
Assignment 1

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1 Breaking Companion Arbiter PUF into a linear model

Let the *challenge* bits be $\mathbf{c} \stackrel{\text{def}}{=} [c_0, c_1, c_2, \dots, c_{31}]$.

We define a new vector $\varphi(c) \stackrel{\text{def}}{=} [\varphi_0, \varphi_1, \varphi_2, \dots, \varphi_{31}]$, such that :-

$$\varphi_i \stackrel{\text{def}}{=} \prod_{k=i}^{31} (1 - 2c_k)$$

We also note that, since $c_i \in \{0, 1\} \Rightarrow (1 - 2c_i) \in \{-1, 1\} \Rightarrow \varphi_i \in \{-1, 1\}$

From Week 1 lecture slides, we know that the lag of an arbiter PUF can be broken up into a linear model given by :-

$$\Delta_{31} = \mathbf{w}^T \cdot \varphi(c) + \beta$$

Let (\mathbf{u}, p) be the linear model for *working* arbiter PUF and (\mathbf{v}, q) be the linear model for *reference* arbiter PUF, i.e.

$$\Delta_w = \mathbf{u}^T \cdot \varphi(c) + p$$

$$\Delta_r = \mathbf{v}^T \cdot \varphi(c) + q$$

The response of our CAR-PUF is 0 if $|\Delta_w - \Delta_r| \leq \tau$ and response is 1 if $|\Delta_w - \Delta_r| > \tau$
We have

$$\Delta_w - \Delta_r = (\mathbf{u}^T \cdot \varphi(c) + p) - (\mathbf{v}^T \cdot \varphi(c) + q)$$

$$\Delta_w - \Delta_r = (\mathbf{u} - \mathbf{v})^T \cdot \varphi(c) + (p - q)$$

Now for the response to be 0,

$$|\Delta_w - \Delta_r| \leq \tau$$

$$|(\mathbf{u} - \mathbf{v})^T \cdot \varphi(c) + (p - q)| \leq \tau$$

Given $\tau > 0$, squaring the above we get

$$\begin{aligned} |(\mathbf{u} - \mathbf{v})^T \cdot \varphi(c) + (p - q)|^2 &\leq \tau^2 \\ \left| \sum_{i=0}^{31} ((u_i - v_i)\varphi_i) + (p - q) \right|^2 &\leq \tau^2 \\ \sum_{i=0}^{31} (u_i - v_i)^2 \varphi_i^2 + (p - q)^2 + 2 \sum_{i=0}^{31} \sum_{j=i+1}^{31} (u_i - v_i)(u_j - v_j) \varphi_i \varphi_j + 2(p - q) \sum_{i=0}^{31} (u_i - v_i) \varphi_i &\leq \tau^2 \end{aligned}$$

Claim : $\varphi_i^2 = 1 \forall i \in \{0, 1, 2, \dots, 31\}$

Proof : Since the challenge bit $c_i \in \{0, 1\} \implies (1 - 2c_i) \in \{-1, 1\}$

By the definition of φ_i , we have

$$\varphi_i \stackrel{\text{def}}{=} \prod_{k=i}^{31} (1 - 2c_k)$$

$$(1 - 2c_i) \in \{-1, 1\} \implies \varphi_i \in \{-1, 1\} \implies \varphi_i^2 = 1$$

Therefore, $\varphi_i^2 = 1 \forall i \in \{0, 1, 2, \dots, 31\}$

Now, our inequality reduces to :-

$$\begin{aligned} \sum_{i=0}^{31} (u_i - v_i)^2 + (p - q)^2 + 2 \sum_{i=0}^{31} \sum_{j=i+1}^{31} (u_i - v_i)(u_j - v_j) \varphi_i \varphi_j + 2(p - q) \sum_{i=0}^{31} (u_i - v_i) \varphi_i &\leq \tau^2 \\ 2 \sum_{i=0}^{31} \sum_{j=i+1}^{31} (u_i - v_i)(u_j - v_j) \varphi_i \varphi_j + 2(p - q) \sum_{i=0}^{31} (u_i - v_i) \varphi_i + \sum_{i=0}^{31} (u_i - v_i)^2 + (p - q)^2 - \tau^2 &\leq 0 \end{aligned}$$

