

Image properties

Quantitative and Functional Imaging

BME 4420/7450

Fall 2022

How are these images different?

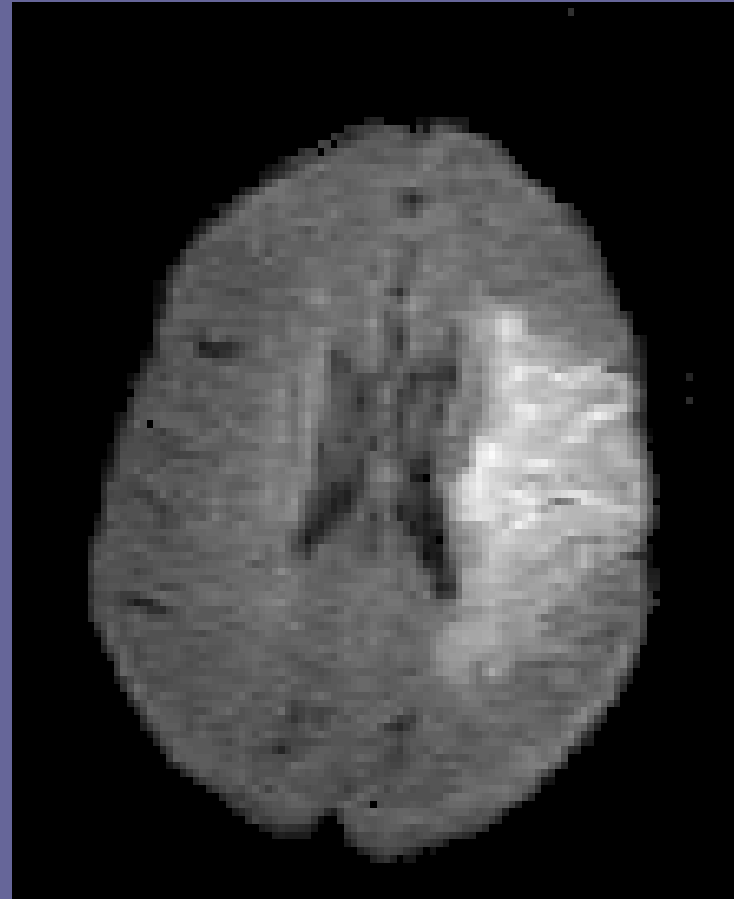
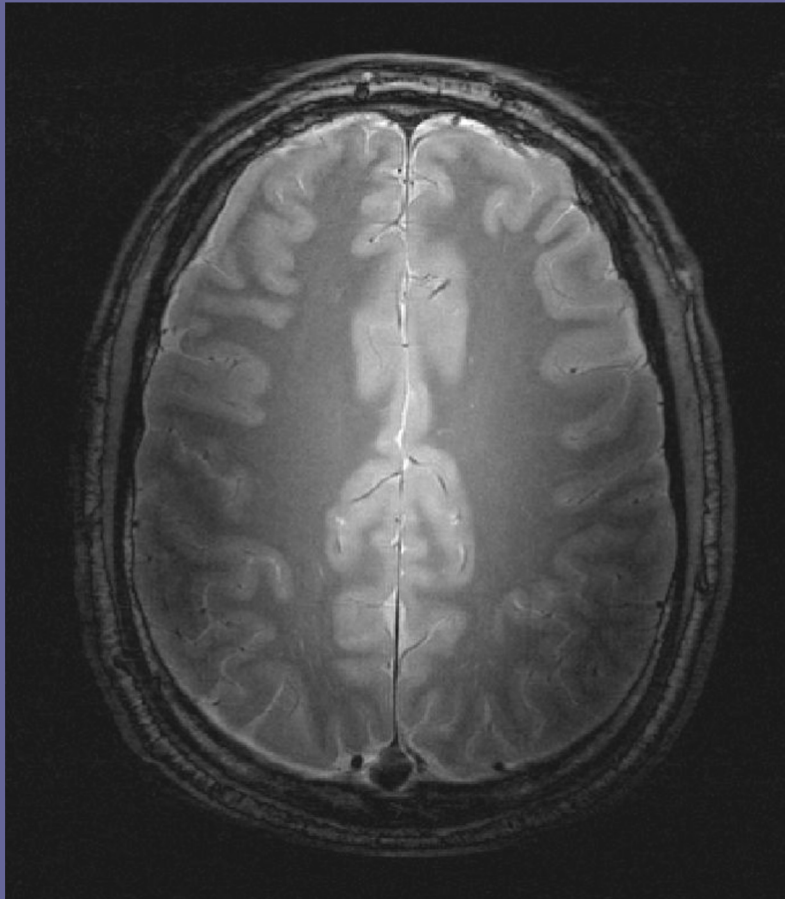


Image information

- An image represents the *spatial distribution of physical properties* in an object
 - Imperfect representation
- How do images differ from the true distributions they map?
- Let's look at some examples...

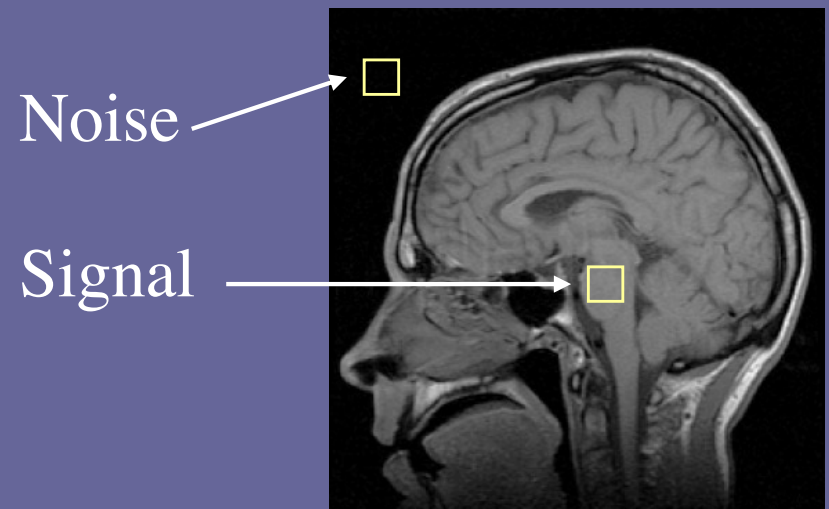
What is the difference?



Signal to noise ratio (SNR)

- Measures the relative intensity of
 - Signal
 - Noise
- Independent of intensity scaling
- A standard measure of sensitivity

