

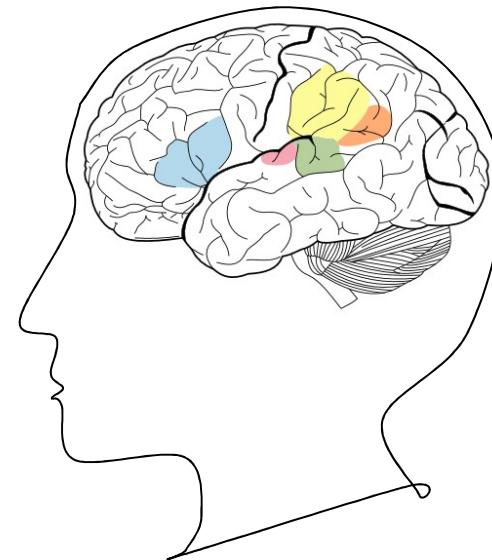
Introduction to Multimedia Compression



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Data is Everything, and Nothing

- ❑ Everyday, we are bombarded by all kinds of information (data)
 - textbooks, news papers, movies, songs, conversations, lectures, preaching, ...
- ❑ What are the purposes of all the information?
 - To cause chemical reactions in your brains
 - To “duplicate” chemical reactions from one brain to the other
 - To ...



Three Aspects of Data

- ❑ To distribute information (data) around to serve your purposes, there are three aspects one should worry about:
 - Quantity
 - Reliability
 - Security

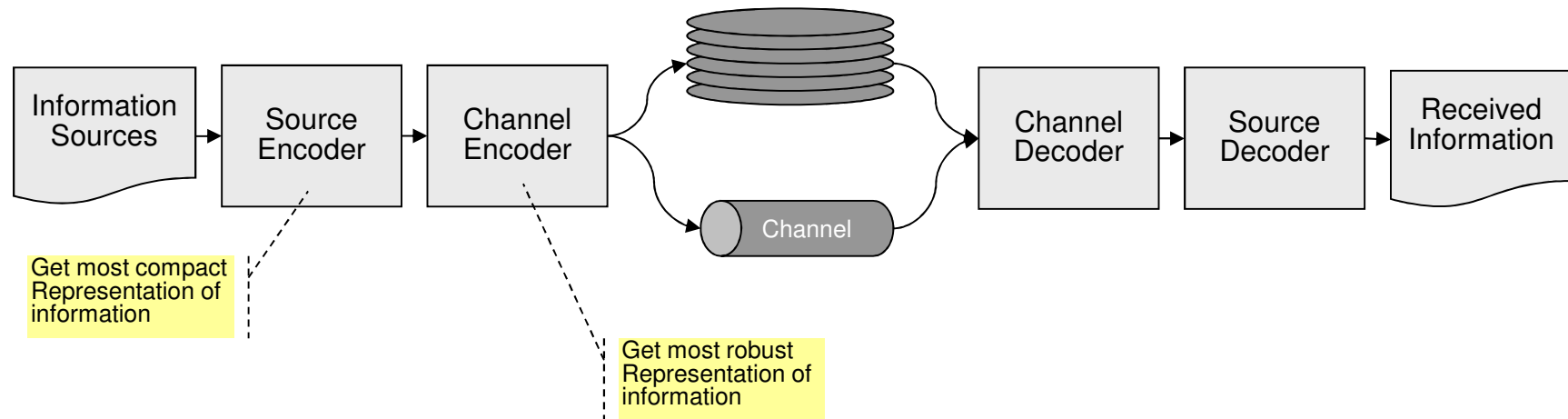
- ❑ Example: “knowledge of multimedia compression”

The Paper That Starts It All ...

- ❑ In 1948, Claude E. Shannon published the revolutionary paper, “A Mathematical Theory of Communication.”
 - Later, in 1949, a book was published based on this paper, but the first word of the title was changed from “A” to “*The*”
- ❑ The paper provides many insights into the essence of the communication problem
 - In particular, Shannon perceived that all communication is essentially digital !

