# QAP, PCP, POE

Oct 1, 2024

Distributed Lab

### Plan

Recap

Quadratic Arithmetic Program

Probabilistically Checkable Proofs

QAP as a Linear PCP

RECAP •0000000000

# Recap

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- **Succinctness** the proof size and verification time is relatively small to the computation size and typically does not depend on the size of the data or statement.
- **Non-interactiveness** to produce the proof, the prover does not need any interaction with the verifier.
- **Zero-Knowledge** the verifier learns nothing about the data used to produce the proof, despite knowing that this data resolves the given problem and that the prover possesses it.

We can do that in a way like the computer does it - boolean circuits.

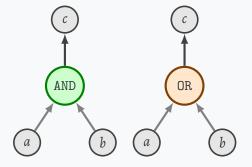


Figure: Boolean AND and OR Gates

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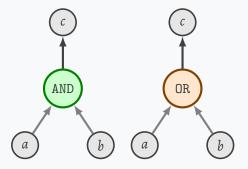


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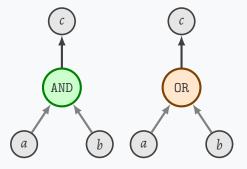


Figure: Boolean AND and OR Gates

> 100000 gates just for SHA256... But nothing stops us from using something more powerful instead of boolean values, gates.

Similar to Boolean Circuits, the **Arithmetic circuits** consist of gates and wires.

- Wires: elements of some finite field  $\mathbb{F}_p$ .
- Gates: addition  $(\oplus)$  and multiplication  $(\odot)$  corresponding to the field.

