

A Study on the Vector Quantization of Medical Imaging System

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= Abstract =

Since PACS (Picture Archiving and Communications System) was installed in the hospital, storing and transmitting digital images have been possible. However, in the present condition, the size of digital medical images is so large that huge storage and fast network were necessary. To increase efficiency of storing and transmitting medical images, the present study suggests the image compression method using full search VQ (Vector Quantization) and TSVQ (Tree Search VQ).

가 200

1.

kbyte - 7 Mbyte

PACS(Picture Archiving and Communicatios System)

PACS
[1].

DICOM(Digital Imaging and
Communications in Medicine)

RLE(Run - Length Encoding)

가 . PACS

JPEG, JPEG2000

가

가

가

MR(Magnetic Resonance)/CT(Com - puted
Tomography)

VQ(Vector Quantization)

□

: , (120 - 752)

134

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2.

(pixel)
 가 ,
 가 [1].
 가
 가 . DCT
 (Discrete Cosine Transform) DWT(Discrete Wavelet
 Transform) (space domain)
 가 (coefficient)
 , 가
 (block) 가 가
 가 가
 VQ
 lossy (quantization)

$Q:R^k \rightarrow C$
 , $C = [y_1, y_2, \dots, y_N]$ (codebook)
 (code) , $i \in [1, 2, \dots, N]$
 N , $y_i \in R^k$
 가 N R^k N
 (cell) R_i . i
 $R_i = \{x \in R^k : Q(x) = y_i\}$
 ,
 $R_i \cap R_j = \emptyset$, $R_i \cup R_j = R^k$
 , R^k (partition) [3].
 X

[1].
 가 (scalar)
 .
 (algorithm)

. Shannon
 가 가
 .
 $d(X, Y_i) \leq d(X, Y_j), \text{ all } j$

2.1.

[3].
 , 가 .
 .
 (templet)
 .
 $N \times k$
 (Euclidean space) R^k
 $Q: R^k \rightarrow C$ (map -
 ping)

2.2. Full Search Vector Quantization

Full Search VQ(Vector Quantization)

가
 가
 . Full Search VQ
 (clustering) Lloyd (Forgy, Isodata, k -
 mean) [4, 5, 6].
 (minimum distortion)
 (nearest neighbor)

2.3. Tree Search Vector Quantization

Shannon VQ [2].
가 가 가

가 가 가
가 가 (balanced binary tree)

lattice - based codes, classified VQ, multistep VQ, product code(gain/sharp and mean removed), predictive VQ, finite - state VQ, tree structured VQ [4].
TSVQ(Tree Search Vector Quantization)

TSVQ P. C. Cosman, K. L가 [2].

2.4.

[2].
TSVQ search VQ full (root node)
가 가 가
(unbalanced binary tree)
가 가
(fixed rate code)

CT (modality) brain transverse
DICOM RAW
PiView(Infinitt, Seoul, Korea)
512 x 512 8bit
full search VQ
TSVQ 4,16,64,256
(average distortion),
(entropy), PSNR(Peak Signal to Noise Ratio)

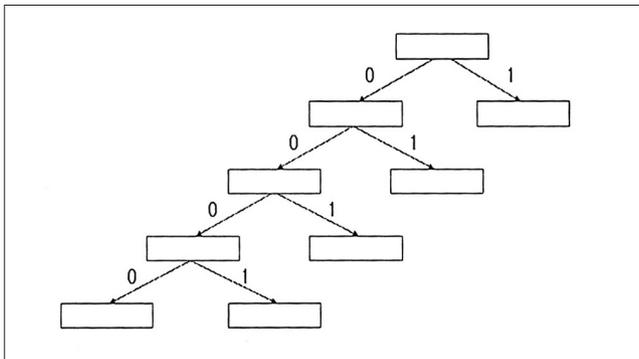


Fig. 1. This figure shows how to grow Unbalanced tree.

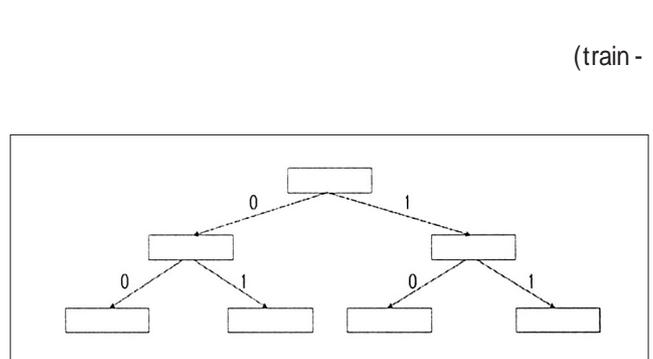


Fig. 2. This figure shows how to grow balanced tree.

