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Removal of Artifacts in Image Using Adaptive Neighboring Estimator Algorithm

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ABSTRACT: Image compression is used to reduce irrelevance and redundancy of the image data to transmit data in an efficient form. Image compression is minimizing the size in bytes of a graphics file without degrading the quality of the image and it allows more images to be stored in a disk [20]. There are several different ways in which image files can be compressed. It can be Lossy Compression and Lossless compression. Although Lossy compression is very efficient, it introduces certain artifacts and noise [2]. The quality of images is reduced after decompression especially if the compression ratio is high. There are many methods to reduce the artifacts [15]. This paper presents Patch Based Removal Method, Semi-Local Gaussian Process and proposed technique which is hybrid of these two algorithms and use fuzzy logic and nearest neighboring algorithm. Experimental results show that the proposed method achieves higher PSNR than both the algorithms and the processed images present good visual quality.

KEYWORDS: Image Compression, Patch Based Deblocking, Semi-Local Gaussian Process, Adaptive Neighboring Estimator Algorithm.

I. INTRODUCTION

Artifact is any error in the perception or representation of any visual. Image compression is the basically reason for the artifacts [7].Different types of artifacts came when different algorithm are used for compressing the image as shown in Fig. 1.1.



Fig. 1.1 Original Image and Artifacts



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II. RELATED WORK

In [7] authors focused upon the low-quality images that were usually not only with low-resolution, but also suffered from compression artifacts, e.g., blocking artifacts. It performs image super-resolution (SR) to a low-quality image, which distorted the visual quality of an image. Here, proper HR and LR image patches were compared. The main problem with this is that it didn't properly convert low resolution to high resolution. Artifacts were removed but some cracks were left on the side edge of blocking parts and some color difference was also there i.e. lacked impainting. In [3] the authors emphasized upon patches of an image rather than the whole complete image. Before this, various difficulties found out with all the non local methods without any heuristic formulas. Here main problem was large area image impainting. In [4] it focused upon the image denoising and results were achieved using orthogonal dictionary. Here 3 patch based algorithm were used i.e. Patch Based Local PCA, Patch Based Global PCA, Patch Based Hierarchical PCA which performed thresholding on the coefficients of the patches. In [21] it emphasized that the fast patch based denoising technique was better than the patch by patch denoising and it was based upon the Geodesic Paths. Here patches of images were used and Euclidean line paths were used to connecting the two patches. Then giving weight to each patch and with the help of probabilities artifacts were removed. The main drawback of this was it was restricted to Gaussian noise, only handled the Poisson noise. In [17] authors focused upon the two techniques called LPF (Low Pass Filter) and ASLPF (Adaptive Spatial Low Pass Filter). LPF was used for only just identifying the artifacts present in the video image. ASLPF (Adaptive Spatial Low Pass Filter) was used to identify the artifacts present in horizontal or vertical position or near edges, it was converted into frames and each frame was then compared. ASLPF removed the artifacts either horizontally, vertically, or near the edges. But it produced some faded colors where algorithm were applied

III. PREVIOUS WORK

Patch based image deblocking method

Patch based image deblocking method using geodesic distance weighted low rank approximation is used by clustering similar patches, a low rank data matrix can be obtained. The singular value thresholding (SVT) algorithm is applied to solve the low rank approximation problem [1]. To compensate for the mismatch of the clustering procedure, patch based geodesic distance is utilized to weight different patches in the data matrix.

Steps:-

- 1. Read the original standard image.
- 2. For the compressed image, set patch as the reference patch and search for patches that are similar to within a neighborhood of position.
- 3. Apply SVT algorithm to solve low rank minimization problem.
- 4. Apply weighting method based on the geodesic distance to compensate for the mismatch of the clustering procedure.
- 5. Generate the final image.

Semi-Local Gaussian Processes Method

Instead of more time consuming training and testing of a single GP model on a large number of images, a set of sparse models are constructed at each point which is made by the corresponding sparse GP underlying global model. Gaussian processes have been used in various image enhancement problems.

