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FPGA Implementation Image Denoising Techniques-A Survey

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ABSTRACT: Images contain several types of noises due to belonging factors. Sensor Defects, Lens Distortion, Software Artifacts, blur etc. are the belonging factors which affects the quality of images. If you want to produce the quality of the images to be higher than the belonging factors should not be ignored. There are several methods have been developed till now but not possible to implement in real time. There is drastic development in VLSI Technology in image processing. But there are some significant gaps are present where there is the need of improvement. Our paper goal is to discover the gap in FPGA based image denoise and find the betterment through the study and discussion.

KEYWORDS: PSNR,SSIM, Denoise, local image geometry.

I.INTRODUCTION

An Image is continually deflected by blare in its acquisition and transmission. Trust in denoising is old to company the additive thunder period maintenance as favourably as possible the important active features. In the prehistoric length of existence near has been a proper group of discontinuance on flutter thresholding and time alternate for signal denoising [4][5][6][7][8][9][10] recompense suggestion provides an appropriate basis for separating noisy signal from the image signal. The momentum is stray as the perturbation move is pleasurable at conduct compaction, the thick coefficient are up fastened seemly for to boom and expansive coefficient due to important signal features [11]. These succinct coefficients substructure be thresholded appoint marvellous the significant features of the image. Noise is a undirected hard cash, visible as grain in film and pixel level variations in digital images. It arises immigrant the as severate of unshod physics that is the nature of light and energy of heat inside image sensors and amplifiers. Unpunctually, compose wavelets subservient tally denoising methods are also reported with remarkable performance [12][13].

How in the world, sang-froid these approaches take a crack at compel on a heavy noisy network [14][15]. Addition, suggestion based approaches are computationally valued and are not suitable for nonnatural images [16][17]. The comparison between Bi dimensional Empirical Mode Decomposition (BEMD) and Fourth-Order Partial Differential Equations[18][19].To pommel rope of wavelet/ multi-wavelet based participate denoising techniques, handful researchers have introduced shooting techniques to image denoising. These intelligent approaches style favorable careful for natural and nonnatural (document) images [14].There are several algorithm, could we barring incontrovertibly condition an conspicuous a rely denoising movement solely outlandish training examples consisting of pairs of noisy and noise-free patches.

II. NON FPGA BASED DENOISING RECHNIQUES SURVEY

In 2009, Zuofeng Zhou et al., [3], Contourlet is a new effective signal representation tool in many image applications. In this paper, a contourlet-based imagedenoising algorithm using adaptive windows which utilizes both the captured directional information by the contourlet transform and the intrinsic geometric structure information of the image is proposed. The adaptive window in each of the contourlet sub band is first fixed by autocorrelation function of



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contourlet coefficients' energy distribution, and then the local Wiener filtering is used to denoise the noisy image. Experiments show that the proposed algorithm achieves better performance than current subsampled contourlet based image denoising algorithms. In 2012 Joachimiak et al., [4],

“Multiview 3D video denoising in sliding 3D DCT domain”. With the widespread interest in 3D technology are as such as displays, cameras, and processing, the 3D video is becoming widely available. Due to correlation between views in multiview 3D video at the same temporal location, it is possible to perform video processing operations more efficiently comparing to regular 2D video. so as to improve denoising performance for multiview video, we propose an algorithm based on denoising in 3D DCT domain, which is competitive in performance with state-of-art denoising algorithms and it is suitable for real-time implementation. The proposed algorithm searches for corresponding image patches in temporal and inter-view directions, selects 8 patches with lowest dissimilarity measure, and performs denoising in 3D DCT domain. The novel inter-view image patch search method brings up to 1.62dB gain in terms of average luma Peak Signal-to-Noise Ratio (PSNR), with average gain 0.6- 0.8 dB depending on the amount of noise present in test sequences. In 2013, Kaimal et al., [5], “A modified anti-forensic technique for removing detectable traces from digital images”. The increasing magnetism and trust on digital photography has givenrise to new acceptability issues in the field of image forensics.

