Detecting Flash Loan Based Attacks in Ethereum

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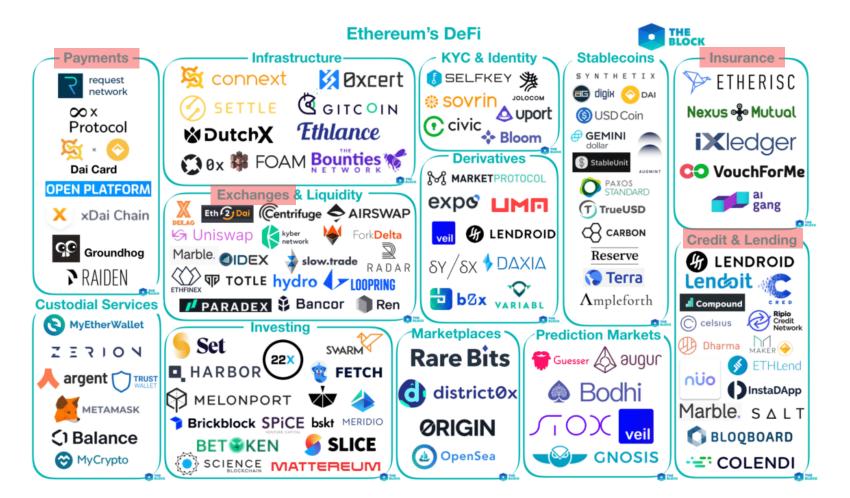
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DeFi ecosystem in Ethereum

• *DeFi (Decentralized Finance):* the blockchain-based form of finance that does not rely on centralized intermediaries.



DeFi ecosystem in Ethereum

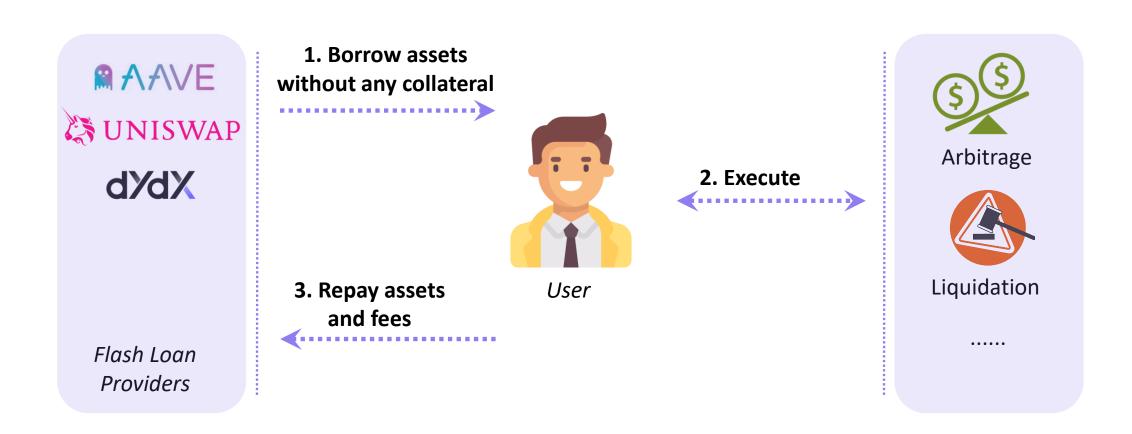
• *DeFi (Decentralized Finance):* the blockchain-based form of finance that does not rely on centralized intermediaries.



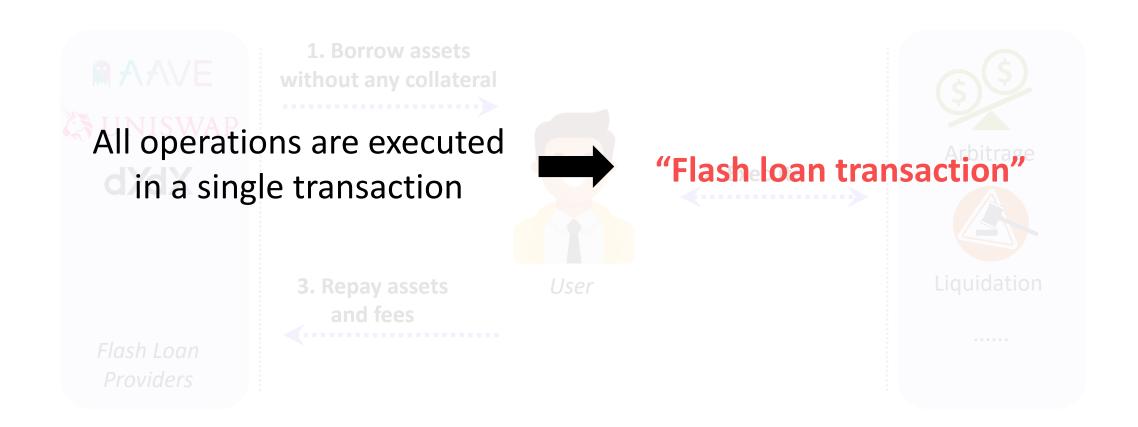
DeFi witnessed rapid growth in early 2021.