Data Distribution Systems

- Shannon was the first person to partition a communication system as follows:



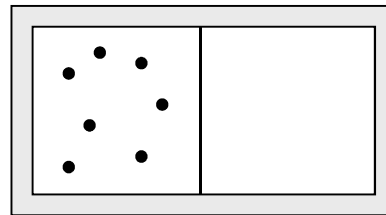
→ The information content of a source and the information capacity of a channel can be identified using the concept of **entropy**

The Origin of Information Theory

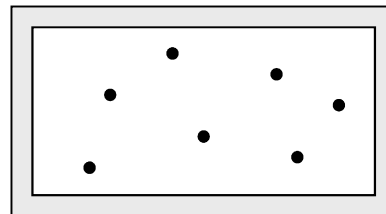
- ❑ The term “entropy” was first used in thermodynamics and in statistical mechanics
 - Some people think that information theory grew out of statistical mechanics because L. Szilard applied an idea of information to solve a physical problem in 1929
- ❑ However, Shannon’s work evolved from the field of electrical communication
 - “entropy” was used in information theory merely due to its mathematical analogy with the entropy of statistical mechanics

Entropy in Thermodynamics

- ❑ In thermodynamics, entropy is a measure of thermal energy of a body of gas



Low entropy



High entropy

- ❑ Statistical mechanics says that an increase in entropy means a decrease in predictability

Linking Back to Information Theory

- ❑ The complexity of a system depends on our knowledge of the system; the more we know about the system, the less words we need to “describe” the system
- ❑ In information theory, the amount of information conveyed by a message increases as the amount of uncertainty as to what message actually will be produced becomes greater

Some “Information”

- ❑ Check the “entropy” of the following messages
 - My dog cannot fly
 - My dog runs faster than a chicken
 - My dog is a lady dog
 - My dog runs slower than a chicken
 - My dog can sing

- ❑ It seems that, a rare message carries more information than a common message

Frequency-based Coding

□ Morse code

- Invented in 1838 by Morse for electrical telegraph, and expanded by Vail in 1844
- To shorten the transmission of messages, English text was coded based on relative frequencies of occurrence
- The efficiency of Morse code can only be improved by 15% using modern theory[†]
- Questions: efficient for all languages?

