

possibly random *process* of (1) uniformly and independently choosing element  $x$  from set  $y$ , or (2) uniformly and independently drawing  $x$  according to distribution  $y$ , or (3) setting object  $x$  equal to object  $y$ , or (4) setting object  $x$  equal to the output of the (possibly probabilistic) algorithm  $y$  (in which case we specify also the input to  $y$ ). By  $\text{Prob}[R_1; \dots; R_n : E]$  we denote the probability of event  $E$ , after the ordered execution of possibly random processes  $R_1, \dots, R_n$ .

We define negligible functions as functions that tend to zero smaller than any inverse of a polynomial.

**Definition 1.** A function  $\delta$  is negligible if for all positive constants  $c$  there exists an integer  $n_c$  such that  $\delta(n) < n^{-c}$ , for all  $n \geq n_c$ .

Intuitively, events with a negligible probability should not be noticed by probabilistic polynomial-time algorithms when the input sizes are large enough. We now are ready to formally define one-way functions.

**Definition 2.** A function  $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$  is one-way if

1. there exists an efficient algorithm  $C$  that, on input  $x$ , returns  $f(x)$ ;
2. for any efficient algorithm  $A$ , the following probability is negligible in  $n$ :

$$\text{Prob}[x \leftarrow \{0, 1\}^n; y \leftarrow f(x); x' \leftarrow A(1^n, y) : f(x') = f(x)].$$

We also define collections of one-way functions.

**Definition 3.** A collection of functions  $F = \{f_n : n \in \mathcal{N}, f_n : \{0, 1\}^n \rightarrow \{0, 1\}^n\}$  is one-way if

1. there exists an efficient algorithm  $C$  that, on input  $n, x$ , returns  $f_n(x)$ , and if
2. for any efficient algorithm  $A$ , the following probability is negligible in  $n$ :

$$\text{Prob}[x \leftarrow \{0, 1\}^n; y \leftarrow f_n(x); x' \leftarrow A(1^n, y) : f_n(x') = f_n(x)].$$

It is possible to prove that one-way functions exist if and only if collections of one-way functions exist. We note that the definition of one-way function essentially implies that almost all inputs to the function produce an output that is hard to invert. A natural relaxation of this intuition is that only a large fraction of the inputs produce inputs that are hard to invert. These functions are called “weak one-way” and will be discussed later in greater detail.

We now recall the definition of “trapdoor” functions as one-way function with the additional property that there exists some information that allows its owner (and only her) to invert the function.

**Definition 4.** A trapdoor function  $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$  is a one-way function for which there exists an efficient algorithm  $E$  and a polynomial  $p$  such that, for any  $n$ , there exists a string  $t_n$  such that  $|t_n| \leq p(n)$  and for all  $x \in \{0, 1\}^*$ ,  $E(f(x), t_n) = x'$  and  $f(x) = f(x')$ .

### 2.3. Weak vs. Strong One-Way Functions

We now formally define “weak” one-way functions and will also refer to (previously defined) one-way functions as “strong” one-way functions. Informally, weak one-way function represent a relaxation of one-way function as they only require that no efficient adversary can invert the function for at least a noticeable fraction of the inputs.

**Definition 6.** *Let  $p$  be a polynomial. A function  $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$  is a  $p$ -weak one-way function if*

1. *there exists an efficient algorithm  $C$  that, on input  $x$ , returns  $f(x)$*
2. *for any efficient algorithm  $A$ , it holds that for all sufficiently large  $n$ ,*  

$$\text{Prob}[x \leftarrow \{0, 1\}^n; y \leftarrow f(x); x' \leftarrow A(1^n, y) : f(x') \neq f(x)] \geq 1/p(n).$$

Similarly as before, we can define a *collection of weak one-way functions*. The following theorem was first stated in the oral presentation of [52].

**Theorem 1.** *A weak one-way function exists if and only if a strong one-way function exists.*

As one-way functions are believed to represent a very minimal notion of cryptographic hardness, this theorem seems to suggest that cryptographic hardness can be amplified from a low (but sufficiently noticeable) level to a high (and sufficiently close to the maximum possible) level.

*Proof.* We start the proof by recalling the transformation from weak to strong one-way functions from [52]. Intuitively, the strong one-way function is the concatenation of sufficiently many application of the weak one-way function. This is reminiscent of analogue theorems in Information Theory; interestingly, as we will see, the proof of this theorem is significantly harder.

More formally, given a  $p$ -weak collection of one-way functions  $F = \{f_n : n \in \mathcal{N}\}$ , where  $f_n : \{0, 1\}^n \rightarrow \{0, 1\}^n$ , we define a collection  $G = \{g_m : m \in \mathcal{N}\}$ , where  $g_m : \{0, 1\}^m \rightarrow \{0, 1\}^m$ , for  $m = 2n^2p(n)$ , is defined as

$$g_m(x_1, \dots, x_{2np(n)}) = (f_n(x_1) \circ \dots \circ f_n(x_{2np(n)})).$$

We now prove that  $G$  is a collection of strong one-way functions. Assume (towards contradiction) that this is not the case. Then there exists an efficient adversary  $A$  and a polynomial  $q$  such that for infinitely many  $m$ , it holds that

$$\text{Prob}[x \leftarrow \{0, 1\}^m; y \leftarrow g_m(x); x' \leftarrow A(1^n, y) : g_m(x') = g_m(x)] \geq 1/q(m).$$

If we present an efficient adversary  $A'$  that, using  $A$ , can invert  $f_n$  with probability at least  $1 - 1/p(n)$  then we contradict the assumption that  $F$  is a collection of  $p$ -weak one-way functions. Consider the following algorithm  $A'$ .

