

Basic principles of medical imaging

Image elements

Image orientation

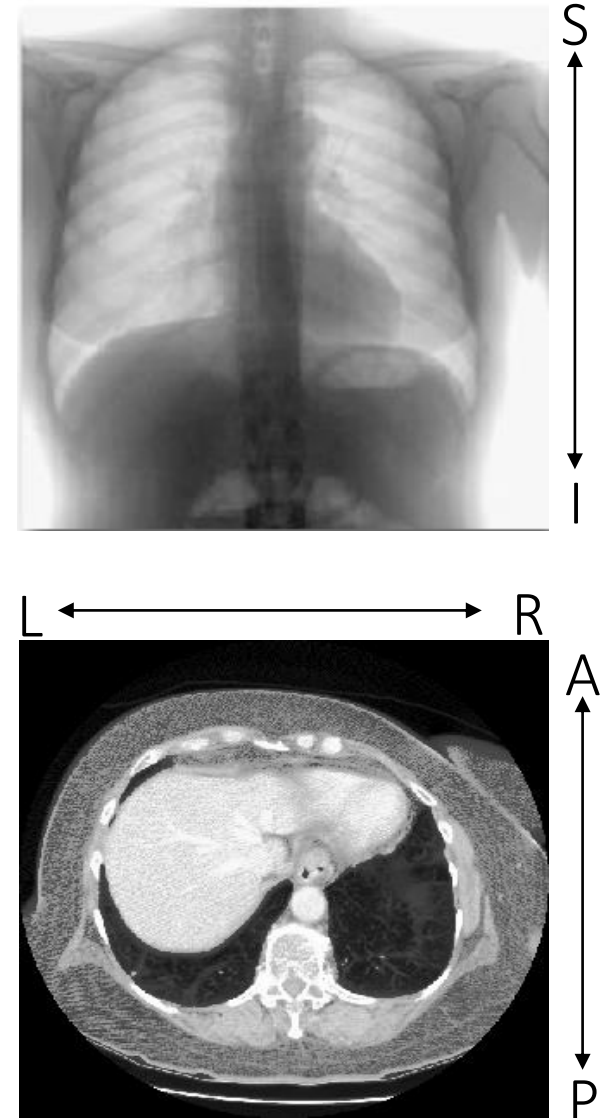
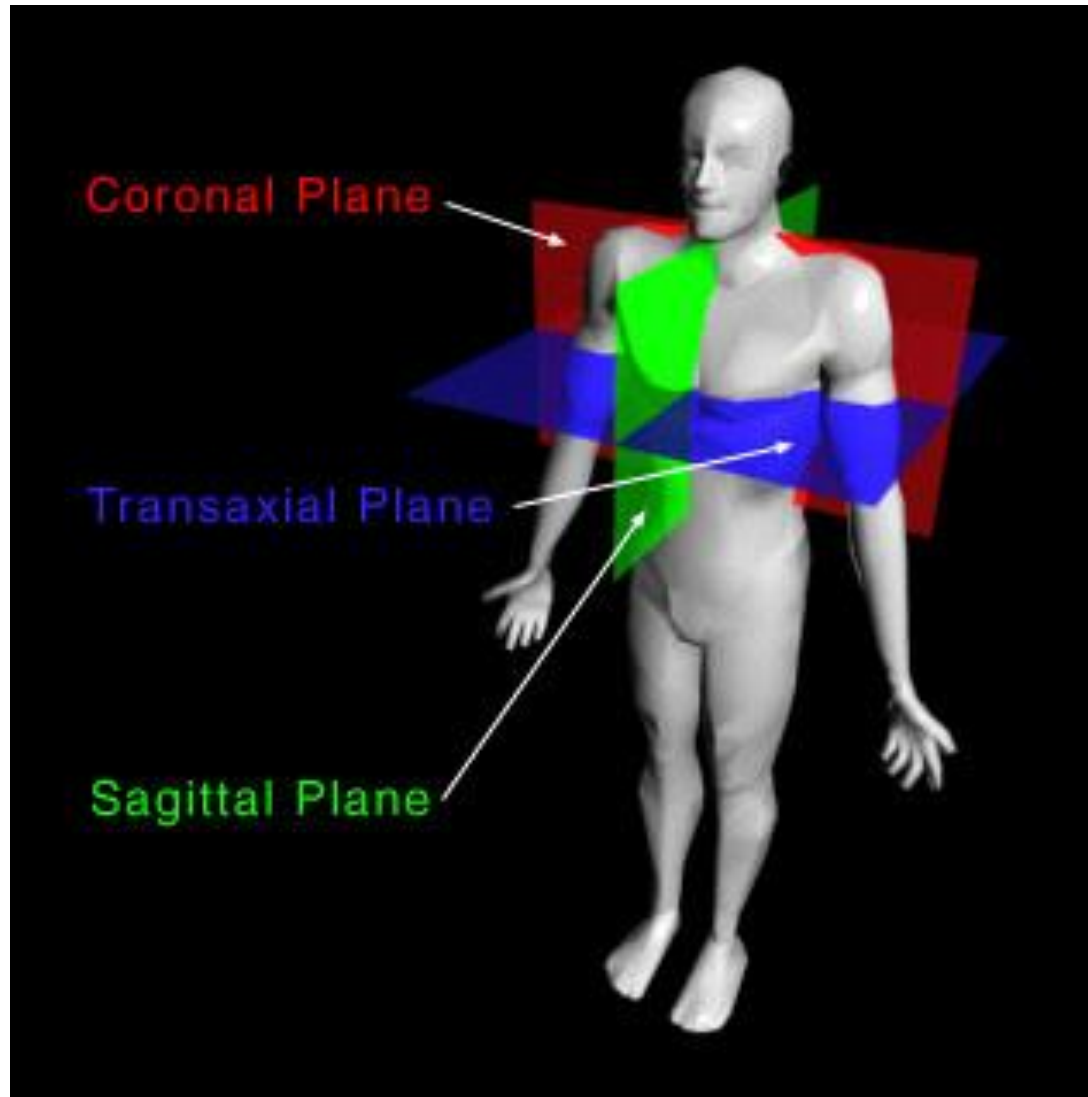


