

Department of Computer Science and Information Engineering
National Taiwan University of Science and Technology

Introduction to Multimedia Signal Processing

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2018

What is multimedia?

- Multimedia includes a combination of :
 - Text
 - Still images
 - Audio
 - Animation
 - Video
 - Interactivity content forms

[2]

Examples

*Aperture, in Geometry, is the Inclination of Lines which meet in a Point.
Aperture in Opticks, is the Hole next to the Object Glass of a Telescope, thro' which the Light and Image of the Object comes into the Tube, and thence it is carried to the Eye.*

Text



Audio



Still Images



Animation



Video Footage

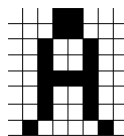


Interactivity

[3]

Text

- When PC's were in their infancy running under MS-DOS
 - They only displayed text in one size and one color.
 - A text charter was 8 pixels high and 8 pixels wide

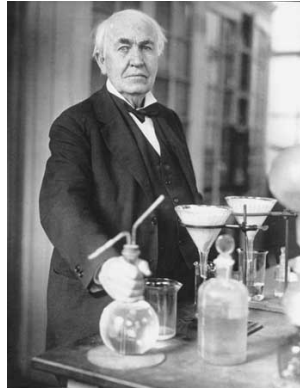


- Nowadays



[4]

Still Images



Thomas Alva Edison 1847 – 1931

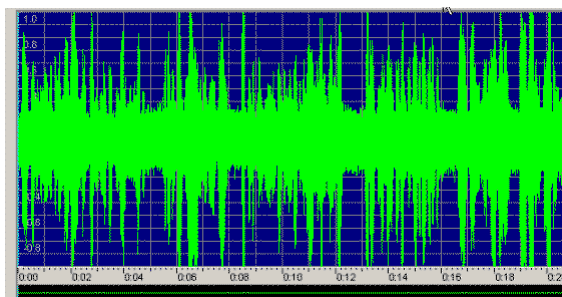


Mark Zuckerberg (1984)
Facebook founder

Processor chips can only copy data in byte multiples (8, 16, 32, or 64 bits).
[5]

Audio

- 1877 Thomas Alva Edison Demonstrated his cylinder phonograph



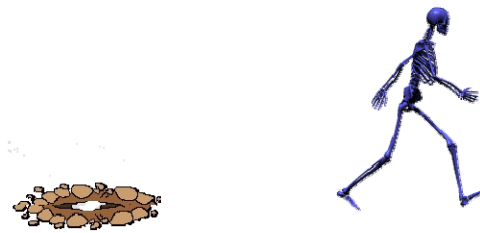
The image above is a map of the above wave file.

[6]



Animation

- Animation is a sequential series of still images that create an illusion of motion.



[7]

Video

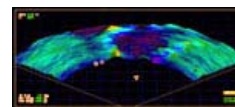
- Unlike an animation that we can create from drawings or images a video movie is created by a photographic process and converted or ported to a computer.

[8]

Introduction to Image and Video Processing

[9]

Images and Videos are Efficient and Effective



Sonar Image

A political ad



<http://marsrovers.jpl.nasa.gov/gallery/press/opportunity/20040125a.html>
JPL Mars' Panorama captured by the Opportunity

- Visual representations are often the most efficient way to represent information
- Images are used in many scenarios:
 - Politics, advertisements, scientific research, military, etc...
- Video signals can effectively tell a temporally evolving story
 - Arise in cinema (motion pictures)
 - Arise in surveillance
 - Arise in medical applications

[10]

Why Do We Process Images?

- Enhancement and restoration
 - Remove artifacts and scratches from an old photo/movie
 - Improve contrast and correct blurred images
- Transmission and storage
 - Images and Video can be more effectively transmitted and stored
- Information analysis and automated recognition
 - Recognizing terrorists
- Evidence
 - Careful image manipulation can reveal information not present
 - Detect image tampering
- Security and rights protection
 - Encryption and watermarking preventing illegal content manipulation

[11]

Compression

- Color image of 600x800 pixels
 - Without compression
 - $600 \times 800 \times 24 \text{ bits/pixel}$
= 11.52K bits = 1.44M bytes
 - After JPEG compression (popularly used on web)
 - only 89K bytes
 - compression ratio ~ 16:1
- Movie
 - 720x480 per frame, 30 frames/sec, 24 bits/pixel
 - Raw video ~ 243M bits/sec
 - DVD ~ about 5M bits/sec
 - Compression ratio ~ 48:1



"Library of Congress" by M.Wu (600x800)

[12]

Denoising



From X.Li <http://www.ee.princeton.edu/~lixin/denoising.htm>

[13]

Deblurring



Blurred & noisy image

Restored image

From Mathworks

[14]

Visual Mosaicing

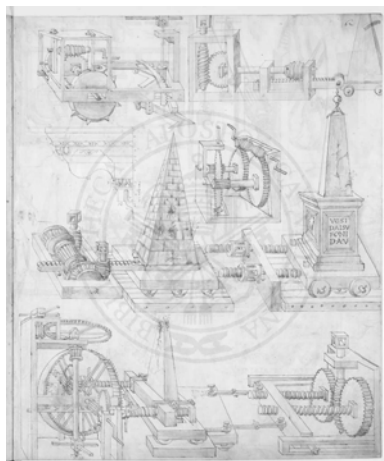
- Stitch photos together without thread or scotch tape



R.Radke – IEEE PRMI journal paper draft 5/01

[15]

Visible Digital Watermarks

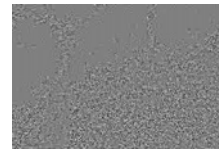
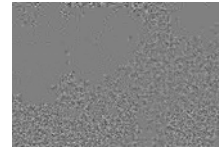


from IBM Watson web page
"Vatican Digital Library"



[16]

Invisible Watermark

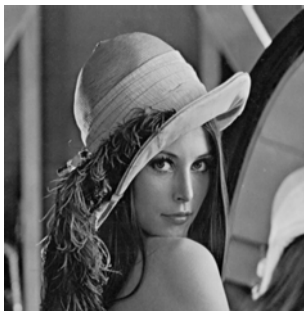


- 1st & 30th Mpeg4.5Mbps frame of original, marked, and their luminance difference
- human visual model for imperceptibility: protect smooth areas and sharp edges

[17]

Error Concealment

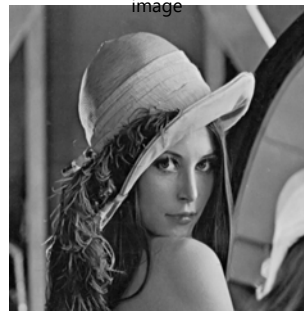
(a) original lenna image



(b) corrupted lenna image



(c) concealed lenna image



25% blocks in a checkerboard pattern are corrupted

corrupted blocks are concealed via edge-directed interpolation

Examples were generated using the source codes provided by W.Zeng.

[18]

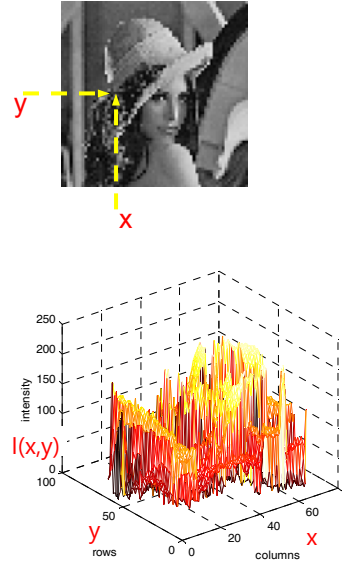
What is An Image?

- Grayscale image
 - A grayscale image is a function $I(x,y)$ of the two spatial coordinates of the image plane.
 - $I(x,y)$ is the intensity of the image at the point (x,y) on the image plane.
 - $I(x,y)$ takes non-negative values
 - assume the image is bounded by a rectangle $[0,a] \times [0,b]$

$$I: [0, a] \times [0, b] \rightarrow [0, \text{inf})$$

- Color image

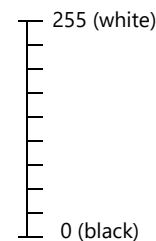
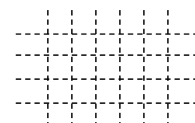
- Can be represented by three functions, $R(x,y)$ for red, $G(x,y)$ for green, and $B(x,y)$ for blue.



[19]

Sampling and Quantization

- Computer handles “discrete” data.
- Sampling
 - Sample the value of the image at the nodes of a regular grid on the image plane.
 - A pixel (picture element) at (i, j) is the image intensity value at grid point indexed by the integer coordinate (i, j) .
- Quantization
 - Is a process of transforming a real valued sampled image to one taking only a finite number of distinct values.
 - Each sampled value in a 256-level grayscale image is represented by 8 bits.