*Input for Algorithm  $A'$ :*  $y \in \{0, 1\}^n$ , where  $y = f_n(x)$ , for a randomly chosen  $x$ .

*Instructions for Algorithm  $A'$ :*

1. repeat  $4n^2p(n)q(m)$  times:
  - for  $i = 1, \dots, 2np(n)$ 
    - randomly choose  $x_j \in \{0, 1\}^n$ , for  $j = 1, \dots, i-1, i+1, \dots, 2np(n)$
    - compute  $y_j = f_n(x_j)$ , for  $j = 1, \dots, i-1, i+1, \dots, m$
    - if  $A$  successfully inverts  $(y_1, \dots, y_{i-1}, y, y_{i+1}, \dots, y_{2np(n)})$  then
      - let  $(x_1, \dots, x_{2np(n)}) = A(y_1, \dots, y_{i-1}, y, y_{i+1}, \dots, y_{2np(n)})$
      - return:  $x_j$  and halt
2. return: ‘failure to invert’.

We define the subset  $\text{BAD} \subseteq \{0, 1\}^n$  of  $x$  such that the probability, over the randomness used by  $A'$ , that in a single iteration of its repeat loop  $A'$  returns  $f_n^{-1}(f_n(x))$  is less than  $1/4np(n)q(m)$ .

We now show that the probability, over the randomness used by  $A'$  and the random choice of  $x$ , that  $A'$  is not successful is ‘essentially’ the probability that  $x$  is BAD. More precisely, we define event  $e(A', x)$  as the event that  $A'$  does not invert  $y = f_n(x)$ , when  $x$  is randomly chosen, and  $A'$  is run on  $f_n(x)$ . Then we have that

$$\begin{aligned}
 \text{Prob}[e(A', x)] &= \text{Prob}[e(A', x) \mid x \in \text{BAD}] \cdot \text{Prob}[x \in \text{BAD}] \\
 &\quad + \text{Prob}[e(A', x) \mid x \notin \text{BAD}] \cdot \text{Prob}[x \notin \text{BAD}] \\
 &\leq 1 \cdot \text{Prob}[x \in \text{BAD}] + (1 - 1/4np(n)q(m))^{4n^2p(n)q(m)} \cdot 1 \\
 &\leq \text{Prob}[x \in \text{BAD}] + e^{-n}
 \end{aligned}$$

If we show that  $\text{Prob}[x \in \text{BAD}] \leq 1/2p(n)$  then we have that  $\text{Prob}[e(A', x)] \leq 1/2p(n) + e^{-n} < 1/p(n)$ , which brings us to contradicting the assumption that  $f_n$  is a weak one-way function. To show that  $\text{Prob}[x \in \text{BAD}] \leq 1/2p(n)$ , assume (towards contradiction) that this is not the case. Then let  $\vec{x} = (x_1, \dots, x_{2np(n)})$  and define the event  $e(A, \vec{x})$  as the event that  $A$  successfully inverts  $\vec{y} = g_m(\vec{x})$ , when  $\vec{x}$  is uniformly chosen. Then we have that the probability of event  $e(A, \vec{x})$  is

$$\begin{aligned}
 &= \text{Prob}\left[e(A, \vec{x}) \mid \bigvee_{i=1}^{2np(n)} x_i \in \text{BAD}\right] \cdot \text{Prob}\left[\bigvee_{i=1}^{2np(n)} x_i \in \text{BAD}\right] \\
 &\quad + \text{Prob}\left[e(A, \vec{x}) \mid \bigwedge_{i=1}^{2np(n)} x_i \notin \text{BAD}\right] \cdot \text{Prob}\left[\bigwedge_{i=1}^{2np(n)} x_i \notin \text{BAD}\right] \\
 &\leq \sum_{i=1}^{2np(n)} \text{Prob}[e(A, \vec{x}) \mid x_i \in \text{BAD}] \cdot \text{Prob}[x_i \in \text{BAD}] \\
 &\quad + \text{Prob}\left[e(A, \vec{x}) \mid \bigwedge_{i=1}^{2np(n)} x_i \notin \text{BAD}\right] \cdot \text{Prob}\left[\bigwedge_{i=1}^{2np(n)} x_i \notin \text{BAD}\right] \\
 &\leq (2np(n)) \cdot \left(\frac{1}{4np(n)q(m)}\right) \cdot 1 + 1 \cdot (1 - 1/2p(n))^{2np(n)} \\
 &\leq \left(\frac{1}{2q(m)}\right) + e^{-n} \\
 &< \left(\frac{1}{q(m)}\right)
 \end{aligned}$$

which negates our original assumption and therefore gives us a contradiction.  $\square$