

# INTRODUCERE

# In cursul de astazi

- Procesarea imaginii (definitii si exemple)
- Scopul procesarii imaginii
- Imaginea digitala
- Sistemul vizual uman
- Digitizare (frecventa Nyquist, alias, convolutie, functia de distributie a punctului)
- Cuantificare
- Modele de culori

# Generalitati

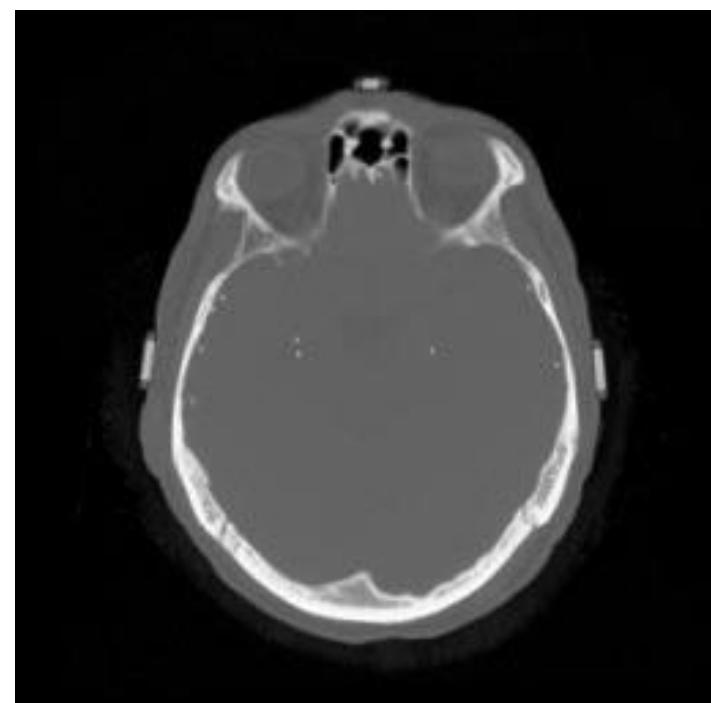
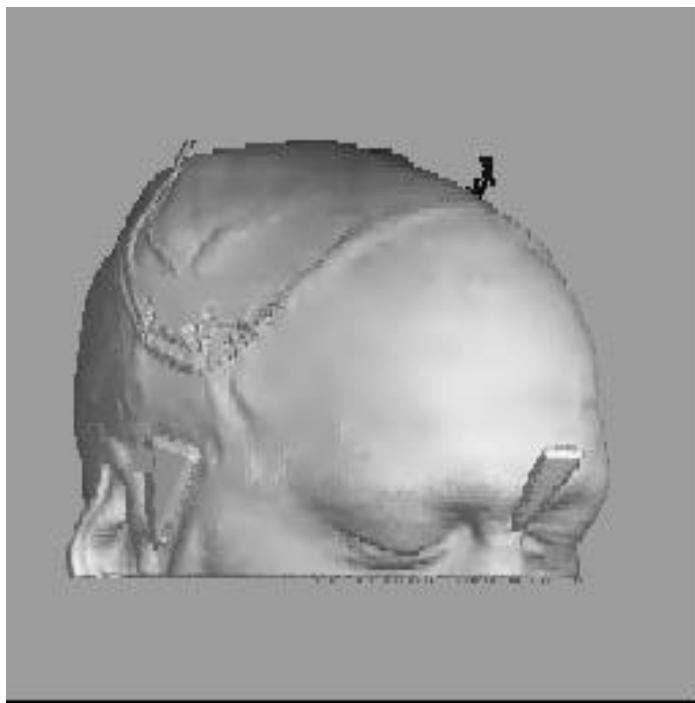
- procesarea imaginilor = manipularea si analiza informatiilor continute in imagini
- exemple simple:
  - utilizarea ochelarilor si a lentilelor;
  - controlul contrastului, luminozitatii, etc. televizorului, monitorului;
  - reflexia peisajului in apa, “fata morgana”, etc.
  - luarea unei fotografii cu o camera digitala

# Utilizari

- industrie (verificarea pieselor, aplicatii CAD/CAM, etc.)
- procesarea informatiilor (recunoasterea caracterelor)
- astronomie
- criminologie (recunoasterea fetelor, amprentelor, etc.)
- etc.

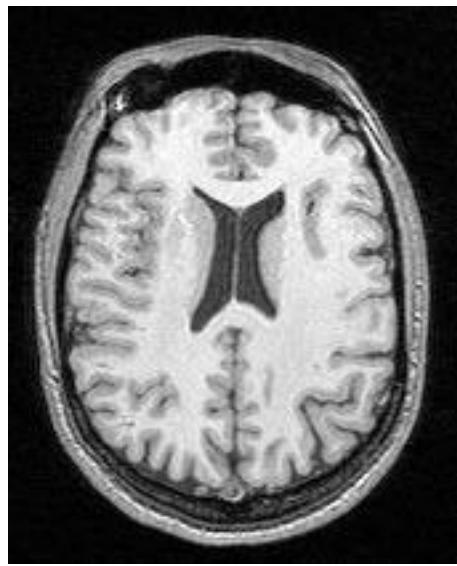
# Utilizari in domeniul medical (1)

- Vizualizare



# Utilizari in domeniul medical (2)

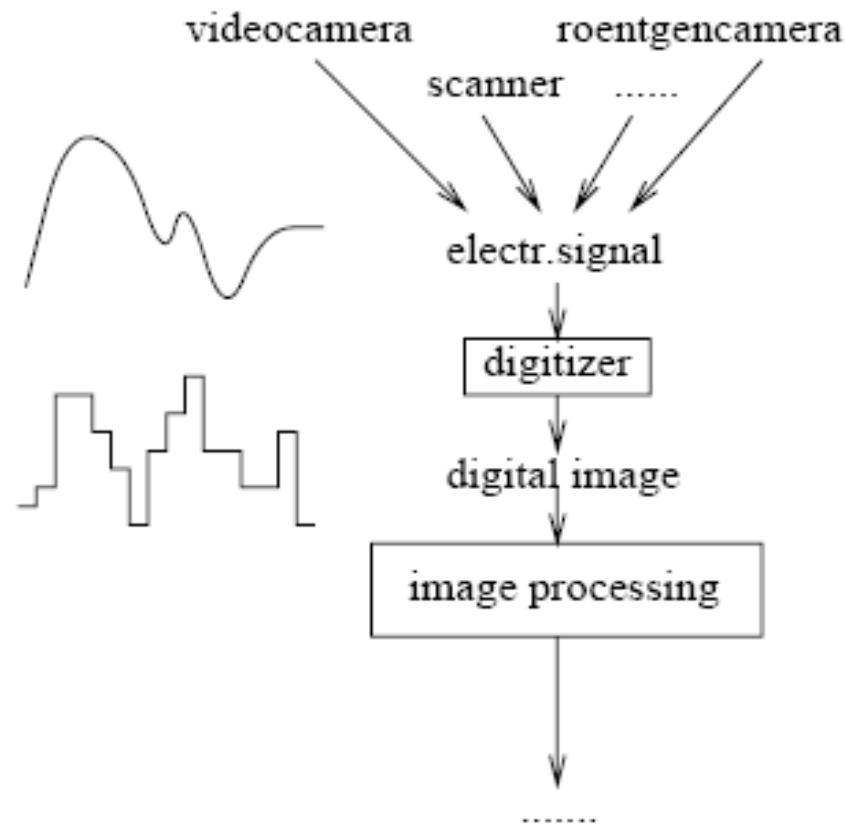
- Diagnosticare asistata de calculator
- Registrarea imaginilor (potrivirea imaginilor)
- Segmentarea imaginilor



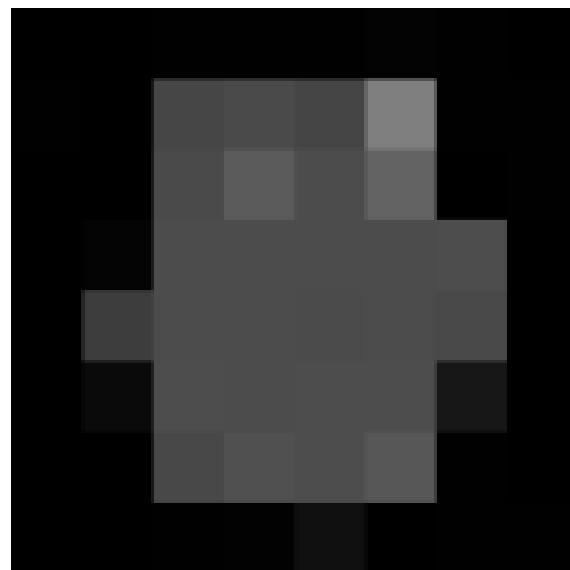
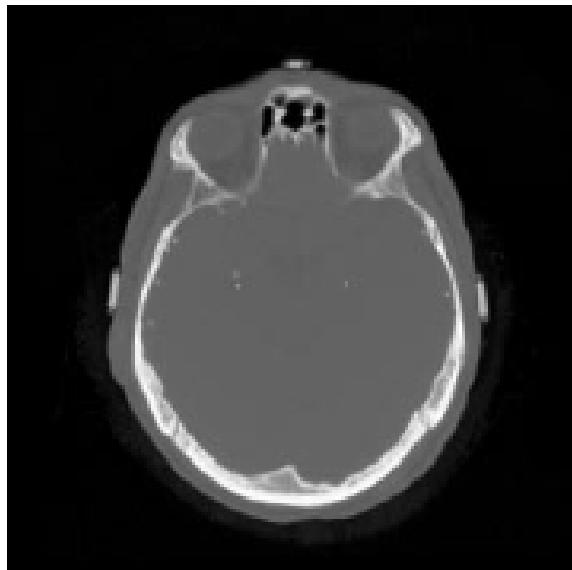
# Scopul procesarii imaginilor

- imbunatatirea imaginilor
- recunoasterea formelor (forma si textura)
- reducerea datelor la informatii mai usor de utilizat
- sinteza imaginii (2D -> 3D)
- combinarea imaginilor
- compresia datelor