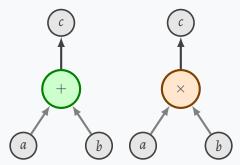


Figure: Addition and Multiplication Gates

### Example

How can we translate if statements?

```
def example (a: bool, b: F, c: F) -> F:
    if a:
        return b * c
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Corresponding equations for the circuit are:

$$r_1 = b \times c$$
,  $r_3 = 1 - a$ ,  $r_5 = r_3 \times r_2$   
 $r_2 = b + c$ ,  $r_4 = a \times r_1$ ,  $r = r_4 + r_5$ 

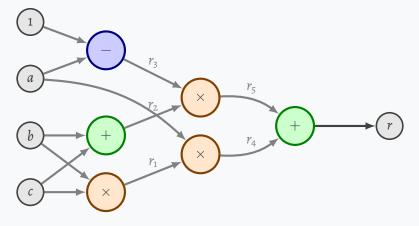


Figure: Example of a circuit evaluating the if statement logic.

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Each **constraint** in the Rank-1 Constraint System must be in the form:

$$\langle \mathbf{a}, \mathbf{w} \rangle \times \langle \mathbf{b}, \mathbf{w} \rangle = \langle \mathbf{c}, \mathbf{w} \rangle$$

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Thus

$$\left(\sum_{i=1}^n a_i w_i\right) \times \left(\sum_{j=1}^n b_j w_j\right) = \sum_{k=1}^n c_k w_k$$

That is, actually, a quadratic equation with multiple variables.

#### Example

Consider the most basic circuit with one multiplication gate:

$$x_1 \times x_2 = r$$
. The witnes vector  $\mathbf{w} = (r, x_1, x_2)$ . So

$$w_2 \times w_3 = w_1$$

$$(o + w_2 + o) \times (o + o + w_3) = w_1 + o + o$$

$$(ow_1 + 1w_2 + ow_3) \times (ow_1 + ow_2 + 1w_3) = 1w_1 + ow_2 + ow_3$$

Therefore the coefficients vectors are:

$$\mathbf{a} = (0, 1, 0), \quad \mathbf{b} = (0, 0, 1), \quad \mathbf{c} = (1, 0, 0).$$

The general form of our constraint is:

$$(a_1w_1 + a_2w_2 + a_3w_3)(b_1w_1 + b_2w_2 + b_3w_3) = c_1w_1 + c_2w_2 + c_3w_3$$

$$r = x_1 \times (x_2 \times x_3) + (1 - x_1) \times (x_2 + x_3)$$

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Thus, the next constraints can be build:

$$x_1 \times x_1 = x_1$$
 (binary check) (1)

$$x_2 \times x_3 = \text{mult} \tag{2}$$

$$x_1 \times \text{mult} = \text{selectMult}$$
 (3)

$$(1-x_1) \times (x_2 + x_3) = r - \mathsf{selectMult} \tag{4}$$

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The witness vector:  $\mathbf{w} = (1, r, x_1, x_2, x_3, \text{mult, selectMult})$ .

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The coefficients vectors:

$$\mathbf{a}_1 = (0, 0, 1, 0, 0, 0, 0), \quad \mathbf{b}_1 = (0, 0, 1, 0, 0, 0, 0), \quad \mathbf{c}_1 = (0, 0, 1, 0, 0, 0, 0)$$

$$\mathbf{a}_2 = (\mathtt{0}, \mathtt{0}, \mathtt{0}, \mathtt{1}, \mathtt{0}, \mathtt{0}, \mathtt{0}), \quad \ \mathbf{b}_2 = (\mathtt{0}, \mathtt{0}, \mathtt{0}, \mathtt{0}, \mathtt{1}, \mathtt{0}, \mathtt{0}), \quad \mathbf{c}_2 = (\mathtt{0}, \mathtt{0}, \mathtt{0}, \mathtt{0}, \mathtt{0}, \mathtt{1}, \mathtt{0})$$

$$\mathbf{a}_3 = (0, 0, 1, 0, 0, 0, 0), \quad \mathbf{b}_3 = (0, 0, 0, 0, 0, 1, 0), \quad \mathbf{c}_3 = (0, 0, 0, 0, 0, 0, 1)$$

$$\mathbf{a}_4 = (1, 0, -1, 0, 0, 0, 0), \quad \mathbf{b}_4 = (0, 0, 0, 1, 1, 0, 0), \quad \mathbf{c}_4 = (0, 1, 0, 0, 0, 0, -1)$$

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- Although Rank-1 Constraint Systems provide a powerful method for representing computations, they are not succinct.
- We need to transform our computations into a form that is more convenient for proving statements about them.

$$a_1, a_2, \ldots, a_m, b_1, b_2, \ldots, b_m, c_1, c_2, \ldots, c_m,$$

We finished with:

$$a_1, a_2, \ldots, a_m, b_1, b_2, \ldots, b_m, c_1, c_2, \ldots, c_m,$$

Of course, they form corresponding matrices:

