



**Electronics and
Computer Science**
University of Southampton



CM214-COMP2008

Data Communications and Networks

Lossy Data Compression

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Objectives



- To understand how “lossy” data compression works
 - Where it is appropriate for use
 - Where it should not be used
- (Peterson & Davie, Section 7.2)



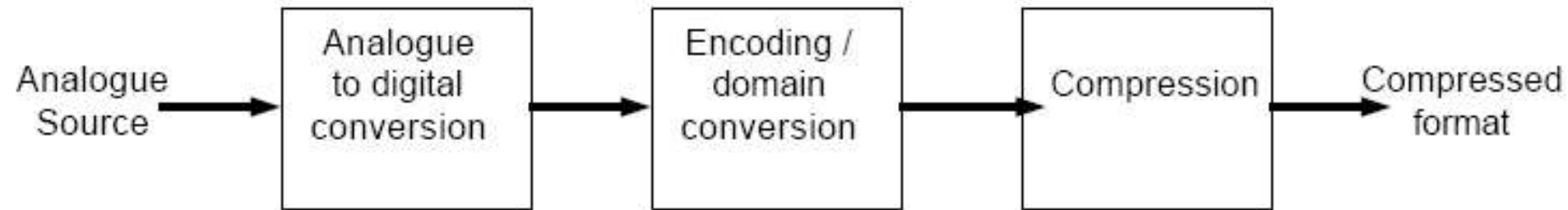
Lossy Compression



- Is only appropriate for data that is, in some way “analogue”
 - Not a general purpose compression algorithm like Lempel-Ziv
- Takes advantage of imperfections in human perception
 - Removes information we don’t notice



Analogue Compression



- We will look at A/D conversion process
- And MP3 / JPEG / MPEG encoding & compression



A/D Conversion



1. Sampling

- Take “snapshot” of continuously varying signal at discrete times

2. Quantisation

- Convert snapshot to a discrete value

3. Coding

- Convert value into *b*-bit number

How often? How many bits?



Nyquist Sampling Theorem



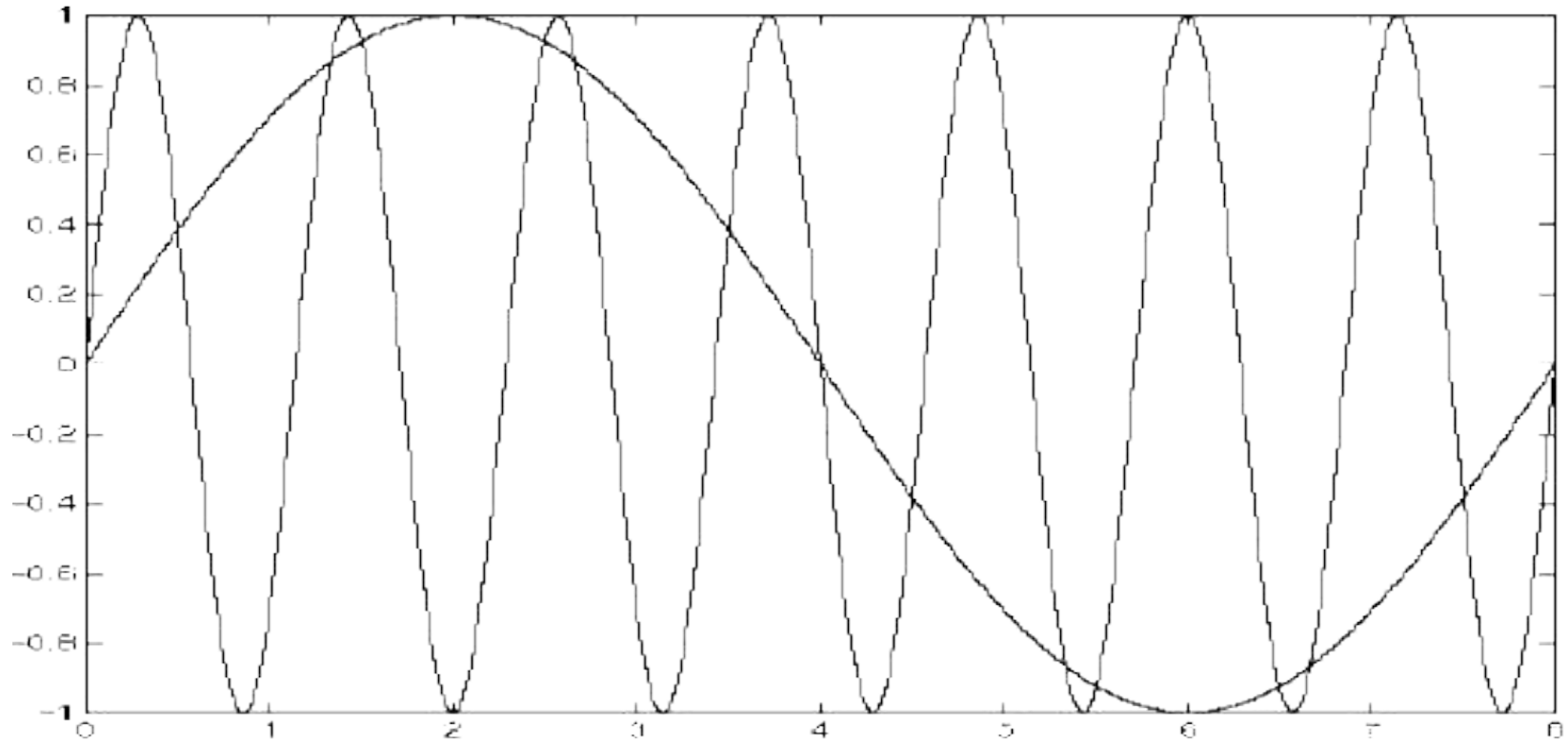
“The minimum sampling rate F_s needed to reconstruct an analogue signal from a sampled signal is the Nyquist rate,

