

Combatting Crypto Scams on the Web with Semantic Graph Analysis

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Outline

- 1. Motivations
- 2. Recent Crypto Scams
- 3. Related Work
- 4. Integrative Blockchain Provenance Analyzer (IBPA)
 - a. Design and Operation
 - b. Results and Evaluation
- 5. Future Work

Motivation: Fix Rampant Misinformation on the Web about Cryptocurrencies

- Blockchain space is highly fragmented
- The space is especially highly susceptible to fraudulent activities
 - Deceptive smart-contracts
 - DeFi scams, & "pump and dumps"
 - "A-List" / "Airdrop" Phishing Scams
- Anonymity of many blockchain tokens act as a vessel for widespread phishing scams that so far are effectively untraceable
- No technologies are widely available to identify crypto addresses involved in a phishing scam as "fraudulent"





High-Profile

Crypto "Phishing" Scams





Tweet



I am giving back to the community.

All Bitcoin sent to the address below will be sent back doubled! If you send \$1,000, I will send back \$2,000. Only doing this for 30 minutes.

bc1qxy2kgdygjrsqtzq2n0yrf2493

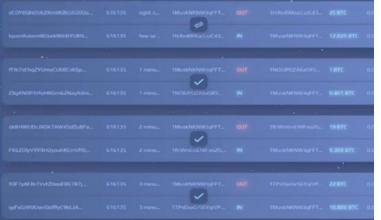
Enjoy!

4:22 PM · 15 Jul 20 · Twitter Web App









More info on SPACEXDROP.NET



Use the QR code or this BTC address to join:

✓ 1MuskNKNWJqFFT5S4MEtiNEvw2DZFHcEcy

To participate you just need to send 0.1 BTC to 20 BTC to the contribution address and we will immediately send you back 0.2 BTC to 20 BTC to the address you sent it from.

You can participate only once.

If you send 0.1+ BTC, you will get 0.2+ BTC back
If you send 0.5+ BTC, you will get 1+ BTC back
If you send 1+ BTC, you will get 2+ BTC back
If you send 5+ BTC, you will get 10+ BTC back
If you send 10+ BTC, you will get 20+ BTC back
If you send 20 BTC, you will get 40 BTC back





WELCOME TO 5000 BTC GIWEAVAY





BACK BONUS SYSTEM

IF YOU SEND 0.1+ BTC, YOU WILL BE AIRDROPPED 0.2+ BTC BACK IF YOU SEND 0.5+BTC, YOU WILL BE AIRDROPPED 1+ BTC BACK+10% BONUS IF YOU SEND 1+ BTC, YOU WILL BE AIRDROPPED 2+ BTC BACK +25% BONUS IF YOU SEND 5+ BTC. YOU WILL BE AIRDROPPED 10+ BTC BACK +40% BONUS IF YOU SEND 10+BTC, YOU WILL BE AIRDROPPED 20+BTC BACK +60% BONUS

5000 BTC GIVEAWAY

TO PARTICIPATE YOU JUST NEED TO SEND BEETWEEN 0.1+ BTC TO 10+ BTC TO THE CONTRIBUTION ADDRESS WE WILL IMMEDIATELY SEND YOU BACK BEETWEEN 0.2+ BTC TO 20+ BTC TO THE ADDRESS YOU SEND IT FROM. EVERY PERSON CAN PARTISIPATE ONLY ONE TIME



1xGdMFTpCkQJWiuZSufa8MKrsrrgbPyxQ

Just Log into your mobile APP; Scan QR Code! Send amount of 0.1+ BTC to 20+ BTC to participate. REMEMBER! You can participate only once!

MORE INFO: MUSKXEVENT.SITE



#NASA #SpaceX #Live

Falcon 9 and Crew Dragon launch | Starlink Mission | Elon Musk | SpaceX Conference | Live















More info on WOZBTO



Use the QR code or this BTC address to join:

✓ 1WoZp2r6NMXbqKk7nYm6WUpWPd ***

To participate you just need to send 0.1 BTC to 20 BTC to the contribution address and we will immediately send you back 0.2 BTC to 40 BTC to the address you sent it from.

You can participate only once.

