



IMAGE COMPRESSION

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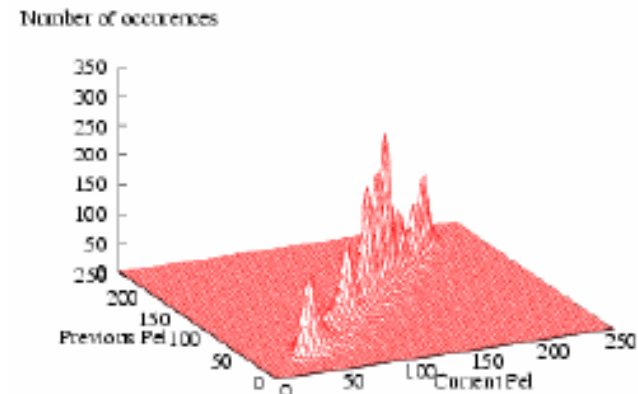
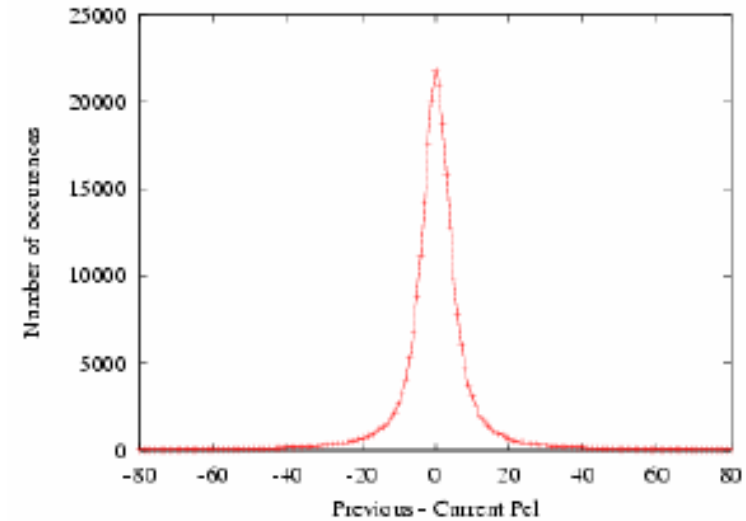
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Why image compression?

- Digital image acquiring devices develop fast → Image data become huge.
 - However: storage and transmission systems are still limited.
- We need methods and algorithms to reduce the size of sound and image to efficiently store and transmit over network.

Spatial dependency

- Neighboring pixels are highly correlated.
- The value of a pixel is probability dependent on those of its neighbors.



Predictive and lossless Encoding

- Entropy coding
 - Based on statistical properties of the information.
 - The original information and decoded information are exactly the same.
- Predictive coding
 - Based on the spatial and temporal dependency.
 - The decoded data may different from the original data.

Entropy

- Information measure
 - A symbol x with probability p contains information measured by:

$$I(x) = -\log(p(x))$$

- Information measure does not depend on the value of symbol.
- Information measure depends on the probability of symbol.
- When the base of log function is 2, the unit of information measure is bit.

Entropy

- Information Entropy
 - Entropy is defined as the average information measure of all symbols in a source. Entropy, H , is defined as follows:

$$H(X) = \sum_{x \in \mathcal{X}} I(x) = \sum_{x \in \mathcal{X}} -p(x) \cdot \log(p(x))$$

- It means that entropy of an information source is a function of its symbol probabilities. Entropy is maximum when all symbols appear with the same probability.

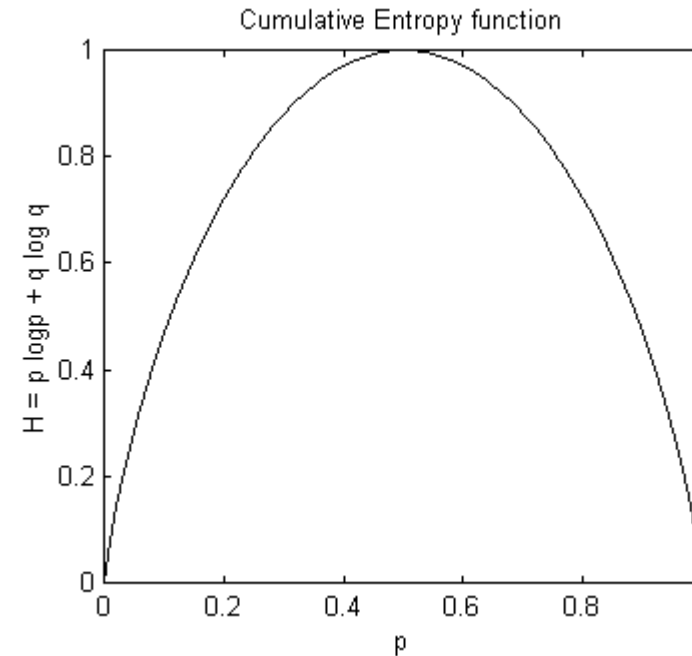
Entropy

- For example, an information source $\chi=\{0,1\}$, $p(0)=1/3$, $p(1)=1-p(0)=2/3$

$$\begin{aligned} H(X) &= -p(0)\log(p(0)) - p(1)\log(p(1)) \\ &= -\frac{1}{3}\log\left(\frac{1}{3}\right) - \frac{2}{3}\log\left(\frac{2}{3}\right) \\ &= 0.646 \end{aligned}$$

Entropy

- Entropy of source with 2 symbols
 - A convex function of p
 - When p comes to 0 or 1, source χ carries smaller information.
 - Maximum at $p=0.5$



Entropy

- Calculate $H(X)$:

$$X = \begin{cases} a & p(a) = 1/2 \\ b & p(b) = 1/4 \\ c & p(c) = 1/8 \\ d & p(d) = 1/8 \end{cases}$$

Entropy – Binary image

```

0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 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 0 0
0 0 0 0 0 0 0 0

```

$$P_0 = \frac{63}{64}$$

$$P_1 = \frac{1}{64}$$

$$H = -\frac{63}{64} \log_2 \frac{63}{64} - \frac{1}{64} \log_2 \frac{1}{64}$$

$$= 0.116 \text{ bits/pixel}$$

```

0 1 1 0 1 0 1 0
1 0 1 0 1 0 0 1
1 1 0 1 0 1 1 0
0 1 0 0 1 1 0 0
1 0 0 0 1 0 1 1
0 0 1 0 1 1 1 1
0 1 0 1 1 1 0 1
0 1 0 0 0 1 0 0

```

$$P_0 = \frac{32}{64} = \frac{1}{2}$$

$$P_1 = \frac{32}{64} = \frac{1}{2}$$

$$H = -\frac{1}{2} \log_2 \frac{1}{2} - \frac{1}{2} \log_2 \frac{1}{2}$$

$$= 1.0 \text{ bits/pixel}$$

Data compression

- Symbols with high probability are assigned with short codes, whereas symbols with low probability are assigned with longer codes.
- The length of coded data is shortened because the average length is reduced.

Data compression theory

- Code C for a source X is a mapping from χ (space of X), to D^* where D^* is a set of code sequence. For example:
 - $C(\text{Red})=00$, $C(\text{Blue})=11$ is code of $\chi=\{\text{Red}, \text{Blue}\}$ and $D=\{0,1\}$
- The average length of C :

$$L(C) = \sum_{x \in \chi} p(x)l(x)$$

Data compression theory

- Code C is called prefix code if there is no code is prefix of any other code.
 - 00 is prefix of 001
 - Code C with $C(1)=0$, $C(2)=10$, $C(3)=110$, $C(4)=111$ is a prefix code.
 - With prefix code, decode process can be done on-the-fly.

Entropy coding

- Example – fixed length code

Symbol x	Probability	Code $C(x)$	Code length $L(c)$
A	0.75	00	2
B	0.125	01	2
C	0.0625	10	2
D	0.0625	11	2

Average bits = $0.75 \cdot 2 + 0.125 \cdot 2 + 0.0625 \cdot 2 + 0.0625 \cdot 2 = 2.0$ bits/symbol

Entropy coding

- For example – variable length code

Symbol x	Probability	Code $C(x)$	Code length $L(c)$
A	0.75	0	1
B	0.125	10	2
C	0.0625	110	3
D	0.0625	111	3

Average bits = $0.75*1 + 0.125*2 + 0.0625*3 + 0.0625*3 = 1.375$ bits/symbol

Entropy coding

- Exp-Golomb codes
 - Used in coding video (motion vectors)
- Huffman codes
 - Data compression, still image compression.
- Arithmetic codes
 - Entropy coding for H.264/AVC
- Run-length codes

Exp-Golomb code

- Easy and simple to implement on hardware devices.
- Suitable for source having laplace or exponential distribution.
- Used for coding the length of motion vectors in video coding.