# Obtinerea imaginilor digitale

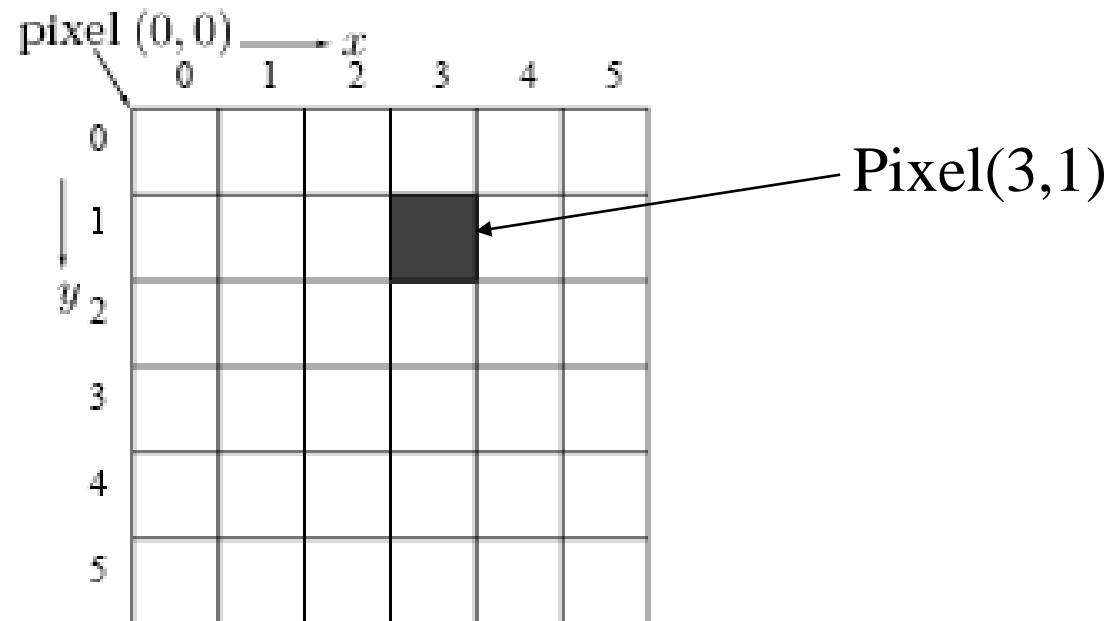


# Esantionare si cuantificare



00	00	01	01	01	03	01	00
01	00	46	4a	45	7f	01	01
00	00	4a	5b	4c	63	00	01
00	04	4c	4c	4c	4c	4d	00
00	3d	4c	4c	4b	4c	49	00
00	0a	4d	4c	4d	4d	17	00
00	00	48	50	4d	57	01	00
00	00	01	02	10	00	01	01

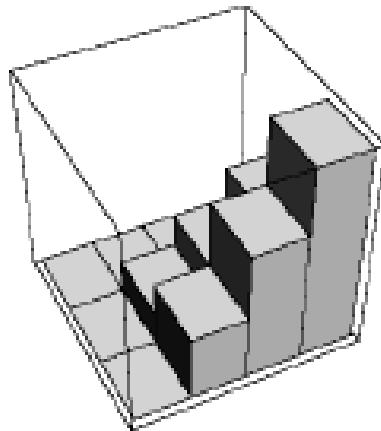
# Convenția coordonatelor



# Reprezentare matematica

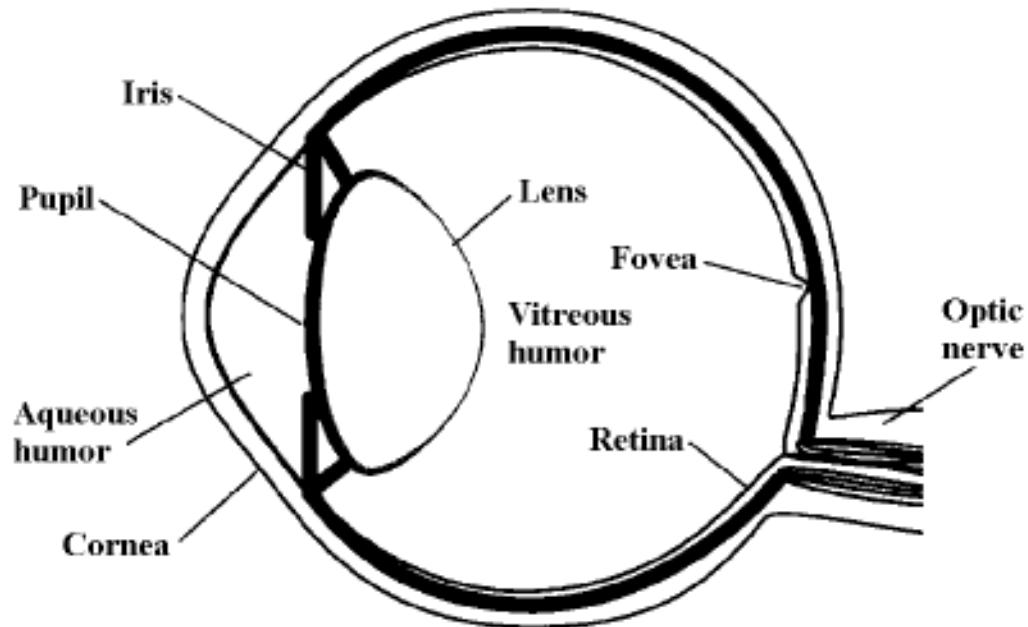
$f(x,y) = xy$  pentru  $(x,y) \in \{0,1,2,3\} \times \{0,1,2\}$

$$M = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 3 \\ 0 & 2 & 4 & 6 \end{pmatrix}$$



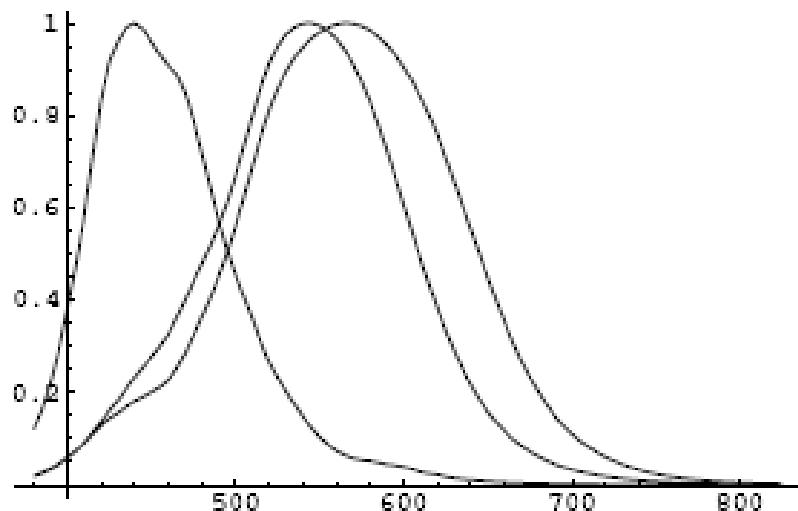
# Sistemul vizual uman

Luminanta <-> luminozitate (stralucire)

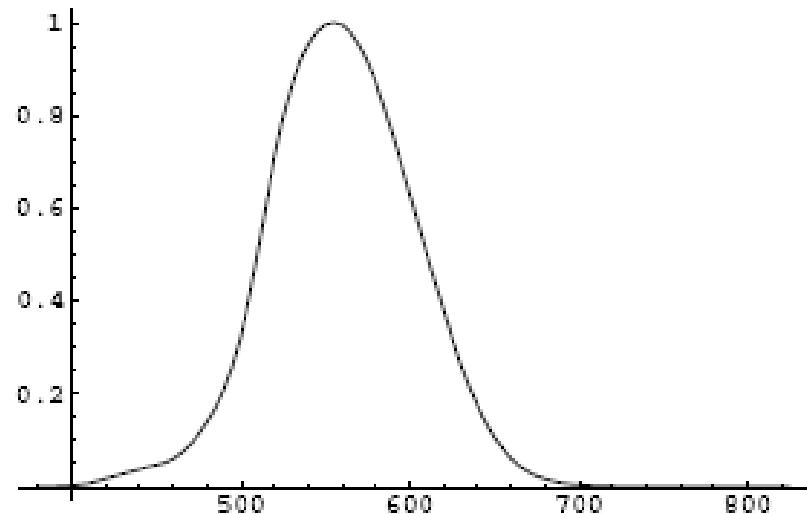


- *bastonase* (cca 100 mil.) – intensitatea luminoasa
- *conuri* (cca 7 mil.) – culorile si detaliile

# Sistemul vizual uman (2)



Senzitivitatea relativă a celor trei tipuri de conuri



Functia de eficientă luminoasă

# Luminanta versus luminozitate

**Luminozitatea unui obiect:**

$$I(\lambda) = \rho(\lambda) E(\lambda),$$

unde:  $\rho(\lambda)$  este reflectivitatea obiectului, [0,1]

$E(\lambda)$  este energia razeelor de lumina de unda  $\lambda$ .

