

A Sea of Coins: The Cryptocurrencies Proliferation in UniswapV2

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Main
ConceptsToken
Creation

Dataset

Returns

Positive
Profits for:1. Token
Security

2. Sandwich

3. Physical
time

Conclusions

- The rise of DEXs like Uniswap has enabled the creation and trading of a huge number of new cryptocurrencies, revolutionizing how assets are exchanged globally.
- We analyze financial properties (e.g., returns, liquidity) and security challenges (e.g., smart contract risks, transaction vulnerabilities) to understand the dynamics of these tokens.

A blockchain is a distributed ledger maintained by a network of computers (nodes).

Key Properties: Decentralization, immutability, and transparency.

Each block holds a batch of transactions and a reference (hash) to the previous block.

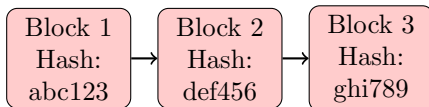


Figure 1: Blockchain

- **What is a Decentralized Exchange (DEX)?**

A peer-to-peer marketplace where crypto traders transact directly, eliminating the need for a central intermediary.

- **Smart Contracts – The Backbone of DEXs:**

Automated code programs that facilitate trades, handle funds, and ensure trustless transactions. These self-executing contracts operate based on predefined rules coded into them.

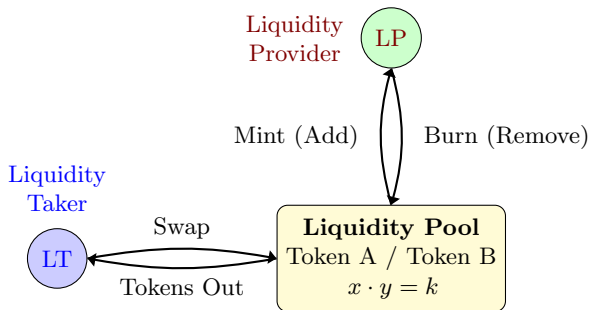


Automated Market Makers (AMMs): A type of DEX that uses a mathematical formula to price assets (**Constant Product Formula**);

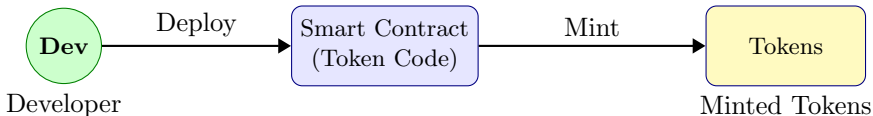
A mempool is a node's "waiting room" for unconfirmed transactions before they're added to a block.

Miners or validators typically pick transactions offering higher gas fees first (in many blockchains). During high activity, the mempool grows, increasing gas fees and wait times.

- **Automated Market Maker (AMM)** using the formula between the quantity of Token A x and of Token B y in the reserves: $x \cdot y = k$.
- **Pool Pair**: Smart contract holding two tokens (e.g., Token A, Token B).



- **Smart Contract Code:** Written in a language (e.g., Solidity).
- **Deploy to Blockchain:** The contract is published (deployed) on an EVM-compatible chain (e.g., Ethereum).
- **Minting:** Upon deployment or via a function call, the token supply is created (minted).
- **Token Standard (e.g., ERC-20):** Defines functions like `totalSupply()`, `balanceOf()`, `transfer()`, etc.
- **Ownership & Distribution:** Deployer or specified addresses receive the initial tokens.



```

contract DOPE is Context, IERC20, Ownable {
    mapping (address => uint256) private _balances;
    address payable private _taxWallet;
    uint8 private constant _decimals = 9;
    uint256 private constant _tTotal = 10000000000 * 10**_decimals;
    string private constant _name = unicode'Decentralization obligatory, practicality
    ↪ essential';
    string private constant _symbol = unicode'DOPE';
    // ... other variables ...
    constructor () {
        _taxWallet = payable(_msgSender());
        _balances[_msgSender()] = _tTotal;
        emit Transfer(address(0), _msgSender(), _tTotal);
    }
    function symbol() public pure returns (string memory) { return _symbol;}
    function totalSupply() public pure override returns (uint256) { return _tTotal;}
    // ... other functions ...
    function _transfer(address from, address to, uint256 amount) private {
        // Swap Rule function
    }
}

```


Retrieve information on:

- Newly created pairs (pair contract address, token0, token1) \leftrightarrow Token creation
- Swaps (amount swapped in/out) \rightarrow Price behavior
- Mints and Burns (amount minted/burned, liquidity tokens) \rightarrow Liquidity behavior

Approach:

- ❶ **Connect to an Ethereum Node:** - Use a node or an RPC provider (e.g., Infura, Alchemy) via web3.
- ❷ **Data:** - Use 17194 new coins created between the 10th October 2024 and the 2nd of December 2024.