Exp-Golomb code

- Structure of Exp-Golomb code is as following:
[M zeros][1][INFO]
 - INFO is M-bit field number
 - $M = \text{floor}(\log_2[\text{code_num}+1])$
 - $\text{INFO} = \text{code_num}+1 - 2^M$
- Decode:
 - Read M characters '0', follow by a '1'
 - Read M bits of INFO
 - $\text{Code_num} = 2^M + \text{INFO} - 1$

Exp-Golomb codes

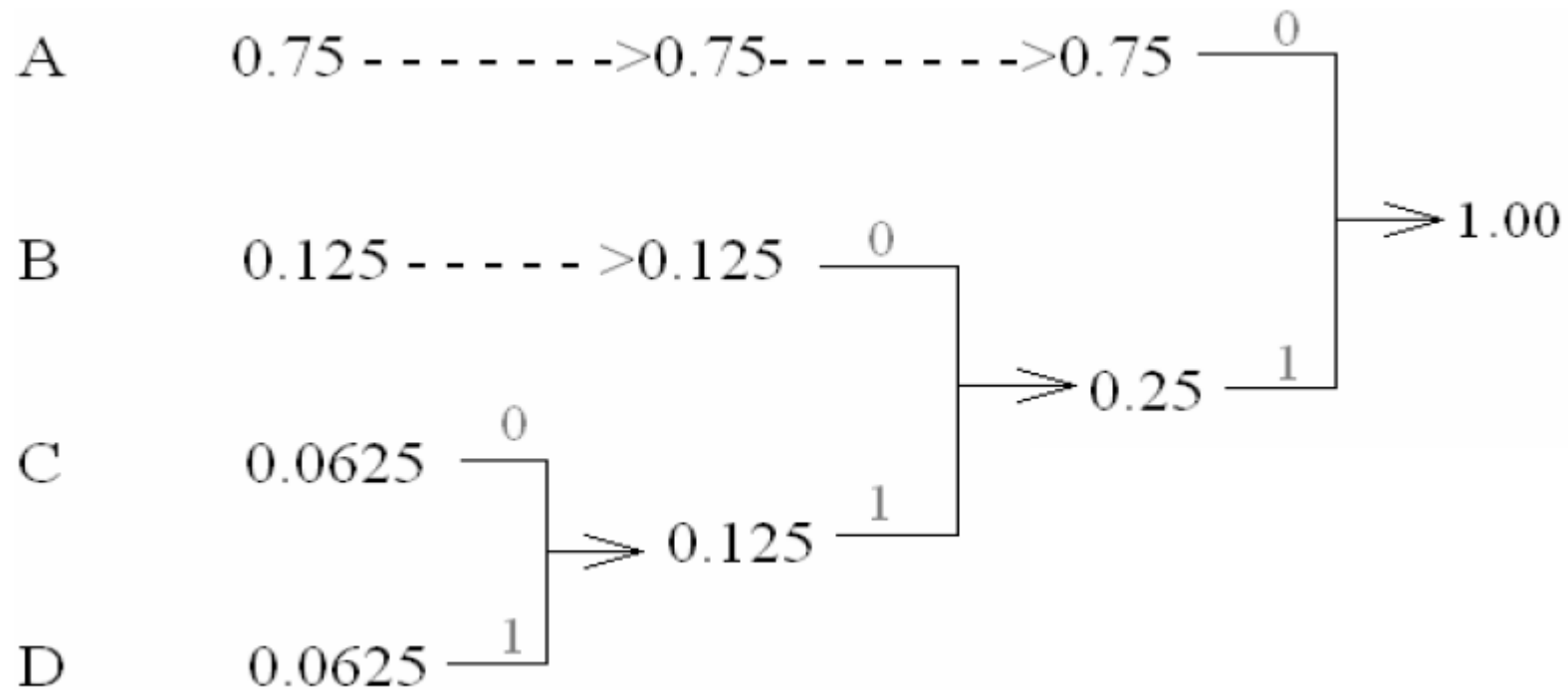
Code	Unsigned number	Signed number
1	0	0
010	1	1
011	2	-1
00100	3	2
00101	4	-2
00110	5	3
00111	6	-3
0001000	7	4
0001001	8	-4
0001010	9	5
0001011	10	-5
0001100	11	6
...

Huffman codes

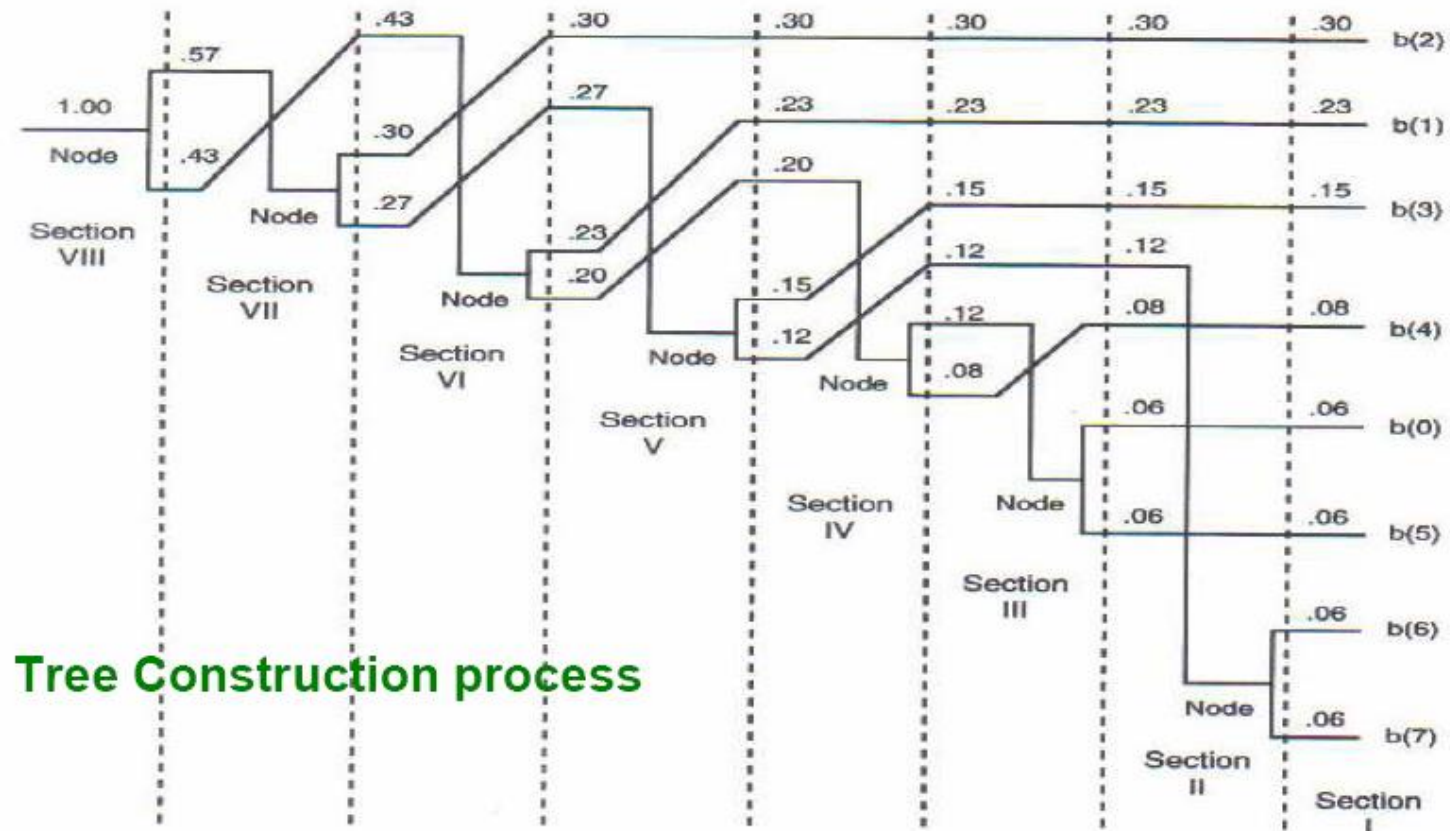
- Were discovered by Huffman in 1952 at MIT.
- These are prefix codes assigned to symbols depending on the probability of the symbols.

