

# OneDigit Schema

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## 1 Notation

Let  $\mathcal{A}$  be the set of alphabet. We assume that  $|\mathcal{A}| = N$ . For the case of password generation,  $\mathcal{A} = \{A, B, \dots, Z\}$  and  $N = 26$ . We denote the set of digits by  $\mathcal{D}$ , i.e.,  $\mathcal{D} = \{0, \dots, 9\}$ . Let's  $\mathcal{C}$  denotes the set of possible challenges. We denote the  $i^{th}$  coordinate of a vector  $\vec{u}$  by  $u_i$ .

## 2 OneDigit Schema

. Given a positive integer  $A$ , we define  $[A] = 0, \dots, A - 1$

### 2.1 Preprocessing step

- Memorize a random map  $f : \mathcal{A} \rightarrow \mathcal{D}$
- Memorize a random string  $s = s_1 \dots s_{d-1} \in \mathcal{D}^{d-1}$

### 2.2 Processing step

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**Algorithm 1** OneDigit schema

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Input: Challenge  $c = c_1 \dots c_l$

$g \stackrel{10}{\equiv} f(c_1) + \dots + f(c_l)$

Output: Response  $sg$

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Before stating the main theorem of this note, we define the notion of strong linearly independence.

**Definition 1.** We say that set of challenges  $\{c_1, \dots, c_p\}$  is strong linearly independent (mod 10) if  $\{c_1, \dots, c_p\}$  is linearly independent (mod 5) and (mod 2). Note that a direct consequence of strong linear independence is linear independence.

**Theorem 2.** Denote the output of OneDigit schema on a challenge  $c$ , by  $p(c)$ . We define  $\mathcal{R} = \{p(c) \mid c \in \mathcal{C}\}$ . For any challenge  $c \in \mathcal{C}$  and any response  $r \in \mathcal{R}$

$$(a) \Pr[p(c) = r] = \frac{1}{10^d}$$

Furthermore, assume that we have made  $k$  observations  $(c_1, p(c_1)), \dots, (c_k, p(c_k))$ . Then,  $\forall g_{k+1} \in \mathcal{D}$  and  $\forall c_{k+1} \in \mathcal{C}$  s.t.  $\{c_1, \dots, c_k, c_{k+1}\}$  is strong linearly independent (mod 10)

$$(b) \Pr[p(c_{k+1}) = sg_{k+1} \mid (p(c_1) = sg_1), \dots, (p(c_k) = sg_k)] = 1/10$$

Part (a) is saying that without having any prior information, the probability of guessing the correct response to any single challenge is  $1/10^d$ . In other words, for any two responses  $r_1$  and  $r_2$

$$\Pr[p(c) = r_1] = \Pr[p(c) = r_2]$$

Now assume that the adversary has observed  $k$  (challenge, response) pairs and she is trying to guess the response to a new challenge  $c_{k+1}$ . After seeing the first (challenge, response) pair, she will know the value of  $s$ . So the only unknown part of  $p(c_{k+1})$  is the single digit  $g_{k+1}$ . Part (b) is saying that for any new challenge  $c_{k+1}$  which forms a strong linearly independent set with  $k$  previously observed challenges, the adversary can not do better than guessing  $g_{k+1}$  randomly.

*Proof.* (a) For any  $c \in \mathcal{C}, r \in \mathcal{R}$ . Let  $r = r_1 \dots r_d$

$$\begin{aligned} \Pr[p(c) = r] &= \Pr[p(c)_1 \dots p(c)_{d-1} = r_1 \dots r_{d-1}] \Pr[p(c)_d = r_d] \\ &= \Pr[s = r_1 \dots r_{d-1}] \Pr[p(c)_d = r_d] \end{aligned}$$

