A Framework and Data Set for Bugs in Ethereum Smart Contracts

Abstract—Ethereum is the largest blockchain platform that supports smart contracts. Users deploy smart contracts by publishing the smart contract's bytecode to the blockchain. Since the data in the blockchain cannot be modified, even if these contracts contain bugs, it is not possible to patch deployed smart contracts with code updates. Moreover, there is currently no comprehensive classification framework for Ethereum smart contract bugs, nor a unified bug judgment standard. This makes it difficult for developers to fully understand the potential dangers when developing smart contracts, and it is difficult for researchers to find ways to detect bugs. In this paper, to fill the gap, we first collect as many smart contract bugs as possible from multiple sources and then divide these bugs into 9 categories by tailoring and expanding the IEEE Standard Classification for Software Anomalies, and give the standards for judging each bug. Besides, we provide a set of smart contracts covering all kinds of bugs that we have counted. Developers can learn smart contract bugs from classification frameworks and data sets. Researchers can develop smart contract analysis tools based on the bug judgment standard we give and use our data set as a benchmark for evaluating smart contract analysis tools.

Index Terms—Blockchain, Ethereum, Solidity, Smart contract bug

I. INTRODUCTION

Blockchain [1] is a decentralized and distributed ledger whose data cannot be modified. It was first used as the underlying storage technology to support Bitcoin [1]. Then, Ethereum [2] introduced smart contracts to the blockchain, thereby expanding the scope of blockchain applications. Ethereum smart contracts are typically developed with highlevel programming languages and then compiled into bytecode, which will be deployed to the blockchain through transactions.

Similar to traditional computer programs, it is difficult to avoid bugs in smart contracts. Recent years have witnessed various bugs in smart contracts, resulting in huge losses. For example, the *re-entrancy bug* [3] in *the DAO* smart contract [4] led to a loss of \$60 million.

The security of smart contracts has attracted the attention of many researchers, and several smart contract analysis tools have been developed to detect the potential bugs in smart contracts. However, recent research [5] shows that almost all smart contract analysis tools [6]–[14] can only detect some kinds of smart contract bugs. This cannot avoid the potential bugs in smart contracts, even smart contract analysis tools have been used. We believe that one of the reasons for this situation is the lack of a collection and classification of all the existing smart contract bugs, which makes developers lack guidance when smart contract analysis tools are developed. At

the same time, recent studies showed that one main reason for the proliferation of smart contract bugs is the lack of a comprehensive classification framework for smart contract bugs [3]. Although a few studies summarized and classified some kinds of bugs in smart contracts [3], [15], [16], they have the following limitations:

- The existing classification criteria for smart contract bugs is not comprehensive. Dingman et al. [15] count 49 kinds of smart contract bugs and classify them using NIST framework, but only 24 kinds of bugs were classified accurately, and the remaining 25 kinds of bugs were classified into other category. Smartdec [17] divides smart contract bugs into three levels: blockchain, language and model, and classifies the bugs in each level, but their classification is not comprehensive enough, and some kinds of bugs (eg., locked ether) are not counted and classified.
- Lack of a unified bug judgment standard. Existing work [3], [15], [16] only describes the causes of various kinds of bugs, but does not give the standards for judging the existence of bugs. This makes it difficult for developers to find a way to detect bugs, or makes different smart contract analysis tools have different standards for detecting the same kind of bug. For example, there is still a lack of smart contract analysis tools that can detect short address attack bug, because there is no work yet to give a standard for judging short address attack bug.
- The existing data set for smart contract bugs is incomplete. For example, SmartContractSecurity [18] counts 33 kinds of smart contract bugs, but only some kinds of bugs provide sample smart contracts; crytic [19] provides sample smart contracts for Solidity security issues, but only covers 12 kinds of bugs; Durieux et al. [20] provide a dataset containing 69 problematic smart contracts, but only covered 10 kinds of bugs. Moreover, these data sets use older language versions and provide fewer smart contracts.

In this paper, to fill in the gap, we first carefully collect known bugs of Ethereum smart contracts from many sources, including, academic literature, networks, blogs, and related open-source projects, and finally obtain 323 records describing Ethereum smart contract bugs. Then, by reviewing the *Ethereum Wiki* [21], *Ethereum Improvement Proposals* [22] and the development documents of *Solidity* [23], we remove the bugs that had been fixed by Ethereum. We also merge the bugs caused by the same cause, and eventually 49 kinds

of bugs left. After that, by tailoring and expanding the *IEEE Standard Classification for Software Anomalies*, we classify 49 kinds of bugs into 9 categories based on the causes of these bugs, and give the judgment standard for each kind of bug. Finally, according to the classification framework, we provide a smart contract data set containing the collected bugs. We call the framework and dataset as *Jiuzhou*, which can be found at https://github.com/xf97/JiuZhou.

