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SCFformer: a binary data hiding method against JPEG compression based on spatial channel fusion Transformer

Key words: Binary data hiding; Against JPEG compression; Discrete cosine transform quantization; SCFformer

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Motivation

- Targeting the inadequacy of current data hiding algorithms, which exhibit significant extraction errors after low-quality-factor JPEG compression and demonstrate limited practical applicability, a novel anti-JPEG compression binary data hiding method based on spatial channel fusion Transformer (SCFformer) is proposed, which can recover secret information with high precision after image processing.

Main idea

- A novel anti-JPEG compression binary data hiding method based on SCFformer is proposed, which can provide high-precision secret information recovery with a low quality factor.
- A discrete cosine transform (DCT) quantization truncation mechanism is proposed, which improves the anti-JPEG compression ability of stego images, improves the flexibility by using the scalable module, and realizes secret data hiding with variable capacity.
- An efficient fusion Transformer is proposed, which improves the ability of capturing the dimensional relationships of the entire space and channels while maintaining high efficiency, thereby achieving a lower secret data extraction error rate (ER) at the same capacity.

Method

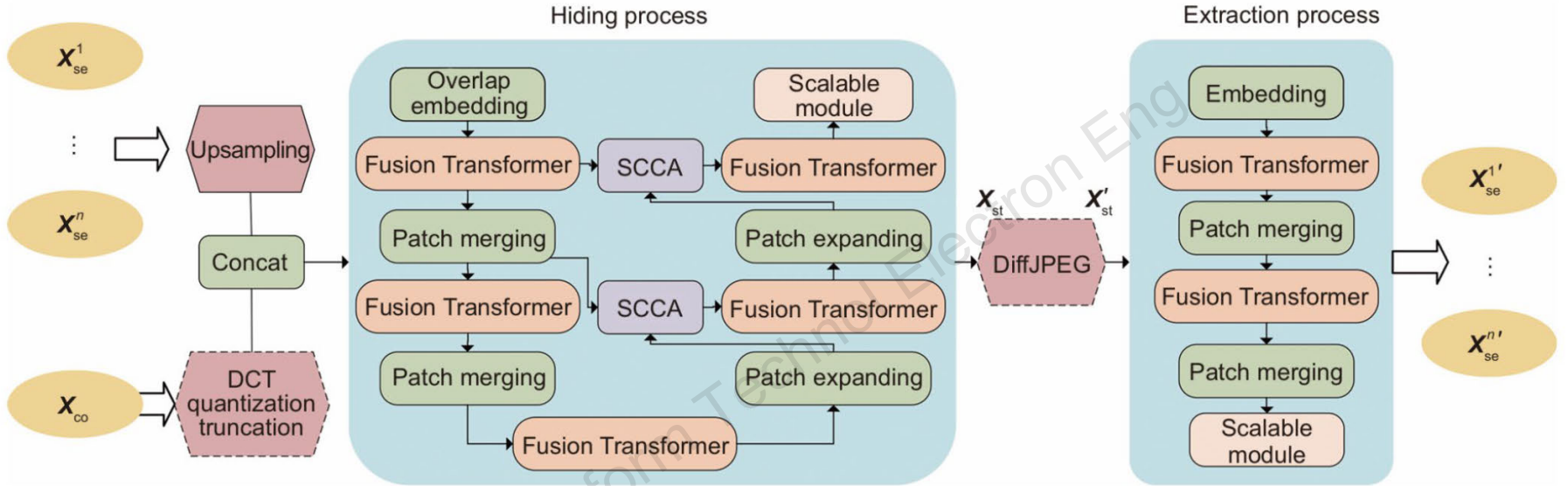


Fig. 1 The overall framework of SCFformer data hiding scheme including mainly two parts, hiding process and extraction process

