

# Quantitative Analysis of Smart Contracts

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# Outline

Smart Contracts

Bugs in Smart Contracts

Language Design

Games and Abstraction

Experimental Results

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- ▶ This is actually a consensus about the results of a computation
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- ▶ Programs run on the Blockchain are called Decentralized Applications (dapps) or Smart Contracts
- ▶ Ethereum supports arbitrary stateful Turing-complete smart contracts



## An Example Contract – Token Transfer

```
1  contract Token {
2      mapping(address=>uint) balances;
3
4      function buy_tokens() payable {
5          balances[msg.sender] += msg.value;
6      }
7
8      function transfer( address to, uint amount ) {
9          if(balances[msg.sender]>=amount) {
10             uint x = balances[msg.sender];
11             uint y = balances[to];
12             balances[msg.sender] = x - amount;
13             balances[to] = y + amount;
14         }
15     }
16 }
```

## Another Example – Three-way Lottery

```
1  contract Lottery {
2
3  address a=0,b=0,c=0;
4
5  function register() payable {
6      require(msg.value == 1);
7      require(a == 0 || b == 0 || c == 0);
8      require(msg.sender!=a && msg.sender!=b && msg.sender!=c);
9      if(a==0)
10         a = msg.sender;
11     else if(b==0)
12         b = msg.sender;
13     else
14         c = msg.sender;
15 }
16
17 mapping(address => uint) hashedChoices;
18
19 function makeChoice(uint choice){
20     require(a!=0 && b!=0 && c!=0);
21     require(msg.sender==a||msg.sender== b||msg.sender==c);
22     require(hashedChoices[msg.sender] == 0);
23     hashedChoices[msg.sender] = choice;
24 }
```

## Another Example – Three-way Lottery

```
1 mapping(address => uint) actualChoices;
2
3 function revealChoice(uint choice)
4 {
5     require(msg.sender==a||msg.sender== b||msg.sender==c);
6     require(hashChoices[a]!=0);
7     require(hashChoices[b]!=0);
8     require(hashChoices[c]!=0);
9     require(sha256(choice) == hashChoices[msg.sender]);
10    actualChoices[msg.sender] = choice;
11 }
```

## Another Example – Three-way Lottery

```
1 address winner = 0;
2
3 function claim()
4 {
5     require(actualChoices[a]!=0);
6     require(actualChoices[b]!=0);
7     require(actualChoices[c]!=0);
8
9     if(actualChoices[a]%3 == actualChoices[b]%3)
10        winner = a;
11     else if((actualChoices[b] + actualChoices[c])%2 == 0)
12        winner = c;
13     else
14        winner = b;
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16     winner.send(this.balance);
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Sometimes the two types coincide, i.e. a coding error leads to an incentive for dishonest interaction.

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- ▶ Well-defined Phases
- ▶ Concurrent Moves using Commitment Schemes

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  - ▶ It has no loops
  - ▶ Each function is assigned a time interval
  - ▶ There is native support for commitment schemes, i.e. some functions get their parameters from different parties
- ▶ We showed that many real-world contracts can be written in our language pretty easily

# How Our Language Looks

```
contract RPS {
id Alice = issuer;
id Bob = null;
numeric bid[0,100] = 0;
numeric AliceWon[0,1] = 0;
numeric BobWon[0,1] = 0;
//0 denotes no choice,
//1 rock, 2 paper,
//3 scissors

function registerBob[1,10]
(payable _bid[0,100] : caller) {
    if(Bob==null) {
        Bob = caller;
        bid=_bid;
    }
    else
        payout(caller, bid);
}
```



```

function play[11,20]
  (numeric AlicesMove[0,3]=0: Alice,
   numeric BobsMove[0,3]=0: Bob,
   payable AlicesBid[0,100]=0: Alice)
  {
    id winner = null;
    if(AlicesBid!=bid)
      winner = Bob;
    else
      //set winner according to RPS rules
    if(winner==null)
      {
        payout(Alice, bid);
        payout(Bob, bid);
      }
    else
      payout(winner, 2*bid);
    //set the values of AliceWon and BobWon accordingly
  }

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- ▶ The objective function can include not only monetary gains and losses, but also mathematical and logical expressions over the value of global variables at the end of the contract.
- ▶ For example, for a party  $p$ , her objective in a lottery can be:

$$p^+ - p^- + 10 \times [\text{winner} == p]$$

where  $p^+$  is the amount she received from the contract and  $p^-$  is the amount she paid. In a correct implementation of the three-way lottery, we expect the value of the contract to be  $10/3$ .

## Contract Values

- ▶ Intuitively, a policy  $\sigma_p$  for a party  $p$  of the contract is a function that, given the sequence of states the contract has visited up until now, suggests an action (or a distribution over actions) for  $p$ .

This action is always in the form of issuing a message, i.e. setting a value for a variable, paying an amount or calling a function. It should always follow the rules of the contract.