In summary, we make the following contributions:

- We propose a comprehensive framework for the bugs in Ethereum smart contracts based on *IEEE Standard Classification for Software Anomalies*. We collect these bugs from many sources and then classify them into 9 categories.
- We give the judgment standard for each bug. According
 to the cause of each bug, the most common occurrence,
 common repair methods, and the false positives and
 omissions generated by various smart contract analysis
 tools when detecting these kinds of bugs, we give the
 judgment standard of smart contract analysis tool when
 detecting each bug.
- We provide a data set of problematic smart contracts.
 The smart contracts in the data set to cover all kinds of bugs. It contains 176 smart contracts, including contracts that contain bugs, contracts that fix bugs and misleading contracts. By reading the smart contracts in the data set, smart contract developers and researchers can understand the programming patterns that are likely to cause bugs and the solutions to avoid bugs.
- We use the data set as a benchmark to evaluate various smart contract analysis tools. Based on our evaluation results, we recommend a set of smart contract analysis tools. This set of smart contract analysis tools can detect as many kinds of bugs as possible, and has a good recall and accuracy.

The rest of this paper is organized as follows: Section 2 introduces the necessary background. Section 3 presents the Ethereum smart contract bug classification framework, describes the characteristics and judgment standard of each bug, and gives the severity level of each bug. Section 4 introduces the data set that matches the bug classification. Section 5 uses the *Jiuzhou* data set to evaluate smart contract analysis tools. Section 6 describes the related work. Finally, Section 7 concludes the paper.

II. BACKGROUND

A. Smart contract

When the conditions specified in the contract are met or the smart contracts are called, Smart contracts can be executed automatically on blockchain [2]. In Ethereum, each smart contract or user is assigned a unique address. Smart contracts can be invoked by sending transactions to the address of the contract. *Ether* is the cryptocurrency used by Ethereum, and both contracts and users can trade *ethers*. To avoid abusing the computational resources, Ethereum charges *gas* from each executed smart contract statement.

B. Solidity

Solidity is the most widely used programming language for developing Ethereum smart contracts [23]. Solidity is a Turing-complete and high-level programming language capable of expressing arbitrarily complex logic. Before deployment, the smart contracts written by Solidity are compiled into byte code of Ethereum virtual machine. Solidity provides many built-in symbols to perform various functions of Ethereum. For example, transfer and send are used to perform the transfer of ethers, and keywords such as require and assert are used to check the status. Solidity is a fast-evolving language. The same keyword may have different semantics in different language versions. In general, when using Solidity to develop smart contracts, developers can specify the Solidity language version used by the contracts.

C. IEEE Standard Classification for Software Anomalies

The IEEE Standard Classification for Software Anomalies [24] provides a unified method for the classification of traditional software anomalies. In its latest version, software anomalies are classified into six categories: data, interface, logic, description, syntax, standards. The standard also provides ranking criteria for the effect, severity, and priority of software anomalies. Researchers can flexibly tailor or extend this standard to adapt to different types of software. In this paper, we classify bugs in smart contracts based on the modification of this standard.

III. A CLASSIFICATION FRAMEWORK FOR SMART CONTRACT BUGS

To build a comprehensive classification framework, we collect smart contract errors from many sources, including academic literature, the Web, blogs, and related open-source projects. Since there is no uniform bug naming standard, the same bug may have different names. Consequently, we first merge similar bugs according to their behavior. Then, according to the cause of the bug, we divided all bugs into 9 categories. Each category contains several sub-categories, and the sub-categories contain specific bugs. Finally, according to the severity of different bugs, we give each bug a severity rating.

A. Collect smart contract bugs

First, we collect smart contract bugs from academic literature, networks, blogs, and other resources. For academic literature, we use *smart contract vulnerabilities*, *smart contract bugs*, *smart contract defects*, *smart contract problems*, and *smart contract anomalies* as search keywords to search for papers published since 2014 in *ACM digital library* [25] and *IEEE Xplore digital library* [26]. The reason for the paper after 2014 was chosen is that Ethereum started ICO (initial coin offering) in 2014. For networks and blogs, we mainly focus on the *Github homepage of Ethereum* [27], *the development documents of Solidity* [23], *the official blogs of Ethereum* [28], *the Gitter chat room* [29], *Ethereum Improvement Proposals* [22] and other resources. Second, related open-source

projects are also our focus since the open-source community plays an important role in the field of software security [30]. Specifically, we use *smart contract bugs*, *smart contract prob*lems, smart contract defects, smart contract vulnerabilities and and smart contract anomalies as search keywords to retrieve related open-source projects on GitHub [31]. Besides, many smart contract analysis tool projects are also open-sourced on GitHub [31], and there are also some documents describing smart contract bugs in these projects. Therefore, we also use smart contract analysis tools and smart contract security as search keywords. We focus on the projects for Ethereum smart contracts. After removing duplicate search results, we obtained a total of 266 projects. Third, many famous Ethereum smart contract analysis tools can detect smart contract bugs. We send emails to the authors of these tools asking what kinds of bugs they detect. We also look at the kinds of bugs detected by the Solidity static analysis feature of Remix [32]. Finally, from the resources mentioned above, we collected 323 records describing Ethereum smart contract bugs.

To continuously collect bugs, we expose various kinds of bugs on *Github* and accept other users to extend new bugs. Besides, we developed a crawler called *BugGetter*¹. *BugGetter* runs regularly (now set to 15 days, adjustable), and sends query requests to *Github* every time it runs. *BugGetter* uses keywords such as *smart contract vulnerabilities*, *smart contract bugs*, *smart contract defects*, *smart contract problems*, *smart contract security*, and *smart contract analysis tools* to construct query requests to *Github*, and parses out the list of projects and the update time of these projects. By comparing previously obtained projects and their update time, *BugGetter* will send us an email if a new project appears or an existing project is updated. After receiving the email, we will manually check all changes and update the collected bug results in time.