$$2F_{\max}”$$

i.e. You must sample twice as fast as the most rapidly varying signal



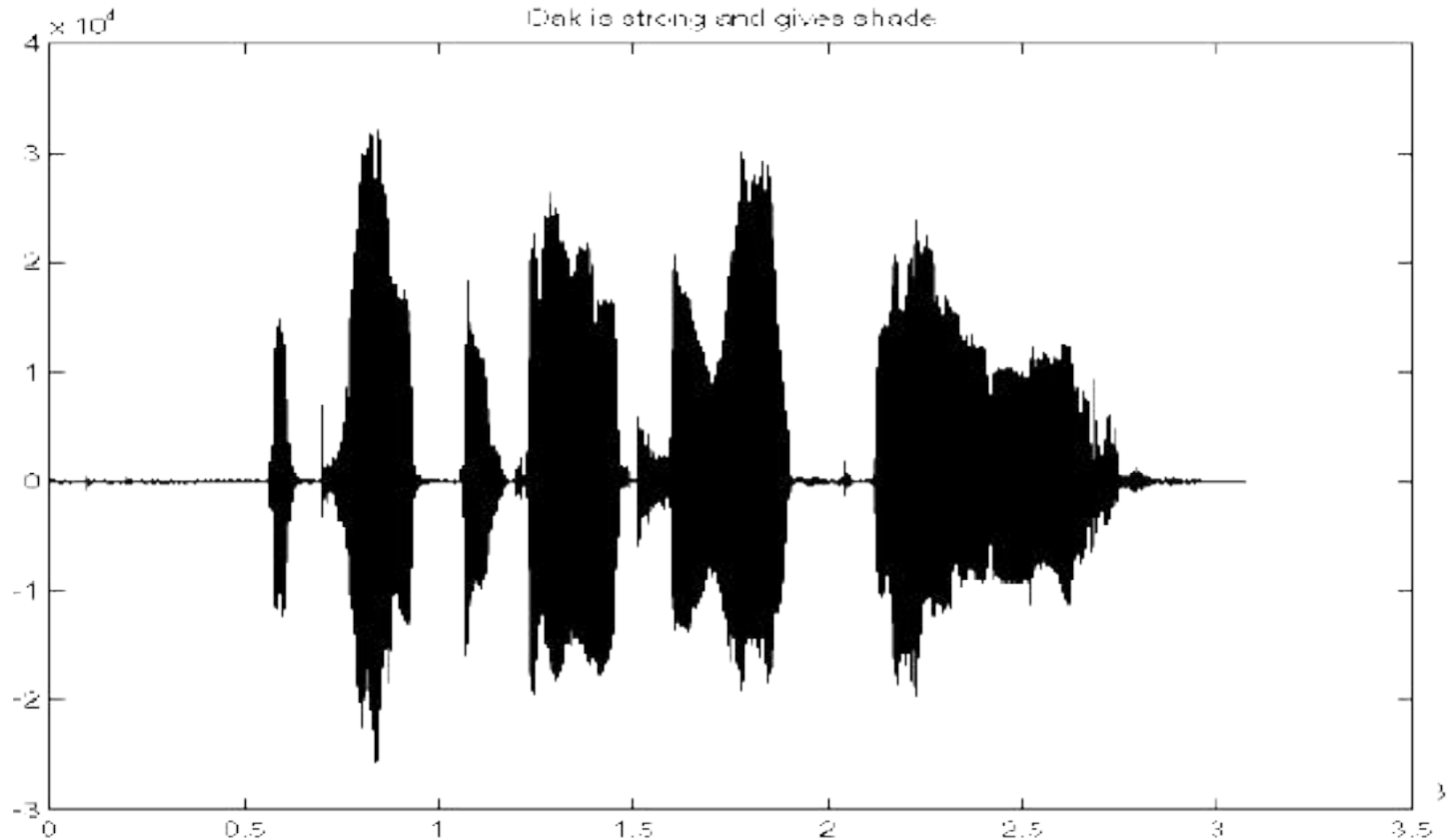
Undersampling



Sample taken every second (1Hz) of signal varying at $7/8\text{Hz}$ appears as $1/8\text{Hz}$ due to undersampling or “aliasing”



