Digital Image Processing

Image Compression

Importance of Image Compression

- Typical image resolution/depth
 - 1024 x 1024 x 24
 - Typical image size in bytes: 3 MB
- Typical video rate/resolution/depth
 - 30 fps, 720×576 , 24 bits
 - Typical video size for 1 minute: More than 2 GB

Data Redundancy

 Redundancy in information theory is the number of bits used to transmit a message minus the number of bits of actual information in the message. Informally, it is the amount of wasted "space" used to transmit certain data

Data Compression

 data compression or source coding is the process of encoding information using fewer bits (or other information-bearing units) than an un-encoded representation would use through use of specific encoding schemes.

Compression Methods

- Lossless Methods
- Lossy Methods

Run Length Coding

- In this method *runs* of data (that is, sequences in which the same data value occurs in many consecutive data elements) are stored as a single data value and count.
- Useful in relatively simple graphic images such as icons, line drawings, and animations.

Huffman Coding

- Huffman coding is an entropy encoding algorithm used for lossless data compression.
- The term refers to the use of a variable-length code table for encoding a source symbol

Example of Variable Length Coding

r _k	$p_r(r_k)$	Code 1	$l_1(r_k)$	Code 2	$l_2(r_k)$
$r_{87} = 87$	0.25	01010111	8	01	2
$r_{128} = 128$	0.47	1000000	8	1	1
$r_{186} = 186$	0.25	11000100	8	000	3
$r_{255} = 255$	0.03	11111111	8	001	3
r_k for $k \neq 87, 128, 186, 255$	0	—	8	—	0

Huffman Coding Example

a2	obability 0.4	1 0.4	2	3 0.4	4 → 0.6
	0.4	0.4	0.4	0.4	-06
a_6 a_1 a_4 a_3	0.3 0.1 0.1 0.06	0.3 0.1 0.1 0.1	0.3 → 0.2 0.1]	0.3 – → 0.3 –	0.4

Huffman Coding Example

Original source				Source reduction						
Symbol	Probability	Code	1	l	2	2		3	4	Ļ
$a_2 \\ a_6 \\ a_1 \\ a_4 \\ a_3 \\ a_5$	0.4 0.3 0.1 0.1 0.06 0.04	1 00 011 0100 01010 01011	0.4 0.3 0.1 0.1 	1 00 011 0100 ← 0101 ←		1 00 010 011	0.4 0.3 —0.3	1 00 - 01 -	0.6 0.4	0 1

Arithmetic Coding

- The output from an arithmetic coding process is a single number less than 1 and greater than or equal to 0
- This single number can be uniquely decoded to create the exact stream of symbols that went into its construction.

Example

• Encode the string "BILL GATES"

Character	Probability
SPACE	1/10
A	1/10
В	1/10
E	1/10
G	1/10
I	1/10
L	2/10
S	1/10
Т	1/10

Arithmetic Coding Example

Character	Probability	Range
SPACE	1/10	0.00 - 0.10
А	1/10	0.10 - 0.20
В	1/10	0.20 - 0.30
E	1/10	0.30 - 0.40
G	1/10	0.40 - 0.50
I	1/10	0.50 - 0.60
L	2/10	0.60 - 0.80
S	1/10	0.80 - 0.90
Т	1/10	0.90 - 1.00

Recovering Data

New Character	Low value	High Value
	0.0	1.0
В	0.2	0.3
I	0.25	0.26
L	0.256	0.258
L	0.2572	0.2576
SPACE	0.25720	0.25724
G	0.257216	0.257220
A	0.2572164	0.2572168
Т	0.25721676	0.2572168
E	0.257216772	0.257216776
S	0.2572167752	0.2572167756

Lossy Compressions

- JPEG
- JPEG2000
- MPEG (Video Compreession)
- H264 (Video Compression)

Steps followed in JPEG

- Color space transformation
- Downsampling
- Block splitting
- Discrete cosine transform
- Quantization
- Entropy coding

Color space transformation

- First, the image should be converted from RGB into YCbCr color space
- The compression is more efficient as the brightness information, which is more important to the eventual perceptual quality of the image, is confined to a single channel, more closely representing the human visual system.

Downsampling

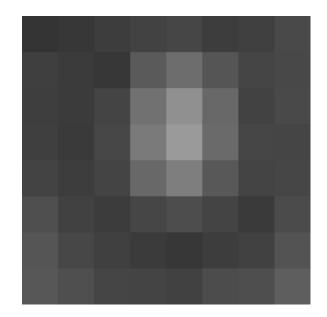
- Due to the densities of color- and brightnesssensitive receptors in the human eye, humans can see considerably more fine detail in the brightness of an image (the Y component) than in the color of an image (the Cb and Cr components).
- Downsample Cb and Cr components only 4:2:0

Block splitting

• After subsampling, each channel must be split into 8×8 blocks of pixels.