B. Merge smart contract bugs

Because there is no uniform bug naming standard, even if the names of the collected bugs are different, some similar bugs may point to the same bug. Consequently, we need to merge the duplicate bugs. The collected bugs generally have two attributes, namely, the behaviors causing the bug and the consequences caused by the bug. If there is a bug A. Let,

- the behaviors causing bug A be b(A),
- the consequences caused by bug A be c(A).

If there are two bugs, A and B. Then A and B are merged according to the following steps:

- 1) $b(A) \neq b(B)$. A and B are not merged.
- 2) $b(A) = b(B), c(A) \neq c(B)$. In this case, c(A) and c(B) respectively cover part of the consequences of the bug. We merge A and B, rename the merged bug, summarize c(A) and c(B), and give the consequences after they are merged
- 3) b(A) = b(B), c(A) = c(B). In this case, we choose the name that better reflects the characteristics of the bug

as the name of the merged bug, and then A and B are merged.

After the duplicate bugs are merged, we verify the validity of each bug (that is, the bug has not been fixed), and delete the fixed bugs. Finally, 49 kinds of bugs are left. We list the correspondence of bugs before and after the merger via https://github.com/xf97/JiuZhou/blob/master/Correspondence.xlsx, which allows us to trace back the process of the merger.

C. Classify smart contract bugs

Classification criteria and results

According to IEEE Standard Classification for Software Anomalies [24] issued in 2010, software anomalies are classified into six categories: data, interface, logic, description, syntax, standards. Among them, we do not consider syntax category, because a smart contract with syntax bugs cannot be compiled into bytecode and cannot be deployed in Ethereum. Besides, the bugs caused by gas, smart contract interactions, ethers exchange, and smart contract support software are Ethereum-specific software anomalies. Consequently, the original classification provided by IEEE Standard Classification for Software Anomalies [24] cannot accurately classify these bugs. To accurately classify all kinds of bugs in smart contracts, we add four new categories: security, performance, interaction, and environment. Therefore, we divide smart contract bugs into the following nine categories lexicographically:

- Data. Bugs in data definition, initialization, mapping, access, or use, as found in a model, specification, or implementation.
- 2) **Description**. Bugs in the description of the software or its use, installation, or operation.
- Environment. Bugs due to errors in the supporting software.
- Interaction. Bugs that cause by interaction with other accounts.
- 5) *Interface*. Bugs in specification or implementation of an interface.
- 6) *Logic*. Bugs in decision logic, branching, sequencing, or computational algorithm, as found in natural language specifications or implementation language.
- Performance. Bugs that cause increased gas consumption.
- Security. Bugs that threaten contract security, such as authentication, privacy/confidentiality, property.
- 9) Standard. Nonconformity with a defined standard.

During the process of merging bugs, by consulting *Ethereum Improvement Proposals* [22], *the Ethereum Wiki* [21] and the development documents of *Solidity* [23], we removed bugs that have been fixed by Ethereum (eg., the *call depth attack*, which was fixed in the *EIP150* [33]). Some kinds of bugs are caused by specific *Solidity* versions. Because it is still possible to use these versions of *Solidity* to develop smart contracts, we list the range of *Solidity* versions that cause these kinds of bugs. Any version of *Solidity* can result in bugs that do not have the *Solidity* versions listed.

¹https://github.com/xf97/BugGetter

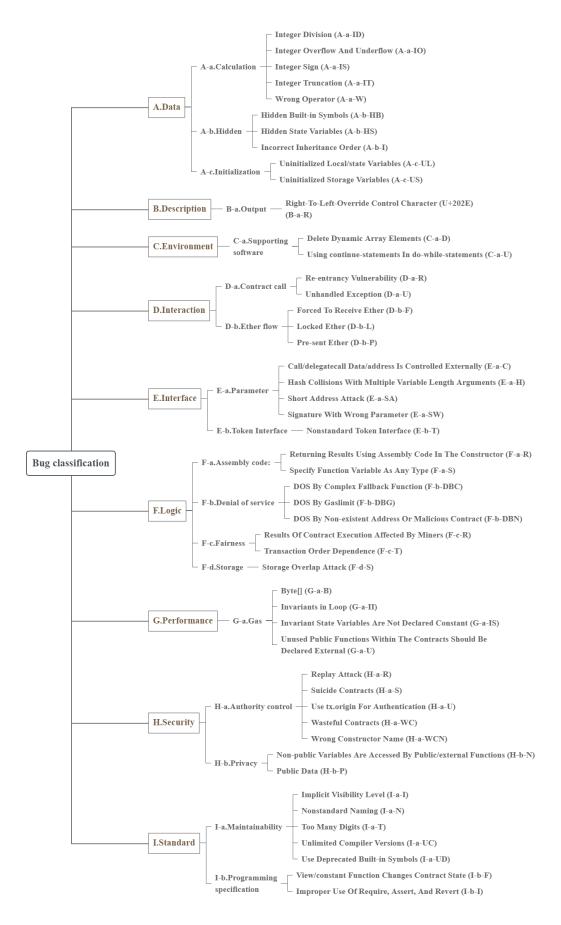


Fig. 1. Smart contract bug statistics and classification

We divide 49 smart contract bugs into 9 categories, each category is divided into several sub-categories, and each sub-category contains several kinds of bugs. Fig 1 shows the classification results of 49 kinds of smart contract bugs. For ease of expression, we assign a corresponding abbreviated name for each bug. The composition rules of the abbreviated name are: category number-subcategory number-final short name. The final short name is composed of the first letter of each word that constitutes the bug name. The original short name is composed of the first letter of each word that constitutes the bug name, and then a substring (this substring can be distinguished from the final short names of other kinds of bugs in the same subcategory) of the original short name as the final short name. Due to space limitations, in this paper, we only introduce some kinds of bugs in detail. For the full version, please access this URL^2 .