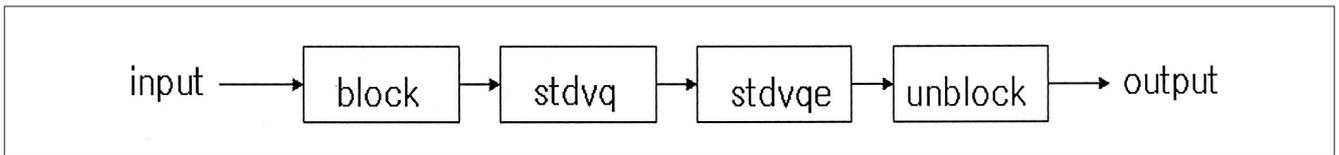


Fig. 3. This figure shows the diagram of full search VQ

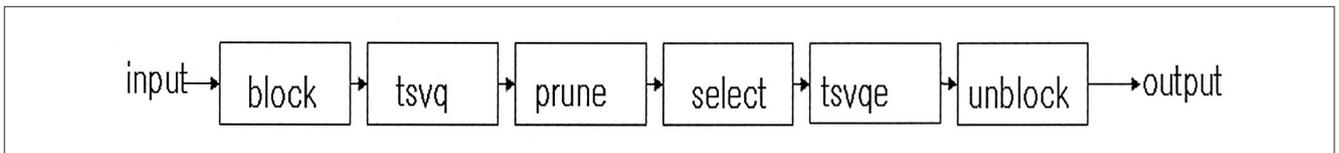


Fig. 4. This figure shows the diagram of TSVQ

ing) . Figure 3, 4 full PSNR
 search VQ TSVQ . (sequence)
 full search VQ
 3. 4
 Table 1 full search VQ Table 2 TSVQ . TSVQ full
 Figure 1 brain . search VQ 가
 16 PNSR [3].
 가 가
 64 PSNR
 (background)

Table 1. The result using full search vector quantization

dimension	num. of encoded vector	average rate	entropy	average distortion	PSNR
4	65536	1.386294	0.772711	45.971858	37.514513
16	16384	1.386294	0.794885	704.781493	31.687617
64	4095	1.386294	0.813372	7015.466799	27.730471
256	1024	1.386294	0.922285	42802.68913	25.897412

Table 2. The result using tree search vector quantization

dimension	num. of node	average rate	entropy	average distortion	PSNR
4	55509	9.862686	1.301292	0.045381	67.536313
16	52479	9.778137	1.225222	0.099946	70.133474
64	3571	6.910156	0.997414	94.055189	46.426487
256	867	5.358398	0.846192	5673.920211	34.664378

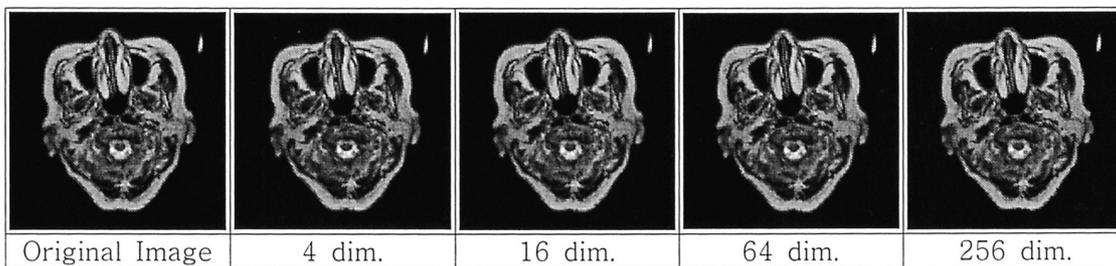


Fig. 5. This figure shows that full search vector quantized images over 16 dimension have distortion.

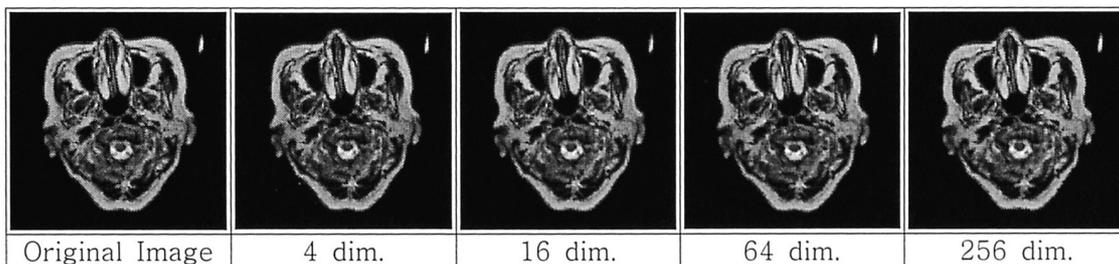


Fig. 6. This figure shows that tree search vector quantized images over 64 dimension have distortion.

Table 3. Compressed ratio about full search VQ when we use 256 KB 512 × 512 CT brain image

dim.	codebook size	training image num.	compress ratio
4	9 K	22	28.4
16	33 K	22	7.75
64	129 K	22	1.98
256	513 K	22	N/A

Table 4. Codebook size about TSVQ when we use 256KB 512 × 512 CT brain image

dim.	codebook size	training image num.	compress ratio
4	506 K	22	N/A
16	1158 K	22	N/A
64	1815 K	22	N/A
256	1832 K	22	N/A

256 TSVQ Figure 6
 Table 3 full search VQ

CT MR

Full search VQ PSNR
 4 30:1 JPEG
 10:1 [1].

Table 4 TSVQ
 가 가
 TSVQ 가
 가 가

Acknowledgments

(02 - PJ3 - PG6 - EV08 - 0001).

4.
 CT brain full search VQ,
 TSVQ
 JPEG
 가 가
 가 12 - bit
 가 0.25bpp
 가
 (block artifact)가
 DICOM WG JPEG2000 가
 JPEG2000 JPEG 16 - bit
 0.25bpp

1. , , , JPEG, JPEG2000
 PET brain , 가
 PACS 2003;9:1-7
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 29, April 1984

= =

PACS가 200kbyte - 7Mbyte 가 . 가

가 JPEG/JPEG2000 VQ(Vector Quantization)

full search VQ TSVQ(Tree Search Vector Quantization) .