

Digital Image Compression using Hybrid Transform with Kekre Transform and Other Orthogonal Transforms

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Abstract: This paper presents image compression technique using hybrid transform. Concept of hybrid wavelet transform can be extended to generate hybrid transform. In hybrid wavelet transform first few rows represent global features of an image and remaining rows represent local features of an image. In Hybrid wavelet matrix rows contributing to global characteristics can be varied. In the limiting case by taking kronecker product of orthogonal component transforms, hybrid transform is generated where all rows of transform matrix represent global features and no local features are present. This hybrid transform matrix is then applied on color image. High frequency contents of transformed image are eliminated and only low frequency contents are retained to get compressed image. RMSE is calculated at different compression ratios to check the performance of hybrid transforms. Various orthogonal transforms like DCT, Walsh, Slant, Hartley, Real-DFT and DST are combined with Kekre transform to generate hybrid transforms. DKT-DCT gives better image quality and lower RMSE than other pairs formed with DKT. Component size 32-8 i.e. 32x32 (Kekre Transform) and 8x8 (DCT) gives best results than other possible size combinations like 8-32, 16-16 and 64-4.

Keywords: Compression Ratio, Hybrid Transform, Image compression, Kekre Transform, Real-DFT

I. Introduction

In today's multimedia applications, digital images are used on large scale. Storage and transmission of such images needs more memory and bandwidth. Also time required to transfer such large amount of information is more. This infeasibility can be avoided by compressing the images. In compression, only visible information is extracted and redundant information is eliminated. [1]. It results in less storage space and less time for transmission. Image compression falls in two classes: lossy and lossless. In lossy image compression some loss of clearness of an image is allowed as it is not detected by human eyes. Discrete Cosine transform (DCT) [2] is a popular transform used in image compression. While using DCT image is divided into blocks and then DCT is applied on blocked image. It introduces blocking artifacts in the compressed image.

Wavelet transform is a mathematical tool that divides the data into different frequency components. High energy compaction property is the key characteristic of wavelets. Wavelets also help to analyze local properties of an image. This feature makes them highly applicable in image compression. Nowadays wavelets are becoming popular for other applications like biometrics applications [3,4], CBIR [5], steganography [6], analysis of DNA, ECG etc. Various wavelet based compression schemes are available and implemented in literature [7]. Different wavelet based image coding schemes include lifting based wavelet transform [8], set partitioning in hierarchical trees (SPIHT) [9,10], spatial orientation tree wavelet (STW), Embedded zero tree wavelet (EZW), Wavelet difference reduction (WDR) and adaptively scanned wavelet difference reduction (ASWDR) [1].

II. Related Work

Various methods have been proposed by different researchers in the literature. In recent years focus is on wavelet based compression methods and hybrid compression techniques. Compression using column and row transform is proposed in [11] by Kekre et al. Column and row wavelet based image compression is presented in [12] by Kekre et al. which uses wavelet generation method proposed in [13]. Use of column transforms or column wavelet transform instead of full transform or full wavelet transform proves to be useful in saving number of computations. Real Discrete Fourier Transform has been proposed in [14] by H.B. Kekre, Tanuja Sarode and Prachi Natu. It considers only real valued functions in Fourier transform and avoids complex functions. Hybrid compression technique using DCT and fractal image compression method has been proposed by Rawat and Meher in [15]. DCT is applied on 8x8 blocked images and then DCT coefficients of each block are quantized. Zigzag scanning is used to extract nonzero coefficients. Further Fractal image compression method is used and then the image will be encoded using Huffman coding. In [16] color image compression using DCT, VQ based coding and a new method that combines DCT and wavelet transform is used. Fractal image coding using optimization techniques like genetic algorithm, ant colony optimization and particle swarm optimization is

given in [17]. An attempt to reduce the computational cost is done by using these optimization techniques. In [18] effective run length coder has been implemented. Their algorithm works on quantized coefficients of DCT.DCT based fractal image compression and wavelet based fractal image compression is proposed in[19]. In[20] wavelet families like Daubechies, Biorthogonal, Coeflit and Symlet have been examined. Combination of wavelet image compression with SPIHT and fractal image compression has been presented in [21].