Number of New Tokens Created

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Main
Concepts

Token
Creation

Dataset

Returns

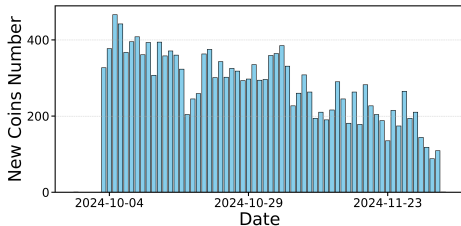
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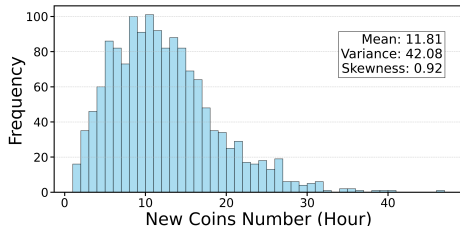
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3. Physical
time

Conclusions

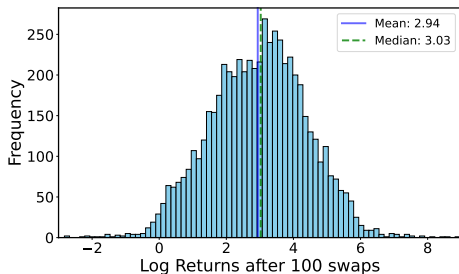


(a) Number of tokens created in a day for 2 months

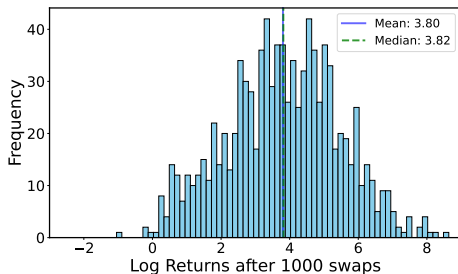


(b) Distribution of the new token number in one hour.

We focused on the cases in which one of the two pairs of the pool was WETH.



(a) Histogram of lagged log returns after 100 swaps.



(b) Histogram of lagged log returns after 1000 swaps.

Figure 3: Histograms of the lagged log returns of the new tokens after different number of swaps. We stress that we have a medium percent returns value $\langle \eta \rangle = 10^5\%$.

Profit distribution for a simple Buy & Hold strategy

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Main
Concepts

Token
Creation

Dataset
Returns

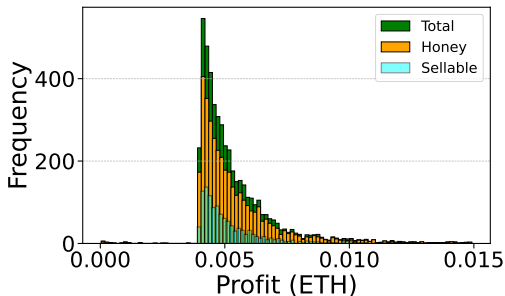
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We do not have considered:

- ① Are the purchased tokens secure?
- ② Presence of sandwich attacks
- ③ Swap time instead Physical time

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Token
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Dataset
Returns

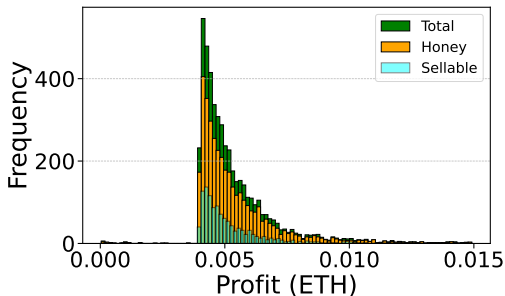
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Definition 1 (Honeypot)

A honeypot is a smart contract that pretends to leak its funds to an arbitrary user (victim), provided that the user sends additional funds to it. However, the funds provided by the user will be trapped and at most the honeypot creator (attacker) will be able to retrieve them.¹

Possible Scam Honeypot Elements:

- Ownership & Renouncement
- Liquidity Addition
- Minting & Burning
- Slippage Tax Modifiable
- Blacklist/Whitelist
- Token Supply
- Swapping Mechanism
- Taxes & Fees
- Liquidity Burned/Locked

¹Christof Ferreira Torres, Mathis Steichen, and Radu State. The art of the scam: Demystifying honeypots in ethereum smart contracts, USENIX Security 19, pages 1591-1607.

