

# 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. For the sake of simplicity, we assume that each challenge  $c \in \mathcal{C}$  does not contain more than four repeated letters. We denote the  $i^{th}$  coordinate of a vector  $\vec{u}$  by  $u_i$ .

## 2 OneDigit Schema

### 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  $k$  (input, output) pairs and she is trying to guess the response to a new challenge  $c_{k+1}$ . After seeing the first (input, output) 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  $k$  observed challenges, the adversary can't 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

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

In the next lemma, we show that the number of solutions to above  $k$  linear equations is  $10^{N-k}$ . Therefore, the value of the ratio (2) is equal to

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

□

**Lemma 3.** *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  $\vec{f}$  such that  $\vec{f}_i = f(a_i)$  s.t.  $a_i$  is the  $i^{th}$  element of  $\mathcal{A}$ . Similarly, for every challenge  $c \in \mathcal{C}$ , we define the  $N$ -dimensional row vector  $\vec{c}$  as follows. The  $i^{th}$  coordinate of  $\vec{c}$ ,  $\vec{c}_i$ , is the number of occurrence of the  $a_i$  in  $c$ . In this vector setting, the last system of equations will be equivalent to

$$\begin{cases} \vec{c}_1 \cdot \vec{f} \equiv g_1 \\ \vdots \\ \vec{c}_k \cdot \vec{f} \equiv g_k \end{cases} \Rightarrow \begin{bmatrix} \vec{c}_1 \\ \vdots \\ \vec{c}_k \end{bmatrix} \cdot \vec{f} \equiv \begin{bmatrix} g_1 \\ \vdots \\ g_k \end{bmatrix} \quad (4)$$

Let's define

$$C = \begin{bmatrix} \vec{c}_1 \\ \vdots \\ \vec{c}_k \end{bmatrix}, \quad \vec{g} = \begin{bmatrix} g_1 \\ \vdots \\ g_k \end{bmatrix}$$

By assumption, rows of matrix  $C$  are linearly independent (mod 2), thus there must be  $k$  columns  $\{C^{j_1}, \dots, C^{j_k}\}$  that are linearly independent (mod 2). Using Prop. 5,  $\{C^{j_1}, \dots, C^{j_k}\}$  are linearly independent (mod 5) as well.

We claim that for any set  $\mathcal{F}_{N-k} = \{\vec{f}_j \in \mathcal{D} : j \in \{j_1, \dots, j_k\}^c\}$ , there will be a unique set  $\mathcal{F}_k = \{\vec{f}_j \in \mathcal{D} : j \in \{j_1, \dots, j_k\}\}$  such that  $\mathcal{F}_k \cup \mathcal{F}_{N-k}$  is a solution for system 4.

Given a set  $\mathcal{F}_{N-k}$ , let's substitute arbitrary values of  $f_j$  for  $j \notin \{j_1, \dots, j_k\}$  in Eq. 3. So the matrix equation will be simplified as follows

$$[C^{j_1}, \dots, C^{j_k}] \begin{bmatrix} f_{j_1} \\ \vdots \\ f_{j_k} \end{bmatrix} \stackrel{10}{\equiv} \begin{bmatrix} g'_{j_1} \\ \vdots \\ g'_{j_k} \end{bmatrix}$$

Therefore

$$[C^{j_1}, \dots, C^{j_k}] \begin{bmatrix} f_{j_1} \\ \vdots \\ f_{j_k} \end{bmatrix} \stackrel{5}{\equiv} \begin{bmatrix} g'_{j_1} \\ \vdots \\ g'_{j_k} \end{bmatrix}, \quad [C^{j_1}, \dots, C^{j_k}] \begin{bmatrix} f_{j_1} \\ \vdots \\ f_{j_k} \end{bmatrix} \stackrel{2}{\equiv} \begin{bmatrix} g'_{j_1} \\ \vdots \\ g'_{j_k} \end{bmatrix} \quad (5)$$

Since matrix  $[C^{j_1}, \dots, C^{j_k}]$  is full rank (mod 5) and (mod 2), the above linear equations respectively has unique solution  $\vec{x}$  (mod 5) and  $\vec{y}$  (mod 2). We claim that, there exists a vector  $\vec{z}$  such that  $\vec{z} \equiv \vec{x}$  (mod 5) and  $\vec{z} \equiv \vec{y}$  (mod 2). To prove our claim, we first need to briefly remind Chinese Remainder Theorem.

**Theorem 4.** (*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  $N$ .

We want to prove that there exists a vector  $\vec{z}$  such that  $\vec{z} \equiv \vec{x}$  (mod 5) and  $\vec{z} \equiv \vec{y}$  (mod 2). Consider the following  $k$  systems of simultaneous congruences:

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

Using Chinese Remainder Theorem, there exist  $z_1, \dots, z_k$  satisfying the above congruences. Furthermore, for all  $i \in [k]$ ,  $z_k \pmod{10}$  is unique. Therefore  $z = [z_1, \dots, z_k]^T$  will be the unique solution to Eq. 5.

So far we have shown that every set  $\mathcal{F}_{N-k} = \{\vec{f}_j \in \mathcal{D} : j \notin \{j_1, \dots, j_k\}\}$ , there is a unique set  $\mathcal{F}_k = \{\vec{f}_j \in \mathcal{D} : j \in \{j_1, \dots, j_k\}\}$  such that  $\mathcal{F}_k \cup \mathcal{F}_{N-k}$  is a solution for system 4. Therefore, the number of solutions to system of linear equations 4 is number of sets  $\mathcal{F}_{N-k} = \{\vec{f}_j \in \mathcal{D} : j \notin \{j_1, \dots, j_k\}\}$  which is equal to  $10^{N-k}$ .  $\square$

**Proposition 5.** Given a set of challenges  $\mathcal{C} = \{c_1, \dots, c_k\}$  s.t.  $\forall i \in [k]$ , the challenge  $c_i$  does not contain more than four repeated letters. If  $\mathcal{C}$  is linearly independent (mod 2), this implies that it is linearly independent (mod 5) as well.

*Proof.*

□

## 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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HUM =  $2 \times$  (initialization) +  $l \times$  (while loop condition) +  $l \times$  (map) +  $l \times$  (add) +  $l \times$  (shift pointer) + (end while) +  $2 \times$  (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	$\sim 18$