Steps:-

Read the original standard image. Let I Be image data

- 1. Let Indata =I
- 2. Calculate egdePert
 - a) [Rmax, Gmax, Bmax] = MaxofInData (InData)
 - AvgEpsilon = (Rmax + Gmax + Bmax)/3
 - b) ProcessData = zeros (M, N,C)
 - c) ProcessData = double (ProcessData)
 - InData = double(InData)
 - d) ArgZ = abs(max(log(double(AvgEpsilon))))



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ProcessData= bitor(ArgZ,abs(InData))

3. Let sig = exp(Argz) Apply Process and sig to Guassian process to calculate final Data.

III. PROPOSED WORK

In proposed method, first of all read an input image and then image is being compressed using any compression technique. It gives blocking artifacts after high compression. Artifacts can be removed by using Patch based image deblocking method, Semi-Local Gaussian Processes Method and proposed method i.e. Adaptive Neighboring Estimator Algorithm After that apply Unsharp Mask Filter and get the final image. The quality can be measured using the various quality metrics like MSE, PSNR, SSIM and SC.

Adaptive Neighboring Estimator Algorithm (ANE)

The algorithm, followed in order to fulfil the aim of this thesis is as follows: **Step1:** Read the compressed image. **Step2:** Calculate Threshold values of various algorithms. **Step 3:** Apply Fuzzy logic. **Step4:** Apply Laplacian α-Blending technique and generate the image.

Mathematical Computation of ANE

Let I be input image

I₁=Patch (I)

 I_2 = Gaussian Semi Local Process (I)

 T_1 =Thresh (l_1)

 T_2 = Thresh (I_2)

Let \propto be blending image

Optimal \propto =Fuzzy (I_1 , I_2 , T_1 , T_2)

$$F_{g} = I_{1}. \propto \dots \dots (i)$$

$$F_b = I_2$$
. (1- ∞) ----- (ii)

Where,

F_=Foreground Image

 F_h =Background Image



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OutImg = $F_g + F_b$

FinalImg = UnSharp Mask(OutImg)

Flowchart Representation





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IV. RESULTS

The values in Table 1 are able to show the different performance metrics by using the patch based method and the values in Table 2 shows the different performance metrics by using the Semi-Local Gaussian method and the values in Table 3 are able to show the different performance metrics by using the Proposed method which clearly shows that this image is better than above two images because PSNR value is very much high than previous two methods and error i.e. MSE value is less. These results are at 70% compressed level

Image Name	Patch Based Method				
	PSNR	MSE	SSIM	SC	
Penguins.jpg	20.7159	6.3125	0.77658	1.0486	
onion.png	18.1972	6.8925	0.57431	1.1746	
Tulips.jpg	22.8404	5.8233	0.72146	1.0403	
lena512color.tiff	23.7522	5.6134	0.76217	1.0363	
bmp.bmp	17.9436	6.9509	0.79906	1.0778	
pears.png	18.2688	6.876	0.55415	1.016	

Table 1: Comparison of PSNR.MSE, SSIM and SC on different compressed images using Patch Based Method

	Semi-Local Gaussian Method				
Image Name					
	PSNR	MSE	SSIM	SC	
Penguins.jpg	20.5558	6.3494	0.73906	0.99593	
onion.png	17.957	6.9478	0.50005	1.0783	
Tulips.jpg	23.0789	5.7684	0.72225	0.99499	
lena512color.tiff	23.592	5.6503	0.73212	0.98435	
bmp.bmp	17.9145	6.9576	0.76953	1.0291	
pears.png	17.9677	6.9453	0.5047	0.9772	

Table 2: Comparison of PSNR.MSE, SSIM and SC on different compressed images using Semi-Local Gausssian Method



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Image Name	Proposed Method				
	PSNR	MSE	SSIM	SC	
Penguins.jpg	22.4198	5.9202	0.82921	1.0248	
onion.png	24.0743	5.5392	0.69767	1.0662	
Tulips.jpg	24.8971	5.3498	0.79896	1.0169	
lena512color.tiff	25.7474	5.154	0.82176	1.0163	
bmp.bmp	27.2655	4.8044	0.86331	1.0241	
pears.png	27.4915	4.7524	0.82787	1.0106	

Table 3: Comparison of PSNR.MSE, SSIM and SC on different compressed images using Proposed Method

V. CONCLUSION

This paper presents a new method for artifacts removal from compressed image. Artifacts round the image degradion and the visual quality of image. Restoration method based on Adaptive Neighboring Estimator Algorithm (ANE) used to attain the good visual quality outcomes of the image. This paper successfully improves the visual outcomes of the compressed image and achieves better results than available method.

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