**Luminanta unui obiect:**

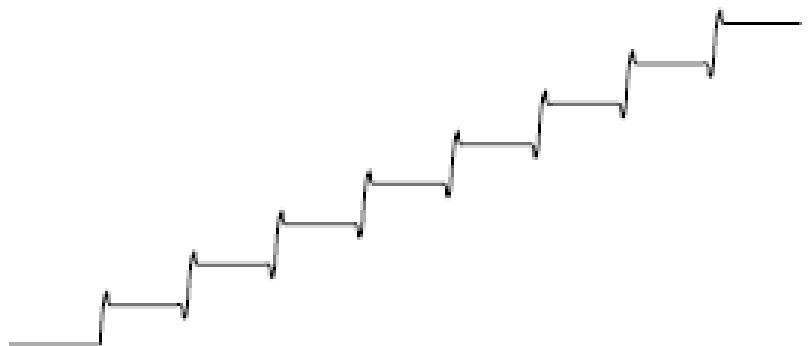
$$L = \int_0^{\infty} I(\lambda) V(\lambda) d\lambda,$$

unde  $V(\lambda)$  este functia eficientei luminoase a sistemului vizual

# Contrast

Contrast - perceptia luminozitatii unei zone in functie de intensitatea ariei inconjuratoare

Efectul de banda Mach – accentuarea schimbarilor de intensitate



# Contrast (2)

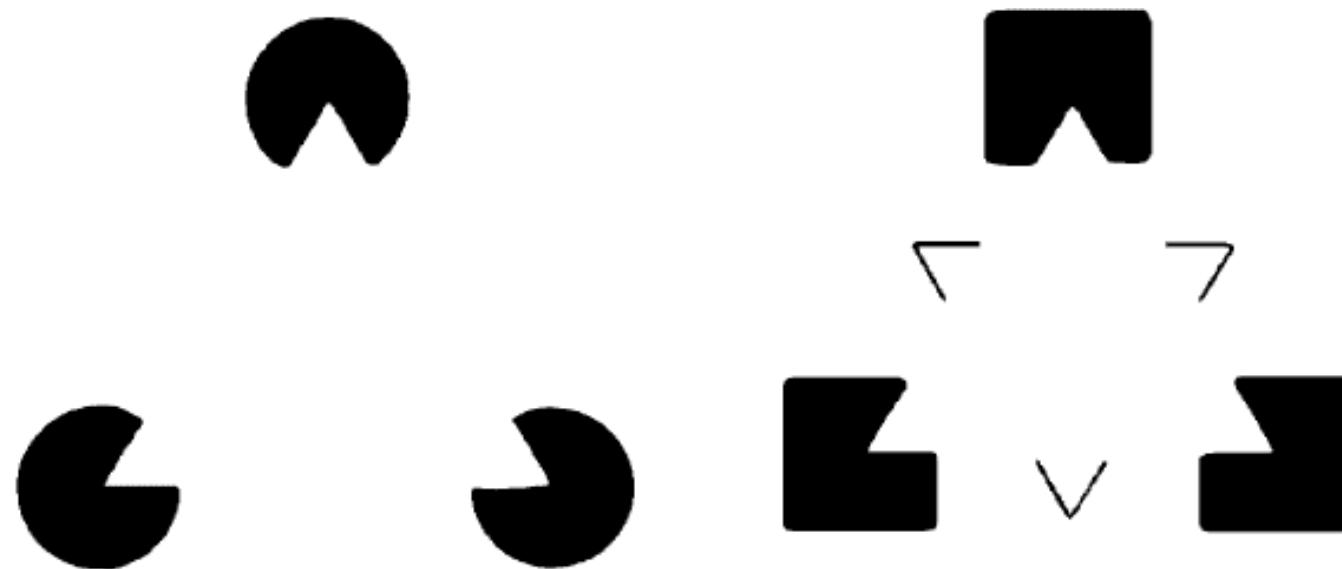
Efectul contrastului simultan



Inelul lui Benussi

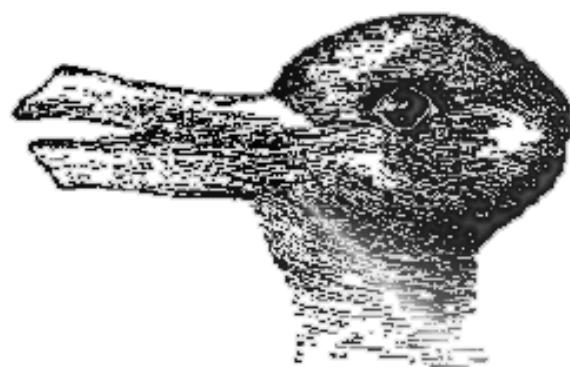
# Contrast (3)

Fenomenul Gestalt



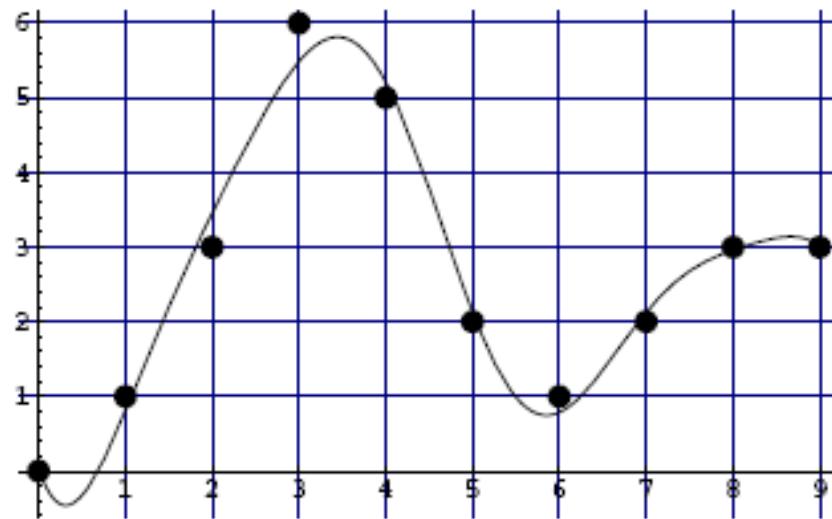
# Contrast (4)

Fenomenul Gestalt

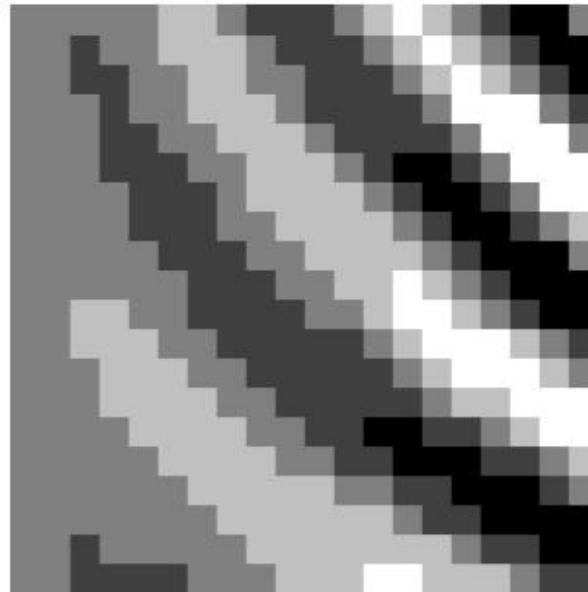
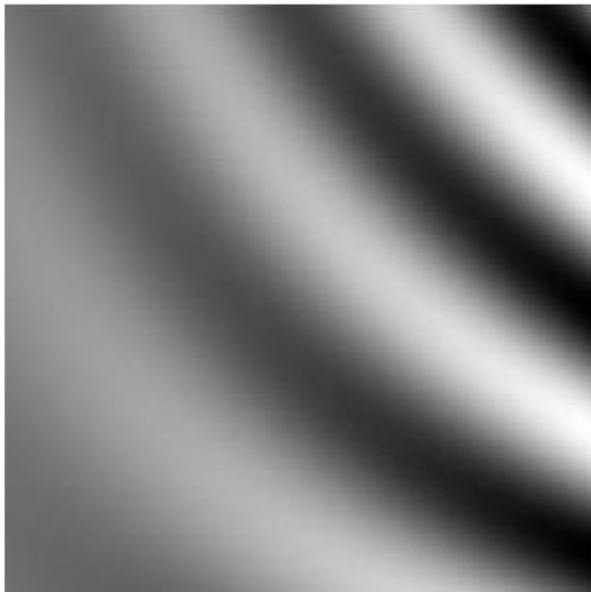
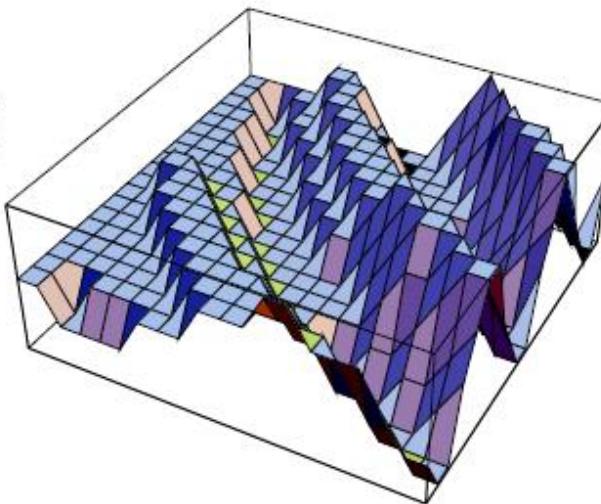
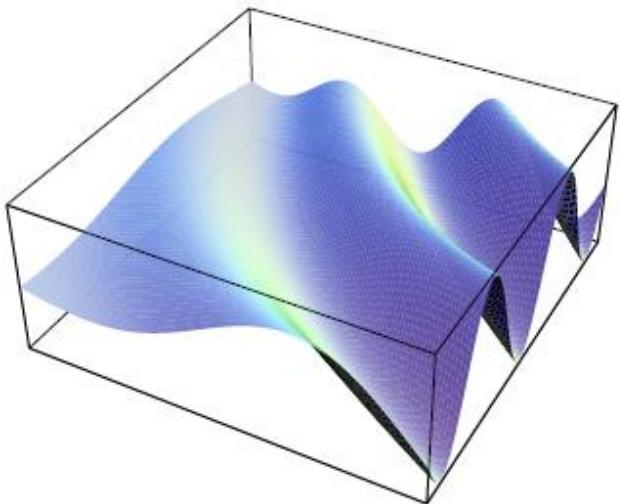