Since each digit of string  $s$  is chosen independently at random, the above formula is equal to

$$\Pr[s_1 = r_1] \dots \Pr[s_{d-1} = r_{d-1}] \Pr[r(c)_d = r_d]$$

The first  $d-1$  probabilities appearing above are each equal to  $1/10$ . Thus we only need to compute  $\Pr[r(c)_d = r_d] = \Pr[f(c_1) + \dots + f(c_l) \equiv r_d \pmod{10}]$ . One way to compute this probability is to count the number of maps  $f$  that satisfy

$$f(c_1) + \dots + f(c_l) \equiv r_d \pmod{10} \quad (1)$$

and divide it by the total number of maps  $f : \mathcal{A} \rightarrow \mathcal{D}$ . What is the number of maps  $f$  that satisfy Eq. 2 ? One can choose  $f(c_1), \dots, f(c_{l-1})$  arbitrarily, then  $f(c_l)$  will be chosen uniquely by  $f(c_l) \equiv r_d - \sum_{i=1}^{l-1} f(c_i) \pmod{10}$ . So the total number of choices of  $f$  will be  $10^{N-l}$  for the letters that are not present in  $c$ ,  $10^{l-1}$  for the first  $l-1$  letters in  $c$  and 1 for the last letter in  $c$ . So the total number of choices is  $10^{N-l} 10^{l-1} = 10^{N-1}$ . Note that the total number of maps  $f : \mathcal{A} \rightarrow \mathcal{D}$  is  $10^N$ . This leads to

$$\Pr[r(c)_d = r_d] = \Pr[f(c_1) + \dots + f(c_l) = r_d] = \frac{10^{N-1}}{10^N} = \frac{1}{10}$$

Consequently, accounting for the fixed string  $s$

$$\Pr[r(c) = r] = \frac{1}{10^{d-1}} \frac{1}{10} = \frac{1}{10^d}$$

- (b) Now assume that the adversary have observed  $k$  (challenge, response) pairs  $(c_1, p(c_1) = sg_1), \dots, (c_k, p(c_k) = sg_k)$ , and we want to compute

$$\Pr[(p(c_{k+1}) = sg_{k+1}) \mid (p(c_1) = sg_1), \dots, (p(c_k) = sg_k)]$$

This is equal to

$$\Pr[(p(c_{k+1})_d = g_{k+1}) \mid (p(c_1)_d = g_1), \dots, (p(c_k)_d = g_k)]$$

which is equal to

$$\frac{\Pr[(p(c_{k+1})_d = g_{k+1}), (p(c_1)_d = g_1), \dots, (p(c_k)_d = g_k)]}{\Pr[(p(c_1)_d = g_1), \dots, (p(c_k)_d = g_k)]} \quad (2)$$

We start by computing the value of denominator. The nominator value can be achieved similarly. In order to compute  $\Pr[(p(c_1)_d = g_1), \dots, (p(c_k)_d = g_k)]$ , we should count the number of mappings  $f$  that satisfy

$$\begin{cases} f(c_{11}) + \dots + f(c_{1l}) \equiv g_1 \pmod{10} \\ \vdots \\ f(c_{k1}) + \dots + f(c_{kl}) \equiv g_k \pmod{10} \end{cases} \quad (3)$$

Lemma 4 shows that the number of solutions to above  $k$  linear equations is  $10^{N-k}$ . Therefore, the value of the ratio (2) is equal to

$$\frac{10^{n-k+1}}{10^{n-k}} = \frac{1}{10}$$

□

The main component of lemma 4 is to count the number of solutions to a system of linear equations mod a prime number  $p$ . Lemma 3 achieves this result.

**Lemma 3.** *Given a prime number  $p$ , matrix  $C_{k \times N}$  with  $k$  linearly independent rows  $(\text{mod } p)$ , and vector  $g_{k \times 1}$ , the number of solutions to  $Cx \stackrel{p}{\equiv} g$  is  $p^{N-k}$ .*

*Proof.* By assumption, rows of matrix  $C$  are linearly independent  $(\text{mod } p)$ , thus there must be  $k$  columns  $\{C^{j_1}, \dots, C^{j_k}\}$  that are linearly independent  $(\text{mod } p)$ . Let's denote the  $j^{th}$  coordinate of vector  $x$  by  $x_j$ . We claim that for any set  $\mathcal{X}_{N-k} = \{x_j \in \{0, \dots, p-1\} : j \notin \{j_1, \dots, j_k\}\}$ , there will be a unique set  $\mathcal{X}_k = \{x_j \in \{0, \dots, p-1\} : j \in \{j_1, \dots, j_k\}\}$  such that  $x = \mathcal{X}_k \cup \mathcal{X}_{N-k}$  is a solution for system  $Cx \stackrel{p}{\equiv} g$ .