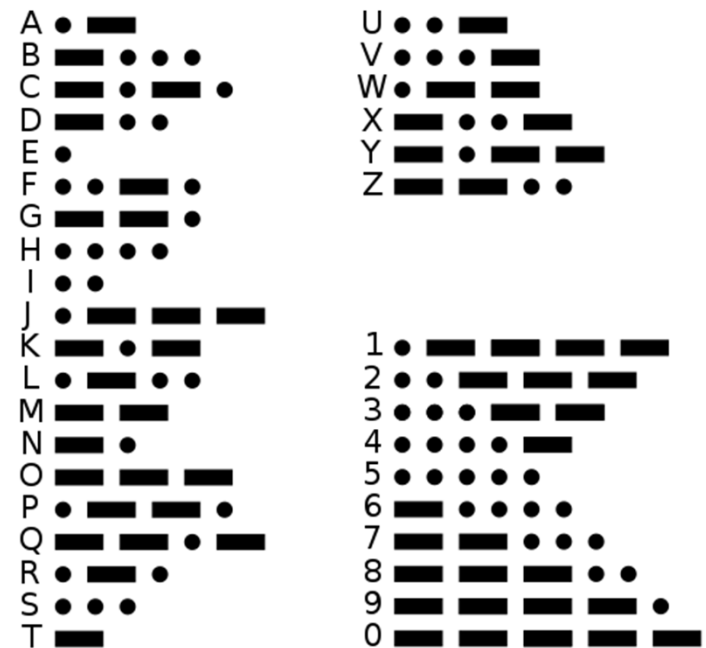


fig. ref.: wikipedia

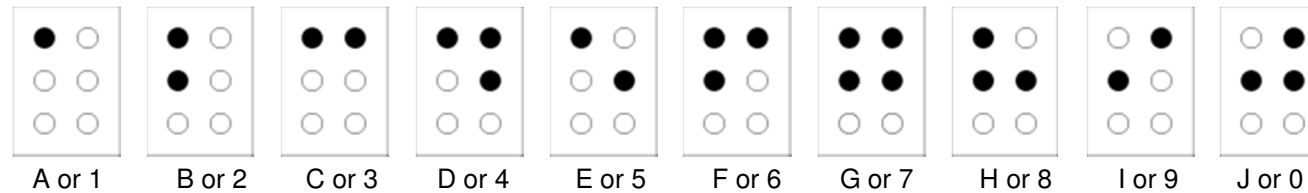
[†] J. R. Pierce, *An Introduction to Information Theory, 2nd. Ed.*, Dover Publications, 1980.

Context-based Coding

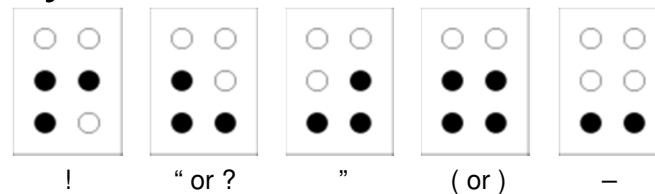
❑ Braille code, by Louis Braille in 1825

❑ Grade 1 Braille

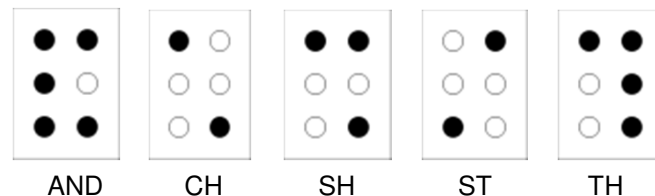
■ Letters and numbers



■ Symbols



❑ Grade 2 Braille



Model-based Coding

- ❑ Statistical structure is not the only way of compression. Describing “things” using “models” is usually less wordy
- ❑ For example, what is the minimal precise description of π ?
 - Shannon’s idea – the unpredictability of patterns of digits in π
 - Kolmogorov’s idea – the size of a program that computes π

How Large Is the Amount of Data?

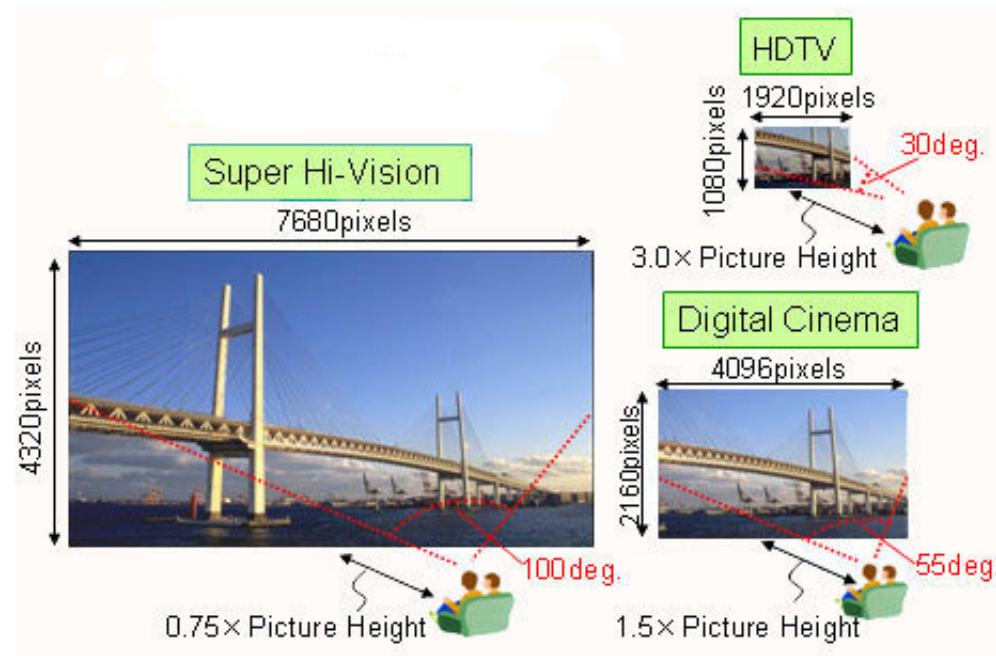
- ❑ 1 second of CD audio:
 - $44100 \text{ samples} \times 2 \text{ channels} \times 16 \text{ bits/sample}$
= 1,411,200 bits
- ❑ 1 second of 1080p HD video:
 - $1920 \times 1080 \text{ pixels} \times 3 \text{ color channels} \times 8 \text{ bits/color sample}$
 $\times 30 \text{ frames}$
= 1,492,992,000 bits