$$SNR = \frac{\text{mean}(\text{Signal})}{\text{std}(\text{Noise})}$$

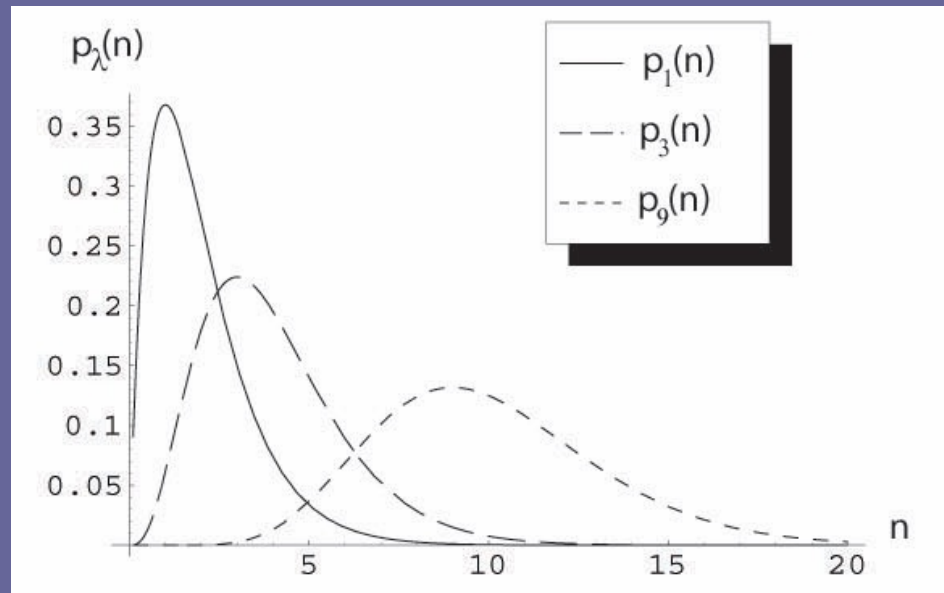


Models of noise

- Poisson noise
 - Counting independent events
 - CT, PET

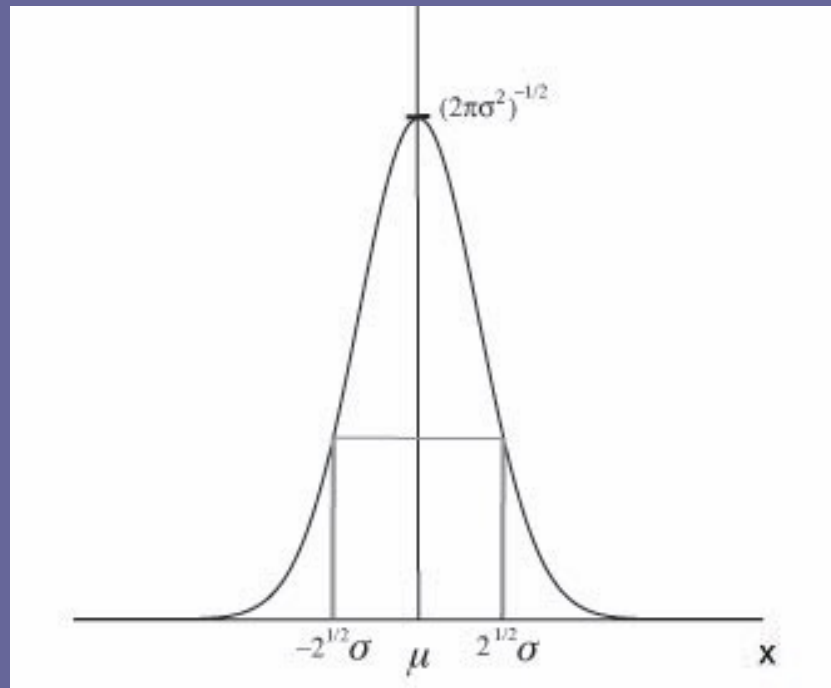
$$p_{\lambda}(n) = \frac{\lambda^n}{n!} e^{-\lambda}$$

λ = mean count



- Gaussian noise
 - 'Normal' distribution
 - Sum of many independent noise sources
 - MRI

$$p(\mathbf{x}) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-(\mathbf{x}-\mu)^2/2\sigma^2}$$



μ = mean value
 σ = std deviation

What's the difference?

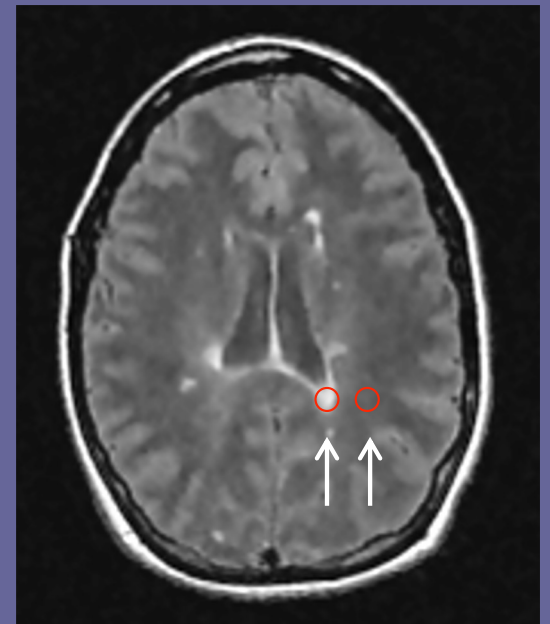


Contrast

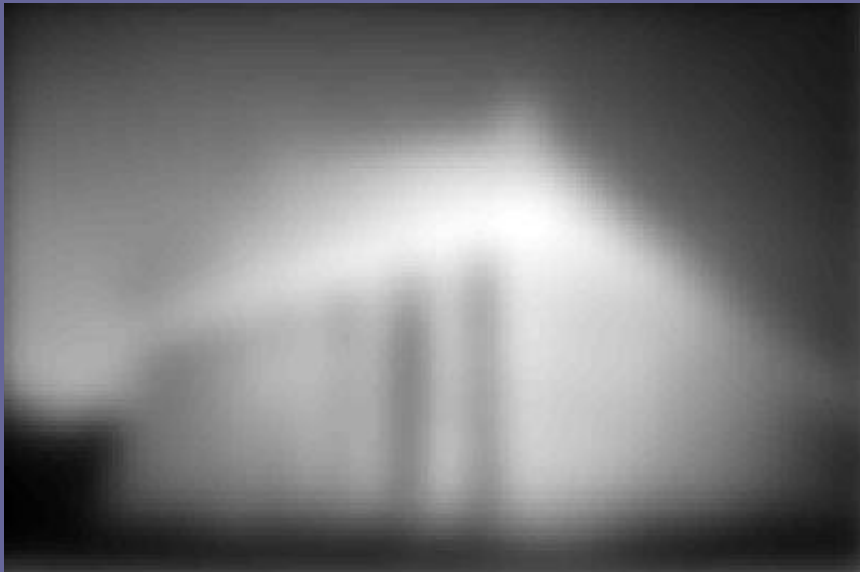
- Image intensity difference (between regions)
- Allows regions to be distinguished
- Helps determine detection limits

$$\begin{aligned}\text{Contrast} &= \Delta S \\ &= S(\text{lesion}) - S(\text{white matter})\end{aligned}$$

$$CNR = \frac{\Delta S}{\sigma} \quad \text{Contrast to noise ratio}$$



What is the difference?