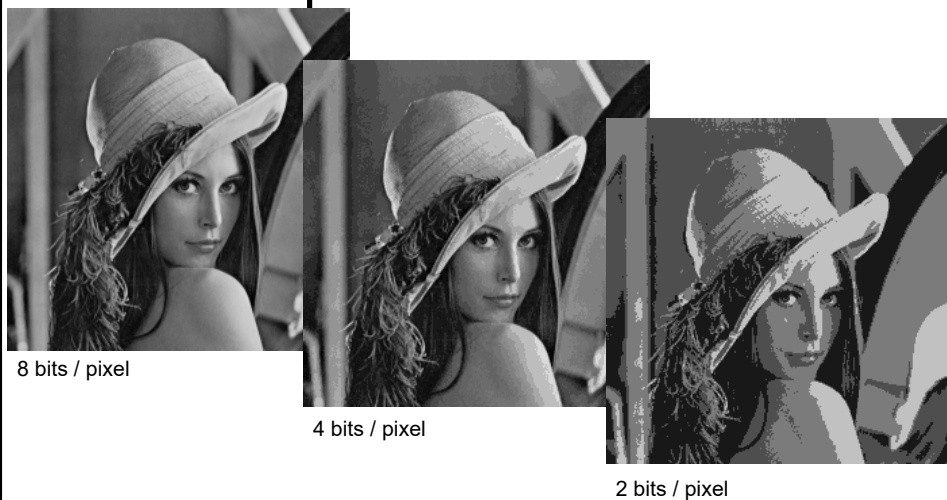
[20]

Examples of Sampling



[21]

Examples of Quantizaion



[22]

Introduction to Multimedia Data Hiding

[23]

Why Hide Data in Multimedia

[24]

Crypto is Useful to Protect MM, but Not Enough ...

- Encryption
 - Helps to protect confidentiality
 - Protection vanishes after decryption
 - Prefer a way to associate copyright info. with multimedia source even after decryption/compression/transmission/etc.
- Digital cryptographic signature
 - Helps to authenticate sender's identity and data integrity
 - Need to attach a separate signature to the data source
 - Audio/image/video allows imperceptible changes
 - Opportunities for new and seamless ways of authentication

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Demands on Info. Security and Protection

- Intellectual property management for digital media
 - Promising electronic marketplace for digital music and movies
 - Napster controversy
- Conventional encryption alone still leaves many problems unsolved
 - Protection from encryption vanishes once data is decrypted
 - Still want establish ownership and restrict illegal re-distributions
 - How to distinguish changes introduced by compression vs. malicious tampering?
 - Bit-by-bit accuracy is not always desired authenticity criterion for MM

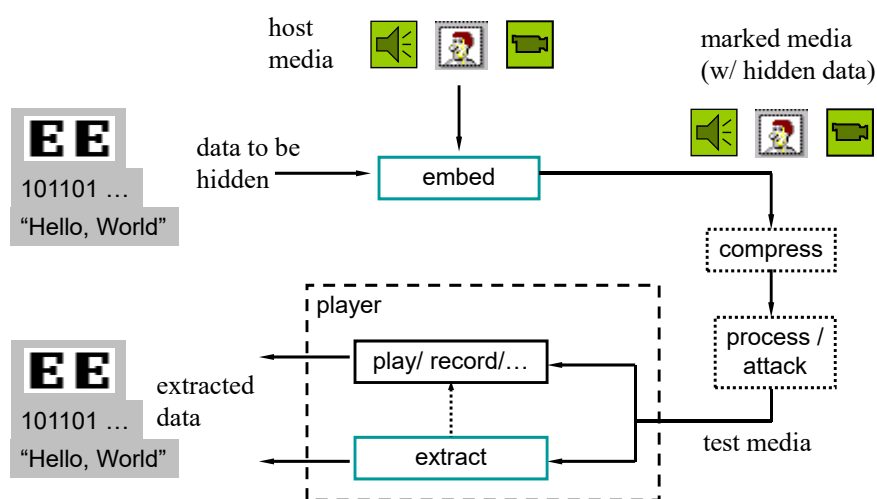
[26]

Multimedia Data Hiding / Digital Watermarking

- What is it?
 - Example: Picture in picture, words in words
 - It is “Steganography”
 - Secondary information in perceptual digital media data
- Why hide information?
 - Convey information without an additional channel
 - Hidden data is tied to content: Can be made difficult to separate data from content.
 - Covert communication: Adversary does not know it is there!

[27]

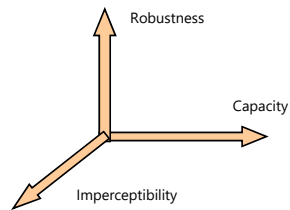
General Framework of Data Hiding



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Issues and Challenges

- Tradeoff among conflicting requirements
 - Imperceptibility
 - Robustness & security
 - Capacity
- Key elements of data hiding
 - Perceptual model
 - Embedding one bit
 - Multiple bits
 - Uneven embedding capacity
 - Robustness and security
 - What data to embed

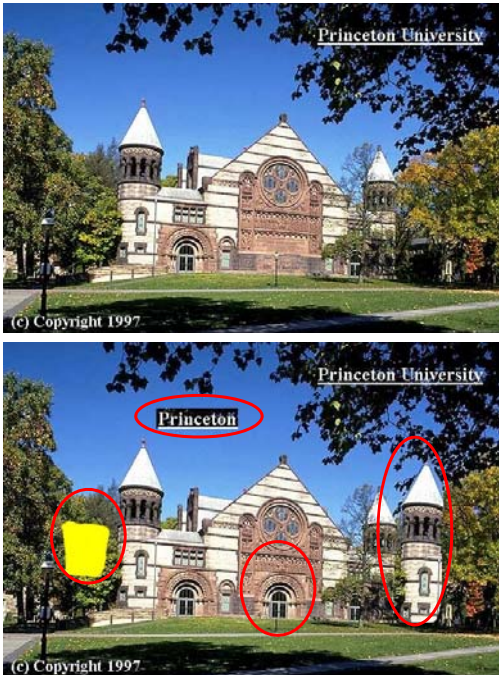


Upper Layers
	Coding of embedded data
	Security
	Error correction
	Uneven capacity equalization
Lower Layers	Multiple-bit embedding
	Imperceptible embedding of one bit

[29]

Examples of Multimedia Data Hiding

[30]




(c) Copyright 1997.

Watermark-based Authentication

- Embed patterns and content features using a lookup-table
- High embedding capacity/security via shuffling
 - ♦ *locate alteration*
 - ♦ *differentiate content vs. non-content change (compression)*

unchanged

content changed

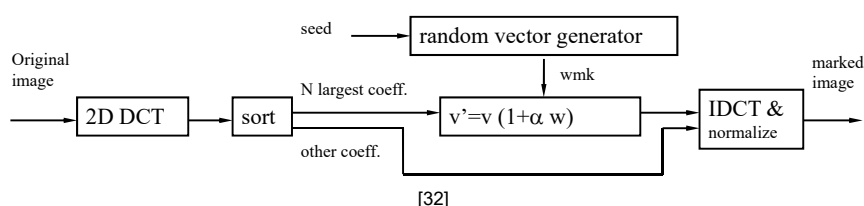


Spread Spectrum Approach: Cox et al (NECI)

- Key points
 - Place wmk in perceptually significant spectrum (for robustness)
 - Modify by a small amount below Just-noticeable-difference (JND)
 - Use long random vector as wmk to avoid artifacts (for imperceptibility & robustness)
- Embedding

$$v'_i = v_i + \alpha v_i w_i = v_i (1 + \alpha w_i) \quad \alpha = 0.1$$

$$w_i \sim \text{iid, zero mean, unit variance}$$
 - Perform DCT on entire image and embed wmk in DCT coeff.
 - Choose N=1000 largest AC coeff. and scale $\{v_i\}$ by a random factor



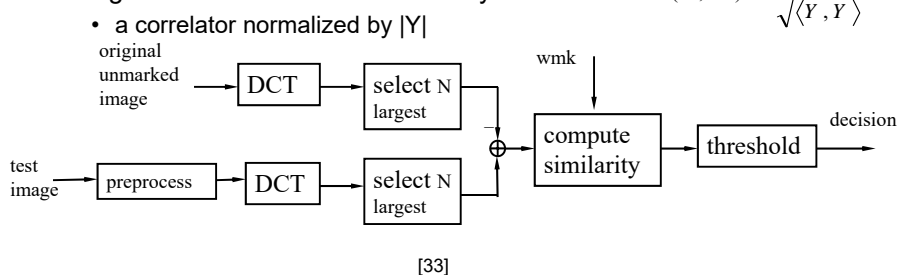
Cox's Scheme (cont'd)

- Detection

- Subtract original image from the test one before running through detector

$$\begin{array}{lcl}
 \text{orig } X & \text{[image icon]} & \\
 X' = X + W + N ? & & \\
 \text{test } X' & \text{[yellow image icon]} & \\
 \end{array}
 \quad
 \begin{array}{c}
 \oplus \\
 \downarrow \\
 Y = X' - X \Rightarrow \begin{cases} H_0: Y = N & \text{no watermark} \\ H_1: Y = W + N & \text{watermark} \end{cases}
 \end{array}$$

- Original detection measure used by Cox et al. $\text{sim}(Y, W) = \frac{\langle Y, W \rangle}{\sqrt{\langle Y, Y \rangle}}$



Theoretical Foundations

- Optimal detection for On-Off Keying (OOK)

$$\begin{cases} H_0: y_i = d_i & (\text{if } b = 0) \\ H_1: y_i = s_i + d_i & (\text{if } b = 1) \end{cases} \text{ for } i = 1 \sim n$$

- OOK under i.i.d. Gaussian noise $\{d_i\}$
 - $b \in \{0, 1\}$ represents absence vs. presence of ownership mark
 - Use a correlator-type detector (recall the review last week)

- Need to determine how to choose $\{s_i\}$

- Neyman-Pearson Detection [Poor's book Sec.2.4]

- False-alarm ~ claiming wmk existence when nothing embedded
- Given max. allowed false-alarm, try to minimize prob. of miss detection
 - Use likelihood ratio as detection statistic
 - Determine threshold according to false-alarm prob.

