2D image compression

2D image compression

 \cdot General methods not satisfactory for images:

RLE - short runs

Statistical methods – similar probabilities for different colors/shades of gray

Dictionary based methods - short repeating patterns (digitalization...)

2D image compression

- $\boldsymbol{\cdot}$ Fourier transform (FT), Discrete FT, Wavelet transform (WT), Space filling curves (SPC) ...
 - exploit pixel correlation (2D)
 - allow lose of information

Image compression

- Representation
- Transformation
- Quantization
- Encoding

Representation

- ·RGB, YUV, CIE, HSV ...
- Y = 0.299R + 0.587G + 0.114B
- U = 0.492(B Y)
- V = 0.877(R Y)
- ·Chrominance (UV) can be lossy \rightarrow highly compressed

Representation - YUV







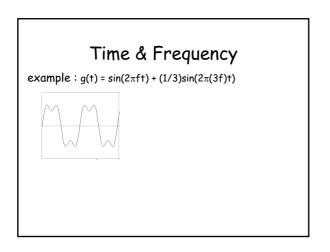


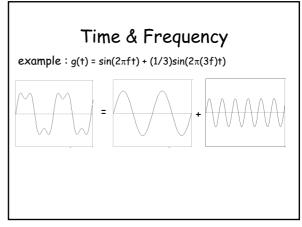
Fourier transform

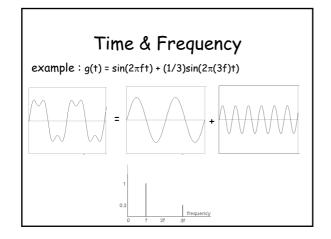
Transform = mathematical tool to solve problems Change a quantity to another form that may exhibit useful features : XCVI \times XII \rightarrow 96 \times 12 = 1152 \rightarrow MCLII

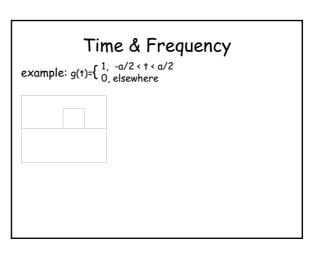
Signal in time domain \rightarrow frequency domain

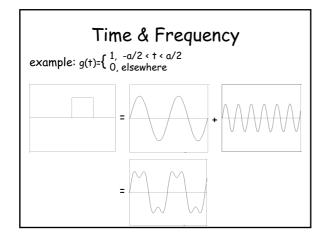
(Time = image space)

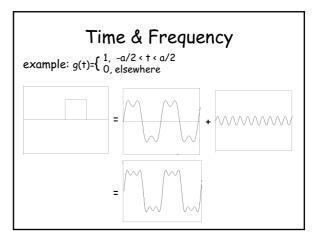


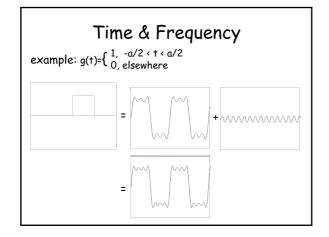


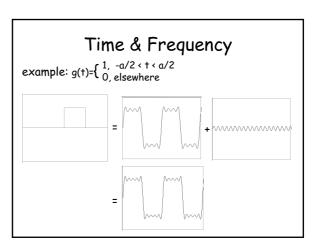


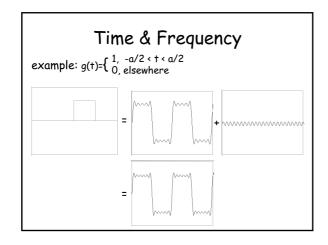


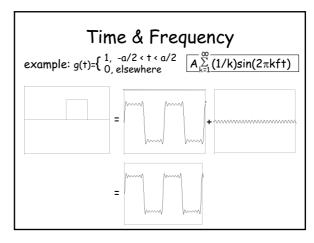


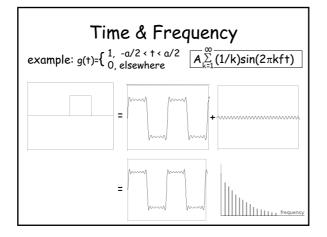


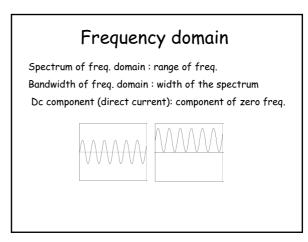












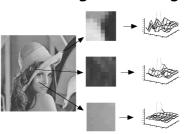
Fourier transform

 $G(f) = \int_{-\infty}^{\infty} g(t)[\cos(2\pi f t) - i \sin(2\pi f t)]dt$ $g(t) = \int_{-\infty}^{\infty} G(f)[\cos(2\pi f t) + i \sin(2\pi f t)]df$

