

# MineHunter: A Practical Cryptomining Traffic Detection Algorithm Based on Time Series Tracking

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### **BACKGROUND**

- Cryptomining is a process in which transactions for various forms of cryptocurrency are verified and added to the blockchain digital ledger.
- Cryptojacking, the unauthorized use of someone else's computer for cryptomining,
  has become a popular attack similar to ransomware since 2018.





# LIMITATIONS OF EXISTING METHODS

### • Malicious mining codes in the websites:

- ➤ Install a plug-in in the user's browser, which analyzes the JavaScript code in the website and the usage of the computing resources.
- > Require the cooperation of users and browser vendors and difficult to deploy on a large scale environment.

### Cryptojacking malware in the host:

- > Similar to the detection method of malware, mainly by deploying anti-virus software on the host.
- ➤ Only support the general computers and difficult to deploy effectively.

### Our solutions:

Instead of deploying at the hosts, **MineHunter** detects the cryptomining traffic at the entrance of enterprise or campus networks by traffic analyzing method.



# CHALLENGES

### Extremely unbalanced datasets.

➤ Data imbalance is the core challenge in the field of traffic anomaly detection. Machine learning algorithms usually require a relatively balanced dataset.

#### Uncontrollable number of alarms.

Traditional network traffic anomaly detection algorithms usually have the problem of high false positives and cannot guarantee the specific number of false positives.

#### Traffic confusion.

Common obfuscation techniques include adding proxy, load encryption, port replacement, and packet padding.

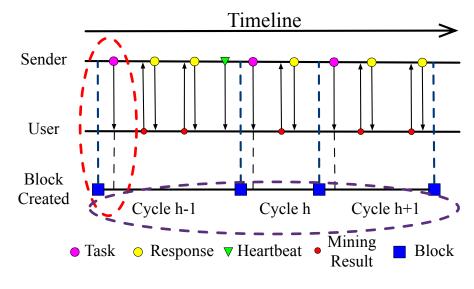
#### Online detection.

Due to the rapid growth of network bandwidth in the actual network environment, there are strict restrictions on the computational complexity of the detection algorithm.

### INTUITIVE IDEA

#### Two essential characteristics.

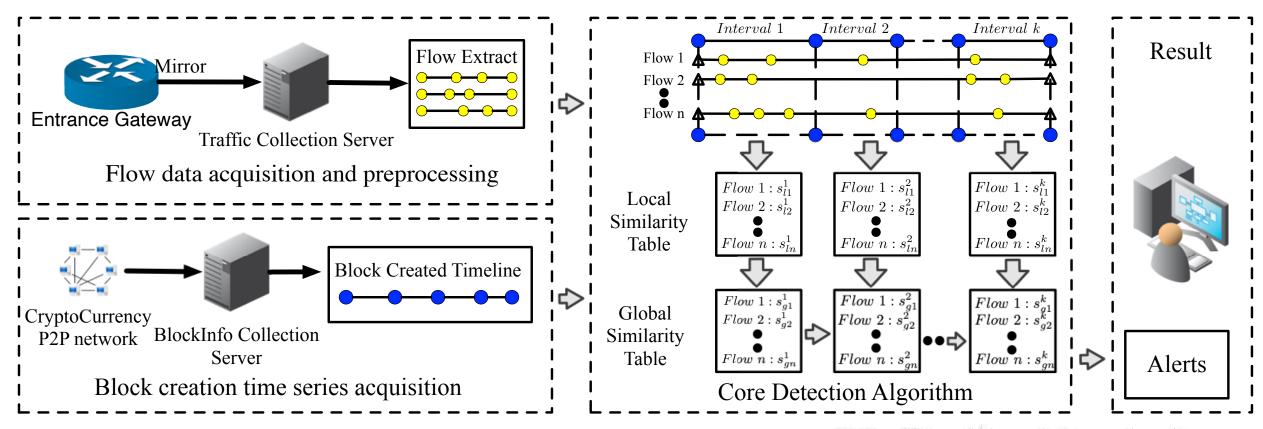
- ➤ One is that the time of task packet issued by a proxy or a mining pool is the same as the time when a new block is created.
- > The other is that cryptomining requires a long period of communication.





### **DETECTOR DESIGN**

#### Overview





# CRYPTOMINING TRAFFIC DETECTION ALGORITHM

### Cryptomining Traffic Detection Algorithm

### Problem & Target Formulation

Flow set:  $F = \{f_1, f_2, ..., f_n\}$ 

Time Series:  $f = \{p_1, p_2, ..., p_m\}$ 

Time Interval:  $[t_s, t_e]$ 

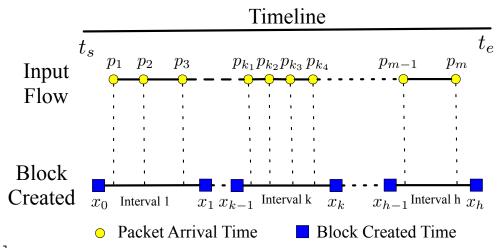
Target: for every f in F within  $[t_s, t_e]$ ,  $MH(f|[t_s, t_e]) = S$ ,  $S \in [0, 1]$ 

### Local Similarity Algorithm

#### **Naïve Algorithm**

Local interval distance: 
$$e(f^k) = \min_{x_{k-1} 
$$dis(p, x_k) = p - x_{k-1}$$$$

Local interval Similarity: 
$$s_l(f^k) = 1 - \frac{e(f^k)}{x_k - x_{k-1}}$$





# CRYPTOMINING TRAFFIC DETECTION ALGORITHM

### Two noisy scenarios

- high-frequency and large-scale data communications.
- > periodic heartbeat signals for a long time.