Discrete cosine transform

52	55	61	66	70	61	64	73
63	59	55	90	109	85	69	72
62	59	68	113	144	104	66	73
63	58	71	122	154	106	70	69
67	61	68	104	126	88	68	70
79	65	60	70	77	68	58	75
85	71	64	59	55	61	65	83
87	79	69	68	65	76	78	94

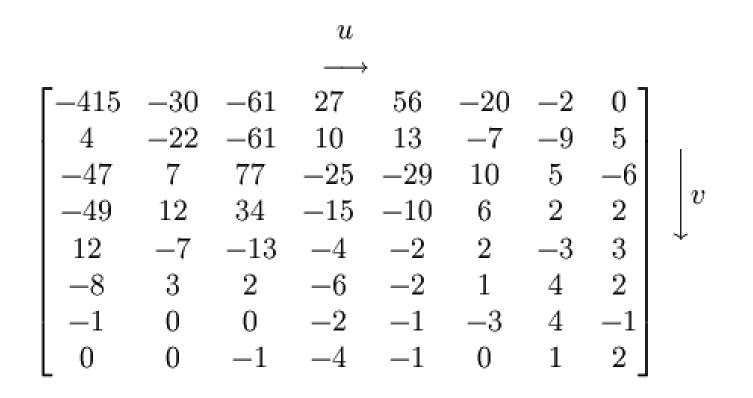


2D DCT

$$G_{u,v} = \alpha(u)\alpha(v)\sum_{x=0}^{7}\sum_{y=0}^{7}g_{x,y}\cos\left[\frac{\pi}{8}\left(x+\frac{1}{2}\right)u\right]\cos\left[\frac{\pi}{8}\left(y+\frac{1}{2}\right)v\right]$$

$$lpha_p(n) = egin{cases} \sqrt{rac{1}{8}}, & ext{if } n=0 \ \sqrt{rac{2}{8}}, & ext{otherwise} \end{cases}$$

DCT Transformed Image



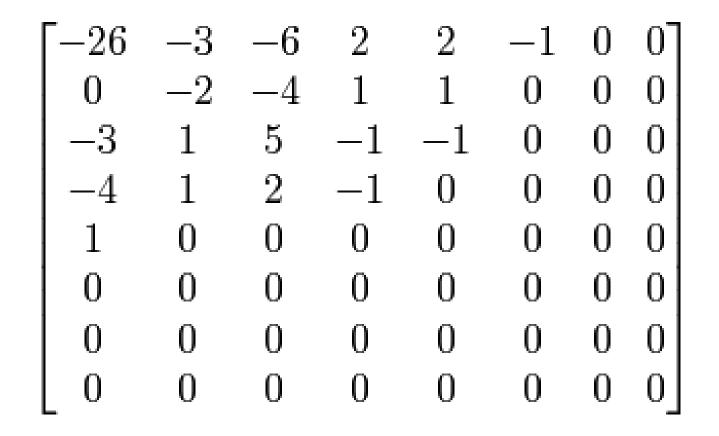
Quantization

 Quantization is done by simply dividing each component in the frequency domain by a constant for that component, and then rounding to the nearest integer

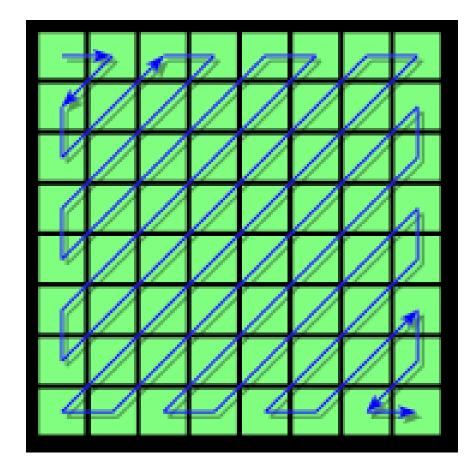
Sample Quantization Table

[16	11	10	16	24	40	51	61]
12	12	14	19	26	58	60	55
14	13	16	24	40	57	69	56
14	17	22	29	51	87	80	62
18	22	37	56	68	109	103	77
24	35	55	64	81	104	113	92
49	64	78	87	103	121	120	101
72	92	95	98	112	100	103	99]

Quantization Results



Entropy coding (Zig-Zag Scanning)



Zig-Zag Scanning

-26							
-3	0						
-3	-2	-6					
2	-4	1	-4				
1	1	5	1	2			
-1	1	-1	2	0	0		
0	0	0	-1	-1	0	0	
0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	
0	0	0	0	0	0		
0	0	0	0	0			
0	0	0	0				
0	0	0					
0	0						
0							

Variable length codes

- Variable length codes are generated for the code words of the zig-zag scan result.
- Trailing zeros are truncated by inserting an EOB symbol after the last non-zero value.

Typical compression ratios

- The compression ration depends on the quantization factors used.
- Typical ratio is 4:1