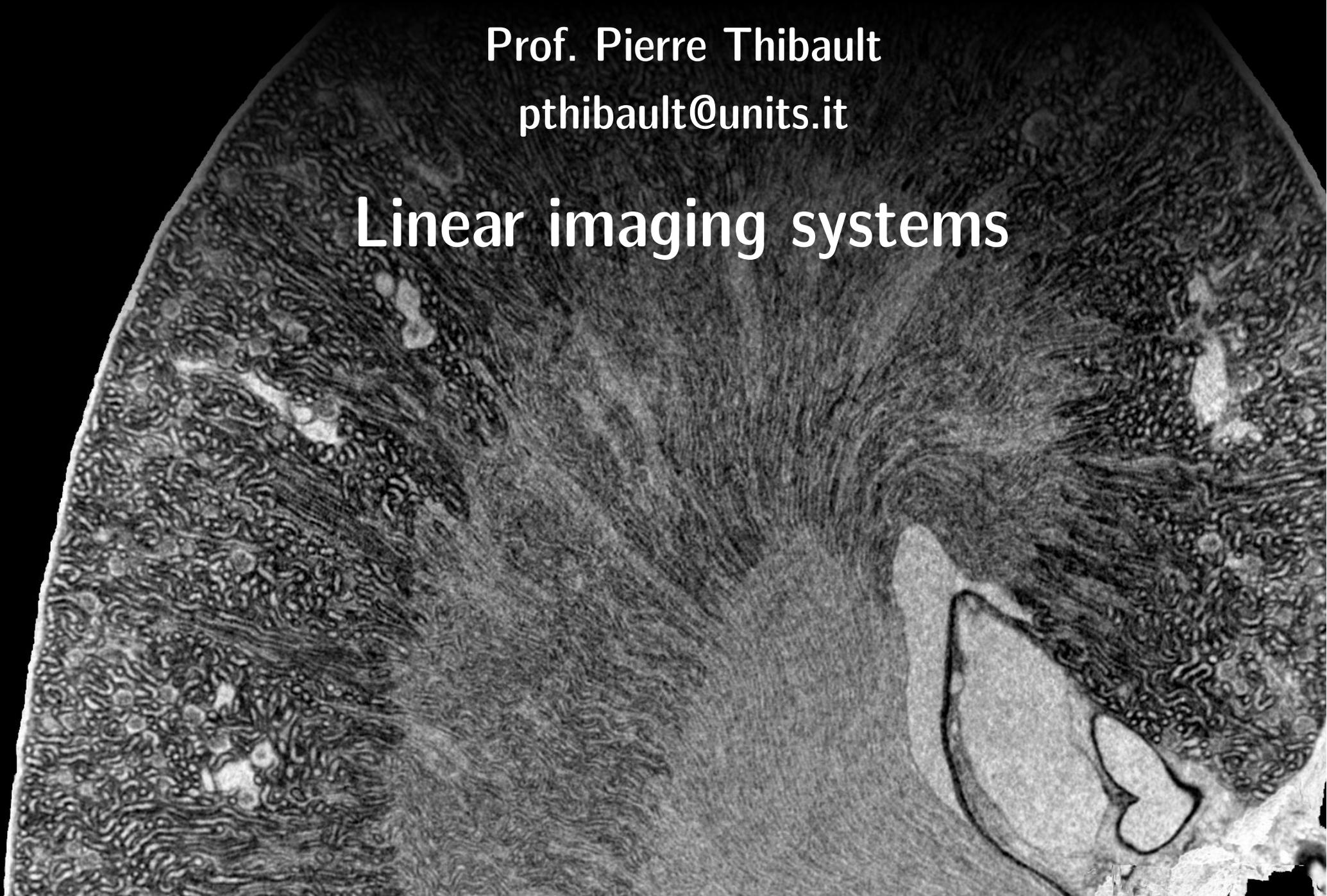


# Image Processing for Physicists

Prof. Pierre Thibault

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## Linear imaging systems



# Overview

- Definition of resolution
- Imaging systems:
  - Linear transfer model
  - Noise

# Resolution

“the smallest detail that can be distinguished”

- No unique definition
  - Numerical aperture ← microscopy, photography, astronomy
  - Pixel size ← detector-limited
  - Other criteria (PSF, MTF) ← generalization
- What is “detail”?
- What is “distinguish”?