Motivation: Not all tokens remain sellable throughout their lifecycle. Some evolve into honeypot contracts, making them effectively unsellable.

Our Approach:

- **Time-Based Analysis:** Observe each token over its entire history.
- **Security Detector Queries:** Use external security checks from:
 - honeypot.is — Honeypot detection service;
 - GoPlus Labs — Token security scoring;
- **Final Categorization:** At the end of its observed lifetime, each token is classified as:
 - **Sellable:** No honeypot behavior detected ~ 2080 tokens $\simeq 12\%$ of the total;
 - **Honeypot:** Displays locked or blocked selling ~ 15114 tokens $\simeq 88\%$ of the total.

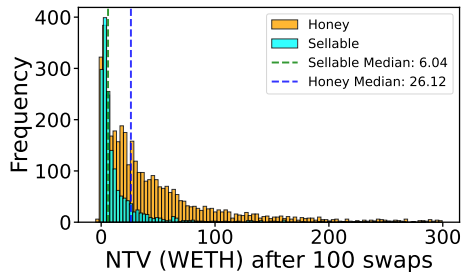
The Net Traded Values (NTV) of the token i is given by:

$$\text{NTV}_t^i = \left(\sum_{t=1}^{N_t} v_t^i \right) \cdot p^i(t = N_t) \quad (1)$$

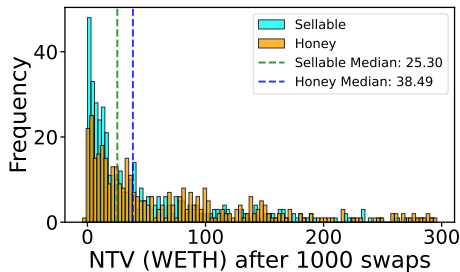
where:

- v_t^i represents the signed token amount exchanged in the t -th swap for the i -th token;
- N_t is the number of swaps considered;
- $p^i(t = N_t)$ is the token price recorded at the N_t -th swap for the i -th token.

We are interested in the amount of money circulating around these new tokens.



(a) Histogram of NTV after 100 swaps.



(b) Histogram of NTV after 1000 swaps.

Figure 4: Histograms of the NTV of the new tokens after different number of swaps.

We define the total new coins market value at fixed swap number N_t as the sum of all the Net Traded Values (NTVs) of the new coins at that time, namely

$$\text{NTV}_{\text{tot}}(N_t) = \sum_i \text{NTV}^i(N_t). \quad (2)$$

N_t	honeypot NTV_{tot} (ETH)	sellable NTV_{tot} (ETH)
10	16857.19 (94%)	1059.91 (6%)
100	170150.11 (86%)	27450.66 (14%)
1000	25485.46 (55%)	20867.45 (45%)

Table 1: NTV values of honeypots and sellable tokens at different swap thresholds.

NTV is also an indicator of a sudden lack of liquidity in the market that is artificially triggered.

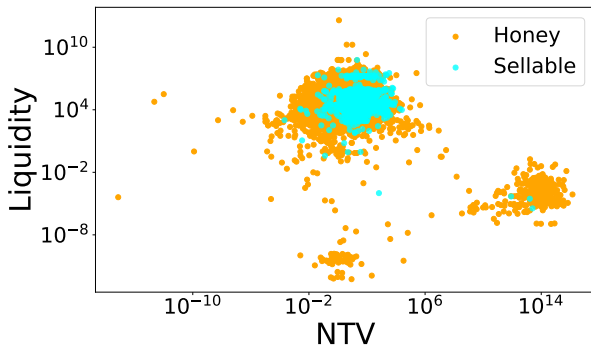
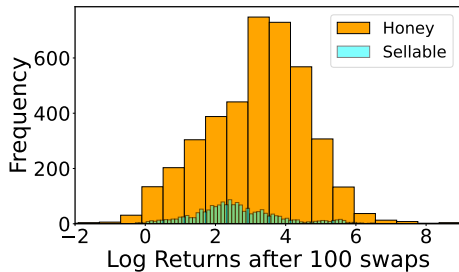
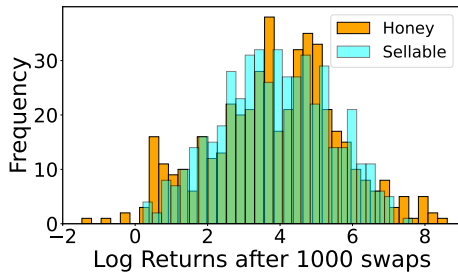


Figure 5: Scatter plot between the maximum value of NTV in the lifetime of all new tokens and the corresponding value of Liquidity in that time.