Image elements

Image volume, slice, voxel and field-of-view (FOV)

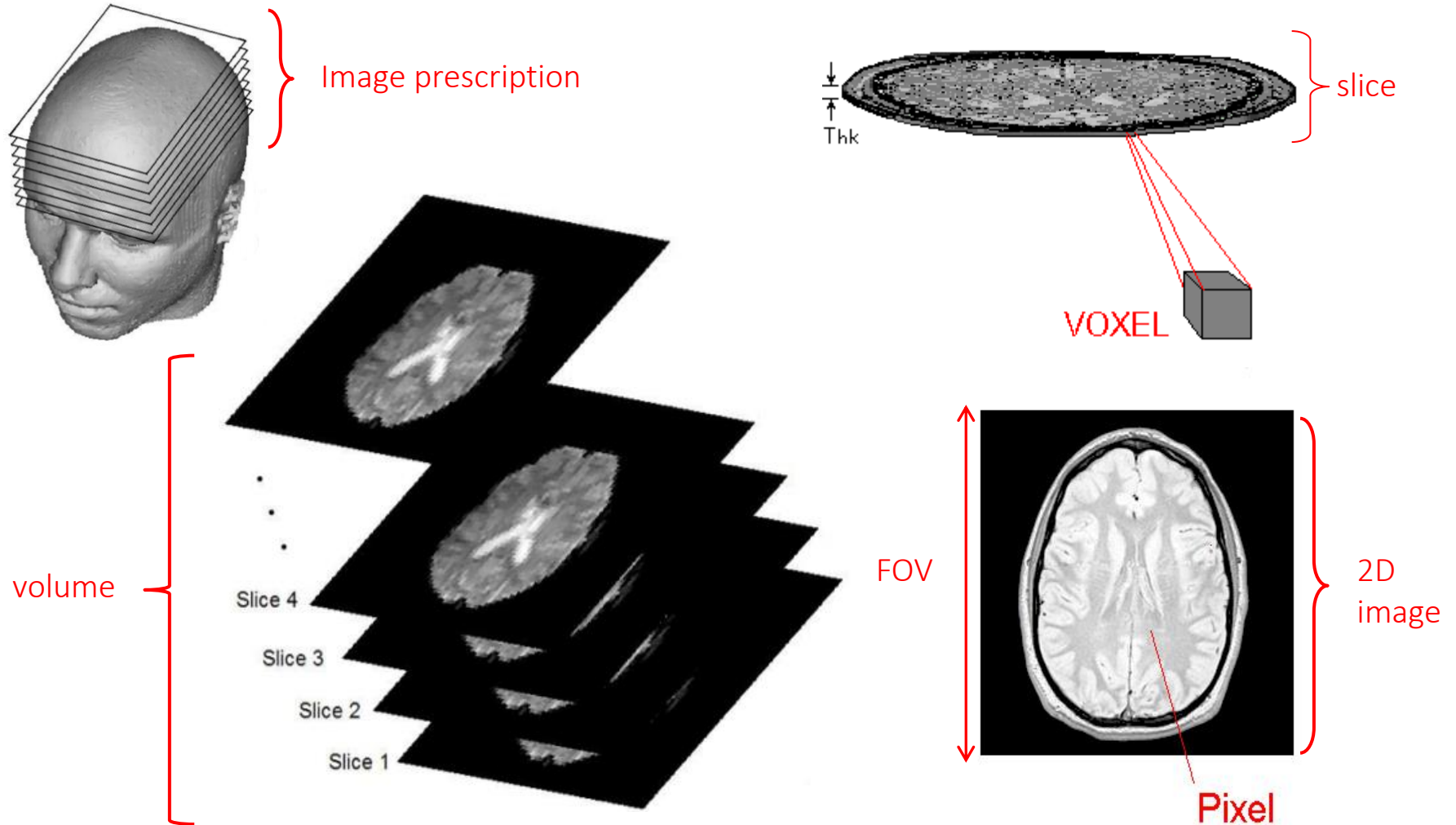


Image elements

Image contrast

Image contrast = difference in image intensity that makes an object distinguishable from another or from the background.

Image contrast depends on the tissue characteristics as well as the contrast sensitivity of the imaging system used.

