

Zero-Knowledge in EasyCrypt

Denis Firsov^{1,2} and Dominique Unruh³

¹Guardtime

²Tallinn University of Technology

³Tartu University

July 10, 2023

Main Goal

Implement a framework in EasyCrypt theorem prover which formally defines notions associated with sigma-protocols and provides generic lemmas which capture common patterns in proofs.

Main Contributions: Quick Overview

- Formal definitions: completeness, special soundness, soundness, proof-of-knowledge, zero-knowledge.
- Generic derivations (in computational and information-theoretical setting):
 - Proof-of-knowledge from special soundness.
 - Soundness from proof-of-knowledge.
 - Zero-knowledge from one-shot simulator.
 - Sequential composition.
- Use cases: Fiat-Shamir, Schnorr, and Blum protocols.
- Stepping stone for end-to-end verified and executable sigma-protocols.

Background: EasyCrypt

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 - a probabilistic, relational Hoare logic (pRHL);
 - a probabilistic Hoare logic (pHL);
 - an ordinary (possibilistic) Hoare logic (HL);
 - an ambient higher-order logic for proving general mathematical facts and connecting judgments in the other logics.

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 - an ordinary (possibilistic) Hoare logic (HL);
 - an ambient higher-order logic for proving general mathematical facts and connecting judgments in the other logics.
- Also, we showed that EasyCrypt programs could be “reflected” into their probabilistic semantics to carry out proofs which rely on more advanced mathematical facts.

Notation

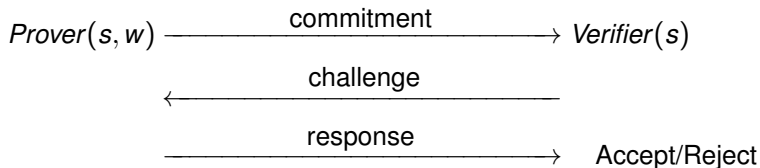
We write

$$\Pr[r \leftarrow X.f(i) @m : P]$$

to denote probability of event P after executing procedure f with argument i of module X at initial state m .

Background: Sigma Protocols

Every sigma protocol is designed to work with a specific formal NP-language. The language is induced by a relation between statements and witnesses. Then a protocol for the relation must allow the prover to convince the verifier that the prover knows a witness for some given statement without revealing anything else about itself.



Sigma protocols: Properties

- *Completeness* ensures the correct operation of the protocol if both prover and verifier follow the protocol honestly.
- *Soundness* ensures that for “wrong” statements (i.e., with no witness) a prover can convince the verifier with only small probability.
- *Proof-of-knowledge* guarantees that any prover that successfully convinces the verifier actually knows a witness (and not only abstractly that it exists).
- *Zero-knowledge* establishes that any cheating verifier cannot learn anything about the witness when running the protocol.

Sigma Protocols: Basic Parameters

```
abstract theory GenericProtocol.  
  
  type statement.  
  type witness.  
  
  type relation = statement → witness → bool.  
  
  op in_language (R:relation) statement: bool  
    = ∃ witness, R statement witness.  
  
  op completeness_relation      : relation.  
  op soundness_relation         : relation.  
  op zk_relation                 : relation.
```

Sigma Protocols: Basic Parameters

```
...
type commitment.
type response.
type challenge.

type transcript = commitment × challenge × response.

op verify_transcript: statement → transcript → bool.

op challenge_set: challenge list.

...
end GenericProtocol.
```

Completeness

Completeness ensures the correct operation of the protocol if both prover and verifier follow the protocol **honestly**.

Completeness: Honest Prover

```
module type HonestProver = {  
  proc commitment (s:statement,w:witness) : commitment  
  proc response(ch:challenge) : response  
}.
```

Completeness: Honest Verifier

```
module type HonestVerifier = {  
  proc challenge(s:statement,c:commitment) : challenge  
  proc verify(r:response) : bool  
}.
```

Completeness: Game

In EasyCrypt we define completeness module to capture the interaction between honest parties:

```
module Completeness (P: HonestProver, V: HonestVerifier) = {  
  
  proc run(s:statement, w:witness) = {  
    var commit, challenge, response, accept;  
    commit    <@ P.commitment(s,w);  
    challenge <@ V.challenge(s,commit);  
    response  <@ P.response(challenge);  
    accept    <@ V.verify(response);  
    return accept;  
  }  
}.  
}
```

Completeness: Property

The module for default honest verifier **HV** is derived automatically. The user must specify honest prover **HP** and a completeness lower-bound δ :

`op` δ : real.

`lemma` statistical_completeness s w m: completeness_relation s w

\Rightarrow `Pr`[out \leftarrow Completeness (HP, HV) .run (s, w) @m: out] $\geq \delta$.

“One-round” completeness must be proved manually.