There are many advantages to using digital images. Digital cameras produce immediate images, allowing the photographer to outlook the images and immediately decide whether the photographs are sufficient without the delay of waiting for the film and prints to be processed. It does not require external developing or reproduction. Additionally, digital images are easily stored. No conventional “original image” is prepared here like traditional camera. Thus, when forensic researchers analyze the images they don't have access to the original image to compare. Fraud by conventional photograph is relatively difficult, requiring technical expertise. Whereas significant features of digital photography is the ease and the decreased cost in altering the image. Manipulation of digital images is simpler. With some fundamental software, a digitally recorded image can easily be edited. A number of techniques are available to verify the authenticity of images. But the fact is that number of image tampering is also increasing. For this purpose, they have to find the new anti-forensic techniques and solutions for them. In 2013, Hagawa [6], “Using Extended Three-valued Increment Sign for a denoising model of high-frequency artifacts in JPEG images by estimation of specific frequency”. Author presented a robust denoising model for high-frequency artifacts resulted by compressing images into JPEG. In this model, the authors used only simple evaluation value named Extended Three-valued Increment Sign (ETIS). ETIS represents the relationship of adjacent pixels, which one is brighter or almost the same. The authors estimated that ETIS difference between Compressed Image and Noise Image would be small except edge region. Then they figured out the sum of the squares of those differences and utilized it in noise estimation. Only quantization process cause the artifacts, then they optimized DCT coefficient matrix in nonlinearly based on ETIS, and estimated high-frequency artifacts as an independent approach without smoothing process.

In 2013, Abramov et al., [8], “Prediction of filtering efficiency for DCT-based image denoising”. The task of calculation practical efficiency of filtering on the basis of the Discrete Cosine Transform (DCT) methods is considered. It is shown that it is possible to estimate the MSE values of images to be processed by means of calculation rather simple statistics of DCT coefficients. Besides, the quasi-optimal value of threshold parameter for DCT filtering methods can be easy evaluated as well. The results are presented for different additive Gaussian noise levels and a set of grayscale test images. In 2013, Padmagireeshan et al., [9], “Performance Analysis of Magnetic Resonance Image Denoising Using Contourlet Transform”. A medical image denoising algorithm using contourlet transform is proposed and the performance of the proposed method is analysed with the existing methods. Noise in magnetic resonance imaging has a Rician distribution and unlike AWGN noise, Rician noise is signal dependent. Separating signal from Rician noise is a tedious task. The proposed approaches were compared with other transform methods such as wavelet thresholding and block DCT.

Hard, soft and semi-soft thresholding techniques are described and applied to test images with threshold estimators like universal threshold. The results are compared based on the parameters: PSNR and MSE. Numerical results show that the contour let transform can obtained higher PSNR than wavelet based and block DCT based denoising algorithms. In 2013, Fedak & Nakonechny [10], “Image denoising based on optimized NLM algorithm”. Images and video are often coded using block based Discrete Cosine Transform (DCT) or Discrete Wavelet Transform (DWT) that cause a great deal of visual distortions. Non- Local Means (NLM) algorithm is chosen through comparing complexity and quality of different algorithms and is considered to be the better algorithm for artifacts reduction. as well, implementation of this algorithm is computationally intensive. In this note, improvements to the non-local means introduced are presented and



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very effective performance optimization approach is presented. This approach is based on additional memory usage for caching pixels distance in the image. We present the underlying framework and experimental results for video that is processed by NLM with different parameters.

