
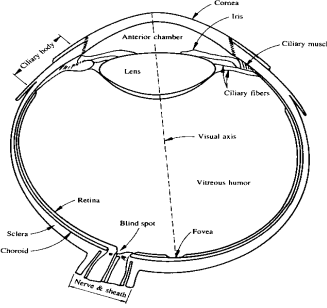



4TN3 Image Processing

Digital Image Fundamentals



Structure of Human Eye

Digital Image Fundamentals

- Elements of visual perception
- Image sensing and acquisition
- Sampling and quantization
- Relationship between pixels






Image formation in the eye

- Lens is flexible and its refraction is controlled by its thickness.
- Thickness is controlled by the tension of muscles.
- Focus on distance objects: lens is relatively flattened, refractive power is minimum.
- Focus on near objects: lens is thicker, refractive power is maximum
- Perception takes place by excitation of receptors which transform radiant energy into electrical impulses that are decoded by the brain.



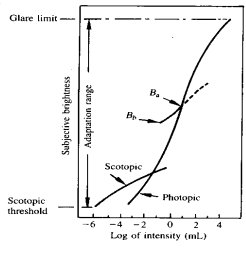

The Human Visual System (HVS)

- A true measure of image processing quality is how well the image appears to the observer.
- The HVS is very complex and is not understood well in a complete sense. However, many of its properties can be identified and used to our advantage.



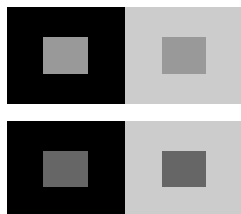
Brightness & Intensity

- The dynamic range of light intensity to which eye can adapt is enormous.
- Brightness is a logarithmic function of light intensity.
- HVS cannot operate over the entire range simultaneously. It accomplishes large variations due to **brightness adaptation**.

Brightness & Intensity

- Relationship between brightness and intensity is not a simple function!



A simple image model

- The image magnitude depends on:
 - The amount of source illumination incident on the scene, $i(x,y)$
 - The amount of illumination reflected by the objects in the scene, $r(x,y)$

$$f(x, y) = i(x, y) \cdot r(x, y)$$

(x,y) : coordinates

Total absorption: $r(x,y)=0$

Total reflection: $r(x,y)=1$

Brightness & Intensity

Mach Band Effect

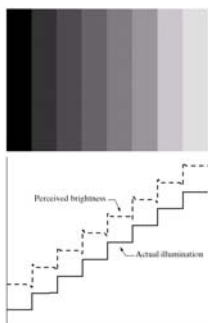


FIGURE 2.7
(a) An example showing that perceived brightness is not a simple function of intensity. The relative vertical positions between the two profiles in (b) have no special significance; they were chosen for clarity.

Sampling & Quantization

- Computer processing: image $f(x,y)$ must be digitized both spatially and in amplitude
- Digitization in spatial coordinates: sampling
- Digitization in amplitude: quantization
- Image: $[f(i,j)]_{N \times M}$
- What should be the values of N , M and the number of gray levels G ?
- Normally: $N=2^n$, $M=2^m$, $G=2^k$

Image sensing and acquisition

- If a sensor can be developed that is capable of detecting energy radiated by a band of the EM spectrum, we can image events in that band.
- Image is generated by energy of the illumination source reflected (natural scenes) or transmitted through objects (X-ray).
- A sensor detects the energy and converts it to electrical signals.
- Sensor should have a material that is responsive to the particular type of energy being detected.

Sampling & Quantization

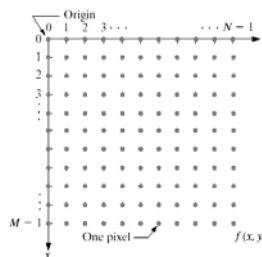


FIGURE 2.18
Coordinate convention used in this book to represent digital images.

Sampling & Quantization

- Number of bits required to store the image:
 $N \times M \times k$.
- The more the values of N,M and G, the better approximation of a continuous image.
- Storage and processing requirements increase as well.

