

# The PCP theorem and the complexity of 2 prover games

A game  $G$  is described four finite nonempty sets  $\mathcal{X}, \mathcal{Y}, \mathcal{A}, \mathcal{B}$ , a probability distribution  $\mu : \mathcal{X} \times \mathcal{Y} \rightarrow \mathbb{R}_+$  (we assume that for all  $x, y$ ,  $\mu(x, y)$  can be represented by abitstring of length at most  $\lceil \log(|\mathcal{X}||\mathcal{Y}|) \rceil$ ) and a table  $V : \mathcal{X} \times \mathcal{Y} \times \mathcal{A} \times \mathcal{B} \rightarrow \{0, 1\}$ . To see it as the input of a computational problem which should represent  $G$  using a finite bitstring. One way to represent such a  $G$  is by the following string:

$$\text{repr}(G) := \text{bin}(|\mathcal{X}|) | \text{bin}(|\mathcal{Y}|) | \text{bin}(|\mathcal{A}|) | \text{bin}(|\mathcal{B}|) | \text{bin}_{\lceil \log(|\mathcal{X}||\mathcal{Y}|) \rceil}(\mu(x, y))_{x,y \in \mathcal{X}, \mathcal{Y}} | (V(x, y, a, b))_{x,y,a,b \in \mathcal{X}, \mathcal{Y}, \mathcal{A}, \mathcal{B}}$$

where  $\text{bin}(n)$  is the binary representation of the integer  $n$ ,  $\text{bin}_k(\alpha)$  for  $\alpha \in (0, 1)$  is the binary representation of  $\alpha$  truncated after  $k$  bits,  $|$  is a separator symbol. To represent the lists for  $\mu$  and  $V$ , we have implicitly chosen a fixed orders on  $\mathcal{X} \times \mathcal{Y}$  and  $\mathcal{X} \times \mathcal{Y} \times \mathcal{A} \times \mathcal{B}$  and the list is represented as a separated sequence of bit-strings in the corresponding order. Note that the size of the string representing  $G$  contains  $O(|\mathcal{X}||\mathcal{Y}||\mathcal{A}||\mathcal{B}|)$  symbols.

Given a game  $G$ , we can define its value

$$\text{val}(G) = \sup_{p,q} \sum_{x,y} \mu(x, y) \sum_{a,b} V(x, y, a, b) p(a|x) q(b|y),$$

where  $p, q$  are such that  $p(\cdot|x), q(\cdot|y)$  are probability distributions for every  $x, y$ . As the function is linear in  $p$  and  $q$ , we can restrict the optimization to  $p, q$  satisfying  $p(a|x), q(b|y) \in \{0, 1\}$ , i.e., deterministic strategies. We can then define the promise problem  $\rho$ -GAPGAMEVAL as follows: for any  $G$  as above if  $\text{val}(G) = 1$ , then  $\text{repr}(G)$  is a YES instance and if  $\text{val}(G) \leq \rho$ , then  $\text{repr}(G)$  is a NO instance

**Proposition 0.1.** *There exists a constant  $\rho < 1$  such that promise problem  $\rho$ -GAPGAMEVAL is NP-hard in the sense that for any  $L \in NP$ , there is a polynomial time function  $f$  such that if  $x \in L$ , then  $f(x)$  is a YES instance of  $\rho$ -GAPGAMEVAL and if  $x \notin L$ ,  $f(x)$  is a NO instance of  $\rho$ -GAPGAMEVAL.*