If you send 0.1+ BTC, you will get 0.2+ BTC back
If you send 0.5+ BTC, you will get 1+ BTC back
If you send 1+ BTC, you will get 2+ BTC back
If you send 5+ BTC, you will get 10+ BTC back
If you send 10+ BTC, you will get 20+ BTC back
If you send 20 BTC, you will get 40 BTC back









B Use the QR code or this BTC address to join:

1WoznScUTGbqGTqqeLPBgAVPgipx64URX

To participate you just need to send 0.1 BTC to 20 BTC to the contribution adress and we will immediately send you back 0.2 BTC to 20 BTC to the adress you sent it from

More info on WozDrop.me

If you send 0.1+ BTC, you will get 0.2+ BTC back

If you send 0.5+ BTC, you will get 1+ BTC back

If you send 1+ BTC, you will get 20+ BTC back

If you send 5+ BTC, you will get 20+ BTC back

If you send 10+ BTC, you will get 20+ BTC back

Can we figure out a way to flag these scam addresses (and addresses likely associated with them) before they offload their wallets at an exchange?



Selected Related Work



Phillips & Wilder: Clustering Replicated Phishing Sites

- Searched for similarities between operators of various, seemingly disconnected scam operations
- Discovered many operations were in fact operated by the same campaigns via clustering
- Stressed need for exchanges to be more vigilant

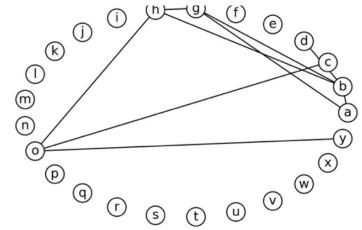




Fig. 5. Blockchain cluster overlap between different campaigns

Bartoletti et al.: Data Mining for Detecting BTC Ponzis

- Fraud detection to cybercrime analysis in BTC is an unexplored field
- Translate data mining techniques from credit card operations to Bitcoin Ponzi Schemes
- Created public dataset for systematic evaluations + comparisons of scams
- Automatic analysis comparing various ML approaches (supervised)
- Unclear how to operate for identifying addresses in real-time



IBPA:

Integrative Blockchain Provenance Analyzer



Goal: Derive Meaning from Raw Ledger Data

Latest Transactions

Hash	Time	Amount (BTC)	Amount (USD)
aad1cf51f0c90f8083ab04	01:44	0.01126765 BTC	\$150.21
b1eba0ee41d83254317a6	01:44	0.20907099 BTC	\$2,787.13
4a84b753f8cc22190718b	01:44	2.23379432 BTC	\$29,778.78
1903d996c52c3d164fbb4	01:44	0.04560389 BTC	\$607.95
c206f9c030f7ceb33c070	01:44	0.06598789 BTC	\$879.69
88092a2709986ffcee42c	01:44	0.00761768 BTC	\$101.55

https://www.blockchain.com/explorer

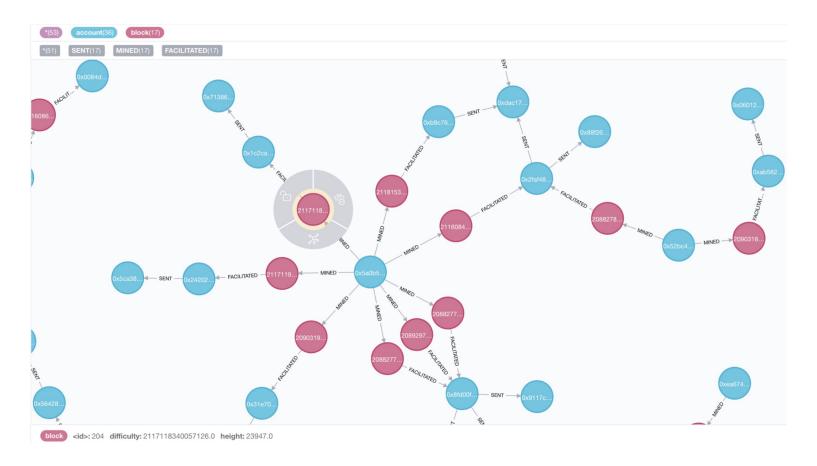
Tx	0x39fc91ef7d1e	From 0xee576686b08ffbaaf8	0 Eth
IX	22 secs ago	To 0x96f9632b25f874769	
_	0x6d2e354ab3e	From 0xc4fffd98fb3caec63cc	0.003 Eth
22 secs ago	To 0x983cdcb1fbb14692e		
Tx	0x9859f26bd57	From 0xc63fc6907d3f9657a	0.059 Eth
IX	22 secs ago	To 0x1457b31c41ae3a13d	
T	0x5dbad46205c	From 0x098306de8c7701999	0 Eth
Tx	22 secs ago	To 0x5fda1bb20918e5797	
	0xce76b73caec	From 0x9983cfd319085687b	0 Eth