• Discrete

$$\begin{split} G(f) &= (1/n) \sum_{t=0}^{n-1} g(t) [\cos(2\pi f t/n) - i \sin(2\pi f t/n)] \;, \quad 0 < f < n-1 \\ g(t) &= (1/n) \sum_{t=0}^{n-1} G(f) [\cos(2\pi f t/n) + i \sin(2\pi f t/n)] \;, \quad 0 < t < n-1 \end{split}$$

FT for digitized image



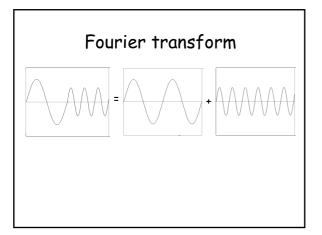
- neighboring pixels "close" values \rightarrow surface almost flat \rightarrow most FT coeff. small (large DC small AC coeff.)
- ·Loose unimportant image info buy cut Gij at right bottom

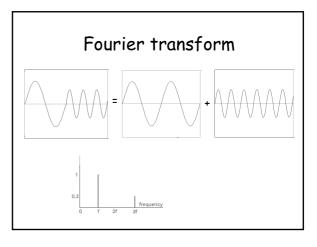
FT for digitized image

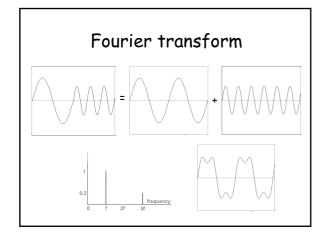
- · image = (continuous) signal of intensity function i(t)
- Sampling: store a finite sequence in memory i(1)...i(n)
- · The bigger the sample, the better the quality? no
- \bullet Sampling theory: sample an image and reconstruct it without loss of quality if we can :
 - Transform i(t) function from time to freq. Domain
 - Find the max freq. Fm
 - Sample i(t) at rate > 2fm
 - Store the sampled values in a bitmap