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- ▶ The value of the contract for a party  $p$  is her guaranteed earnings/losses assuming that all other parties are adversarial.
- ▶ Formally, if we let  $u_p(\sigma_p, \sigma_{-p})$  be the earnings of party  $p$  when she follows a policy  $\sigma_p$  and others follow  $\sigma_{-p}$ , then her contract value is:

$$\sup_{\sigma_p} \inf_{\sigma_{-p}} u_p(\sigma_p, \sigma_{-p}).$$

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- ▶ A finite set  $S$  of states,
- ▶ An initial (start) state  $s_0$ ,
- ▶ A finite set  $A$  of actions,
- ▶ Two action assignment functions  $\Gamma_i : S \rightarrow 2^A \setminus \{\emptyset\}$ .  
Intuitively,  $\Gamma_i$  decides which actions are available to player  $i$  at each state;

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Intuitively,  $\Gamma_i$  decides which actions are available to player  $i$  at each state;
- ▶ and a transition function  $\delta : S \times A \times A \rightarrow S$  that assigns to every state  $s \in S$  and action pair  $a_1 \in \Gamma_1(s), a_2 \in \Gamma_2(s)$  a successor state  $\delta(s, a_1, a_2) \in S$ .

# Gameplay

- ▶ The game starts at state  $s_0$ . At each state  $s_i \in S$ , player 1 chooses an action  $a_1^i \in \Gamma_1(s_i)$  and player 2 chooses an action  $a_2^i \in \Gamma_2(s_i)$ . The choices are made simultaneously and independently. The game subsequently transitions to the new state  $s_{i+1} = \delta(s_i, a_1, a_2)$  and the same process continues.

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- ▶ This leads to an infinite sequence of tuples  $p = (s_i, a_1^i, a_2^i)_{i=0}^{\infty}$  which is called a *play* of the game.
- ▶ The notion of a policy (strategy) is defined in the usual way

# Utilities and Game Values

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- ▶ A *Game* is a pair  $(G, u)$  where  $G$  is a game structure and  $u$  is a utility function for player 1. We assume that player 1 is trying to maximize  $u$ , while player 2 aims to minimize it.
- ▶ The L-step finite-horizon value of a game  $(G, u)$  is defined as

$$V_L(G, u) := \sup_{\sigma_1} \inf_{\sigma_2} \mathbb{E}^{(\sigma_1, \sigma_2)} [u_L(p)],$$

where  $\sigma_i$  iterates over all possible mixed strategies of player  $i$ . This models the fact that player 1 is trying to maximize the utility in the first L steps of the run, while player 2 is minimizing it.

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- ▶ The number of parties is bounded
- ▶ The number of function calls by each party at each time frame is also bounded.
  - ▶ In real-life contracts one's ability to call many functions is limited by the capacity of a block in the blockchain.



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- ▶ We then create two new games  $(G^\uparrow, u^\uparrow)$  and  $(G^\downarrow, u^\downarrow)$ . These games are obtained by merging the states that are in the same partition to form a single “abstract” state.

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- ▶ We first partition the states into several sets.
- ▶ We then create two new games  $(G^\uparrow, u^\uparrow)$  and  $(G^\downarrow, u^\downarrow)$ . These games are obtained by merging the states that are in the same partition to form a single “abstract” state.
- ▶ In  $u^\uparrow$ , an abstract state’s utility is the maximum utility among normal states that it corresponds to. In  $u^\downarrow$ , we take the minimum utility.
- ▶ When at an abstract state  $s$  the players play actions  $a_1$  and  $a_2$ , the game can transition to any abstract state that contains the result of a transition with the same actions from a normal state in  $s$ . In  $G^\uparrow$  player 1 chooses the resulting transition (solves the nondeterminism), while in  $G^\downarrow$  player 2 does this.

# Soundness, Refinement and Completeness in the Limit

It is easy to check that our abstraction has the following three properties (for proof and formal treatment see the paper):

- ▶ **Soundness.** The value of  $(G^\downarrow, u^\downarrow)$  is always a lower-bound for that of  $(G, u)$  and the value of  $(G^\uparrow, u^\uparrow)$  is always an upper-bound. Hence, solving each abstracted game pair gives us an interval that contains the value of the original game.
- ▶ **Refinement.** If we refine our partitions by breaking each set into smaller sets, the interval shrinks.
- ▶ **Completeness in the Limit.** For any  $\epsilon > 0$ , we can refine our partitions such that the length of the interval becomes less than  $\epsilon$ .



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- ▶ We implemented the approach with some heuristics for refining abstractions
- ▶ We tested our approach on several real-world bugs (including the two examples shown before).
- ▶ In each case, we coded both the correct and the buggy variant of the contract.



- ▶ The original contracts had up to more than  $10^{23}$  concrete states.

Sale						
Size	Abstractions					
<b>&gt; <math>4.6 \cdot 10^{22}</math></b>	Correct Program			Buggy Variant		
	states	[ <i>l</i> , <i>u</i> ]	time	states	[ <i>l</i> , <i>u</i> ]	time
	17010	[0 , 2000]	226	17010	[0 , 2000]	275
	75762	[723 , 1472]	1241	81202	[1167 , 2000]	1733
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	1040	[0 , 2000]	20	6561	[0 , 2000]	237
	32880	[844 , 1793]	562	131520	[1716 , 2000]	3979
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- ▶ We refined our abstraction to the point that the intervals for values of the two contracts became disjoint.

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- ▶ The runtimes were a few hours in some cases, but we are happy with it given that each contract should only be checked once before its deployment.

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