

## Handout 6: Equivalence of GM and Semantic Security

The TA's came up with a simple proof that if a cryptosystem is GM-secure, it is also semantically secure. Below is their original write-up of this proof.

**Notation:** Note that in this write-up, the cryptosystem in question is denoted as  $\mathcal{C}$ , the key-generation algorithm is also denoted as  $\mathcal{C}$  and the public key is denoted by  $E$ . Furthermore, when  $E$  is a public key, the notation  $E(m)$  is used to denote the encryption of message  $m$  using public key  $E$ . (This notation is quite natural if you think of the key-generation procedure as producing the code of the encryption algorithm with the public-key hard-coded in.)

## GM Security $\implies$ Semantic Security

We show that  $\neg$  Semantic Security  $\implies \neg$  GM-Security. Let  $\{M_n\}$  be message spaces,  $f$  be a polynomial-time computable function, and  $\{A_n\}$  be circuits such that for a fixed  $c > 0$  and infinitely many  $n$

$$\Pr[A_n(E, \alpha) = f(m) \mid m \leftarrow M_n, E \leftarrow \mathcal{C}(1^n), \alpha \leftarrow E(m)] \geq \tilde{p} + \frac{1}{n^c} \quad (1)$$

where  $\tilde{p} = \mathbb{E}_{E \leftarrow \mathcal{C}(1^n)}[p_E]$  is the expected prediction probability without the knowledge of  $\alpha$ .

Consider the following algorithm  $T_n : (E, m_0, m_1, \alpha) \rightarrow \{0, 1\}$ .

1. Let  $\beta \leftarrow A_n(E, \alpha)$ .
2. If  $\beta = f(m_0)$  but  $\beta \neq f(m_1)$ , output 0.
3. If  $\beta = f(m_1)$  but  $\beta \neq f(m_0)$ , output 1.
4. Otherwise, output a random value from  $\{0, 1\}$  with probability  $\frac{1}{2}$  each.

The test is very intuitive. We simply run  $A_n$  on the challenge  $\alpha$ . Since we expect  $A_n$  to correctly predict the value of  $f$ , we compare its output  $\beta$  with  $f(m_0)$  and  $f(m_1)$ . Note that the test is clearly polynomial time since all the steps (including computations of  $f$ ) are polynomial time.

If exactly one of the tests succeed, we output the corresponding message. Otherwise, we flip a coin as we did not learn anything. For specific  $m_0$  and  $m_1$ , let

$$q(m_0, m_1) = \Pr[T_n(E, m_0, m_1, \alpha) = i \mid i \in_r \{0, 1\}, E \leftarrow \mathcal{C}(1^n), \alpha \leftarrow E(m_i)]$$

be the probability that  $T_n$  distinguishes encryptions of  $m_0$  and  $m_1$ .

We show that  $T_n$  violates the GM-security of  $\mathcal{C}$  by finding two particular messages  $m_0$  and  $m_1$  that are distinguished by  $T_n$ , i.e.  $q(m_0, m_1) \geq \frac{1}{2} + \frac{1}{2n^c}$  (same  $c$  as in (1)). To show the existence of such  $m_0$  and  $m_1$  we use the probabilistic method. We pick both  $m_0$  and  $m_1$  *independently at random according to the given probability distribution*  $M_n$  (that violates the Semantic Security in (1)). We then argue that  $T_n$  has non-negligible *expected* advantage in distinguishing a random encryption of  $m_0$  or  $m_1$ , i.e.  $q := \mathbb{E}_{m_0, m_1}[q(m_0, m_1)] \geq \frac{1}{2} + \frac{1}{2n^c}$ . Hence, the required  $m_0$  and  $m_1$  exist.

It remains to prove the bound on  $q$ . We note that since the algorithm  $T_n$  is symmetric in  $m_0$  and  $m_1$ ,  $q$  equals to the expected probability that  $T_n$  outputs 0 if  $\alpha$  is an encryption of  $m_0$ , i.e. without loss of generality we can assume that  $i = 0$ . Now, our experiment can be viewed as the following. Pick  $m_0 \leftarrow M_n$ ,  $E \leftarrow \mathcal{C}(1^n)$ ,  $\alpha \leftarrow E(m_0)$ ,  $\beta \leftarrow A_n(E, \alpha)$ . Now we pick a *brand new* message  $m_1 \leftarrow M_n$  and run steps 2–4 of  $T_n$ .  $q$  is the probability that we output 0. Before computing  $q$ , we claim that

$$\Pr[\beta = f(m_0)] \geq \tilde{p} + \frac{1}{n^c}; \quad \Pr[\beta = f(m_1)] \leq \tilde{p} \quad (2)$$

Indeed, the first bound follows directly from (1), as  $\beta \leftarrow A_n(E, \alpha)$  and  $\alpha \leftarrow E(m_0)$ . For the second bound, we observe that for any *fixed*  $E$ , the message  $m_1$  is chosen *independent* of  $m_0$ ,  $\alpha \leftarrow E(m_0)$  and, therefore,  $\beta \leftarrow A_n(E, \alpha)$ . Hence, for any fixed  $E$  the probability that  $f(m_1)$  equals to  $\beta$  is at most the probability that it equals to any pre-specified element, which is at most  $p_E$ . Since for a fixed  $E$ , our probability is stochastically dominated by  $p_E$ , we can take the expectation over  $E$  to obtain the claimed bound.

Now, using the fact  $\Pr[A \wedge B] + \Pr[A \wedge \bar{B}] = \Pr[A]$ , we can compute the probability  $q$  of outputting 0 in the following way:

$$\begin{aligned} q &= \Pr[\beta = f(m_0) \wedge \beta \neq f(m_1)] + \frac{1}{2}(\Pr[\beta = f(m_0) = f(m_1)] + \Pr[\beta \notin \{f(m_0), f(m_1)\}]) \\ &= \frac{1}{2}(\Pr[\beta = f(m_0) \wedge \beta \neq f(m_1)] + \Pr[\beta = f(m_0) \wedge \beta = f(m_1)]) + \\ &\quad \frac{1}{2}(\Pr[\beta = f(m_0) \wedge \beta \neq f(m_1)] + \Pr[\beta \neq f(m_0) \wedge \beta \neq f(m_1)]) \\ &= \frac{1}{2}(\Pr[\beta = f(m_0)] + \Pr[\beta \neq f(m_1)]) = \frac{1}{2} + \frac{1}{2}(\Pr[\beta = f(m_0)] - \Pr[\beta = f(m_1)]) \\ &\stackrel{(2)}{\geq} \frac{1}{2} + \frac{1}{2} \left( \left( \tilde{p} + \frac{1}{n^c} \right) - \tilde{p} \right) = \frac{1}{2} + \frac{1}{2n^c} \end{aligned}$$

This concludes the proof.  $\square$