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Performance of Cox's Scheme

- Robustness

- (claimed) scaling, JPEG, dithering, cropping, “printing-xeroxing-scanning”, multiple watermarking

Distortion	none	scale 25%	JPG 10%	JPG 5%	dither	crop 25%	print- xerox- scan
similarity	32.0	13.4	22.8	13.9	10.5	14.6	7.0

threshold = 6.0 (determined by setting false alarm probability)

- No big surprise with high robustness

- equiv. to conveying just 1-bit {0,1} with $O(10^3)$ samples

- Comment

- Must store original unmarked image \Rightarrow “private wmk”, “non-blind” detect.
- Perform image registration if necessary
- Adjustable parameters: N and α

[35]

Invisible Robust Wmk: Improved Schemes

- Apply better Human-Perceptual-Model

- Global scaling factor is not suitable for all coefficients
- Explicitly computes Just-noticeable-difference (JND)
 - JND \sim max amount each freq. coeff. can be modified imperceptibly
 - Use α_i for each coeff. \Rightarrow finely tune wmk strength

$$v_i' = v_i \cdot (1 + \alpha_i \cdot w_i)$$
- Better tradeoff between imperceptibility and robustness
 - Try to add a watermark as strong as possible

- Block-DCT based schemes:

- Podilchuk-Zeng & Swanson et al.
- Existing visual model for block DCT: JPEG Quality Factors

[36]

Compare Cox & Podilchuk Schemes



Original

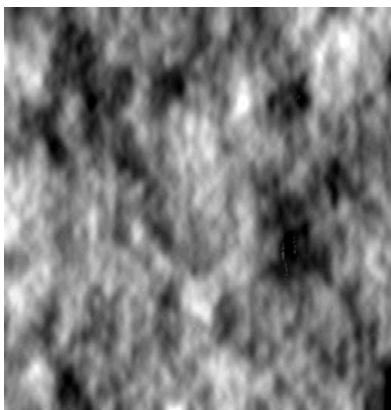
Cox
whole image DCT
Embed in 1000 largest coeff.

Podilchuk
block-DCT
Embed to all

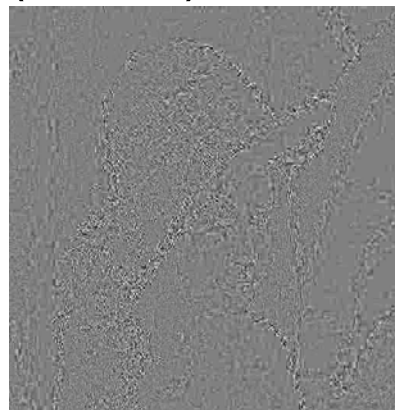
"embeddables"

[37]

Compare Cox & Podilchuk Schemes (cont'd)



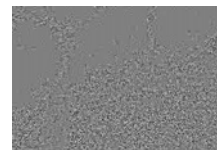
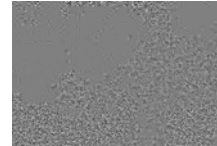
Cox



Podilchuk

[38]

Video Example



- 1st & 30th Mpeg4.5Mbps frame of original, marked, and their luminance difference
- human visual model for imperceptibility: protect smooth areas and sharp edges

[39]

Summary and Conclusions

- Data hiding in digital multimedia for a variety of purposes, involving multiple disciplines
- Tradeoff among many criteria
- Important to think both as designer and as attacker
- Emerging problems
 - how to effectively combine with other security mechanisms
- Data hiding in market
 - digital cameras with authentication watermark module
 - plug-in for image editors
 - video watermark proposals for DVD copy control
 - on-going effort for digital music
 - “Digital Rights Management (DRM)” for multimedia data

[40]

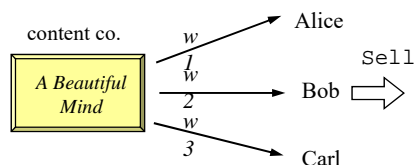
Suggested reading

- I. Cox, J. Kilian, T. Leighton, T. Shamoon: “Secure Spread Spectrum Watermarking for Multimedia”, IEEE Transaction on Image Processing, vol.6, no.12, pp.1673-1687, 1997.
 - Download from IEEE online journal, or
<http://www.neci.nj.nec.com/homepages/ingemar/papers/ip97.ps>
- C. Podilchuk and W. Zeng, “Image Adaptive Watermarking Using Visual Models,” IEEE Journal Selected Areas of Communications (JSAC), vol.16, no.4, May, 1998.
 - Download from IEEE online journal
- M. Wu and B. Liu: “Multimedia Data Hiding”, Springer Verlag, to appear Dec. 2002. An earlier dissertation version is at
http://www.ece.umd.edu/~minwu/research/phd_thesis.html.

[41]

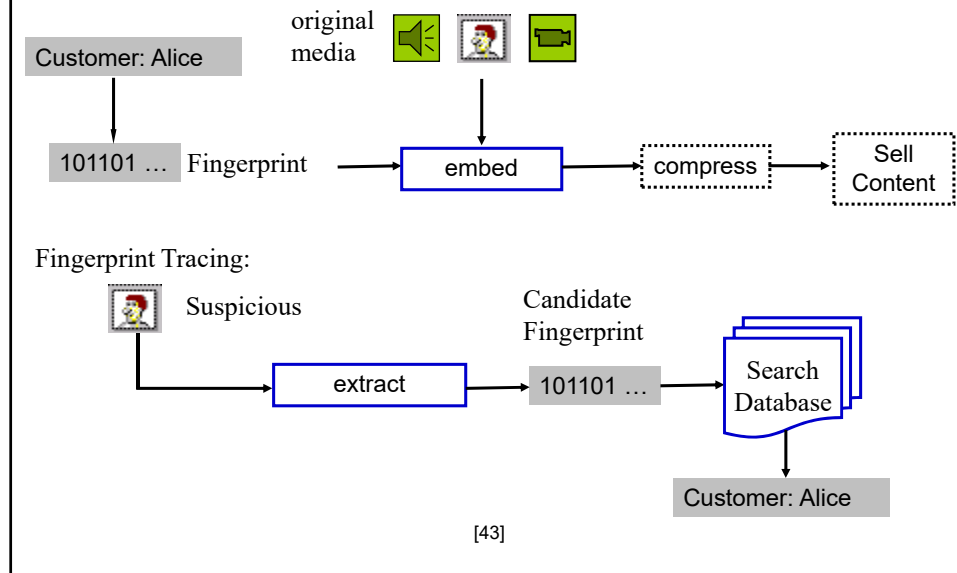
Digital Fingerprinting and Tracing Traitors

- Recent advances in communications allow for convenient information sharing
- Severe damage is created by unauthorized information leakage
 - e.g., pirated content or classified document
- Promising countermeasure: robustly embed digital fingerprints
 - Insert ID or “fingerprint” to identify each customer
 - Prevent improper redistribution of multimedia content



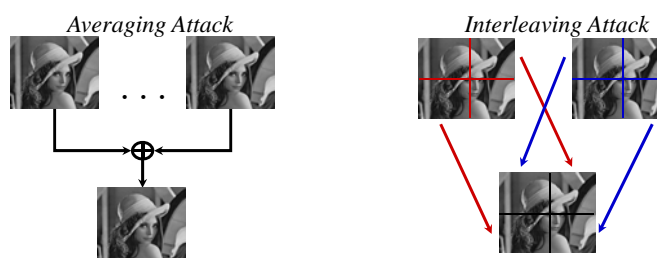
[42]

Embedded Fingerprinting for Multimedia



Collusion Scenarios

- **Collusion: A cost-effective attack against multimedia fingerprints**
 - Users with same content but different fingerprints come together to produce a new copy with diminished or attenuated fingerprints
- **Colluders cannot arbitrarily manipulate embedded fingerprint bits**
 - Different from Boneh-Shaw's marking assumption (1995)
- **Result of fair collusion: energy of embedded fingerprints decreases**



Anti-Collusion Fingerprinting

- Build anti-collusion fingerprinting to trace traitors and colluders
- Potential impact
 - Gather digital evidence and trace culprits
 - Deter unauthorized dissemination in the first place
 - let the bad guys know their high risk of being caught
 - Potential civilian use for digital rights management (DRM)
 - DRM business ~ \$96M in 2000 and projected \$3.5B in 2005

[45]

Introduction to data compression

[46]

Why data compression?

