



# Applying deep learning to known-plaintext attack on chaotic image encryption schemes

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# 1. Introduction

The **known plaintext attack**, as one cryptanalysis method, is crucial to evaluate the security of image encryption.

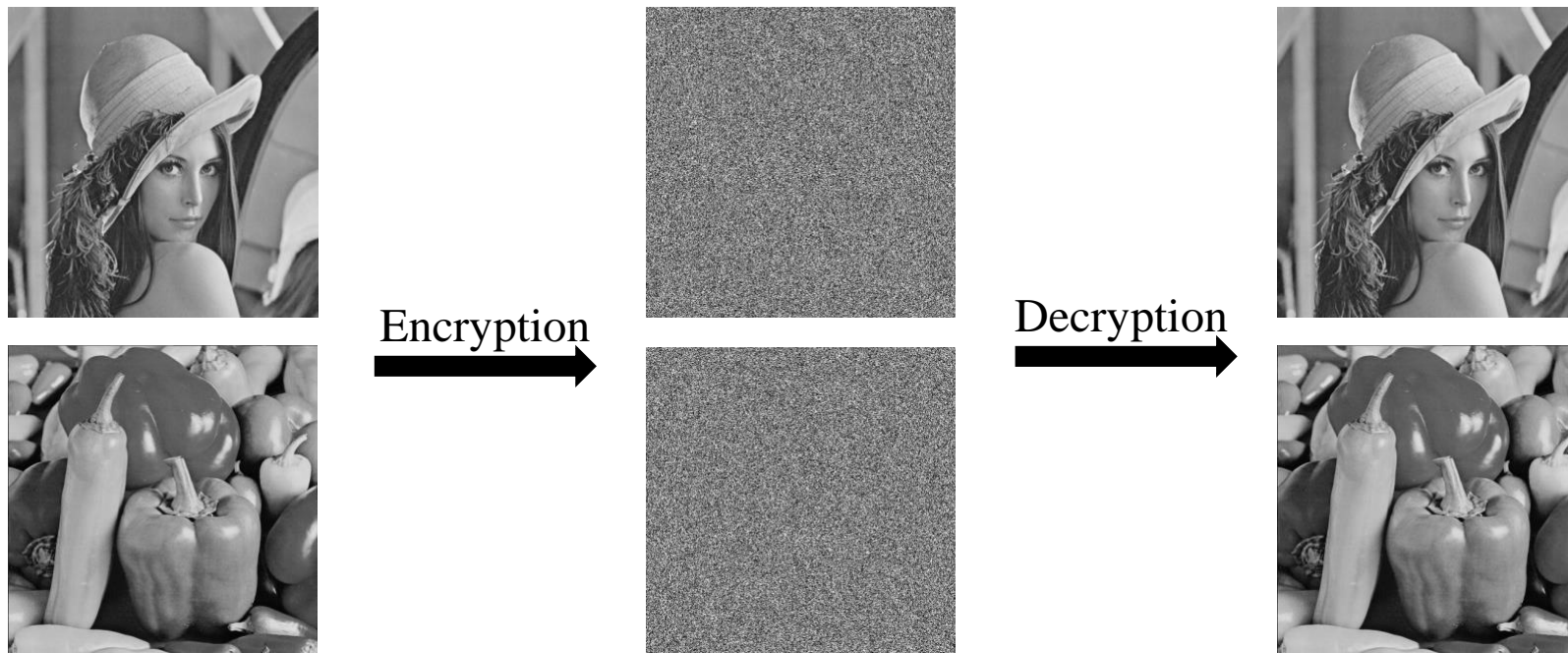


Fig. 1 Some “plaintext-ciphertext” image pairs.



## 2. Traditional Known plaintext attack

The traditional known plaintext attack works are usually based on some **mathematical means**, such as **differential attack**. It has some **shortcomings**:

- It is **complicated** to design an attack scheme.
- Usually, one attack method is only designed for a **specific** chaotic encryption system, which is **hard to be applied to other chaotic encryption systems**.