Some kinds of smart contract bugs

In this part, we will introduce the causes, consequences and judgment standards of seven kinds of bugs in detail, and illustrate these bugs with examples. These kinds of bugs are either bugs that are difficult to understand the attack process, or bugs not mentioned in other work [3], [15], [16]. Table 1 lists these 7 kinds of bugs.

TABLE I
7 KINDS OF BUGS INTRODUCED IN DETAIL

Bug name	A-a-IS, A-a-W, A-c-US, F-c-T,
abbreviation	D-a-R, E-a-SA, E-a-SW

Integer Sign (A-a-IS) [8]:

- Cause: In Solidity, Converting int type to uint type (and vice versa) may produce incorrect results.
- Consequence: This kind of bugs may result in incorrect integer operation results, which will affect the function of the contract. When the wrong result is used to indicate the number of ethers (or tokens), this kind of bugs will cause economic losses.
- Example: Consider the smart contract in Fig 2, if the attacker calls the withdrawOnce function and specifies that the value of amount is a negative number, then this call will pass the check and transfer out the ether that exceeds the limit (1 ether).
- *Judgment standard*: This kind of bugs exists when the following characteristics are present in the contract:
 - Forcibly convert an *int* variable with a negative value to a *uint* variable.
 - The contract does not check whether this int variable is negative.

Wrong Operator (A-a-W) [18]:

• Cause: Before Solidity version 0.5.0, users can use =+ and =- operators in the integer operation without compiling errors (up to and including version 0.4.26).

```
1 pragma solidity 0.6.2:
 2 - contract signednessError{
         mapping(address => bool) public transferred;
         address public owner;
         constructor() public payable{
             owner = msg.sender;
require(msg.value > 0 && msg.value % 1 ether == 0);
         function withdrawOnce (int amount) public {
10 +
             if ( amount > 1 ether || transferred [msg.sender]) {
11
                  revert();
              msg.sender.transfer(uint(amount));
13
              transferred [msg.sender] = true ;
15
16 }
```

Fig. 2. A contract that contains integer sign bug

```
pragma solidity 0.4.26;
       contract wrongOpe{
   3
           uint256 private myNum;
   4
           address public owner;
   5
           uint256 public winNum;
           uint256 public constant OpeNum = 1:
   6
           constructor(uint256 _num, uint256 _win) public payable{
   8
               myNum = _num;
winNum = _win;
   9
               owner = msg.sender;
  10
  11
               require(msg.value == 1 ether);
  12
  13 +
           function addOne() external payable{
  14
               require(msg.value == 1 wei);
<u>15</u>
               myNum =+ OpeNum; //wrong operator
  16
               isWin(); }
  17 +
           function subOne() external payable{
  18
               require(msg.value == 1 wei);
               myNum =- OpeNum; //wrong operator
  19
               isWin(): }
  20
  21 +
           function isWin() internal{
               if(myNum == winNum)
  22
  23
                   msg.sender.transfer(address(this).balance); }
```

Fig. 3. A contract contains wrong operator bug

- Consequence: This kind of bugs may result in incorrect integer operation results, which will affect the function of the contract. When the wrong result is used to indicate the number of ethers (or tokens), this kind of bugs will cause economic losses.
- Example: Consider the smart contract in Fig 3, the user can adjust the value of myNum by calling addOne/subOne function. When myNum and WinNum are equal, the user gets all the ethers of the contract. But because the developers write the wrong operators (Line 15, 19), the value of myNum will not change.
- *Judgment standard*: This kind of bugs exists when the following characteristics are present in the contract:
 - There is a = + or = operator in the contract.

Uninitialized Storage Variables (A-c-US) [18]:

- *Cause*: The uninitialized storage variable serves as a reference to the first state variable, which may cause the state variable to be inadvertently modified (up to and including version 0.4.26).
- *Consequence*: This kind of bugs may cause key state variables to be rewritten inadvertently, and eventually, the function of the contract will be affected.
- *Example*: Consider the smart contract in Fig 4, when the user calls the function *func*, the owner will be re-assigned

²need to do

to 0x0.

```
1 pragma solidity 0.4.26
2 - contract Uninitialized{
        address owner = msg.sender;
3
        struct St{
4 -
5
            uint a;
6
7 -
        function func() {
8
            St st;
9
            st.a = 0x0; //owner is override to 0.
10
11
   }
```

Fig. 4. A contract contains Uninitialized storage variables bug

- emphJudgment standard: The developers do not initialize the storage variables in the contract.
 - The developers do not initialize the storage variables in the contract.