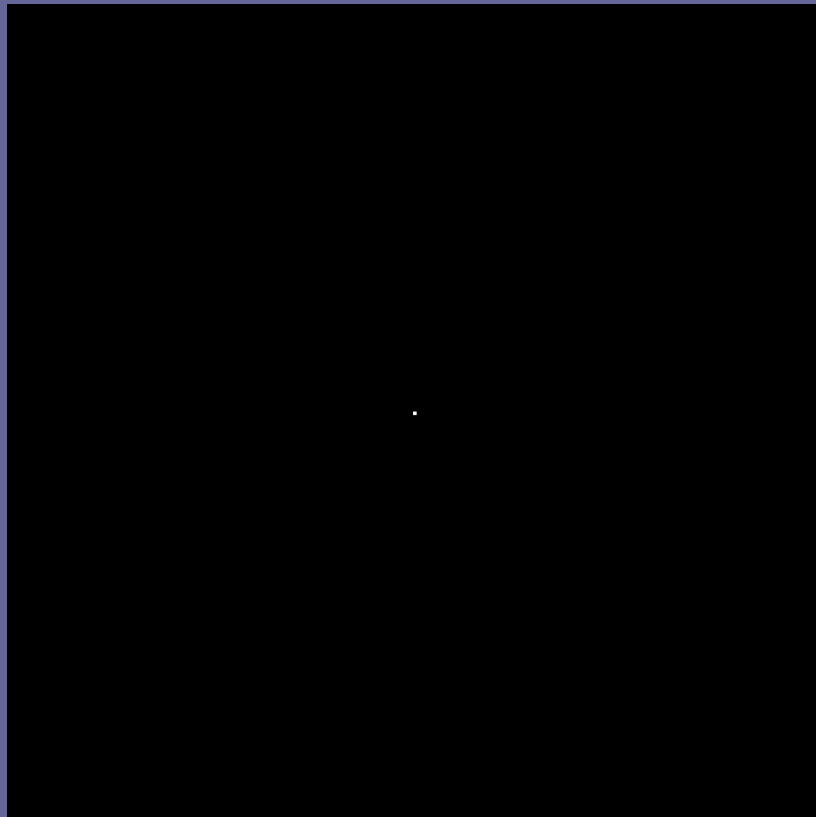


Resolution

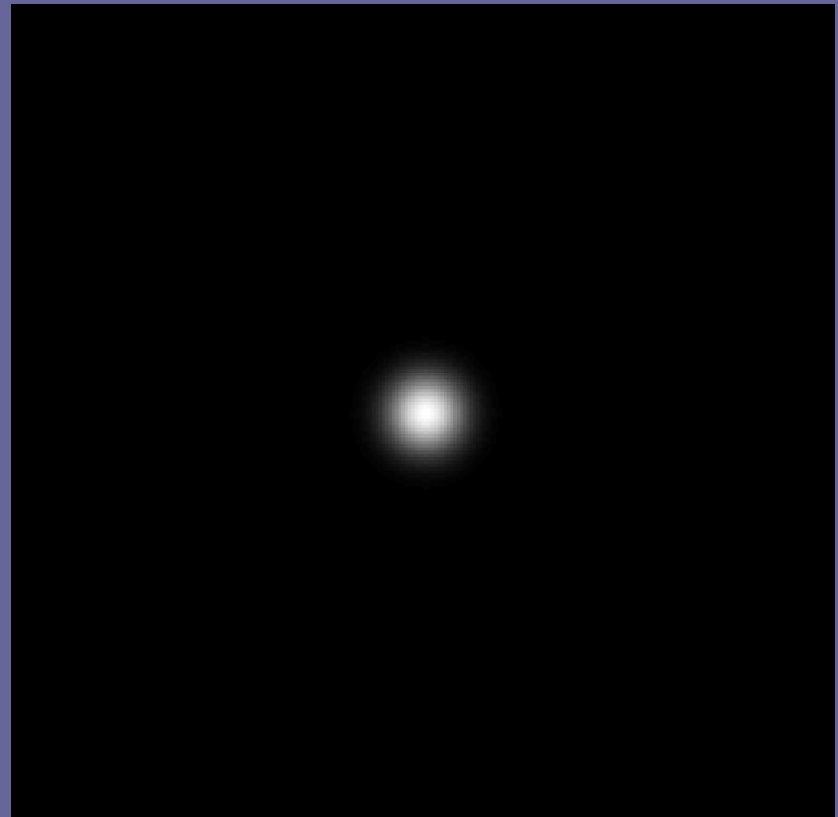
- Determines the smallest detectable feature in an image
- Not necessarily the pixel size
- May be different in x, y (and z)
- Characterized by the width of the *point spread function* (PSF)

Images are blurred representations

Ideal image



Blurred image



Convolution and image blurring

- The convolution of two functions, $f(x)$ and $h(x)$, is defined as

$$\begin{aligned} g(x) &\equiv \int_{-\infty}^{\infty} f(x') h(x - x') dx' \\ &= f(x) * h(x) \end{aligned}$$

- Notice that the convolution is *commutative*: making the change of variables $x' = x - x''$,

$$\int_{-\infty}^{\infty} f(x') h(x - x') dx' = - \int_{\infty}^{-\infty} f(x - x'') h(x'') dx''$$

- Notice that the convolution is *commutative*: making the change of variables $x' = x - x''$,

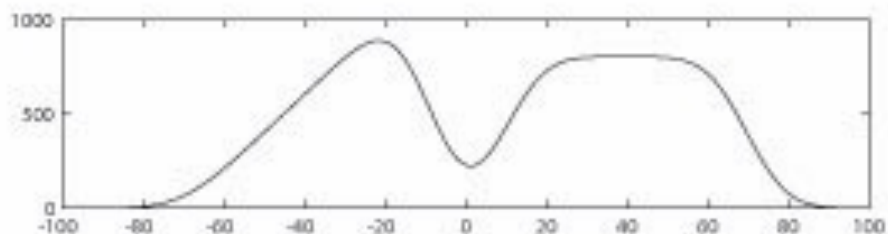
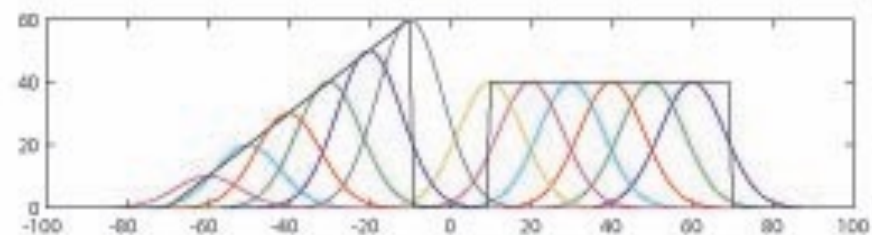
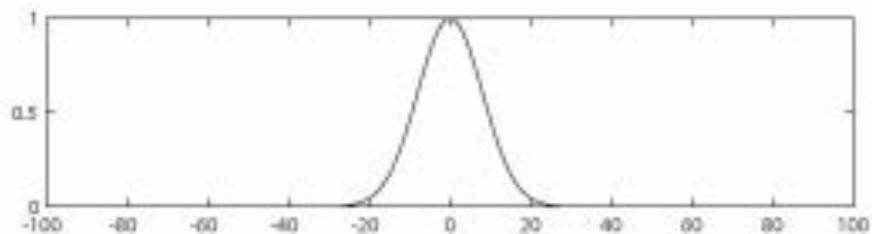
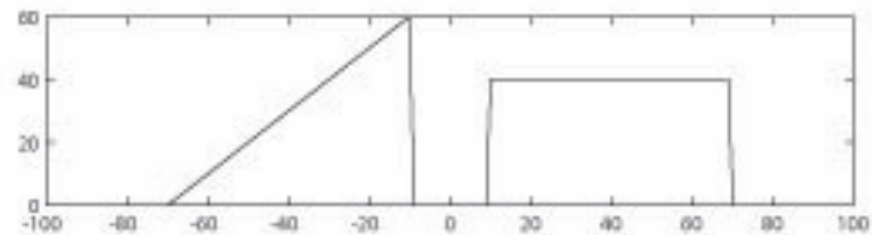
$$\begin{aligned}\int_{-\infty}^{\infty} f(x')h(x - x')dx' &= -\int_{\infty}^{-\infty} f(x - x'')h(x'')dx'' \\ &= \int_{-\infty}^{\infty} h(x'')f(x - x'')dx''\end{aligned}$$

so

$$f(x) * h(x) = h(x) * f(x)$$

and the order doesn't matter.