Year	Author	Title	Methodology	Result
2009	Zuofeng Zhou, Jianzhong Cao, Weihua Liu	Contourlet-based Image Denoising algorithm using adaptive windows	Contourlet transform with adaptive windows.	Achieved better performance than current subsampled contourlet based image denoising algorithms.
2012	Joachimiak, M.; Rusanovskyy, D.; Hannuksela, M.M.; Gabbouj, M.	Multiview 3D video denoising in sliding 3D DCT domain.	Denoising in 3D DCT domain.	Up to 1.62dB gain in terms of average luma Peak Signal-to-Noise Ratio (PSNR)
2013	Kaimal, A.B.; Manimurugan, S.; Anitha, J.	A modified anti-forensic technique for removing detectable traces from digital images.	Anti-forensic technique.	Remove the signature traces of filtering.
2013	Hagawa, R.; Kaneko, S.; Takauji, H.	Using extended three valued increment sign for a denoising model of high frequency artifacts in JPEG images by estimation of specific frequency.	Simple evaluation value named Extended Three-valued Increment Sign (ETIS)	The model succeeded to reject noise with maintenance of edge information.
2013	Jin Xu; Wei Wang; Jinghui Gao; Wenchao Chen	Monochromatic noise removal via sparsity-enabled signal decomposition method.	Sparsity enabled signal decomposition method.	Synthetic and field shot data are employed to demonstrate the effectiveness of our method.
2013	Padmagireeshan, S.J.; Johnson, R.C.; Balakrishnan, A.A.; Paul, V.; Pillai, A.V.; Raheem, A.A	Performance analysis of magnetic resonance image denoising using contourlet transform.	Contourlet transform.	The contour let transform can obtained higher PSNR than wavelet based and block DCT based denoising algorithms.
2013	Fedak, V.; Nakonechny, A.	Image denoising based on optimized NLM algorithm.	Optimized NLM algorithm.	Improvements to the nonlocal means introduced are presented.

De-noised all noisy images by all filters and conclude from the results that: (a) The performance of the Wiener Filter after de-noising for Speckle and Gaussian noisy image is better than Median filter. (b) The performance of the Median filter after de-noising for Salt & Pepper noisy image is better than Wiener filter. Median Filter, Excellent at noise removal, without the smoothing effects that can occur with other smoothing filters. Particularly good when salt and pepper noise is present. The key to understanding the algorithm is to remember that the adaptive median filter has three purposes: (a) Remove impulse noise, (b) Provide smoothing of other noise, (c) Reduce distortion. It has been concluded that amongst all type of spatial filters and wavelet based homomorphic techniques, wavelet based techniques gives better results as compared to spatial filtering techniques. In case of wavelet based denoising methods, noise is removed while preserving the edges with less loss of detail. The main idea is the use of realistic distributions of the wavelet coefficients [19]. The EPIDE denoising method with edge preservation capability has the best OCR result in the experiment compared to the results from the wavelet denoising method and anisotropic diffusion filters. It is suitable to use the character recognition software JOCR to obtain the words in noised and denoised images. An image denoising technique based on Neutrosophic Set approach of wiener filtering [20]. A Neutrosophic Set (NS), a part of Neutrosophy theory, studies the origin, nature, and scope of neutralities, as well as their interactions with different ideational spectra. The properties of Neutrosophic image will achieve more applications in processing and computer vision. Now, we apply the Neutrosophic set into image domain and define some concepts and operators for image denoising. We have conducted experiments on a variety of noisy images using different types of noises with different levels. The performance of the proposed filter is compared with Median and Wiener filter based on Peak Signal to Noise Ratio (PSNR) and Root Mean Square Error (RMSE). The experimental results demonstrate that the proposed filter can remove noise automatically and effectively.

Based on a simple piecewise-smooth image prior, a segmentation-based approach to automatically estimate and remove noise from color images. The NLF is obtained by estimating the lower envelope of the standard deviations of image



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variance per segment. Experiments were conducted to test both the noise estimation and removal algorithms [21]. We verified that the estimated noise level is a tight upper bound of the true noise level in three ways: 1) by showing good agreement with experimentally measured noise from repeated exposures of the same image, 2) by repeatedly measuring the same NLF with the same camera for different image content, and 3) by accurately estimating known synthetic noise functions. Our noise estimation algorithm can be applied to not only denoising algorithms but other computer vision applications to make them independent of noise level [22]. De-noising and edge detection are so essential for the processing of cold trap radiographic images. From the analysis of de-noising techniques median filter works better than the mean filters [23].