# Achizitia imaginilor digitale

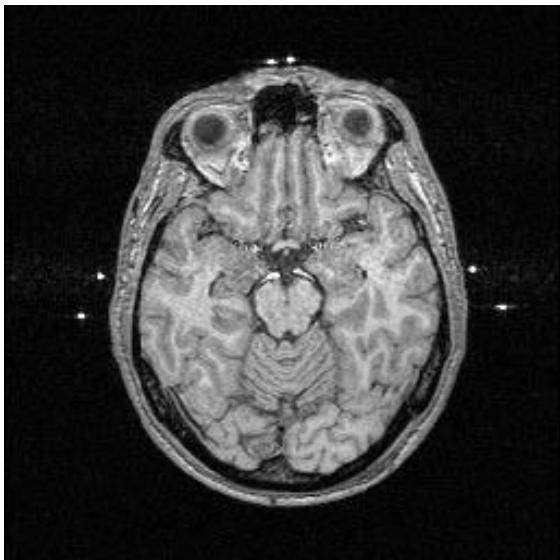
# Esantionarea unui semnal 1D



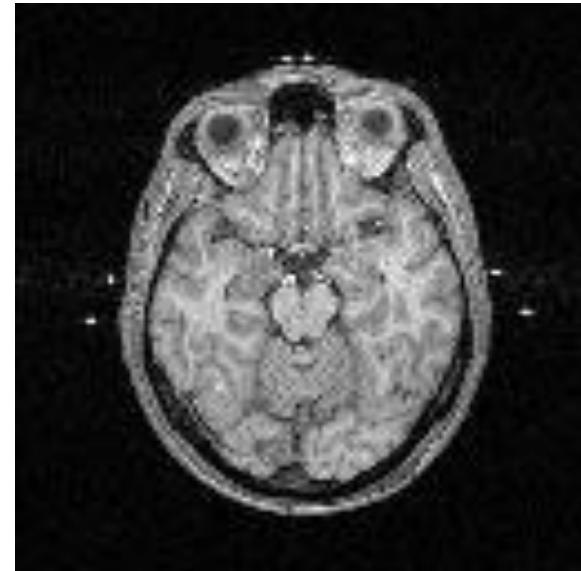
# Esantionarea unui semnal 2D



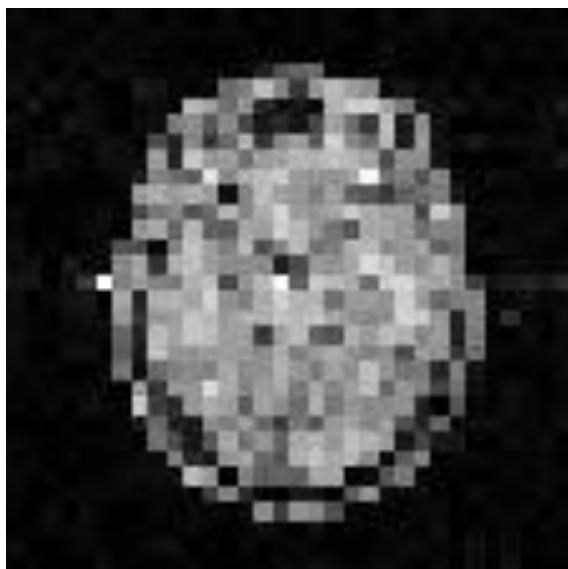
# Rezolutia spatiala



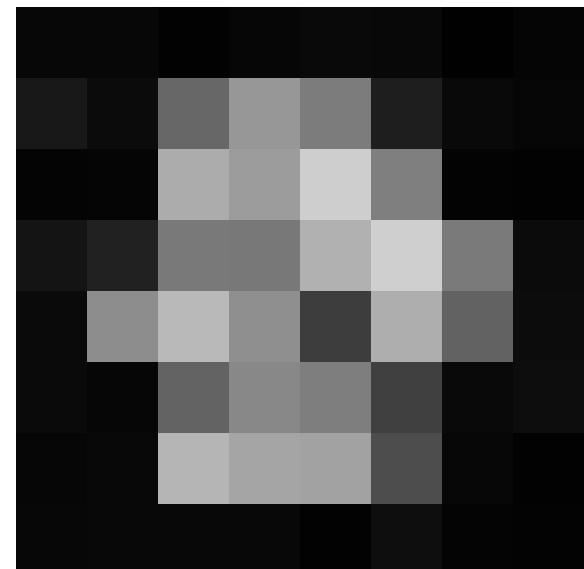
← 256 x 256



128 x 128 →

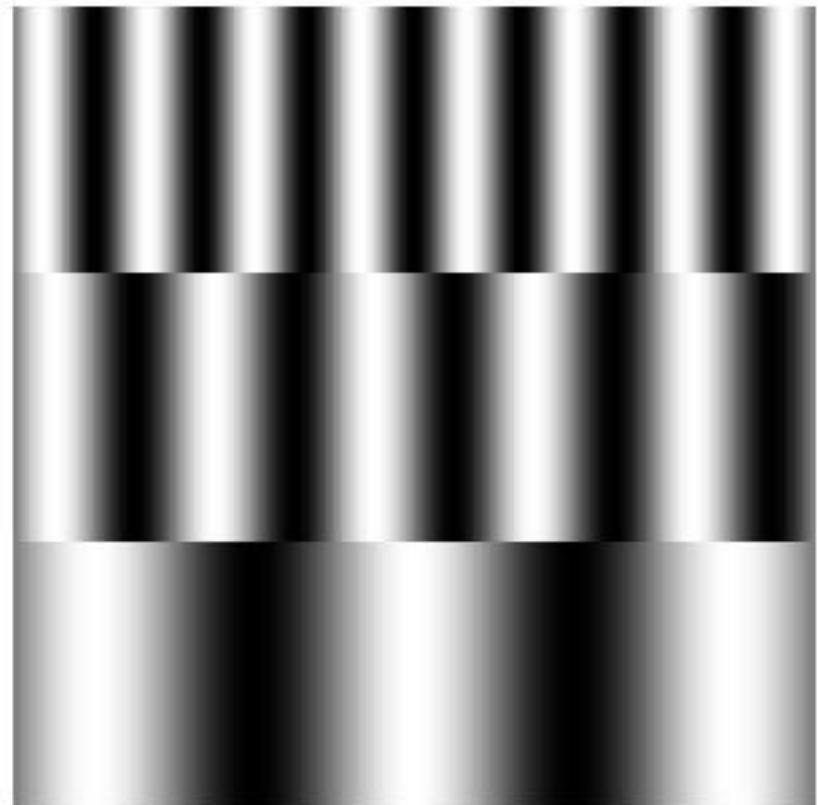
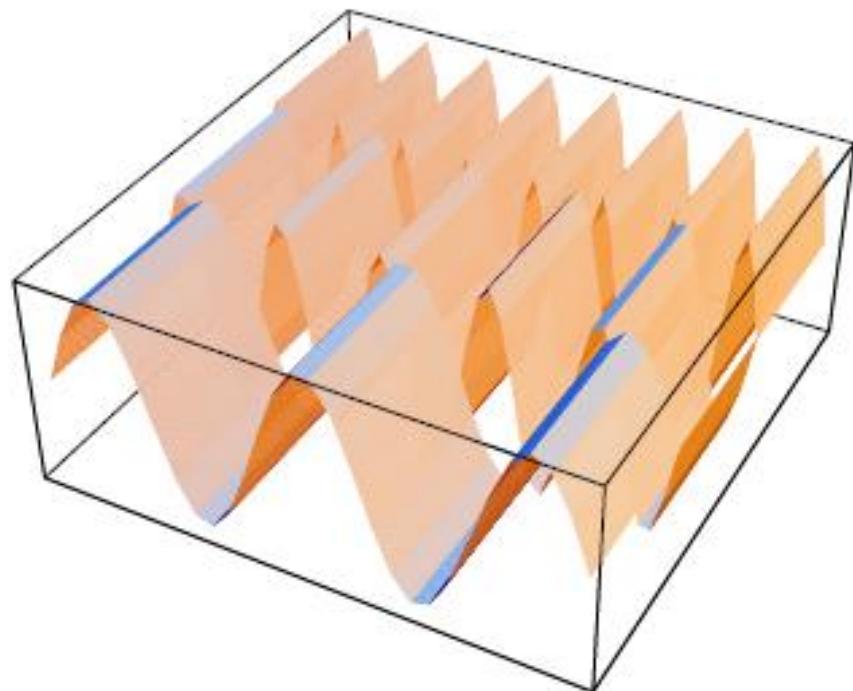


← 32 x 32



8 x 8 →

# Frecventa spatiala



# Frecventa

“Frecventa” = frecventa undei sinusoidale corespunzatoare

$$f(x) = \sin(ax), \text{ unde } v=a/(2\pi) \text{ este frecventa (spatiala)}$$

Frecventa cea mai joasa = 0 => imagine constanta

Rezolutia este limitata => detaliul cel mai fin este limitat =>  
=> frecventa cea mai inalta este limitata

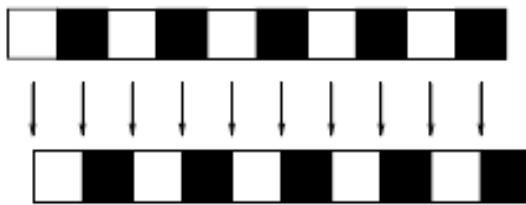
# Teorema esantionarii

*Pentru a capta cele mai mici detalii (cele mai înalte frecvențe), frecvența de esantionare trebuie să fie  $2\xi$  sau mai mare, unde  $\xi$  reprezintă cea mai înalta frecvență din imaginea originală*

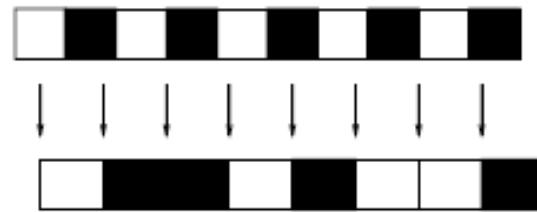
Frecvența Nyquist =  $2\xi$

# Alias (1)

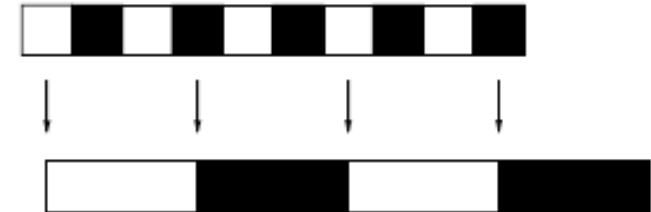
$$\xi_n = 1/2 \text{ per pixel} \Rightarrow \text{Frecventa Nyquist} = 2 \cdot 1/2 = 1$$



