ZKP from MPC

April 16, 2019

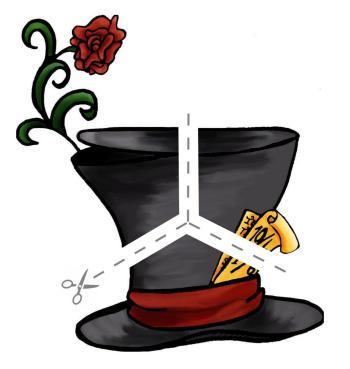
TOY SCENARIO

Alice: I know how to make hats.

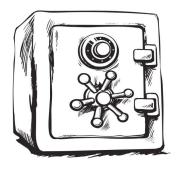
Bob: Prove it!

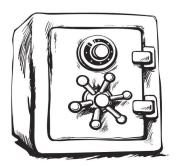
Alice: I'm not showing you; you might steal my design.



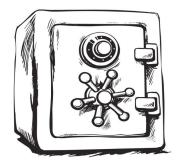


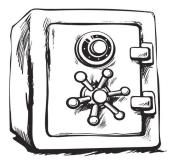
Step 1: Alice cuts up her hat (read: proof) into pieces.

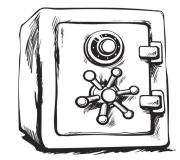


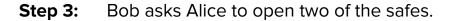


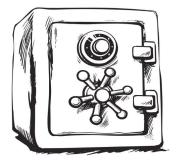
- Step 2:
- **2:** Alice stores each piece into a different safe and then gives the safes to Bob.

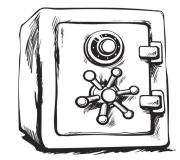








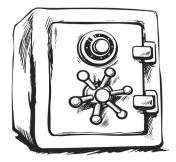






p 3: Bob asks Alice to open two of the safes.

He checks to see if the pieces fit together (read: the views of the hat are consistent).



TOY SCENARIO: HAT LEMMA

Denote the hat pieces by H_1, \dots, H_n . We say that a collection of hat pieces are **consistent** if they fit together and are clearly from the same hat.

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Lemma. H_1, \dots, H_n are consistent if and only if H_i and H_j are pairwise consistent for each $1 \le i \le j \le n$.

Corollary. If H_1, \dots, H_n are not consistent, then there exists at least one pair of H_i and H_j that are not consistent with each other.

- **Corollary.** If H_1, \dots, H_n are not consistent, then there exists at least one pair of H_i and H_i that are not consistent with each other.
- **Question.** If Alice is lying and her hat pieces are not consistent, what is the probability that Bob discovers her deceit?

- **Corollary.** If H_1, \dots, H_n are not consistent, then there exists at least one pair of H_i and H_i that are not consistent with each other.
- **Question.** If Alice is lying and her hat pieces are not consistent, what is the probability that Bob discovers her deceit?

Answer.

$$\Pr\left(\text{Bob discovers Alice's deceit}\right) \ge {\binom{n}{2}}^{-1}$$

- **Corollary.** If H_1, \dots, H_n are not consistent, then there exists at least one pair of H_i and H_i that are not consistent with each other.
- **Question.** If Alice is lying and her hat pieces are not consistent, what is the probability that Bob discovers her deceit?

Answer. $\Pr\left(\text{Bob discovers Alice's deceit}\right) > \frac{1}{n^2}$

TOY SCENARIO: LIMITATION

Bob still learns something about Alice's hat designs from her cut-up pieces.

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Bob still learns something about Alice's hat designs from her cut-up pieces.

Question. Is there a way to split up her proof so that no two shares can be combined to reveal any information?

Answer. Use secure multi-party computation!

ZKP from MPC

Zero-Knowledge from Secure Multiparty Computation*

Yuval Ishai[†] Eyal Kushilevitz[‡] Rafail Ostrovsky[§] Amit Sahai[¶]

Abstract

A zero-knowledge proof allows a prover to convince a verifier of an assertion without revealing any further information beyond the fact that the assertion is true. Secure multiparty computation allows n mutually suspicious players to jointly compute a function of their local inputs without revealing to any t corrupted players additional information beyond the output of the function.

We present a new general connection between these two fundamental notions. Specifically, we present a general construction of a zero-knowledge proof for an NP relation R(x, w) which only makes a *black-box* use of any secure protocol for a related *multi-party* functionality f. The latter protocol is only required to be secure against a small number of "honest but curious" players. We also present a variant of the basic construction that can leverage security against a large number of *malicious* players to obtain better efficiency.