Huffman codes

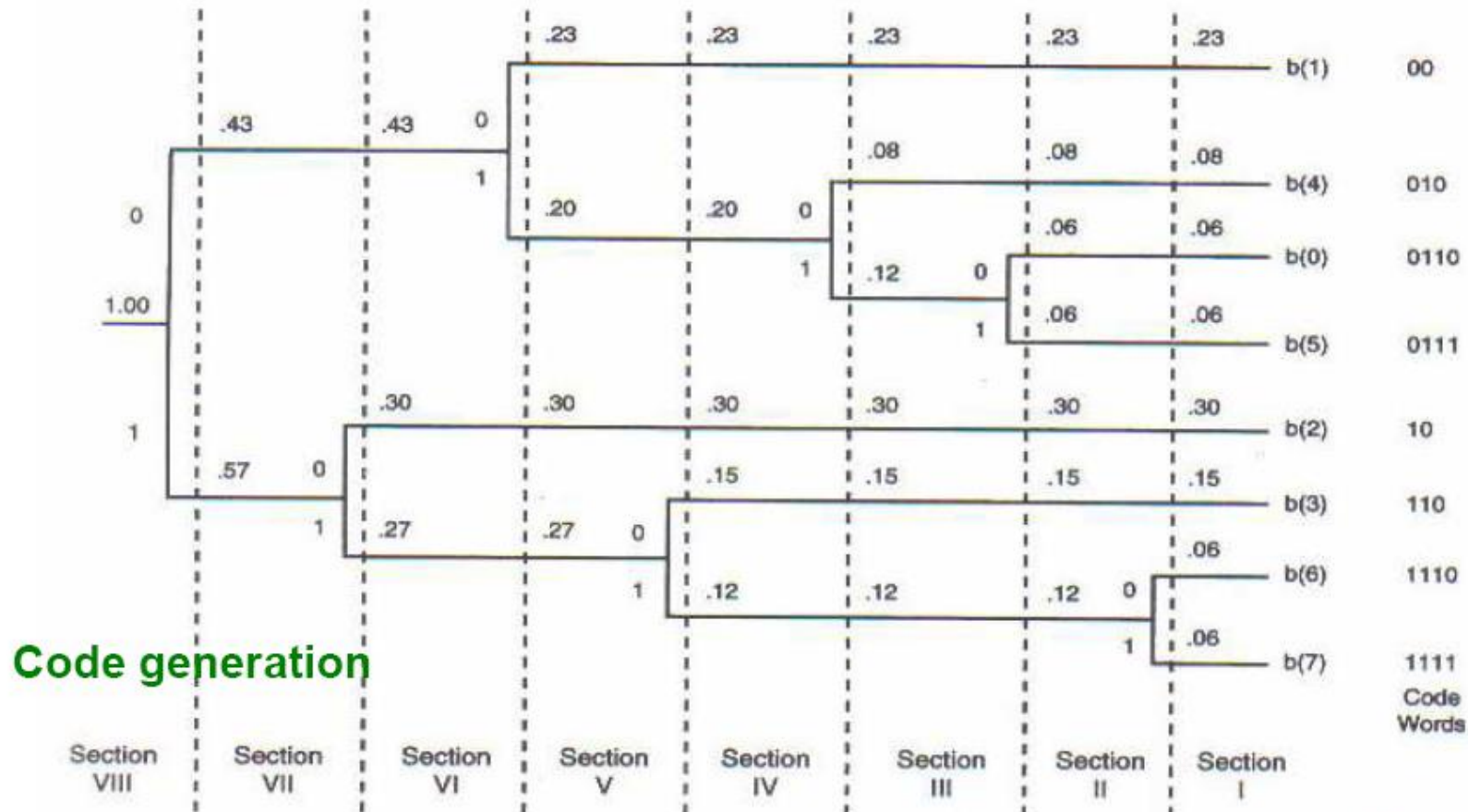
- Generating Huffman codes from a source



Huffman codes



Huffman codes



Huffman codes

Disadvantages of Huffman codes:

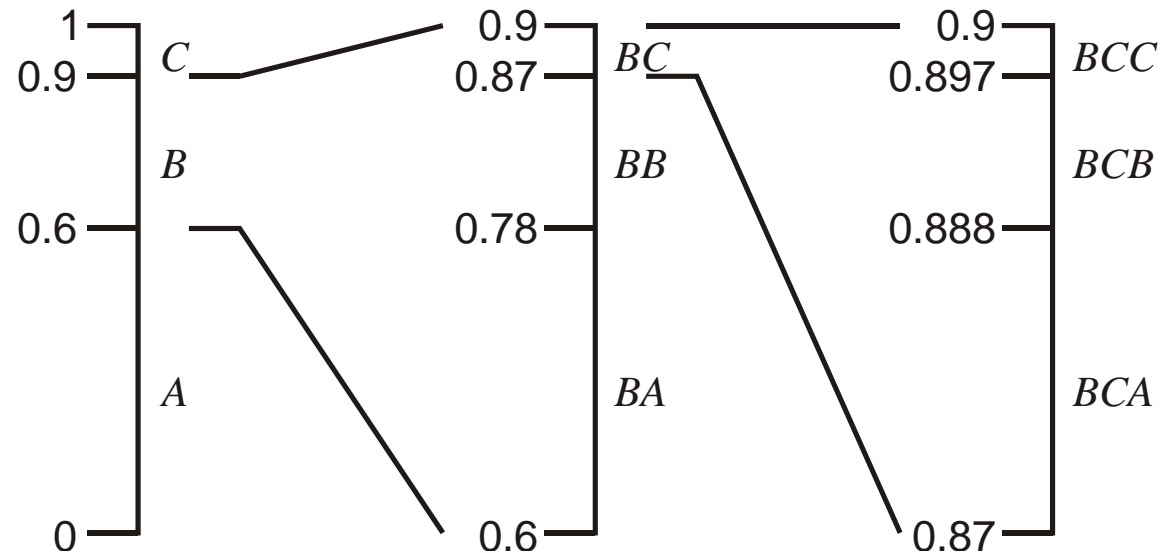
- Huffman codes must have integer number of bits.
- Eg: an information source has 2 symbols a and b with probability $p(a) = 0.9999$, $p(b) = 0.0001$. Huffman assigns each symbol with a code word length 1. Therefore the average code length is 1 whereas the entropy of the source is 0.00147 bits/symbol.
- Probability of symbol must be known before encoding and encoding, the symbol-code mapping must be agreed between encoder and decoder.
- When the number of symbol is large, the mapping table becomes very large.

Arithmetic codes

- Arithmetic Coding
 - Code words can be non-integer length, and the probability of symbols can be calculated during encoding and decoding process.
 - Instead of building the mapping between symbol and code word, arithmetic coding finds the mapping from the sequence of symbol to a real number in $[0,1]$.

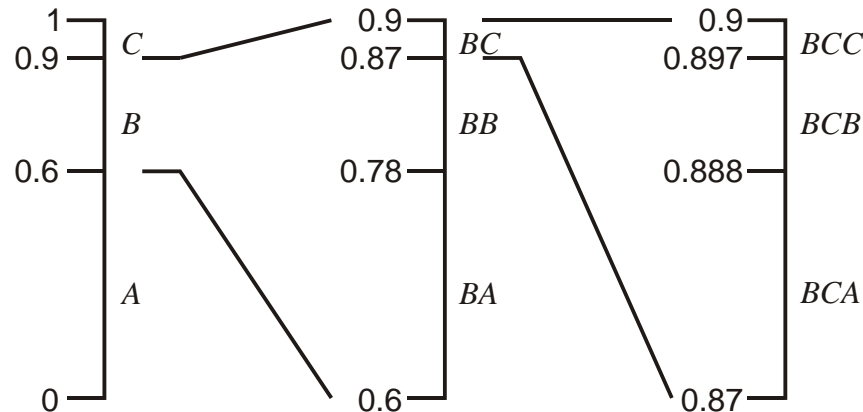
Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- The sequence starts with A, B, and C is mapped to a half-open interval $[0, 0.6)$, $[0.6, 0.9)$, and $[0.9, 1)$ depending on its probability. In this example interval $[0.6, 0.9)$ is first used for the symbol B.



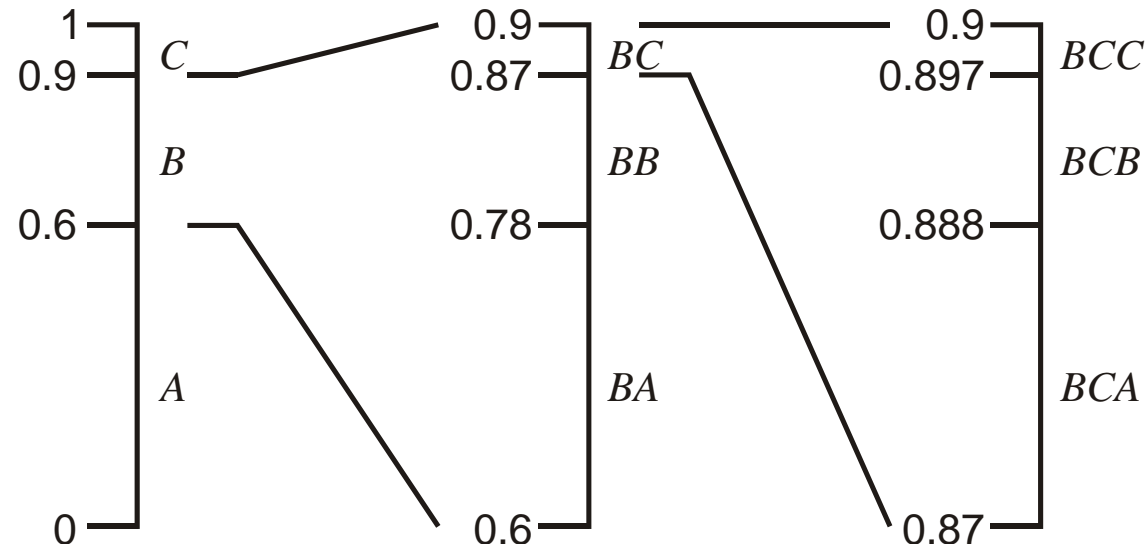
Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- This interval is further divided to smaller interval $[0.6, 0.78)$, $[0.78, 0.87)$, $[0.87, 0.9)$, for sequences start with BA , BB , BC . Note that: the ratio among the sub-interval is equal to the ratio among probabilities of symbols:
 - $0.18 : 0.09 : 0.03 = 6 : 3 : 1 = p(A) : p(B) : p(C)$



Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- Repeat this step, interval $[0.87, 0.9)$ is divided to 3 smaller intervals. Sequences start with BCA are mapped to $[0.87, 0.888)$.