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \quad B = \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{bmatrix} \quad C = \begin{bmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{m1} & c_{m2} & \dots & c_{mn} \end{bmatrix}$$

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An example of a single "if" statement:

$$\begin{array}{l} \mathbf{a}_1 = (\mathtt{0},\mathtt{0},\mathtt{1},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0}) \\ \mathbf{a}_2 = (\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{1},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0}) \\ \mathbf{a}_3 = (\mathtt{0},\mathtt{0},\mathtt{1},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0}) \\ \mathbf{a}_4 = (\mathtt{1},\mathtt{0},\mathtt{-1},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0},\mathtt{0}) \end{array} \qquad A = \begin{bmatrix} \mathtt{0} & \mathtt{0} & \mathtt{1} & \mathtt{0} & \mathtt{0} & \mathtt{0} & \mathtt{0} \\ \mathtt{0} & \mathtt{0} & \mathtt{1} & \mathtt{0} & \mathtt{0} & \mathtt{0} & \mathtt{0} \\ \mathtt{0} & \mathtt{0} & \mathtt{1} & \mathtt{0} & \mathtt{0} & \mathtt{0} & \mathtt{0} \\ \mathtt{1} & \mathtt{0} & \mathtt{-1} & \mathtt{0} & \mathtt{0} & \mathtt{0} & \mathtt{0} \end{bmatrix}$$

We finished with:

$$a_1, a_2, \ldots, a_m, b_1, b_2, \ldots, b_m, c_1, c_2, \ldots, c_m,$$

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Pleeeeeenty of zeroes, doesn't it? And this is just one out of 3 matrices...

$$\mathbf{w} = (1, r, x_1, x_2, x_3, \text{mult, selectMult})$$

Let's take a closer look at the matrix columns:

$$\begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & -1 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Consider 4th constraint:  $(1 - x_1) \times (x_2 + x_3) = r$  – selectMult

So, every column is a mapping of constraint number to a coefficient for the witness element.

The previous witness vector:

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$$L(x) = \sum_{i=0}^{n} y_i \ell_i(x), \quad \ell_i(x) = \prod_{j=0, j \neq i}^{n} \frac{x - x_j}{x_i - x_j}.$$

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There are *n* columns and *m* constraints. So, it results in *n* polynomials such that:

$$A_j(i) = a_{i,j}, i \in \{1, 2, ..., m\}, j \in \{1, 2, ..., n\}$$

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The same is true for matrices B and C, with 3n polynomials in total, n for each of the coefficients matrices:

$$A_1(x), A_2(x), \ldots, A_n(x), B_1(x), B_2(x), \ldots, B_n(x), C_1(x), C_2(x), \ldots, C_n(x)$$

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#### Note

We could have assigned any unique index from F to each constraint (say,  $t_i$  for each  $i \in \{1, ..., m\}$ ) and interpolate through these points:

$$A_j(t_i) = a_{i,j}, i \in \{1, 2, \ldots, m\}, j \in \{1, 2, \ldots, n\}$$

# Example

Considering the witness vector **w** and matrix A from the previous example, for the variable  $x_1$ , the next set of points can be derived:

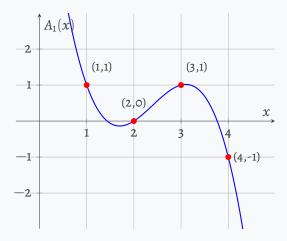
$$\{(1,1), (2,0), (3,1), (4,-1)\}$$

The Lagrange interpolation polynomial for this set of points:

$$\ell_1(x) = -\frac{(x-2)(x-3)(x-4)}{6}, \qquad \ell_2(x) = \frac{(x-1)(x-3)(x-4)}{2},$$
 
$$\ell_3(x) = -\frac{(x-1)(x-2)(x-4)}{2}, \qquad \ell_4(x) = \frac{(x-1)(x-2)(x-3)}{6}.$$

Thus, the polynomial is given by:

$$A_{x_1}(x) = 1 \cdot \ell_1(x) + 0 \cdot \ell_2(x) + 1 \cdot \ell_3(x) + (-1) \cdot \ell_4(x)$$
$$= -\frac{5}{6}x^3 + 6x^2 - \frac{79}{6}x + 9$$



