Improvement of Echocardiography Image based on Hybrid Technique in Data Mining

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Abstract: Image enhancement is preprocessing method for quality improvement of image Echocardiography is critical subject in information mining and machine learning and it can be broadly utilized as a part of many fields. In this dissertation a hybrid of improvement method based on wavelet transform function and neural networks is proposed. SOM were used to find correlation between noised and original DWT coefficients and approximations. Experiments results showed capability of proposed method to remove noise in terms of peak signal to noise ratio and visual quality. Different architectures and different activation functions is considered. The experimental results show the mean with the traditional enhancement methods, in this technique threshold-based enhancement digital image enhancement algorithm for mixed digital image enhancement is relatively clear, especially in the extra noise, extra complex cases”, can show its better presentations. In the improvements process in order to effort better enhancements effect, the systems takes more time to pay; the other for color digital images processings has not been a good results. Therefore, focus on late goals and improves the efficiencies of color image enhancements. after all, the algorithm has a drawbacks of needings more computing time when select a largers hybrid generations. This will be a key problem to solution in the following work.

Keywords: Digital Image, Image Enhancement, SOM, DWT, PSNR

I. INTRODUCTION

Digital images are often corrupted by noise during their acquisition and transmission. A fundamental challenge in image enhancement is to reduce noise while maintaining the desired image features such as edges, textures, and fine details. An enhancement process is normally applied to digital images to obtain more information or details contained in the images. This process is very useful for consumer electronic products which the process is normally applied in preprocessing and post processing stages. However, most recorded images are poor in contrast and having non-uniform illumination enhancement techniques use a monotonic transfer function to map the grey levels of the input images and reshape the output histogram.

II. RELATED WORK

This chapter gives an extensive literature survey on the existing image denoising method based on wavelet transform and artificial neural network and some other optimization techniques. In this study various research implement and journal and know about image denoising, wavelet transform, Neural Network and optimization techniques. All methodology and process are not described here. But some related work in the field of image compression in concern of wavelet and neural network, discuss by the name of authors and their respective title.

Yi Niue, Xiaolin Wu, Guangeming Shei Et al. this article introduces a new contrast enhancement algorithm of tone-preserving entropy maximization. Its design objective is to present the maximal amount of information content in the enhanced image, or being optimal in an information theoretical sense, while preventing the loss of tone continuity.

ParisaGifani, Haemide Behnaem, FarzaniHadedadi, Zahera AlizadeehSani, and Maeryam Shoejaieifard Et al. they introduce a novel framework that optimizes TSR enhancement of echo-cardio-graphic images by utilizing temporal information and sparse representation. The goal of this method is to increase the frame rate of echo-cardio-graphic videos, and therefore enable more accurate analyses of moving structures. For the discussed method, they first derived temporal information by extracting intensity variation time curves (IVTCs) assessed for each pixel. They then designed both low-resolution and high-resolution over complete dictionaries based on prior knowledge of the temporal signals and a set of pre-specified known functions.

Laurent Navarero, Guaneg Deneg and Guy Coubreebaisse Et al. in this implement a new extension of logarithmic image preprocessing (LIP) model, called Symmetric Logarithmic Image Processing (SLIP) Digital images are often corrupted by noise during their acquisition and transmission. A fundamental challenge in image enhancement is to reduce noise while maintaining the desired image features such as edges, textures, and fine details.
An enhancement process is normally applied to digital images to obtain more information or details contained in the images. This process is very useful for consumer electronic products which the process is normally applied in preprocessing and post processing stages. However, most recorded images are poor in contrast and having non-uniform illumination. These conditions occur due to the insufficient lighting sources and improper focusing during the image acquisition process. Insufficient illumination makes the brightness in the image unevenly distributed. Thus, image enhancement has been employed to improve the interpretability or perception of information in images. In addition, it provides better input images for further processing tasks.

V. Murugean, R. Bala-subraemanian Et al. In particular, there are two common types of noise namely Gaussian noise and Impulse noise, which are introduced during the acquisition and transmission processes. Noisy images can be found in many applications. Noise is also introduced in digital images, when a damaged image is scanned. Digital cameras may introduce noise because of CCD sensor malfunction, electronic interference or flaws in data transmission. In the last two decades, many methods have been introduced in the literature to remove either Gaussian or Impulse noise. For most users and in most applications the main purpose of image enhancement is to bring conspicuity to features and details that are otherwise obscured in the original image.

The image enhancement techniques can be roughly divided into two types: filter-based techniques and histogram-based techniques. Filter-based techniques increase image contrast by boosting high frequency components; for examples, using high-pass filters to sharpen edges and textures, or using homomorphic filtering and bilateral filtering to suppress the low-pass luminance signal and emphasize the high-pass reflectance signal. Another well-known filter-based technique that uses Gaussian filters to simulate the collaboration of retina and cortex of human visual system. The filter-based techniques can significantly improve image contrast, but they cannot preserve relative ranking of neighboring pixel values and thus are prone to “halo” artifacts.