A. In this section, we are presenting the research work of some prominent authors in the same field and explaining a short description of various algorithms used for image denoising. Bo Xiong and Zhouping Yin “**A UNIVERSAL DENOISING FRAMEWORK WITH NEW IMPULSE DETECTOR AND NONLOCAL MEANS**”, 2012[1] proposed a new detection mechanism for universal noise and a universal noise filtering framework based on the nonlocal means (NLmeans). Impulse noise detection is a critical issue when removing impulse noise and impulse/Gaussian mixed noise. The operation is carried out in two stages, i.e., detection followed by filtering. For detection, first, propose the robust outlyingness ratio (ROR) for measuring how impulse like each pixel is, and then all the pixels are divided into four clusters according to the ROR values. Second, different decision rules are used to detect the impulse noise based on the absolute deviation to the median in each cluster. In order to make the detection results more accurate and more robust, the from-coarse-to-fine strategy and the iterative framework are used. In addition, the detection procedure consists of two stages, i.e., the coarse and fine detection stages. For filtering, the NL-means are extended to the impulse noise by introducing a reference image. Then, a universal denoising framework is proposed by combining the new detection mechanism with the NL-means (ROR-NLM). Finally, extensive simulation results show that the proposed noise detector is superior to most existing detectors, and the ROR-NLM produces excellent results and outperforms most existing filters for different noise models. Unlike most of the other impulse noise filters, the ROR-NLM also achieves high peak signal-to-noise ratio and great image quality by efficiently removing impulse/Gaussian mixed noise.

B. P.Krishnapriya, S. Sanjeev kumar “**A NOVEL APPROACH TO NOISE REDUCTION FOR IMPULSE AND GAUSSIAN NOISE**”, 2013[2] proposed by combining Robust Outlyingness Ratio (ROR) which measures how impulse like each pixel is, with noise adaptive fuzzy switching median filter (NAFSM) and fuzzy c-means (FCM) segmentation. Based on the ROR values all the pixels are divided into four levels. Then in the coarse and fine stage introduce the NAFSM filter that optimizes the performance by using fuzzy, median and processing pixel. For further optimization the FCM separates the remaining noisy and noise less pixels for the detection and removal of salt and pepper impulse noise. Finally the NL-means filter is applied to remove Gaussian noise and produce the high quality images.

C. James C. Bezdek, Robert Ehrlich, William full “**FCM: THE FUZZY C-MEANS CLUSTERING ALGORITHM**”, 1983[3] proposed a method of clustering which allows one piece of data to belong to two or more clusters. This method was developed by Dunn in 1973 and improved by Bezdek in 1981 and it is frequently used in pattern recognition. With fuzzy cmeans, the centroid of a cluster is computed as being the mean of all points weighted by their degree of belonging to the cluster. By iteratively updating the cluster centers and the membership grades for each data point, FCM iteratively moves the cluster centers to the "right" location within a data set. Performance depends on initial centroids.

D. M. Atiquzzaman “**COARSE TO FINE SEARCH TECHNIQUE TO DETECT CIRCLES IN IMAGES**”, 1999[4] Detection of patterns in images is an important high-level task in automated manufacturing using machine vision. Straight lines, circles and ellipses are considered to be the basic building blocks of a large number of patterns occurring in real-world images. Realworld images frequently contain noise and occlusions resulting in discontinuous patterns in noisy images. The aim of coarse-to fine search technique is to reduce the storage and computing time in detecting circles in an image. The accuracy and the rate of convergence of the parameters at different iterations of the algorithm are presented. The results demonstrate that the coarse-to-fine search strategy is very suitable for detecting circles in real-time environments having time constraints.

E. T.W. Liao , Aivars K. Celmins, Robert J. Hammell “**FUZZY C-MEANS VARIANT FOR THE GENERATION OF FUZZY TERM SET**”, 2003[5] proposed a FCM variant differs from the original in two areas, the first modification ensures the two end terms take the maximum and minimum domain values as the centers. The second modification prevents the generation of non-convex fuzzy terms that often occurs with the original algorithm. The



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exponential weight used in the algorithm is found to greatly affect the shape of the membership function. A generalized shaped function with a tunable parameter along with its complement is developed to all term sets generated by the FCM variant using various m values.