Esantionare cu  
 $v=1$   
 $\Rightarrow$ se obtine  
imaginea  
originala

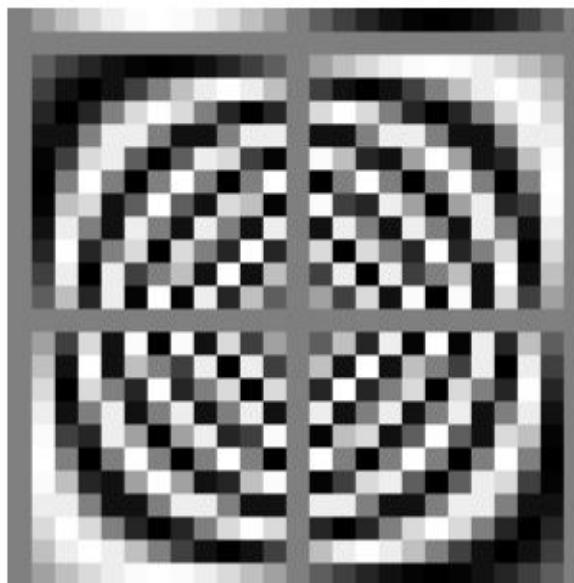
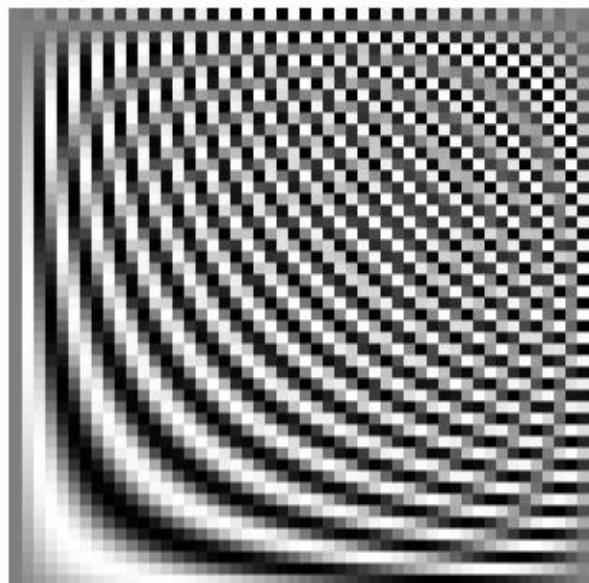
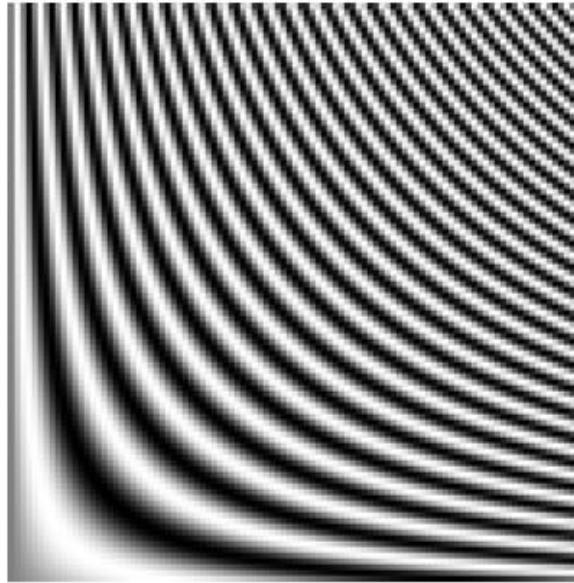
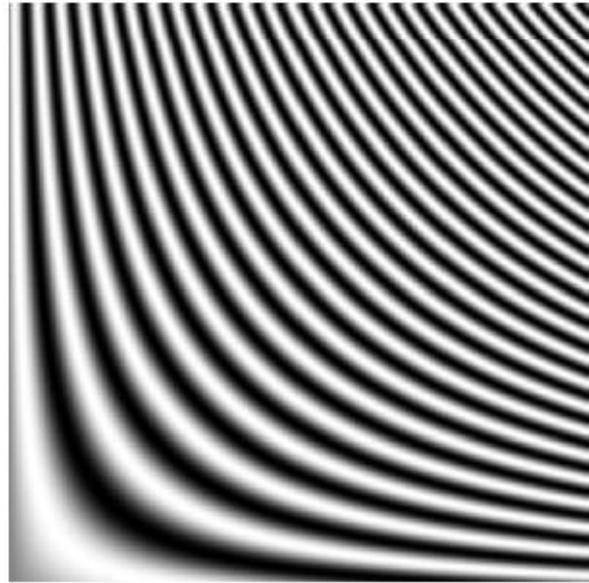


Esantionare cu  
 $v=4/5$   
 $\Rightarrow$ pierderea unor  
detalii



Esantionare cu  
 $v=1/3$   
 $\Rightarrow$ un sablon care  
nu era prezent in  
imaginea originala

# Alias (2)



Fie functia:

$$f(x, y) = \sin(2\pi x(26-y))$$

definita in domeniul:

$$D_f = \{(x, y) \in [0, 1] \times [1, 25]\}$$

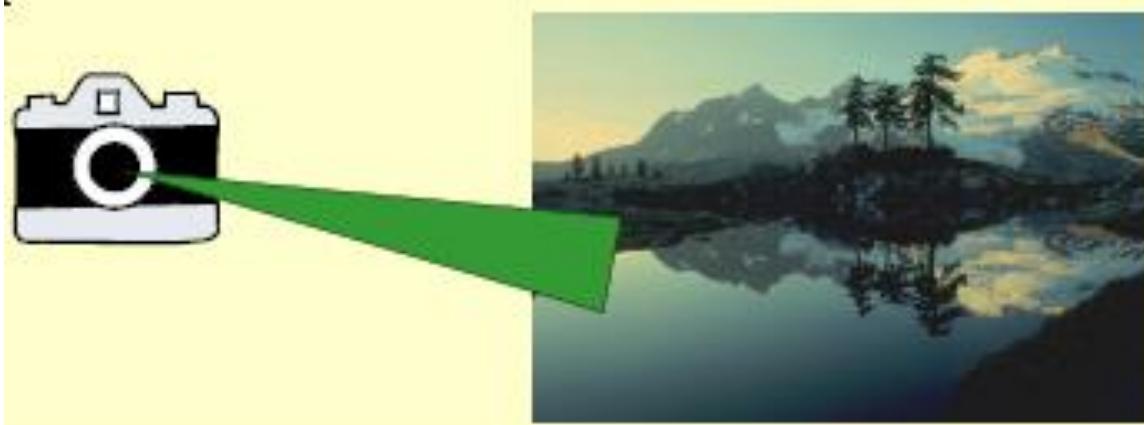
Observam ca:

$$v = 25/\text{linie sus}$$

$$v = 1/\text{linie jos}$$

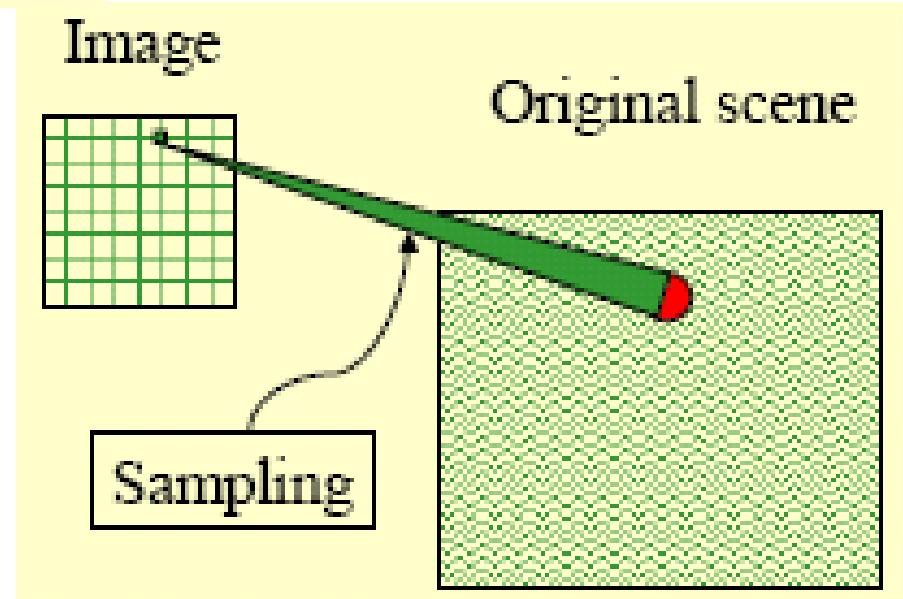
Esantionare la diferite  
frecvențe: 400, 100, 50, 25

# Esantionare



- esantionarea ia în considerare o vecinătate

- esantionare cu precizia infinită = fals



# Esantionare si convolutie

- modelarea matematica = convolutia:  $f=\text{imag orig.}$ ,  $g=\text{nucleul}$
- convolutia a doua functii integrabile  $f,g:\mathbb{R} \rightarrow \mathbb{R}$  (im.esant.)

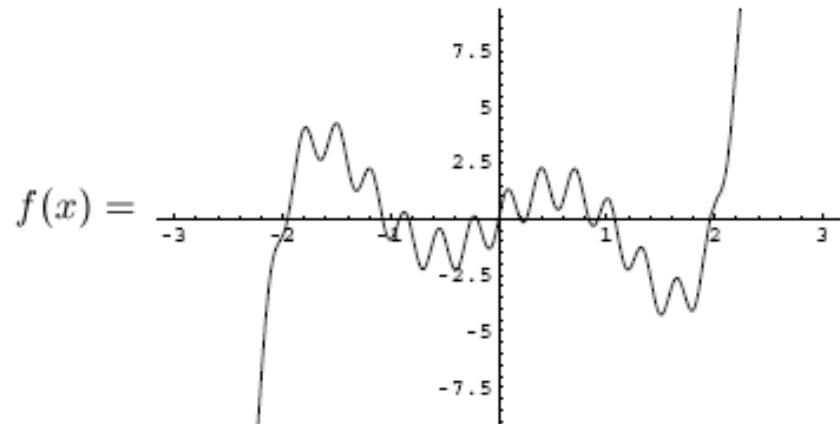
$$(f * g)(x) = \int_{-\infty}^{\infty} f(a) g(x-a) da$$

- convolutia a doua imagini  $f$  si  $g$  bidimensionale

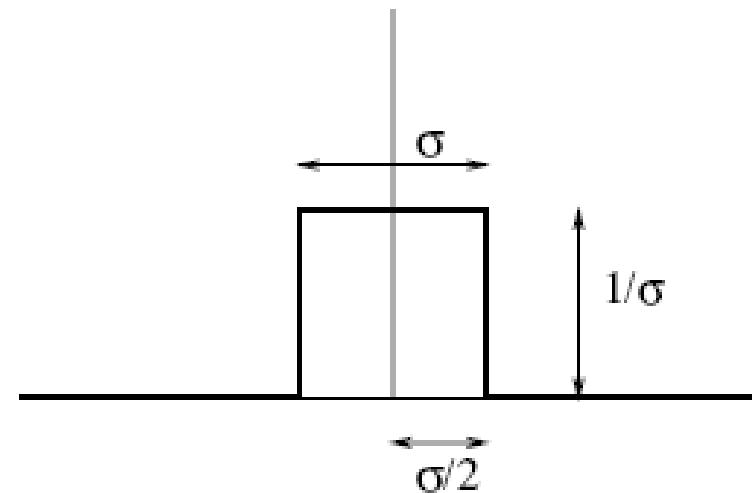
$$(f * g)(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(a, b) g(x-a, y-b) da db$$

