

Fuzzy contrast enhancement for images in the compressed domain

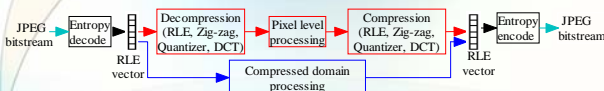
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Brief Overview

- Compressed domain image processing algorithms – in encoded JPEG image domain
 - Provide a powerful computation alternative to classical
- This field is in its beginning
 - The algorithms reported in the literature are mostly based on linear arithmetic point operations (addition, subtraction, multiplication)
- Addressed problem
 - Implementing a nonlinear operator using compressed domain processing – digital image enhancement using fuzzy theory
- Advantage:
 - No need to decompress/ recompress the whole image prior to enhance/ after enhance
 - For the 8x8 size blocks processed in the compressed domain, the enhancement algorithm implies a single comparison of the DC coefficient with the threshold (instead of 64 - at pixel level)

JPEG domain image enhancement



- There are two ways to enhance the images compressed with JPEG
 - Compressed domain processing = no decompression
 - Pixel level processing = enhance the image after decompression
- In our algorithm the enhancement unit is added to the JPEG decoder just before the decompression
 - Adaptive minimal decompression:
 - full decompression no longer needed, but
 - decompression used for blocks with many details for improved accuracy

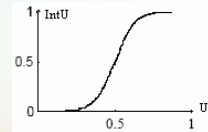
Contrast improvement with fuzzy theory

Analytical and graphical representation of the fuzzy contrast enhanced function:

$$U = \{u(i, j) \in [0, 1], \text{ where } i = 0, 1, \dots, H-1, j = 0, 1, \dots, W-1\}$$

$$IntU(i, j) = \begin{cases} 2 \cdot [u(i, j)]^2, & \text{if } 0 \leq u(i, j) \leq 0.5 \\ 1 - 2 \cdot [1 - u(i, j)]^2, & \text{if } 0.5 \leq u(i, j) \leq 1 \end{cases}$$

where $i = 0, 1, \dots, H-1; j = 0, 1, \dots, W-1$



- U is original image and $IntU$ is enhanced image
- For contrast enhancement with fuzzy functions - the square of each pixels' luminance is needed => a nonlinear operation

Convolution of the image in the compressed domain

- the square of each pixels' luminance is needed – nonlinear operation; implemented in the compressed domain using convolution:

$$U_{dct}(x_1, x_2) = \frac{1}{4 \cdot Q(x_1, x_2)} \sum_{y_1, y_2} C(i, x_1) \cdot C(j, x_2) \cdot u(i, j) \cdot u(i, j) = \sum_{y_1, y_2, w_1, w_2} U_{dc}(y_1, y_2) \cdot U_{dc}(w_1, w_2) \cdot W_0(y_1, y_2, w_1, w_2, x_1, x_2)$$

where: $W_0(y_1, y_2, w_1, w_2, x_1, x_2) = \frac{Q(y_1, y_2) \cdot Q(w_1, w_2)}{256 \cdot 64 \cdot Q(x_1, x_2)} \cdot W(x_1, y_1, w_1) \cdot W(x_2, y_2, w_2)$

with: $W(x, y, w) = \sum_i C(i, x) \cdot C(i, y) \cdot C(i, w)$

$$C(i, x) = A(x) \cos \frac{(2-i+1) \cdot x \cdot \pi}{16}; \quad A(x) = \begin{cases} \frac{1}{\sqrt{2}}, & \text{for } x=0 \\ 1, & \text{for } x \neq 0 \end{cases}$$

$U_{dct}(i, j)$ – DCT coefficients in the block
 $U_{dct, sq}(i, j)$ – DCT of the block matrix
 $U_{sq}(i, j) = U^2(i, j)$

- U_{dct} – is zero for most values
- In the $W_0(y_1, y_2, w_1, w_2, x_1, x_2)$ function only about 4% of the terms are non-zero => we can efficiently compute this sum

Fuzzy image enhancement in the compressed domain

$$IntU_{dct} = \begin{cases} U_{dct, sq} + 2 \cdot U_{dct} - 128, & \text{if } DC \leq 0 \\ 2 \cdot U_{dct} - U_{dct, sq} - 128, & \text{if } DC > 0 \end{cases}$$

$IntU_{dct}$ = the enhanced 8x8 image block in the compressed domain

- Is based on the contrast improvement method with fuzzy theory
- The luminance values are scaled symmetrically towards 0 => the threshold for the fuzzy algorithm in the compressed domain should have the value 0
- The algorithm is applied on the DCT coefficients - only the DC coefficient will be compared with the threshold for each 8x8 block

Image enhancement algorithm

- for each 8x8 pixels block of the JPEG image
 - The average amount of A.C. energy E_{AC} is computed

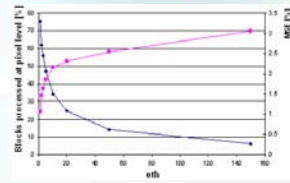
$$E_{AC} = \frac{\sum_{i,j} |U_{dct,AC}(i,j)|^2}{63}$$

- If $E_{AC} < e_{th}$ - then the block can be considered to have quasi-uniform luminance and it will be processed in the compressed domain
- If $E_{AC} > e_{th}$ - then the block has a significant content of details and it will be decompressed, every pixel from the block being processed separately (point-wise).

