

Techniques for Video Compression and Analysis (5LSE0), Module 01 - A

Introduction to Coding and Information Measures

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slides version 1.0

Lecture Information

- * Lecturers
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- * Information: <http://vca.ele.tue.nl/courses/5LSE0> (version 2022-23)
 - Manuscript
 - Lecture slides
 - Software Practical Exercises
- * Examination: Oral Examn (70%) + Practical exercises (30%)

Purpose of the course 5LSE0

- * Learn the power of signal transformations
 - For video compression, like as in transform coding
 - For video analysis, like as in the Gabor wavelet transform
 - For combining them with Deep Learning
- * Exploit the concept of “condensing information”
 - This is useful for coding and compression
 - But also applicable to video analysis: what is the key informative feature
 - Use specific transformation for that, like the PCA,
- * Therefore, this course is divided in two parts
 - A. Video compression using Discrete Cosine Transformation
 - Learn the concept of entropy as information measure both for compression and video analysis
 - B. Analysis techniques, e.g. wavelet transform, for finding feature
 - Expand the tool set of computer vision and for Deep Learning

Mod 01 - A / Part 1: Introduction / Motivation for Compression

What is Digital Signal Coding?

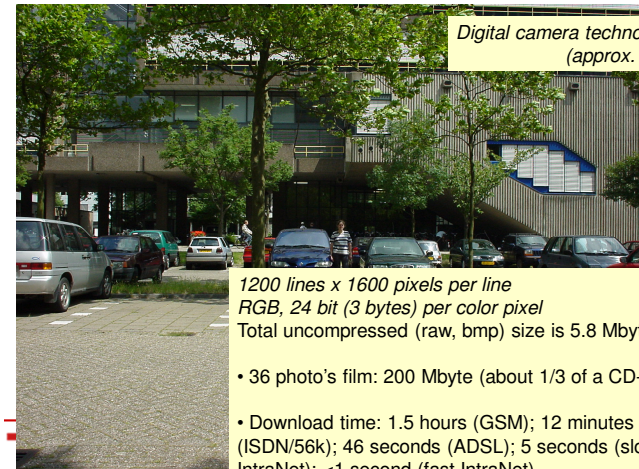
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- * Digital signals: **audio, image(s), video (multimedia)**
 - Will not discuss *document* compression
 - Focus on image compression (video and images are signals)
- * Coding/Compression = **compact representation**:
 - Less space required on storage media (hard disks, CD, DVD, BD, embedded memory)
 - Smaller transmission bandwidth required (Internet, terrestrial broadcast, satellite links, mobile phones)
 - Allows for faster uploading/downloading of multimedia files (Internet, authoring environment)

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Why Digital Signal Coding? – (1)

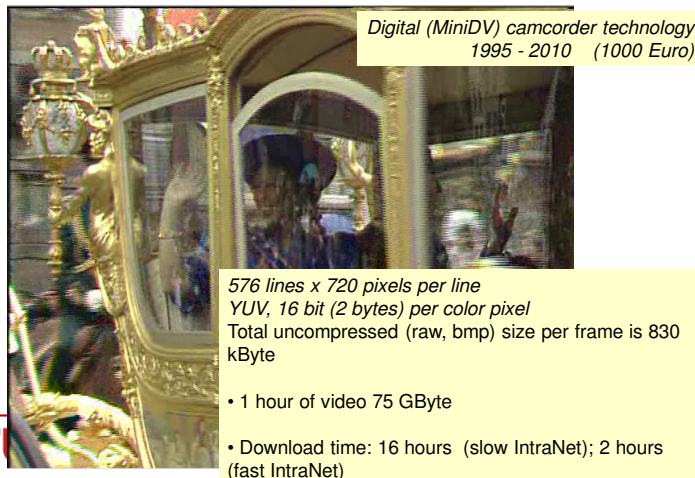
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Why Digital Signal Coding? – (2)

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Usage of Digital Signal Coding

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- * Coding/Compression => **compact representation**:
 - Less space required on storage media
 - Smaller transmission bandwidth required
 - Allows for faster uploading/downloading of multimedia files / communication
- * **Examples of “products”/systems using compression**:
 - Digital audio: MP3
 - Digital image and video cameras: JPEG, MPEG, MPEG-4 AVC
 - Internet movie downloading/streaming DivX, Qtime, Real players
 - Image compression for mobile phones: Whatsapp
 - Digital Cinema: JPEG-2000

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Objectives of this course

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*** Objectives of this course:**

