



A New Technique for Elimination of Harmonics Using Three Phase Shunt Active Filter

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ABSTRACT: This paper describes the current-detection algorithm based on the time-domain approach for three-phase shunt active power filters (APFs) to eliminate harmonics, and/or correct power factor, and/or balance asymmetrical loads is analyzed in this paper. A basic overview and evaluation of the performance of existing current-detection algorithms for active power filters are presented. According to different complicated power quality issues and various compensation purposes, a novel current-detection algorithm is then proposed. Comparing with existing algorithms, this algorithm has shorter response time delay and clear physical meaning. Different compensating current references can, thus, be accurately and easily obtained by adopting the proposed algorithm. It ensures that the shunt APF can very well achieve different compensation purposes. Simulation results obtained with MATLAB and testing results on an experimental shunt APF are presented to validate the proposed algorithm.

KEYWORDS: Active Power Filter(APF), Current-detection algorithm, harmonic compensation, reactive power compensation, time domain.

I. INTRODUCTION

Image denoising is a technique which removes out noise which is added in the original image. Noise reduction is an important part of image processing systems. An image is always affected by noise. Image quality may get disturbed while capturing, processing and storing the image. Noise is nothing but the real world signals and which are not part of the original signal. In images, noise suppression is a particularly delicate task. In this task, noise reduction and the preservation of actual image features are the main focusing parts. The wavelet transform provides a multiresolution [1] representation using a set of analyzing functions that are dilations and translations of a few functions (wavelets). A denoising method is used to improve the quality of image corrupted by a lot of noise due to the undesired conditions for image acquisition. The image quality is measured by the peak signal-to-noise ratio (PSNR) or signal-to noise ratio (SNR). Traditionally, this is achieved by linear processing such as Wiener filtering [3]. Recently introduced Dual-Tree Complex wavelet transforms and Multilevel Dual-Tree Complex DWT can give best results in image denoising applications.

II. RELATED WORK

The transform is an function of change image one forum into forum .it classified many types the one transform is wavelet transform main function of wavelet transform is real and imaginary values into orthonormal series in this transform used for specific applications. Compare to slant and haar transform the both algorithms are used for image denoising depend upon applications function of slant transform One shape image into another shape image using transformation the image horizontal or vertical shapes using mathematical representation. The wavelet transform is computed separately for different segments of the time-domain signal at different frequencies. Multi-resolution analysis: analyzes the signal at different frequencies giving different resolutions [2]. Multi-resolution analysis is designed to give good time resolution and poor frequency resolution at high frequencies and good frequency resolution and poor time resolution at low frequencies. Good for signal having high frequency components for short durations and low frequency components for long duration, e.g. Images and video frames This is a newly introduced technique of DWT. Orthogonal wavelet decompositions, based on separable, multirate filtering systems have been widely used in image and signal processing, largely for data compression. Kingsbury introduced a very elegant computational structure, the Dual-Tree complex wavelet transform which displays near-shift invariant properties. Other constructions can be found such as in .Kingsbury pointed out the problems of Mallat-type algorithms. These algorithms have the lack of shift invariance.