- Transmission and storage

Telephony	8000 samples/second × 12 bits/sample =
(200–3400 Hz):	96 kbps
Wideband speech	16,000 samples/second × 14 bits/sample =
(50–7000 Hz):	224 kbps
Wideband audio	44,100 samples/second × 2 channels ×
(20–20,000 Hz):	16 bits/sample = 1.412 Mbps
Images:	512 × 512 pixel color image × 24 bits/pixel =
	6.3 Mbits/image
Video:	640 × 480 pixel color image × 24 bits/pixel ×
	30 images/second = 221 Mbps
HDTV:	1280 × 720 pixel color image × 60 images/second
	× 24 bits/pixel = 1.3 Gbps

Approximate Bit Rates for Uncompressed Sources

- For uncompressed video

- CD-ROM (650MB) could store $650\text{MB} \times 8 / 221\text{Mbps} \approx 23.5$ seconds
- DVD-5 (4.7GB) could store about 3 minutes

[47]

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Approximate Bit Rates for Uncompressed Sources

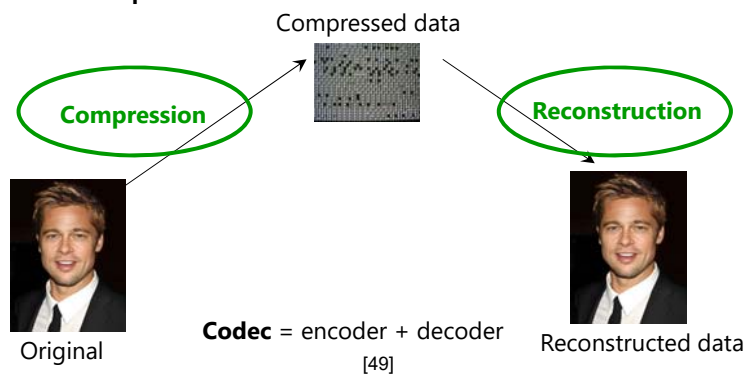
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[48]

What is data compression?

- To represent information in a compact form (as few bits as possible)
- Technique



Technique (cont.)

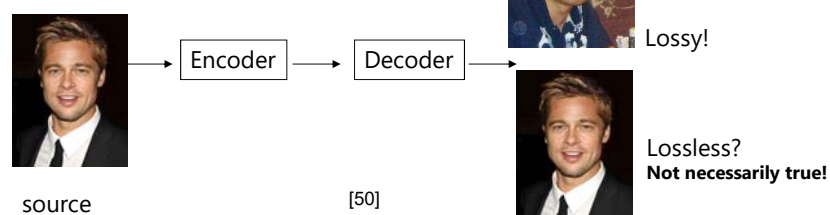
- Lossless
 - The reconstruction is *identical* to the original.

Do not send money!

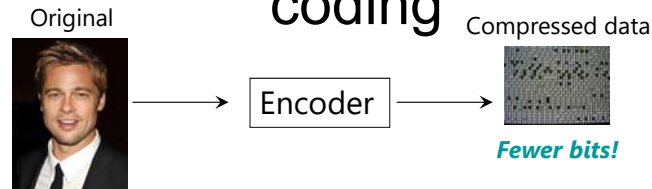
Do now send money!

- Lossy
 - Involves loss of information

Example: image codec



Two phases: modeling and coding



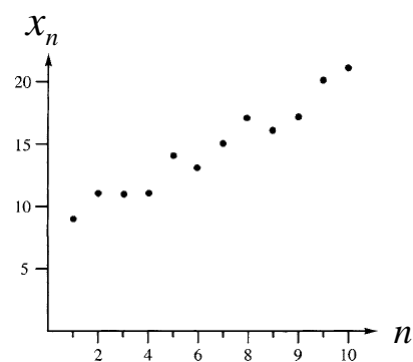
- Modeling
 - Discover the structure in the data
 - Extract information about any redundancy
- Coding
 - Describe the model and the **residual** (how the data differ from the model)

[51]

Example: Modeling

Data: x_n

9	11	11	11	14	13	15	17	16	17	20	21
---	----	----	----	----	----	----	----	----	----	----	----



Model: $\hat{x}_n = n + 8 \quad n = 1, 2, \dots$

[52]

Example: Coding

Original data x_n

9	11	11	11	14	13	15	17	16	17	20	21
---	----	----	----	----	----	----	----	----	----	----	----

Model $\hat{x}_n = n + 8$

9	10	11	12	13	14	15	16	17	18	19	20
---	----	----	----	----	----	----	----	----	----	----	----

Residual $e_n = x_n - \hat{x}_n$

0	1	0	-1	1	-1	0	1	-1	-1	1	1
---	---	---	----	---	----	---	---	----	----	---	---

- $\{-1, 0, 1\}$
- 2 bits * 12 samples = 24 bits (compared with 60 bits before compression)

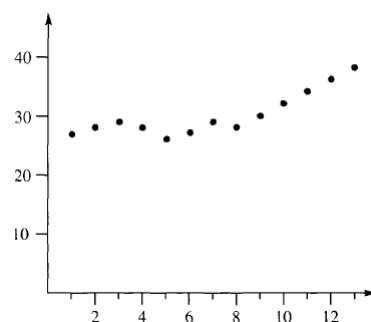
We use the model to **predict** the value, then encode the residual!

[53]

Example (2)

27	28	29	28	26	27	29	28	30	32	34	36	38
----	----	----	----	----	----	----	----	----	----	----	----	----

27	1	1	-1	-2	1	2	-1	2	2	2	2	2
----	---	---	----	----	---	---	----	---	---	---	---	---

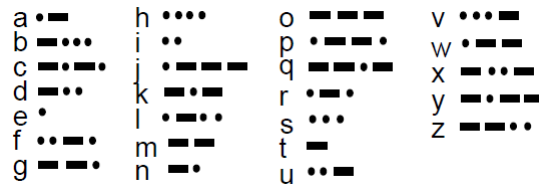


$$\hat{x}_n = x_{n-1} \quad n = 2, 3, \dots$$

[54]

Example (4)

- Morse Code (1838)



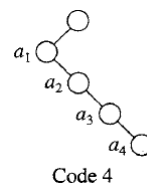
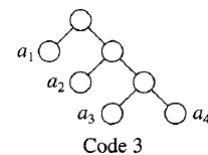
Shorter codes are assigned to letters that occur more frequently!

[55]

Prefix Codes

- No codeword is a *prefix* to another codeword
- **Uniquely decodable**

Letters	Code 2	Code 3	Code 4
a_1	0	0	0
a_2	1	10	01
a_3	00	110	011
a_4	11	111	0111



[56]

Coding (cont.)

- For compression
 - Uniquely decodable
 - Short codewords
 - Instantaneous (easier to decode)
- 1. What sets of code lengths are possible?
- 2. Can we always use prefix codes?

[57]

Introduction to Huffman codes

[58]

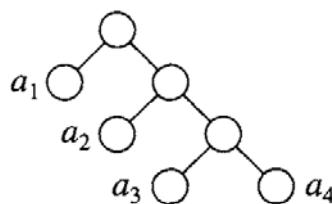
Outline

- Huffman codes
 - Basic form and the algorithm
 - Length of Huffman codes
 - Extended form

[59]

Huffman Coding

- Observations of prefix codes



- Frequent symbols have shorter codewords
- The two symbols that occur **least frequently** have the same length

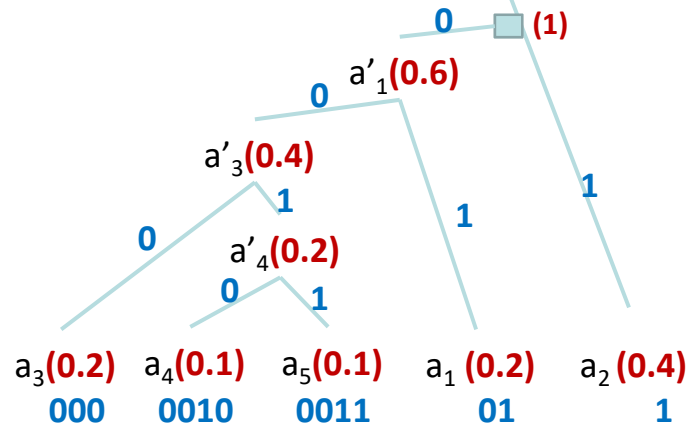
Huffman procedure:

Two least frequent symbols differ only in the last bit. Ex: *m0*, *m1*

[60]

Algorithm

- $A = \{a_1, a_2, a_3, a_4, a_5\}$
- $P(a_1) = 0.2, P(a_2) = 0.4, P(a_3) = 0.2, P(a_4) = 0.1, P(a_5) = 0.1$

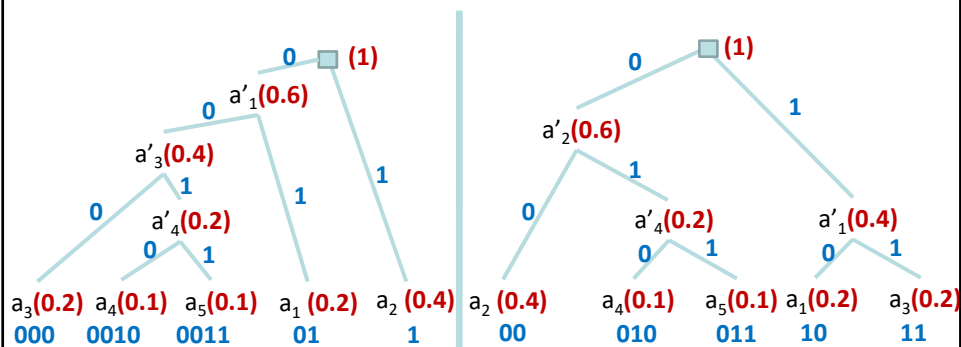


[61]

Algorithm

Letter	Probability	Codeword
a_2	0.4	1
a_1	0.2	01
a_3	0.2	000
a_4	0.1	0010
a_5	0.1	0011

- Entropy: 2.122 bits/symbol
- Average length: 2.2 bits/symbol



Which code is preferred? *The one with minimum variance!*

Length of Huffman Codes

- Lower and upper bounds

$$H(\mathcal{S}) \leq \bar{l} < H(\mathcal{S}) + 1$$

- $H(\mathcal{S})$: entropy
- \bar{l} : average code length

[63]

Extended Huffman Codes (1)

- Consider small alphabet and skewed probabilities

– Example:

symbol	Prob.	codeword
a	0.9	0
b	0.1	1

1 bit / letter
No compression!