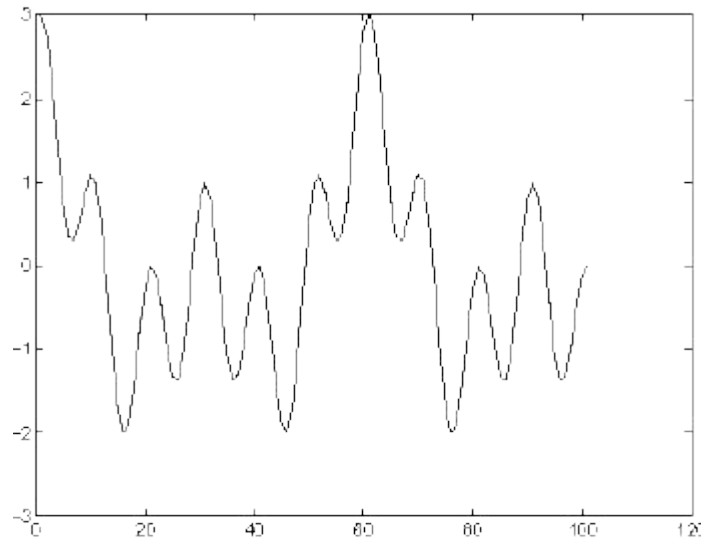
Real Speech



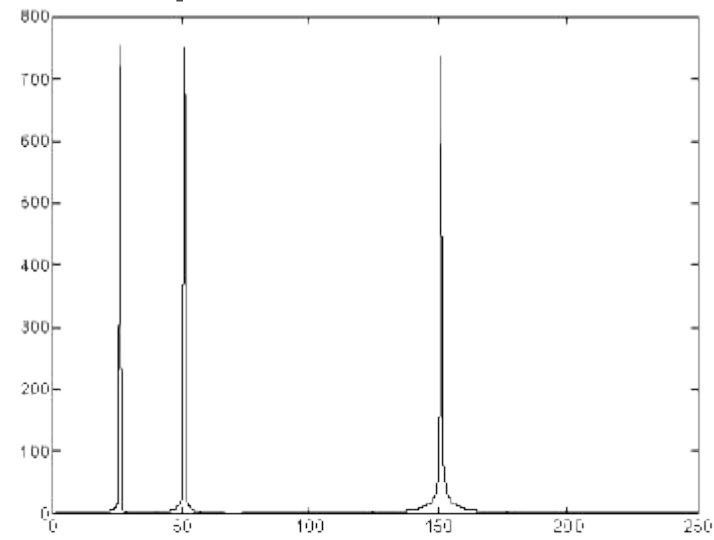


Domain Conversion - 1

- For compression we convert from time domain to frequency domain
 - From set of time varying signals to set of sine waves of various frequencies