# 3. Proposed Approach

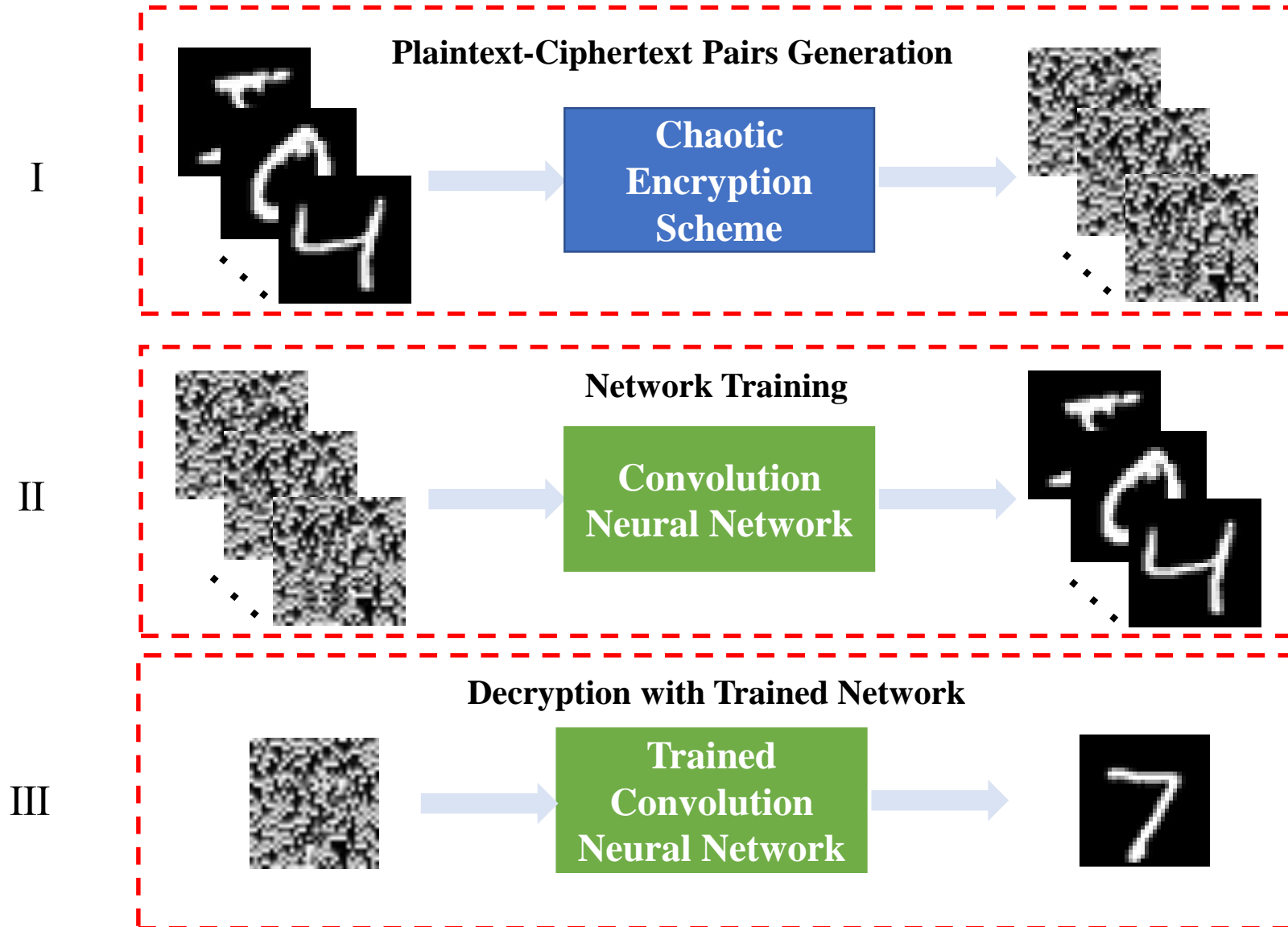


Fig. 2 The overall framework of deep learning-based known-plaintext attack on chaotic cryptosystem.



The **advantages** of the proposed deep learning-based known-plaintext attack method:

- It is **easy** to **design a convolution neural network** for the **known plaintext attack**.
- Different from the traditional known-plaintext attack methods for chaotic cryptosystems, **a convolutional neural network** can be employed to **decrypt different chaotic cryptosystems**.