- Block multiple symbols together

symbol	Prob.	codeword
aa	0.81	0
ab	0.09	10
ba	0.09	110
bb	0.01	111

0.645 bit / letter

- Bounds: $H(\mathcal{S}) \leq R \leq H(\mathcal{S}) + \frac{1}{n}$

Number of letters per block

[64]

Extended Huffman Codes (2)

Another example:

Letters	$P(a_i)$	codeword
a_1	0.80	1
a_2	0.02	11
a_3	0.18	10

$$H = 0.816 \text{ bits/symbol}$$

$$\bar{l} = 1.2 \text{ bits/symbol}$$

Letters	$P(a_i)$	codeword
$a_1 a_1$	0.6400	0
$a_1 a_2$	0.0160	10101
$a_1 a_3$	0.1440	11
$a_2 a_1$	0.0160	101000
$a_2 a_2$	0.0004	10100101
$a_2 a_3$	0.0036	1010011
$a_3 a_1$	0.1440	100
$a_3 a_2$	0.0036	10100100
$a_3 a_3$	0.0324	1011

$$H = 1.6315 \text{ bits/block} = 0.816 \text{ bits/symbol}$$

$$\bar{l} = 1.7228 / 2 = 0.8614 \text{ bits/symbol}$$

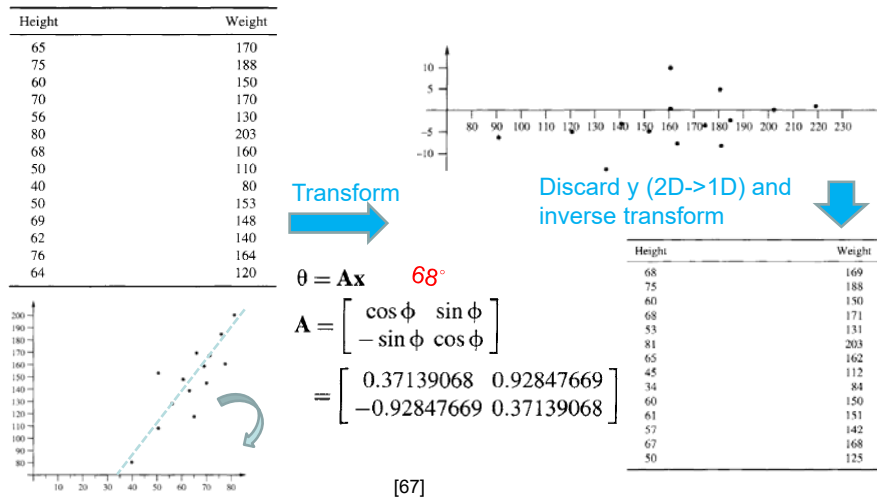
[65]

Introduction to Transform Coding

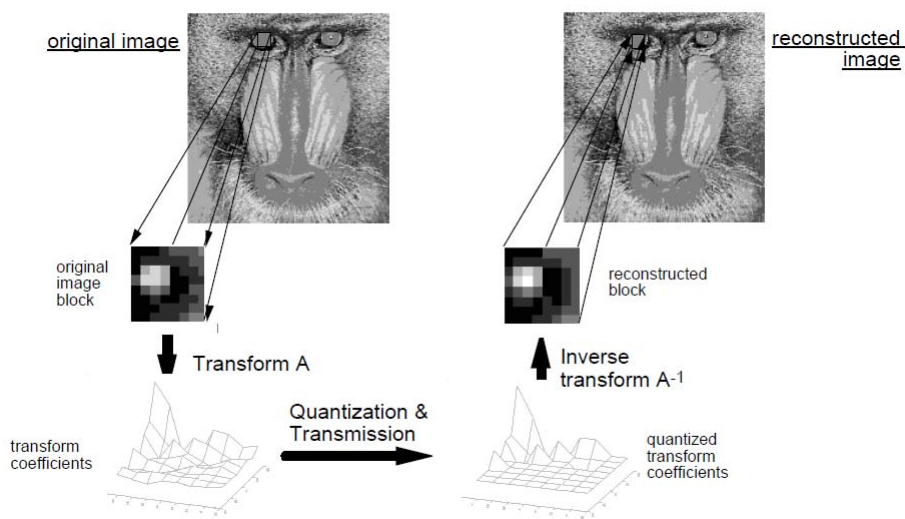
[66]

Introduction

- To compact most of the information into a few elements
- The weight-height example:

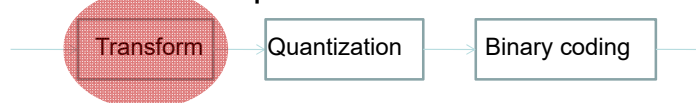


Another example



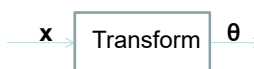
Transform Coding

- Three steps
 - Divide a data sequence into blocks of size N and transform each block using a reversible mapping
 - Quantize the transformed sequence
 - Encode the quantized values



[69]

The Transform



- Forward transform

$$\theta_n = \sum_{i=0}^{N-1} x_i a_{n,i} \quad \text{or} \quad \theta = \mathbf{A}\mathbf{x} \quad [\mathbf{A}]_{i,j} = a_{i,j}$$

θ is an N -dimensional vector, \mathbf{A} is an $N \times N$ matrix, and \mathbf{x} is an N -dimensional vector.

- Inverse transform

$$x_n = \sum_{i=0}^{N-1} \theta_i b_{n,i} \quad \text{or} \quad \mathbf{x} = \mathbf{B}\theta \quad [\mathbf{B}]_{i,j} = b_{i,j}$$

- Orthonormal transforms

$$-\mathbf{A}^{-1} = \mathbf{A}^T$$

$$\begin{bmatrix} x_0 \\ x_1 \end{bmatrix} = \begin{bmatrix} b_{00} & b_{10} \\ b_{01} & b_{11} \end{bmatrix} \begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix} = \theta_0 \begin{bmatrix} b_{00} \\ b_{01} \end{bmatrix} + \theta_1 \begin{bmatrix} b_{10} \\ b_{11} \end{bmatrix}$$

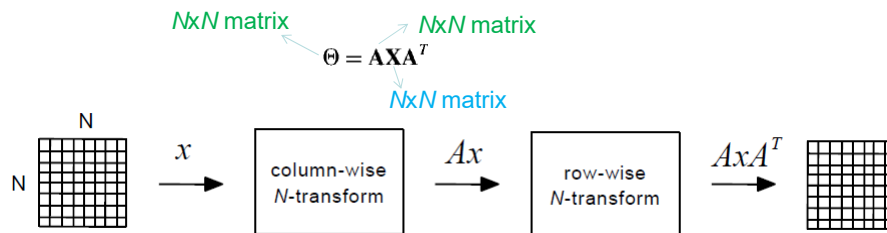
- Linearity: \mathbf{x} is represented as a linear combination of “basis vectors”
- Parseval’s Theorem holds: the total energy is preserved

[70]

Two-dimensional transform

- Separable transforms
 - Take the transform along one dimension, then repeat the operation along the other direction
 - Ex: rows \rightarrow columns, or columns \rightarrow rows

- Forward transform



- Backward transform

$$X = B\Theta B^T$$

[71]

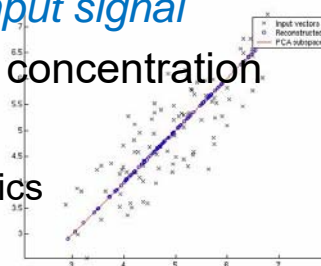
Transforms of interest

- Data-dependent
 - Discrete Karhunen Løve transform (KLT)
- Data-independent
 - Discrete cosine transform (DCT)

[72]

KLT

- Also known as *Principal Component Analysis* (PCA), or the *Hotelling transform*
- Transforms *correlated* variables into *uncorrelated* variables
- Basis vectors are *eigenvectors of the covariance matrix of the input signal*
- Achieves *optimum* energy concentration
- Disadvantages
 - Dependent on signal statistics
 - Not separable