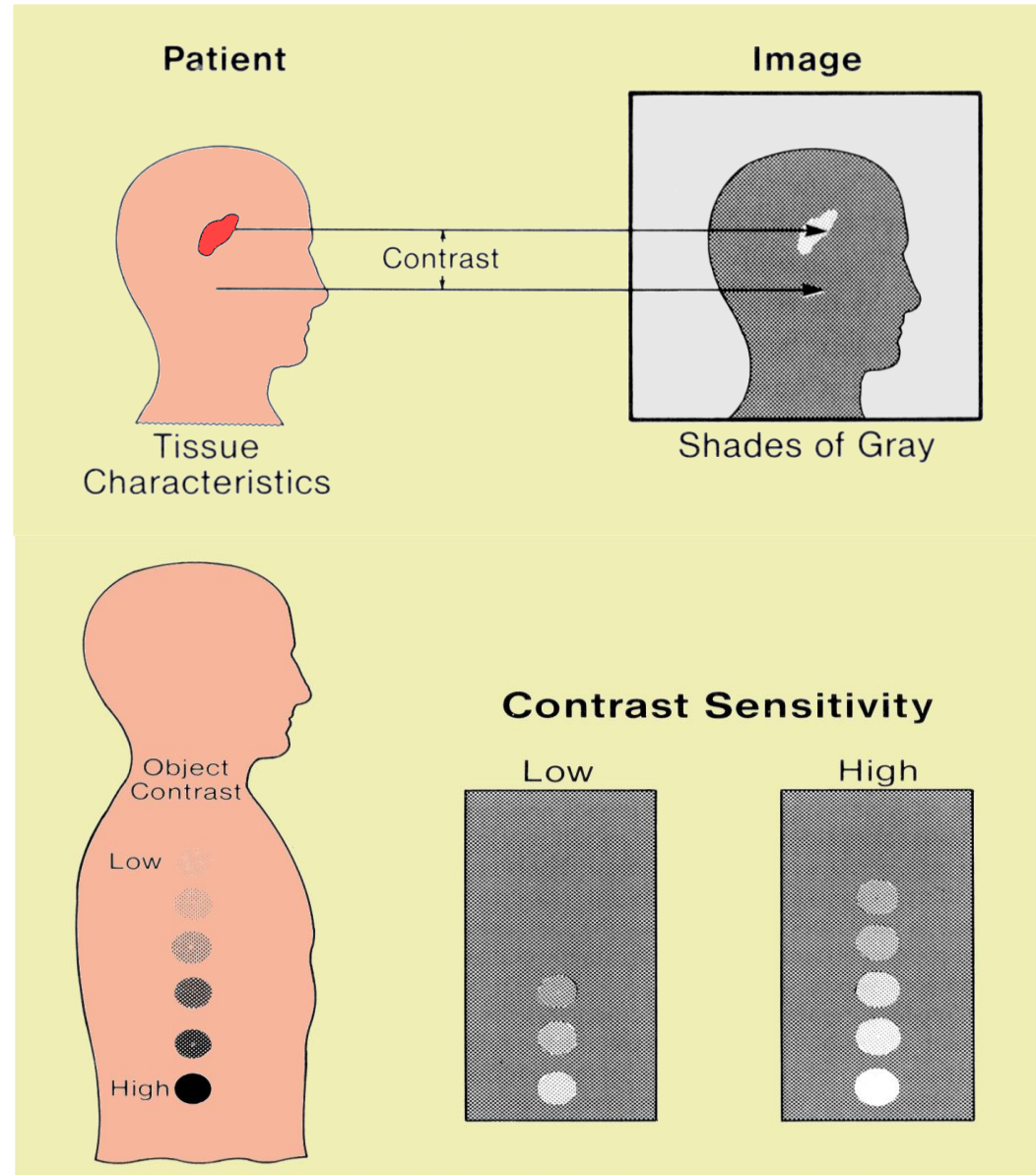
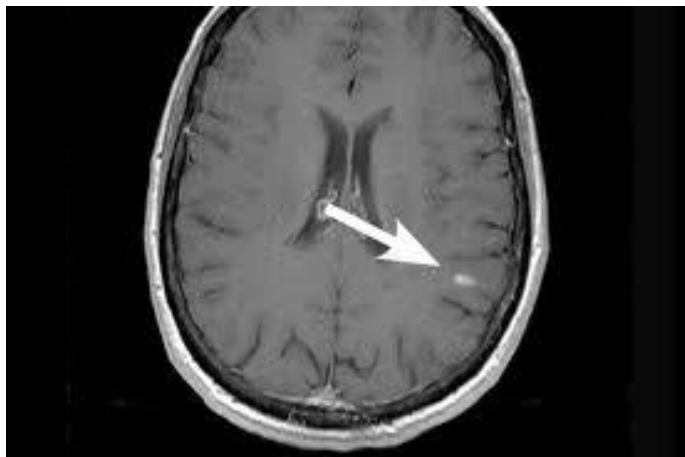
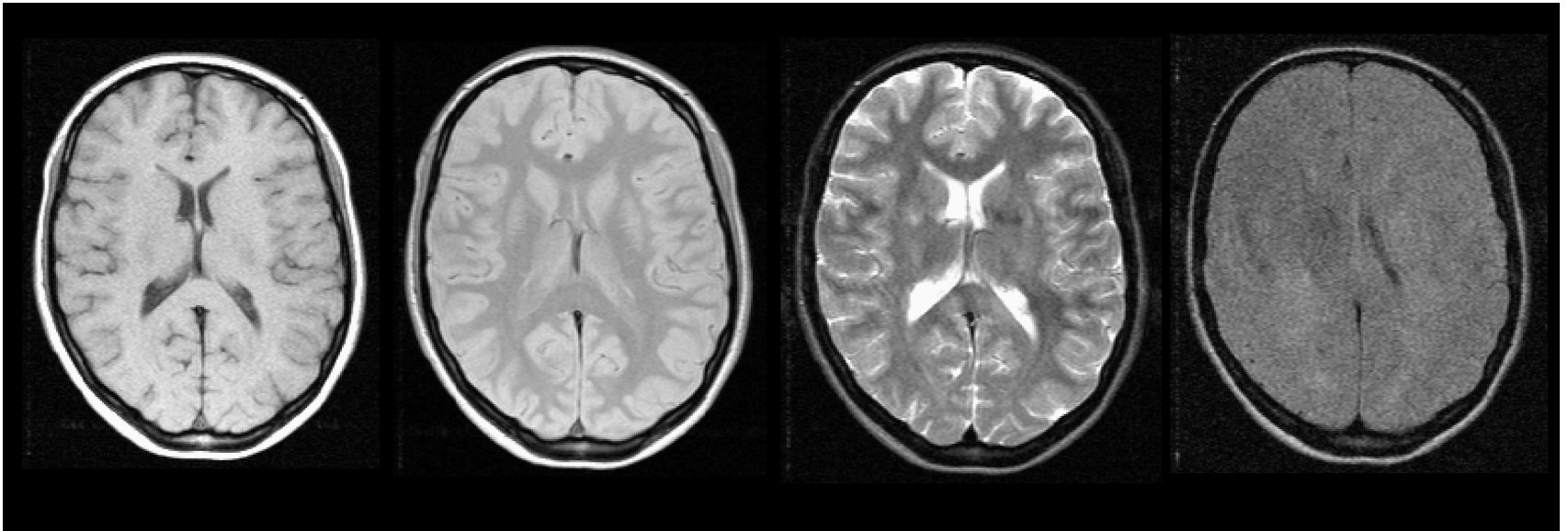