#### Solutions:

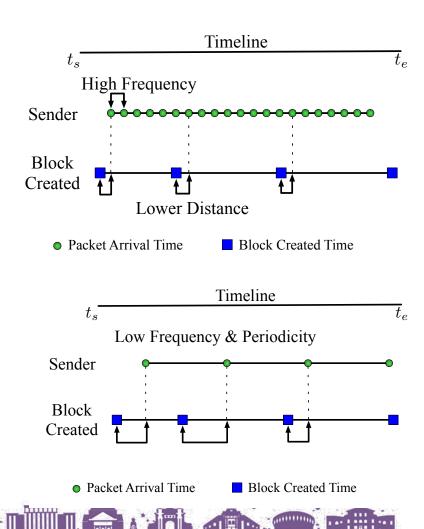
Local similarity algorithm based on credible probability estimation

$$s_l(f^k) = \alpha * (1 - \frac{e(f^k)}{x_k - x_{k-1}})$$

Random Sequence:  $m_k$  packets,  $n_k$  interval length,  $e_k$  interval distance

$$P(e = e_k) = \left(\frac{n_k - e_k}{n_k}\right)^{m_k} - \left(\frac{n_k - e_k - 1}{n_k}\right)^{m_k}$$

$$\alpha = P(e > e(f^k))$$



# CRYPTOMINING TRAFFIC DETECTION ALGORITHM

### • An exemple of $\alpha$

> Red: Cryptomining flow

> Green: high-frequency noise

> Yellow: low-frequency periodic noise

### Global Similarity Table (GST)

- > Iterative algorithm
  - \* addition increment
  - subtraction decrement

# Packets Distance	1	2	5	10	60	120
0	0.992	0.983	0.959	0.920	0.605	0.366
1	0.984	0.967	0.919	0.846	0.365	0.133
2	0.976	0.951	0.881	0.777	0.219	0.048
3	0.968	0.935	0.844	0.713	0.131	0.017
4	0.960	0.919	0.808	0.654	0.078	0.006
5	0.952	0.903	0.773	0.599	0.046	0.002
10	0.912	0.826	0.618	0.382	0.003	0.001
15	0.872	0.751	0.488	0.239	0.001	0.001
20	0.832	0.681	0.382	0.147	0.001	0.001



### EVALUATION

### Background Traffic

Duration	Active host	Total	Total	
time	number	packet	throughput	
		number		
Oct 23,	4096	30 billion	28 TeraByte	
2020-Nov				
23, 2020				
Maximum	Maximum	Average	Average	
packets per	bits per	flow num-	packet	
second	second	ber per	numbe per	
		day	day	
280533 pps	1.3 Gbit/s	4.7 million	0.9 billion	

#### Ethical Considerations

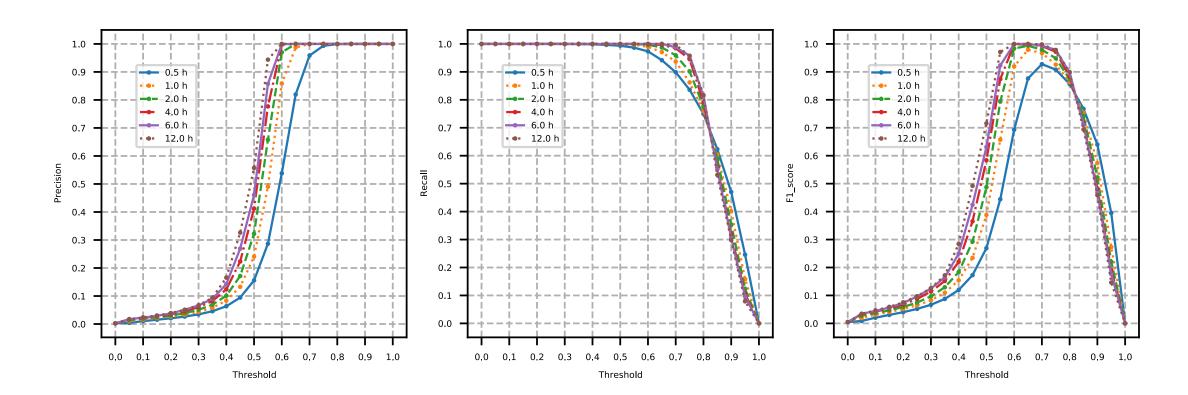
- ➤ IP addresses anonymized, Payload removed.
- Accordance with the policies defined by our institution.