Effects of Reducing Gray Levels

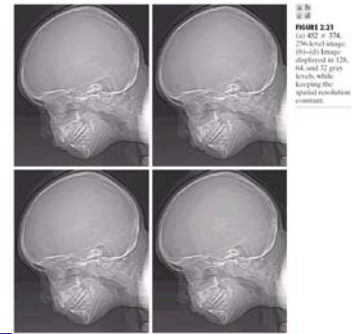


FIGURE 2.21 (Continued) Grayscale image displayed in 128, 64, and 32 gray levels, while keeping the spatial resolution constant.

Effects of Reducing Spatial Resolution

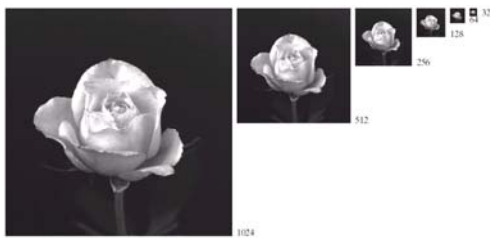


FIGURE 2.19 A 1024×1024 , 8-bit image subsampled down to size 32×32 pixels. The number of allowable gray levels was kept at 256.

Effects of Reducing Gray Levels

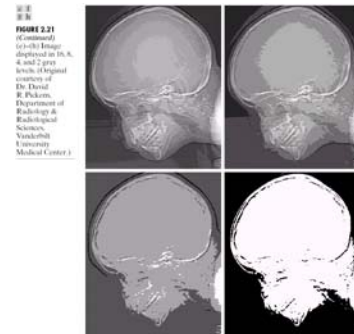


FIGURE 2.21 (Continued) Grayscale image displayed in 16, 8, 4, and 2 gray levels of original contrast of Dr. David R. Pickens, Department of Radiology & Radiological Sciences, Vanderbilt University Medical Center.

Effects of Reducing Spatial Resolution

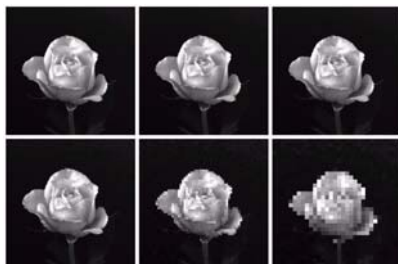
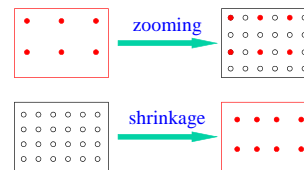


FIGURE 2.20 (a) 1024×1024 , 8-bit image (b) 512×512 image resampled into 1024×1024 pixels by row and column duplication. (c) through (f) 256×256 , 128×128 , 64×64 , and 32×32 images resampled into 1024×1024 pixels.

Zooming and Shrinkage

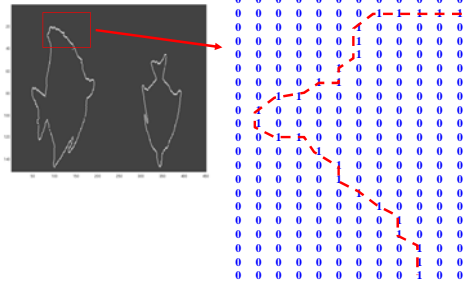
- **Zooming:** increasing the resolution (size) of an image.
- **Shrinkage:** decreasing the resolution of an image.



Zooming

- Pixel replication: interpolated pixel in the enlarged image is set to the gray-level of its nearest pixel in the original image.
- Bilinear interpolation: interpolated pixel in the enlarged image is set to the average its four nearest pixels in the original image.

Relationship between pixels



Zooming: example

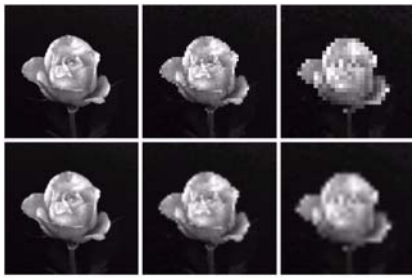


FIGURE 2.25 Top row: images zoomed from $128 \times 128 \times 64$, and 32×32 pixels to 1024×1024 pixels, using nearest neighbor gray-level interpolation. Bottom row: same sequence, but using bilinear interpolation.