Completeness: Sequential composition

One-round completeness implies completeness for sequential composition *generically*. Below, the module **CompletenessAmp** runs honest-interaction sequentially n -times.

lemma `completeness_seq m s w n: completeness_relation s w \wedge 1 \leq n`

\Rightarrow `Pr[out \leftarrow CompletenessAmp(HP, HV).run(s, w, n)@m: out] \geq δ^n .`

Rewinding

Definition

The module A is rewindable if

- 1 There exists an injective mapping f from the type \mathcal{G}_A to some parameter type $sbits$.
- 2 The module A must have a terminating procedure $getState$, so that the execution of $A.getState()$ in state \mathbf{m} must return the value $f(\mathcal{G}_A^{\mathbf{m}})$ without changing the state.

$$\Pr [r \leftarrow A.getState() @ \mathbf{m} : \mathcal{G}_A^{\text{fin}} = \mathcal{G}_A^{\mathbf{m}} \wedge r = f(\mathcal{G}_A^{\mathbf{m}})] = 1.$$

- 3 The module A must have a terminating procedure $setState$, so that whenever it gets an argument $x : sbits$ and sets $\mathcal{G}_A^{\mathbf{m}}$ to $f^{-1}(x)$ if $f^{-1}(x)$ is defined. Formally, let g be of type \mathcal{G}_A then

$$\Pr [r \leftarrow A.setState(f g) @ \mathbf{m} : \mathcal{G}_A^{\text{fin}} = g] = 1.$$

Zero-Knowledge

Zero-knowledge establishes that any **malicious rewindable** verifier cannot learn anything about the witness when running the protocol.

Zero-Knowledge: Rewindable Malicious Verifier

```
module type RewMaliciousVerifier = {  
  
  proc challenge(s:statement, c:commitment): challenge  
  proc summitup (r:response)   : summary  
  
  proc getState()              : sbits  
  proc setState(b:sbits)      : unit  
  
}.
```

Zero-Knowledge: Distinguisher

```
module type ZKDistinguisher = {  
    proc guess(s:statement,w:witness,sum:summary) : bool  
}.  
.
```

Zero-Knowledge: Real Experiment

```
module ZKReal(P: HonestProver,V: MaliciousVerifier,D: ZKDistinguisher)={  
  
  proc run(s:statement, w:witness) = {  
    var commit, challenge, response, summary, guess;  
  
    commit    <@ P.commitment(s,w);  
    challenge <@ V.challenge(s,commit);  
    response  <@ P.response(challenge);  
    summary   <@ V.summitup(s,response);  
  
    guess     <@ D.guess(s,w,summary);  
    return guess;  
  }  
}.
```

Zero-Knowledge: Simulator for Ideal Game

```
module type Simulator(V: RewMaliciousVerifier) = {  
  proc simulate(s: statement) : summary  
}.  

```

Zero-Knowledge: Ideal Experiment

```
module ZKIdeal(S: Simulator, V: RewMaliciousVerifier, D: ZKDistinguisher) = {  
  proc run(s: statement, w: witness) = {  
    var summary, guess;  
  
    summary <@ S(V).simulate(s);  
    guess <@ D.guess(s, w, summary);  
  
    return guess;  
  }  
}.
```


Zero-Knowledge: Desired Property

There must exist an efficient simulator **Sim** so that for any rewindable malicious verifier **V**, and distinguisher **D** the absolute difference between real and ideal games is bounded from above by ϵ :

```
op  $\epsilon$ : real.
```

```
lemma statistical_zk s w m: zk_relation s w  $\Rightarrow$   
  let real_prob = Pr[out  $\leftarrow$  ZKReal(HP,V,D).run(s,w)@m: out] in  
  let ideal_prob = Pr[out  $\leftarrow$  ZKIdeal(Sim,V,D).run(s,w)@m: out] in  
  
  |ideal_prob - real_prob|  $\leq \epsilon$ .
```

Zero-Knowledge: Direct proofs are hard!

Proving zero-knowledge directly could be challenging. Alternative common strategy is to derive zero-knowledge from “one-shot” simulator.

Zero-Knowledge: One-Shot Simulator

- One-shot simulator **Sim1** is a simulator which in addition to summary returns a “success-event”:

```
module type Simulator1(V: RewMaliciousVerifier) = {  
  proc run(s: statement) : bool × summary  
}.
```

- We ask for the lower-bound σ on that “success-event”

```
op  $\sigma$  : real.
```

```
lemma sim1_lower_bound stat m:  
  Pr[ (succ, _)  $\leftarrow$  Sim1(V).run(stat)@m: succ ]  $\geq \sigma$ .
```

- We also ask simulator to rewind itself and the malicious verifier in case it was not successful:

```
lemma rewind_sim istate m: (glob Sim1(V)) = istate  
  Pr[ (succ, _)  $\leftarrow$  Sim1(V).run(s)@m:  
    !succ  $\Rightarrow$  (glob Sim1(V)) = istate ] = 1.
```