Learning process can be described as

$$\begin{cases} T(Q, K, V) = VC_T(K, Q), \\ C_T(K, Q) = \text{Softmax}\left(\frac{K^T Q}{\tau}\right), \end{cases}$$

$$\begin{cases} E_{\text{block}}(X, Q_1, K_1, V_1) = E(Q_1, K_1, V_1) + X, \\ MF_1(E_{\text{block}}) = MF(\text{LN}(E_{\text{block}})), \\ T_{\text{block}}(E_{\text{block}}, Q_2, K_2, V_2) \\ = T(MF_1(E_{\text{block}}) + E_{\text{block}}) + MF_1(E_{\text{block}}), \\ MF_2(T_{\text{block}}) = MF(\text{LN}(T_{\text{block}})), \\ FT(T_{\text{block}}) = MF_2(T_{\text{block}}) + T_{\text{block}}. \end{cases}$$

(3)

Method

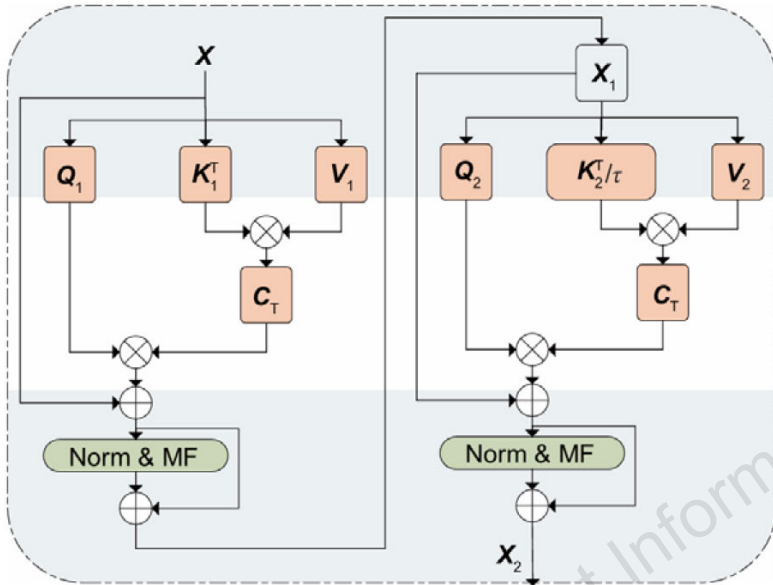


Fig. 2 Efficient fusion Transformer module including an efficient spatial attention block and a channel attention block, followed by a specification and mix-feedforward (MF) neural network

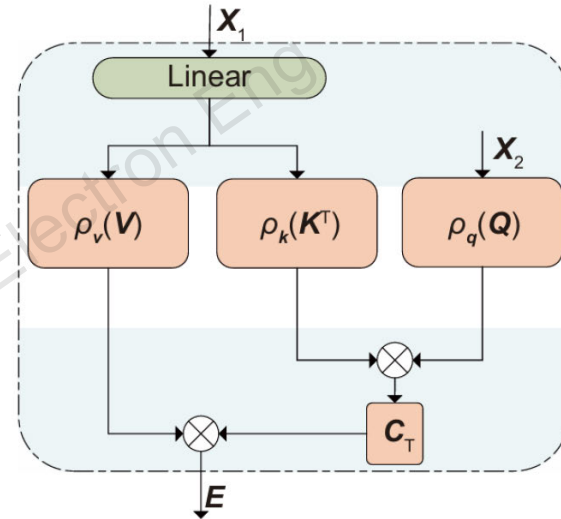


Fig. 3 The SCCA module containing a linear layer and an efficient attention module

