Adaptive Unithreshold Two-Bit Block Motion Estimation

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Abstract—Image compression is essential technique to realize real time communication and motion estimation is the slowest part in the video processing. To reduce the complexity, one bit and two bit transform algorithms are presented past, but this algorithms are some problems, particularly when each pixel transform into binary, it increase encoding time. In this work, we improve the 2 bit transform algorithm and reduce the complexity while keeping subjective quality. When we do bit transforming, we compare the current block and previous frame's collocated block. If the block is almost same, we just calculate current block's average and variance. If it is not, we find the mean and variance separately. By doing this method, we can approximately gain 10 % complexity decrease.

Index Terms—video coding, motion estimation, fast motion, bit transform, block matching

I. INTRODUCTION

Digital video compression is the most important part to store the video data and transmit those in many broad casting services, and the video coding standard such as MPEG-x, H.264 has been standardized for reducing the data. Motion estimation (ME) is the most efficient part in the video coding and has a 50% complexity of entire process [1].

Full search (FS) is the fundamental algorithm and used widely in ME. It has the limited search area and finds the best candidate in that area. This algorithm exhausts all search points and has a mount of complexity; however, we can select best candidate, and it is easy to realize in hardware. To resolve these problems, a lot of algorithms have been proposed. ME is classified into 3 method largely: successive elimination algorithm (SEA), partial distortion search (PDS), bit transform (BT). First, SEA removes special search point if any specific condition is satisfied [2]. It has no loss of subjective quality compared with FS. Second, PDS is the early termination algorithm by taking some sample in the candidate block and predicting entire block distortion [3], [4]. Third, BT changes 8 bit image into binary image and revises the conventional block matching to process parallelly [5]-[9].

II. CONVENTIONAL ALGORITHMS

A. FS Algorithm

S

FS is the most famous technique in ME and use the block distortion of current block and candidate block. While searching the entire search point, it find the best search point having the minimum sum of difference (SAD) in the limited area. It exhausts all candidate points and takes a long time; however, we can get the optimal motion vector.

$$SAD(m,n) = \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \left| f^{t}(i,j) - f^{t-1}(i+m,j+n) \right|, \qquad (1)$$

where $-s \le m, n \le s$

where, f is the intensity of pixel, (i,j) means the pixel position. N-1 is the block size, and (m,n) is the translation of motion vector ,and s is the search area.

B. One-Bit Binary Block Matching ME

Binary block matching ME was developed for decrease the complexity by converting 8bit pixel into a few bits. It speeds up by applying data packing; however, it may have the wrong motion vector because its distortion range from 0 to 64 in case of 1BT 8x8 block matching (FS distortion range: 0~16320).

1BT use the next kernel which is the 17x17 band-pass filter.

$$K_{i,j} = \begin{cases} \frac{1}{25} , \text{ if } i, j \in [0, 4, 8, 12, 16] \\ 0 , \text{ otherwise} \end{cases}$$
(2)

where, i,j is the index of x, y axis. By (2), each image is filtered, and then we make a binary image using (3)

$$B(i, j) = \begin{cases} 1, \text{if } F(i, j) \ge \hat{F}(i, j) \\ 0, \text{otherwise.} \end{cases}$$
(3)

where F is an original image, and \hat{F} is a filtered image, and B means a binary image. Above process is called 1BT. After transforming, we can calculate the number of non-matching point (NNMP) which is the block matching criteria.

$$NNMP(m,n) = \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} B^{i}(i,j) \oplus B^{i-1}(i+m,j+n).$$
(4)
where $-s \le m, n \le s$

where, \oplus is the exclusive operator.

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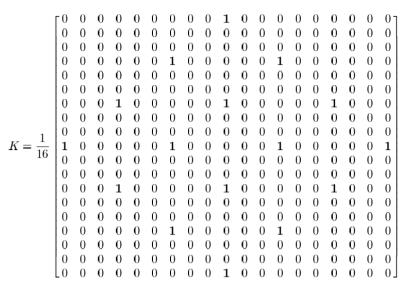


Figure 1. Multiplication-free kernel

C. Multiplication-Free 1-Bit Binary Block Matching ME

In One-bit algorithm, we cannot avoid the multiplication operation which increases the complexity of ME. To resolve this problem, we make the number of non-zeros be a power of 2 which can be realized in shift operation; however, the modified kernel generated in loss. The kernel shape is like a diamond and is shown in Fig. 1. The filter has an addition of 16 pixels is shifting the result, so it can be processed in integer operation.

D. Two-Bit Binary Block Matching ME

Seeing (4), we can notice that the distortion of block matching limited to 64, so it has possibility to find wrong motion vector. To improve this problem, 2BT was proposed. 2BT cover the 40x40 window block by block and calculate the mean and variance of the window using next equation.

$$\mu = E[I_{tw}]$$

$$\sigma^{2} = E[I_{tw}^{2}] - E^{2}[I_{tw}]$$
(5)

where, I_{tw} is the intensity of pixel in window, and E means the average. However, as the standard deviation has high computational burden, we simply approximate to this equation.

$$\sigma_a = 15 + 0.0125\sigma^2$$
. (6)

where, σ_a is the approximated standard deviation. Using these values, we change the pixel into binary.

$$B_{1}(i, j) = \begin{cases} 1, \text{if } I(i, j) \ge \mu \\ 0, \text{otherwise.} \end{cases}$$

$$B_{2}(i, j) = \begin{cases} 1, \text{if } I(i, j) \ge \mu + \sigma_{a} \text{ or } I(i, j) \le \mu - \sigma_{a} \\ 0, \text{otherwise.} \end{cases}$$

$$(7)$$

where $B_1(i, j)$ and $B_2(i, j)$ are each bit plane image.