- In words, the convolution is the sum of many weighted and shifted copies of $h(x)$.



$$f(x)$$

$$h(x)$$

$$\int_{-\infty}^{\infty} f(x') h(x - x') dx'$$

$$= f(x) * h(x)$$

Describing a point source

- We can describe an idealized point source using the *Dirac delta* function, $\delta(x)$
- The delta function is defined by

$$\delta(x - a) = \begin{cases} 0, & \text{for } x \neq a \\ \infty, & \text{for } x = a \end{cases}$$

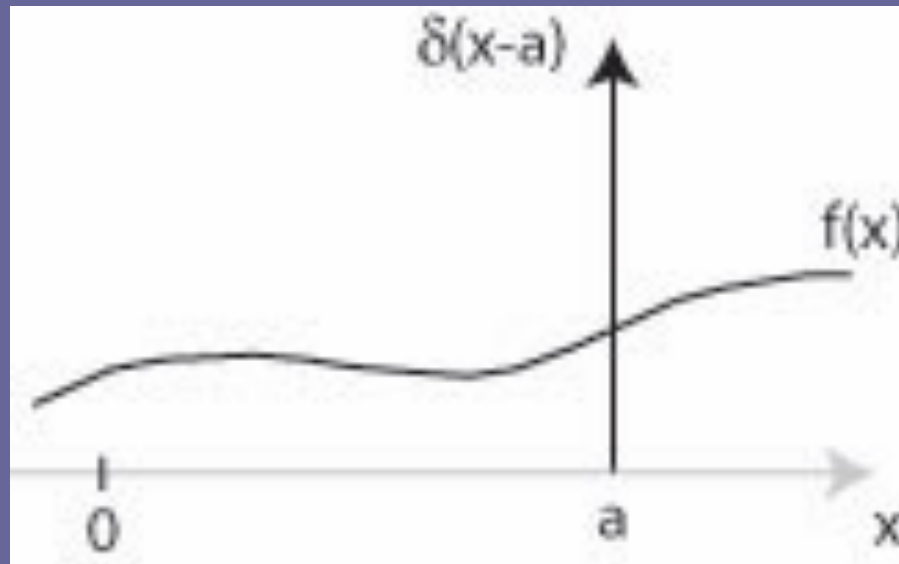
and

$$\int_{-\infty}^{\infty} \delta(x - a) dx = 1$$

so the delta function has infinite height but unit area.

- A basic property of the delta function is

$$\begin{aligned}\int_{-\infty}^{\infty} f(x) \cdot \delta(x-a) dx &= \int_{-\infty}^{\infty} f(a) \cdot \delta(x-a) dx \\ &= f(a) \cdot \int_{-\infty}^{\infty} \delta(x-a) dx \\ &= f(a)\end{aligned}$$



- Convolution of a function with $\delta(x)$ gives

$$\begin{aligned} f(x) * \delta(x) &= \int_{-\infty}^{\infty} f(x') \delta(x - x') dx' \\ &= f(x) \end{aligned}$$

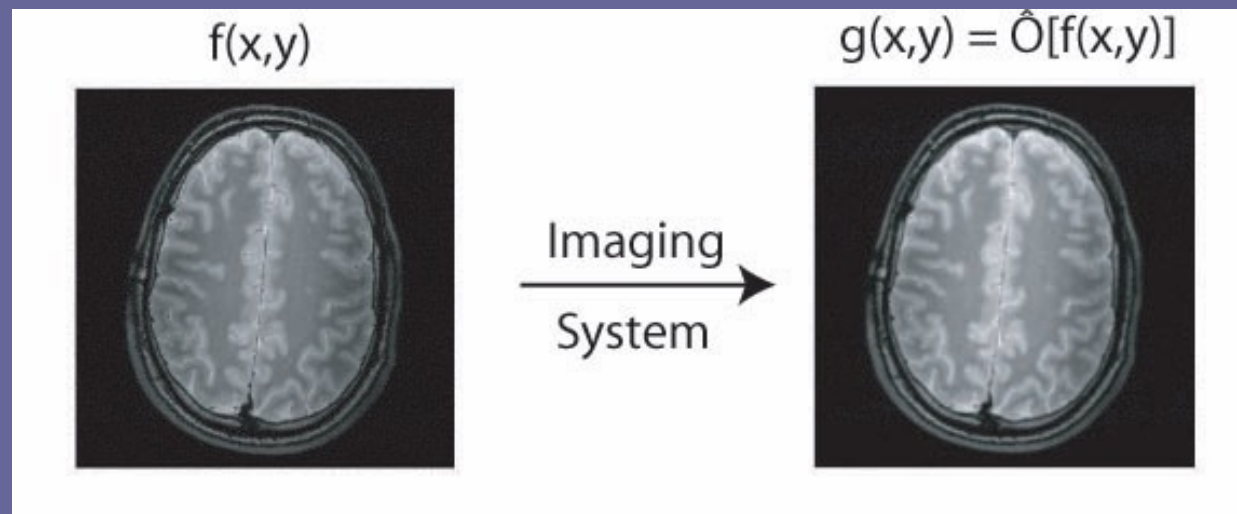
- The delta function can be considered the limit of a series of conventional (e.g., gaussian) functions:

$$\delta(x) = \lim_{\alpha \rightarrow \infty} \sqrt{\frac{\alpha}{\pi}} \cdot e^{-\alpha x^2}$$

In the limit, this has infinite amplitude (at $x=0$) and infinitesimal width.

Linear, shift invariant systems

- Imaging systems acquire signals from a spatial distribution of sources (an input) and produce an image (an output)



Let's denote the relation between input and output by the operator \hat{O}

- A system is *linear* if and only if

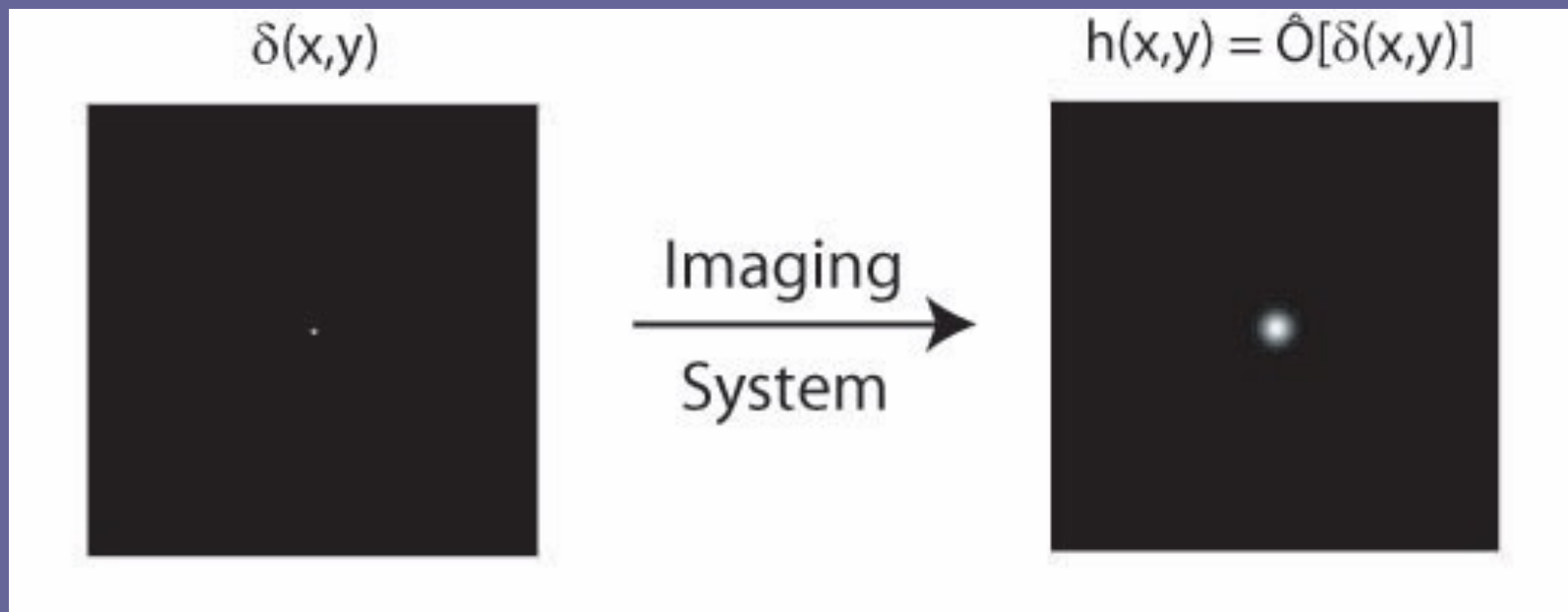
$$\hat{O}[\alpha f_1(x, y) + \beta f_2(x, y)] = \alpha \cdot \hat{O}[f_1(x, y)] + \beta \cdot \hat{O}[f_2(x, y)]$$