As an application, one can translate previous results on the efficiency of secure multiparty computation to the domain of zero-knowledge, improving over previous constructions of efficient zeroknowledge proofs. In particular, if verifying R on a witness of length m can be done by a circuit Cof size s, and assuming one-way functions exist, we get the following types of zero-knowledge proof protocols:

- Approaching the witness length. If C has constant depth over ∧, ∨, ⊕, ¬ gates of unbounded fan-in, we get a zero-knowledge proof protocol with communication complexity m · poly(k) · polylog(s), where k is a security parameter.
- "Constant-rate" zero-knowledge. For an *arbitrary* circuit C of size s and a bounded fan-in, we get a zero-knowledge protocol with communication complexity $O(s) + poly(k, \log s)$. Thus, for large circuits, the ratio between the communication complexity and the circuit size approaches a constant. This improves over the O(ks) complexity of the best previous protocols.

Keywords: Cryptography, zero-knowledge, secure computation, black-box reductions

PRELIMINARIES

Definition. An **NP-relation** R(x,w) is an efficiently decidable binary relation that is polynomially bounded (i.e. $|w| \le p(|x|)$ where p polynomial).

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Examples. Let x = (V,E) be a graph

- I. w is a Hamiltonian path in x
- II. w is a 3-coloring of x

Let $x \in (\mathbf{Z}/n\mathbf{Z})^{\times}$ be relatively prime to n III. w is a square root of x (i.e. $w^2 \cong x \mod n$)

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Remark. Any NP-relation *R* defines an NP-language: $L = \{ x : \exists w, R(x,w) = 1 \}.$

PRELIMINARIES: GOAL



Bob: Prove it!

Alice: I don't want to share *w* with you.



Assume that there is an SMPC algorithm *f* that computes:

$$\mathsf{f}(\mathsf{x},\mathsf{w}_1,\cdots,\mathsf{w}_n) \equiv R(\mathsf{x},\mathsf{w}_1 \oplus \cdots \oplus \mathsf{w}_n)$$

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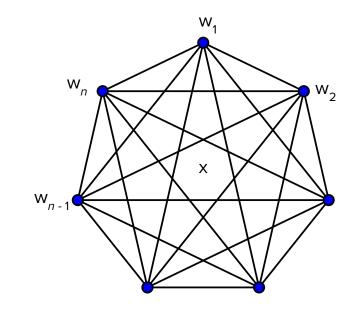
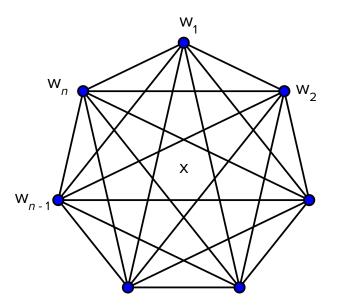
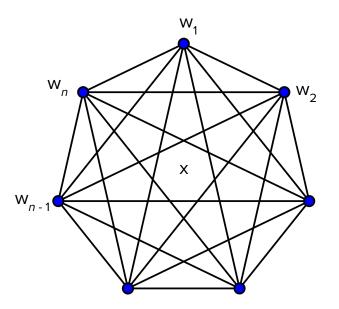


Figure. Each party has a secret share of *w*, where $w \equiv w_1 \oplus \cdots \oplus w_n$ Together, they jointly and privately verify *R*(x, w).

Step 1: Alice simulates the SMPC protocol that verifies that w is a witness to $x \in L$.

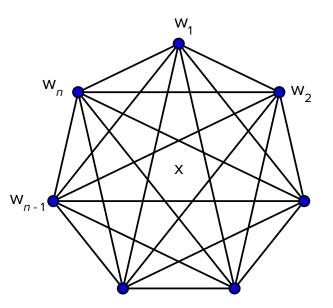


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- Remark. After running the protocol, each party will have a **view** of the messages it received and sent along with the randomness it used.

These views will act as cut-up pieces of the proof that $x \in L$.



Recall that a **commitment scheme**, COM, is a protocol that allows one to commit a message while hiding the message from others. Later, one is able to reveal the original message.

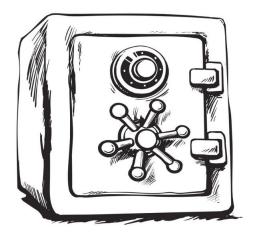
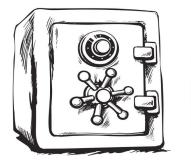
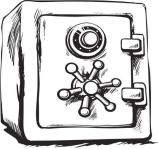
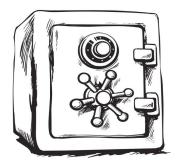


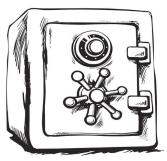
Figure. A commitment protocol is analogous to sending a message locked in a safe. Later in time, the sender can open the safe to reveal the committed message.