We choose our feature vectors to be :- $\{\varphi_i; 0 \leq i \leq 31\} \cup \{\varphi_i \varphi_j; 0 \leq i \leq 31, i + 1 \leq j \leq 31\}$

With these feature vectors we note that the above model is linear, and we define our feature map

$\phi : \{0, 1\}^{32} \rightarrow \mathbb{R}^D$ where $D = 32 + \binom{32}{2} = 528$

Now we represent in terms of matrix as follows :

$$\begin{bmatrix} 2(p - q)(u_0 - v_0) \\ \vdots \\ 2(p - q)(u_{31} - v_{31}) \\ 2(u_0 - v_0)(u_1 - v_1) \\ 2(u_0 - v_0)(u_2 - v_2) \\ \vdots \\ 2(u_{30} - v_{30})(u_{31} - v_{31}) \end{bmatrix}^T \begin{bmatrix} \varphi_0 \\ \vdots \\ \varphi_{31} \\ \varphi_0 \varphi_1 \\ \varphi_0 \varphi_2 \\ \vdots \\ \varphi_{30} \varphi_{31} \end{bmatrix} + \sum_{i=0}^{31} (u_i - v_i)^2 + (p - q)^2 - \tau^2 = \mathbf{W}^T \cdot \phi(\mathbf{c}) + b$$

where

$$\mathbf{W} = \begin{bmatrix} 2(p - q)(u_0 - v_0) \\ \vdots \\ 2(p - q)(u_{31} - v_{31}) \\ 2(u_0 - v_0)(u_1 - v_1) \\ 2(u_0 - v_0)(u_2 - v_2) \\ \vdots \\ 2(u_{30} - v_{30})(u_{31} - v_{31}) \end{bmatrix}_{528} \quad \phi(\mathbf{c}) = \begin{bmatrix} \varphi_0 \\ \vdots \\ \varphi_{31} \\ \varphi_0 \varphi_1 \\ \varphi_0 \varphi_2 \\ \vdots \\ \varphi_{30} \varphi_{31} \end{bmatrix}_{528} \quad b = \sum_{i=0}^{31} (u_i - v_i)^2 + (p - q)^2 - \tau^2$$

The response of this CAR-PUF is 0 if $\mathbf{W}^T \cdot \phi(\mathbf{c}) + b \leq 0$ and response is 1 if $\mathbf{W}^T \cdot \phi(\mathbf{c}) + b > 0$
Hence, the response to the challenge is given by the following expression:

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

We have defined φ_i as:

$$\begin{aligned} \varphi_i &\stackrel{\text{def}}{=} \prod_{k=i}^{31} (1 - 2c_k) \implies \varphi_i \varphi_j = \prod_{k=i}^{j-1} (1 - 2c_k) \cdot \prod_{t=j}^{31} (1 - 2c_t)^2 \\ \varphi_i \varphi_j &= \prod_{k=i}^{j-1} (1 - 2c_k) \text{ since } (1 - 2c_t)^2 = 1 \quad \forall t \in \{0, 1, 2, \dots, 31\} \end{aligned}$$

Therefore, our map $\phi(\mathbf{c})$ is given by :

$$\phi(\mathbf{c}) = \begin{bmatrix} \prod_{k=0}^{31} (1 - 2c_k) \\ \prod_{k=1}^{31} (1 - 2c_k) \\ \vdots \\ \prod_{k=31}^{31} (1 - 2c_k) \\ \prod_{k=0}^0 (1 - 2c_k) \\ \prod_{k=0}^1 (1 - 2c_k) \\ \vdots \\ \prod_{k=30}^{30} (1 - 2c_k) \end{bmatrix}_{528} \quad \text{where } \mathbf{c} = \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_{31} \end{bmatrix}_{32}$$