Note: the value e_{th} represents the selection threshold between the quasi-uniform blocks and the blocks with a significant number of details

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The e_{th} threshold selection


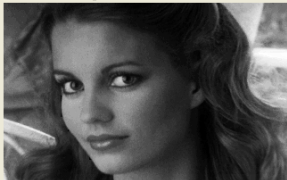


- The value e_{th} is chosen from the experiments, taking into account the image statistics
- A very small e_{th} value will always lead to very good quality of the output images,
 - but the number of processed blocks in the compressed domain will be quite small,
 - so the complexity of the computing algorithm is not significantly small compared with the direct processing on the pixel level
- An appropriate value will lead to the increase of the number of blocks processed in the compressed domain
 - \Rightarrow in this way we can obtain a fast fuzzy algorithm for image enhancement

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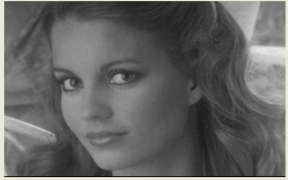
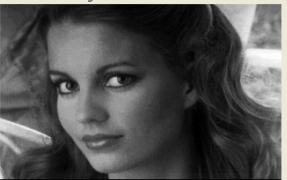
Experimental results

Energy threshold (DCT): Number of blocks processed in the compressed domain: 931
 Image luminance: Number of blocks processed at pixel level: 69 MSE = 6.49 %
 Blocks processed at pixel level [%]: 6.90 %

Original image:  Processed image: 

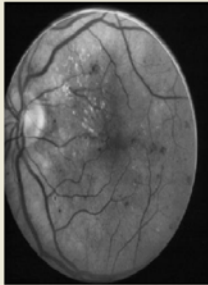
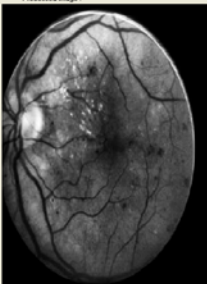
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Energy threshold (DCT): Number of blocks processed in the compressed domain: 825
 Image luminance: Number of blocks processed at pixel level: 175 MSE = 5.38 %
 Blocks processed at pixel level [%]: 17.50 %

Original image:  Processed image: 


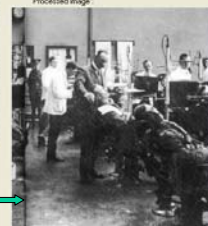
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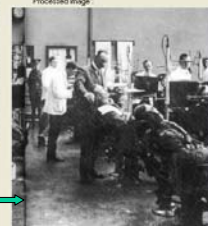
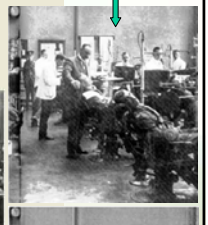
Energy threshold (DCT): Number of blocks processed in the compressed domain: 1820
 Image luminance: Number of blocks processed at pixel level: 156 MSE = 2.02 %
 Blocks processed at pixel level [%]: 7.89 %


Original image:  Processed image: 

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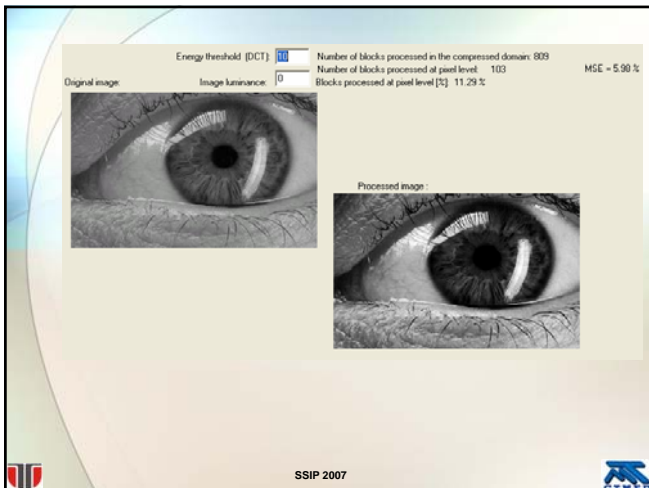
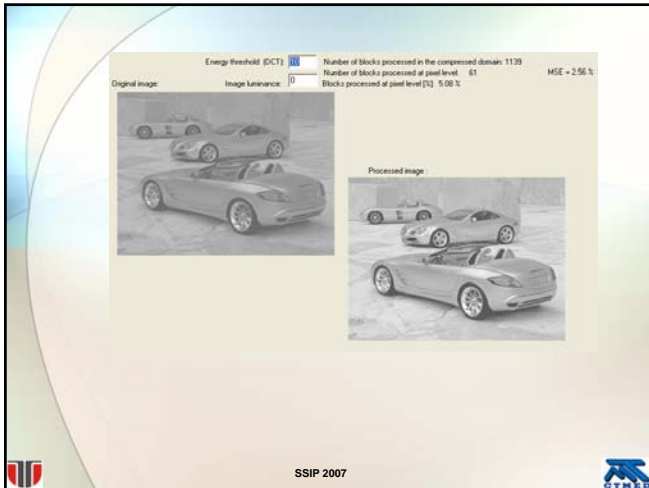
Energy threshold (DCT): Number of blocks processed in the compressed domain: 1291
 Image luminance: Number of blocks processed at pixel level: 223 MSE = 3.77 %
 Blocks processed at pixel level [%]: 15.07 %

Original image:  Processed image: 

Our algorithm \rightarrow  Histogram equalization \rightarrow 

Contrast enhancement by Tang [5] \rightarrow 

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Conclusion

- The experimental performance results show that:
 - the computational complexity of the enhancement can be significantly reduced:
 - at most 25% blocks (for complex images) are processed at pixel level
 - preserving the qualitative performance of classical algorithm

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