III. Proposed Method

In this paper hybrid transform is generated from two orthogonal transforms and applied on the image. In [22] generation of hybrid wavelet transform from two orthogonal transforms has been proposed.Kekre transform [23]is used as base transform to generate hybrid transform. Consider two orthogonal transforms A and B having size MxM and NxN respectively. To generate hybrid wavelet of A and B, first ‘m’ rows of resultant matrix is calculatedby repeating each column of ‘N’ times and multiplying it with each element of first row of B. These ‘M’ rows represent global characteristics in hybrid wavelet transform. Remaining rows are obtained by translating the rows of matrix B from second row onwards. These rows contribute local features of an image. Generated hybrid wavelet transform is as follows:

$b_{11} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \end{pmatrix}$...	$b_{1n} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \end{pmatrix}$	$b_{11} \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \end{pmatrix}$...	$b_{1n} \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \end{pmatrix}$	$b_{11} \begin{pmatrix} a_{1m} \\ a_{2m} \\ \vdots \end{pmatrix}$	$b_{1n} \begin{pmatrix} a_{1m} \\ a_{2m} \\ \vdots \end{pmatrix}$
b_{21}	...	b_{2n}	0	0	0	0	...	0
0	0	0	b_{21}	...	b_{2n}	0	...	0
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
0	0	0	0	0	0	\vdots	b_{21}	...	b_{2n}
b_{31}	...	b_{3n}	0	0	0	\vdots	0	...	0
0	0	0	b_{31}	...	b_{3n}	\vdots	0	...	0
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
0	0	0	0	0	0	\vdots	b_{31}	...	b_{3n}
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
b_{n1}	...	b_{nn}	0	0	0	\vdots	0	0	0
0	0	0	b_{n1}	...	b_{nn}	\vdots	0	0	0
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
0	0	0	0	0	0	\vdots	b_{n1}	...	b_{nn}

To increase contribution of global features, hybrid wavelet formation is done differently as in [24]. First N rows of T_{AB} are formed by repeating each column of ‘A’ ‘N’ times and multiplying each column by each element of first row of B. Next M rows are formed by repeating each column of ‘A’ N times and multiplying each column by each element in second row of B.Remaining (N-2)M rows are obtained by performing shift and rotate operation on third row onwards of Matrix ‘B’ by appending zeroes to it. As contribution of global characteristics increases, contribution of local features decreases.Above hybrid wavelet transform matrix changes to

$b_{11} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \end{pmatrix}$...	$b_{1n} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \end{pmatrix}$	$b_{11} \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \end{pmatrix}$...	$b_{1n} \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \end{pmatrix}$	$b_{11} \begin{pmatrix} a_{1m} \\ a_{2m} \\ \vdots \end{pmatrix}$	$b_{1n} \begin{pmatrix} a_{1m} \\ a_{2m} \\ \vdots \end{pmatrix}$
$b_{21} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \end{pmatrix}$...	$b_{2n} \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \end{pmatrix}$	$b_{21} \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \end{pmatrix}$...	$b_{2n} \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \end{pmatrix}$	$b_{21} \begin{pmatrix} a_{1m} \\ a_{2m} \\ \vdots \end{pmatrix}$	$b_{2n} \begin{pmatrix} a_{1m} \\ a_{2m} \\ \vdots \end{pmatrix}$

b_{31}	.	b_{3n}	0	0	0	...	0	0	0
0	0	0	b_{31}	.	b_{3n}	...	0	0	0
.
0	0	0	0	0	0	...	b_{31}	.	b_{3n}
.
b_{n1}	.	b_{nn}	0	0	0	...	0	0	0
0	0	0	b_{n1}	.	b_{nn}	...	0	0	0
.
0	0	0	0	0	0	0	b_{n1}	.	b_{nn}