Transaction Order Dependence (F-c-T) [34]:

- Cause: Miners can decide which transactions are packaged into the blocks and the order in which transactions are packaged. The current main impact of this kind of bugs is the approve function in the ERC20 token standard.
- Consequence: This kind of bugs will enable miners to influence the results of transaction execution. If the results of the previous transactions will have an impact on the results of the subsequent transactions, miners can influence the results of transactions by controlling the order in which the transactions are packaged.
- Example: The approve function allows one address to approve another address to spend tokens on his behalf. The standard implementation of the approve function is shown in Fig 5. We assume that Alice and Tom are two Ethereum users, and Tom runs an Ethereum node. The following steps reveal how Tom used the transaction order dependence bug to monetize:
 - Step 1: Assume Alice has approved Tom to spend n
 of the tokens she holds. Now Alice decides to change
 Tom's quota to m tokens, so Alice sends a transaction
 to modify Tom's quota.
 - Step 2: Since Tom runs an Ethereum node, he knows that Alice will change his quota to m tokens. Then, Tom sends a transaction (eg., Using the transferFrom function to transfer n tokens to himself) to spend Alice's n tokens, and pays a lot of gas to make his transaction executed first.
 - Step 3: The node that obtains the accounting right packs transactions. Because Tom pays more gas, Tom's transaction will be executed before Alice's transaction. Therefore, Tom spent n tokens of Alice first and then is granted a quota of m tokens by Alice, which caused Alice to suffer losses, losses.
- *Judgment standard*: The developers do not initialize the storage variables in the contract.
 - The contract contains the approve function in the ERC20 token standard.

```
function approve(address spender, uint256 value)
public returns (bool) {
    require(spender != address(0));
    _allowed[msg.sender][spender] = value;
    emit Approval(msg.sender, spender, value);
    return true;
}
```

Fig. 5. The standard implementation of the approve function in the ERC20 token standard.

 In the approve function, the quota of the approved address is set from one nonzero value to another nonzero value.

Re-entrancy Vulnerability (D-a-R) [4], [14]:

- Cause: When the call-statement is used to call other contracts, the callee can call back the caller and enter the caller again.
- Consequence: This kind of bugs is one of the most dangerous smart contract bugs, which will cause the contract balance (ethers) to be stolen by attackers.
- Example: We use an example to illustrate the re-entrancy vulnerability. In Fig 6, the contract Re is a contract with a re-entrancy vulnerability, and the balance variable is a map used to record the correspondence between the address and the number of tokens. The attacker deploys the contract Attack, and the value of the parameter _reAddr is set to the address of the contract Re. In this way, the re variable becomes an instance of the contract Re. Then,
 - Step 1: The attacker calls the attack function to deposit ethers into the contract Re and then calls the Re.withdraw function to retrieve the deposited ethers.
 - Step 2: The contract Re executes the withdraw function and uses a call-statement to send ethers to the contract Attack. At this time, the power of control is transferred to the contract Attack, and the contract Attack responds to the transfer using the Attack.fallback function.
 - Step 3: The Attack.fallback function calls the Re.withdraw function to withdraw the ethers again.
 Therefore, the statement (deduct statement) deducting the number of tokens held by the contract Attack will not be executed.
- *Judgment standard*: When there are the following four characteristics in the contract, it will cause the *reentrancy bug*:
 - The *call-statement* is used to send ethers.
 - The amount of gas to be carried is not specified.
 - No callee's response function is specified.
 - Ethers are transferred first and callee's balance is deduced later.

Short Address Attack (E-a-SA) [35]:

• *Cause*: When Ethereum packs transaction data, if the data contains the address type and the length of the address type is less than 20 bits, subsequent data will be used to make up the length of the address type.

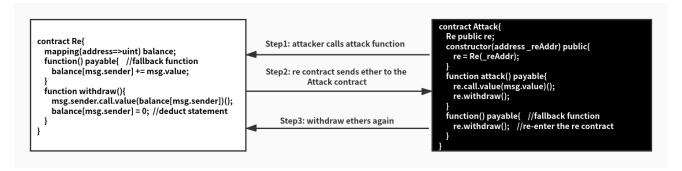


Fig. 6. An example of re-entrancy vulnerability

- Consequence: This kind of bugs may cause the attacker to transfer out tokens (ethers) equivalent to several times the amount he holds.
- Example: We use an example to illustrate short address attacks.
 - Step 1: Tom deploys token contract A on Ethereum, which contains the sendcoin function. The code of sendcoin is shown in Fig 7.

```
function sendCodin(address _to, uint256 _amount) returns(bool){
   if(balance[msg.sender] < _amount)
        return false;
   //using safemath for uint256
   balance[msg.sender] = balance[msg.sender].sub(_amount);
   balance[_to] = balance[_to].add(_amount);
   Tranfer(msg.sender, _to, _amount);
   return true;
}</pre>
```

Fig. 7. Objective function of short address attack

- Step 2: Jack buys 100 tokens of contract A, then registers for an Ethereum account with the last two digits zero (eg. 0x1234567890123456789012 345678901234567800).
- Step 3: Jack calls the function SendCoin with the given parameters, _to: 0x1234567890123456789
 0123456789012345678 (missing last two digits 0), _amount: 50.
- Step 4: The value of _amount is less than 100, so it passes the check. However, because the bits of _to is insufficient, the first two bits (0) of _amount will be added to the _to when the transaction data is packed. Therefore, in order to make up for the length of _amount, the Ethereum virtual machine will add 0 to the last two bits. In the end, the value of _amount is expanded by four times.
- Judgment standard: When there are the following four characteristics in the contract, it will cause the short address attack:
 - The contract uses a function to transfer ethers or tokens.
 - The number of tokens (ethers) and the address for receiving tokens (ethers) are provided by external users.