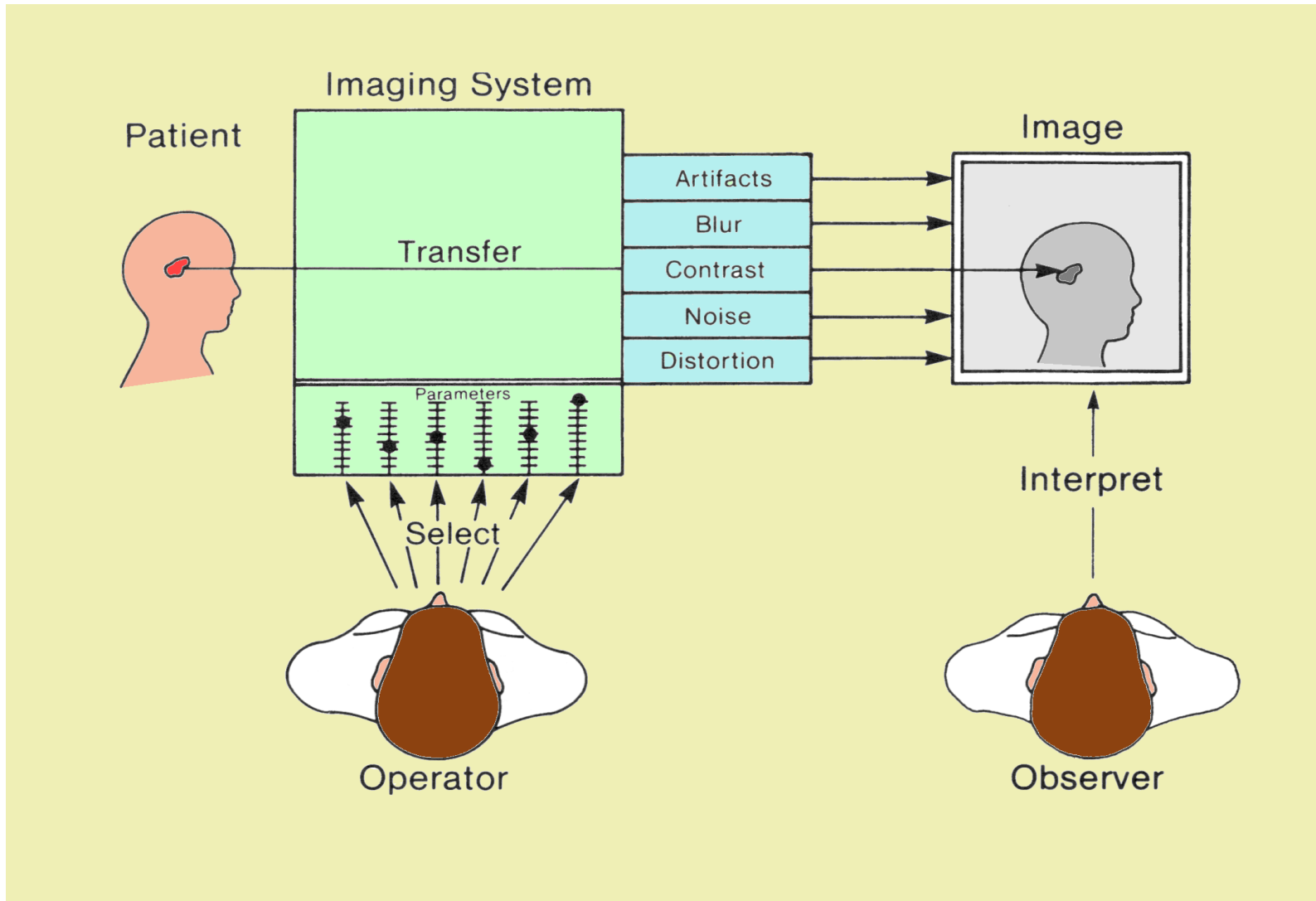
Image elements

Image contrast



Imaging principles

The imaging process

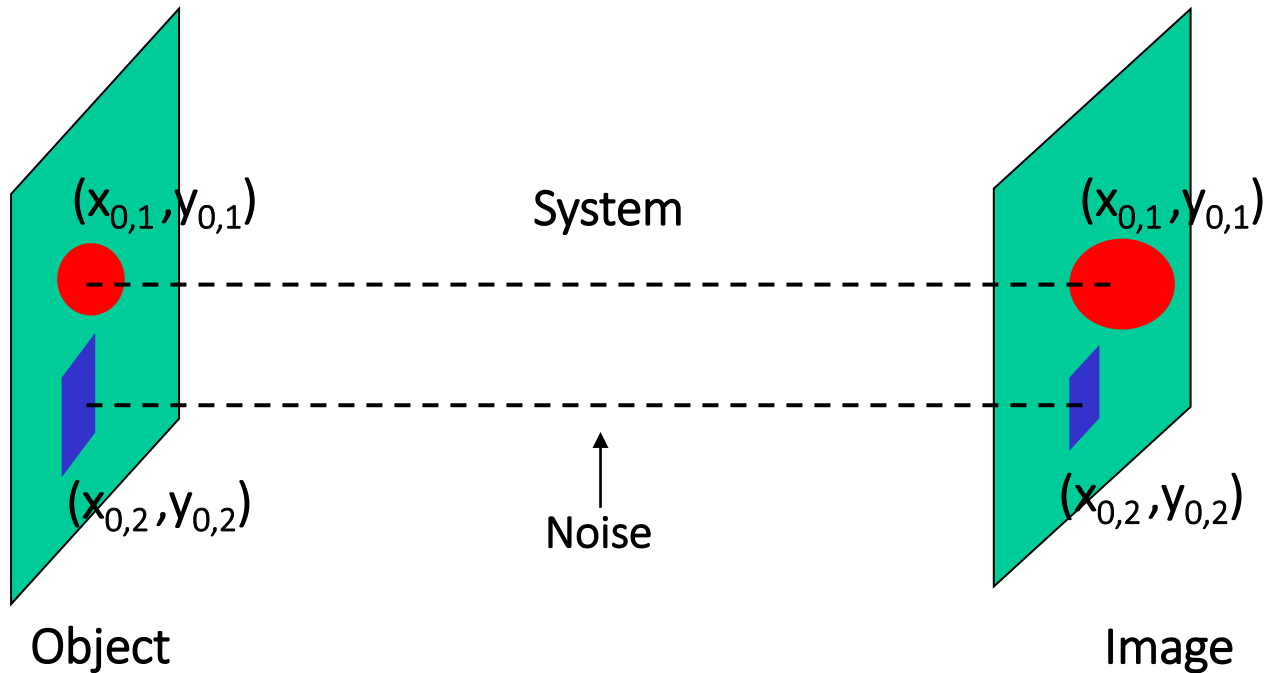


Imaging principles

In general:

$$I(x, y) = S\{O(x, y)\} + N(x, y)$$

$$I(x, y) = S\left\{ \iint_{-\infty \rightarrow +\infty} O(x_0, y_0) \delta(x - x_0, y - y_0) dx_0 dy_0 \right\} + N(x, y)$$



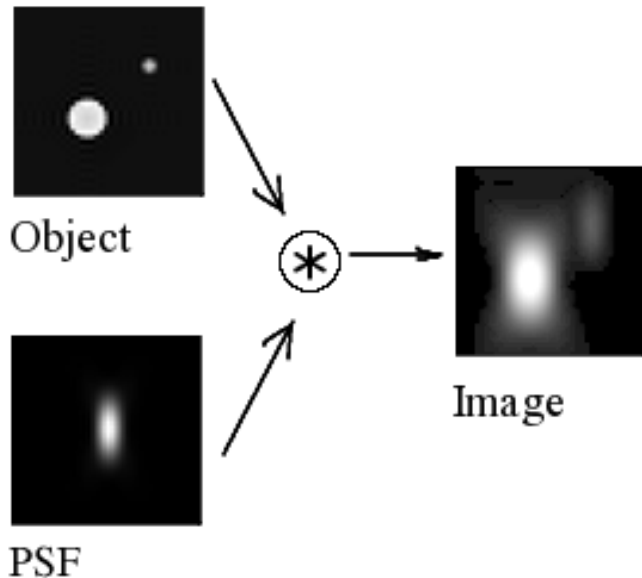
Imaging principles

For a spatially invariant system :

$$I(x, y) = \iint_{-\infty \rightarrow +\infty} O(x_0, y_0) PSF(x - x_0, y - y_0) dx_0 dy_0 + N(x, y)$$

$$I(x, y) = O(x, y) \otimes PSF(x, y) + Noise$$

PSF = Point Spread Function



Determinant factors (depend on modality):

$$h_{total} = h_{sensor} \otimes h_{sampling} \otimes h_{reconstruction} \otimes h_{filtering}$$

Ideal PSF:

$$h(x, y, z) = \delta(x, y, z) \Rightarrow I = O, \delta = \text{Dirac's delta}$$

Most common PSF shapes: Gaussian, sinc, ...