Its market capitalization currently stabilizes at around \$40 billion.

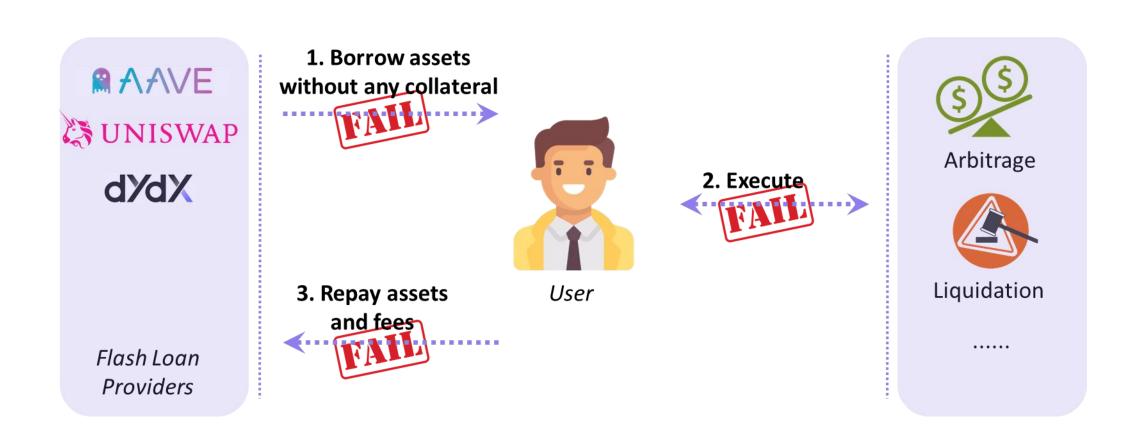
• Flash loan is an *uncollateralized loan* based on the *transaction atomicity*.



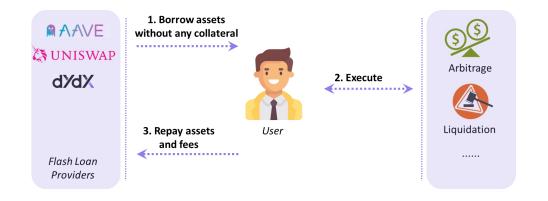
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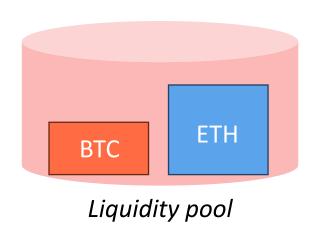


If the user repays, the transaction will be successful and the flash loan provider will earn a fee.



If the user fails to repay, the transaction will be aborted, but the flash loan provider will not lose anything.

 Automated Market Makers (AMMs) are widely used in DeFi protocols to trade assets, e.g., the constant product formula.

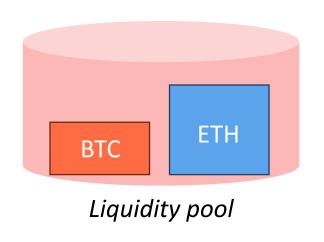




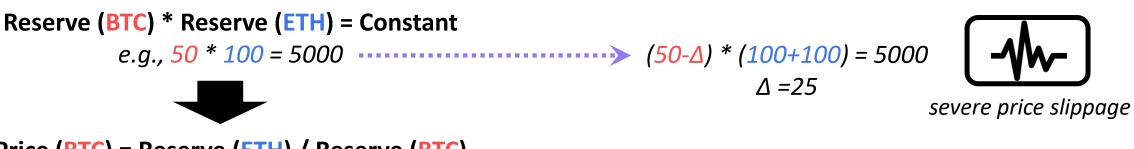




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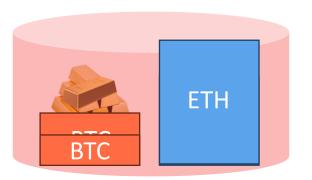