- There is no operation to check the length of the received tokens (ethers) address in the function.

Signature With Wrong Parameter (E-a-SW) [36]:

- *Cause*: When the parameters of the *ecrecover()* are wrong, the *ecrecover()* will return 0x0.
- Consequence: This kind of bugs will allow the attacker to pass the authentication and then the attacker can manipulate the token (ethers) held by the 0x0 address.
- Example: Considering the smart contract in Fig 8, when the attacker gives the wrong parameters (v, r, s) and the value of the specified parameter $_id$ is 0x0, the attacker can pass the identity verification (Line 10), which eventually leads to the ethers in the contract are destroyed.

```
1 pragma solidity 0.6.2;
2 - contract SigWrongPara{
        bytes32 private idHash;
3
        constructor() public payable{
4 +
            require(msg.value > 0);
 5
 6
            idHash = keccak256(abi.encode(msg.sender));
 7
8
        function getMyMoney(address payable <u>id</u>, uint8 v,
9 +
          bytes32 r, bytes32 s) external returns(bool){
            if(_id != ecrecover(idHash, v, r, s))
10
11
                 return false;
12
             id.transfer(address(this).balance);
13
            return true:
14
15 }
```

Fig. 8. A contract contains a signature with wrong parameter bug

- Judgment standard:
 - There is an operation in the contract that uses the ecrecover() to calculate the public key address,
 - it does not deal with the case where the ecrecover returns 0x0.

D. Severity grading of smart contract bugs

To give developers and researchers a clear understanding of the consequence of each bug, we grade the severity of each bug. According to the *IEEE Standard Classification for Software Anomalies* [24], we classify the effect of software anomalies into the following four categories:

Functionality. The required function cannot be performed correctly (or an unwanted function is performed).

TABLE II
A CLASSIFICATION OF SEVERITY LEVELS OF EACH BUG

Name	Severity	Name	Severity	Name	Severity
A-a-ID	High	A-a-IO	High	A-a-IS	High
A-a-IT	High	A-a-W	High	A-b-HB	High
A-b-HS	High	A-b-I	High	A-c-UL	High
A-c-US	High	B-a-R	Middle	C-a-D	Middle
C-a-U	Middle	D-a-R	Critical	D-a-U	High
D-b-F	Middle	D-b-L	Critical	D-b-P	Middle
E-a-C	High	E-a-H	High	E-a-SA	High
E-a-SW	High	E-b-T	High	F-a-R	Middle
F-a-S	Middle	F-b -DBC	Middle	F-b -DBG	Middle
F-b -DBN	High	F-c-R	Middle	F-c-T	Middle
F-d-S	High	G-a-B	Low	G-a-II	Middle
G-a-IS	Middle	G-a-U	Middle	H-a-R	High
H-a-S	Critical	H-a-U	High	H-a-WC	Critical
H-a- WCN	High	H-b-N	High	H-b-P	High
I-a-I	Low	I-a-N	Low	I-a-T	High
I-a-UC	Low	I-a-UD	Low	I-b-I	Middle
I-b-F	Middle				

- *Performance*. Failure to meet performance requirements, such as rising operating costs.
- Security. Failure to meet security requirements, such as failure of authority control, privacy breaches, property theft, etc.
- Serviceability. Failure to meet maintainability requirements, such as reduced code readability.

According to the harmfulness of the above four effects, the grading criteria of these bugs are described as follows:

- Critical: These kinds of bugs must affect security.
- High: These kinds of bugs may affect security or necessarily affect functionality.
- **Middle**: These kinds of bugs may affect functionality or necessarily affect performance.
- Low: These kinds of bugs may affect performance or necessarily affect serviceability.

According to the grading criteria, the severity level of each bug is shown in Table II.

IV. Jiuzhou: A DATA SET FOR SMART CONTRACT BUGS

A. An overview of Jiuzhou data set

Jiuzhou provides examples of each bug to help smart contract researchers and developers better understand the bugs and use these contracts as a benchmark to evaluate the capabilities of smart contract analysis tools. Jiuzhou provides 176 smart contracts, covering all smart contract bugs counted in this paper, including smart contracts containing bugs, smart

contracts without bugs, and some misleading contracts that we manually write to mislead smart contract analysis tools.

For each kind of bug, *Jiuzhou* provides at least a contract with the bug and a contract without the bug. For certain contract-context related bugs, we provide *misleading contracts*. The role of *misleading contracts* is to induce smart contract analysis tools to misreport or omit. Currently, we only develop misleading contracts manually. In the future, we will study how to automatically generate misleading contracts.