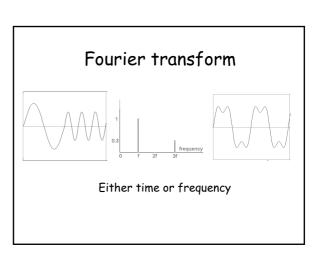
Fourier transform

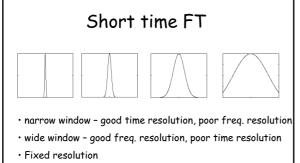


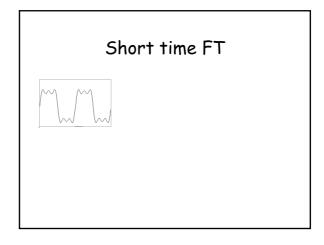


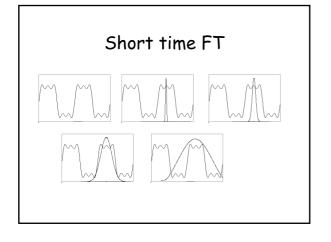


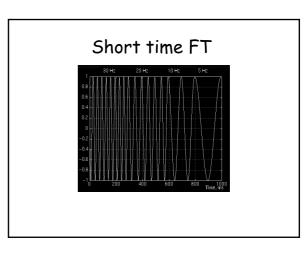


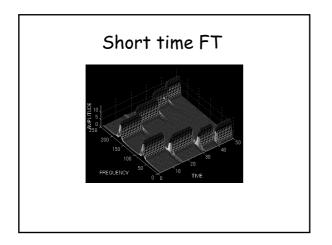


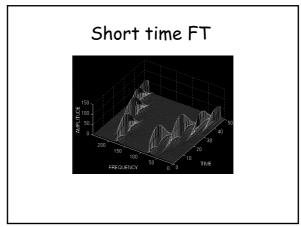


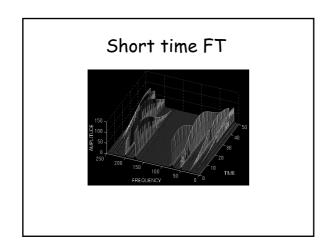


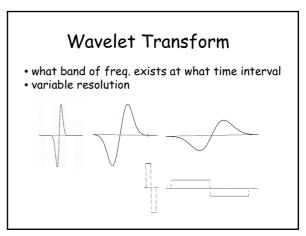


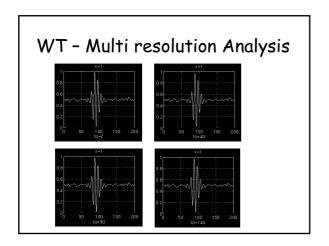


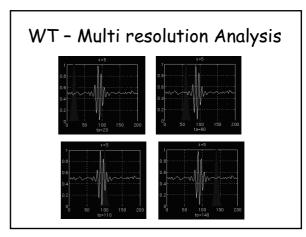


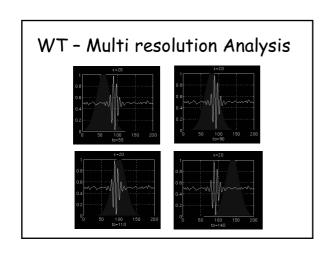


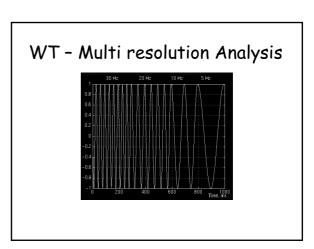




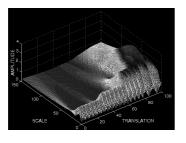




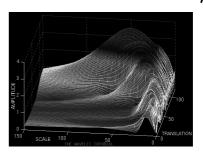




WT - Multi resolution Analysis



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WT - Multi resolution Analysis

Wavelet function $\psi_{ au,s} = rac{1}{\sqrt{s}} \psi\left(rac{t- au}{s}
ight)$

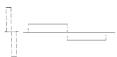
$$\text{Mexican hat} \quad \psi(t) = \frac{1}{\sqrt{2\pi}\sigma^3} \left(e^{\frac{-t^2}{2\sigma^2}}. \left(\frac{t^2}{\sigma^2} - 1 \right) \right)$$



Harr
$$\psi(x) = \begin{cases} 1 & 0 \le x < 1/2 \\ -1 & 1/2 \le x < 1 \\ 0 & x < 0, x > 1 \end{cases}$$

$$0 \le x < 1/2$$

 $1/2 \le x < 1$
 $x < 0, x > 1$



Wavelet Transform

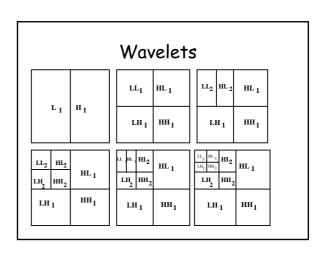
$$\begin{split} CWT_x^{\psi}(\tau,s) &= \Psi_x^{\psi}(\tau,s) = \frac{1}{\sqrt{|s|}} \int x(t) \psi^* \left(\frac{t-\tau}{s}\right) \, dt \\ \psi_{\tau,s} &= \frac{1}{\sqrt{s}} \psi \left(\frac{t-\tau}{s}\right) \end{split}$$

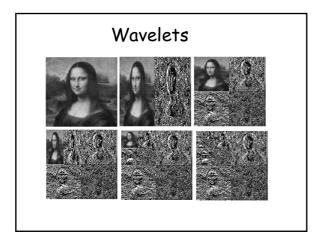
• inverse

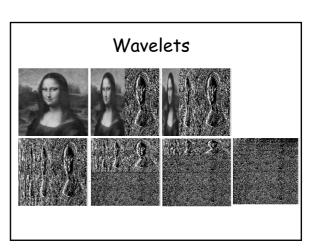
$$x(t) = \frac{1}{c_\psi^2} \int_s \int_\tau \Psi_x^\psi(\tau,s) \frac{1}{s^2} \psi\left(\frac{t-\tau}{s}\right) \, d\tau \, ds$$

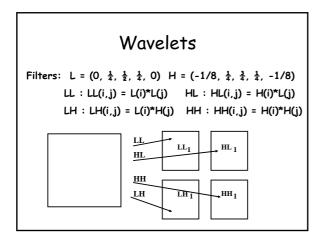
- scale = 1/freq.
 Discretization : non uniform sampling (high scale → low freq. → decrease sampling)