Results

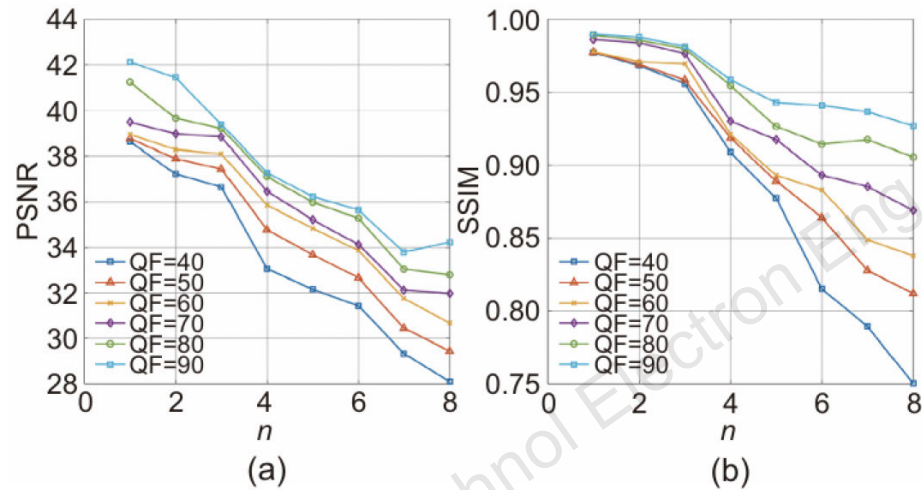


Fig. 8 PSNR (a) and SSIM (b)

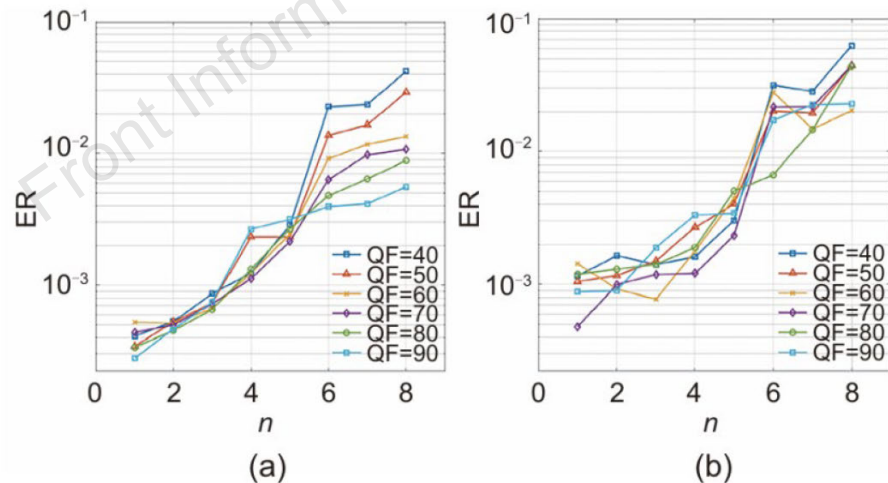


Fig. 9 Error rate (ER) under different n 's using SCF-former (a) and U-Net (b)

Results

Table 4 Generalization performance results in four datasets

Dataset	Type	MSE	ER	PSNR		SSIM	
				Max	Average	Max	Average
ImageNet1K	Variation <i>A</i>	0.000 180	0.000 721	<u>45.77</u>	38.85	<u>0.990</u>	0.976
	Variation <i>B</i>	0.000 195	0.000 972	<u>43.29</u>	38.12	<u>0.986</u>	0.967
	Variation <i>C</i>	0.000 572	0.001 216	41.33	33.61	0.978	0.928
CelebA	Variation <i>A</i>	0.000 064	0.000 139	46.06	42.91	<u>0.990</u>	0.980
	Variation <i>B</i>	<u>0.000 090</u>	<u>0.000 140</u>	43.68	<u>41.12</u>	<u>0.983</u>	0.968
	Variation <i>C</i>	0.000 198	0.000 420	42.33	37.98	0.974	0.954
DIV2K	Variation <i>A</i>	0.000 222	0.000 801	43.91	38.40	0.988	<u>0.978</u>
	Variation <i>B</i>	0.000 241	0.000 725	41.48	36.97	0.988	<u>0.971</u>
	Variation <i>C</i>	0.000 741	0.001 615	38.14	32.09	0.973	0.933
MS-coco	Variation <i>A</i>	0.000 172	0.000 801	45.58	39.35	0.991	0.977
	Variation <i>B</i>	0.000 194	0.000 537	42.98	38.23	0.987	0.967
	Variation <i>C</i>	0.000 540	0.001 154	41.23	33.64	0.931	0.977

The best results are in bold and the suboptimal results are underlined

Conclusions

In this work, we propose a novel anti-JPEG compression binary data hiding method based on spatial channel fusion Transformer (SCFformer), aiming to provide a large-capacity and secure data hiding method to achieve high-precision secret information recovery with a low quality factor. The design of SCFformer can provide some inspiration for further research on hiding a large amount of binary data in JPEG images.



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