- Suppose our system is linear and we use it to image an ideal point source

$$f(x, y) = \delta(x, y) = \delta(x) \cdot \delta(y)$$

- The image of a point source is called the *point spread function* (PSF)

- The image of a point source is called the *point spread function* (PSF)

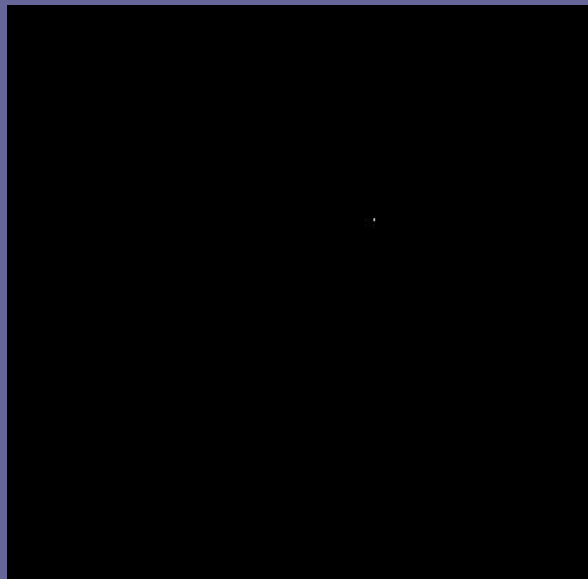


we'll write the PSF as $h(x,y)$.

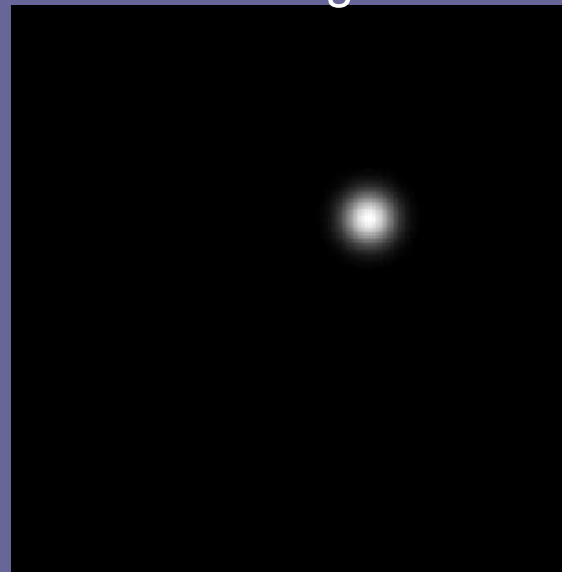
- Suppose that our system is not only linear, but also *shift invariant*
 - The only effect of shifting the input is to shift the output by the corresponding amount
 - For a delta function point source,

$$h(x - x_0, y - y_0) = \hat{O}[\delta(x - x_0, y - y_0)]$$

Point source



Image



- Suppose that our system is not only linear, but also *shift invariant*
 - The only effect of shifting the input is to shift the output by the corresponding amount
 - For a delta function point source,

$$h(x - x_0, y - y_0) = \hat{O}[\delta(x - x_0, y - y_0)]$$

- To save writing, we'll use the notation
input \rightarrow *output*

so

$$\delta(x - x_0, y - y_0) \rightarrow h(x - x_0, y - y_0)$$

- Given an array of point sources with different magnitudes, linearity and shift invariance imply

$$\sum_n f(x_n, y_n) \cdot \delta(x - x_n, y - y_n) \rightarrow$$

$$\sum_n f(x_n, y_n) \cdot h(x - x_n, y - y_n)$$

- In the limit that δ function spacing goes to zero, the sums are replaced by integrals:

$$\int f(x', y') \cdot \delta(x - x', y - y') dx' dy' \rightarrow$$

$$\int f(x', y') \cdot h(x - x', y - y') dx' dy'$$

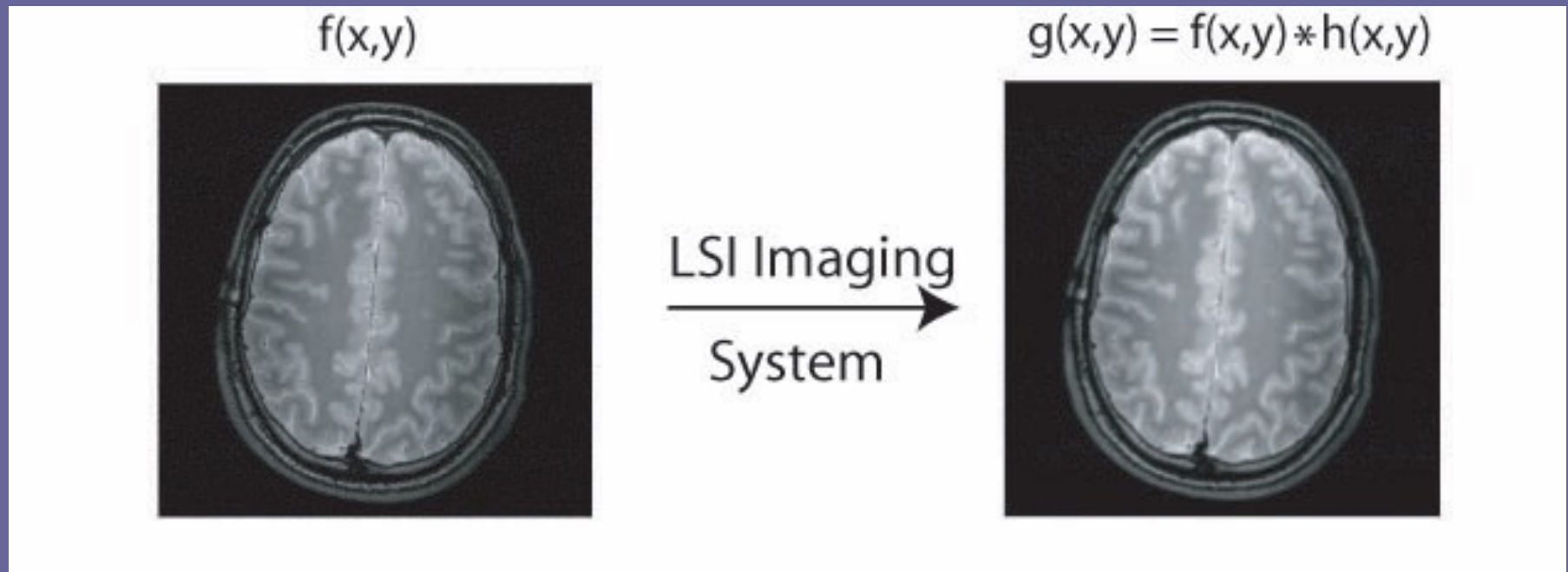
- In the limit that δ function spacing goes to zero, the sums are replaced by integrals:

$$\int f(x', y') \cdot \delta(x - x', y - y') dx' dy' \rightarrow$$
$$\int f(x', y') \cdot h(x - x', y - y') dx' dy'$$

or

$$f(x, y) \rightarrow f(x, y) * h(x, y)$$

The output is equal to the true distribution, convolved (i.e., blurred) by the PSF.