Step 2: Alice commits each of the views using a commitment scheme, sending the commits to Bob.

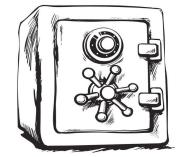


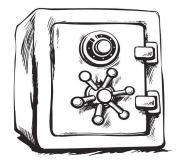




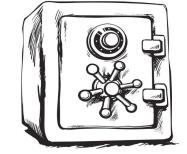


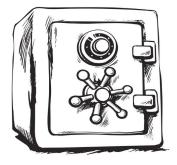
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revealing two views of the MPC protocol.





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- **Decision.** If the views are inconsistent, Bob rejects. Otherwise, Bob accepts.





Remark. The **soundness error** (i.e. probability that Bob accepts a invalid proof) at this point is as before:

$$\Pr\left(\text{Bob accepts a false proof}\right) < 1 - \frac{1}{n^2}$$

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We could make this probability negligibly small (say, w.p. 2^{-k}) through repetitions (say, kn² times).

HIGH-LEVEL OVERVIEW: UPSHOT

Main Result. We can build ZKP protocols given black-box access to MPC protocols.

Let *P* be the prover and *V* be the verifier. Let *L* be an NP-language.

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- **Definition** A **zero-knowledge proof** (ZKP) is a protocol (*P*,*V*) where for each (informal). $x \in L$, the prover tells the verifier essentially nothing but $x \in L$.
- **Remark.** For a given input x, the prover and verifier will exchange messages according to some underlying probability distribution, say A_{x} , that depends on x.

We call a collection of distributions $\{A_x\}_{x \in X}$ a **probability ensemble** indexed by X.

FORMALITIES

Definition. Two probability ensembles $\{A_x\}_{x \in X}$ and $\{B_x\}_{x \in X}$ are **computationally indistinguishable** if for any non-uniform efficient distinguisher D,*

$$\left| \Pr\left[D(A_x) = 1 \right] - \Pr\left[D(B_x) = 1 \right] \right| \le \varepsilon(|x|)$$

where $\varepsilon(\cdot)$ is a *negligible* function.**

*by efficient, we mean probabilistic polytime (PPT) algorithm, and by non-uniform, we mean that the algorithm can depend on the length of x.

**by negligible, we mean that for all c > 0, asymptotically, $\epsilon(n) < o(n^{-c})$.

FORMALITIES

Definition. A protocol (P, V) is a **zero-knowledge proof protocol** for the NP relation R (with corresponding language L), if it satisfies:

- I. **completeness:** if $x \in L$, and if both players follow the protocol, the verifier always accepts
- II. **soundness:** for every malicious and computationally unbounded prover P^* , if $x \notin L$, the verifier accepts with negligible probability $\varepsilon(|x|)$
- III. **zero-knowledge:** for any malicious PPT verifier V^* , there is a PPT simulator M^* , such that the view of V^* is computationally indistinguishable from the output distribution $M^*(x)$.

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The verifier V^* is told x, and then interacts with Alice. All new information is encapsulated in a string: $View_{V^*}(x,w)$ the collection of messages and random bits that V^* saw through the interaction.



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World II

In the simulation world, a random string: $M^*(x)$ is generated from the input x. There is no access to Alice.

Remark. What does it mean for the view of V^* to be computationally indistinguishable from the output distribution $M^*(x)$?

This means that it is not possible to computationally determine which world we're actually in:

$$\left| \Pr\left[D(\mathbf{View}_{V^*}(x, w)) = 1 \right] - \Pr\left[D(M^*(x)) = 1 \right] \right| \le \delta(|x|)$$

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It follows that Bob will not learn anything about *w* that he can efficiently recover even after interacting with Alice.

FORMALITIES: MPC PRIMITIVES

Definition An *n*-party MPC protocol Π_f computes *f* with *t*-privacy if no (informal). matter how a subset of *t* corrupted players collude, they can gain no additional information beyond their shared secrets.

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Remark. We can have different versions of *t*-privacy:

- perfect *t*-privacy: same distribution
- statistical *t*-privacy: statistical indistinguishability
- computational *t*-privacy: computational indistinguishability

FORMALITIES: MPC PRIMITIVES

Definition. Let $1 \le t \le n$. An MPC protocol \prod_f realizes f with **perfect t-privacy** if there is a PPT simulator SIM such that for any input x, w_1, \cdots, w_n , and for any set of corrupted players $T \subseteq [n]$ of size t, the joint view of those T players is distributed identically to: $SIM(T, x, (w_i)_{i \in T}, f_T(x, w_1, \cdots, w_n)).$

Similar definitions for **statistical** *t*-**privacy** and **computational** *t*-**privacy**, with respect to a security parameter *k*.