(a) Histogram of lagged log returns after 100 swaps.



(b) Histogram of lagged log returns after 1000 swaps.

Figure 6: Histograms of the lagged log returns of new sellable and honey tokens.

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Main
Concepts

Token
Creation

Dataset
Returns

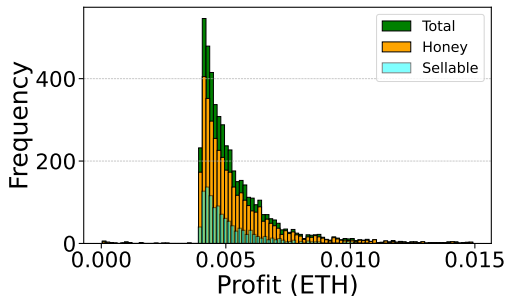
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time

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Sandwich Strategy Overview

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Concepts

Token
Creation

Dataset
Returns

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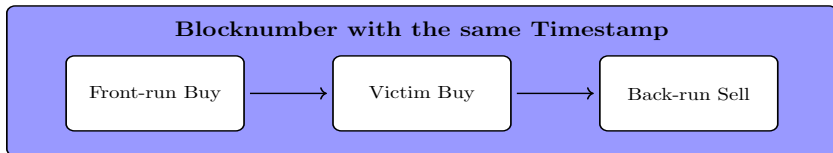
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Conclusions

High Profit for the presence of sandwich whose Tx's are deployed in the same blocknumber.

- **Mempool Monitoring:** Identify a pending swap transaction (buying token y for some ETH x).
- **Front-Run Attack:** Submit a transaction with higher gas to buy token y first.
- **Victim's Purchase:** The original, slower transaction then occurs, pushing the price up.
- **Back-Run Attack:** Finally, sell the token y at the new higher price, capturing profit.



With a initial reserve x of the ETH in the pair liquidity position:

- The first swap of the front-run attack sells Δx_a ETH;
- The swap of the slower trader sells Δx_ϵ ETH;
- The second swap of the back-run attack sells all of his tokens bought and obtains Δx_{tot} ETH.

The maximum of the gain $s = \Delta x_{tot} - \Delta x_a$, with a pair pool fee $r = 1 - f$ ($f = 0.3\% \ll 1$) and in the approx $f\Delta x_a \simeq O(\Delta x_\epsilon)$:

$$\Delta x_a^{\max} = \frac{\Delta x_\epsilon}{f} - x ; \quad (3)$$

$$s_{\max} = \frac{(\Delta x_\epsilon - fx)^2}{\Delta x_\epsilon} , \quad (4)$$

Optimal Sandwich - Example

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Main
ConceptsToken
CreationDataset
ReturnsPositive
Profits for:1. Token
Security

2. Sandwich

3. Physical
time

Conclusions

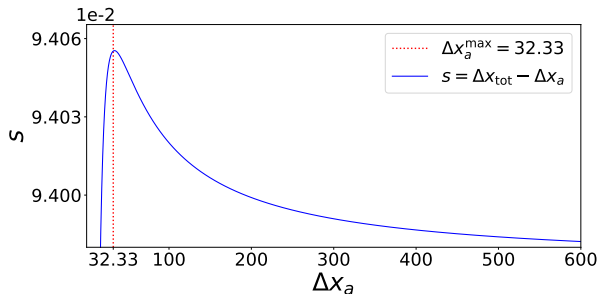


Figure 7: Visualization of the gain s as a function of the attacker input amount Δx_a . The plot is generated using the parameter values $x = 1$, $r = 0.997$, and $\Delta x_\epsilon = 0.1$. The red vertical line corresponds to the estimated maximum point.