Price (BTC) = Reserve (ETH) / Reserve (BTC)

$$e.g., price (BTC) = 2 ETH/BTC$$

Price (BTC) = $100/25 = 4 ETH/BTC$

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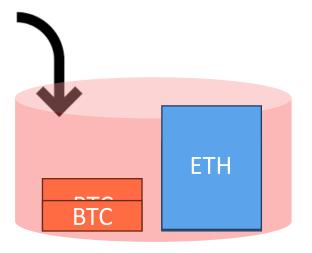
"The market price of every particular commodity is regulated by the proportion between the quantity which is actually brought to market"
----Adam Smith

 Attackers can utilize borrowed assets from flash loans to disrupt asset prices, and make a profit.



1. Exchange large amounts of borrowed ETH

3. Exchange the same amount of BTC for more ETH

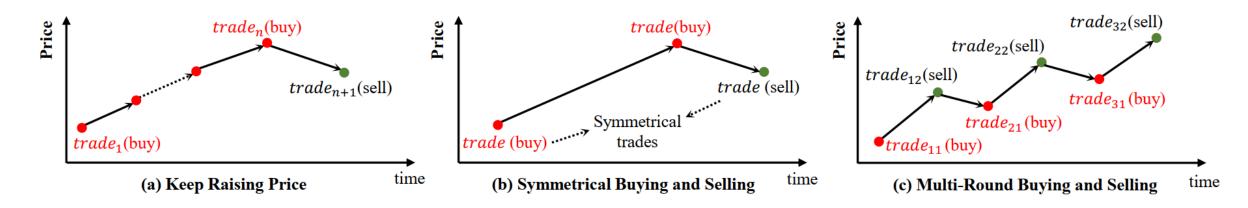


2. Decrease the quantity of BTC to promote its price

FlpAttack: flash loan based price manipulation attack

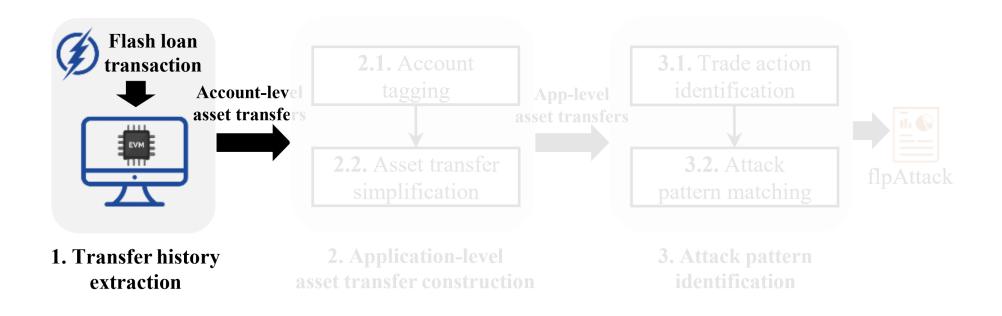
Empirical study on flash loan based attacks

- We collect 44 real-world flash loan based attacks in the past two years, including 23 attacks against Ethereum and 21 attacks against BNB Smart Chain.
- According to whether the attacker makes a profit from the deliberately-made price difference with flash loans, we divide these attacks into 22 price manipulation attacks and 22 non-price manipulation attacks.
- We summarize 3 attack patterns from 17 real-world flpAttacks.



Approach

• We propose *LeiShen* to automatically detect flpAttacks, which takes a flash loan transaction as input and returns a detailed report regarding attack patterns as output.



Transfer history extraction

- We identify flash loan transactions from 3 popular flash loan providers, i.e., Uniswap, AAVE and dYdX.
- A flash loan transaction can be identified by its called functions or event logs.

Provider	Function	Event	
Uniswap	swap uniswapV2Call		
AAVE	flashLoan	FlashLoan	
dYdX	Operate Withdraw callFunction Deposit	LogOperation LogWithdraw LogCall LogDeposit	

 We replay each flash loan transaction in a modified Geth client to obtain the transfer history of Ether and ERC20 tokens.

