

# An Algebraic Framework for Updatable and Universal (zk)SNARKs

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**Carla Ràfols and *Arantxa Zapico***

**4th ZKProof Workshop**



**Universitat  
Pompeu Fabra  
Barcelona**



# Pairing-Based (zk)SNARKs

*State of the art*

Interactive Proof-Systems [GMR89]

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Pinnocchio [PGHR13] → ZeroCash

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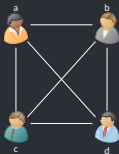
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Multiparty Computation (Zcash Ceremony)

One ceremony per circuit !!!

# Updatable and Universal SNARKs

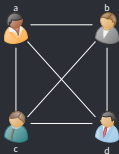
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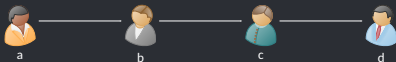


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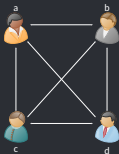


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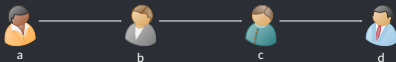


# Updatable and Universal SNARKs

- Multiparty Computation Model:



- Updatable Model:



Circuit Specific vs. Universal

# Motivation

*Ishai's wisdom*

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*"This calls for a modular approach that allows for an easier navigation in the huge design space. A higher level of modularity and abstraction is useful (...)"*

# Updatable and Universal (zk)SNARKs

## *Common Design Principle*



# Updatable and Universal (zk)SNARKs

*Common Design Principle*

AHP/  
PHP/  
...

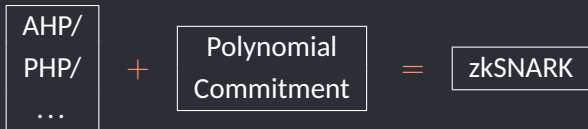
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## *Common Design Principle*



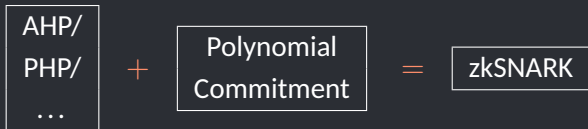
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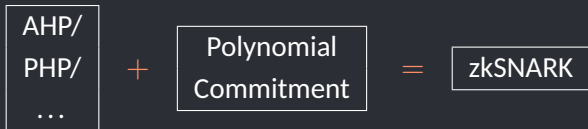
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Holographic:

# Updatable and Universal (zk)SNARKs

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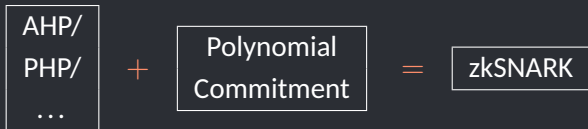


### Holographic:

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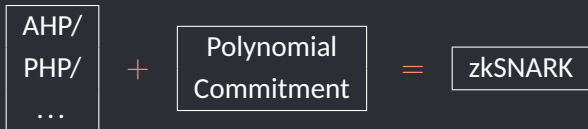


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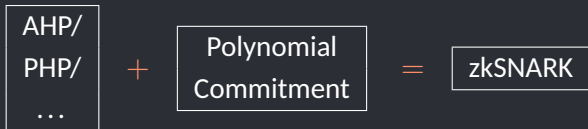


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Sonic[MBKM19] — Plonk[GWC19] — Marlin[CHMMVW20] —  
Lunar[CFFQH20] — Claymore[SZ21]



# Motivation

*This proposal*

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¿Can we break down further the information theoretic component?

# Organization

- Break Circuit Satisfiability in three main algebraic components:

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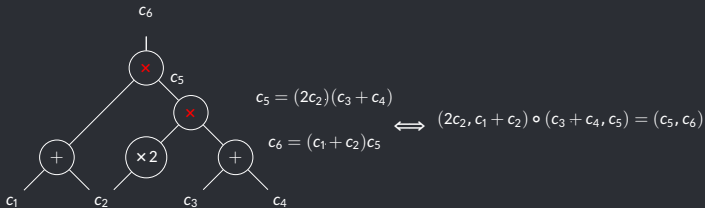
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- Definition of Verifiable Subspace Sampling