Hybrid transform is the limiting case of hybrid wavelet transform generated above in which only global characteristics are focused and local features are not considered. It is generated as $A \otimes B$. Hybrid transform matrix is given below:

a_{11}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$...	a_{11}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$	a_{12}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$...	a_{12}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$...	a_{1m}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$...	a_{1m}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$
a_{21}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$...	a_{21}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$	a_{22}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$...	a_{22}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$...	a_{1m}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$...	a_{1m}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$
.
.
.
a_{m1}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$.	a_{m1}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$	a_{m2}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$.	a_{m2}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$.	a_{mm}	$\begin{pmatrix} b_{11} \\ b_{21} \\ \vdots \end{pmatrix}$.	a_{mm}	$\begin{pmatrix} b_{1n} \\ b_{2n} \\ \vdots \end{pmatrix}$

Above hybrid transform matrix is applied on R, G, and B plane of image separately. To compress the image low frequency contents of image are retained and high frequency contents are made zero. Compression ratio is varied from 2 to 32. In each case Root Mean Square Error between original image and compressed image is calculated.

IV. Experiments and Results

Proposed method is experimented on 20 different images using Matlab 7 and AMD dual core processor with 4 GB RAM.



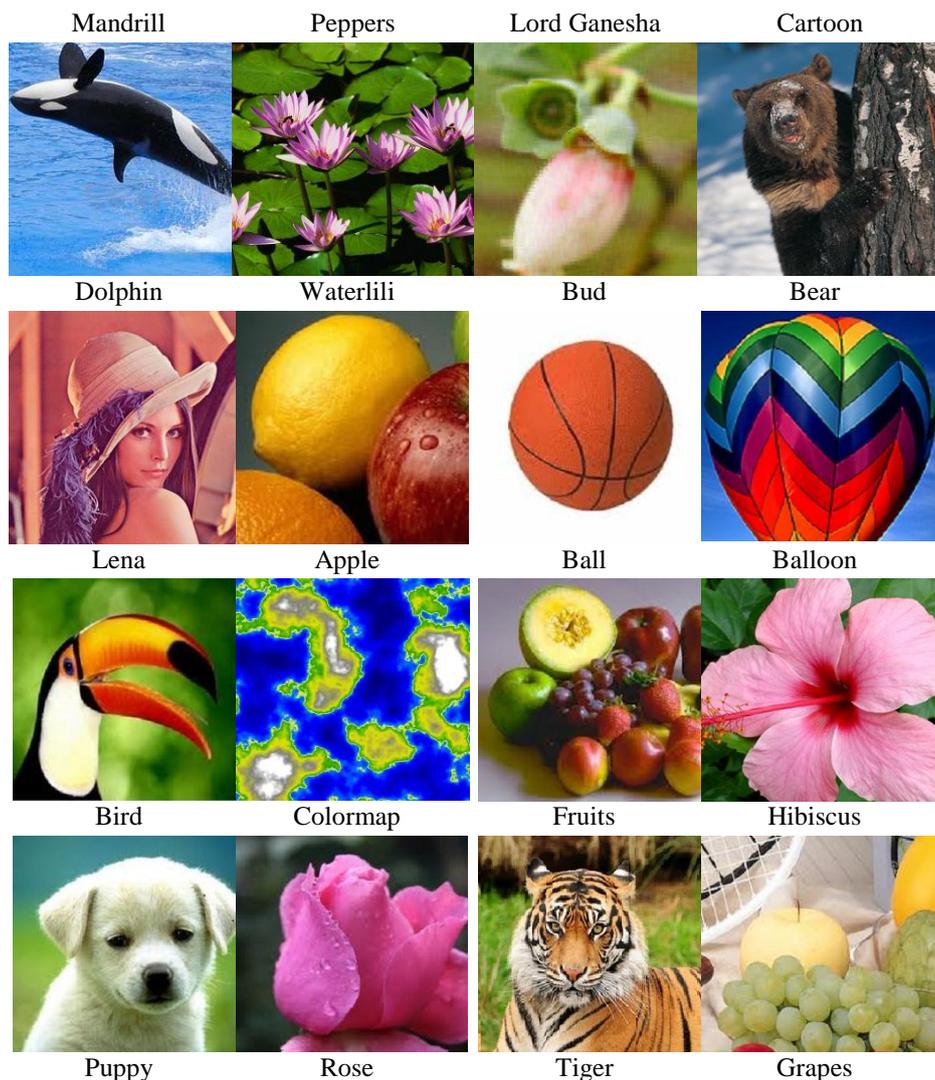


Fig.1 Set of twenty test images of different classes used for experimental purpose