**Illustration:** The Lagrange interpolation polynomial for points  $\{(1,1), (2,0), (3,1), (4,-1)\}$ visualized over R.

# Question

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Consider two polynomials p(x) and q(x):

$$p(x) = -\frac{1}{2}x^2 + \frac{3}{2}x,$$
  $q(x) = \frac{1}{3}x^3 - 2x^2 + \frac{8}{3}x + 1.$ 

With corresponding sets of points:

$$\{(0,0),(1,1),(2,1),(3,0)\}, \{(0,1),(1,2),(2,1),(3,0)\}$$

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The sum of these polynomials can be calculated as:

$$r(x) = \frac{1}{3}x^3 - 2\frac{1}{2}x^2 + 4\frac{1}{6}x + 1$$

The resulting polynomial r(x) corresponds to the set of points:

$$\{(0,1),(1,3),(2,2),(3,0)\}$$

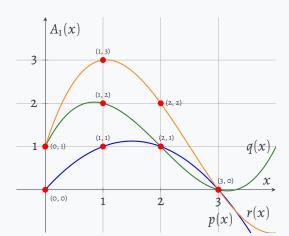


Figure: Addition of two polynomials

Now, using coefficients encoded with polynomials, we can build a constraint number  $X \in \{1, \dots m\}$  in the next way:

$$(w_1A_1(X) + w_2A_2(X) + \dots + w_nA_n(X)) \times \times (w_1B_1(X) + w_2B_2(X) + \dots + w_nB_n(X)) = = (w_1C_1(X) + w_2C_2(X) + \dots + w_nC_n(X))$$

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Or written more concisely:

$$\left(\sum_{i=1}^n w_i A_i(X)\right) \times \left(\sum_{i=1}^n w_i B_i(X)\right) = \left(\sum_{i=1}^n w_i C_i(X)\right)$$

$$\left(\sum_{i=1}^n w_i A_i(j)\right) \times \left(\sum_{i=1}^n w_i B_i(j)\right) = \left(\sum_{i=1}^n w_i C_i(j)\right) \ \forall j \in \{1, \ldots, m\}$$

Hold on, but why does it hold? Let us substitute any X = j into this equation:

$$\left(\sum_{i=1}^n w_i A_i(j)\right) \times \left(\sum_{i=1}^n w_i B_i(j)\right) = \left(\sum_{i=1}^n w_i C_i(j)\right) \ \forall j \in \{1, \ldots, m\}$$

Recall that we interpolated polynomials to have  $A_i(j) = a_{j,i}$ . Therefore, the equation above can be reduced to:

$$\left(\sum_{i=1}^n w_i a_{j,i}\right) \times \left(\sum_{i=1}^n w_i b_{j,i}\right) = \left(\sum_{i=1}^n w_i c_{j,i}\right) \ \forall j \in \{1,\ldots,m\}$$

Hold on, but why does it hold? Let us substitute any X = i into this equation:

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But hold on again! Notice that  $\sum_{i=1}^n w_i a_{j,i} = \langle \mathbf{w}, \mathbf{a}_i \rangle$  and therefore we have:

$$\langle \mathbf{w}, \mathbf{a}_j \rangle \times \langle \mathbf{w}, \mathbf{b}_j \rangle = \langle \mathbf{w}, \mathbf{c}_j \rangle \ \forall j \in \{1, \ldots, m\},$$

so we ended up with the initial *m* constraint equations!

Now let us define polynomials A(X), B(X), C(X) for easier notation:

$$A(X) = \sum_{i=1}^{n} w_i A_i(X), \quad B(X) = \sum_{i=1}^{n} w_i B_i(X), \quad C(X) = \sum_{i=1}^{n} w_i C_i(X)$$

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Therefore:

$$A(X) \times B(X) = C(X)$$

Now, we can define a polynomial M(X), that has zeros at all elements from the set  $\Omega = \{1, ..., m\}$ 

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It means, that M(X) can be devide by vanishing polynomial  $Z_{\Omega}(X)$ without a remainder!

$$Z_{\Omega}(X) = \prod_{i=1}^{m} (X - i), \qquad H(X) = \frac{M(X)}{Z_{\Omega}(X)}$$

# Definition (Quadratic Arithmetic Program)

Suppose that *m* R1CS constraints with a witness of size *n* are written in a form

$$A\mathbf{w} \odot B\mathbf{w} = C\mathbf{w}$$
,  $A$ ,  $B$ ,  $C \in \mathbb{F}^{m \times n}$ 

Then, the **Quadratic Arithmetic Program** consists of 3n polynomials  $A_1, \ldots, A_n, B_1, \ldots, B_n, C_1, \ldots, C_n$  such that:

$$A_j(i) = a_{i,j}, B_j(i) = b_{i,j}, C_j(i) = c_{i,j}, \forall i \in \{1, ..., m\} \forall j \in \{1, ..., n\}$$

