after moving this amount of mining power into this blockchain, the new comer will control 50% mining power in total. However, this will not detect an attacker who gains 90% mining power from cloud, and 10% from another source. A more cautious blockchain may set a tighter threshold, e.g. 5%, however, this may cause many false positive alarms.

There are two limitations of this method. First, it may introduce false positive detections, and it is hard to identify which blockchain will be the victim upon detection. Second, it is not cost effective, as it requires significant communication overhead to monitor all possible stronger blockchains and cloud mining services in real-time.

**6.1.2 Reactions.** Upon detection, a potential victim can react to manage potential risks. Several reactions can be taken to reduce the potential damage from the sucker punch attacks. The first reaction is to increase the number  $nConfirm$  of block confirmations. As shown in Figure 3d, in our experiment setting (Table 4), with the increase of required number of confirmations, the related revenue decreases. Second, decreasing the maximum amount of cash out in a single transaction. As shown in Figure 3b, the higher the transaction value is, the more relative revenue an attacker can gain. Thus, decreasing the maximum value of a transaction for cash out would discourage a rational miner to launch sucker punch attacks.

Our model can be used to provide recommendations on the above mentioned parameters. For example, Figure 6 shows the impact of the value  $v_{tx}$  of transactions and the number  $nConfirm$  of confirmations on the 51% on ETC. This analysis is produced by using our SPA-MDP model, and all other parameters are set up according to the attack happened. This shows that if the value of transactions was limited to 9,000 ETC (approximately \$38340.0) per transaction, then the attacker will not get any net revenue. On the other side, if the exchange wants to allow a maximum of 52,800

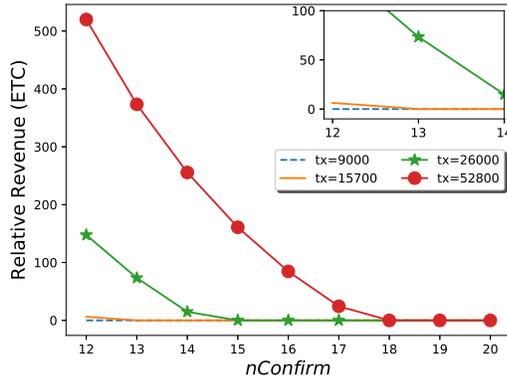


Figure 6: Impacts of  $v_{tx}$  and  $nConfirm$  on the ETC attack.

ETC (approximately \$224928.0), then our model recommends that the  $nConfirm$  should be increased to 18 to eliminate the incentive of a rational miner to launch such attacks.

In addition, decreasing the maximum frequency of cash out would also limit the potential damage from an attacker, as it reduces the daily withdraw limit.

Last, if a potential attack is considered very likely, then the potential victim can halt the cash out temporarily, to increase the cost of the attack.

## 6.2 Long term solutions

Though easy to deploy, aforementioned quick remedies are not sufficient. First, they sacrifice the usability of blockchains. Second, all of them only minimize the effect of the potential attacks, rather than eliminating them.

Improving the PoW protocol from the protocol-level is also a promising approach to combat our attacks. There are limited works aiming at minimizing the effects of powerful miners being malicious. For example, RepuCoin [19] aims at mitigating the 51% attacks in PoW protocols by introducing the “reputation”. In RepuCoin, the weight of each miner is decided by the reputation rather than the mining power. The reputation of a miner depends on the mining power, but also takes the past contribution of miners into consideration. In this way, “sucker punch” cannot raise the reputation in a short time period, and the “sucker punch”-style attacks become much harder.

## 7 RELATED WORK

Different models, such as models based on the Markov decision process and models based on game theory, have been proposed to analyse the PoW blockchains.

*PoW security model.* Refined properties [20, 21, 22], such as *Common prefix* and *chain quality*, have been proposed to evaluate proof-of-work (PoW) in Bitcoin with different network settings. Based on these refined properties, a framework [23] for quantitative analysis on the chain quality and resistance of attacks have been recently proposed, with a focus on four metrics including chain quality,

incentive compatibility, subversion gain, and censorship susceptibility.

*MDP-based PoW Modelling.* Markov decision process (MDP)-based model has been previously employed to analyze different aspects of blockchains, including Selfish Mining attacks on PoW-based blockchains [24, 25], and the consensus guarantee of various PoW-based blockchains [26].

*Game-theoretic analysis of PoW.* Utilising game-theoretic models to analyse PoW blockchains is another direction. *The Miner’s Dilemma* [27] describes miners’ incentive in attacking PoW blockchains through game theory. Carlsten et al. [28] introduced a game-theoretic model to analyze the instability of the block reward with only transaction fees. Kwon et al. [29] extended the “block withholding attack” to the “fork after withholding attack” and analyzed them with the Nash Equilibrium theory. Kwon et al. [30] show that a coin can intentionally weaken the security and decentralization level of the other rival coin when mining hardware is shared between them, allowing for automatic mining.

## 8 CONCLUSION

Honest majority is the most important assumption of PoW-based blockchains. However, this assumption does not always hold in practice. We designed our SPA-MDP model to present two possible cases that can break this assumption, including migrating mining power from one chain to the other, and renting cloud mining power.

Our evaluation provided an estimated amount of the cost and the net revenue of each attack. We showed that it is feasible to launch both attacks, and rational miners do have a reason to do so since they are purely profit driven. By using our model, we also provided an analysis on the strategy the attacker uses, in the recent 51% attacks on ETC in January 2019.

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