III. PROBLEM STATEMENT

The basic idea behind this thesis is the estimation of the uncorrupted image from the distorted or noisy image, and is also referred to as image “denoising”. There are various methods to help restore an image from noisy distortion. Selecting the appropriate method play a major role in getting the desired image. The denoising methods tend to be problem specific. For example, a method that is used to Denoises. Satellites images may not be suitable for denoising medical or images. Each method is compared and classified in terms of its efficiency. In order to quantify the performance of the various denoising algorithms, a high quality image is taken and some known noise is added to it. This can then be given as input to the denoising algorithm, which produces an image close to the original high quality image. The performances of each algorithm is compared by computing Signal to Noise Ratio (SNR) besides the visual interpretation. Also we find in general problems in image denoising process used wavelet transform and artifical neural networks model.

The medical image has more noise.

A. The neighborhood smoothing method: Using the average gray value of the pixel and its neighborhood look upon as the gray value of the pixel, this method is simple, but it will make the image blurred boundaries. Therefore, in order to better image denoising. After some research denoising algorithm. Proposed a threshold based on digital image denoising hybrid algorithms. It has several features:

B. Bad peak signal to noise ratio in images of rich textures and higher visual quality in the region of texture area.

C. Difficult to design adaptable size of coded blocks according to the level of wavelet packet decomposition.

IV. PROPOSED ALGORITHM

In this section discuss the proposed algorithm of image enhanced technology based on the combination of wavelet transform function and SOM neural network model. Also used wavelet function for the optimization of unwanted attribute of image enhanced technology. For the optimization of data used wavelet transform function. Wavelet transform function optimized the data and process the data for classification. The process of classification used SOM neural network model. The steps of algorithm given below.

A. Input the selected feature from wavelet transform function and passes through SOM

B. The selected value of features F mapped in relation of Fi: \( \mathbb{R}^d \) here d is size of feature vector

C. Sampling of input features vector for the measuring similarity of features

\[
\text{sim features} = \frac{\sum_{n=1}^{d} \sum_{m=1}^{d} \delta_{nm} \cdot u(n,x,y)}{m \ast k}
\]

D. Derive the kernel function of real valued function \( K(x,y) \)

E. Measure the value of hypothesis of kernel function \( \kappa(x) = v \theta + \sum_{\text{pixels data} v/j \text{ area}} \left( \frac{\text{mevul of mah}}{s^2} \right) \) this kernel function of SOM

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F. Used product function of similar data in \((Fx \cdot Fy)\) and estimate final \(C\) data for training of class

G. \(L_0 = \text{SimFet} \prod_{i=1}^{n} \min (Fx - Fy)\) where \(L_0\) is training parameter of kernel function.

H. Assigned the trained feature as label \(c_1, c_2, \ldots, c_n\)

I. Else Find level of class in terms of product of similarity \(C = R \times X\)

J. Estimate margin value of kernel for similar class

K. End the process and feature are classified out data of image

L. Finally gets de noise image and calculate the value of peak signal to noise ratio and sim value

V. EXPERIMENTAL RESULT ANALYSIS

For the performance evaluation of histogram equalization method for Baby, Lena, Family, and Barbara image. This all image is gray scale image size is 512 * 512. Histogram equalization is basic method for image enhancement. The performance measuring parameter is peak signal to noise ratio and AMBE. Here we use various types of image enhancement techniques such as ME, MHE, Channel filter, and proposed algorithm.

Figure 1: shows that the main implementation window of image enhancement. It show blank image field with EM and Proposed techniques button.

Figure 2: Enhanced Ultra Sound Output Image Window Of Image 7 Week Using Em Methods.
Enhanced Ultra Sound Output Image Window Of Image 12 Week Using Proposed Methods

| IMAGE  | Name Of Method | Peak signal to noise ratio | similarity index measure |
|--------|----------------|----------------------------|--------------------------|
| 12 Week| EM             | 22.05                      | 0.220                    |
|        | Proposed       | 21.84                      | 0.218                    |

Table 1: Comparative values of 12 Week image using EM and Proposed Method. Find the resultant values of peak signal to noise ratio and similarity index measure.

Comparative result analysis for 12 Week image using EM and Proposed Methods

Figure 4: Comparative Result Graph For 12 Week Using EM And Proposed Methods, Here We Found The Value Of Peak Signal To Noise Ratio And Similarity Index Measure.
VI. CONCLUSION AND FUTURE WORK

In this dissertation, we proposed a hybrid method for image enhancement for echo-cardiographic images. In this thesis, better results are compared to traditional methods of improvement. However, the computational time of the process is increased. In this technique, threshold-based enhancement of digital images is relatively clear, especially in the extra noise, extra complex cases, and can show its better presentations. In the improvements process, in order to effort better enhancements, the systems take more time to pay; the other for color digital images processing has not been a good result. Therefore, focus on the latest goals and improvements of color image enhancements, after all, the algorithm has drawbacks of needing more computing time when select a larger hybrid generations. This will be a key problem to solution in the following work. We used in future optimization technique for the minimize of time and upgrade of quality of image.

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