- convolutia este comutativa  $f * g = g * f$

# Convolutia unei functii unidimensionale

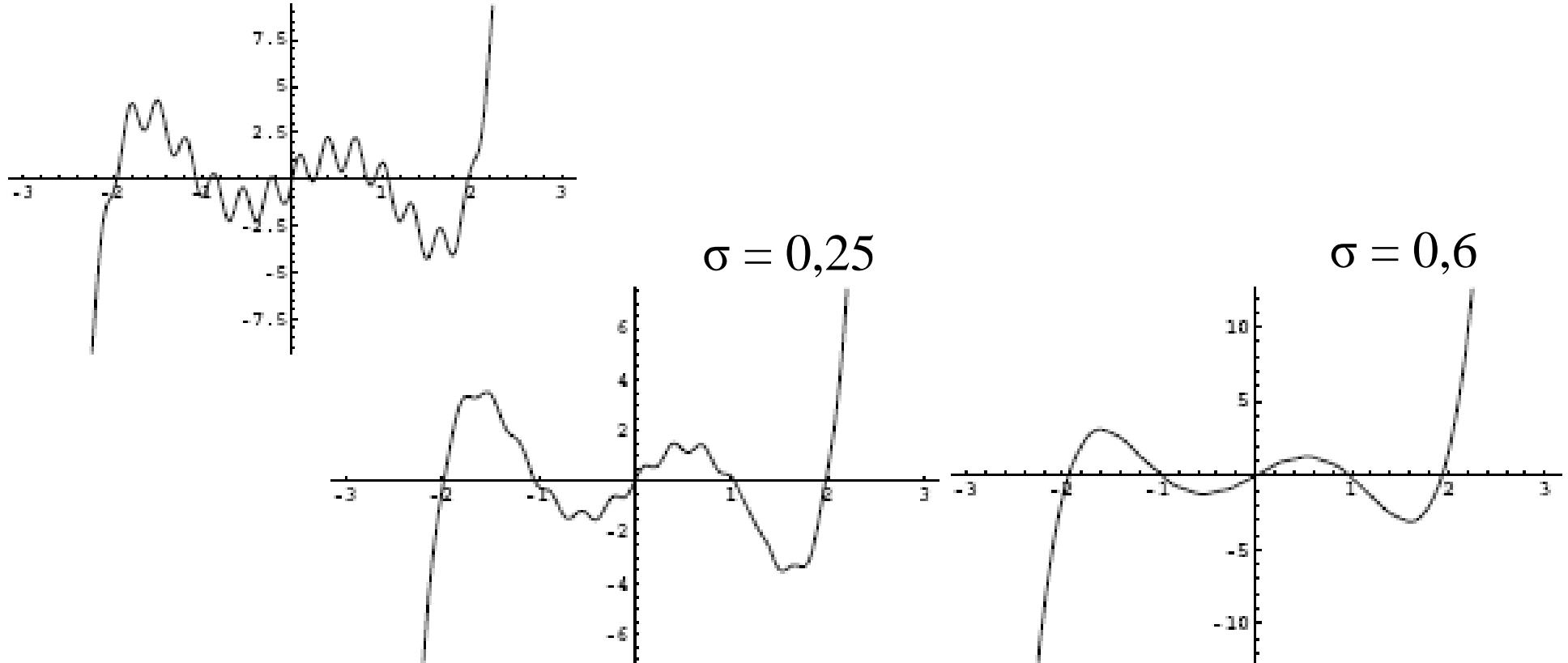


$$g(x) = \begin{cases} 1/\sigma & \text{if } |x| \leq \sigma/2 \\ 0 & \text{if } |x| > \sigma/2 \end{cases} =$$



$$(g * f)(x) = \int_{-\infty}^{\infty} g(a) f(x-a) da = \frac{1}{\sigma} \int_{-\frac{\sigma}{2}}^{\frac{\sigma}{2}} f(x-a) da$$

# Convolutia unei functii unidimensionale



$\sigma$  creste  $\Rightarrow$  functia devine mai neteda, dispar detalii  $\Rightarrow$  filtru trece jos

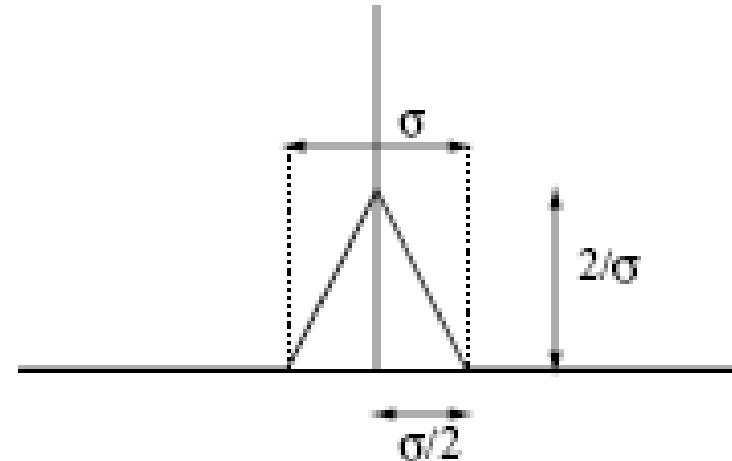
# Nucleul ideal

- $g = \text{nucleul convolutiei}$
- daca  $\sigma$  tinde la zero  $\Rightarrow$  functia Dirac:  
$$\begin{cases} \delta(x) = 0 & \forall x \in \mathbb{R} \setminus 0 \\ \int_{-\infty}^{\infty} \delta(x) dx = 1. \end{cases}$$
- convolutia cu  $\delta$  lasa functia neschimbata

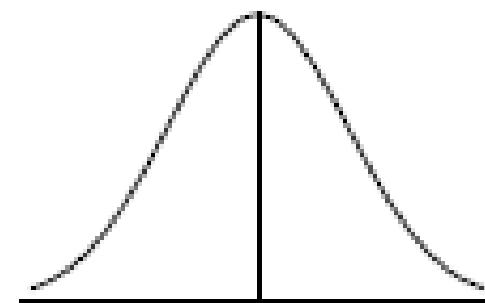
$$(\delta * f)(x) = \lim_{\sigma \rightarrow 0} \frac{1}{\sigma} \int_{-\frac{\sigma}{2}}^{\frac{\sigma}{2}} f(x-a) da = f(x)$$

# Nuclee de convolutie 1D

$$\begin{cases} \frac{2}{\sigma} - \frac{4x^2}{\sigma^2|x|} & \text{if } |x| \leq \sigma/2 \\ 0 & \text{if } |x| > \sigma/2 \end{cases}$$



$$\frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x}{\sigma}\right)^2}$$



# Nuclee de convolutie 2D

- functia dreptunghiulara

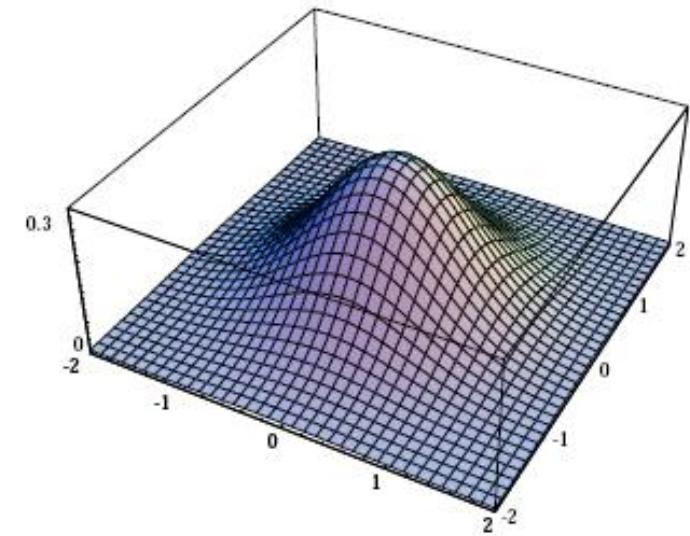
$$g(x, y) = \begin{cases} 1/\sigma^2 & \text{if } |x| \leq \sigma/2 \text{ and } |y| \leq \sigma/2 \\ 0 & \text{otherwise} \end{cases}$$