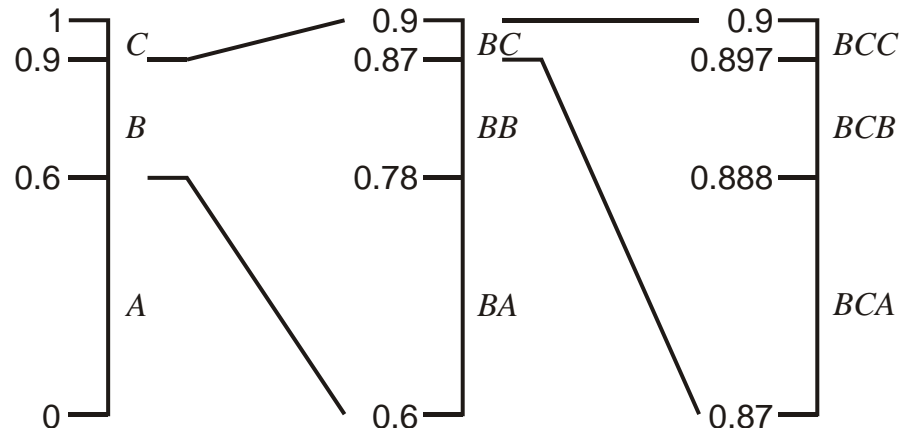


Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- 2 ends of interval $[0.87, 0.888)$ is presented by a binary number:

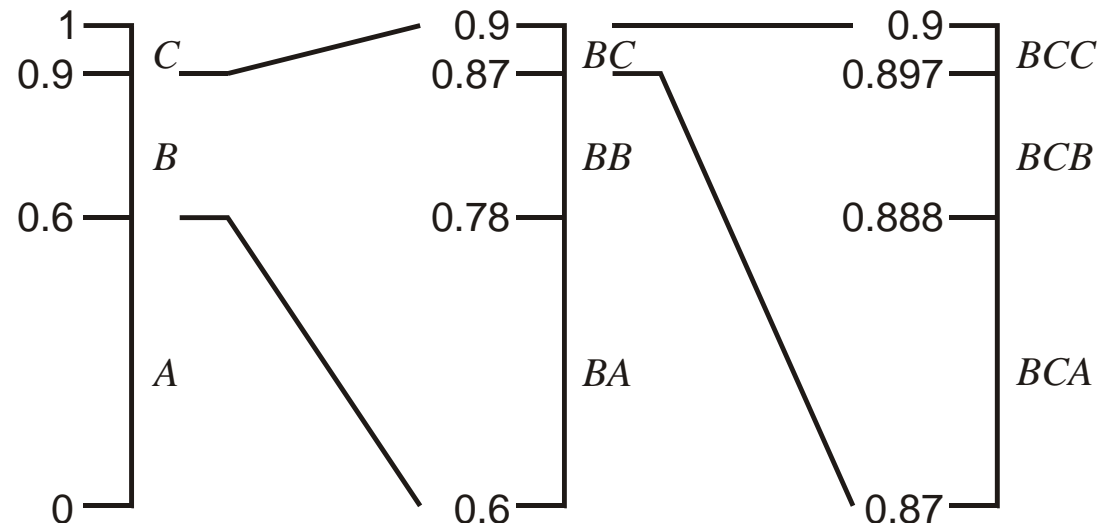
$$0.87 = \frac{1}{2} + \frac{1}{2^2} + \frac{1}{2^4} + \frac{1}{2^5} + \dots = 0.11011\dots,$$

$$0.888 = \frac{1}{2} + \frac{1}{2^2} + \frac{1}{2^3} + \frac{1}{2^7} + \dots = 0.11100\dots.$$



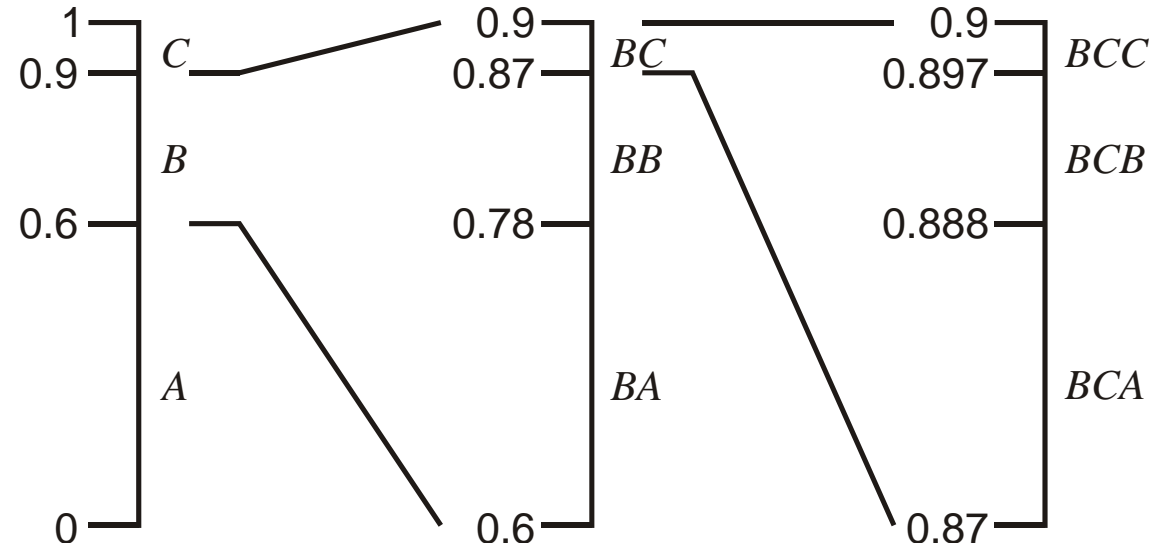
Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- Encoder can send any number in the interval $[0.87, 0.9)$ to indicate the encoded sequence starts with BCA. For example, it can send 3 bits 111 corresponding to 0.875 in decimal.



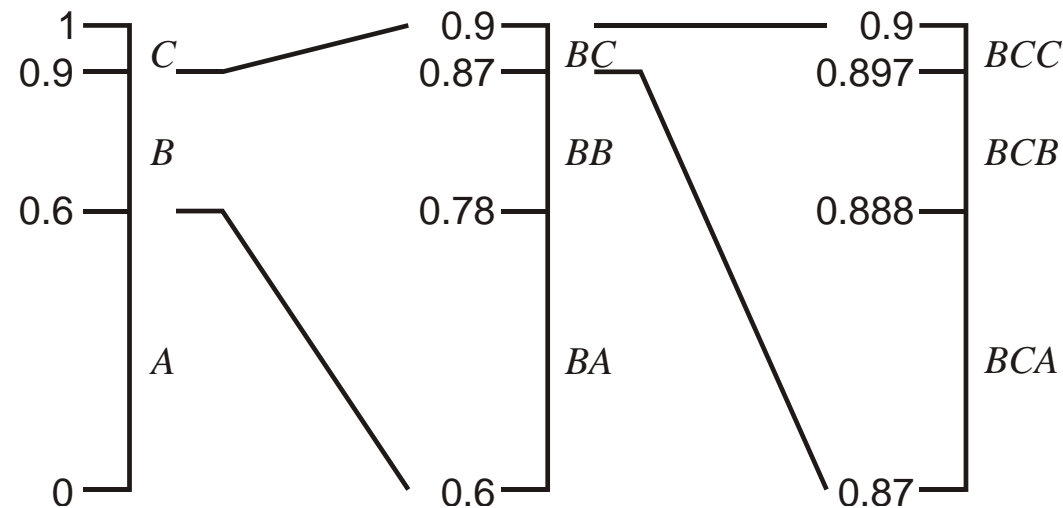
Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- After receiving 111, decoder processes the interval dividing procedure and know that the decoded sequence starts by BCA. Because it knows that 0.875 is in the interval $[0.87, 0.888)$



Arithmetic codes – Example 1

- Given a source $\{A, B, C\}$, and a sequence BCA to be encoded.
- $p(A) = 0.6$, $p(B) = 0.3$, $p(C) = 0.1$
- However, 0.875 can be the presentation of B, BA, or BAC. So it needs an additional symbol to indicate the end of sequence (EOS).



Arithmetic codes – Example 2

Example 4.3.1:

Consider a three-letter alphabet $\mathcal{A} = \{a_1, a_2, a_3\}$ with $P(a_1) = 0.7$, $P(a_2) = 0.1$, and $P(a_3) = 0.2$. Using the mapping of Equation (4.1), $F_X(1) = 0.7$, $F_X(2) = 0.8$, and $F_X(3) = 1$. This partitions the unit interval as shown in Figure 4.1.

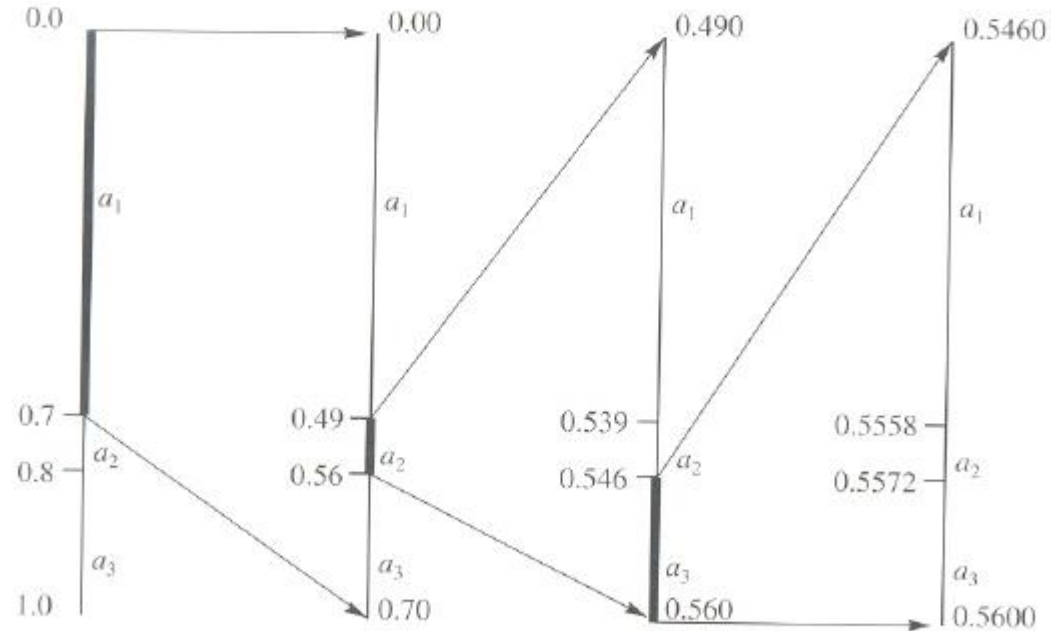


FIGURE 4.1 Restricting the interval containing the tag for the input sequence $\{a_1, a_2, a_3, \dots\}$.