Sometimes, large data amount is a technique against piracy

The Future Is Here Already

❑ Super Hi-Vision† (8K system)

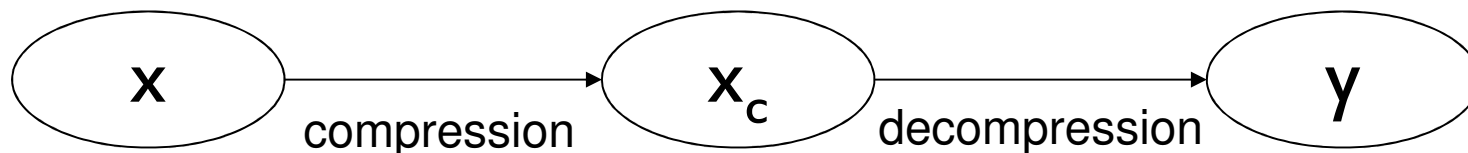
- $7680 \times 4320 = 33$ Mega pixels per frame
- Physical data rate (video-only): 180-600 Mbps



† http://www.nhk.or.jp/digital/en/super_hi/

Data Compression Concept

- ❑ X – original data, X_c – compressed representation, y – reconstruction



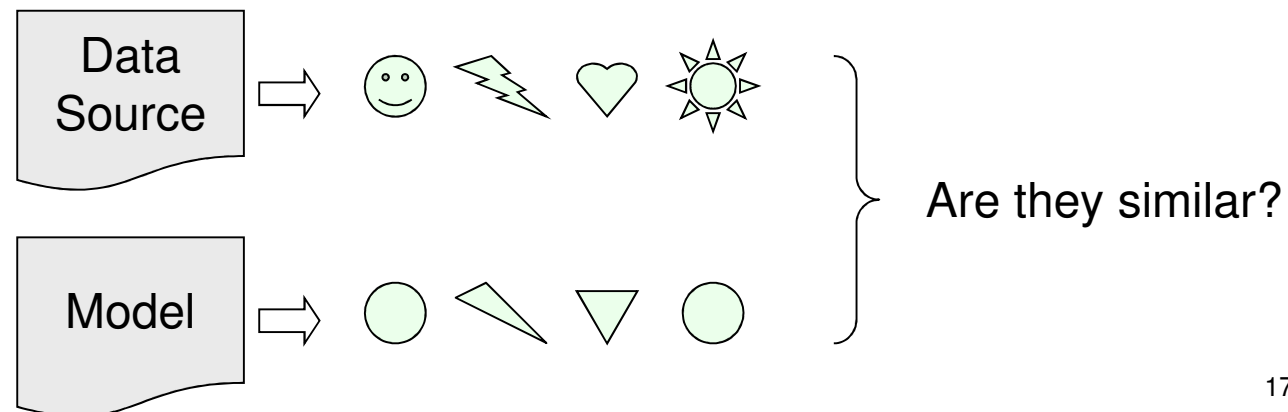
- ❑ Lossless compression: when y is equal to x
- ❑ Lossy compression: when y is different from x
- ❑ Compression ratio: $|X| : |X_c|$ or $(|X| - |X_c| / |X|) * 100\%$
 - For example, $|X| = 65536$ bytes, $|X_c| = 16384$ bytes, the compression ratio is 4:1 or 75%.
- ❑ Data rate: for time-varying data, the number of bits per second (or sample) required to represent the data