Latest Transactions

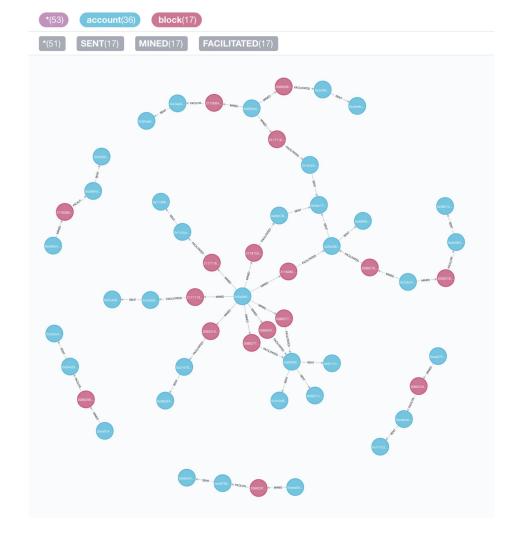
https://etherscan.io



Can We Use Provenance to Identify Suspicious Addresses?



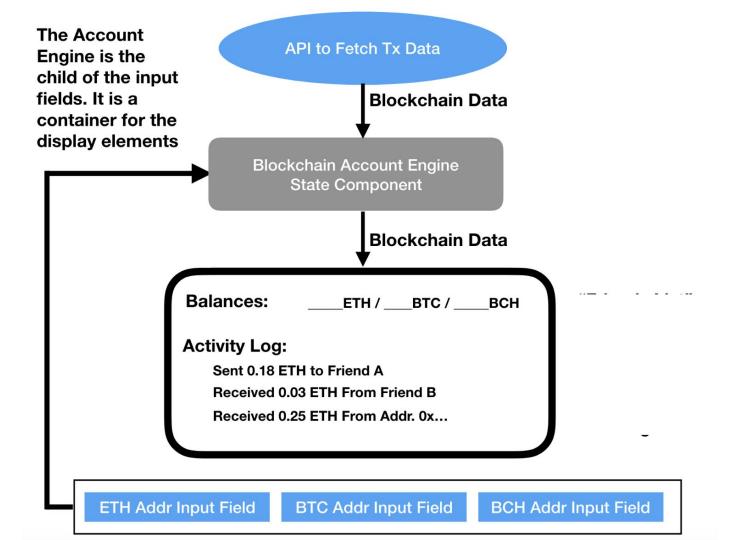
Answer: Yes we can

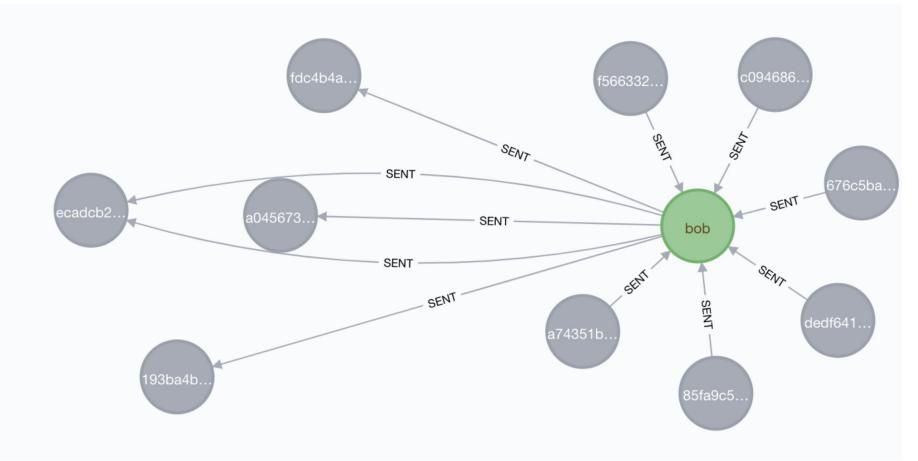


Design of the IBPA: Extract Transform and Load

- Extracts raw tx data from a given blockchain network via API calls, initiated by the user on the frontend
- *Transforms* that data, filtering the metrics / keys that matter (what were the final input and output addresses + amounts to the transaction)
- Loads that data into a Neo4j database for further querying by the IBPA
- Analysis: Runs tests using CypherQL+JS to return a predicted answer