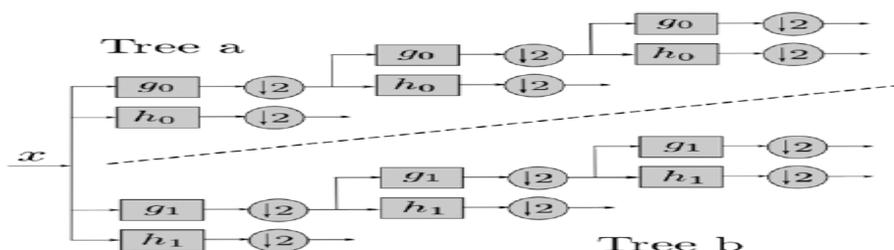


Fig 1. Structure for Wavelet Transform

III. DUAL TREE COMPLEX WAVELET TRANSFORM (DTCWT)

In this system complex value extended to discrete forum .our work with complex wavelets for motion estimation [5] showed that complex wavelets could provide approximate shift invariance. Unfortunately we were unable to obtain PR and good frequency characteristics using short support complex FIR filters in a single tree (e.g., Fig.2). However, we observed that we can also achieve approximate shift invariance with a real DWT, by doubling the sampling rate at each level of the tree. For this to work, the samples must be evenly spaced. One way to double all the sampling rates in a conventional wavelet tree, such as tree *a* Fig.3, is to eliminate the down-sampling by 2 after the level 1 filters. For linear phase filters, this requires odd-length filters in one tree and even-length filters in the other. Greater symmetry between the two trees occurs if each tree uses odd and even filters alternately from level to level, but this is not essential. Note the vertical alignment of these bases at each scale, such that the tree *b* scaling functions interpolate midway between those of tree *a*, while the tree *b* wavelets are aligned with those of tree *a*, but with a quadrature phase shift in the underlying oscillation .To invert the DT CWT, each tree in Fig.2 is inverted separately using orthogonal filters $G_{...}$, designed for perfect reconstruction with the corresponding analysis filters $H_{...}$ in the 2-band reconstruction block, shown lower right. Finally the two tree outputs are averaged in order to obtain an approximately shift invariant system.

This system is a wavelet frame with redundancy two; and if the filters are designed such that the analysis and reconstruction filters have very similar frequency responses (i.e., are almost orthogonal), then it is an almost tight frame, which means that energy is approximately preserved when signals are transformed into the DT CWT domain[3]. The basic functions were obtained by injecting unit pulses separately into the inverse DT CWT at each scale in turn. The real and imaginary parts were obtained by injecting the unit pulses into trees *a* and *b* in turn.

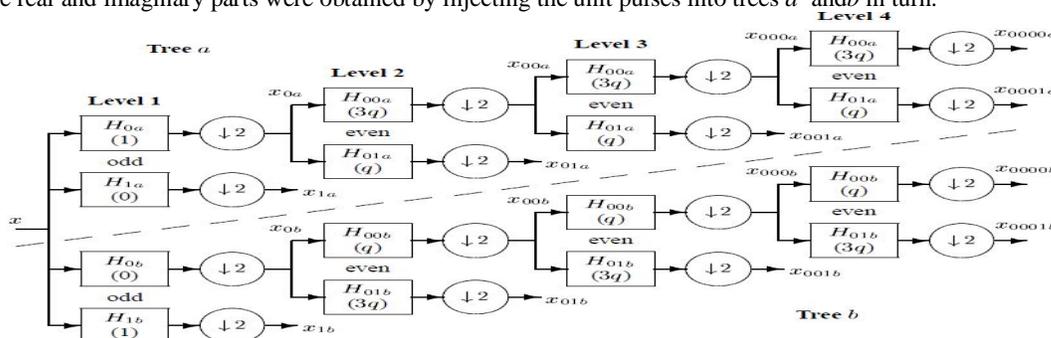


Fig 2. Structure for Dual Tree Complex Wavelet Transform

IV. DTCWT WITH ORTHOGONAL SHIFT PROPERTY

In order to examine the shift invariant properties of the dual tree in either the odd/even or Q-shift forms, consider what happens when we choose to retain the coefficients of just one type (wavelet or scaling function) from just one level of

the dual tree. For example, we might choose to retain only the level-3 wavelet coefficients x_{001a} and x_{001b} , and set all others to zero. If the signal y , reconstructed from just these coefficients [4], is free of aliasing then we define the transform to be shift invariant at that level. This is because absence of aliasing implies that a given sub band is a linear time-invariant (LTI) system and therefore is fully characterized by a single impulse response and a corresponding z -transfer function. In this context we define a sub band as comprising all coefficients from *both* trees at a given level and of a given type (either wavelet or scaling function)[9]. Figure.3 shows the simplified analysis and reconstruction parts of the dual tree when coefficients of just one type and level are retained. All down-sampling and up-sampling operations are moved to the outputs of the analysis filter banks and the inputs of the reconstruction filter banks, respectively, and the cascaded filter transfer functions are combined.