Arithmetic codes

- Binary presentation:

i	2^{-i}	X	Bit
1	0.5	0.3203125	0
2	0.25	0.3203125	1
3	0.125	0.0703125	0
4	0.0625	0.0703125	1
5	0.03125	0.0078125	0
6	0.015625	0.0078125	0
7	0.0078125	0.0078125	1
8	0.00390625	0	0

Arithmetic codes

- Concerned issues for developing arithmetic codes for industrial applications:
 - Using integer numbers instead of floating point numbers.
 - Subinterval shinks rapidly during encoding process → rescale mechanism.
 - Reference to CABAC entropy coding of H.264/AVC.

Arithmetic codes

- Encode the following sequence using arithmetic codes:
badcd
- Distribution of symbols:

$$X = \begin{cases} a & p(a) = 1/2 \\ b & p(b) = 1/4 \\ c & p(c) = 1/8 \\ d & p(d) = 1/8 \end{cases}$$

Run-length coding

- Similar consecutive symbols are grouped and presented by a code
- Start from a specific code, following by a symbol and the number indicating the number of its copies.
- It is a entropy coding techning used in H.264, CAVLC

Run-length coding

- The simplest case of Run-length coding is to group the similar consecutive symbol and replace by a pair [N][S]
 - N – number of symbols
 - S – symbol
- For example: aaaabbbbddaaddccccccca
- Is is compressed to: 4a3b2d2a2d7c1a

Run-length coding

- It is applied to sub-sequence with the number of similar consecutive symbols more than 3.
- Replace the sub-sequence by [E][S][N-1]
 - E: A specific symbol
 - S: Symbol
 - N: Number of symbols

Predictive coding

- The temporal and spatial dependencies are very high.
- Data are not directly encoded, they are predicted from the their neighbors.
- The difference from data and their predicted values are quantized and encoded by entropy code.

Predictive coding

- Combine with transform to form hybrid-coding methods.
- It is fundamental in multimedia data compression.
- Quantization:
 - Build the relationship between the data lost and coding efficiency.
 - Plays vital role in multimedia communication in public channels.

Predictive coding

- For example: previous pixels are used to predict the following pixels:

90	96(+6)	92(-4)
80	100(+20)	100(+0)
81	82(+1)	80(-2)

- In theoretical speaking: this method remove the dependency between two consecutive pixels and reduce the entropy.
- Better prediction, entropy of the difference are smaller.

Predictive coding

“Lena” image



1-D prediction error of “Lena” image

Entropy = 3.06 bits/pixel

Predictive coding

- 2-D relationship

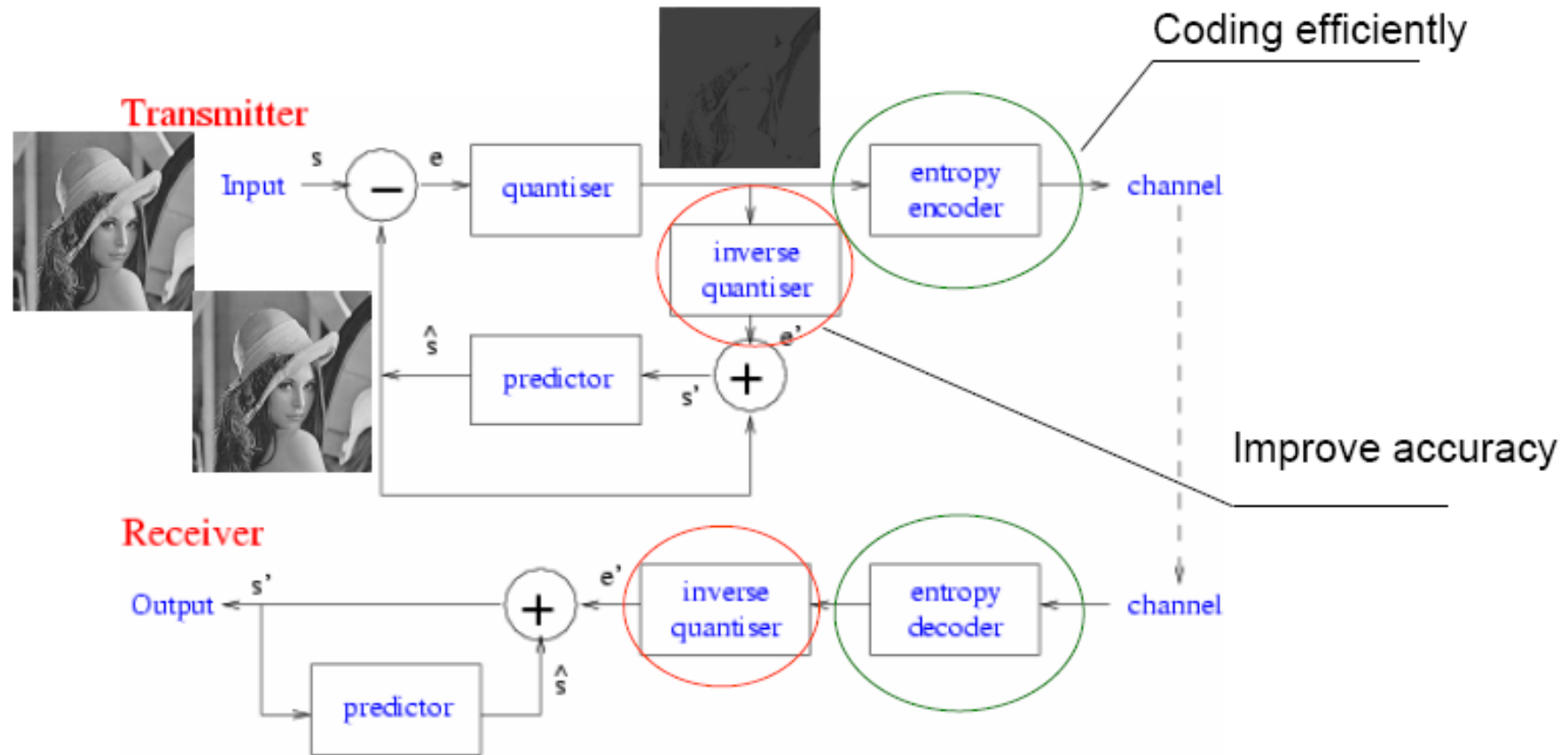
B	C	D
A	X	

$$\hat{x} = k_1 A + k_2 B + k_3 C + k_4 D$$



Entropy = 1.44. bits/pixel

Predictive coding



Quantization

- Data lost during quantization process.
- However, if the quantization is performed carefully in frequency domain, it is not easy to distinguish the difference.