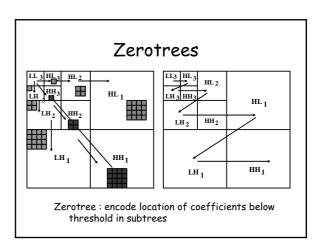
Wavelets (1, 2, 3, 4, 5, 6, 7, 8) $L = \sqrt{\epsilon^0} \left(\frac{1}{2}, \frac{1}{2}\right) H = \sqrt{\epsilon^0} \left(-\frac{1}{2}, \frac{1}{2}\right) :$ $\sqrt{\epsilon^0} \left(3/2, 7/2, 11/2, 15/2, -1/2, -1/2, -1/2, -1/2\right)$ $L = \sqrt{\epsilon^1} \left(\frac{1}{2}, \frac{1}{2}\right) H = \sqrt{\epsilon^1} \left(-\frac{1}{2}, \frac{1}{2}\right) :$ $\sqrt{\epsilon^1} \left(10/4, 26/4, -4/4, -4/4, -1/2, -1/2, -1/2, -1/2\right)$ $L = \sqrt{\epsilon^2} \left(\frac{1}{2}, \frac{1}{2}\right) H = \sqrt{\epsilon^2} \left(-\frac{1}{2}, \frac{1}{2}\right) :$ $\sqrt{\epsilon^2} \left(36/8, -16/8, -4/4, -4/4, -1/2, -1/2, -1/2, -1/2\right)$ $\left(\frac{36/8}{\sqrt{\epsilon^0}}, -\frac{16/8}{\sqrt{\epsilon^0}}, -\frac{4/4}{\sqrt{\epsilon^1}}, -\frac{4/4}{\sqrt{\epsilon^1}}, -\frac{1/2}{\sqrt{\epsilon^2}}, -\frac{1/2}{\sqrt{\epsilon^2}}, -\frac{1/2}{\sqrt{\epsilon^2}}\right)$ $L = \left(0, \frac{1}{4}, \frac{1}{2}, \frac{1}{4}, 0\right) H = \left(-1/8, \frac{1}{4}, \frac{3}{4}, \frac{1}{4}, -1/8\right)$











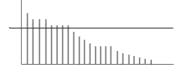
Wavelet coding using zerotrees

- •Encode = several passes with exp. decrease thresholds
- \cdot Each pass send significance bits for new significant coeff
- Additionally send refinement bits for coeff that became significant in an earlier pass (no need to send their location)
- $\boldsymbol{\cdot}$ Decoder can reconstruct geometry associated with any bits prefix



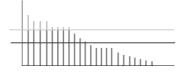
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Wavelets

- · Effective decorrelation
- · Correlation: Coarse level fine level

JPEG image compression

- lossy (also lossless mode)
- · works best for continuous-tone images,
- Advantage use many parameters, user can adjust amount of data lost (and thus compression ration)
- · Progressive and hierarchical coding
- · Main idea -lose data for which the human eye is not sensitive

JPEG image compression

- Representation RGB → luminance/chrominance
- Downsampling (color images only), for chrominance components only
 • DCT 8x8 blocks (data units)

$$G_{f} = (\frac{1}{2}) C_{f} \sum_{t=0}^{7} P_{t} \cos((2t+1)f\pi/16), C_{f} = \begin{cases} \frac{1}{\sqrt{2}} & f=0\\ 1 & f=1...7 \end{cases}$$

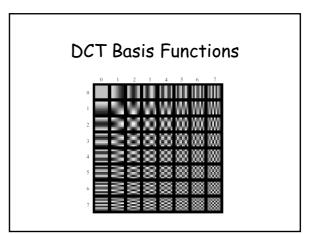
- real numbers
- fast implementation
- Separable (row/column)

JPEG image compression

- DCT on blocks (not entire image): small block
 - faster
 - correlation exists between neighboring pixels

large block

- better compression in "flat" regions
- Prepares data for losing information



JPEG image compression

- Quantization 64 different coefficients

 - Lossy step.High freq. has large QC.
 - Each QC is JPEG parameter, can be specified by the user.
 - Predefined tables for each color component

JPEG image compression

Encoding — RLE, Huffman conding, arithmetic
• Scan each block in zig-zag order



- \bullet DC component first DC + differences (Huffman) : in continuous-tone images average of pixels in adjacent data
- AC component RLE + Huffman/Arithmetic coder: many zeros, few nonzero
- Headers

JPEG image compression

- Progressive mode (for quick coarse preview)

 "scans" of high freq., each sharper image
 Slow encoding (all step for each scan)

 Hierarchical mode (when high resolution image should be output in low resolution)
 - store image several times, several resolutions
- ·Lossless mode
 - differences from prediction based on neighbors
 - Huffman/arithmetic coder