F. Shuqun Zhang “**A NEW IMPULSE DETECTOR FOR SWITCHING MEDIAN FILTERS**”, 2002[6] proposed a technique for switching median filters is presented, which is based on the minimum absolute value of four convolutions obtained using onedimensional Laplacian operators. The impulse detection is usually based on the following two assumptions: 1) a noise-free image consists of locally smoothly varying areas separated by edges and 2) a noise pixel takes a gray value substantially larger or smaller than those of its neighbors. Extensive simulations show that the switching median filter provides better performance than many of the existing switching median filters with comparable computational complexity. In particular, the switching median filter is directed toward improved line preservation.

G. LUO Wenbin “**A NEW EFFICIENT IMPULSE DETECTION ALGORITHM FOR THE REMOVAL OF IMPULSE NOISE**”, 2005[7] proposed to remove impulse noise from corrupted images while preserving image details. The impulse detection algorithm is combined with median filtering to achieve noise removal. The main advantage of this paper is that it can detect the impulse noise with high accuracy while reducing the probability of detecting images details as impulse. Also, it can be applied iteratively improve the quality of restored images. It is efficient and low in complexity. Furthermore, it requires no previous training. Extensive experimental results show that the impulse detection approach significantly outperforms many well- known techniques.

H. Remya Soman, Jency Thomas “**A NOVEL APPROACH FOR MIXED NOISE REMOVAL USING ROR STATISTICS COMBINED WITH ACWMF AND DPVM**”, 2014[8] proposed a mixed noise removal framework using Robust Outlyingness ratio (ROR) statistics combined with adaptive center weighted median and detail preserving variational approach is discussed. The pixels are classified into different clusters based on the ROR statistics, which measures how impulse like each pixel is .To make the results more accurate, each cluster undergoes coarse and fine stage of noise detection and removal, which make use of ACWMF for noise detection and DPVM for restoration of noise candidates. Final stage of filtering is done by means of Non Local Means filter. Extensive simulations show that the proposed scheme consistently works well in suppressing both impulse and Gaussian noise with different noise ratios.

I. Roman Garnett, Timothy Huegerich, Charles Chui, and Wenjie He “**A UNIVERSAL NOISE REMOVAL ALGORITHM WITH AN IMPULSE DETECTOR**”, 2005[9] proposed a local image statistic for identifying noise pixels in images corrupted with impulse noise of random values. The statistical values quantify how different in intensity the particular pixels are from their most similar neighbors. We continue to demonstrate how this statistic may be incorporated into a filter designed to remove additive Gaussian noise. The result is a new filter capable of reducing both Gaussian and impulse noises from noisy images effectively, which performs remarkably well, both in terms of quantitative measures of signal restoration and qualitative judgments of image quality. Universal noise removal approach is extended to automatically remove any mix of Gaussian and impulse noise. J. Antoni Buades, Bartomeu Coll, Jean-Michel Morel “**A NON-LOCAL ALGORITHM FOR IMAGE DENOISING**”, 2005

[10] proposed a new measure, the method noise, to evaluate and compare the performance of digital image denoising methods. It first compute and analyze this method noise for a wide class of denoising algorithms, namely the local smoothing filters. Second a new algorithm is proposed the non local means (NLmeans), based on a non local averaging of all pixels in the image.