- Fundamental video/image compression, basics Information Theory
- Key Techniques by which compression can be achieved
- What is inside the MPEG standard, DVB, DV, JPEG, etc.
- Introduce new signal transformations: DCT (compression, Wavelet transform (compression and analysis), PCA (analysis)
- Introduce metrics that benefit video analysis
- Discuss video analysis cases using signal transforms
- Show video analysis where models and metrics are used
- Practical experiments with compression/quality and transform video analysis, software package

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Compression is now an embedded system

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Signal to be coded $x(t), x(t,s)$

Pre-process, Transform Coding

* "Coding" includes Compression, Security, and Error Control techniques

* This course focuses on Compression and Transform Analysis only

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Lossless vs. Lossy Compression

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Lossy or Non-reversible Compression (if identical: reversible)
 $y(i,j) \neq x(i,j)$

Examples: JPEG, MPEG, MP3, DivX
Can trade compression for amount of difference (quality)

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Some Ballpark Figures

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*** JPEG Images**

- » Digital camera: 1600x1200 pixels at 24 bits/color pixel ==> 46 Mbit/file = 5.8 MByte/file
- » "Good quality" JPEG: 600 kByte
 - » Bit rate: 2.5 bit / color pixel (bpp), Compr. Factor: 9.6

*** MPEG Video**

- » Digital Broadcast TV: 720x576 at 16 bit/color pixel at 25 pictures/second = 166 Mbit/sec = 20 MByte/sec
- » "Commercial tv station" MPEG: 4.5 MBit/sec
 - » Bit rate: 0.43 bit / pixel (bpp), Compr factor: 36.8
- » Full HD video: 1920 pixels x 1080 lines at 16 bit/color pixels, 30Hz frames/s (fps), compressed to 8-16 Mbit/s

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Compression is not "for free" – (1)

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Compression is not "for free" – (2)

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0.6 bit per pixel (Cf=13)

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Compression is not "for free" – (3)

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0.3 bit per pixel (Cf=26)

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Compression is not "for free" – (4)

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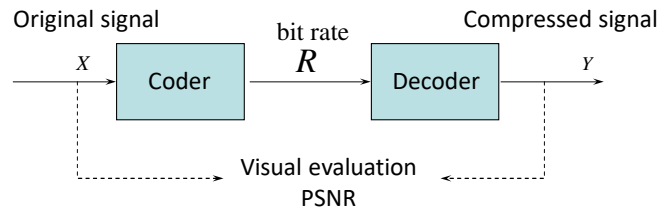
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0.2 bit per pixel (Cf=40)

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Quality of Compressed Signals – (1)

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- * Visual evaluation (always important)
- * Numerical evaluation: PSNR



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Quality of Compressed Signals – (2)

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Number Score	Quality Scale	Impairment Scale
5	Excellent	Imperceptible
4	Good	(Just) perceptible but not annoying
3	Fair	(Perceptible and) slightly annoying
2	Poor	Annoying (but not objectionable)
1	Unsatisfactory (Bad)	Very annoying (Objectionable)

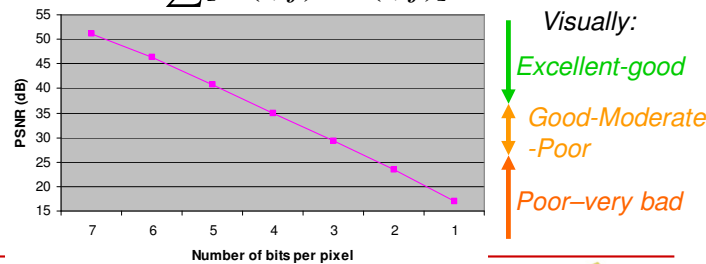
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Quality of Compressed Signals

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PSNR = Peak Signal - to - Noise Ratio =

$$10 \log_{10} \frac{255^2}{\sum [X(i, j) - Y(i, j)]^2} \quad (\text{dB} = \text{dB})$$



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Obvious Compression of Video/Images

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Two obvious ways to compress signals

1. Reduce the sampling rate (subsampling)
 - Fewer audio samples per second
 - Fewer pixels per picture line and fewer lines (images)
 - Fewer pictures per second (video)
2. Reduce the No. of amplitude levels per signal sample:
 - Pulse coded modulation or PCM
 - Obviously we will develop far more advanced techniques in this course

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Case 2: Reduce the #Amplitudes – (1) ²¹



R(ed)G(reen)B(lue) samples, 24 bits (16,777,200 diff. colors)

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Reduce the #Amplitudes – (4) ²²



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But real coding is always better...! ²³



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Mod 01 - A Part 2: Introduction / Relation to statistics of images

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Questions to be Answered 25

Three questions will play a central role:

1. **Why can signals be compressed?**
2. How much can signals be compressed?
3. Which signal processing/information theory algorithms are most efficient in reaching maximum compression?