Initially Walsh transform is used as matrix 'A' and Kekretransform is used as matrix 'B'. Thus Walsh-DKT hybrid transform is generated and applied on the images. Performance is measured in terms of RMSE at different compression ratios extending up to 32. Then hybrid transform is generated by reversing the matrix 'A' and 'B'. i.e. by using DKT-Walsh hybrid transform and again performance is measured. It is compared with the performance of Walsh-DKT hybrid transform. Similarly performance of DCT-DKT and DKT-DCT hybrid transform is compared. Resulting graphs related to these performances are shown below.

Fig. 2 compares performance of Walsh-DKT and DKT-Walsh hybrid transform. Various size combinations of component transforms are tried to select combination giving least RMSE. It has been observed that DKT-Walsh hybrid transform gives less error as compared to Walsh-DKT. Minimum RMSE at various compression ratios is plotted against compression ratio. Among four different size combinations 64-4size DKT-Walsh gives lower value of error upto compression ratio 10. For higher compression ratios 16 and 32, DKT-Walsh for size 32-8 gives lower RMSE. It is observed that as compression ratio increases RMSE increases. The lowest RMSE is 1.57 at compression ratio 2. It increases to 13.94 for compression ratio 32.

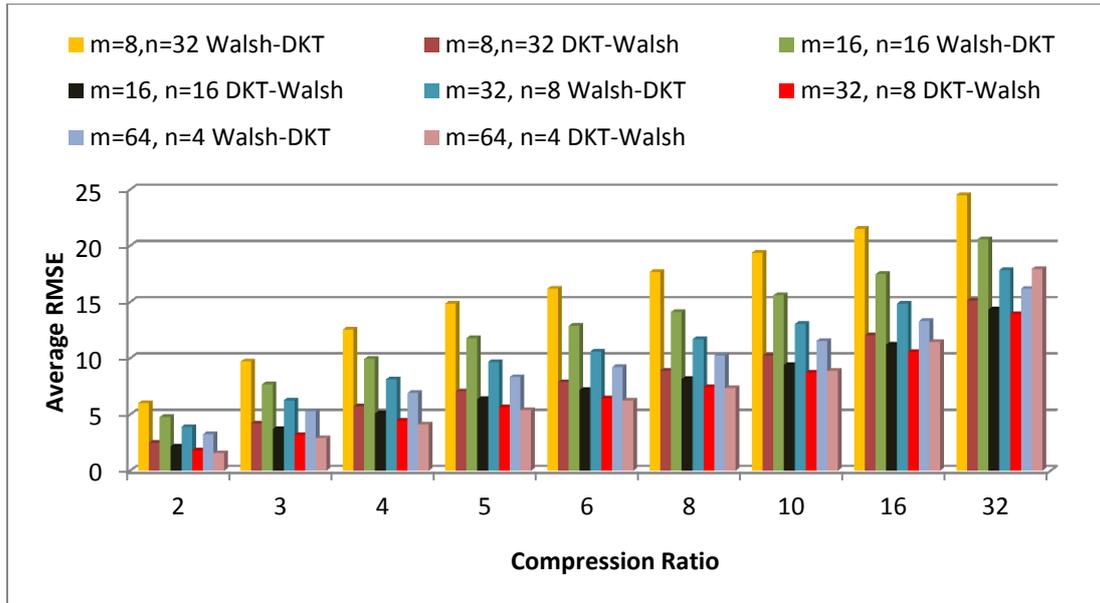


Fig.2 Comparison of Avg. RMSE vs Compression Ratio exchanging ‘A’ and ‘B’ matrices as Walsh-DKT and DKT-Walsh Hybrid Transform

Fig. 3 compares results obtained by DCT-DKT and DKT-DCT hybrid transform using various sizes of component transforms. DKT-DCT gives less error than DCT-DKT for all combinations. Up to compression ratio 6, component size 32-8 and 64-4 of DKT-DCT gives nearly same error. Thereafter from compression ratio 8 to 16 DKT-DCT using component size 32-8 gives less error than size 64-4. At compression ratio 32, component size 16-16 of DKT-DCT gives better results than others. The lowest RMSE is 1.34 at compression ratio 2. It increases to 11.90 for compression ratio 32. Thus giving less error as compared to DKT-Walsh for all combinations.