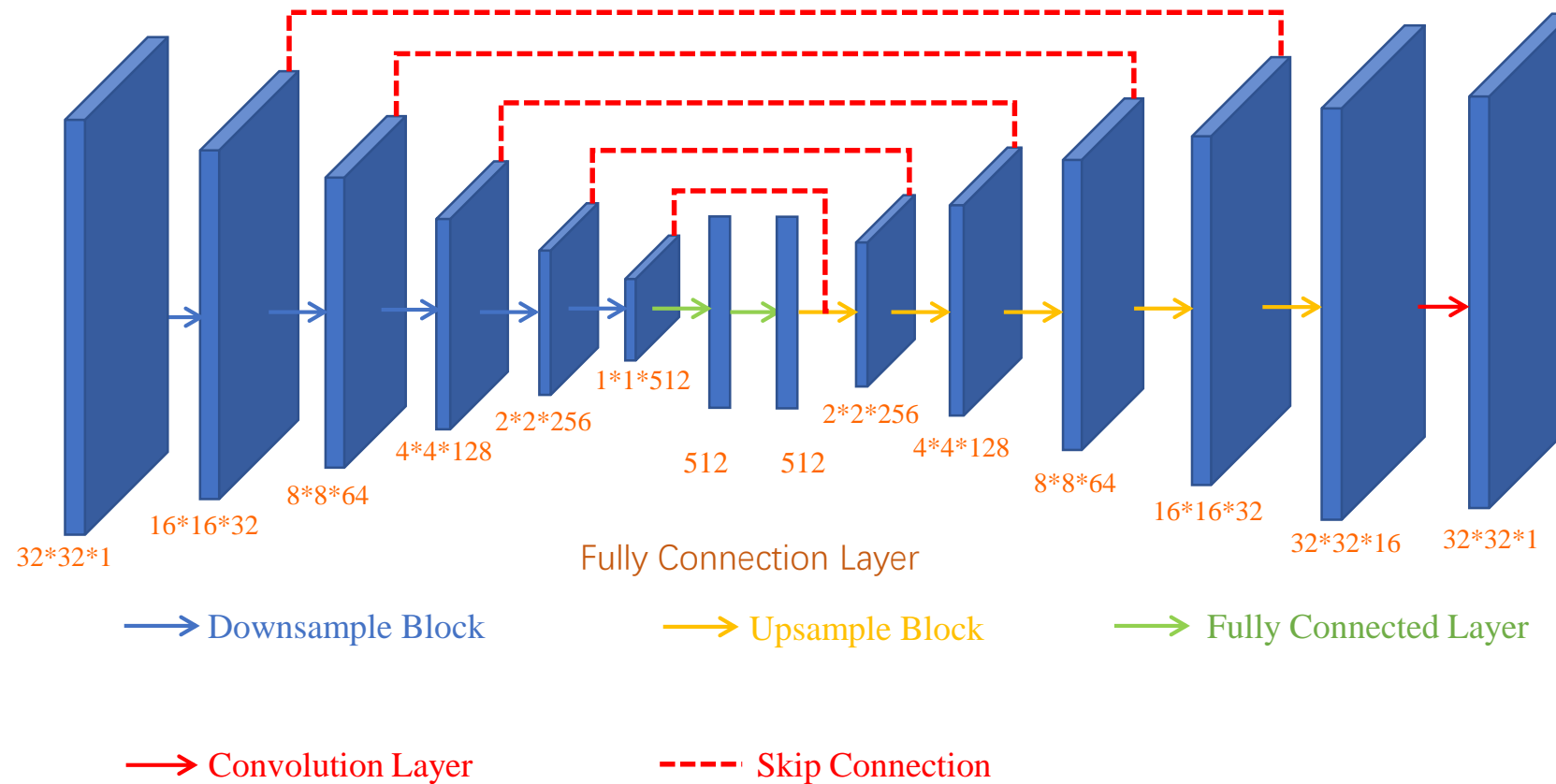


Fig. 3 The architecture of the proposed image decryption encoder-decoder network IDEDNet.