2 Outcomes with different models while tuning hyperparameters

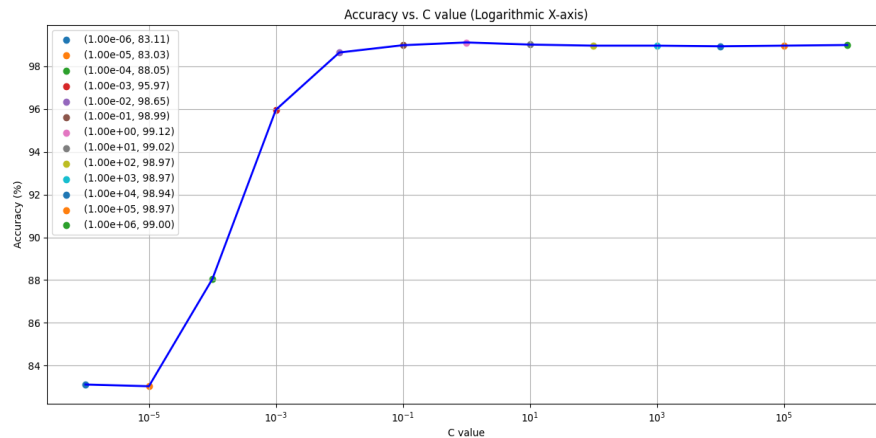
2.1 Changing loss hyperparameter in LinearSVC:

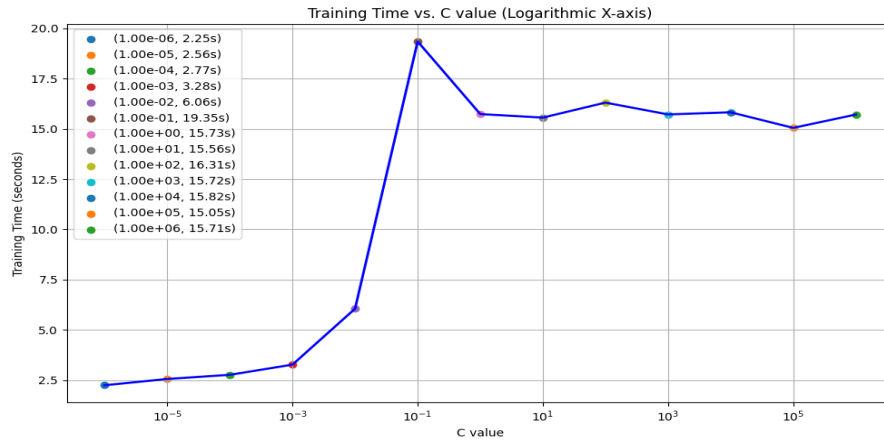
Accuracies and Training time on changing the loss function in the LinearSVC model, with default values of hyperparameters and max_iter = 10,000 are as shown:

Loss	Test Accuracy	Training Time
Hinge	98.82	14.23s
Squared Hinge	99.14	16.27s

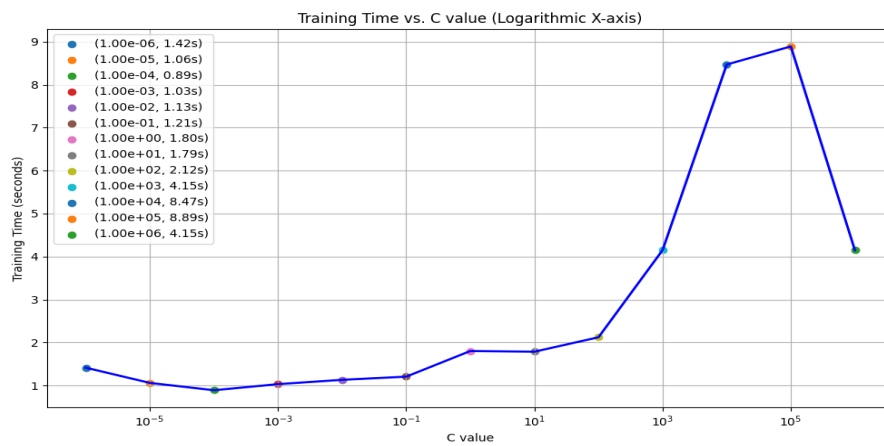
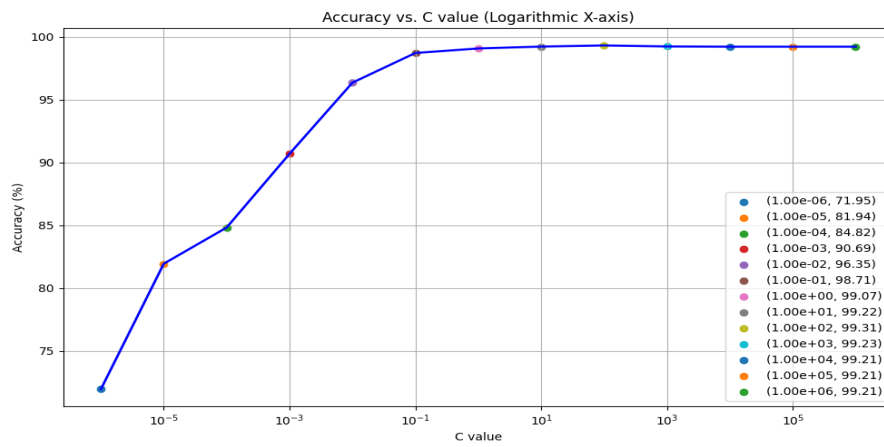
2.2 Tuning C hyperparameter in LinearSVC and LogisticRegression:

2.2.1 Linear SVC



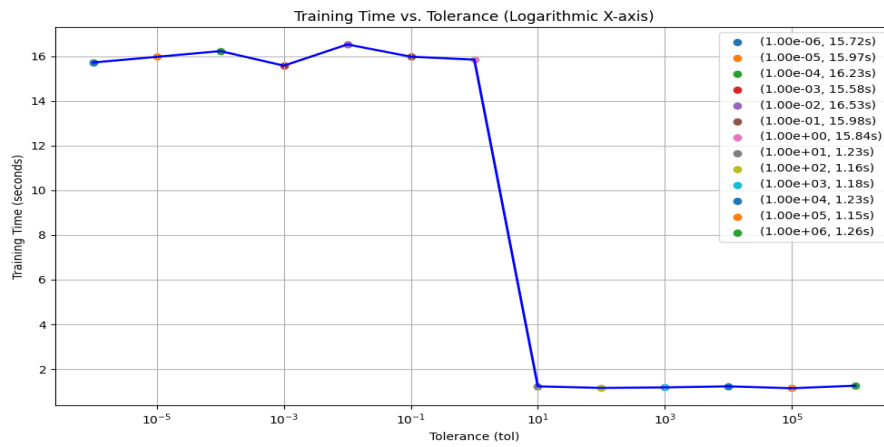
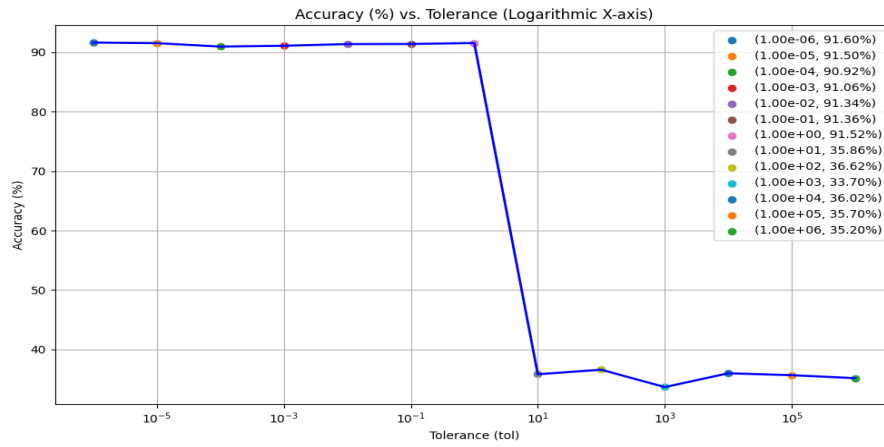