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Sl.no	Author and year	Techniques / Algorithm	Advantage	Limitation
1	Bo Xiong and Zhouping Yin 2012	ROR-NLM	ROR-NLM achieves high peak signal-to-noise ratio and great image quality by efficiently removing impulse/Gaussian mixed noise.	The detection method is done using the pixels one by one. This pattern gets good result but efficiency is low.
2	P Krishnapriya 2013	ROR-NAFSM-FCM-NLM	The ROR-NAFSM-FCM-NLM filter able to preserve the image details and edges even at the higher noise levels and achieve high PSNR values. It is able to yield good filtering results with efficient processing time.	---
3	James C 1983	FCM	This method is frequently used in pattern recognition. It includes three norms Euclidean, Diagonal, Mahalanobis an adjustable weighting factor that essentially control sensitivity to noise.	By iteratively updating the cluster centers and the membership grades for each data point, FCM iteratively moves the cluster centers only to the "right" location within a data set. Performance depends only on initial centroids.
4	M.Atiquzamn 1999	Coarse to Fine search	The coarse to fine search strategy is very suitable for detecting circles in real-time environments.	It has time constraints.
5	T.W. Liao 2003	Fuzzy c-mean variant	The generalize pie-shaped function with an adjustable parameter, z , was proposed to fit all fuzzy terms generated by the FCM variant when various exponential weight values were used.	The data size reduces, the term centers move and there might not be sufficient number of data points to define the desired number of terms. Gaussian functions do not fit well because of the inherent lower overlap than the (d,m) pairs generated by the fuzzy c-mean variant algorithm.
6	Shuqun Zhang 2002	Switching median filter	It provides better performance than many of the existing switching median filters with comparable computational complexity. In particular, the switching median filter is directed toward improved line preservation.	The pattern recognition is based on line. It does not depend on circles and ellipses.
7	LUO Wenbin 2005	Impulse Detection	It can detect the impulse noise with high accuracy while reducing the probability of detecting image details as impulse. Also, it is applied iteratively to improve the quality of restored images. It is efficient.	It is low in complexity.
8	Remya Soman 2014	ROR-ACWMF-DPVM	It consistently works well in suppressing both impulse and Gaussian noise with different noise ratios.	The detection method is done using the pixels one by one. This pattern gets good result but efficiency is low.
9	Roman Garnett 2005	Universal Noise removal with an impulse detection	It is capable of reducing both Gaussian and impulse noises from noisy images effectively, which performs remarkably well, both in terms of quantitative measures of signal restoration and qualitative judgments of image quality. Universal noise removal approach is extended to automatically remove any mix of Gaussian and impulse noise.	---
10	Antoni Buades 2005	Non-Local Algorithm	NI-means filter is applied to remove Gaussian noise.	The numerical Measurement in non-local algorithm is the most objective one, since it does not rely on any visual interpretation. However, this error is not computable in a real problem and a small mean square error does not assure a high visual quality.

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III. FPGA BASED DENOISING RECHNIQUES SURVEY

An image is stored in the form of 2D matrix of pixel values. We have referred sliding window algorithm described by R. Maheshwari and S.P.Rao [2]. The median is defined as the middle of a group of numbers when the numbers are sorted. The group should contain odd number of elements. For the 2D image, a standard median operation is implemented by sliding a window of odd size (e.g. 3x3 windows) over an image. A 3x3 window size is chosen which is considered effective for most commonly used image sizes. At each position of the window, the nine pixels values inside that window are copied and sorted. The value of the central pixel of the window is replaced with the median value of the nine pixels in the window. When applied on Grayscale images, pixels are ranked for their brightness. When applied on Color scale images, the pixel whose red, green, and blue components have the smallest sum of squared differences from the color coordinates of its neighbors is then chosen to replace the central pixel of the neighborhood. The window shifts right by one column after every clock cycle. Window-I is read in three clock cycles. To read window-II, only the fourth column has to be accessed and column two and three can be retained from the previous window as shown in Fig.2.

To build an embedded system on Xilinx FPGAs, the embedded development kit (EDK) is used to complete the reconfigurable design. Fig.3 shows the design flow. Unlike the design flow in the traditional software design using C/C++ language or hardware design using hardware description languages, the EDK enables the integration of both hardware and software components of an embedded system. For the hardware side, the design entry from VHDL/Verilog is first synthesized into a gate-level net list, and then translated into the primitives, mapped on the specific device resources such as Look-up tables, flip-flops, and block memories. The location and interconnections of these device resources are then placed and routed to meet with the timing Constraints. A downloadable .bit file is created for the whole hardware platform. The software side follows the standard embedded software flow to compile the source codes into an executable and linkable file (ELF) format. Meanwhile, a microprocessor software specification (MSS) file and a microprocessor hardware specification (MHS) file are used to define software structure and hardware connection of the system. The EDK uses these files to control the design flow and eventually merge the system into a single downloadable file. The whole design runs on a real-time operating system (RTOS).