Using (7), we can get the block matching distortion.

$$NNMP(m,n) = \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} \left\{ B_1'(i,j) \oplus B_1^{i-1}(i+m,j+n) \right\}$$

$$\| \left\{ B_2'(i,j) \oplus B_2^{i-1}(i+m,j+n) \right\}$$

where $-s \le m, n \le s$ (8)

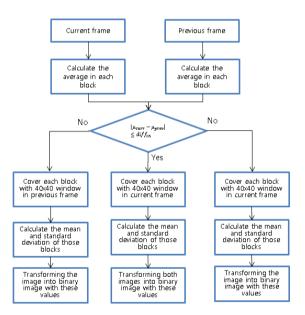


Figure 2. The flow chart of proposed algorithm.

TABLE I. EXPERIMENTAL ENVIRONMENT

Software	Off Codec		
Test Sequence	CIF (352x288)		
Block size	8, 16		
Search area	$\pm 8, \pm 16$		
Kernel size	40x40, 80x80		
Complexity	The number of operations		

PSNR	FS	2BT	Proposed	FS	2BT	Proposed
COASTGUARD	31.70	30.49	30.51	30.68	29.98	29.99
FOREMAN	33.07	30.08	30.35	32.33	30.22	30.26
MOBILE	25.87	24.96	24.98	25.18	24.83	24.85
STEFAN	25.48	24.32	24.35	24.92	24.35	24.35
TABLE	34.08	32.42	32.54	32.77	31.52	31.54
CONTAINER	38.43	37.03	37.72	38.34	37.97	37.98
HALL	36.01	33.96	34.07	34.83	33.78	33.88
AVERAGE	32.09	30.46	30.64	31.29	30.38	30.41

TABLE II. THE PSNR COMPARISON WITH CONVENTIONAL ALGORITHMS

TABLE III. THE COMPLEXITY COMPARISON WITH CONVENTIONAL ALGORITHMS

		1			1	
Complexity	FS	2BT	Proposed	FS	2BT	Proposed
COASTGUARD	77163	74778	60059	1135827	716394	655903.2
FOREMAN	77163	74778	61494	1135827	716394	660972.2
MOBILE	77163	74778	60043	1135827	716394	655479.2
STEFAN	77163	74778	63687	1135827	716394	664633.5
TABLE	77163	74778	60357	1135827	716394	657326.6
CONTAINER	77163	74778	59467	1135827	716394	655230.4
HALL	77163	74778	59565	1135827	716394	655430.5
AVERAGE	77163	74778	60667	1135827	716394	657854

where, || is the bit operator "OR". Equation (8) is revised from (4) and consider two bit plane and increase the distortion range, so we can get the better motion vector than 1BT.

E. Proposed Algorithm

2BT extends the distortion range to acquire more accurate motion vector than 1BT algorithms; however, it is not ignorable to process the mean and variance. Through this algorithm, we do not only increase the encoder speed, but also keep its subjective quality. The procedure of my algorithm is shown in Fig. 2. In advance, cover every block with 40x40 window each frame and simply calculate the average of each block each frame. If the difference between current and previous block's average is in the threshold value, we just calculate the average and variance of current blocks and then transform into binary image with those values. Otherwise, we calculate separately those values.

III. EXPERIMENTAL RESULTS

The Table I shows the experimental setting. Test sequences which have a different motion are used and the size of it is 352x288 (CIF). We experiment in the block size of 8x8, 16x16, and Depending on block size the search range is ± 8 , ± 16 . To find the best motion

vector, we used the spiral method which search from the point of (0,0) to neighboring point. For the comparison with original and reconstructed image, we adopted the PSNR measure which is used widely for the image processing. We also used the complexity method which is measured by following equation

$$Complexity = Add + Mul *8 + Bit _op + Com (9)$$

In (9), Add means add operation and the Mul is multiplication, and Bit_op a shift operation and Com a comparison. The Table II shows the PSNR results of proposed and conventional algorithms. Compared to FS, 2BT changes the 8bit pixels into 2bit pixel, so the PSNR of 2BT dropped to 1dB. Our proposed algorithm increased to 0.2dB in PSNR.

The Table III compared the complexity of algorithms and the complexity is measured as operations. FS has a large burden because it finds every search point and calculates the distortion of the block. The 2BT improved the complexity using bit operation and reducing the data of pixels.

IV. CONCLUSION

Our algorithm can calculate one block's variance and average depending on the similarity between current and previous blocks, and the complexity is reduced compared with 2BT. Its trend is same as 16x16 block case. The proposed algorithm calculates one block's average and variance in transforming binary values if the current and past frame's block is almost same. Our algorithms show not the improvement of PSNR but also reduce the complexity of algorithm, and it can be adopted for a variety of improved 2BT algorithms.

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