B. Smart contract sources

We collect smart contracts from the following three sources:

- Other smart contract data sets (eg., [18], [19]).
- Sample code for smart contract analysis tool papers (eg. [8]), or sample code for smart contract audit checklists (eg. [37]). However, most of the sample code only contains the function or part of the contract, so we need to supplement these sample codes as a complete smart contract.
- We manually write smart contracts based on the characteristics of these bugs. For some kinds of bugs with only text descriptions but no sample code, we manually write smart contracts.

In addition to the sample codes obtained from smart contract analysis tool papers and smart contract audit checklists, we also modify some smart contracts collected from other smart contract data sets. Because these smart contracts of other data sets are developed using some older *Solidity* versions, we manually rewrite these smart contracts using the latest version of *Solidity* that contains these kinds of bugs. Table III shows the distribution of the number of unchanged smart contracts, modified smart contracts, and smart contracts that we developed manually.

TABLE III
NUMBER DISTRIBUTION OF THREE SMART CONTRACTS

	Unchanged smart contracts	Modified smart contracts	Handwritten smart contract	
Num	21	69	86	

C. Comparison with other data sets

All smart contracts of *Jiuzhou* data set are developed using the latest version (0.4.26, 0.5.16 or 0.6.2) of *Solidity* containing bugs. Compared with several commonly used smart contract datasets [18], [19], [38], [39], *Jiuzhou* provides more smart contracts, uses the newer *Solidity* versions, and covers more kinds of smart contract bugs. A comparison of *Jiuzhou* with other commonly used data sets is shown in Table IV.

D. Possible use of the Jiuzhou data set

The Jiuzhou data set has the following possible uses:

- For smart contract developers, they can learn about smart contract bugs by reading these smart contracts.
- For smart contract analysis tool developers, the *Jiuzhou* data set can guide them to develop smart contract analysis

TABLE IV
COMPARISON OF *Jiuzhou* WITH OTHER DATA SETS

Data set	Number of contracts	Kinds of bugs	Solidity version
Jiuzhou	176	49	0.4.24 to 0.6.2
ethernaut [38]	21	21	0.4.18 to 0.4.24
not-so-smart -contracts [19]	25	12	0.4.9 to 0.4.23
SWC-registry [18]	114	33	0.4.0 to 0.5.0
capturetheether [39]	19	6	0.4.21

tools and they can learn about smart contract programming patterns that are prone to false positives by reading misleading contracts.

 For smart contract analysis tool evaluators, they can use these smart contracts as a benchmark to evaluate the capabilities of smart contract analysis tools.

V. EVALUATION OF SMART CONTRACT ANALYSIS TOOLS

A. Overview

We use smart contracts in the *Jiuzhou* data set as a benchmark to evaluate the capabilities of several smart contract analysis tools. An analysis tool with good capability should be able to analyze as many kinds of bugs as possible, and it also has good accuracy and recall rate. We use the following indicators to measure the capabilities of the analysis tools:

- **Coverage**. Coverage refers to the proportion of various bugs that can be detected by the analysis tool in the various bugs of *Jiuzhou* statistics.
- Accuracy and recall. We use equation 1 and equation 2 to calculate the recall rate and accuracy. tp means that the tool analyzes the existence of the bug and the bug does exist. fp means that the tool analyzes the existence of the bug but the bug does not exist. fp means that the bug exists but the tool does not report the bug. The definitions of tp, fp, and fn are shown in Table V.

TABLE V
DEFINITION OF tp, fp, fn

Analysis	exist		non-exist
exist	tp	1	fn
non-exist	fp	1	tn

• **Total score**. The total score is calculated by equation 3. The product of the coverage and recall represents the actual recall rate, and the product of the coverage and accuracy represents the actual accuracy rate. The sum of the actual recall rate and the actual accuracy rate is used as an indicator of the capability of a tool.

$$Recall = (tp \div (tp + fn)) \tag{1}$$

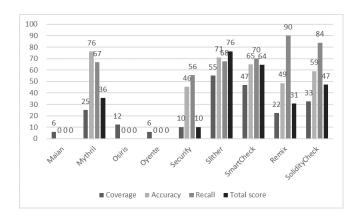


Fig. 9. Coverage, recall and accuracy of various tools

$$Accuracy = (tp \div (tp + fp)) \tag{2}$$

$$Total\ score = Coverage \times (Recall + Accuracy)$$
 (3)

The resources we investigated include 12 smart contract analysis tools. Among them, the nine tools are selected for evaluation, as shown in Table VI. The remaining three tools, i.e., *contractFuzzer* [11], *manticore* [40] and *Solidity-analyzer* [41] we encountered problems during the installation, so we do not evaluate them in this paper.

TABLE VI
THE SELECTED NINE SMART CONTRACT ANALYSIS TOOLS FOR
EVALUATION

Tool	Maian [42], Mythril [43], Osiris [44], Oyente [45], Securify [46],
1001	Slither [47], SmartCheck [48], Remix ³ [49], SolidityCheck [50]

B. Coverage

We obtain the detected bugs by consulting the relevant documents of the analysis tools. For the analysis tools of missing documents, we ask the developers via email. Fig 9 shows the coverage of various tools.

C. Accuracy and recall

The test cases we chose are the smart contracts corresponding to the kinds of bugs that a tool claims to be able to detect. We use each analysis tool to test these contracts, and then calculate the recall and accuracy of each tool. The results are shown in Fig 9.