# Groth16: Overview Example

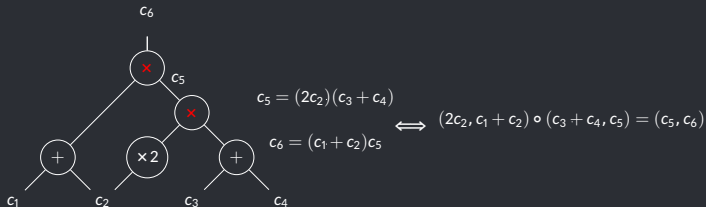
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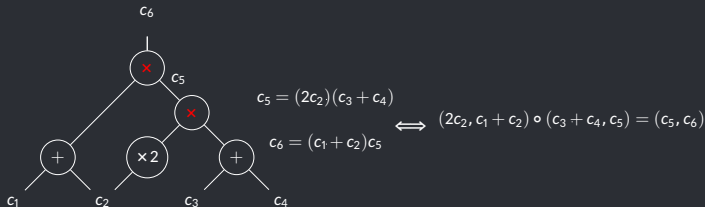
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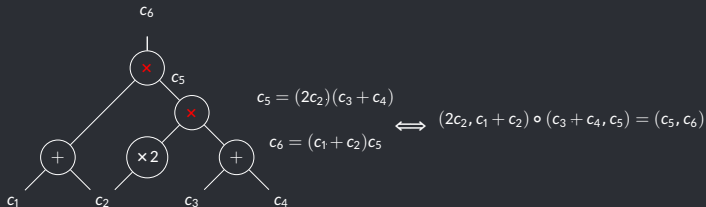
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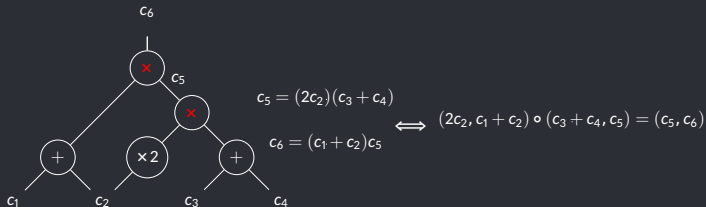
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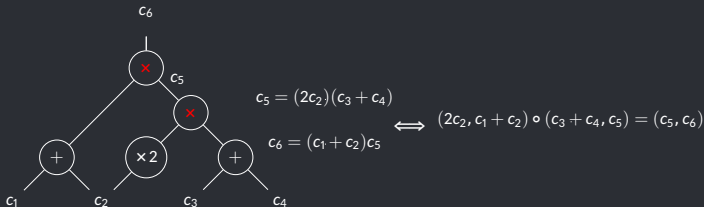
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# Main Tool: “Compressed” Linear Algebra

## *Hadamard Product*

---

<sup>1</sup>Completeness, soundness:  $\mathcal{R}$  arbitrary set, efficiency:  $\mathcal{R}$  multiplicative subgroup.

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Let  $\mathcal{R} = \{r_1, \dots, r_m\} \subset \mathbb{F}_p^*$ .

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Linear Algebra World

$$\vec{y} = (y_1, \dots, y_m)$$

Polynomial World

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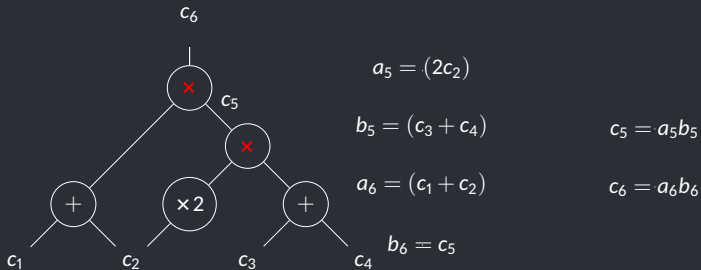
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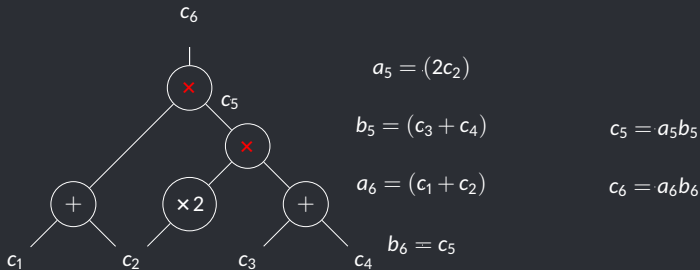
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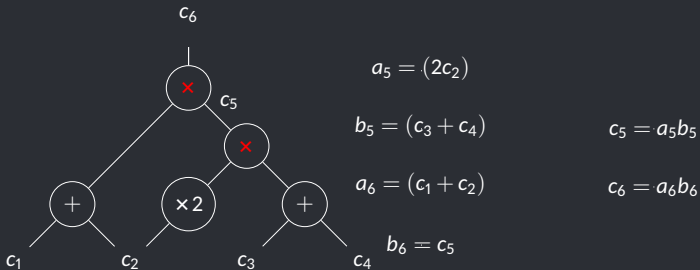


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$a_i$ 's left inputs,  $b_i$ 's right inputs,  $c_i$ 's outputs.

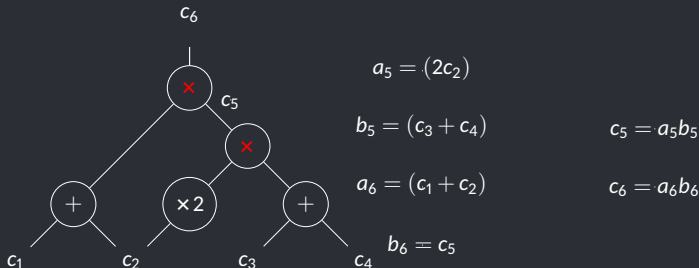
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**Hadamard product:**  $(a_5, a_6) \circ (b_5, b_6) = (c_5, c_6)$