Quantization



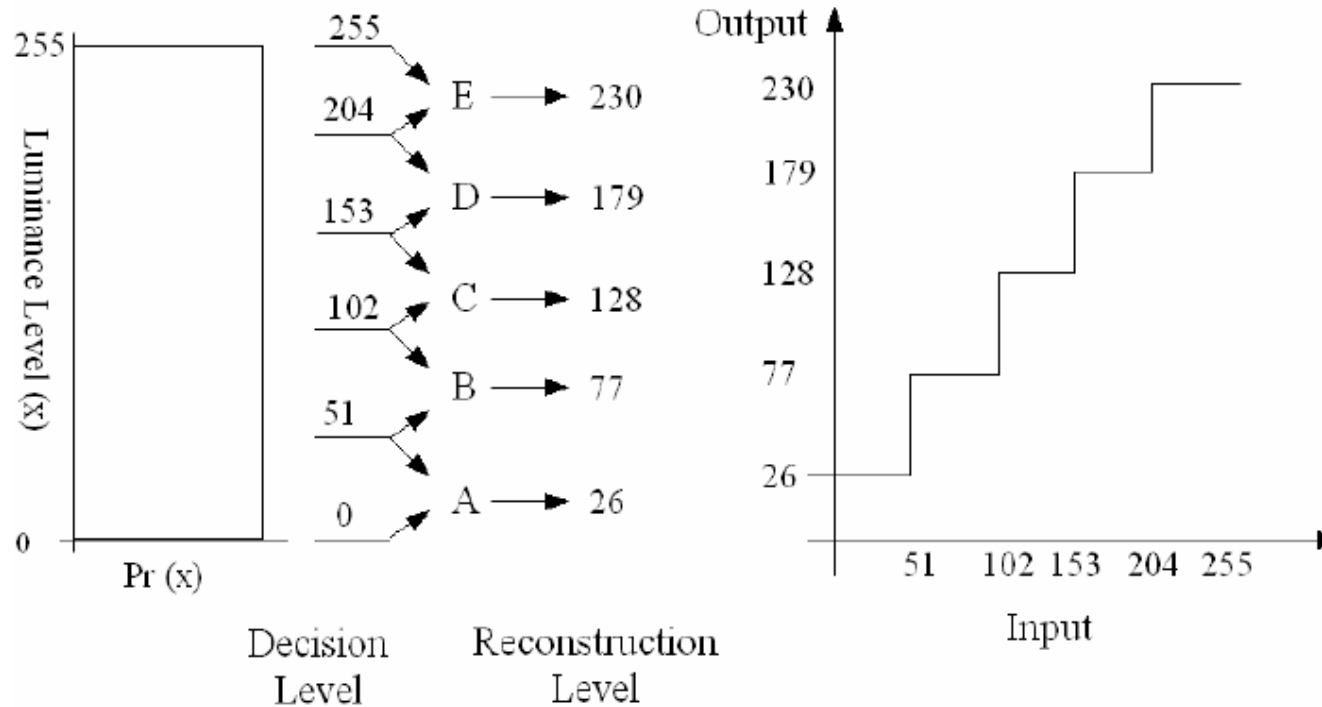
Original



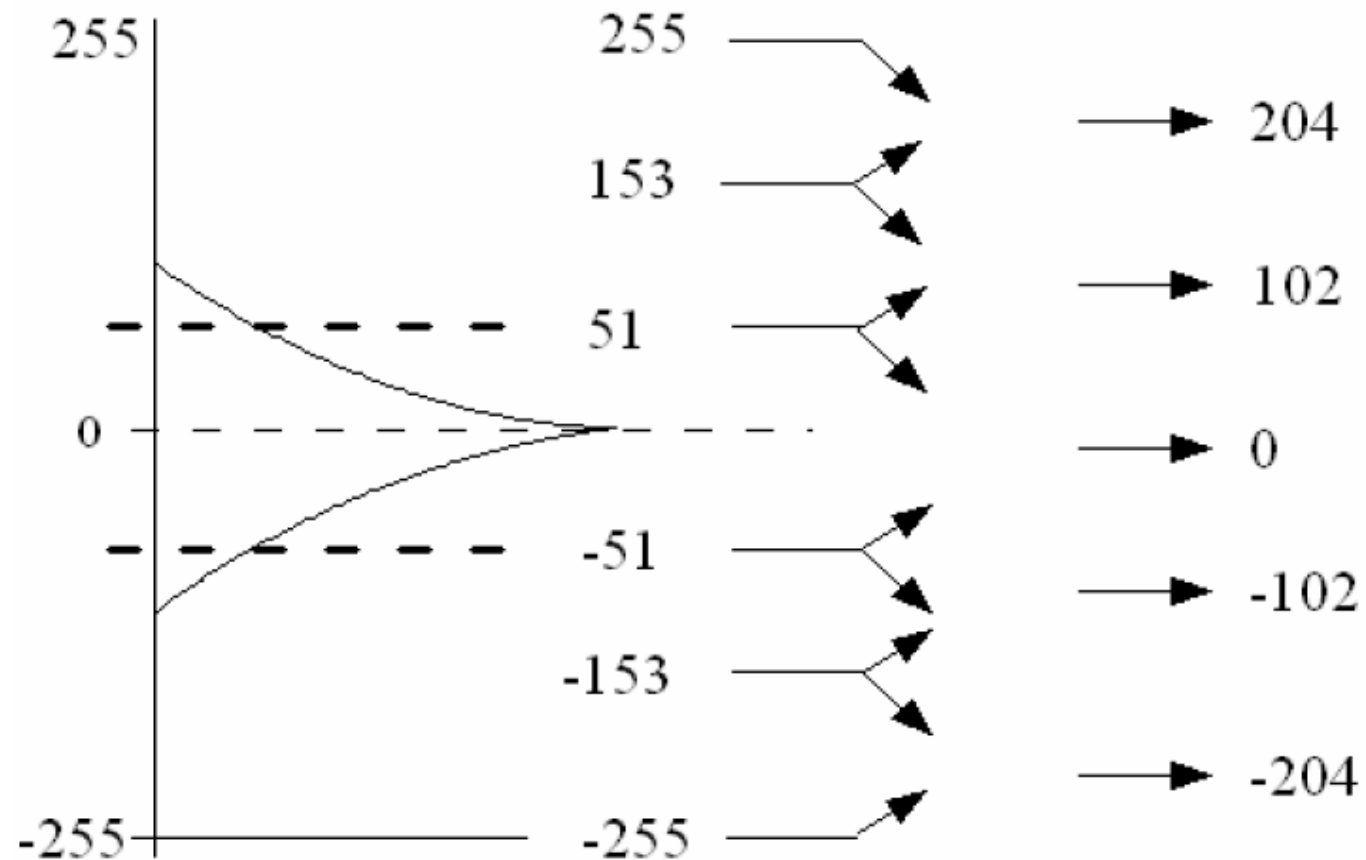
JPEG (compressed 70%)

Quantization

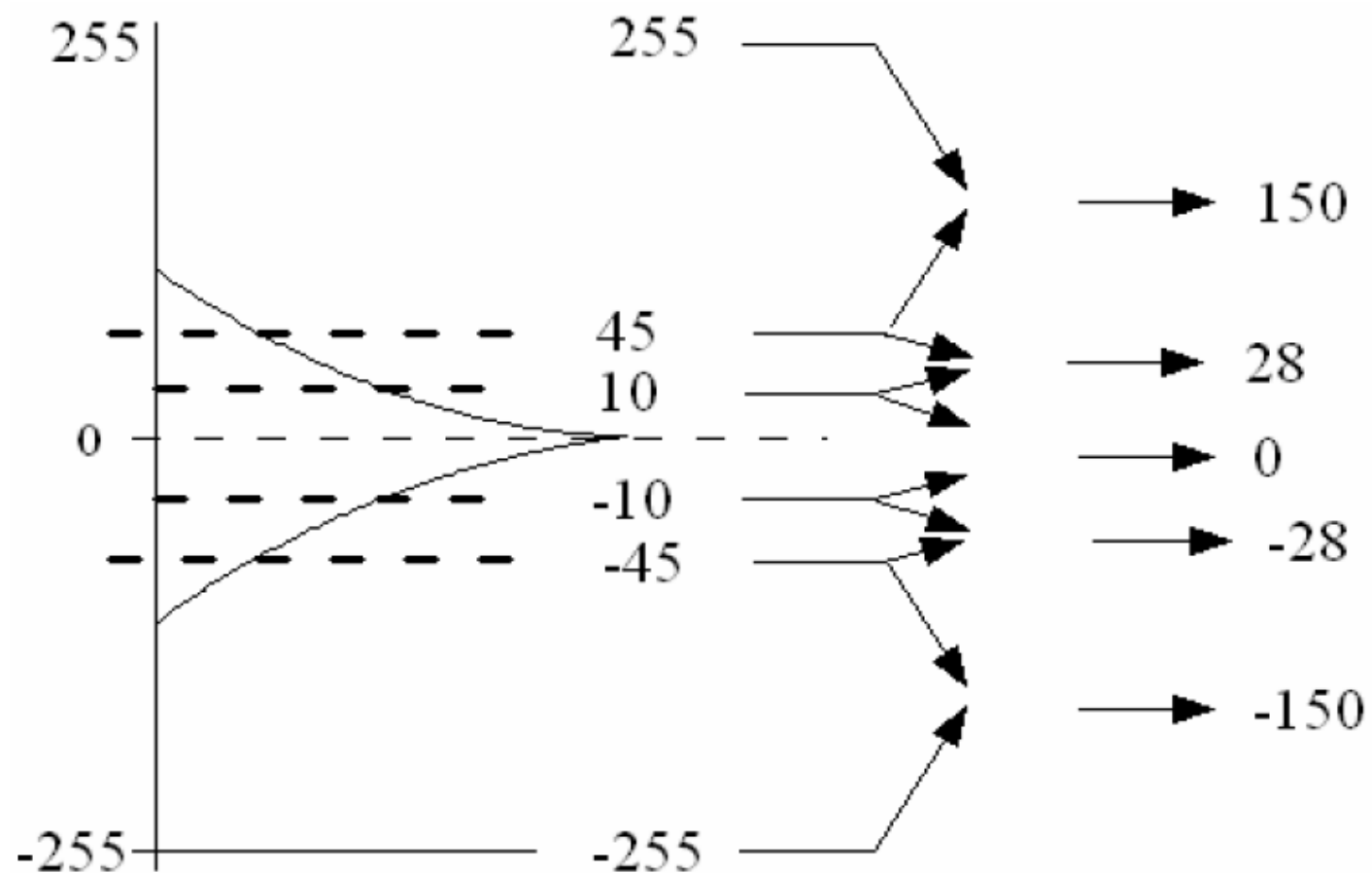
- Uniform quantization: use the same quantization step. Max quantization error is $\frac{1}{2}$ quantization step



Uniform quantization



Non-uniform quantization



Biến đổi DCT 2D

$$A = \begin{bmatrix} \frac{1}{2} \cos(0) & \frac{1}{2} \cos(0) & \frac{1}{2} \cos(0) & \frac{1}{2} \cos(0) \\ \sqrt{\frac{1}{2}} \cos\left(\frac{\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{3\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{5\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{7\pi}{8}\right) \\ \sqrt{\frac{1}{2}} \cos\left(\frac{2\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{6\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{10\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{14\pi}{8}\right) \\ \sqrt{\frac{1}{2}} \cos\left(\frac{3\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{9\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{15\pi}{8}\right) & \sqrt{\frac{1}{2}} \cos\left(\frac{21\pi}{8}\right) \end{bmatrix}$$