Zero-knowledge protocol Π_R in the commitment-hybrid model

- 1. The prover picks at random $w_1, \ldots, w_n \in \{0, 1\}^m$ whose exclusive-or equals the witness w. She emulates "in her head" the execution of Π_f on input (x, w_1, \ldots, w_n) (this involves choosing randomness for the *n* players and running the protocol). Based on this execution, the prover prepares the views V_1, \ldots, V_n of the *n* players; she separately commits to each of these *n* views.
- 2. Verifier picks at random distinct player indices $i, j \in [n]$ and sends them to the prover.
- 3. Prover "opens" the commitments corresponding to the two views V_i, V_j .
- 4. Verifier accepts if and only if:
 - (a) the prover indeed successfully opened the two requested views,
 - (b) the outputs of both P_i and P_j (which are determined by their views) are 1, and
 - (c) the two opened views are consistent with each other (with respect to x and Π_f , see Definition 2.2).

Theorem. Let Π_f be a correct and computational 2-private MPC protocol. Then Π_R from the previous slide is a zero-knowledge proof protocol for the NP-relation *R* with soundness error $\varepsilon \le 1 - n^{-2}$.

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 Proof.
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 SOUNDNESS. Follows from "Hat Lemma".

 ZERO-KNOWLEDGE. Construct the following simulation M*(x).

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World II

- 1. M^* uses the same randomness that V^* uses to choose $1 \le i \le j \le n$.
- M* uniformly at random selects w_i and w_j and runs: SIM(T={i,j}, x, (w_i, w_j), 1) to generate simulated views for i, j.

Theorem. Let Π_f be a correct and computational 2-private MPC protocol. Then Π_R from the previous slide is a zero-knowledge proof protocol for the NP-relation *R* with soundness error $\varepsilon \le 1 - n^{-2}$.

Proof.COMPLETENESS. Follows from correctness of ΠSOUNDNESS. Follows from "Hat Lemma".ZERO-KNOWLEDGE. Follows from 2-privacy of SIM.

Question. What are issues with the current description of the ZK protocol?

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- Issue 1.Soundness error 1 n-2 is not great.Potential solution? Repetitions of protocol.
 - Do multiple rounds reveal information?
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- **Issue 2.** Assumption of ideal primitives for MPC and COM.

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Proof.COMPLETENESS + SOUNDNESS. Follows from single round.ZERO-KNOWLEDGE. Need indistinguishability from repetitions.

Definition. Two probability ensembles $\{A_x\}_{x \in X}$ and $\{B_x\}_{x \in X}$ are **indistinguishable by polynomial-time sampling** if for any non-uniform efficient distinguisher *D* and m = p(|x|),

$$\left| \Pr\left[D(A_x^{(1)}, \dots, A_x^{(m)}) = 1 \right] - \Pr\left[D(B_x^{(1)}, \dots, B_x^{(m)}) = 1 \right] \right| < \varepsilon(|x|)$$

where $\varepsilon(\cdot)$ is a *negligible* function and p is a polynomial.

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$$A^1, A^2, \dots, A^{m-1}, A^m$$

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- **2**. **A**¹, **A**², ... , **A**^{m-1}, **B**^m
- 3. A¹, A², ..., B^{m-1}, B^m

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- **Question.** What is the probability that *D*' distinguishes between neighboring hybrid sequences?
- Idea.The sum of the probability that D' distinguishes neighboring
hybrid sequences telescopes to the probability that D'
distinguishes original two sequences.

- **Theorem.** Two probability ensembles $\{A_x\}$ and $\{B_x\}$ are computationally indistinguishable if and only if they are indistinguishable by polynomial-time sampling.
- **Question.** What is the probability that *D*' distinguishes between neighboring hybrid sequences?
- Idea.The sum of the probability that D' distinguishes neighboring
hybrid sequences telescopes to the probability that D'
distinguishes original two sequences.
- Ergo.D' can distinguish one of these m neighboring probabilities with
probability greater than $\varepsilon(|x|)/m$.
Use to construct single distinguisher!

REMARKS ON EXTENSIONS

If you're interested, the rest of paper goes into:

- 1. More efficient technique using *t*-robustness assumptions (allows verifier to open more than two safes).
- 2. Incorporation of imperfect MPC and commitment protocols into security analysis.
- 3. More on efficiency and coin-flipping.

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