- functia lui Gauss

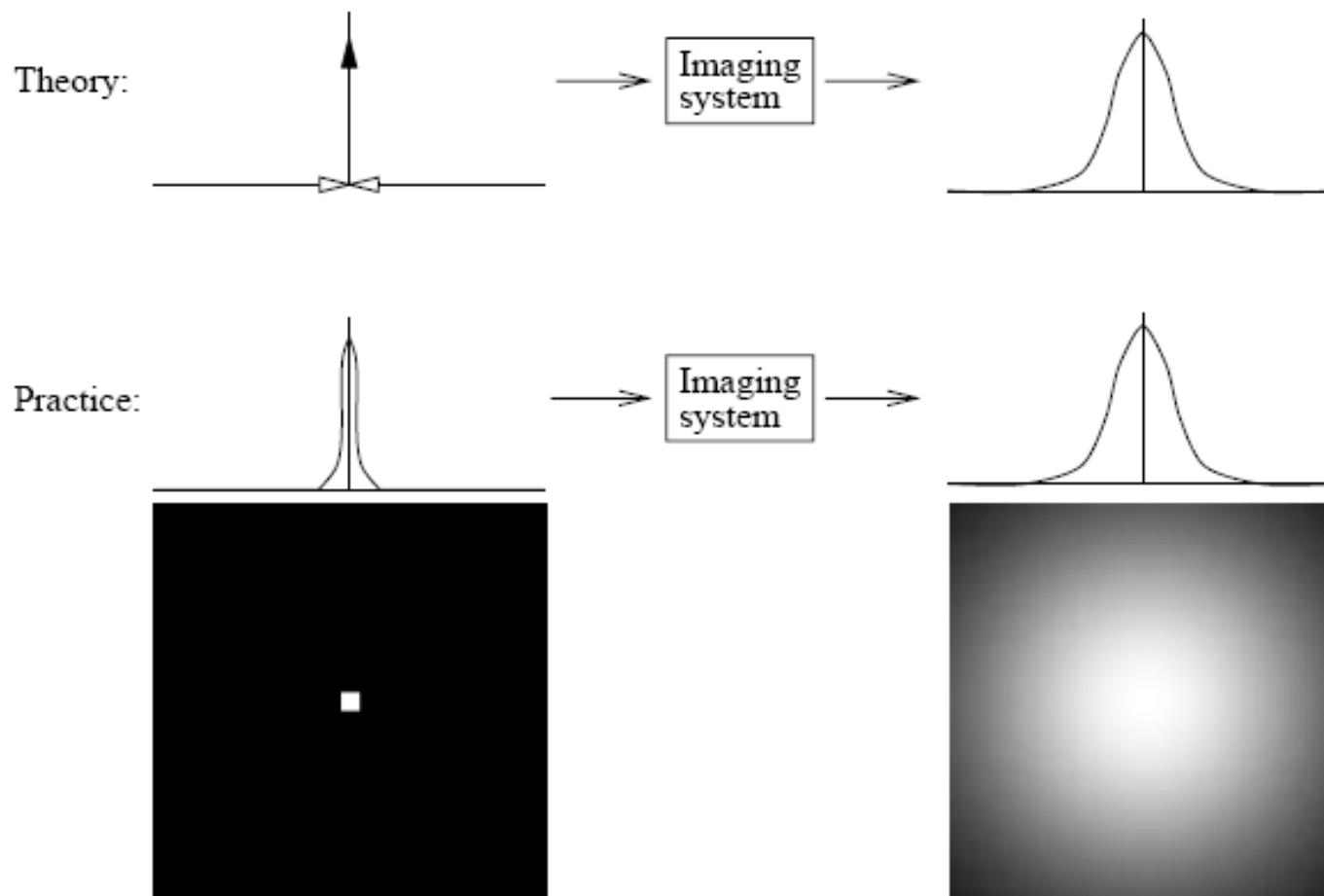
$$g(x, y) = \frac{1}{\sigma^2 2\pi} e^{-\frac{1}{2} \frac{x^2 + y^2}{\sigma^2}}$$

Volumul de sub nucleu este 1

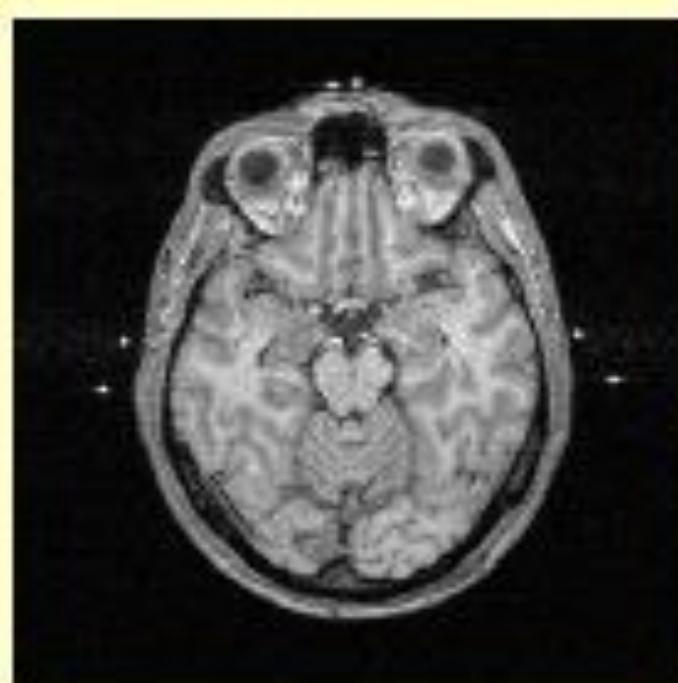
$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) dx dy = 1$$



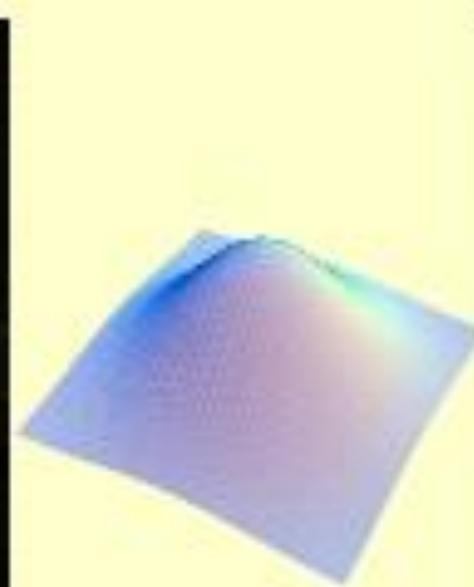
# Functia de distributie a punctului



# Exemplu

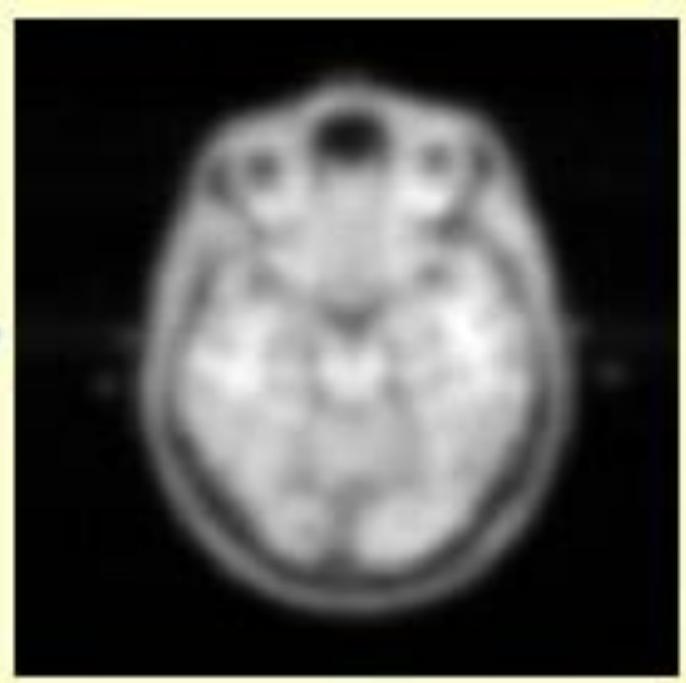


Original scene



PSF

$\approx$   
Convolution kernel



Acquired image

$$f$$

\*

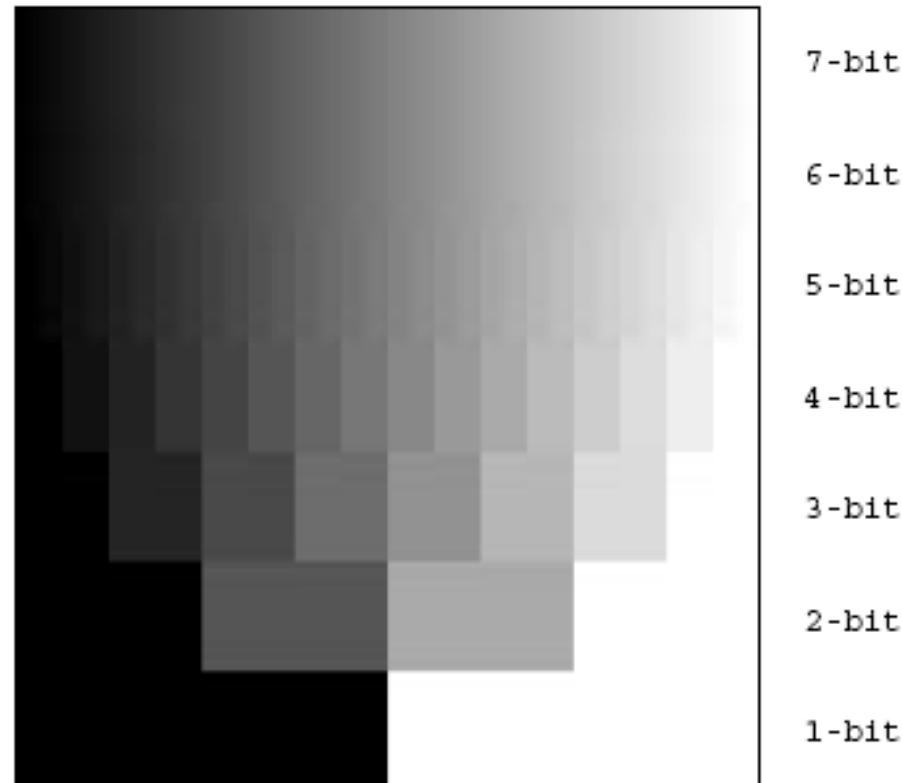
$$g$$

=



# Cuantificare

- val. de gri sunt cuantificate la un numar finit de valori de gri
- $n$  biti  $\Rightarrow 2^n$  nuante de gri
- pe 8 biti  $\Rightarrow 0 \div 255$
- pe 32 biti  $\Rightarrow 0 \div 2^{32}-1$
- intregi, intregi fara semn sau nr.reale (float)