Relationship between pixels

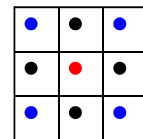
- Neighbors
- Adjacency
- Path
- Connectivity
- Region
- Boundary
- Distance

Shrinkage

- Shrinkage by an integer number can be done by deleting some of the rows and columns of the image.
- Shrinkage by a non-integer factor can be done as the inverse of zooming.

Basic relationships between pixels

- A pixel p at coordinates (x,y) has four horizontal and vertical neighbors:
 $N_4(P) = \{(x+1,y), (x-1,y), (x,y+1), (x,y-1)\}$
- The four diagonal neighbors of P
 $N_D(P) = \{(x+1,y+1), (x-1,y-1), (x-1,y+1), (x+1,y-1)\}$
- The eight point neighbors of P
 $N_8(P) = N_4(P) \cup N_D(P)$



Adjacency

- Two pixels are adjacent if they are neighbors and their gray levels are **similar**
- V: set of gray levels
- Similar gray level means that the gray levels of both pixels belong to set V
- Exp:
 - Binary images: $V=\{1\}$
 - Gray level image: $V=\{32,33, \dots,63,64\}$

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Region

- R: a subset of pixels in an image.
- R is called a region if every pixel in R is connected to any other pixel in R.
- Boundary (border or contour) of a region: set of pixels in the region that have one or more neighbors that are not in R.

0	1	1	0
0	1	1	1
0	0	1	0
0	0	0	0

0	1	1	0
0	1	1	1
0	0	1	0
1	0	0	0

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Adjacency

- 4-adjacency: Two pixels p and q with values from V are 4-adjacent if q is in $N_4(p)$
- 8-adjacency: Two pixels p and q with values from V are 8-adjacent if q is in $N_8(p)$
- 4-adjacency: broken paths
- 8-adjacency: multiple paths

0	1	-1
0	1	0
0	0	1

0	1	-1
0	1	0
0	0	-1

0	1	-1
0	1	0
0	0	1

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Connected Components

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Adjacency

- m-adjacency: Two pixels p and q with values from V are m-adjacent if:
 - q is in $N_4(p)$ or
 - q is in $N_D(p)$ and the intersection of $N_4(p)$ and $N_4(q)$ has no pixels with values in V.

0	1	1
0	1	0
0	0	1

0	1	1
0	1	0
0	0	1

0	1	-1
0	1	0
0	0	1

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Distance measures

- For pixels p,q, and z with coordinates (x,y), (s,t) and (v,w), respectively, D is a distance functions if:
 - $D(p, q) \geq 0$
 - $D(p, q) = D(q, p)$
 - $D(p, z) \leq D(p, q) + D(q, z)$
 - $D_e = [(x-s)^2 + (y-t)^2]^{1/2}$

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Distance measures

- D_4 distance

$$D_4(p, q) = |x - s| + |y - t|$$

- D_8 distance

$$D_8(p, q) = \max\{|x - s|, |y - t|\}$$

Pixel values D_4 distances D_8 distances

0	1	1	2	1	2	1	1	1
0	①	0	1	0	1	1	0	1
0	0	1	2	1	2	1	1	1



End of Lecture



Distance measures

- D_m distance: length of the shortest m-path between two pixels
- D_4, D_8 distance between p and q are independent of the pixels along the path
- D_m depends on the value of the pixels between p and q

0	0	①	0	0	①
1	1	0	0	1	0
①	0	0	①	0	0

$D_m=3$ $D_m=2$



Linear & Non-linear operations

- H : an operator whose inputs and outputs are images
- H is linear if for any two images f and g and any two scalars a and b

$$H(af+bg) = aH(f) + bH(g)$$