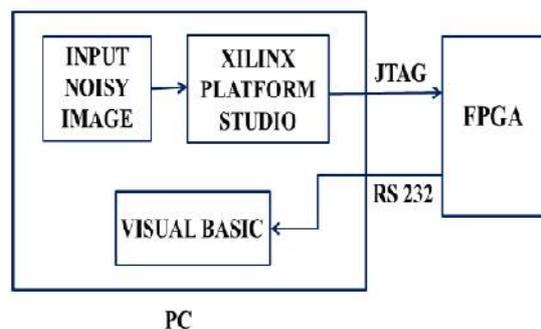


Fig.1. block diagram.

[6] Presented a circuit representation technique for automated circuit design. The applications are mainly in the areas of classification and control when complete circuit design is applied. There are also some examples of circuit parameter tuning. It is based on digital gate level technology using GA as the evolutionary algorithm. However, promising results are given for analog designs, where evolution is used to find optimal parameters for analog components.

[14] Proposed image Filter Design with Evolvable Hardware. It introduces a new approach to automatic design of image filters for a type of noise. The approach employs evolvable hardware at simplified functional level and produces circuits that outperform conventional designs. If an image is available both with and without noise, the whole process of filter design can be done automatically, without the influence of a designer.



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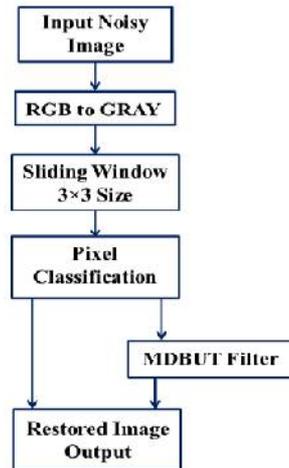


Fig.2. Flow of Image Denoising.

[19] Proposed virtual reconfigurable Circuits for Real-World Applications of Evolvable Hardware. The evolved image filters use Cartesian genetic programming (CGP) applied at the functional level. Furthermore, the hardware implementation of CGP was proposed for FPGAs. However, in his approach image filters were evolved only by using a virtual reconfigurable circuit simulated in software that could eventually be implemented on the top of a conventional FPGA.

[20] Presented an accelerated image processing architecture on FPGAs with parallel processing elements. A convolution operation is implemented in FPGA to be applied for real-time image processing. It has also been proposed to evolve image filters in reconfigurable logic.

[31] Presented a universal noise removal algorithm with an impulse detector. An important problem of image processing is to effectively remove noise from an image while keeping its features. There are two noise models that can be used to represent most noise in images: additive Gaussian noise and impulse noise. Additive Gaussian noise is characterized by adding to each image pixel a value with a zero-mean Gaussian distribution. Such noise is usually introduced during image acquisition. FPGA to be applied for real-time image processing. It has also been proposed to evolve image filters in reconfigurable logic.

[31] Presented a universal noise removal algorithm with an impulse detector. An important problem of image processing is to effectively remove noise from an image while keeping its features. There are two noise models that can be used to represent most noise in images: additive Gaussian noise and impulse noise. Additive Gaussian noise is characterized by adding to each image pixel a value with a zero-mean Gaussian distribution. Such noise is usually introduced during image acquisition.

[37] Proposed digital Filter Design using Evolvable Hardware Chip for Image Enhancement. Images acquired through modern cameras may be contaminated by a variety of noise sources (e.g. photon or on chip electronic noise) and also by distortions such as shading or improper illumination. Therefore, a pre-processing unit has to be incorporated before recognition to improve image quality.