Application-level asset transfer construction

```
T_1 = (\text{``0xb017''}, \text{``0x65bf''}, 5,637, WETH)
T_2 = (\text{``0x65bf''}, \text{``0x57f8''}, 5,637, WETH)
T_3 = (\text{``0xc02a''}, \text{``0x57f8''}, 5,637, ETH)
T_4 = (\text{``0x57f8''}, \text{``0x65bf''}, 5,637, ETH)
T_5 = (\text{``0x65bf''}, \text{``0x31e0''}, 5,637, ETH)
T_6 = (\text{``0x31e0''}, \text{``0x4d2f''}, 5,637, ETH)
T_7 = (\text{``0x4d2f''}, \text{``0x31e0''}, 51, WBTC)
T_8 = (\text{``0x31e0''}, \text{``0x65bf''}, 51, WBTC)
T_9 = (\text{``0x65bf''}, \text{``0x8b3d''}, 51, WBTC)
```

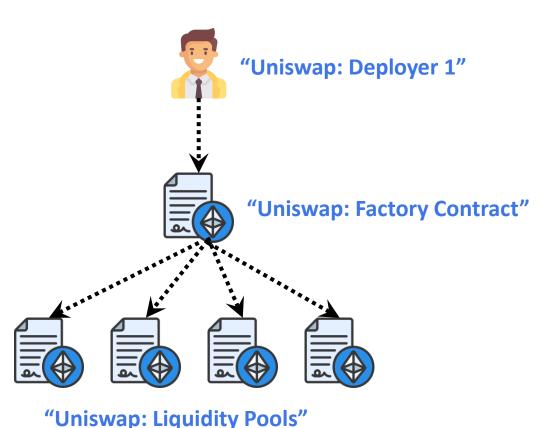
Account-level asset transfers in the bZx attack

Q1: What is the identity of these 160-bit pseudo-anonymous address?

Q2: How to identify user's trade actions and hence identify attack patterns from these complex transfers?

Application-level asset transfer construction

• we collect 52,500 tagged accounts of 119 DeFi applications from Etherscan, and find that 52,482 (>99%) tagged accounts follow the same tagging rule, i.e., accounts with the creation relationship have the same application name tag.



```
T_1 = (\text{``0xb017''}, \text{``0x65bf''}, 5,637, WETH)
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```

Account-level asset transfers

```
tagT_1 = ("bZx", "Kyber", 5,637, WETH)

tagT_2 = ("Kyber", "Kyber", 5,637, WETH)

tagT_3 = ("Wrapped Ether", "Kyber", 5,637, ETH)

tagT_4 = ("Kyber", "Kyber", 5,637, ETH)

tagT_5 = ("Kyber", "Kyber", 5,637, ETH)

tagT_6 = ("Kyber", "Uniswap", 5,637, ETH)

tagT_7 = ("Uniswap", "Kyber", 51, WBTC)

tagT_8 = ("Kyber", "Kyber", 51, WBTC)

tagT_9 = ("Kyber", "bZx", 51, WBTC)
```

Clustering related accounts

Application-level asset transfer construction

• Simplify application-level asset transfers with three heuristic rules, including remove intra-app transfers, remove WETH related transfers and merge inter-app transfers.

```
tagT_1 = ("bZx", "Kyber", 5,637, WETH)
tagT_2 = ("Kyber", "Kyber", 5,637, WETH)
tagT_3 = ("Wrapped Ether", "Kyber", 5,637, ETH)
tagT_4 = ("Kyber", "Kyber", 5,637, ETH)
tagT_5 = ("Kyber", "Kyber", 5,637, ETH)
tagT_6 = ("Kyber", "Uniswap", 5,637, ETH)
tagT_7 = ("Uniswap", "Kyber", 51, WBTC)
tagT_8 = ("Kyber", "Kyber", 51, WBTC)
tagT_9 = ("Kyber", "bZx", 51, WBTC)
```



```
appT_1 = (\text{``bZx''}, \text{``Uniswap''}, 5,637, WETH)