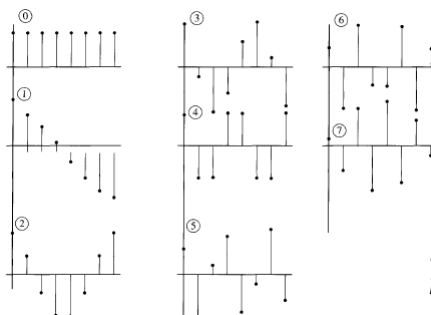


[73]

DCT (1)

- Part of many standards (JPEG, MPEG, H.261, ...)
- $$\text{Th}[c]_{i,j} = \begin{cases} \sqrt{\frac{1}{N}} \cos \frac{(2j+1)i\pi}{2N} & i = 0, j = 0, 1, \dots, N-1 \\ \sqrt{\frac{2}{N}} \cos \frac{(2j+1)i\pi}{2N} & i = 1, 2, \dots, N-1, j = 0, 1, \dots, N-1 \end{cases}$$

- Visualiz

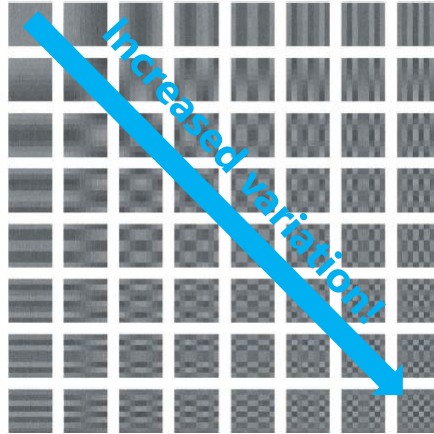


The frequency increases as we go from top to bottom!

[74]

DCT (2)

- 2-D basis matrices of DCT

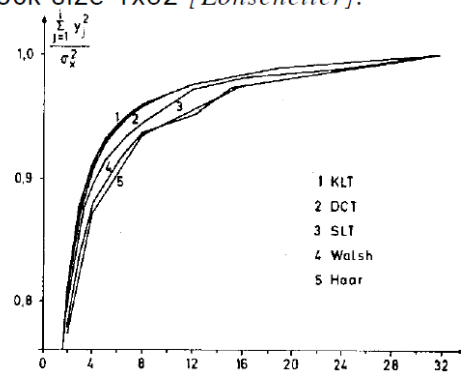


[75]

DCT (3)

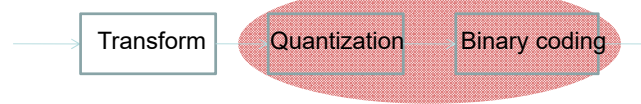
- Performs close to the optimum KLT in terms of compaction

Energy concentration measured for typical natural images, block size 1x32 [Lohscheller]:

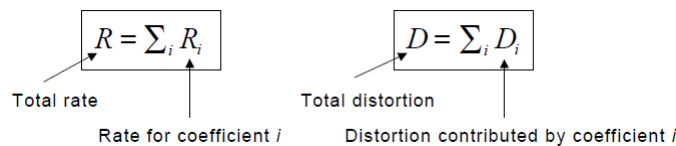


[76]

Quantization and coding of transform coefficients (1)



- The bit-rate allocation problem
 - Divide bit-rate R among transform coefficients such that resulting distortion D is minimized



[77]

Quantization and coding (2)

- The optimal bit allocation

$$R_k = R + \frac{1}{2} \log_2 \frac{\sigma_{\theta_k}^2}{\prod_{k=1}^M (\sigma_{\theta_k}^2)^{\frac{1}{M}}}$$

variance of the transform coefficient θ_k

- The zonal sampling
 - Compute $\sigma_{\theta_k}^2$ for each coefficient
 - Set $R_k = 0$ for all k
 - Sort $\{\sigma_{\theta_k}^2\}$. Suppose $\sigma_{\theta_1}^2$ is the
 - Increment R_1 by 1, divide $\sigma_{\theta_1}^2$
 - Decrement R_b by 1.

- If $R_b=0$, stop;
- Otherwise, go to step 3

[78]

$R_b = 72$ bits

8	7	5	3	1	1	0	0
7	5	3	2	1	0	0	0
4	3	2	1	1	0	0	0
3	3	2	1	1	0	0	0
2	1	1	1	0	0	0	0
1	1	0	0	0	0	0	0
1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0

8x8 DCT of an image

Discarded!

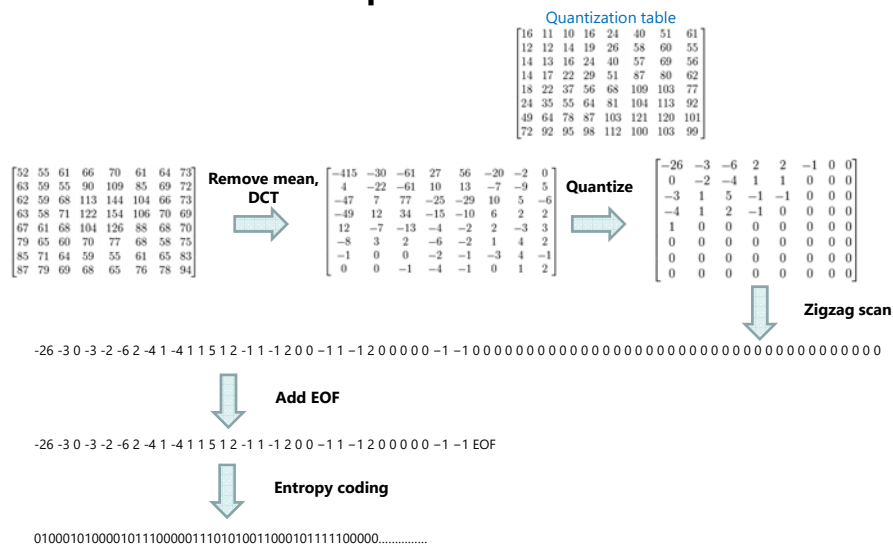
Quantization and coding (3)

- The threshold coding
 - Transform coefficients that fall below a threshold are discarded
 - Example: a 8x8 block



[79]

Example: Encode



[80]

Example: Decode

010001010000101110000011101010011000101111100000.....



Entropy decoding

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



Put into a block

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

Q^{-1}

-416	-33	-60	32	48	-40	0	0
0	-24	-56	19	26	0	0	0
-42	13	80	-24	-40	0	0	0
-56	17	44	-29	0	0	0	0
18	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

Inverse DCT,
Add mean

Original block

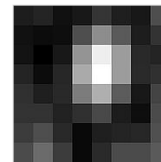
52	55	61	66	70	61	64	73
63	59	55	90	109	85	69	72
62	59	68	113	144	104	66	73
63	58	71	122	154	106	70	69
67	61	68	104	126	88	68	70
79	65	60	70	77	68	58	75
85	71	64	59	55	61	65	83
87	79	69	68	65	76	78	94

Reconstructed block

60	63	55	58	70	61	58	80
58	56	56	83	108	88	63	71
60	52	62	113	150	116	70	67
66	56	68	122	156	116	69	72
69	62	65	100	120	86	59	76
68	68	61	68	78	60	53	78
74	82	67	54	63	64	65	83
83	96	77	56	70	83	83	89

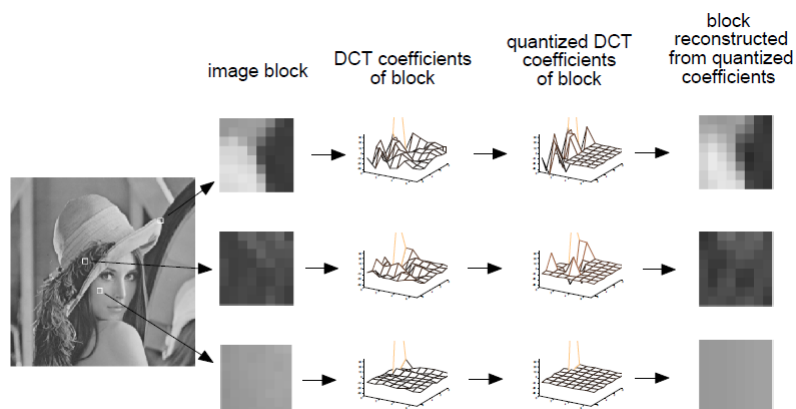
Quantization table

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



[81]

One more example



[82]

Introduction to Video compression algorithm

[83]

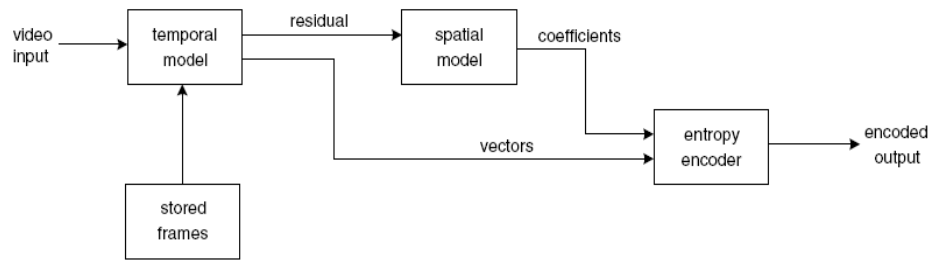
Video compression algorithm



- **Temporal compression:**
 - continuous frames look similar, only the changes from one frame to the next need to be encoded.
- **Spatial compression:**
 - Sensitivity of the human visual system.