The three tools *Maian*, *Osiris*, and *Oyente* cannot analyze the smart contract of Solidity 0.4.26 and subsequent versions, so no bug is detected, resulting in the recall and accuracy of these three tools of 0.

³We use the *solidity static analysis* of Remix

D. Total score and recommendation tools

According to equation 3, we calculate the total score of each tool, and the results are shown in Fig 9. Based on the total score, the best performing analysis tool is *Slither*. The analysis tool with the highest accuracy is Mythril, Remix with the highest recall. Based on the performance of each tool, We recommend using Slither, SmartCheck, Mythril, and Remix to analyze the contract together. This group of tools can detect 37 kinds of bugs, and has a good accuracy and recall rate. Besides, the solidity static analysis function and the free plugin of Mythril can be used in the Remix IDE, SmartCheck also provides online services. At the same time, the installation and use of the Slither is not complicated too. This means that these four tools can be used conveniently. It is worth noting that there are still 12 kinds of bugs (eg., short address attack) that cannot be detected by any of these 9 analysis tools, and developers have to manually check these kinds of bugs.

VI. RELATED WORK

A. Statistics of Ethereum smart contract bugs

Some studies have focused on statistical smart contract bugs. Destefanis et al. [51] propose the need to establish the blockchain software engineering by researching the accident of the freeze of the Ethereum parity wallet. Wohrer et al. [52] describe six kinds of smart contract security models that can be applied by smart contract developers to prevent possible attacks. Delmolino et al. [53] summarize four common smart contract programming pitfalls by investigating students' mistakes in learning smart contract programming. Atezi et al. [3] summarize 11 kinds of programming traps that may lead to security bugs. They believe that one of the main reasons for the continuous proliferation of smart contract bugs is the lack of inductive documentation for smart contract bugs. Chen et al. [16] collect smart contracts from Stack Exchange and Ethereum, define 20 kinds of code smell for smart contracts through manual analysis of smart contracts. Wang et al. [54] propose a research framework for smart contracts based on a six-layer architecture and describe the bugs existing in smart contracts in terms of contract vulnerability, limitations of the blockchain, privacy, and law. Through interviews with smart contract developers, Zou et al. [55] reveal that smart contract developers still face many challenges when developing contracts, such as rudimentary development tools, limited programming languages, and difficulties in dealing with performance issues. Sayeed et al. [56] divide the attacks on Ethereum smart contracts into four categories according to the attack principle and introduce 7 kinds of smart contract bugs, and then they provide suggestions for implementing secure smart contracts. Feist et al. [57] describe 45 kinds of smart contract bugs and implement slither, a static analysis tool of smart contract, to detect these bugs. However, these studies only provide statistics for smart contract bugs but do not classify smart contract bugs.

B. Category Ethereum Smart Contract Bugs

Some researches have contributed to the classification of smart contract bugs. Dingman et al. [15] first count the existing bugs of Ethereum smart contract and then classify them using the NIST framework. They count 49 kinds of bugs and then classify 24 of them. Tikhomirov et al. [12] divide 20 kinds of smart contract bugs into security, functional, operational, and developmental, and give the severity of various bugs. Zhang et al. [58] divide 20 kinds of smart contract bugs into three categories: security, functional, and potential threats in the code according to the hazards of the bugs, and develop a smart contract analysis tool SolidityCheck to detect these bugs. Smartdec [17] divides the Ethereum smart contract bugs into three major categories: blockchain, language, and model. Each major category contains several sub-classes, and each sub-class contains specific bugs. Their classification covers a total of 33 kinds of smart contract bugs. However, these classifications have three main limitations. First, they have not include all kinds of smart contract bugs, Second, they may include bugs that have been fixed officially. Third, they do not provide supporting smart contract data sets.

C. Ethereum problem smart contract data sets

Some organizations and researchers provide problem smart contract data sets. SmartContractSecurity provides a list of smart contract bugs, including 33 kinds of bugs and problem smart contracts, but SmartContractSecurity does not classify these bugs, and some kinds of bugs also lack sample smart contracts [18]. crytic provides some examples of Solidity security issues covering 12 kinds of bugs, but 11 of them have not been updated for two years [19]. OpenZeppelin provides a wargame based on Web3 and Solidity called ethernaut [38]. ethernaut contains 21 problem smart contracts, but ethernaut does not describe which bugs these smart contracts contain. Durieux et al. [20] collect 47,587 Ethereum smart contracts, and then manually mark the smart contract bugs in 69 of these contracts, and then based on the smart contract bug classification provided by DASP [59], smart contract bugs in 69 contracts are divided into ten categories. In general, these smart contract data sets do not cover all kinds of smart contract bugs, and the number of smart contracts provided is relatively small.

VII. CONCLUSION

In this paper, we first count existing Ethereum smart contract bugs. Then we classify these bugs according to *IEEE Standard Classification for Software Anomalies* and give the judgment standard for each bug. Second, according to our bug statistics and classification, we provide a matching smart contract data set. Finally, we use the smart contracts in *Jiuzhou* data set as a benchmark to evaluate smart contract analysis tools, and recommend a set of smart contract analysis tools.

For future work, first we plan to generate misleading contracts automatically. Second, we try to study methods to automatically collect smart contract bugs and accurately classify them.

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