appT_2 = (\text{``Uniswap''}, \text{``bZx''}, 51, WBTC)
```

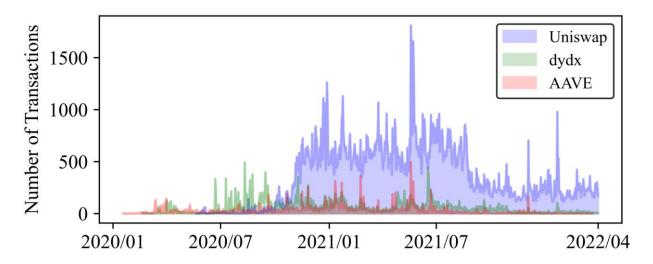
- Remove intra-app transfers: the sender and the receiver have the same tag.
- Remove WETH related transfers: the sender or the receiver has the tag "Wrapped Ether".
- Merge inter-app transfers: in two consecutive transfers, the receiver in the first transfer and the sender in the second transfer have the same tag.

Attack pattern identification

- we identify 3 types of key trades from application-level asset transfers.
 - In a swap, a trader swaps an asset for another asset with a recipient.
 - In a mint liquidity, a trader deposits assets to a DeFi application to mint new assets.
 - In a remove liquidity, a trader returns minted assets to a DeFi application and takes previously deposited assets back.
- With detected trades, we check whether they conform to an attack pattern.

Evaluation

- We implemented LeiShen with *Go* language with ~5,400 LOC, and used modified Geth (v1.10.14) to synchronize Ethereum blockchain and replay transactions.
- We filtered 272,984 flash loan transactions from the first 14,500,000 blocks.
- Average detection time for a flash loan transaction: 10 milliseconds
- For 75% of the transactions: < 16 milliseconds



- **Uniswap:** 208,342
- dydx: 41,741
- AAVE: 22,959

Evaluation

Detecting known flpAttacks

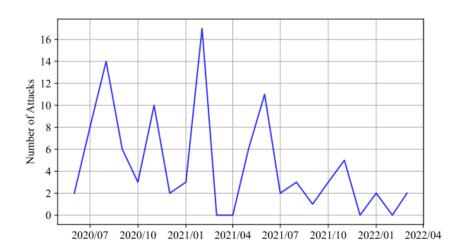
- we detect 14 attacks out of 17 attacks with patterns.

Detecting unknown flpAttacks

- among 180 detected "attacks", 142 attacks were verified as true positives, with an overall precision of 78.9%.
- 109 attacks are previously unknown.

ID	Attacks	DeFi- Ranger	Explorer+ LeiShen	LeiShen
1	bZx-1			√
2	bZx-2		\checkmark	\checkmark
3	Balancer		\checkmark	\checkmark
4	Eminence			\checkmark
5	Harvest Finance	\checkmark	\checkmark	\checkmark
6	Cheese Bank	\checkmark		\checkmark
7	Value DeFi	\checkmark		
8	Yearn Finance	\checkmark		\checkmark
9	Spartan Protocol			\checkmark
10	XToken-1			
11	PancakeBunny Panca			
12	JulSwap			
13	Belt Finance	\checkmark		\checkmark
14	xWin Finance	\checkmark	\checkmark	\checkmark
15	Wault Finance			\checkmark
16	Twindex			
17	AutoShark-2			\checkmark
18	MY FARM PET			
19	PancakeHunny	,		,
20	AutoShark-3	✓.		✓.
21	Ploutoz Finance	\checkmark		\checkmark
22	Saddle Finance	✓		✓

Attack pattern	N	TP	FP	P(%)
KRP	21	21	0	100%
SBS	79	68	11	86.1%
MBS	107	60	47	56.1%



Analyzing unknown flpAttacks

Attacked Applications

 Top3 attacked applications are decentralized exchanges.

Attackers' behaviors

- After finishing an attack, some attackers call the self-destruct function to remove the storage from the Ethereum state.
- Almost all attackers transfer their attack profit with the method of money laundering, such as multi-level intermediary accounts and mixing services.

Profitability

- The maximum yield rate and net profit is $2.2 * 10^5\%$ and >\$6M.
- All attacks result in a total profit of >\$21.8M.

Attacked applications	Attacks	Attackers	Attack contracts	Attacked assets
Balancer	31	5	14	13
Uniswap	16	6	8	5
Yearn	11	1	1	1

	Yield rate (%)	Net profit (\$)
Mean	0.3%	3,509
Min.	0.003%	23
Max.	$2.2*10^5\%$	6,102,198
TOP 10% in AVG	$5.7*10^4\%$	257,078
TOP 20% in AVG	$3.0*10^4\%$	135,522

Conclusion

- We conduct an empirical study on real-world flash loan attacks in the past two years and present three attack patterns to reveal attackers' behaviors in flpAttacks.
- We propose an *LeiShen* to automatically detect flpAttacks conforming to the three attack patterns by utilizing asset transfers of a flash loan transaction.
- We identify a total of 142 true attacks in the first 14,500,000 blocks in Ethereum, including 109 previously-unknown attacks.



LeiShen is the god in charge of thunder and lightning in ancient Chinese mythology.

(https://github.com/tony4paper/LeiShen)

Q&A THANKYOU!