### CryptoMining Traffic

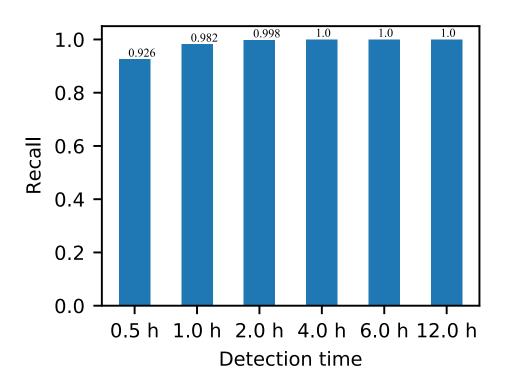
- ➤ 21 Monero mining pool nodes
- > cover nearly 80% computing power
- > all through TLS protocol
- > duration time same as background traffic
- Merge traffic by mergecap
- > Replay the traffic for detection



- Challenge 1: Extremely unbalanced data
- Detection case number: 21 \* 48 \*32 = 30000 cases for ti=0.5 h
- Evaluation results of Minehunter (2h-0.6, precision 97%, recall 99.7%)

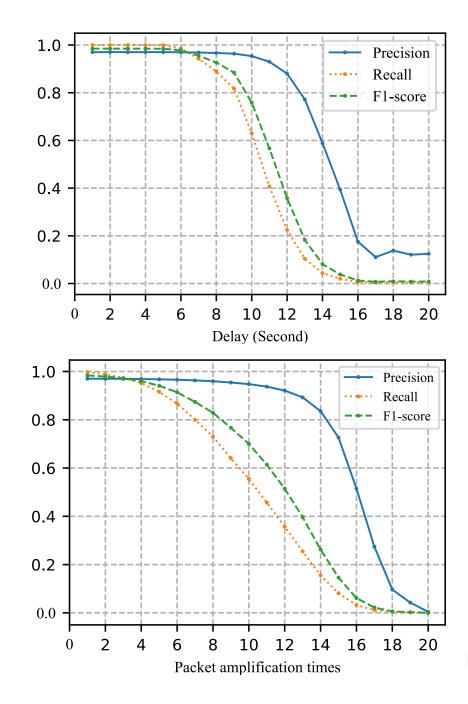


- Challenge 2: Uncontrollable number of alarms
- Alert Condition: Check from the head of the table, and stop checking if a false alarm is found.
- When the detection time is set to 2h, the algorithm's recall can reach 99.8%.

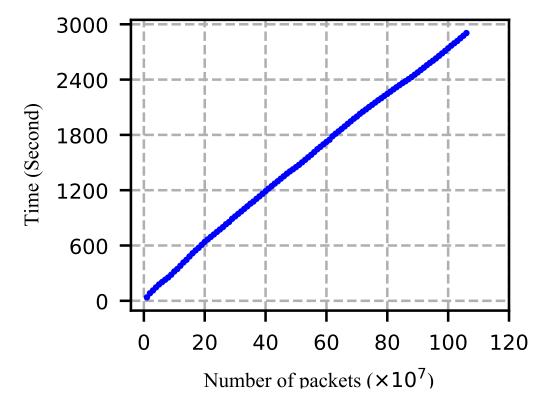




- Challenge 3: Traffic confusion
- Common method: proxy, load encryption, port replacement, and packet padding
- "White Box":
  - Packet Delay:
    - ❖ When the delay time is less than 10s, the overall performance of the algorithm is less affected.
  - ➤ Packet amplification:
    - ❖ The algorithm can effectively combat packet amplification by 10 times.

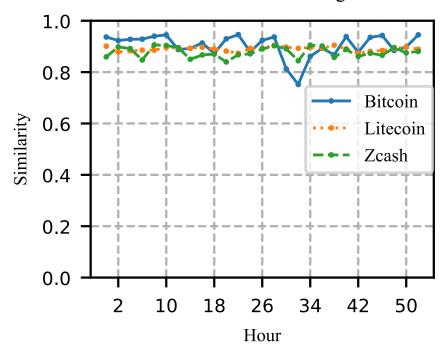


- Challenge 4: Online detection
- Average Speed: 350,000 pps



### Scalability

- ➤ Different cryptocurrencies
- ➤ Websites with embedded mining code



Mining Service	Cryptocurrency	Protocol	Proxy IP	Similarity
CryptoLoot[6]	Uplexa	TLSv1.2	45.79.218.212	0.80
Crypto Webminer[7]	Sumokoin	TLSv1.2	185.163.119.151	0.78
Monerominer.rock[22]	Masari	TLSv1.2	157.230.173.68	0.93

# CONCLUSION

- In this work, we propose **MineHunter**, a practical cryptomining traffic detection algorithm, which can be deployed at the entrance of enterprise or campus networks.
- Our algorithm has attempted to solve the four core challenges faced in the actual network environment,
  including extremely unbalanced datasets, controllable alarms, traffic confusion, and efficiency.
- We conduct a large-scale evaluation experiment in a campus network environment within one month. The experimental results show that our algorithm can achieve 97.0% precision and 99.7% recall on the extremely unbalanced dataset.





### THANKS FOR LISTENING



Public codes and datasets: https://github.com/zsz147/MineHunter For more information, please contact me.

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