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Main
Concepts

Token
Creation

Dataset
Returns

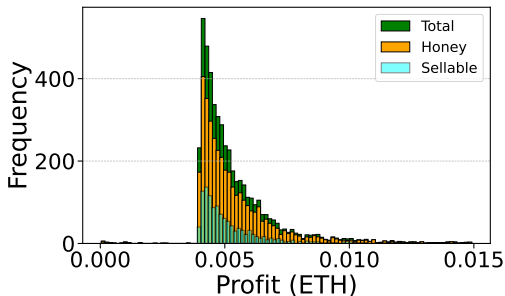
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Clustering in swap time

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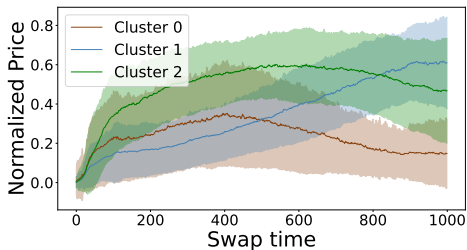
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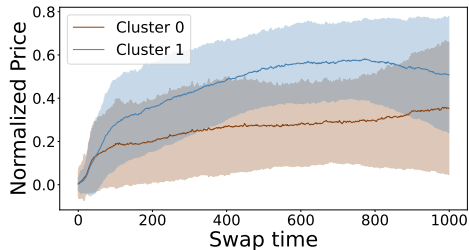
3. Physical
time

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The k-means algorithm partitions the trajectories into k distinct clusters by minimizing the within-cluster variance, thereby grouping similar trajectories based on their features.



(a) Clustering resulting trajectories for $k = 3$.



(b) Clustering resulting trajectories for $k = 2$.

Figure 8: The figures show the mean and the one standard deviation bands of the trajectories of length 10^3 resulting from the clustering, with $k = 2$ and $k = 3$.

Clustering in swap time of Honeypots and Sellable Tokens

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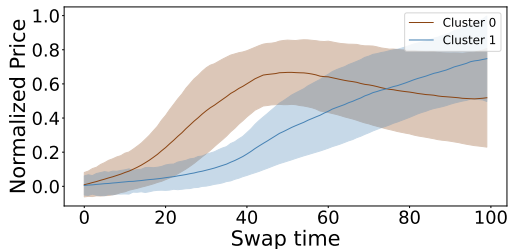
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ConceptsToken
CreationDataset
ReturnsPositive
Profits for:1. Token
Security

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time

Conclusions

In the Table, we show the distribution of the honeypots and sellable tokens.



Cluster	Honeypots	Sellable
0	24%	19%
1	46%	11%

Figure 9: Clustering resulting trajectories with $k = 2$ for 100 swaps.

We clearly observe a dominant presence of honeypots in cluster 1. This is expected, as honeypot tokens prevent selling, leading to a characteristic initial price behavior.

Now, we applied a dynamic time warping (DTW) method to categorize tokens into clusters based on their temporal price evolution.

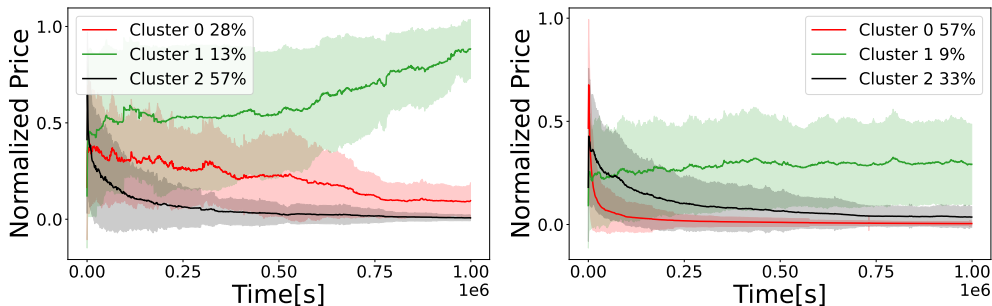


Figure 10: At left, the Honeypot Tokens; instead, at right, the Safe Tokens.

The initial pick is associated with the BOT presence.

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Main
ConceptsToken
CreationDataset
ReturnsPositive
Profits for:1. Token
Security

2. Sandwich

3. Physical
time

Conclusions

- The positive profits of new token returns suggest it may be convenient to trade in this currency.
- However, traders must consider:
 - The presence of honeypots.
 - The risk of sandwich attacks.
 - The necessity of a physical time strategy.
- The emergence of BOT activity significantly reduces potential profits, making careful strategic planning essential.

Optimal Sandwich: Example

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Δx	Δy	x_{new}	y_{new}	New Price	Old Price	Liquidity
25.0000	-268464000	28.3199	35758537	7.920e-07	1.091e-08	31823
0.0500	-62833	28.3699	35695704	7.948e-07	7.920e-07	31823
-25.0316	268463996	3.3382	304159700	1.098e-08	7.948e-07	31865

In this case, $\Delta x_a^{\max} = 10$ ETH and $s_{\max} = 0.032$ ETH, with an initial reserve $x = 3.32$ ETH.

Sandwich on the token AIDLE in the pair WETH/AIDLE executed in the block number 21560868.