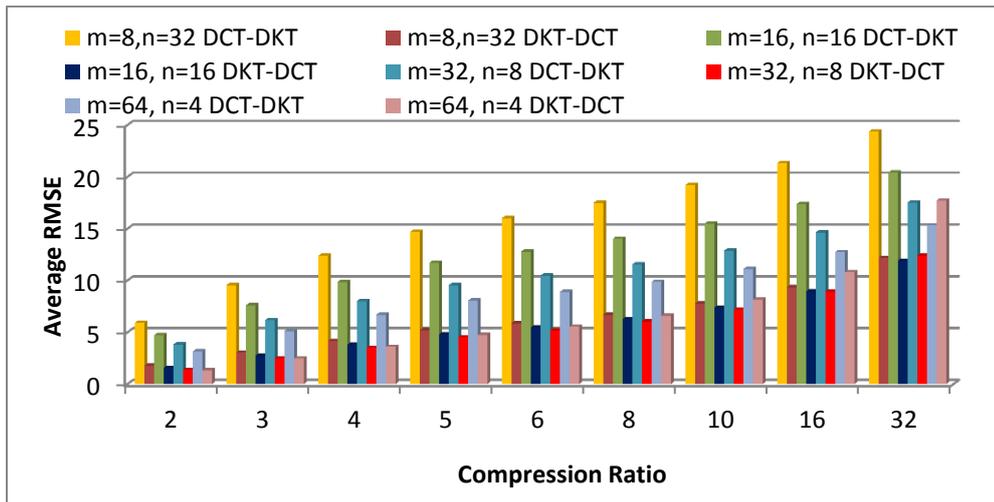


Fig.3 Comparison of Avg. RMSE against Compression Ratio by exchanging the base transform and local transform as Walsh-DCT and DCT-Walsh Hybrid Wavelet Transform

As observed in above two cases DKT-DCT gives better results than DKT-Walsh. Now with Kekre transform other orthogonal transforms like Real-DFT, Hartley, Discrete Sine Transform (DST) and Slant transforms are used. Their performances are compared for different component sizes and are plotted in graphs below.

Fig. 4 compares results obtained by different transforms combined with Kekre transform. Kekre transform of size 32x32 and second component transform of size 8x8 is selected. Among six different hybrid transforms DKT-DCT gives lower RMSE. Acceptable image quality is obtained at compression ratio 32. Performance of DKT-DCT is closely followed by DKT-RealDFT combination.