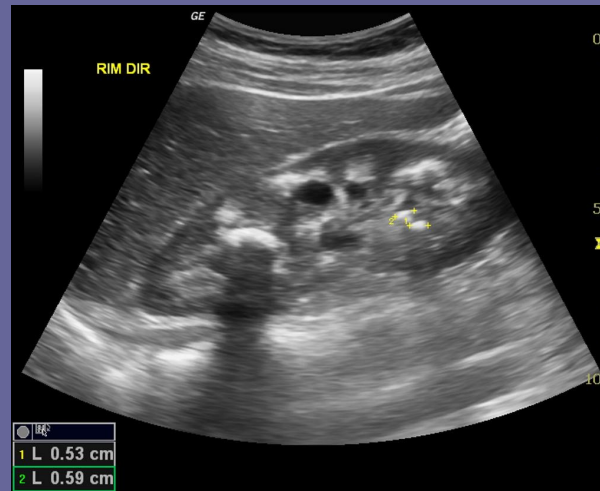


- Equivalent statements:
 - An imaging system is LSI
 - The image is equal to a convolution of the input with a PSF

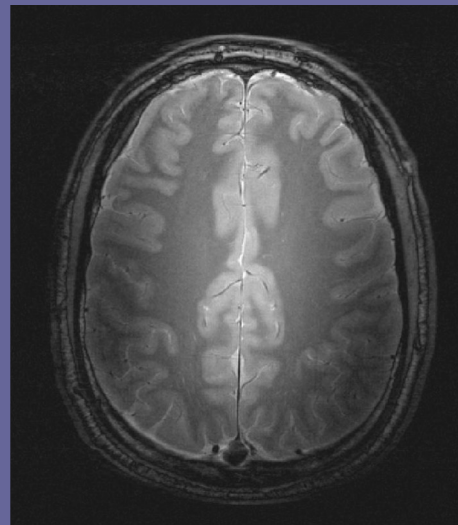
Which imaging modality made these?



Sureshabu, J Nuc Med Tech (2005)



Case courtesy of Dr Bruno Di Muzio, Radiopaedia.org

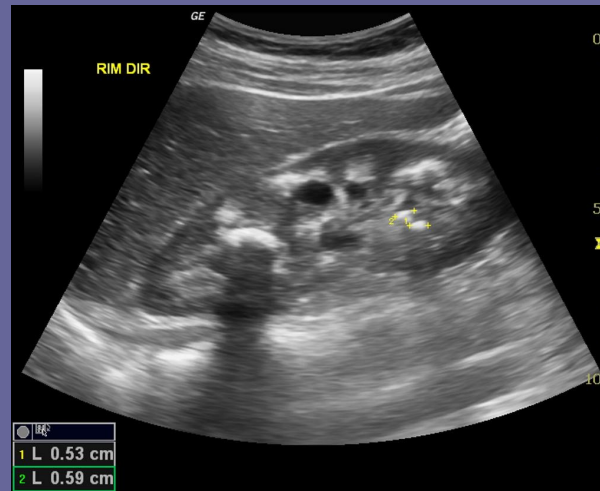


Kinahan (AAPM)

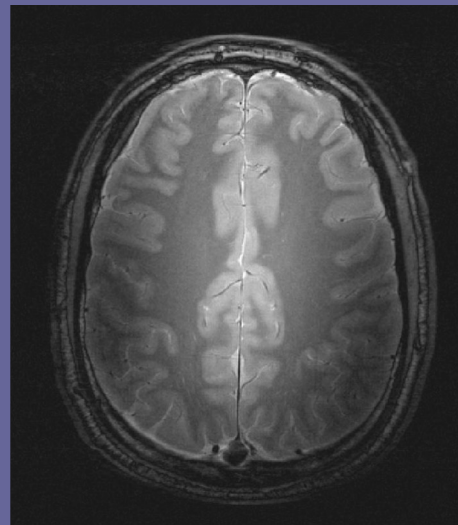
In-class exercise: Choose an imaging modality—is it LSI?



Sureshababu, J Nuc Med Tech (2005)



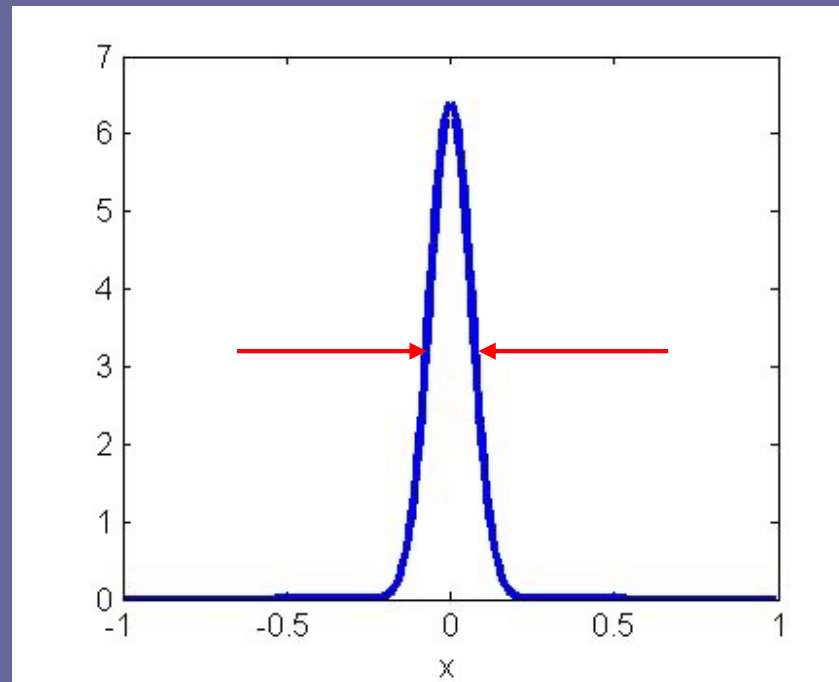
Case courtesy of Dr Bruno Di Muzio, Radiopaedia.org