– Được xấp xỉ thành:

$$A = \begin{bmatrix} 0.5 & 0.5 & 0.5 & 0.5 \\ 0.653 & 0.271 & -0.271 & -0.653 \\ 0.5 & -0.5 & -0.5 & 0.5 \\ 0.271 & -0.653 & 0.653 & -0.271 \end{bmatrix}$$

Biến đổi DCT 2D

- Ví dụ: Khối X được lấy từ ảnh, và nó được chuyển đổi như sau:

	$j = 0$	1	2	3
$i = 0$	5	11	8	10
1	9	8	4	12
2	1	10	11	4
3	19	6	15	7

$$\mathbf{Y}' = \mathbf{AX} = \begin{bmatrix} 17 & 17.5 & 19 & 16.5 \\ -6.981 & 2.725 & -6.467 & 4.125 \\ 7 & -0.5 & 4 & 0.5 \\ -9.015 & 2.660 & 2.679 & -4.414 \end{bmatrix}$$

$$\mathbf{Y} = \mathbf{AXA}^T = \begin{bmatrix} 35.0 & -0.079 & -1.5 & 1.115 \\ -3.299 & -4.768 & 0.443 & -9.010 \\ 5.5 & 3.029 & 2.0 & 4.699 \\ -4.045 & -3.010 & -9.384 & -1.232 \end{bmatrix}$$

Biến đổi Hadamard

- Là một dạng biến đổi Fourier đơn giản
- Được sử dụng cho nén frame I trong nén video
- H_m là ma trận $2^m \times 2^m$ phần tử, được định nghĩa đệ quy như sau:

$$H_0 = [1] \quad H_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \quad H_m = \frac{1}{\sqrt{2}} \begin{bmatrix} H_m & H_m \\ H_m & -H_m \end{bmatrix}$$

Biến đổi Hadamard

$$H_2 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

$$Y = H_2 \cdot X \cdot H_2^T = H_2 \cdot X \cdot H_2$$

JPEG

- JPEG is a compression standard for still images. It was accepted as an international standard in 1992:
 - Developed by Joint Photographic Expert Group of ISO/IEC
 - Is to compress color or monochrome images.
 - Compression rate is about 10:1

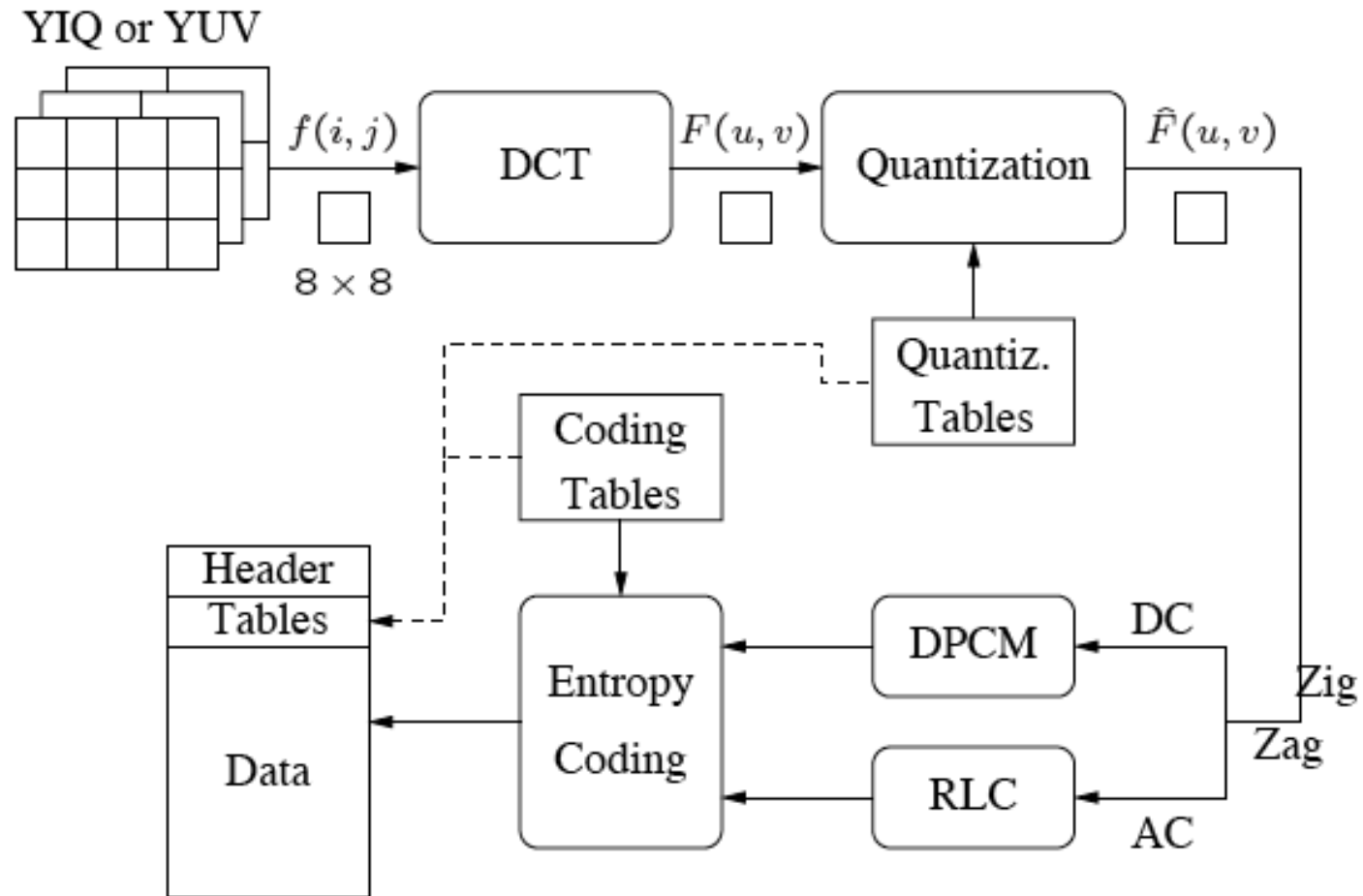
JPEG

- JPEG is lossiness compression method:
 - Based on DCT
 - Compression process is independent from:
 - Image size
 - Color model
 - Image complexity
 - The video compression standard based on JPEG is Motion JPEG (MJPEG)

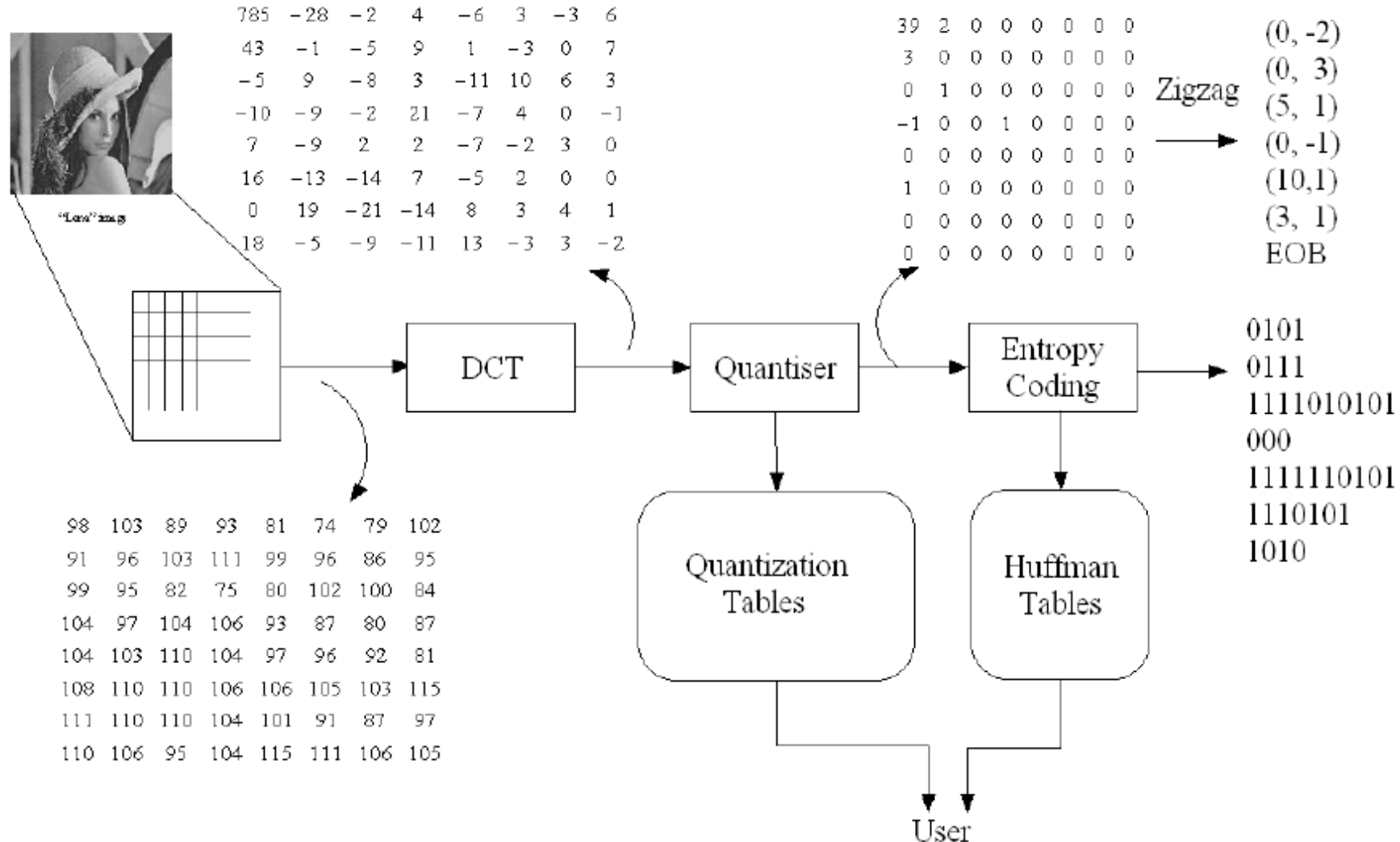
JPEG

- Human eyes sense luminance much better than chrominance:
 - Therefore, JPEG uses sampling of luminance twice as chrominance (4:2:0)