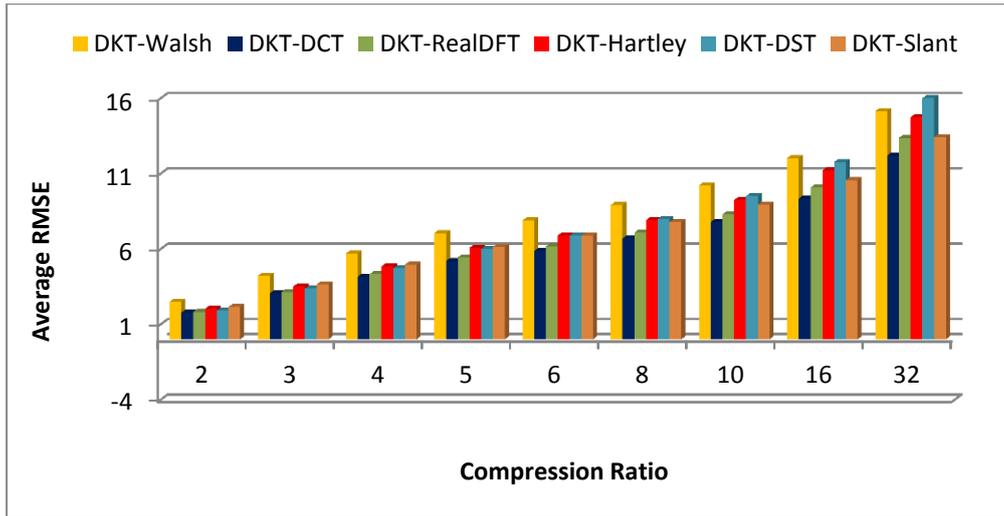


Fig.4 Average RMSE against Compression Ratio for Kekre Hybrid Wavelet Transform where 8x8 Kekre Transform acting as a base transform and different local component transforms of size 32x32

Fig. 5 compares performance of all hybrid transforms with both component transforms of size 16. At this component size also DKT-DCT shows superior performance amongst all. DKT-Real-DFT follows DKT-DCT upto compression ratio 8. For higher compression ratios from 10 to 32 DKT-Slant gives second best performance.

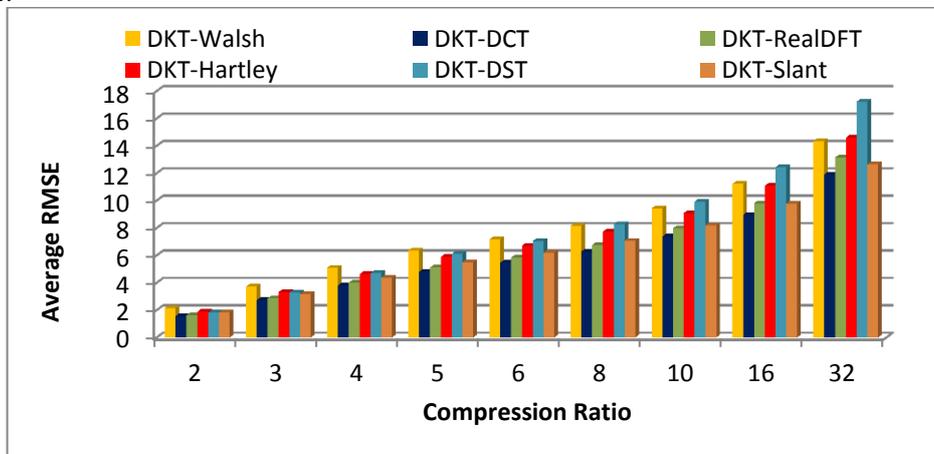


Fig.5 Average RMSE against Compression Ratio for Kekre Hybrid Wavelet Transform where 16x16 Kekre Transform acting as a base transform and different local component transforms of size 16x16

Further, as shown in Fig. 6, component size is changed to 32x32 for Kekre transform and 8x8 for second component. DKT-DCT gives lower error followed by DKT-Slant hybrid transform.

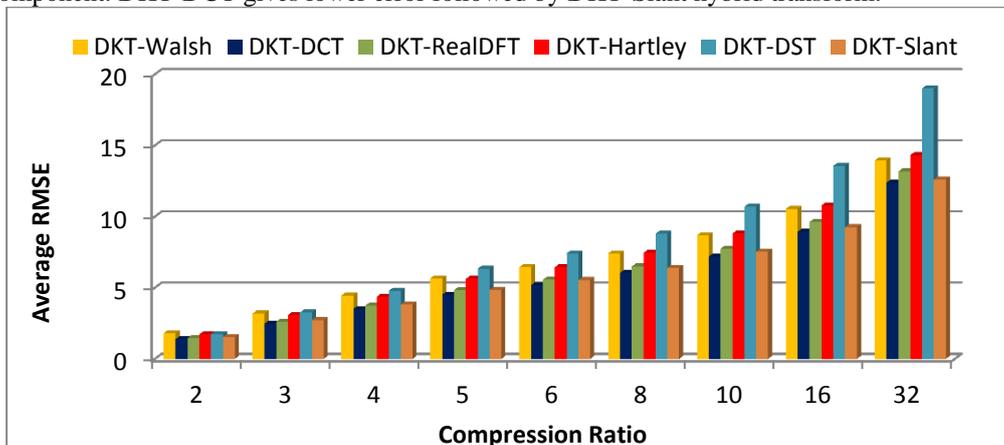


Fig.6 Average RMSE against Compression Ratio for Kekre Hybrid Wavelet Transform where 32x32 Kekre Transform acting as a base transform and different local component transforms of size 8x8

