

# CS771 Report

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## Question 1:

**Theorem 1.** For any arbiter PUF, there exists a linear model  $(\mathbf{w}, b) \in \mathbb{R}^{32} \times \mathbb{R}$  such that the delay  $\Delta$  of the PUF on the challenge  $\mathbf{c}$  is given by

$$\Delta = \mathbf{w}^T \mathbf{x} + b$$

where  $\mathbf{x} \in \mathbb{R}^{32}$  is the feature vector defined by:

$$x_i = (1 - 2c_i)(1 - 2c_{i+1})(1 - 2c_{i+2}) \dots (1 - 2c_{32})$$

for each  $i = 1, 2, \dots, 32$ .

We show how to extend this to the 2-arbiter case.

**Theorem 2.** For a CAR-PUF, there exists a linear model  $(\mathbf{W}, b) \in \mathbb{R}^{528} \times \mathbb{R}$  such that the response of the CAR-PUF on challenge  $\mathbf{c}$  is given by:

$$y = \frac{1 + \text{sgn}(\mathbf{W}^T \phi(\mathbf{c}) + b)}{2}$$

Where,

$$\phi(\mathbf{c}) = (z_{1,1}, z_{1,2} \dots z_{1,32}, z_{2,2}, \dots, z_{2,32}, \dots, z_{32,32})$$

And

$$z_{i,j} = \prod_{l=i}^j (1 - 2c_l)$$

*Proof.* By Theorem 1, there must exist models  $(\mathbf{u}, p), (\mathbf{v}, q)$  which perfectly model the response of the working and reference PUFs exactly.

Thus, we can predict the difference of their delays by a linear model as follows:

$$\Delta_w - \Delta_r = \mathbf{u}^T \mathbf{x} + p - \mathbf{v}^T \mathbf{x} - q = (\mathbf{u} - \mathbf{v})^T \mathbf{x} + (p - q) = \mathbf{m}^T \mathbf{x} + r$$

where  $\mathbf{m} = \mathbf{u} - \mathbf{v}$  and  $r = p - q$ .

Now, notice that the response of the CAR-PUF is 1 exactly when

$$|\Delta_w - \Delta_r| > \tau \iff (\Delta_w - \Delta_r)^2 - \tau^2 > 0$$

Simplifying this expression:

$$\begin{aligned} (\Delta_w - \Delta_r)^2 - \tau^2 &= (\mathbf{m}^T \mathbf{x} + r)^2 - \tau^2 \\ &= \left( \sum_{i=1}^{32} m_i x_i + r \right)^2 - \tau^2 \\ &= \left( \sum_{i=1}^{32} m_i x_i \right)^2 + 2r \left( \sum_{i=1}^{32} m_i x_i \right) + r^2 - \tau^2 \\ &= \sum_{i=1}^{32} m_i^2 x_i^2 + 2 \sum_{1 \leq i < j \leq 32} m_i m_j x_i x_j + 2r \left( \sum_{i=1}^{32} m_i x_i \right) + r^2 - \tau^2 \end{aligned}$$

Since  $x_i \in \{-1, 1\}$ ,  $x_i^2 = 1$ .

$$= \sum_{i=1}^{32} m_i^2 + 2 \sum_{1 \leq i < j \leq 32} m_i m_j x_i x_j + 2r \left( \sum_{i=1}^{32} m_i x_i \right) + r^2 - \tau^2$$

Let  $z_{i,j} = \prod_{l=i}^j (1 - 2c_l)$ , for all  $j \geq i$ . Then, we note the following:

$$x_i = z_{i,32}$$

And,

$$\begin{aligned} x_i x_j &= \prod_{l=i}^{32} (1 - 2c_l) \prod_{l=j}^{32} (1 - 2c_l) \\ &= \prod_{l=i}^{j-1} (1 - 2c_l) \prod_{l=j}^{32} (1 - 2c_l)^2 \\ &= \prod_{l=i}^{j-1} (1 - 2c_l) \\ &= z_{i,j-1} \end{aligned}$$

Thus, letting  $\alpha_{i,j-1} = 2m_i m_j$ ,  $\alpha_{i,32} = 2rm_i$ ,  $\gamma = \sum_{i=1}^{32} m_i^2 + r^2 - \tau^2$ , we have:

$$\begin{aligned} (\Delta_w - \Delta_r)^2 - \tau^2 &= \sum_{1 \leq i < j \leq 32} 2m_i m_j x_i x_j + \sum_{i=1}^{32} 2rm_i x_i + \sum_{i=1}^{32} m_i^2 + r^2 - \tau^2 \\ &= \sum_{1 \leq i < j \leq 32} \alpha_{i,j-1} z_{i,j-1} + \sum_{1 \leq i \leq 32} \alpha_{i,32} z_{i,32} + \gamma \\ &= \sum_{1 \leq i < j \leq 31} \alpha_{i,j} z_{i,j} + \sum_{1 \leq i \leq 32} \alpha_{i,32} z_{i,32} + \gamma \\ &= \sum_{1 \leq i, j \leq 32} \alpha_{i,j} z_{i,j} + \gamma \end{aligned}$$

Which is a linear model in  $\phi(\mathbf{c})$ . There are  $\binom{32}{2} + 32 = 528$   $z_{i,j}$ 's. For the linear model  $(\mathbf{W}, b)$  defined by  $\alpha_{i,j}$ 's and  $\gamma$ , the response for a challenge is precisely  $\frac{1 + \text{sgn}(\mathbf{W}^T \phi(\mathbf{c}) + b)}{2}$ .  $\square$

### Question 3:

**a**

LinearSVC		
loss hyperparameter	Accuracy	Time (sec)
hinge	0.98864	15.700767
squared hinge	0.99116	17.999646

**b**

Changing C		
LinearSVC		
C Value	Accuracy	Time (sec)
0.1	0.98990	20.783399
1.0	0.99172	17.527513
10.0	0.98974	17.341398
100.0	0.98996	17.572503

**ii**

LogisticRegression		
C Value	Accuracy	Time (sec)
0.1	0.98710	2.004277
1.0	0.99070	2.509324
10.0	0.99220	2.880717
100.0	0.99310	4.015238

**d**

Changing Penalty/Regularization		
LinearSVC		
Penalty	Accuracy	Time (sec)
l1	0.99124	163.014093
l2	0.99190	17.716546

ii	LogisticRegression		
	Penalty	Accuracy	Time (sec)
	l1	0.99180	246.677855
	l2	0.99070	3.020177