[84]

Video compression algorithm



Video encoder block diagram

[85]

Temporal compression Block-matching algorithm

Previous Image



Measurement window is compared with a shifted array of pixels in the other image, to determine the best match

Current Image

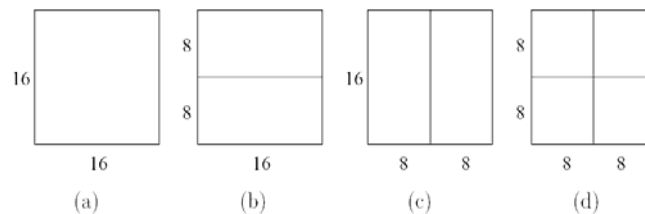


Rectangular array of pixels is selected as a measurement window

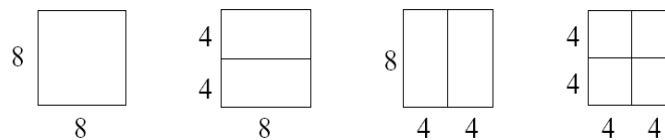
[86]

H.264 Motion Compensation

- First Stage.



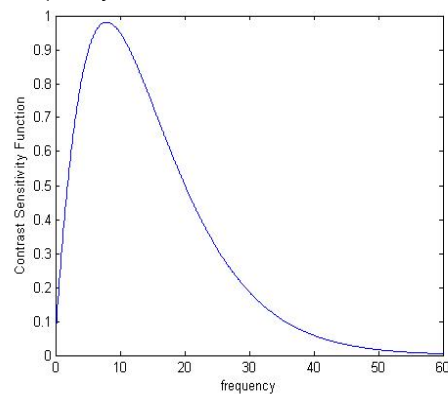
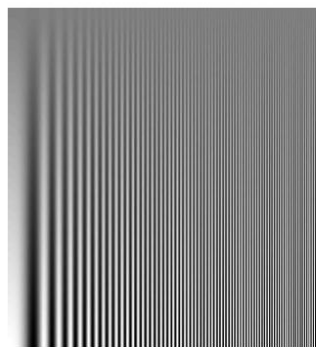
- Second Stage: If (d) has the lowest cost in the first stage



[87]

Spatial compression

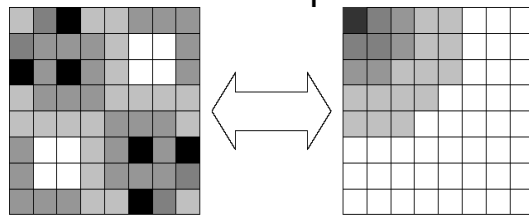
Sensitivity of the Human Visual System to Contrast Changes, as a Function of Frequency.



- The Human Visual System (HVS) is less sensitive to errors in low and high frequency.
- High and low frequencies may be quantized more coarsely

[88]

Spatial compression 8x8 DCT operation



(a) Original picture.

(b) Corresponding DCT mapping of (a)

- The Discrete Cosine Transform (DCT) :
 - decomposes the signal into spatial frequencies, which then allow further processing techniques to reduce the precision of the DCT coefficients consistent with the Human Visual System (HVS) model.
 - concentrates most energy in a few bins (e.g., lots of coefficients near zero).
- Larger blocks → more computation
- Smaller blocks → transform stage less effective

[89]

Optimal Tiling for Video Compression

- Key ideas
 - Use larger dictionaries of tilings than in H.264.
 - Use recursive tree pruning algorithms to efficiently find the best rectangular tiling for an image.
 - Apply such algorithms to two stages of video compression.
 - Motion Compensation.
 - Transform.

[90]

- Achievements

- Algorithm I:

- Use larger dictionary for motion compensation.
 - Our tiling selection method results in up to 19% savings in bit rate as compared to H.264 tiling selection.

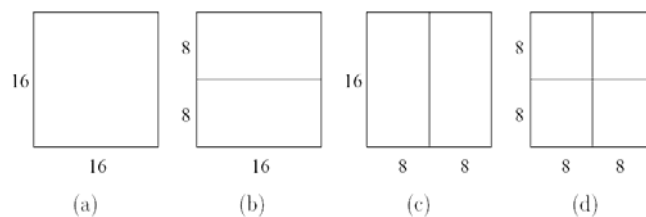
- Algorithm II:

- Use larger dictionary for both motion compensation and transform stages.
 - Our tiling selection method results in up to 23% savings in bit rate as compared to H.264 tiling selection.

[91]

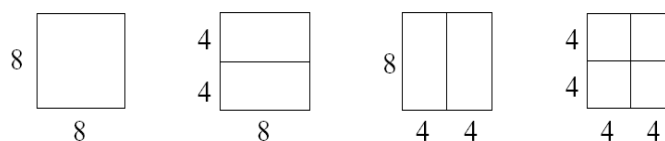
H.264 Motion Compensation

- First Stage.



	H.264
# tilings	259
# tiles	41

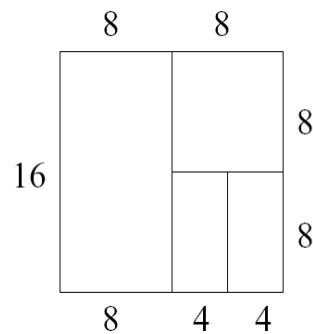
- Second Stage: If (d) has the lowest cost in the first stage



[92]

Dyadic Tilings

- In addition, allow all dyadic two-way splits down to 4×4 tiles. For example,

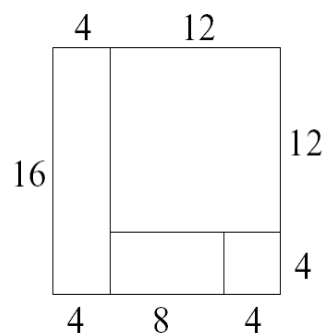


	Dyadic
# tilings	6857
# tiles	49

[93]

Multitree Tilings

- In addition, allow all two-way splits down to 4×4 tiles. For example,



	Multitree
# tilings	68480
# tiles	100

[94]

Number of Tilings and Tiles for a 16X16 Block

	H.264	Dyadic	Multitree
# tilings	259	6857	68480
# tiles	41	49	100

- The H.264 set of tilings is a small subset of the dyadic and multitree tilings.
- The number of tiles, and therefore complexity, increases much more slowly.

[95]

Examples--Carphone Sequence



Original

H.264

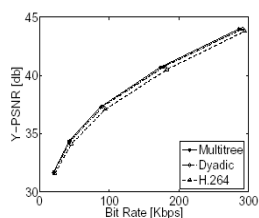
Dyadic

Multitree

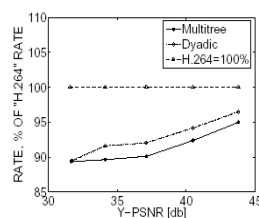
[96]

Results: Fixed 4x4 Transform

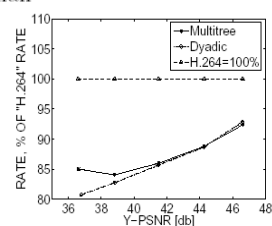
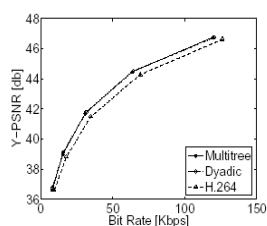
RD Curves



RD Curves in % of H.264



Foreman



Salesman

[97]

Conclusion

- The use of finer tilings result in more overhead bits.
- We show, however, that our optimal tiling search is so effective that the increase in the number of overhead bits is more than compensated for by the reduction in the number of bits require to encode the motion vectors and transform coefficients.

[98]

Some multimedia research topics

[99]

Here comes the opportunity!



Prizes

Thanks to generous support from our sponsors, Google, HP Labs, Nokia, 3Dlife, CeWe, and Yahoo!, the Multimedia Grand Challenge is happy to offer a first prize of \$1500.



- \$1500 (~NT47688) if we earn the first prize
- Make your resume stand out

[100]

<div style="text-align: center;">  </div> <p>.....</p> <p>.....</p> <p>.....</p> <p>Education</p> <p>master, NTUST</p> <p>.....</p> <p>.....</p> <p>Publication</p> <p>xxx, "An fast approach for automatic photo organization", <i>ACM Multimedia</i>, 2010.</p> <p>Experiences</p> <p>.....</p> <p>.....</p> <p>.....</p>	<div style="text-align: center;">  </div> <p>.....</p> <p>.....</p> <p>.....</p> <p>Education</p> <p>master, NXU</p> <p>.....</p> <p>.....</p> <p>Experiences</p> <p>.....</p> <p>.....</p> <p>.....</p>
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[101]

10 Challenges

Nokia Challenge: Photo Location & Orientation	Radvision Challenge: VideoConf Experience	Google Challenge: Video Genre Classification	Radvision Challenge: Content Adaptation	Google Challenge: Personal Diaries
HP Challenge: Visual Communication	Yahoo! Challenge: Video Segmentation	Yahoo! Challenge: Novel Image Understanding	3DLife Challenge: Sport Camera Networks	CeWe Challenge: Photo Set Theme Identification

- Identified by leading companies!
 - two from Google
 - two from Yahoo!
 - two from Radvision
 - one from Nokia, HP, 3DLife, and CeWe

[102]

Can you tell where are these photos taken?