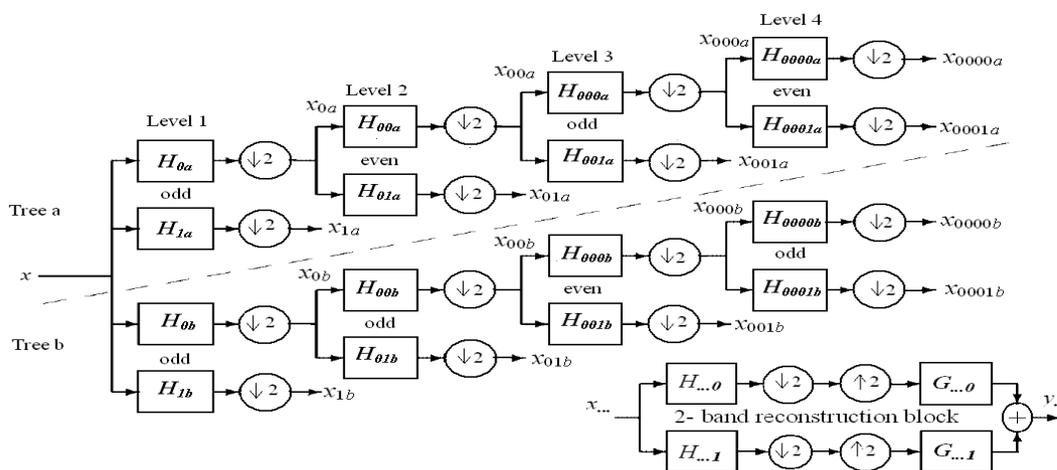


Fig 3 .Structure for Dual Tree Complex Wavelet Transform with OSP

V.SIMULATION RESULT OF DUAL TREE COMPLEX WAVELET TRANSFORM

The main process of wavelet transform is real and imaginary value into orthonormal series .orthonormal means continuous and discrete forum the dual tree complex wavelet transform define as dual real trees combined to perform functionThe experimental results of dual tree complex wavelet transform aregiventhe mathematical representation based simulation results are shown in forum of mat lab representation.The input image signal to noise ratio level is 20.7561 at the next stage of iteration 240 fig (b) Signal to noise ratio level is 26.3078.at the level of iteration 680 signal to noise ratio level is 27.4036 next level of iterations final stage of iteration 960 signal to noise ratio level is 28.0805 simulation results of DTCWT compare to existing algorithm its shown an better results and originality of the image can be improved.

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Simulate Blur and Noise



Processing...
At Iteration: 240



(a) PSNR=20.7561 (b) PSNR = 26.3078

Processing...
At Iteration: 660



Processing...
At Iteration: 960



(c) PSNR = 27.4036

(d) PSNR = 28.0805

V. SIMULATION RESULT OF DTCWT WITH OSP

The simulation results of dual tree complex wavelet transform with orthogonal shift properties is given in this algorithm based on wavelet transform in this techniques used to denoise an images compare to wavelet transform and dual tree complex wavelet transform in this algorithm is a best one for image denoising.

Simulate Blur and Noise



Processing...
At Iteration: 400



(a) PSNR = 20.7561

(b) PSNR = 28.3515

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(c) PSNR = 29.3447



(d) PSNR = 29.4845

The comparison table based on proposed algorithms dual tree complex wavelet transform and dual tree complex wavelet transform and orthogonal shift properties .compared to in this algorithms best one is dual tree complex wavelet transform with orthogonal shift properties.

Iterations	Dual Tree Complex Wavelet Transform(DTCWT)	Dual Tree Complex Wavelet Transform with OSP
	Input image SNR 20.7561	Input image SNR 20.7561
40	25.183349	26.44251
120	25.976274	27.27508
200	26.243046	27.55519
280	26.348001	27.66540
360	26.856187	28.19899
440	27.100302	28.45531
520	27.234780	28.59651
600	27.312901	28.67854
680	27.403688	28.77387
760	27.640600	29.02236
880	27.947419	29.34479
960	28.080527	29.48455

Table1. Comparison table for proposed algorithms

V.II DTCWT& DTCWT WITH OSP USING RGB IMAGES

In This dual tree complex wavelet transform is applicable for rgb images denoising process we experimentally shown in gray level images and rgb images mostly rgb images are corrupted by noise during image transmission and acquisition it is efficient method for recover an images in this algorithm used to recover an all type of images and also shown an better experimental results.

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Simulate Blur and Noise



a) PSNR=23.8711

Processing...
At Iteration: 240



b) PSNR=29.4378

Processing...
At Iteration: 680



c) PSNR= 30.9369



d) PSNR=32.7483

The comparison table based on proposed algorithms dual tree complex wavelet transform and dual tree complex wavelet transform and orthogonal shift properties using rgb images.

Iterations	Dual Tree Complex Wavelet Transform(DTCWT)	Dual Tree Complex Wavelet Transform with OSP
	Input image SNR 23.871135	Input image SNR 23.871135
40	27.933655	29.330338
120	28.936945	30.383792
200	29.327960	30.794358
280	29.510111	30.985616
360	30.123044	31.629196
440	30.450137	31.972644
520	30.640630	32.162172
600	30.735843	32.272635
680	30.936995	32.483845
760	31.144792	32.702031
880	31.247382	32.809751
960	31.188875	32.748321

VI. CONCLUSION