Next size variation done in component transforms is 64x64 and 4x4. Results are plotted in Fig. 7. Here RMSE value increases than previous cases. DKT-DCT, DKT-RealDFT and DKT-Slant show nearly equal value of RMSE.

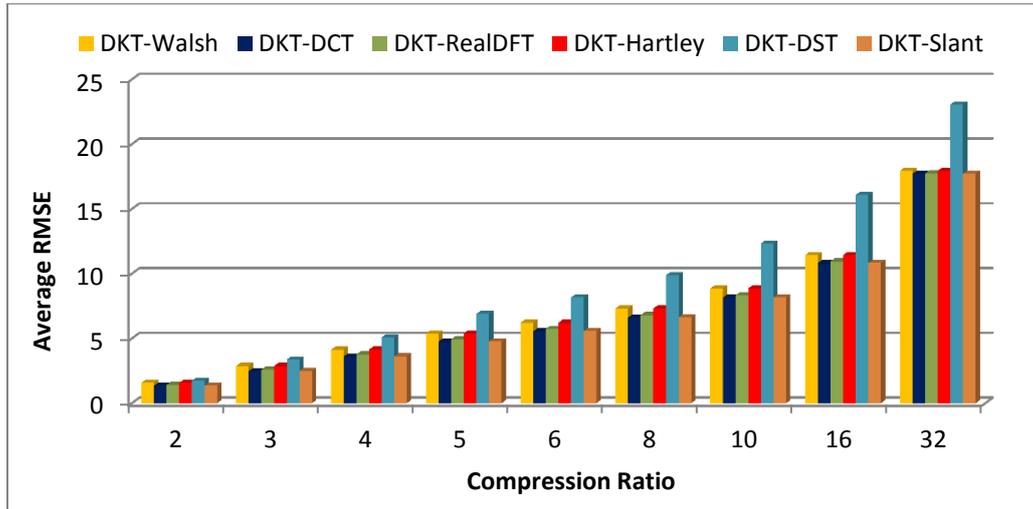


Fig.7 Average RMSE against Compression Ratio for Kekre Hybrid Wavelet Transform where 64x64 Kekre Transform acting as a base transform and different local component transforms of size 4x4

From above graphs it has been observed that DKT-DCT gives superior performance than other hybrid transforms irrespective of size of component transforms. Resulting compressed images using different hybrid transforms at different compression ratios are shown in Fig.8 with their respective RMSE values. In DKT-DST blocking effect is more prominent than other hybrid transforms and it is observed at compression ratio 32.

		Compression Ratio					
		Lemon	2	4	8	16	32
DKT-Walsh							
RMSE	Original Image	0.674	1.843	3.310	4.994	6.941	
DKT-DCT							
RMSE	Original Image	0.375	1.138	2.325	3.90	5.92	
DKT-RealDFT							
RMSE	Original Image	0.408	1.278	2.625	4.353	6.458	
DKT-DST							
RMSE	Original Image	0.619	2.141	4.664	8.174	12.606	

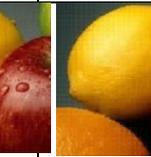
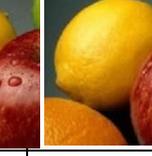
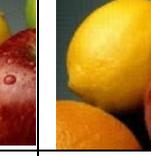
DKT-Hartley						
RMSE	Original Image	0.576	1.735	3.315	5.171	7.212
DKT-Slant						
RMSE	Original Image	0.464	1.355	2.6	4.13	6.049

Fig.8 Compressed images obtained using different hybrid transforms at various compression ratios with their respective RMSE values

Fig. 9 shows average RMSE at different bit rates. Less number of bits used to represent the image pixels indicates higher compression. All combinations give acceptable image quality of compressed images at lowest bit rate 0.25 bpp except DKT-DST.