Lossless and Lossy Compressions

- ❑ Text compression techniques are often lossless
 - Any counter examples?
- ❑ Image, audio, video compression techniques are often lossy
 - Any counter examples?
- ❑ Distortion: the difference between the original and the reconstruction
 - If the distortion is small, we say the “quality” or “fidelity” is high. Or, we say the reconstruction is a “high-definition” copy of the original

Modeling and Coding

- ❑ One of the most powerful tools in data compression is called “data modeling”
 - Model – a systematic way to describe data
- ❑ A common data compression scheme is to “encode” a description of the model, and a description of how the data differ from the model (aka, residual)
 - By encode, we mean to put the data in binary digits



Example 1: Linear Model

- Data sequence y_i ,

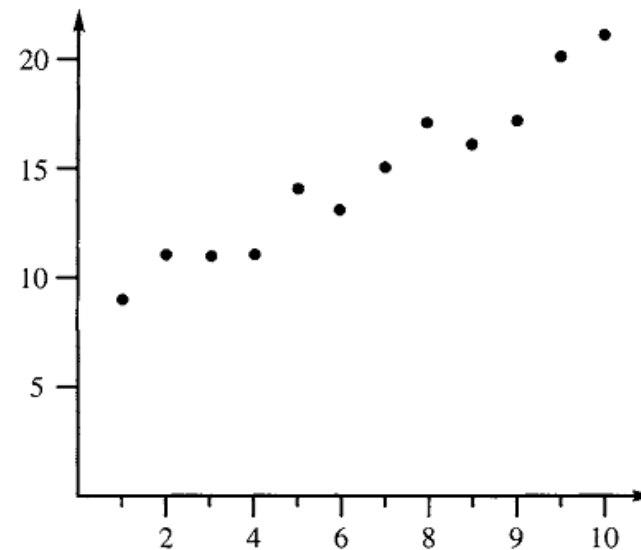
9	11	11	11	14	13	15	17	16	17	20	21	...
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- Model: $\hat{y}_n = n + 8, n = 1, 2, \dots$

Model parameter: 1, 8

- Residual:

$$\begin{aligned} e_n &= y_n - \hat{y}_n \\ &= 0, 1, 0, -1, 1, -1, 0, \dots \end{aligned}$$



Example 2: Differential Model

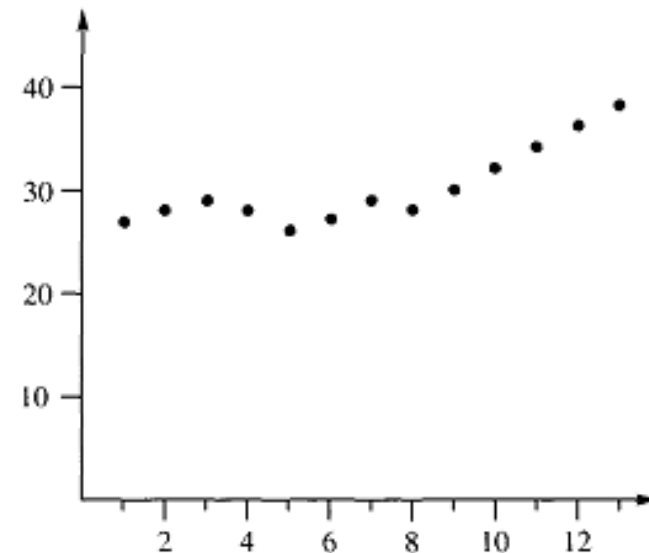
- Data sequence y_i ,

27	28	29	26	27	29	28	30	32	34	36	38	...
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- Model: $\hat{y}_1 = 0, \hat{y}_n = y_{n-1}, n = 2, 3, \dots$