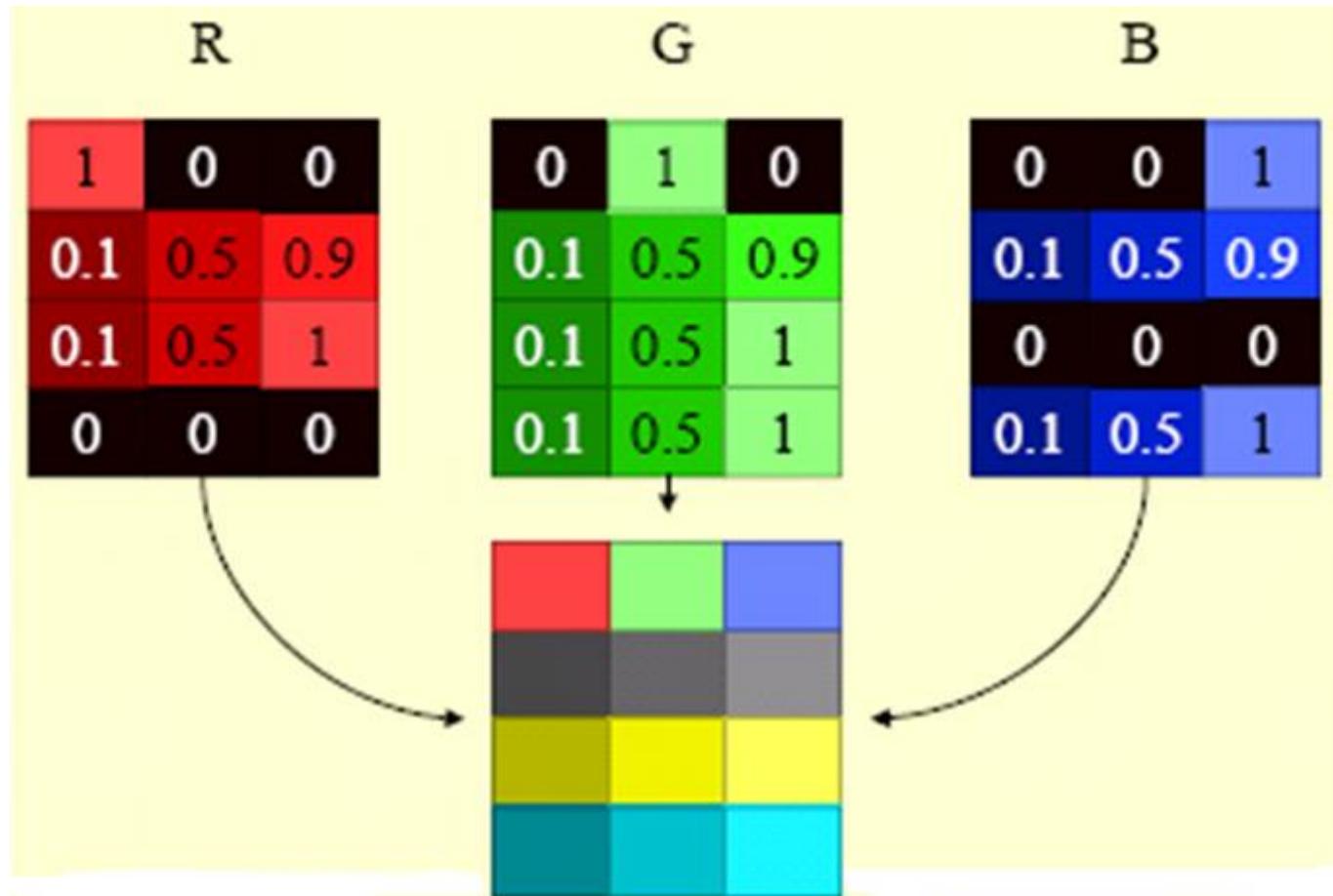


# Modelul color

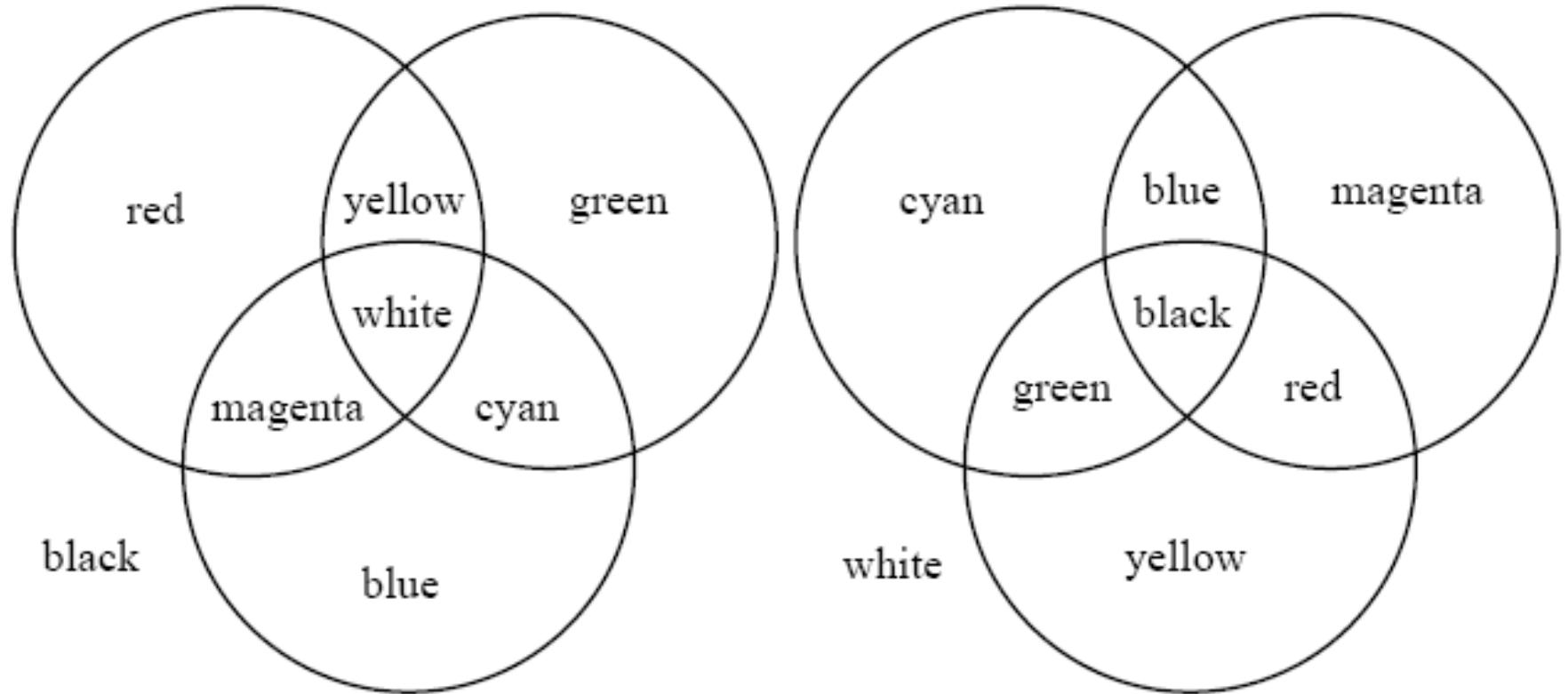
- modelul RGB (aditiv, emisiv)
- modelul CMY (substractiv, reflectant)
- modelul CMYK
- modelul HSV
- etc.

# Modelul RGB

Red (700 nm), Green (546,1 nm), Blue (435,8 nm)

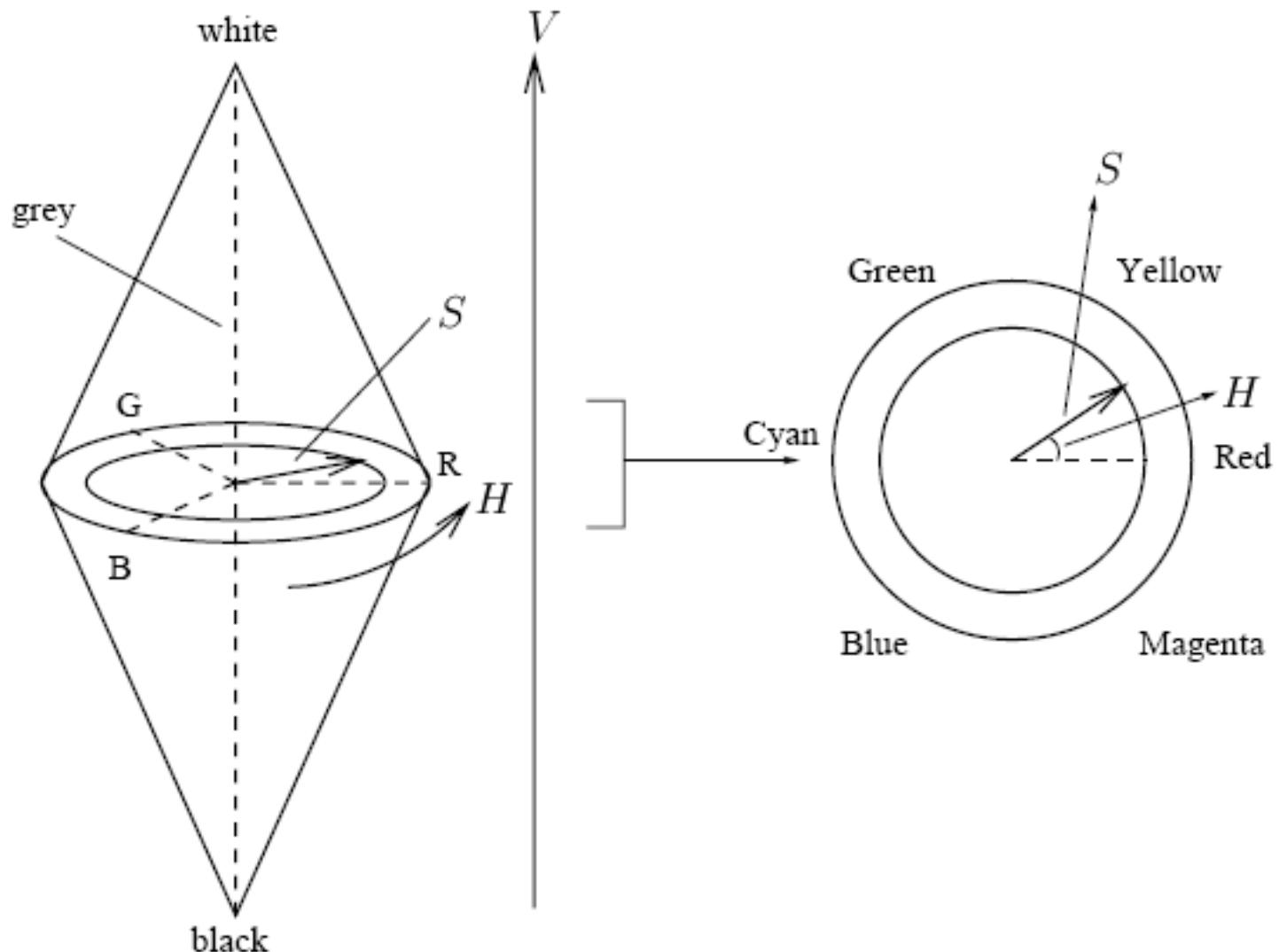


# Modelul RGB si CMY



RGB (aditiv, emisiv) CMY (subtractiv, reflectant)

# Modelul HSV



# Modelul HSV