- Loss Function (L1 Loss).

$$\mathcal{L}_1 = \frac{1}{N} \sum_{i=1}^N |O(C_i; \theta) - P(C_i)|$$

- Attacked chaotic encryption schemes.

(1) Song et al. [5]; (2) Pak et al. [10]; (3) H. N. Abdullah et al. [4].

- Evaluation metric (Pearson correlation coefficient).

$$Corr = \frac{(O - \bar{O})(P - \bar{P})}{\sigma(O)\sigma(P)}$$

[5] Y. Song, J. Song, and J. Qu, “A secure image encryption algorithm based on multiple one-dimensional chaotic systems,” in ICCS, 2016.

[10] C. Pak and L. Huang, “A new color image encryption using combination of the 1d chaotic map,” *Signal Process*, vol. 138, pp. 129–137, 2017.

[4] H. N. Abdullah and H. A. Abdullah, “Image encryption using hybrid chaotic map,” in ICCIT, 2017.





Table. 1 The ciphertext reconstruction result of known-plaintext attack method based on deep learning on MNIST and MNIST-Fashion datasets.

Network	Encryption Scheme	dataset	Training correlation coefficient	Testing correlation coefficient	Epoch	Time/Epoch
IDEDNet	Song et al. [5]	MNIST and Fashion	97.6%	94.2%	300	5.8s
	Pak et al. [8]		97.7%	94.5%		5.7s
	H. N. Abdullah et al. [4]		98.6%	96.7%		5.8s