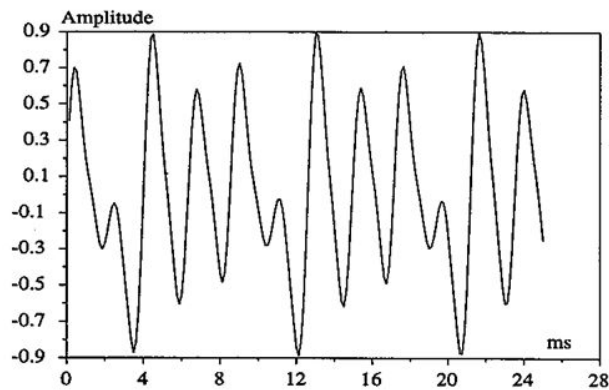
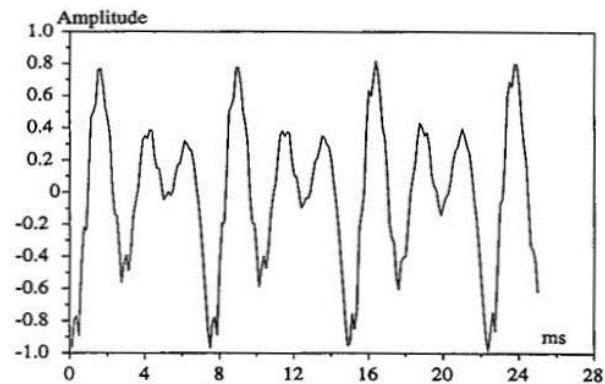
- Residual:

$$\begin{aligned} e_n &= y_n - \hat{y}_n \\ &= 27, 1, 1, -1, -2, 1, 2, \dots \end{aligned}$$

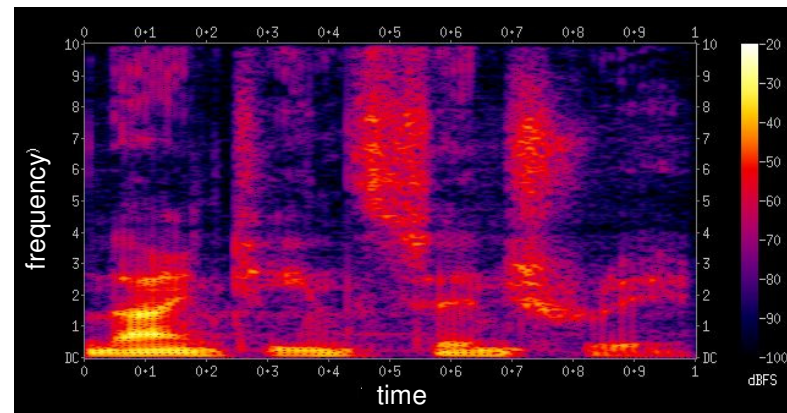


What About Speech Models?

- Typical speech signals:



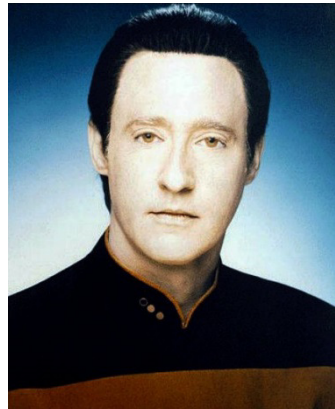
waveform



spectrogram

What About Image Models?

- Typical image signals:



Example 3: Variable Length Coding

- Given a sequence of symbols:

a n a r a n n a r a n f a r f a a r w a y

- If fixed length coding (FLC) is used: 3 bits per symbol
- If variable length coding (VLC) is used: 2.58 bits per symbol
→ 1.16 : 1 compression ratio

TABLE 1.1 A code with codewords of varying length.

<i>a</i>	1
<i>n</i>	001
<i>∅</i>	01100
<i>f</i>	0100
<i>n</i>	0111
<i>r</i>	000
<i>w</i>	01101
<i>y</i>	0101