Image properties

Spatial resolution

R = smallest distance between two point sources for which the sources can be resolved.

R is related with the system point spread function (PSF):

A common shape for the PSF is a Gaussian - in 1D:
(in this case, $R \sim \text{FWHM}$ in each direction)

The total imaging PSF results from several contributions:

... and so does the final image resolution:

$$I(x, y, z) = \text{PSF}(x, y, z) \otimes O(x, y, z)$$

$$h(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{\sigma^2}\right),$$

$$\text{FWHM} = 2\sqrt{2\ln(2\sigma)} \cong 2.36\sigma$$

$$\text{PSF}_{\text{total}} = \text{PSF}_1 \otimes \dots \otimes \text{PSF}_N$$

$$R_{\text{final}} = \sqrt{R_1^2 + \dots + R_N^2}$$

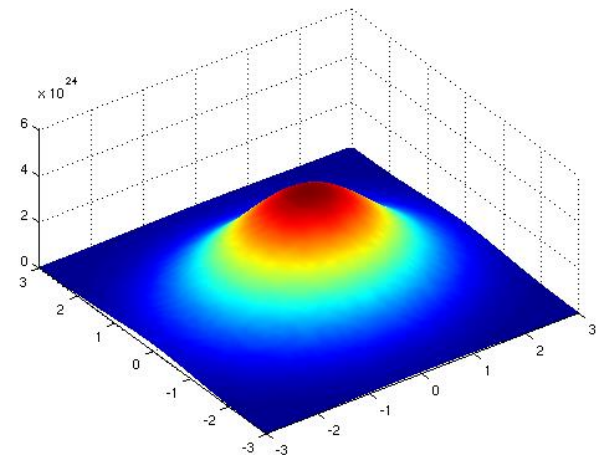
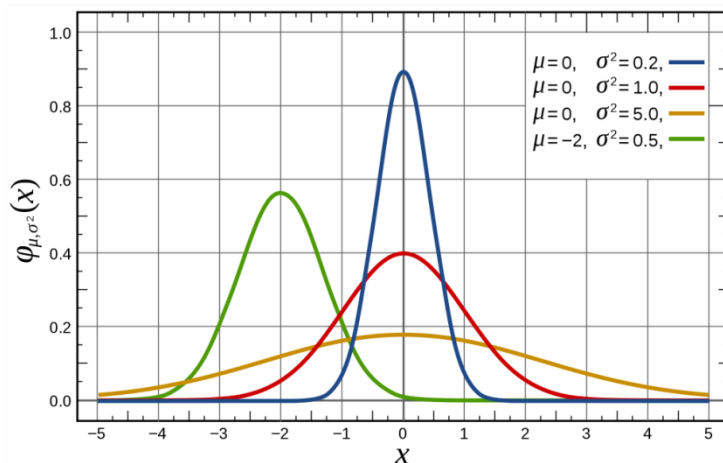


Image properties

Spatial resolution

Image resolution \sim pixel size

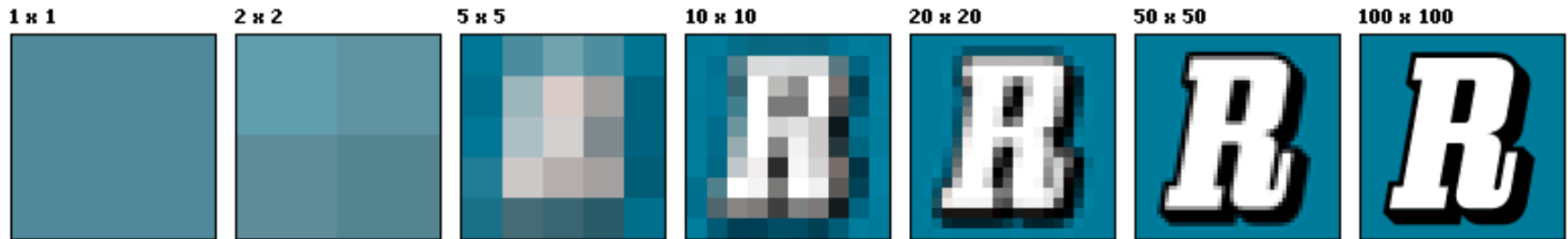


Image resolution = smallest distance between two point sources for which the sources can be resolved.

Image resolution depends on the imaging system and parameters used.