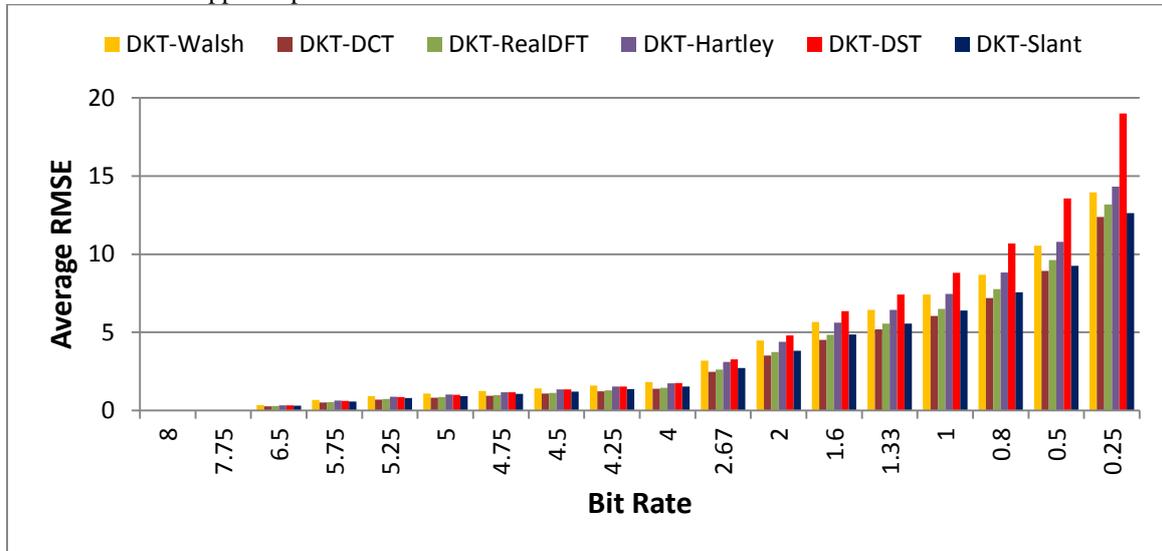


Fig.9 Performance of different hybrid transform in terms of RMSE at different bit rates

Comparison of RMSE values obtained by DKT-DCT having size MNxMN using various component sizes is tabulated in Table 1.

TABLE 1: RMSE values obtained by DKT-DCT hybrid transform at different compression ratios using variation in sizes of component orthogonal transforms

Compression Ratio	Avg. RMSE using DKT of size 'm', DCT of size 'n'			
	m=8, n=32	m=16, n=16	m=32, n=8	m=64, n=4
2	1.7736	1.5654	1.3850	1.3463
3	3.0379	2.7324	2.4764	2.4757
4	4.1601	3.7986	3.5075	3.5999
5	5.1960	4.8038	4.5132	4.7465
6	5.8628	5.4586	5.1836	5.5443
8	6.6895	6.2783	6.0375	6.6058
10	7.7736	7.3639	7.1918	8.1472
16	9.3435	8.9583	8.9328	10.8028
32	12.1830	11.9080	12.3893	17.7049

V. Conclusion

Proposed method of image compression uses hybrid transform which is generated using kronecker product of two orthogonal component transforms. Various pairs of component transforms have been tried and it has been observed that DKT-DCT gives superior results as compared to DKT-Walsh. Other transforms like Real-DFT, Hartley, Slant and Discrete Sine transform are used with Kekre transform. DKT-DCT pair gives minimum RMSE with acceptable image quality than other pairs formed with DKT. Component transforms of different sizes are used to generate hybrid transform of size 256x256. DKT-DCT hybrid transform with component size 32x32 and 8x8 respectively gives better performance. At compression ratio 16, RMSE of 8.93 is obtained. Even at compression ratio 32, acceptable image quality is obtained with RMSE 11.90.

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