Answer: Because signal amplitudes are statistically redundant

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Why can Signals be Compressed? – (1) 26

Example: Signal amplitudes being statistically redundant

Example: What is required no. bits/sample for transmission?

- Signal has integer amplitudes in range [0,7]
- 3 bit per signal sample (3 bps or bpp)

Signal	Code
0	000
1	001
2	010
3	011
4	100
5	101
6	110
7	111

But statistics reveal the redundancy....

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Why can Signals be Compressed? – (2) 27

Because signal amplitudes are statistically redundant

Entropy Coding or Variable Length Coding

Example:

- Signal has integer amplitudes in range [0,7]
- **2 bits per signal sample**

Signal value	Probability	Codeword
0	0.125	100
1	0	
2	0.5	0
3	0	
4	0.125	101
5	0.125	110
6	0	
7	0.125	111

Question: What is the **shortest** average codelength?

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Why can Signals be Compressed? – (3) 28

Because infinite accuracy of signal amplitudes is (perceptually) irrelevant : example color/detail irrelevancy

24 bit (16777200 different colors)

8 bit (256 different colors)
Compression factor 3

The difference between these two pictures is **perceptually irrelevant** because our visual system is insensitive for such small differences in color and fine details. But the files differ a factor of 3 in size!

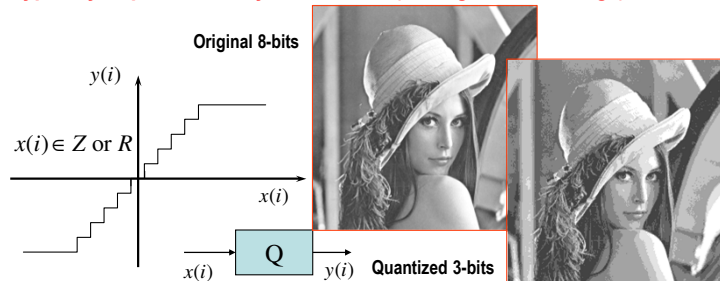
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Why can Signals be Compressed? – (4) ²⁹

Because infinite accuracy of signal amplitudes is (perceptually) irrelevant

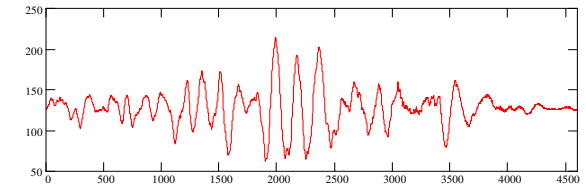
Typically implemented by a Quantizer (intelligent “rounding”)



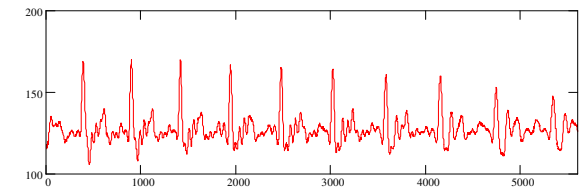
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Why can Signals be Compressed? – (5) ³⁰

Because signal amplitudes are mutually dependent



Signals are stochastic, change over time!



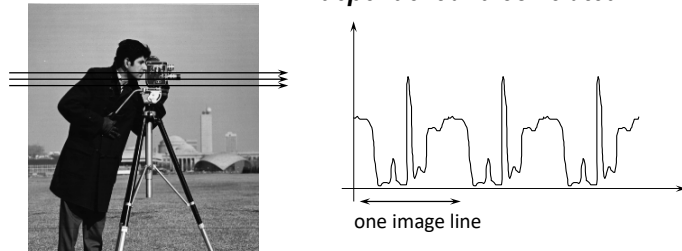
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Why can Signals be Compressed? – (6) ³¹

Because signal amplitudes are mutually dependent

Example of video signal: change over image, but locally dependent and correlated!



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Why can Signals be Compressed? – (7) ³²

Because signal amplitudes are mutually dependent
Dependency is commonly expressed by the **autocorrelation function**:

$$R_X(k) = E[X(n)X(n-k)]$$

$$R_X(k) = \frac{1}{N} \sum_{n=0}^N X(n)X(n-k)$$

For images, we have a 2D correlation

$$R_X(k, l) = E[X(n, m)X(n-k, m-l)]$$

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Summarizing ...

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