2.2.2 Logistic Regression

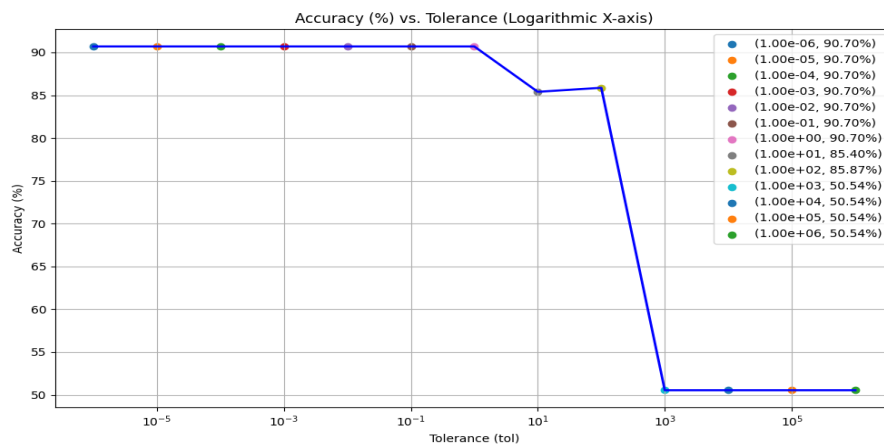


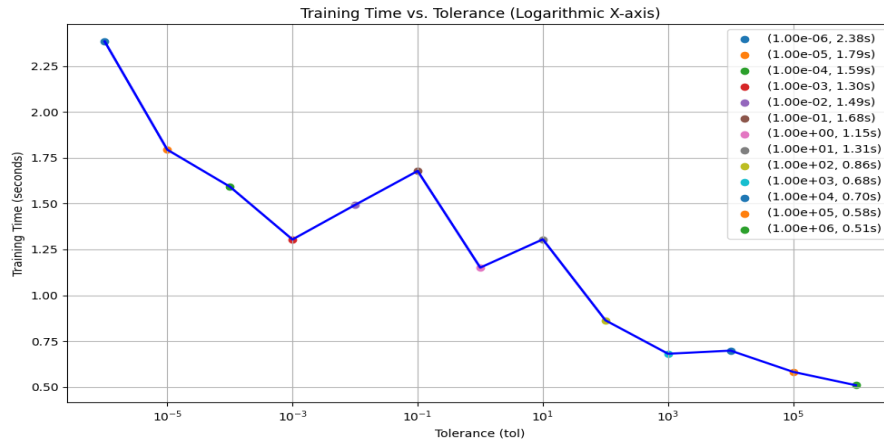
2.3 Tuning tol hyperparameter in LinearSVC and LogisticRegression:

2.3.1 Linear SVC



2.3.2 Logistic Regression





2.4 Changing penalty (regularization) hyperparameter in LinearSVC and LogisticRegression:

Model	Test Accuracy	Training Time
LinearSVC	L1: 99.13	L1: 170s
	L2: 99.17	L2: 15.90s
Logistic Regression	L1: 99.18	L1: 234.79s
	L2: 99.08	L2: 2.85s