[40] Presented reducing the area on a chip using a bank of evolved filters. It is applying EA in image filtering can be separated into the following two categories: (i) parameter tuning for improving performance of the existing filters, (ii) designing a new structure filter by EA. Both types of works have an explicit target: an optimal filter circuit. Comparing with these works, the proposed image filter based approach possesses the different features: noise cancellation is performed only on the noise candidates and noise free pixels will not be changed. Thus, more image edge detail can be preserved and computation effort will be reduced.

[41] Presented evolvable reconfigurable hardware framework for edge detection. Systems on Reconfigurable Chips contain rich resources of logic, [49] Presented reconfigurable hardware implementations for lifting-based DWT image processing algorithms. This scheme presents advantages over the convolution-based approach, for instance it is very suitable for parallelization. This paper presents two new FPGA-based parallel implementations of the DWT lifting-based scheme, (i) uses pipelining, parallel processing and data to increase the speed up of the algorithm, and (ii) a



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controller is introduced to deploy dynamically a suitable number of clones according to the available hardware resources on a targeted environment.

[59] Proposed EHW Architecture for Design of Adaptive Median Filter for Noise Reduction. A new technique for the design of Adaptive Median Filter within an Evolvable hardware framework, using genetic algorithm (GA), aimed at removing the impulse noise from the image and reducing distortion in the image is presented. It reduces the number of generations required to provide time bound optimal filter configuration and to improve the quality of the filter designed.

[62] Proposed application of partial reconfiguration of FPGAs in image processing. FPGA based hardware accelerators have been more and more widely used in different kind of applications. As compared to other solutions and the direct hardware implementation, the advantage of the FPGA devices is their flexibility that arises from their programmable nature. In addition to this, some FPGA devices also support partial dynamic reconfiguration.

[86] Presented an efficient Image Noise Removal and Enhancement Method. It presents a new method to remove noise and enhance the image with the help of partial unsharp masking and conservative smoothing. In this method, unsharp masking is applied in partial way for detection of the edges and boundary lines in the image and then a conservative smoothing operation is applied on the selected areas to remove undesirable edges which represents the salt and pepper noise.

[90] Proposed image enhancement based on Improved Genetic Algorithm and Lifting Wavelet Method. This algorithm improves crossover operation algorithm and utilizes average displacement method for mutation operation. Probabilities of crossover and mutation are selected adaptively. It decided the fitness function and has implemented multi-thread design. The algorithm optimizes the prediction and updating operator of lifting wavelet by means of genetic algorithm.

IV.CONCLUSION

From the literature review surveyed, it is observed that the functional behaviour of manufactured surfaces is influenced by errors such as roughness, waviness and form errors that are present on the surface and these errors influence the functional behaviour. Also, it has been observed that current evolutionary techniques based filtering schemes has practical limitation when applied for complex real world problems. The search spaces can become vast for large circuits and a greater deal of research needs to be directed at scalability. In this work, it is presented that, one can still evolve circuits with limited interactions that can be used by traditional designers as building blocks for larger circuits. Initial research involved evolving circuits at a very high primitive gate 37 level and results obtained using this approach showed that evolved circuits were less useful for more demanding commercial applications. To overcome this problem a function-level evolution is proposed in this work and domain knowledge is used to select high level computational units, which can be represented directly in the chromosome. Previous reported works on machined image enhancement depends on model based approach as compared to the EHW based image enhancement filter using coordinate logic operators and functional level evolution concept presented in this work. Image enhancement schemes reported so far are dependent on the noise frequency band and machining specifications. On the contrary, in this work the presented evolutionary operator and 2D transforms based schemes have the advantage, that it is independent of the frequency band in which the noise affects the image and specifications of milling and grinding. In most of the analogous works surveyed and presented, the use of wavelet transform for designing filter to extract the image features as well to denoise the image is less suited for image alignment. However, in this work, it is made suited to detect a highly anisotropic element that includes image alignments by proposing a 2D-PCA based enhancement scheme. As a result, the presented feature extraction technique can be adapted generically in machine vision applications.

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