Then,  $\mathbf{w} \in \mathbb{F}^n$  is a valid assignment for the given QAP and **target polynomial**  $Z(X) = \prod_{i=1}^m (X-i)$  if and only if there exists such a polynomial H(X) such that

$$\left(\sum_{i=1}^n w_i A_i(X)\right) \left(\sum_{i=1}^n w_i B_i(X)\right) - \left(\sum_{i=1}^n w_i C_i(X)\right) = Z(X)H(X)$$

# **Probabilistically Checkable Proofs**

PROBABILISTICALLY CHECKABLE PROOFS

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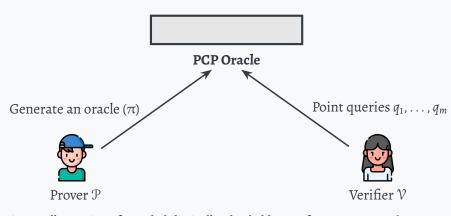


Figure: Illustration of a Probabilistically Checkable Proof (PCP) system. The prover  $\mathcal{P}$  generates a PCP oracle  $\pi$  that is queried by the verifier  $\mathcal{V}$  at specific points  $q_1, \ldots, q_m$ .

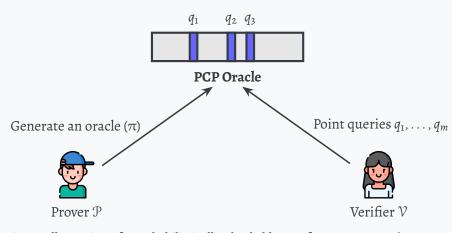


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Three main extensions of PCPs that are frequently used in SNARKs are:

- **IPCP** (**Interactive PCP**): The prover commits to the PCP oracle and then, based on the interaction between the prover and verifier, the verifier queries the oracle and decides whether to accept the proof.
- **IOP** (**Interactive Oracle Proof**): The prover and verifier interact and on each round, the prover commits to a new oracle. The verifier queries the oracle and decides whether to accept the proof.
- **LPCP** (**Linear PCP**): The prover commits to a linear function and the verifier queries the function at specific points.

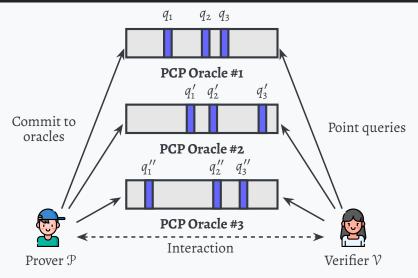


Figure: Illustration of an Interactive Oracle Proof (IOP). On each round i ( $1 \le i \le r$ ), V sends a message  $m_i$ , and P commits to a new oracle  $\pi_i$ , which V can query at  $\mathbf{q}_i = (q_{i,1}, \dots, q_{i,m})$ .

### Definition (Linear PCP)

A **Linear PCP** is a PCP where the prover commits to a linear function  $\pi = (\pi_1, \dots, \pi_k)$  and the verifier queries the function at specific points  $q_1, \dots, q_r$ . Then, the prover responds with the values of the function at these points:

$$\langle \pi_1, \mathbf{q}_1 \rangle$$
,  $\langle \pi_2, \mathbf{q}_2 \rangle$ , ...,  $\langle \pi_r, \mathbf{q}_r \rangle$ .

# Example (QAP as a Linear PCP)

Recall that key QAP equation is:

$$L(x) \times R(x) - O(x) = Z(x)H(x).$$

Now, consider the following **linear PCP for QAP**:

- 1. P commits to an extended witness w and coefficients  $\mathbf{h} = (h_1, \dots, h_n) \text{ of } H(x).$
- 2.  $\mathcal{V}$  samples  $\gamma \stackrel{R}{\leftarrow} \mathbb{F}$  and sends query  $\gamma = (\gamma, \gamma^2, \dots, \gamma^n)$  to  $\mathcal{P}$ .
- 3. P reveals the following values:

$$\begin{split} \pi_1 &\leftarrow \langle \mathbf{w}, \mathbf{L}(\gamma) \rangle, & \pi_2 &\leftarrow \langle \mathbf{w}, \mathbf{R}(\gamma) \rangle, \\ \pi_3 &\leftarrow \langle \mathbf{w}, \mathbf{O}(\gamma) \rangle, & \pi_4 &\leftarrow Z(\gamma) \cdot \langle \mathbf{h}, \gamma \rangle. \end{split}$$

4.  $\mathcal{V}$  checks whether  $\pi_1\pi_2-\pi_3=\pi_4$ .

### **Ouestion**

Why is it safe to use such a check?

The polynomials L(x), R(x) and O(x) are interpolated polynomials using |C| (number of gates) points, so:

$$\deg(L) \leqslant |C|$$
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If an adversary A does not know a valid witness  $\mathbf{w}$ , he can compute a polynomial  $M(x) \leftarrow \mathcal{A}(\cdot)$  that satisfies a verifier  $\mathcal{V}$ :

$$\Pr_{s \leftarrow \mathbb{F}} [\widetilde{M}(s) = M(s)] \leqslant \frac{2|C|}{|\mathbb{F}|}$$

Thanks for your attention!