Given an arbitrary set  $\mathcal{X}_{N-k}$ , let's substitute values of  $x_j$  for  $j \notin \{j_1, \dots, j_k\}$  in  $x$  and simplify the equation  $Cx \stackrel{p}{\equiv} g$  :

$$[C^{j_1}, \dots, C^{j_k}] \begin{bmatrix} x_{j_1} \\ \vdots \\ x_{j_k} \end{bmatrix} \stackrel{p}{\equiv} \begin{bmatrix} g'_1 \\ \vdots \\ g'_k \end{bmatrix}$$

Where  $g'_l = g_l - \sum_{i \notin \{j_1, \dots, j_k\}} C_{li} x_i$  for all  $l \in \{1, \dots, k\}$ .

Since matrix  $[C^{j_1}, \dots, C^{j_k}]$  is full rank (mod  $p$ ), the above linear equation has unique solution  $x \pmod{p}$ . So far we have proved that for every set  $\mathcal{X}_{N-k} = \{x_j \in \{0, \dots, p-1\} : j \notin \{j_1, \dots, j_k\}\}$ , there is a unique set  $\mathcal{X}_k = \{x_j \in \{0, \dots, p-1\} : j \in \{j_1, \dots, j_k\}\}$  such that  $x = \mathcal{X}_k \cup \mathcal{X}_{N-k}$  is a solution for system  $Cx \stackrel{p}{\equiv} g$ . This concludes that the number of solutions to  $Cx \stackrel{p}{\equiv} g$  is equal to the number of possible sets  $\mathcal{X}_{N-k}$  which is equal to  $p^{N-k}$ .  $\square$

**Lemma 4.** *Given a function  $f : \mathcal{A} \rightarrow \mathcal{D}$  and set  $\{c_1, \dots, c_k\} \subseteq \mathcal{C}$  strong linearly independent and  $g_1, \dots, g_{k+1} \in \mathcal{D}$ , the system of linear equations 3, has  $10^{N-k}$  solutions.*

*Proof.* Assume there is an ordering  $a_1, \dots, a_N$  on elements of  $\mathcal{A}$ . Let's define the  $N$ -dimensional column vector  $f$  such that  $f_i = f(a_i)$ . Given challenges  $\{c_1, \dots, c_k\}$ , for every challenge  $c_i$ , we define the  $N$ -dimensional row vector  $C_i$  as follows. The  $i^{th}$  coordinate of  $C_i$ , is the number of occurrence of  $a_i$  in  $c$ . In this vector setting, the last system of equations will be equivalent to

$$\begin{cases} C_1 \cdot f \stackrel{10}{\equiv} g_1 \\ \vdots \\ C_k \cdot f \stackrel{10}{\equiv} g_k \end{cases} \Rightarrow \begin{bmatrix} C_1 \\ \vdots \\ C_k \end{bmatrix} f \stackrel{10}{\equiv} \begin{bmatrix} g_1 \\ \vdots \\ g_k \end{bmatrix} \Rightarrow Cf \stackrel{10}{\equiv} g \quad (4)$$

Defining

$$C = \begin{bmatrix} c_1 \\ \vdots \\ c_k \end{bmatrix}, \quad g = \begin{bmatrix} g_1 \\ \vdots \\ g_k \end{bmatrix}$$

We want to count the number of solutions to  $Cf \stackrel{10}{\equiv} g$ . Let's denote  $F_2$  to be the set of solutions to  $Cf \stackrel{2}{\equiv} g$  and similarly  $F_5$  to be the set of solutions to  $Cf \stackrel{5}{\equiv} g$ . Using lemma 3,  $|F_2| = 2^{N-k}$  and  $|F_5| = 5^{N-k}$ . We claim that for every  $x \in F_2$  and  $y \in F_5$  there is a unique vector  $z \pmod{10}$  such that  $z$  is a solution for  $Cz \stackrel{10}{\equiv} g$ . Furthermore, for every such vector  $z \pmod{10}$ , there is a unique pair  $x \in F_2$  and  $y \in F_5$ .