Zero-Knowledge: One-Shot Simulator

The absolute difference between success-probabilities of the real game conditioned on the success-event and the ideal game must be bounded from above by ϵ :

```
lemma sim1_zk_cond3 s w m: zk_relation s w  $\Rightarrow$   
  let sim1_real  
    = Pr[(succ, out)  $\leftarrow$  ZKReal'(HP,V,D).run(s,w)@m: succ  $\wedge$  out] in  
  let sim1_ideal = Pr[out  $\leftarrow$  ZKIdeal(Sim1,V,D).main(s,w)@m: out] in  
  let succ_event = Pr[(succ, _)  $\leftarrow$  Sim1(V).run(s)@m: succ] in
```

$$|\text{sim1_real} / \text{succ_event} - \text{sim1_ideal}| \leq \epsilon.$$

Zero-Knowledge: Many-Shot Simulator for One-Round ZK

Given such one-shot simulator we define simulator **SimN** which runs one-shot simulator until it succeeds, but at most **N** times. Then we generically conclude the following statistical zero-knowledge:

```
lemma statistical_zk s w m: zk_relation s w  $\Rightarrow$   
  let real_prob = Pr[out  $\leftarrow$  ZKReal(HP, V, D).run(s, w)@m: out] in  
  let ideal_prob = Pr[out  $\leftarrow$  ZKIdeal(SimN, V, D).run(s, w)@m: out] in  
  
  |ideal_prob - real_prob|  $\leq$   $\epsilon + 2 \cdot (1 - \sigma)^N$ .
```

Zero-Knowledge: Sequential Composition

From one-round zero-knowledge we can conclude multiple-round zero-knowledge generically!

Zero-Knowledge: Sequential Composition

We define “sequentially” composed “real” experiment:

```
module ZKRealAmp(P:HonestProver,V:MaliciousVerifier,D:ZKDistinguisher)={
  proc run(s: statement, w: witness) = {
    var commit, challenge, response, summary, guess,i;
    i ← 0;

    while(i < n){
      commit    <@ P.commitment(s,w);
      challenge <@ V.challenge(s,commit);
      response  <@ P.response(challenge);
      summary   <@ V.summitup(response);
      i ← i + 1;
    }

    guess <@ D.guess(s,w,summary);
    return guess;
  }
}.
```

Zero-Knowledge: Sequential Composition

Ideal game for sequentially composed ZK does not change.

Zero-Knowledge: Multiple-Run Simulator

Generic transformation of one-run to multiple-run simulator:

```
module SimAmp(S: Simulator, V: RewMaliciousVerifier) = {
  proc simulate(s: statement) = {
    var summary, i;
    i ← 0;

    while(i < n) {
      summary <@ S(V).simulate(s);
      i ← i + 1;
    }

    return summary;
  }
}.
```

Zero-Knowledge: Sequential Composition Generically

If **Sim** is δ -one-run simulator then **SimAmp(Sim)** is a $n\delta$ -multiple-run simulator for sequentially composed ZK:

```
lemma zk_seq s w m:  
  let ideal_prob = Pr[out  $\leftarrow$  ZKIdeal(SimAmp(Sim), V, D).run(s, w)@m: out] in  
  let real_prob  = Pr[out  $\leftarrow$  ZKRealAmp(P, V, D).run(s, w)@m: out] in  
  
  |ideal_prob - real_prob|  $\leq$  n  $\cdot$   $\delta$ .
```

More generic derivations

- Proof-of-knowledge from special soundness.
- Soundness from proof-of-knowledge.

Use cases

- Schnorr protocol (discrete logarithm).
- Fiat-Shamir protocol (quadratic residue).
- Blum protocol (Hamiltonian cycles, NP-complete).

Fiat-Shamir protocol

- Completeness + sequential composition (50 lines of code);
- Special Soundness (60 lines of code);
- Proof-of-Knowledge (40 lines of code);
- Soundness + sequential composition (30 lines of code);
- One-Shot Simulator (200 lines of code).
- Zero-Knowledge + sequential composition (50 lines of code).

Conclusions

- It is relatively simple to instantiate and derive properties of sigma-protocols in our EasyCrypt framework.
- The downside of EasyCrypt formalizations is that the resulting protocols are not executable.
- In EasyCrypt formalizations are usually done at the very high-level of abstraction.
- For example, protocols are usually developed in context of abstract groups, particular distributions, etc.
- The naive compilation from high-level to low-level is not guaranteed to preserve cryptographic properties like zero-knowledge.

Work in progress: Sigma Protocols in Jasmin

- In the further work we implement sigma-protocols in assembly via Jasmin toolchain.
- Jasmin is a low-level programming language for high-assurance and high-speed cryptography.
- Jasmin programs can be extracted to EasyCrypt to address functional correctness, cryptographic security, or security against timing attacks.
- We derive properties for the sigma protocols in Jasmin by carrying them over from the our ZK framework.

Thank you!