[103]

Nokia Challenge

- Goal
 - Try to derive exact camera poses (**location** and **orientation**) of given photos that are lacking location annotation
- You can assume the availability of nearby photos/video with known location that can be used to derive unknown camera poses; other ideas that do not require existing content will be welcome.
- Applications
 - <http://betalabs.nokia.com/apps/nokia-image-space>

[104]



<geolocation alt="0" lat="43.731639899896" lng="7.421372230007" />
<orientation pitch="0" roll="0" yaw="104" />

緯度, 經度

<geolocation alt="0" lat="43.731634200204" lng="7.421395112594" />
<orientation pitch="0" roll="0" yaw="100" />

<geolocation alt="0" lat="43.731594553817" lng="7.421487732589" />
<orientation pitch="0" roll="0" yaw="276" />

<geolocation alt="0" lat="43.731658423895" lng="7.421072325625" />
<orientation pitch="7.5" roll="0" yaw="240" />

[105]

Nokia Challenge

- Matlab tools and datasets are available!
 - photos captured with the Nokia 6210 Navigator
 - each image has an associated GPS/orientation measurement



```
<?xml version="1.0" encoding="utf-8"?>
<imageset>
  <image height="600" id="0" src="demoset/image0.jpg" width="800">
    <geolocation alt="0" lat="43.731777279237" lng="7.42103980385" />
    <orientation pitch="0" roll="0" yaw="41" />
  </image>

  <image height="600" id="1" src="demoset/image1.jpg" width="800">
    <geolocation alt="0" lat="43.731745092741" lng="7.42101088629" />
    <orientation pitch="15" roll="0" yaw="102" />
  </image>

  <image height="600" id="2" src="demoset/image2.jpg" width="800">
    <geolocation alt="0" lat="43.73173595647" lng="7.421045168269" />
    <orientation pitch="7.5" roll="0" yaw="111" />
  </image>
  .
  .
  .
```

[106]

Google Challenge: Indexing and Fast Interactive Searching in *Personal Diaries* (個人日記)

[107]

Google Challenge

- Diaries can be ***any combination*** of audio, video, geographic location, photos, phone logs, and whatever other multimedia data the user generates or accesses.
- Goal
 - develop good schema, algorithms, UI, etc., that will be useful for diaries from audio-only through full-featured multimedia.

[108]



Google Challenge

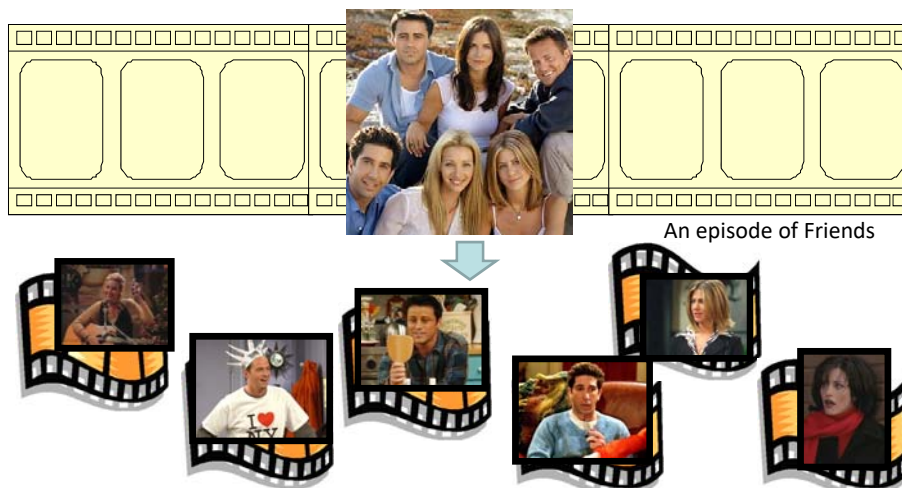
- Some thoughts:
 - What would be a good design for blogs?
 - We now have photos, comments, game records, ... on facebook. Could we design new tools for users to create their own “e-diary”?

Yahoo! Challenge: Robust Automatic Segmentation of Video According to Narrative Themes

故事形式的

[111]

One example:



[112]

Yahoo! Challenge

- Goal
 - develop methods, techniques, and algorithms to automatically generate **narrative themes** for a given video, as well as present the content in an easy-to-consume manner to end-users in a search engine experience
- Applications
 - allow users consume pieces of a video that would be of interest to them
 - let users kill time during lunch breaks in creative ways

[113]

Another example:



2010/03/30 中天新聞



政治・財經



社會・地方



影視・體育



生活・醫療

1. Train a 馬英九總統 detector, use it to identify video segments, and tag them with “政治”
2. Identify the sound of the anchor man (主播), from there identify keywords such as “娛樂”, “社會”, etc.
3.

[114]

One more example:



2010/03/28 La New vs. 兄弟 @ 新莊



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告

1. Background music is different, tag the video segment “廣告”
2.

[115]

- My suggestions

- select a certain type of videos

- Examples:

- sitcom or some popular tv show such as 康熙來了
- your favorite movie
- a wedding video of your family (進場, 致詞, 敬酒, 玩遊戲...)
- sport videos
- educational videos
- ...

- You need to specify not only the themes, but also how your are going to segment the input video into themes *automatically*

[116]

Yahoo! Challenge: Novel Image Understanding

[117]

Yahoo! Challenge

- Move beyond simple image classification
- Goal
 - develop novel and useful ways to organize and structure image content

[118]

Yahoo! Challenge

- Example

- Sort celebrity pictures by their subject's age or hair style



Yahoo! Challenge

- Example

- discover how a logo/advertisement/product has evolved over time



[120]

Yahoo! Challenge

- There are many ways to organize photos!
 - What are the ways that are not obvious?
 - What can we do better than we can do today?

[121]

HP Challenge: High Impact Visual Communication

[122]

HP Challenge


- Goal
 - find a solution which can create **a collage** (美術拼貼) and generate **a textual description** that tells the story of a set of photos
- How do we create a high impact picture that can convey information across cultural boundaries and find a thousand words that best describe such a picture?

[123]


HP Challenge

- Input
 - a digital photo collection, such as photos taken during a vacation
- Outputs
 - the most appealing collage picture (1600x1200 pixels) that best represents the original collection
 - a description (<100 words) of the collage picture

[124]



A show at xxx pub. Hang out with my friends....



my favorite cartoon: Sponge bob and square pants....

[125]

HP Challenge

- 6 datasets are available!
 - each with 20 photos

[126]

CeWe Challenge: Automatic Theme Identification of Photo Sets for Digital Print Products

[127]

CeWe Challenge

- About CeWe
 - the Number One services partner for first-class trade brands on the European photographic market
 - supplies both stores and Internet retailers with photographic products

posters, calendars or ***photo books***

[128]

CeWe Challenge

- Goal
 - simplify the “style selection” step
- In the CEWE PHOTOBOOK software about 100 styles are available to suit different user tastes and different types of photo books.
 - events (party, holiday, BBQ...)
 - seasons (Christmas, summer, Easter, new year...)
 - design styles (classical, funky, cute, ...)

[129]

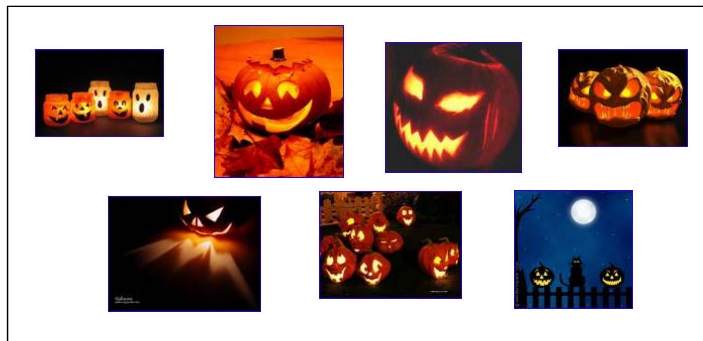
CeWe Challenge

- Based on the photo contents
 - not necessarily to automatically determine the one and only perfect style,
 - but rather to provide the user with a reasonable selection of styles he or she can choose from.

[130]

CeWe Challenge

- Example
 - What would be good background colors for the set of photos?



[131]

Radvision Challenge: Video Conferencing To Surpass “In-Person” Meeting Experience

[132]

Radvision Challenge

- Goal
 - developing new technologies and ideas to surpass (超越, 改善) the “in-person” meeting experience

[133]

Radvision Challenge

- Example:
 - Come up with ways to maintain a long-term relationship



[134]

For more information...



<http://www.acmmm.org/2017/>

[135]