JPEG Encoding



JPEG encoding based on DCT



JPEG encoding

- Encoding step of JPEG:
 - Transform colors from RGB color model to YUV or YIQ.
 - DCT transform for all 8x8 block
 - Quantization
 - For each 8x8 block, scan coefficients with zig-zag pattern and use run-length coding.
 - Entropy coding

DCT transform for each block

- Image is divided into 8x8 blocks
 - 2D DCT transform is applied for all blocks.
 - When compress images with low bitrate. There are blocky discontinuity appearing at the edges of blocks.

Quantization

- Quantization matrix $N_{j,k}$:

For luminance:

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

For chrominance:

17	18	24	47	99	99	99	99
18	21	26	66	99	99	99	99
24	26	56	99	99	99	99	99
47	66	99	99	99	99	99	99
99	99	99	99	99	99	99	99
99	99	99	99	99	99	99	99
99	99	99	99	99	99	99	99
99	99	99	99	99	99	99	99

JPEG encoding based on DCT

- Quantization:

$$\hat{Y}_{j,k} = Q[Y_{j,k}] = \text{round}\left(\frac{Y_{j,k}}{N_{j,k} \times Q_s}\right)$$

- Q_s is user parameter indicating the quality of compression.

DCT transform in smooth area



An 8×8 block from the Y image of 'Lena'

200	202	189	188	189	175	175	175	515	65	-12	4	1	2	-8	5
200	203	198	188	189	182	178	175	-16	3	2	0	0	-11	-2	3
203	200	200	195	200	187	185	175	-12	6	11	-1	3	0	1	-2
200	200	200	200	197	187	187	187	-8	3	-4	2	-2	-3	-5	-2
200	205	200	200	195	188	187	175	0	-2	7	-5	4	0	-1	-4
200	200	200	200	200	190	187	175	0	-3	-1	0	4	1	-1	0
205	200	199	200	191	187	187	175	3	-2	-3	3	3	-1	-1	3
210	200	200	200	188	185	187	186	-2	5	-2	4	-2	2	-3	0
$f(i, j)$								$F(u, v)$							

Fig. 9.2: JPEG compression for a smooth image block.

Quantization

32	6	-1	0	0	0	0	0
-1	0	0	0	0	0	0	0
-1	0	1	0	0	0	0	0
-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	0

$\hat{F}(u, v)$

512	66	-10	0	0	0	0	0
-12	0	0	0	0	0	0	0
-14	0	16	0	0	0	0	0
-14	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	0

$\tilde{F}(u, v)$

Inverse DCT transform

```
199 196 191 186 182 178 177 176
201 199 196 192 188 183 180 178
203 203 202 200 195 189 183 180
202 203 204 203 198 191 183 179
200 201 202 201 196 189 182 177
200 200 199 197 192 186 181 177
204 202 199 195 190 186 183 181
207 204 200 194 190 187 185 184
```

$$\tilde{f}(i, j)$$

```
1  6 -2  2  7 -3 -2 -1
-1  4  2 -4  1 -1 -2 -3
0 -3 -2 -5  5 -2  2 -5
-2 -3 -4 -3 -1 -4  4  8
0  4 -2 -1 -1 -1  5 -2
0  0  1  3  8  4  6 -2
1 -2  0  5  1  1  4 -6
3 -4  0  6 -2 -2  2  2
```

$$\epsilon(i, j) = f(i, j) - \tilde{f}(i, j)$$

DCT transform in detailed area



Another 8×8 block from the Y image of 'Lena'

```

70 70 100 70 87 87 150 187
85 100 96 79 87 154 87 113
100 85 116 79 70 87 86 196
136 69 87 200 79 71 117 96
161 70 87 200 103 71 96 113
161 123 147 133 113 113 85 161
146 147 175 100 103 103 163 187
156 146 189 70 113 161 163 197

```

$f(i, j)$

```

-80 -40 89 -73 44 32 53 -3
-135 -59 -26 6 14 -3 -13 -28
47 -76 66 -3 -108 -78 33 59
-2 10 -18 0 33 11 -21 1
-1 -9 -22 8 32 65 -36 -1
5 -20 28 -46 3 24 -30 24
6 -20 37 -28 12 -35 33 17
-5 -23 33 -30 17 -5 -4 20

```

$F(u, v)$

Quantization and inverse quantization

-5	-4	9	-5	2	1	1	0
-11	-5	-2	0	1	0	0	-1
3	-6	4	0	-3	-1	0	1
0	1	-1	0	1	0	0	0
0	0	-1	0	0	1	0	0
0	-1	1	-1	0	0	0	0
0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0

$\hat{F}(u, v)$

-80	-44	90	-80	48	40	51	0
-132	-60	-28	0	26	0	0	-55
42	-78	64	0	-120	-57	0	56
0	17	-22	0	51	0	0	0
0	0	-37	0	0	109	0	0
0	-35	55	-64	0	0	0	0
0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0

$\tilde{F}(u, v)$

Inverse DCT and the errors

```

70  60 106  94  62 103 146 176
85 101  85  75 102 127  93 144
98  99  92 102  74  98  89 167
132 53 111 180  55  70 106 145
173 57 114 207 111  89  84  90
164 123 131 135 133  92  85 162
141 159 169  73 106 101 149 224
150 141 195  79 107 147 210 153

```

$\tilde{f}(i, j)$

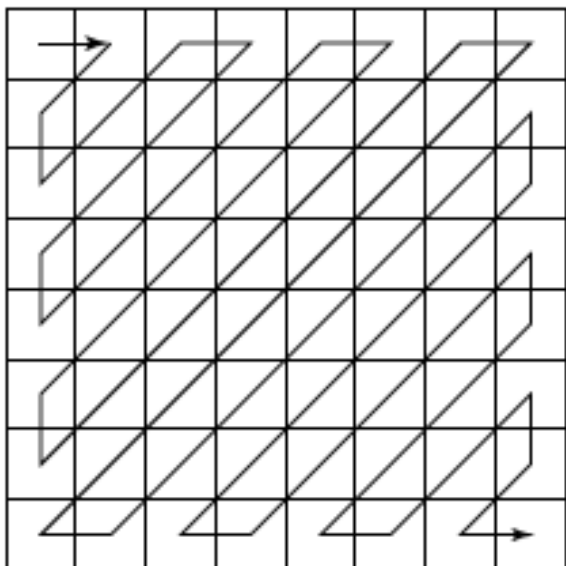
```

0  10  -6 -24  25 -16  4  11
0  -1  11  4 -15  27  -6 -31
2 -14  24 -23  -4 -11  -3  29
4  16 -24  20  24  1  11 -49
-12 13 -27  -7  -8 -18  12  23
-3  0  16  -2 -20  21  0  -1
5 -12  6  27  -3  2  14 -37
6  5  -6  -9  6  14 -47  44

```

$\epsilon(i, j) = f(i, j) - \tilde{f}(i, j)$

Zig-zag scan



32	6	-1	0	0	0	0	0
-1	0	0	0	0	0	0	0
-1	0	1	0	0	0	0	0
-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	0

(**32**, 6, -1, -1, 0, -1, 0, 0, 0, -1, 0, 0, 1, 0, 0, ..., 0)

Hệ số DC

Các hệ số AC

Run-length coding for AC coefficients

- Run-length coding represent each AC coefficient (after zig-zag scanning) as a pair (Runlength, value).
 - Runlength: number of zeros before a non-zero coef.
 - Value: value of non-zero number.
- For example:
 - Sequence: 32, 6, -1, -1, 0, -1, 0, 0, 0, -1, 0, 0, 1, 0, 0, ..., 0
 - After run-length coding: (0,6) (0,-1) (0,-1) (1,-1) (3,-1) (2,1) (0,0). Skip the first coefficient (DC).

Coding for DC coefficient

- Each block has a DC coefficient
- DC values vary not much in neighbor blocks. Use zig-zag scan to read DC values block by block.
- Use **Differential Pulse Code Modulation** (DPCM) for coding DC values:
 - The current DC value is predicted from the previous one.
 - For example, first 5 DC values: 150, 155, 149, 152, 144.
 - DPCM codes are: 150, 5, -6, 3, -8.

Coding for DC coefficient

- Run-length coded values of AC and DC coefficients are entropy coded:
 - Each value is presented by a pair of symbols (Size, amplitude)
 - Size: Number of bits to present the value and coded by Huffman codes.
 - Amplitude has uniform distribution, then no compression needed.