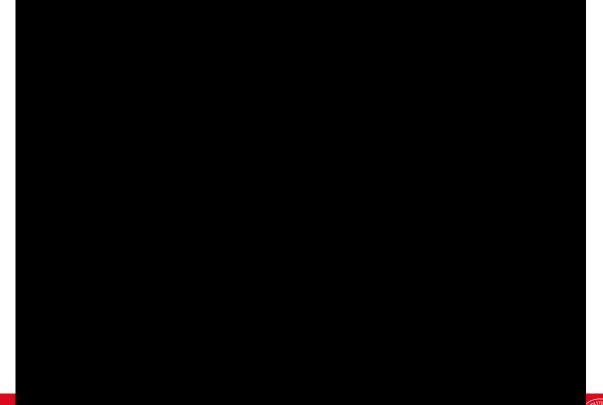




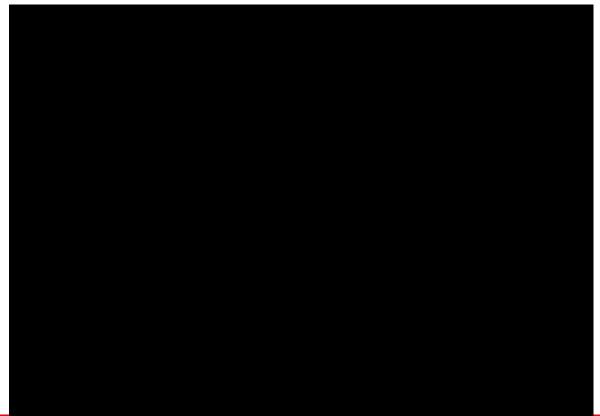




IBPA Operation: Scam Addresses



IBPA Operation: Control Addresses





IBPA Results:

Avg	Median	Incoming	Outgoing	Incoming	Outgoing	IBPA Score
Tx/Hr	Tokens	Nodes	Nodes	Rels	Rels	& Classification
0.015	0.0133	7	6	7	6	1 (regular)
0.015	0.6035	5	5	5	5	1 (regular)
< 0.01	0.0003	9	0	9	0	1 (regular)
< 0.01	0.7922	1	1	1	1	0 (regular)
< 0.01	0	0	5	5	5	1 (regular)
0.07	0.01	7	2	13	11	5 (scam)
3.00	0.01	7	3	6	5	9 (scam)
8.73	0.01	11	1	11	5	11 (scam)
0.41	0.005	6	1	6	1	7 (scam)
4.00	0.022	6	3	11	24	3 (scam)

Table 1: IBPA results for regular (non-scam) and scam accounts. Any score >= 3 results in a failing score and the corresponding address is flagged as a suspicious address meeting fraudulent criteria.



IBPA Evaluation

- Does an effective job at quickly calculating a reasonable suspicion flag for a selected address
- Creates an easy-to-query graph representation of the surrounding nodes of interest
- Can be easily modified to quickly evaluate a large cluster of addresses
- Is a necessary first step in logically tackling identification of fraudulent actors



Future Work



Expanding the System

- Comprehensive data extraction pipeline
 - Scour the web intelligently for similar scams to plug into the evaluator
 - Combine with other semantically annotated data on the Web for enhanced provenance analysis
- Building out for additional network use-cases
 - Works for standard L1's, but what about L2 with state channels?
 - What about other privacy-enabled L1's?
- Large-scale testing
- Customize for exchange use / Metamask integration?

Summary

- There are many cryptocurrency scams on the Web
- We need tools and techniques to identify whether the addresses advertised are scam addresses or not
- We utilized network analysis techniques to identify scam addresses with good accuracy
- We have an initial encoding of the transaction graphs represented primarily using the provenance ontology
- We are expanding this work to include other information available on the "Semantic Web"

Questions/Comments?

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