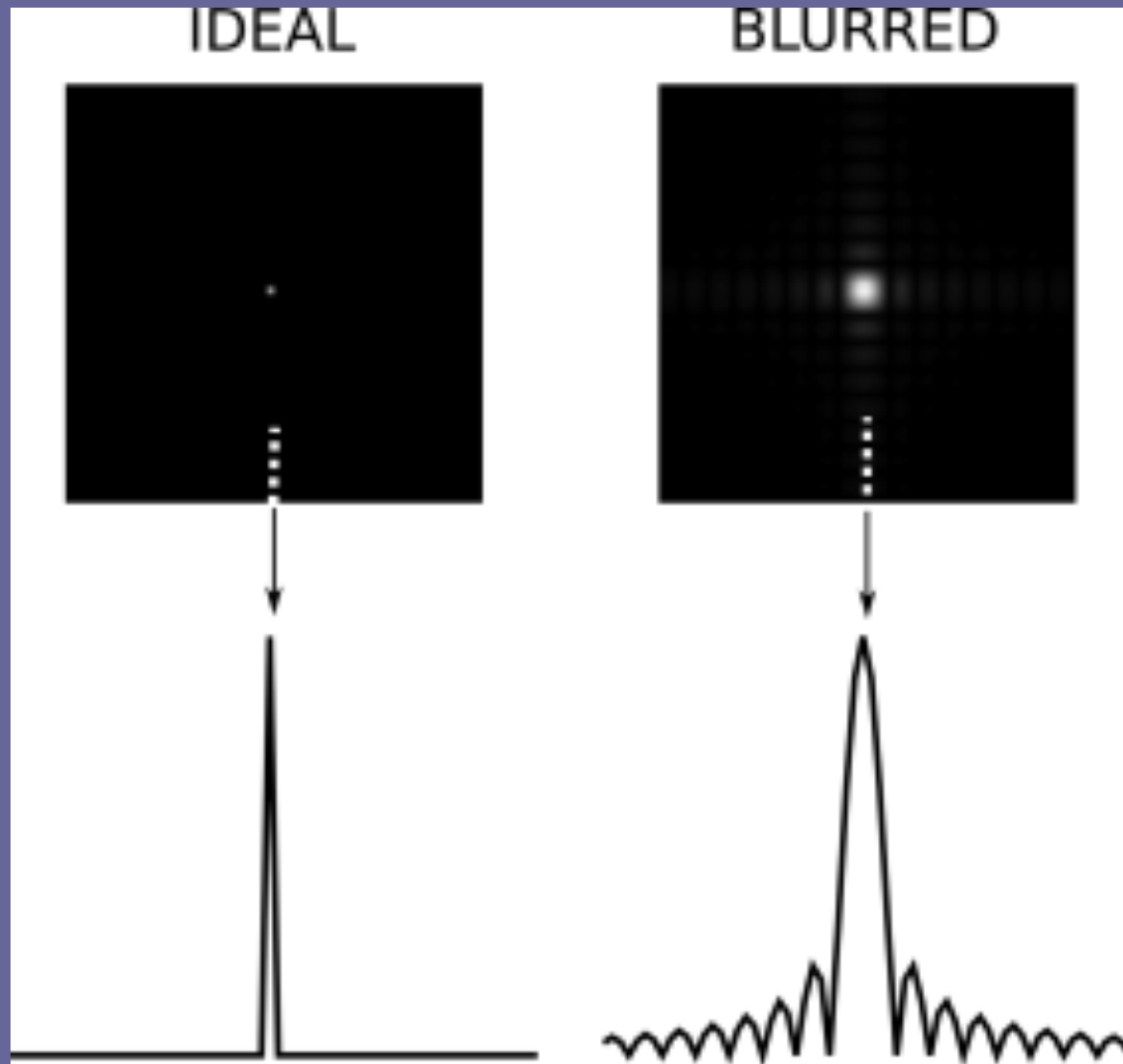
Kinahan (AAPM)

The PSF determines resolution

- A common definition of resolution is the full width at half maximum (FWHM) of the PSF