Time Domain



Frequency Domain



Domain Conversion – 2



- Conversion assumes signals look like lots of superimposed sine waves
 - All “natural” signals do!
 - (corollary –only works for natural signals)
- From Nyquist, we need to sample at twice the rate of the highest frequency sine wave



Fourier Transform



- Time to frequency domain conversion is achieved using the Fourier Transform
 - If signal is: Discrete Fourier Transform is:

$$x_n = \frac{1}{N} \sum_{k=0}^{N-1} X_k e^{-j2\pi kn/N} \quad X_k = \sum_{n=0}^{N-1} x_n e^{j2\pi kn/N}$$

- In reality, use FFT algorithm from library routine or hardware implementation (e.g. DSPs in cameras & phones)



Why Frequency Domain?



- Because easier to identify features humans cannot perceive
 - E.g. For MP3 audio we can remove
 - All frequencies outside human range
 - “time domain” masked sounds (quiet sound at same time as a loud one)
 - “frequency domain” masked sounds
 - Based on **empirical** perception tests of humans **NOT** on mathematical algorithms



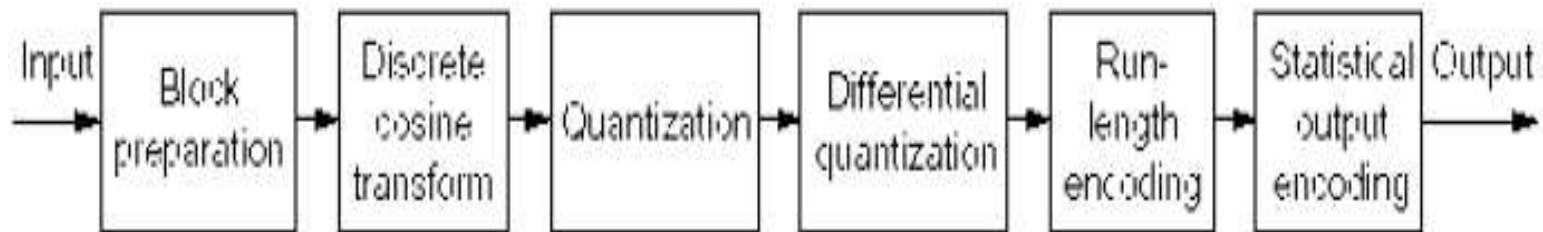
JPEG Compression



- Similar to MP3, uses empirical & physiological knowledge to remove information that humans cannot perceive
 - More conventional techniques then compress resulting data
- Can control amount of data “loss”
- Too much results in no tonal variation