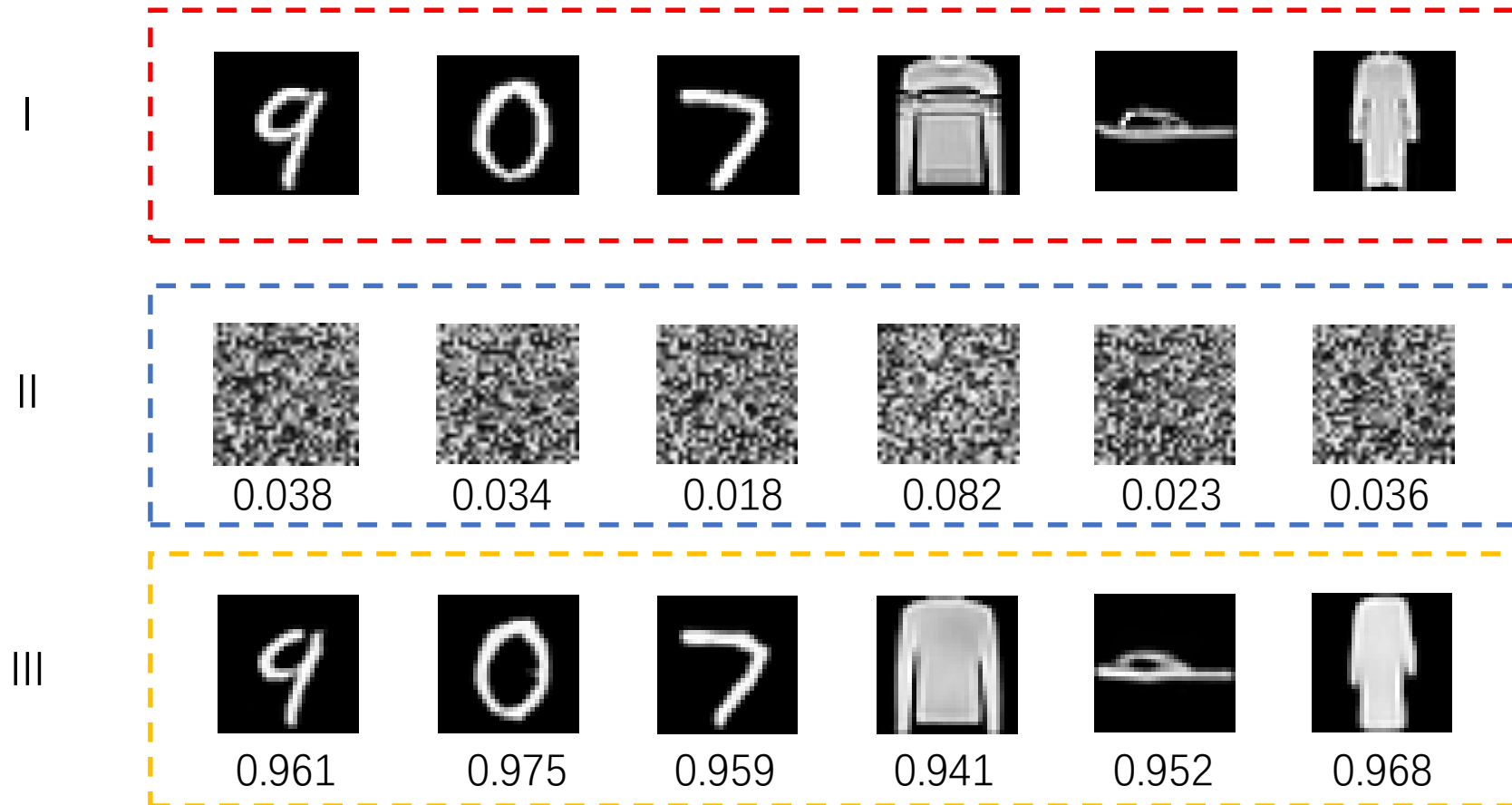
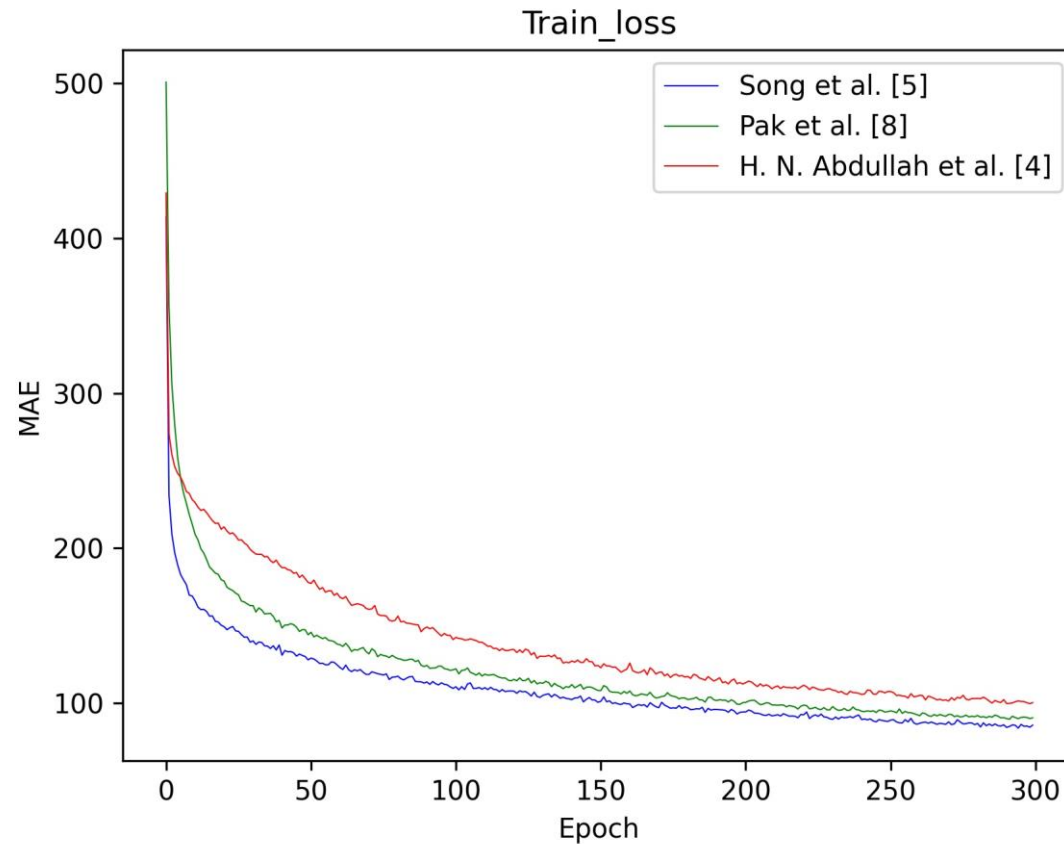
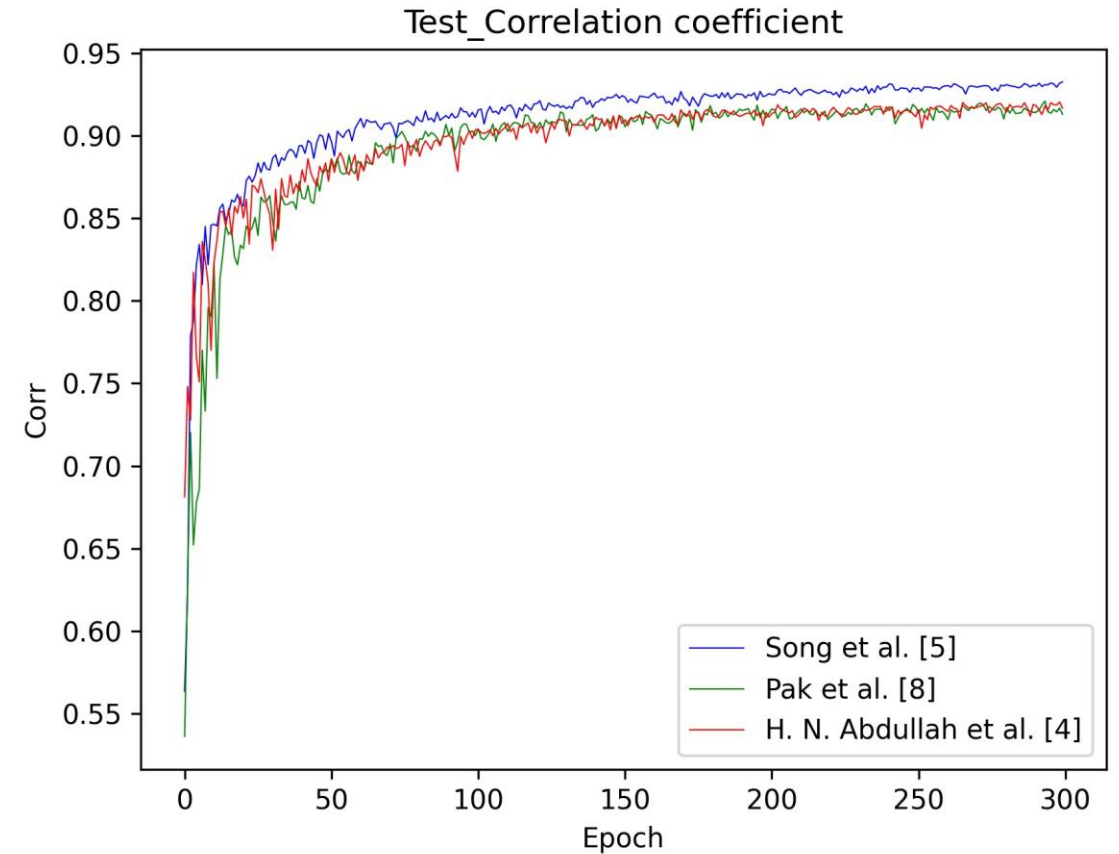


Fig. 4 visualization results on Song et al. [5]. (I) Plaintext image; (II) Ciphertext image; (III) Decrypted image; The number under the image represents the correlation coefficient between the image and the plaintext.





(a)



(b)

Fig. 5 The change curve of training and testing process on the mixed dataset of MNIST and MNIST-Fashion: (a) Training L1Loss, (b) Testing Correlation Coefficient. The blue, green and red lines represent the chaotic encryption schemes of Song et al. [5], Pak et al. [10], H. N. Abdullah et al. [4].



- Compared with the traditional known-plaintext attack methods specific to a certain chaotic cryptosystem, our method is more **cost-effective, flexible**;
- The chaotic cryptanalysis method based on deep learning can be introduced to **multiple chaotic cryptosystems** and even to **the field of non-chaotic cryptosystems**;
- It also proposes a **new research direction** in the field of multimedia security, i.e., how to **prevent cryptography attack methods based on deep learning**.



# Q&A