**Proof** We are going to use the NP-hardness of  $\rho$ -GAP3SAT (see the chapter on PCP theorem in the Arora-Barak book <https://theory.cs.princeton.edu/complexity/book.pdf>). An instance of GAP3SAT is given by a set of variables labeled by  $[n]$ , and a set of constraints labeled by  $[m]$ . A constraint  $i \in [m]$  contains three variables  $v_1(i), v_2(i), v_3(i) \in [n]$  and each variable appears with a negation or not we represent this with  $w_1(i), w_2(i), w_3(i) \in \{0, 1\}$ . For example, a constraint  $x_1 \vee \bar{x}_{10} \vee x_{12}$  is represented by  $v_1 = 1, v_2 = 10, v_3 = 12$  and  $w_1 = 0, w_2 = 1, w_3 = 0$ . The game we construct is as follows:  $\mathcal{X} = [n], \mathcal{Y} = [m], \mathcal{A} = \{0, 1\}, \mathcal{B} = \{0, 1\}^3$ . Then  $\mu(v, i) = \frac{1}{3m}$  if variable  $v \in \{v_1(i), v_2(i), v_3(i)\}$  and otherwise 0. Also we set  $V(v, i, a, (b_1, b_2, b_3)) = 1$  when  $b_1, b_2, b_3$  satisfies the constraint  $i$  (i.e.,  $\bar{b}_1^{w_1(i)} \vee \bar{b}_2^{w_2(i)} \vee \bar{b}_3^{w_3(i)} = 1$ , where the notation  $\bar{b}^w$  refers to  $b$  if  $w = 0$  and  $\bar{b}$  if  $w = 1$ ) and  $a = b_j$  where  $v = v_j(i)$ . And  $V$  is set to 0 otherwise. Note that all the operations take a time which is polynomial in  $n$  and  $m$  so this is a valid reduction.

Now assume that the instance of GAP3SAT is satisfiable. Then the game has a strategy than wins with probability 1: just take a satisfying assignment and both players answer according to this. Conversely, assume the game has a winning probability  $1 - \delta$ . Then let us construct an assignment of the variables. We may assume that the strategy achieving  $1 - \delta$  is deterministic. Thus, the first player's strategy is described by a function  $\sigma : [n] \rightarrow \{0, 1\}$  and we interpret this as an assignment to the variables. Then the probability of losing the game can be written as

$$\frac{1}{3m} \sum_{i \in [m]} \sum_{j=1}^3 \mathbf{1}_{\sigma(v_j(i)) \neq b_j(i) \text{ OR } \bar{b}_1^{w_1(i)} \vee \bar{b}_2^{w_2(i)} \vee \bar{b}_3^{w_3(i)} = 0}$$

We know that this quantity is  $\leq \delta$ . We are on the other hand interested in

$$\begin{aligned} \frac{1}{m} \sum_{i \in [m]} \mathbf{1}_{\frac{w_1(i)}{\sigma(v_1(i))} \sqrt{\frac{w_2(i)}{\sigma(v_2(i))}} \sqrt{\frac{w_3(i)}{\sigma(v_3(i))}} = 0} &\leq \frac{1}{m} \sum_{i \in [m]} \mathbf{1}_{\bar{b}_1(i)^{w_1(i)} \bar{b}_2(i)^{w_2(i)} \bar{b}_3(i)^{w_3(i)} = 0} \mathbf{1}_{\sigma(v_j(i)) = b_j(i) \forall j \in \{1,2,3\}} \\ &+ \frac{1}{m} \sum_{i \in [m]} \sum_{j=1}^3 \mathbf{1}_{\sigma(v_j(i)) \neq b_j(i)} \\ &\leq 6\delta. \end{aligned}$$

So the formula is  $(1 - 6\delta)$ -satisfiable and this concludes the proof of the converse.  $\square$

Note that we can even obtain the NP-hardness for any constant  $\rho < 1$ . This follows immediately from the parallel repetition theorem. In fact, we will give a reduction from  $\rho$ -GAPGAMEVAL to  $\epsilon$ -GAPGAMEVAL for any  $\epsilon > 0$ . Take an instance  $G$  for  $\rho$ -GAPGAMEVAL and then consider the game  $G'$  obtained by parallel repetition  $G$  a constant  $c(\rho, \epsilon)$  number of times. Then  $G'$  can be obtained in polynomial time from  $G$ , and if  $G$  had a value of 1, then so does  $G'$ , and if  $G$  had a value  $\leq \rho$ , then  $G'$  has a value  $\leq \epsilon$  if  $c(\rho, \epsilon)$  is chosen appropriately.