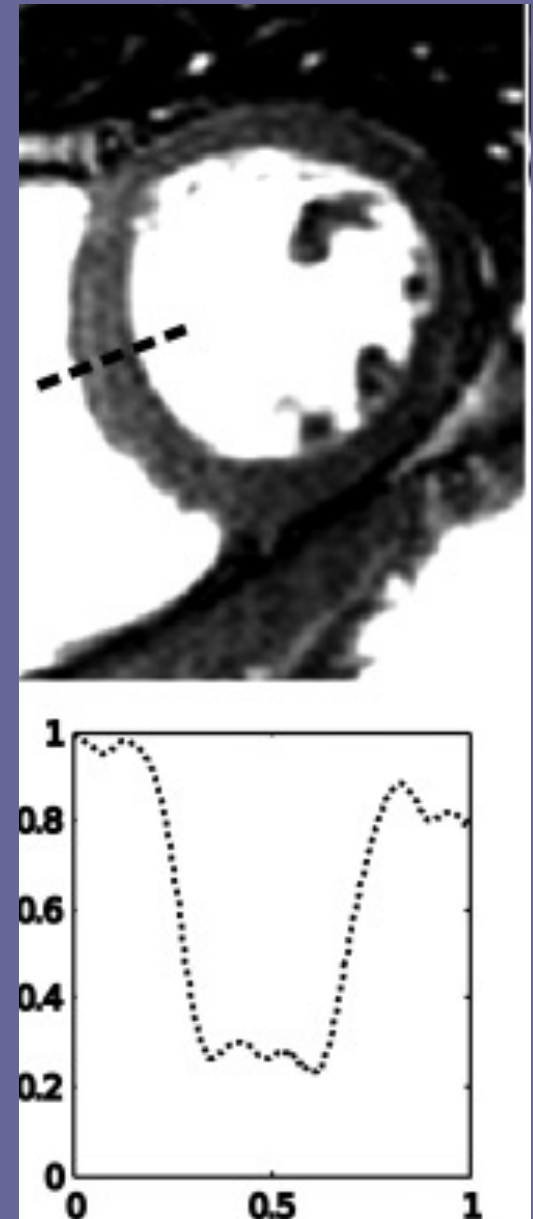


The Point Spread Function describes blurring

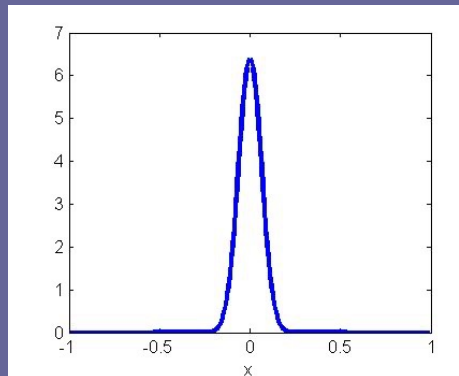


Effects of the PSF

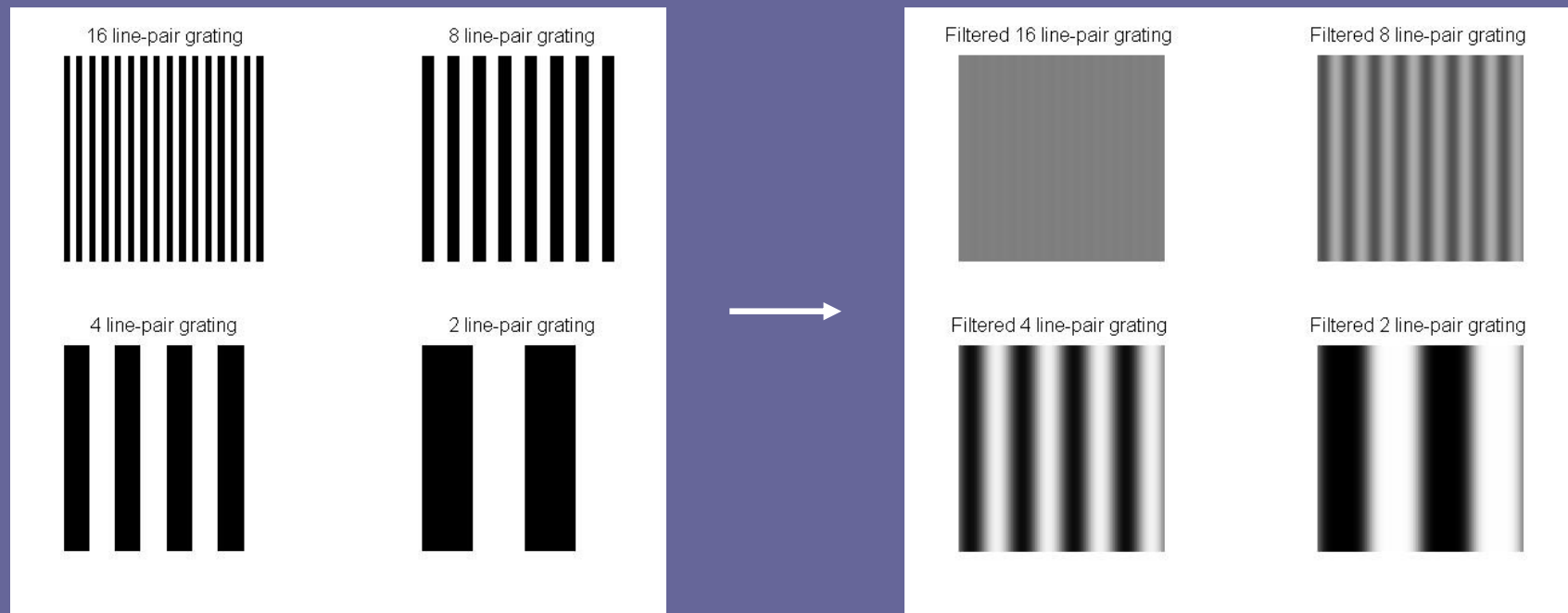
- PSF width causes blurring in the image
- Oscillation in the PSF causes ringing in the image
 - Edge artifacts that can be mistaken for real structure
 - Most obvious near strong contrast boundaries
- Solutions?



Resolution affects contrast



PSF



Partial volume averaging

- Each volume element (*voxel*) of the image represents a volume of tissue
 - Voxels are not cubes—they are broadened by the PSF
- Tissue properties may not be uniform in the voxel
- Measurements from the image data usually don't reflect true properties of heterogeneous voxels
- Measurements from the data may not give good estimates of volume-averaged properties

Artifacts

- Image reconstruction errors
 - Missing signal
 - Extra signal
 - Displaced signal
- Violation of assumptions of reconstruction
 - Motion
 - Model relating signal to tissue properties is too simple...
- Artifacts interfere with analysis (usually don't average out)

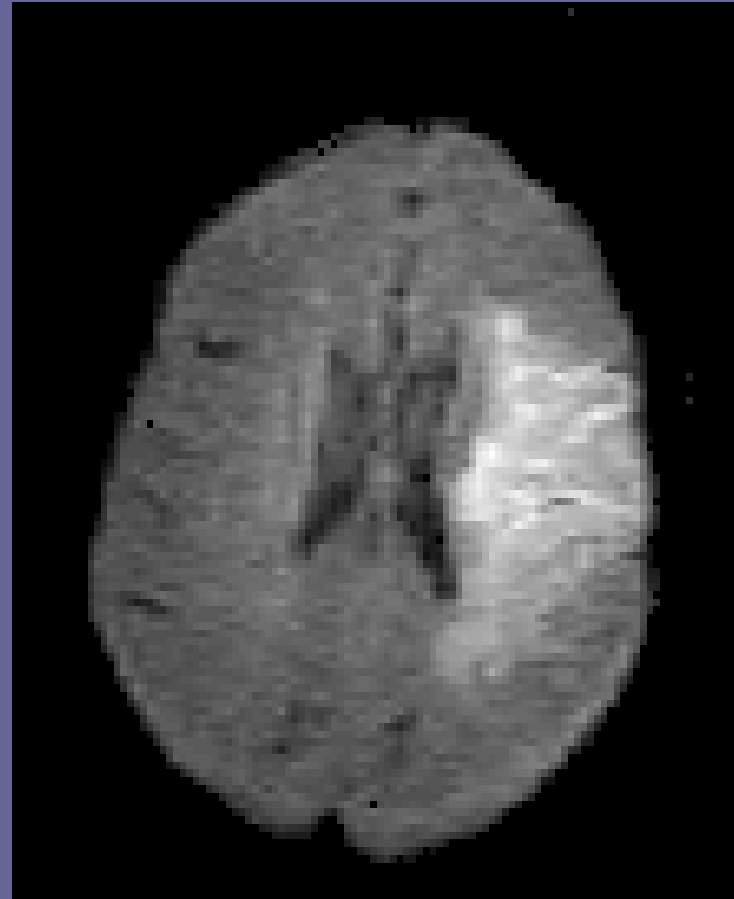
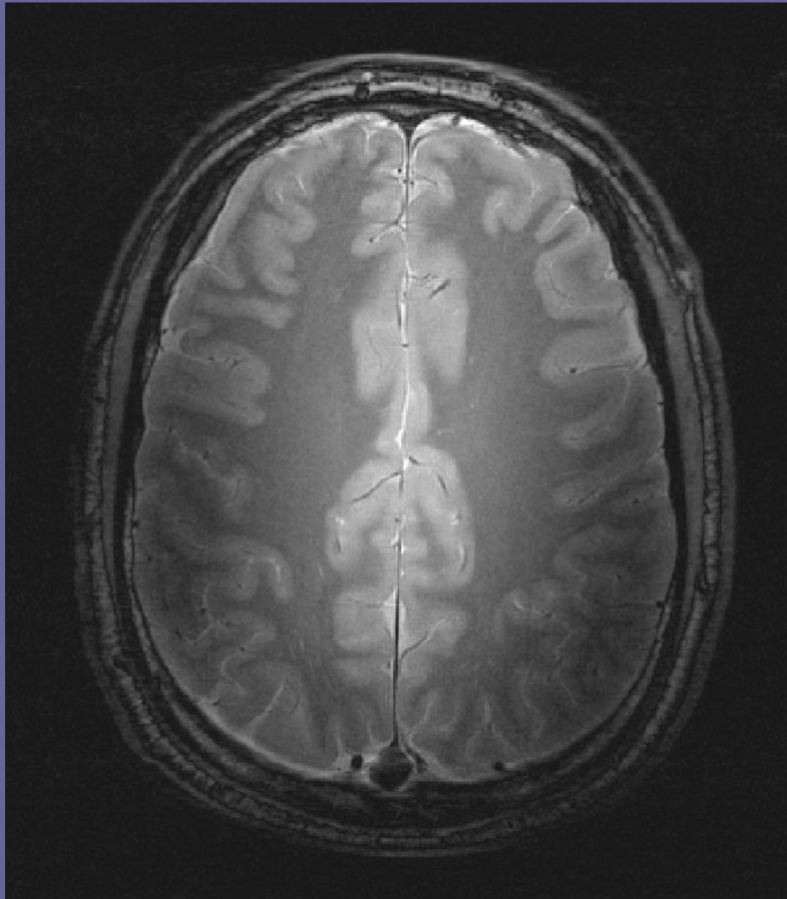


Artifacts

- Image reconstruction errors
 - Missing signal
 - Extra signal
 - Displaced signal
- Violation of assumptions of reconstruction
 - Motion
 - Model relating signal to tissue properties is too simple...
- Artifacts interfere with analysis (usually don't average out)



How are these images different?



Summary

- Information in an image depends on
 - SNR
 - Contrast
 - Resolution
 - Artifacts
- These are the most important aspects of image quality

References

- Hansen MS, Kellman P. Image reconstruction: An overview for clinicians. Journal of Magnetic Resonance Imaging. 25 JUN 2014 DOI: 10.1002/jmri.24687
- Suetens P. Fundamentals of Medical Imaging (Cambridge, 2002).