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**Linear relations:**

$$\begin{pmatrix} a_5 \\ a_6 \end{pmatrix} = \begin{pmatrix} 0 & 2 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \end{pmatrix} \vec{c}, \quad \begin{pmatrix} b_5 \\ b_6 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \vec{c},$$



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Number of linear constraints:  $2 \times \text{\#mult. gates}$

# “Compressed” Linear Algebra

## Inner Product

- Define  $\mathbf{W} = \begin{pmatrix} \mathbf{I} & \mathbf{0} & -\mathbf{F} \\ \mathbf{0} & \mathbf{I} & -\mathbf{G} \end{pmatrix}$ , check that  $\mathbf{W} \cdot (\vec{a}, \vec{b}, \vec{c}) = \vec{0}$

---

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$\{\vec{w}_i(X)\}_{i=1}^m$ can be computed by the indexer, but $2m$ inner products	

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- Sample a random vector  $\vec{d}$  in the rowspace of the matrix
- Check **one** inner product  $\vec{d} \cdot (\vec{a}, \vec{b}, \vec{c}) = 0$ .

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$$\vec{d} \left\{ \text{Samp}(x)^T \begin{pmatrix} \mathbf{I} & \mathbf{0} & -\mathbf{F} \\ \mathbf{0} & \mathbf{I} & -\mathbf{G} \end{pmatrix} \begin{pmatrix} \vec{a} \\ \vec{b} \\ \vec{c} \end{pmatrix} = \vec{0} \right.$$

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- Sample a random vector  $\vec{d}$  in the rowspace of the matrix **from the function  $\text{Samp}(X)$  and some challenge  $x$ .**

# Verifiable Subspace Sampling

## *Algebraic Intuition*

$$\vec{d} \left\{ \text{Samp}(x)^T \begin{pmatrix} \mathbf{I} & \mathbf{0} & -\mathbf{F} \\ \mathbf{0} & \mathbf{I} & -\mathbf{G} \end{pmatrix} \begin{pmatrix} \vec{a} \\ \vec{b} \\ \vec{c} \end{pmatrix} = \vec{0} \right.$$

- Sample a random vector  $\vec{d}$  in the rowspace of the matrix **from the function  $\text{Samp}(X)$  and some challenge  $x$** .
- Check **one** inner product  $\vec{d} \cdot (\vec{a}, \vec{b}, \vec{c}) = 0$ .

## Verifiable Subspace Sampling

*In the polynomial world*

$$D(X) = (\text{Samp}(x)^T \mathbf{W}) \vec{\lambda}(X) = \vec{d} \cdot \vec{\lambda}(X)$$

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## Who samples $x$ ?

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- *Prover*: yeah, sure...
- *Verifier*: Who evaluates  $P(X, Y) = (\text{Samp}(\vec{Y})^\top \mathbf{W}) \vec{\lambda}(X)$  in  $Y = x$ ?

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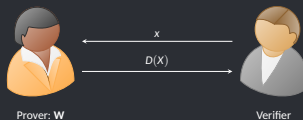
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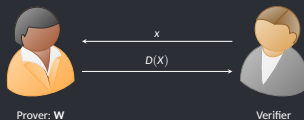
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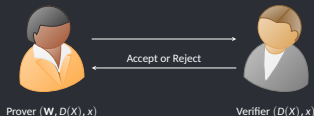
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- *Decision phase:* Verifier accepts only if  $D(x)$  encodes vector in the row space of  $\mathbf{W}$  sampled according to  $x$ .



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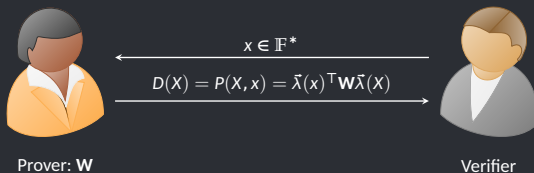
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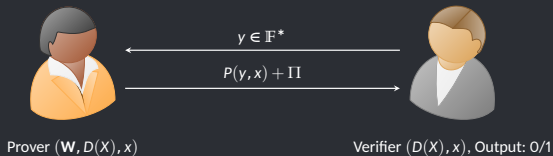
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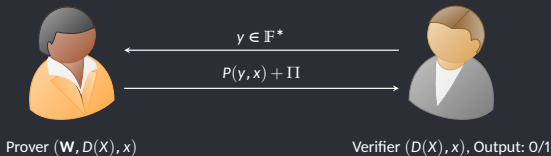
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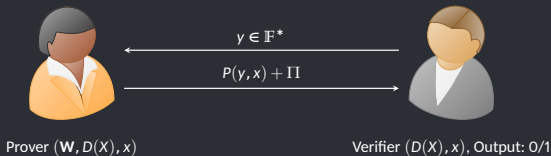
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- **Decision phase:** Verifier accepts only if  $D(X)$  encodes  $\text{Samp}(\vec{x})^\top \mathbf{W}$ .
- $\Pi$  is a **proof** that  $P(y, x)$  is correctly evaluated (Signature of Correct Computation of Sonic).

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Thank you!<sup>3</sup>

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<sup>3</sup>I am looking for a research stay/internship :) [arantxa.zapico@upf.edu](mailto:arantxa.zapico@upf.edu)