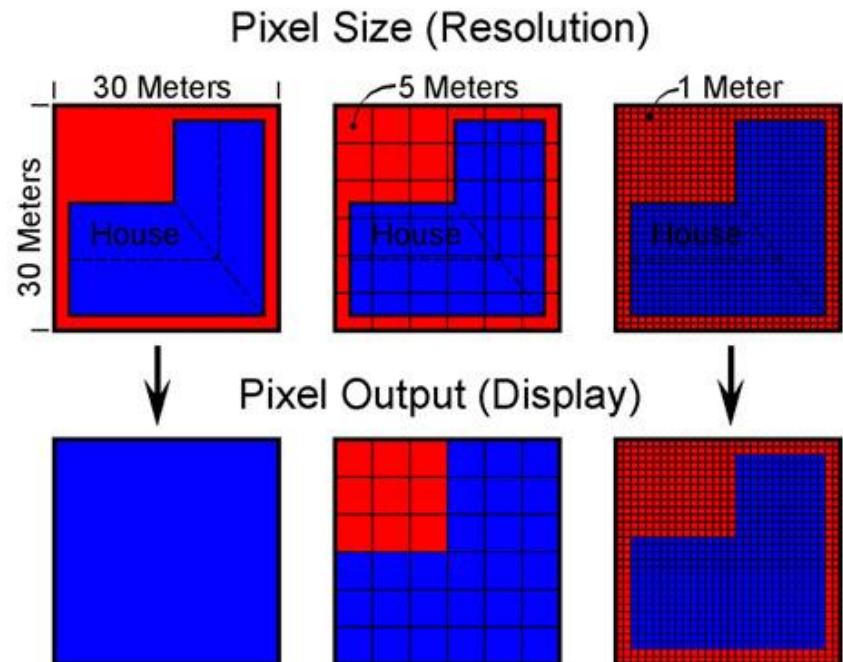


Image properties

Spatial resolution



Image properties

Signal-to-noise-ratio (SNR):

SNR ~ compares the level of a desired signal to the level of background noise.

“Signal processing” definition: $\text{SNR} = \frac{P_{\text{signal}}}{P_{\text{noise}}}$, (power of signal over power of noise)

“Medical imaging” definition: $\text{SNR} = \frac{\mu}{\sigma}$ (mean signal over SD of noise)

Image SNR depends on
the available object signal and
on the imaging system noise
characteristics.

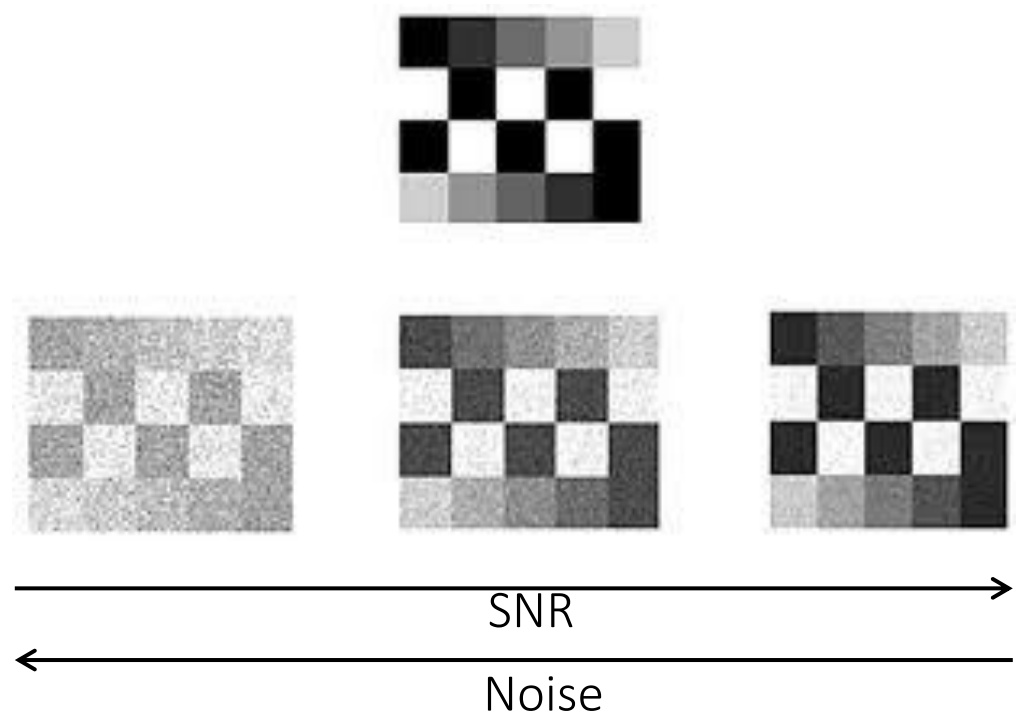


Image properties

Signal-to-noise-ratio (SNR):