Given  $x \in F_2$ , and  $y \in F_5$ , we prove that there exists a  $k$ -dimensional vector  $z$  such that  $z \equiv x \pmod{5}$  and  $z \equiv y \pmod{2}$ . Consider the following  $k$  systems of simultaneous congruences:

$$\begin{array}{ccc} z_1 \stackrel{5}{\equiv} x_1 & \dots & z_k \stackrel{5}{\equiv} x_k \\ z_1 \stackrel{2}{\equiv} y_1 & & z_k \stackrel{2}{\equiv} y_k \end{array}$$

Using Chinese Remainder Theorem, there exist  $z_1, \dots, z_k$  satisfying the above congruences. Therefore

$$\begin{aligned} z \stackrel{5}{\equiv} x &\Rightarrow Cz \stackrel{5}{\equiv} Cx \stackrel{5}{\equiv} g \\ &\Rightarrow Cz \stackrel{10}{\equiv} g \\ z \stackrel{2}{\equiv} y &\Rightarrow Cz \stackrel{2}{\equiv} Cy \stackrel{2}{\equiv} g \end{aligned}$$

Furthermore, for all  $i \in [k]$ ,  $z_i \pmod{10}$  is unique. Therefore  $z = [z_1, \dots, z_k]^T$  will be the unique solution to  $Cf \stackrel{10}{\equiv} g$ . So far we have shown that for every pair  $(x, y) \in F_5 \times F_2$  there is one and only one vector  $z$  satisfying  $Cz \stackrel{10}{\equiv}$ . Therefore, the number of solutions to  $Cz \stackrel{10}{\equiv}$  is equal to  $|F_2||F_5| = 10^{N-k}$   $\square$

**Lemma 5.** (*Chinese Remainder Theorem*) Suppose  $n_1, \dots, n_k$  are positive integers that are pairwise coprime. Then, for any given sequence of integers  $a_1, \dots, a_k$ , there exists an integer  $x$  solving the following system of simultaneous congruences.

$$\begin{cases} x \equiv a_1 & (\text{mod } n_1) \\ \vdots \\ x \equiv a_k & (\text{mod } n_k) \end{cases}$$

Furthermore, any two solutions of this system are congruent modulo the product  $N = n_1 \dots n_k$ . Hence, there is a unique (non-negative) solution less than  $N$ .

## 2.3 HUM

In order to calculate the HUM, we first need to write the steps of Alg. 2.2 in more details

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### Algorithm 2 OneDigit schema

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Input: Challenge  $c = c_1 \dots c_l$ 
Set  $i = 1$ , SUM = 0
While not EndOfChallenge :
    Compute  $f(c_i)$  (Applying the map)
    SUM  $\stackrel{10}{\equiv}$  SUM +  $f(c_i)$  (Add to the running sum)
     $i = i + 1$  (shift pointer)

Print fixed string  $s$ 
Print SUM

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$\text{HUM} = 2 \times (\text{initialization}) + l \times (\text{while loop condition}) + l \times (\text{map}) + l \times (\text{add}) + l \times (\text{shift pointer}) + (\text{end while}) + 2 \times (\text{print}) = 4l + 5$

## 2.4 Security: $Q$ value

Theorem. 2 is saying that as long as a new challenge forms a strong linearly independent set with already observed challenges, the adversary can not predict the response to this new challenge. In order to compute the value of  $Q$  for OneDigit schema we should answer the following question:

- ◊ Given a set of challenges  $\mathcal{C}$ , at each round, a new challenge  $c \in \mathcal{C}$  is chosen uniformly at random. Let  $\mathcal{C}_i$  be the set of challenges chosen till round  $i$ . What is the maximum  $i$  such that  $\mathcal{C}_i$  is a strong linearly independent set?

To answer the above question, we ran the following experiment. We chose the challenge set  $\mathcal{C}$  to be the set of all the valid website names. At each iteration, our program choses a challenge  $c$  uniformly at random from  $\mathcal{C}$  and checks if  $c$  along with challenges chosen so far, forms a strong linearly independent set. If yes, it saves the number of iterations as the  $Q$  value. Otherwise, it will continue. We repeated this procedure for 1000 times and took the average of all the saved  $Q$  values. The following table shows the result:

Number of trials	Q
1000	$\sim 18$