JPEG Process



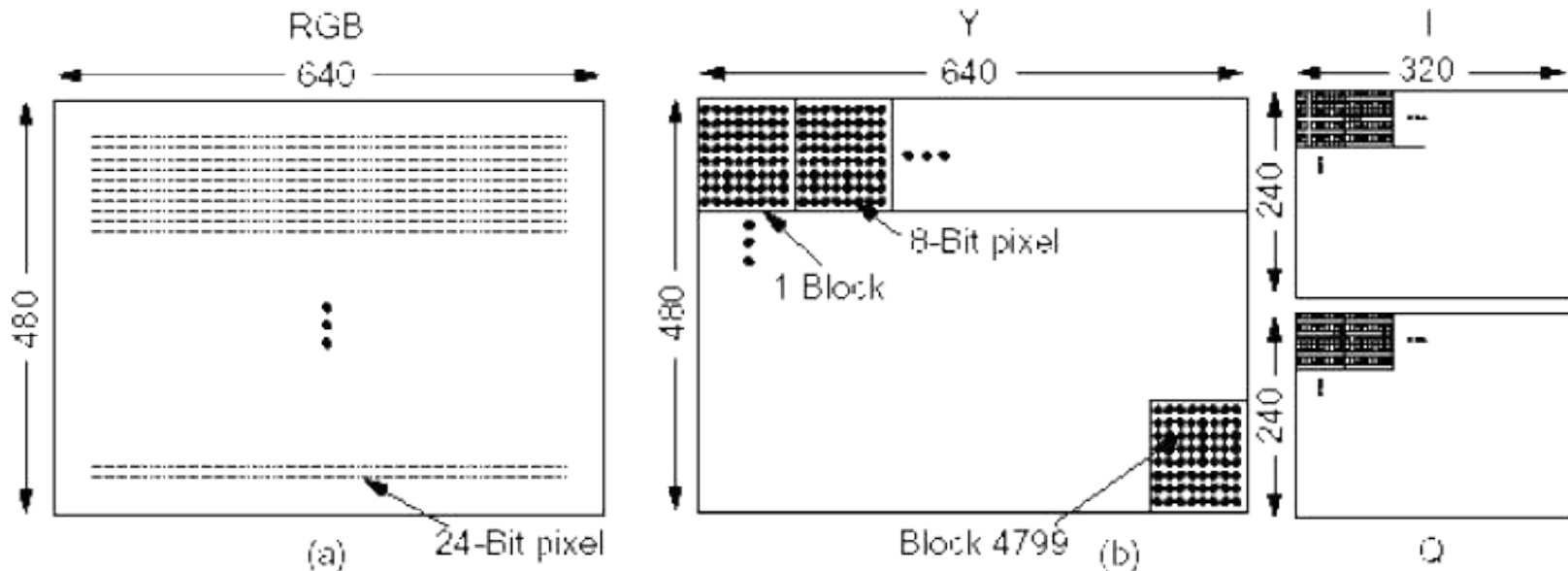
- We will look at each stage in turn



Block Preparation



- Take a 24 bit RGB image & convert to “luminance” & “chrominance”
 - Latter loses information through averaging





Conversion to Luminance & Chrominance



- From the RGB pixel values we calculate Y, I & Q values
 - (see previous slide)
- $Y = 0.30R + 0.59G + 0.11B$
- $I = 0.60R - 0.28G - 0.32B$
- $Q = 0.21R - 0.52G - 0.31B$
 - Why these values?
 - Physiological characteristics of *human* eyes



Block Conversion



- Each 8 x 8 block goes through Discrete Cosine Transform (DCT)
 - Similar to Fourier Transform
 - (See P&D for the maths)
- The calculated DCT coefficients then quantised according to a quantisation table



Compression



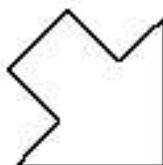
DCT Coefficients

150	80	40	14	4	2	1	0
92	75	36	10	6	1	0	0
52	38	26	8	7	4	0	0
12	8	6	4	2	1	0	0
4	3	2	0	0	0	0	0
2	2	1	1	0	0	0	0
1	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0

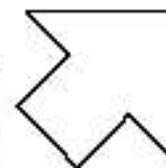
Quantized coefficients

150	80	20	4	1	0	0	0
92	75	18	3	1	0	0	0
26	19	13	2	1	0	0	0
3	2	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

Quantization table



1	1	2	4	8	16	32	64
1	1	2	4	8	16	32	64
2	2	2	4	8	16	32	64
4	4	4	4	8	16	32	64
8	8	8	8	8	16	32	64
16	16	16	16	16	16	32	64
32	32	32	32	32	32	32	64
64	64	64	64	64	64	64	64





Quantisation



- The quantisation tables can be adjusted by the user
- The table is then run-length encoded in a zig-zag pattern
- These results then Huffman encoded
 - This is how JPEG provides variable compression rates
 - And why we can't specify required compression ratio at outset



MPEG Compression



- MPEG (Motion Picture Experts Group) includes video & audio streams
 - (Audio is MP3)
 - Synchronised to embedded time code
- Non-symmetrical encoding / decoding times
 - Need to identify scene changes
 - Need to identify moving features



MPEG Frame Types



- Video has four different frame types
 - Intracoded (I)
 - Full frame encoded in a JPEG like form
 - Predictive (I)
 - Difference from previous frame
 - Bidirectional (B)
 - Difference from previous & next frames
 - DC coded
 - Block average for fast forward (optional)



MPEG Transmission



- Frames are sent such that dependent frames are transmitted after the frame they depend on
 - Original frames – I B B P B B I
 - Transmit as – I P B B I B B
- Difficult in real time
 - Video conferencing often just frame by frame JPEG compression (M-JPEG)



Summary



- Lossy compression applies to the data transmitted over the network at application level
- Appropriate only for “analogue” data
- Can require high processing overhead
- Can result in very high compression ratios