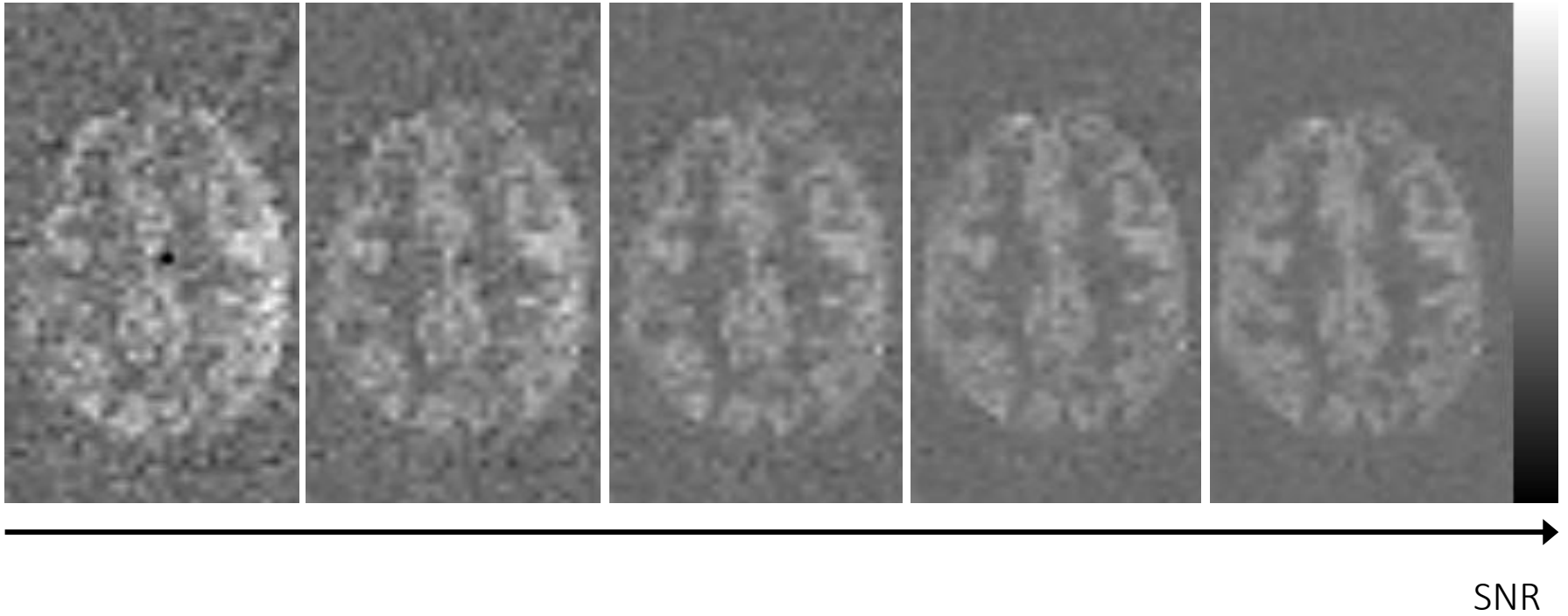


Image properties

Contrast-to-noise ratio (CNR):

CNR ~ compares the image contrast between regions-of-interest A and B to the level of background noise.

$$C = \frac{|S_A - S_B|}{\sigma_o}$$

Image CNR depends on
the tissue contrast,
the object size and the
contrast sensitivity, image
resolution and SNR of the
imaging system.

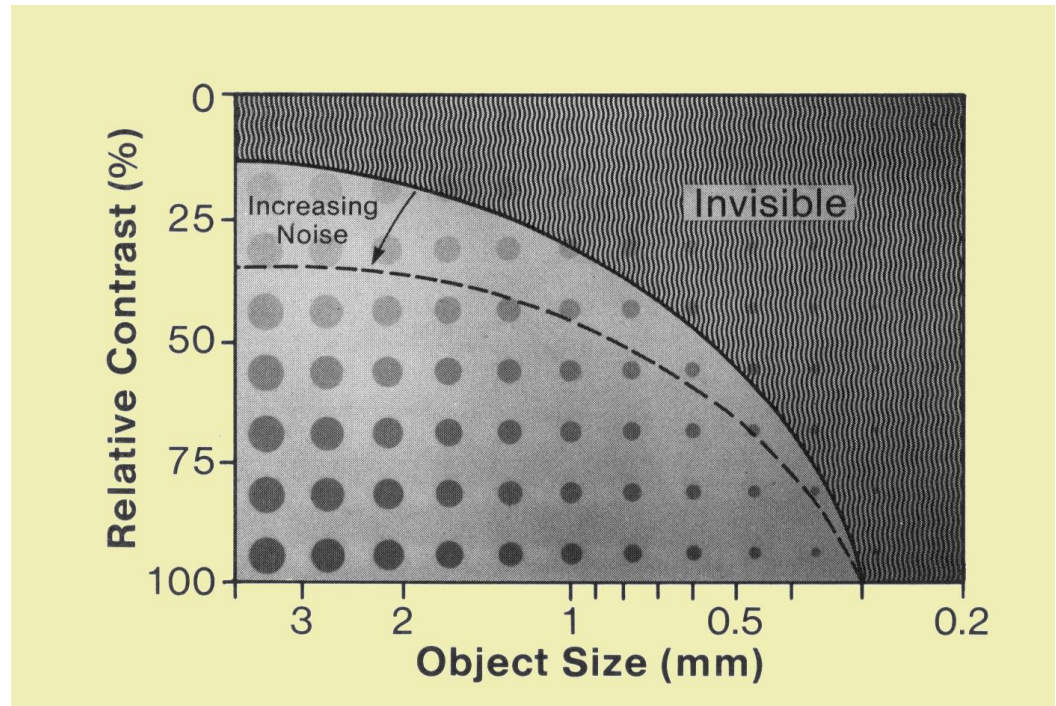


Image properties

Main image characteristics:

Spatial resolution = minimum point separation (R):	PSF
Signal-to-noise-ratio (SNR):	Object, Noise
Contrast-to-noise ratio (CNR):	Object, Noise, PSF

CT, PET, SPECT:

$$SNR = N/\sigma \propto \sqrt{N}$$

N = number of photons X ou γ ,
following a Poisson distribution,
 σ = standard deviation

MRI:

$$SNR = \hat{s}/\sigma_{\xi}, SNR_N \propto SNR\sqrt{N}$$

s = mean signal;
 σ_{ξ} = standard deviation of the noise
 N = number of repetitions