# Resolution

1280 x 1280



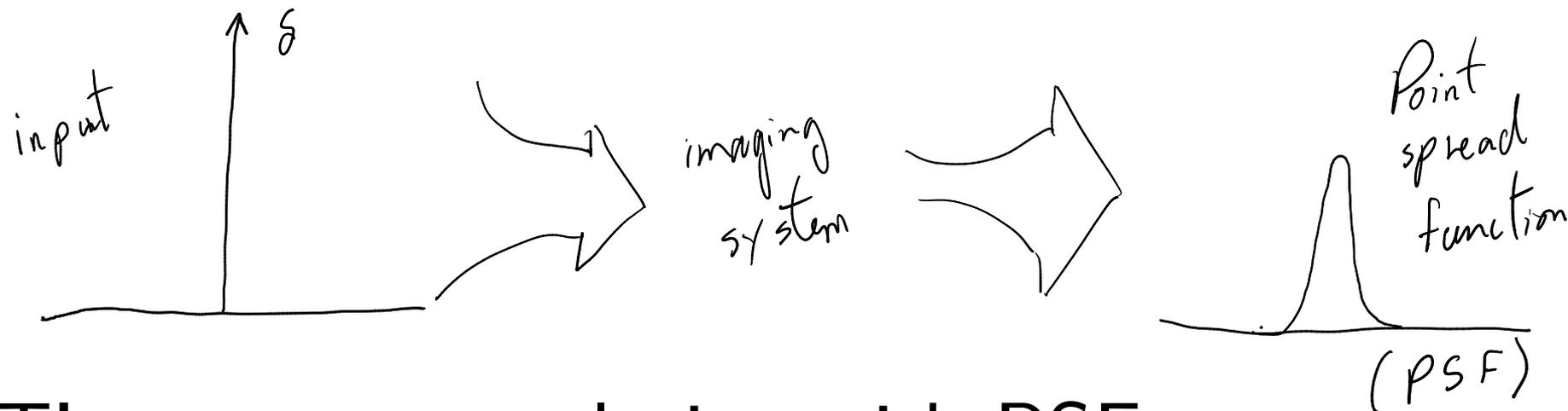
640 x 640



- **not** simply given by pixel size (i.e. sampling rate)
- light quality, optics quality, detector quality, algorithm quality, noise, ...

# Linear translation-invariant systems

- Point spread function (“impulse response”)



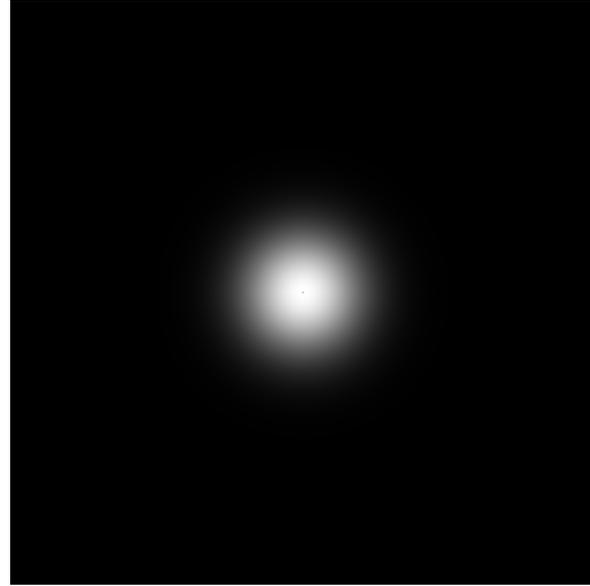
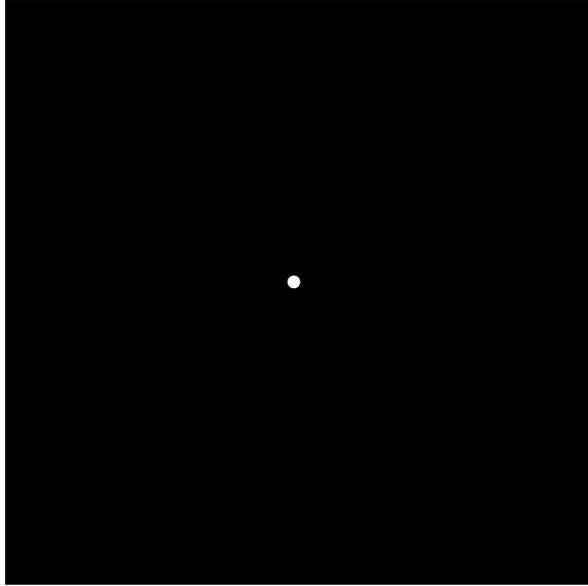
- LTI system: convolution with PSF

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

↓ Imaging system

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

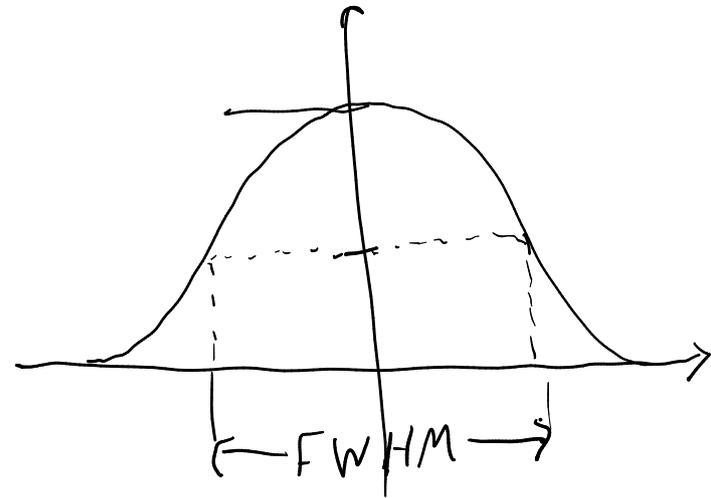
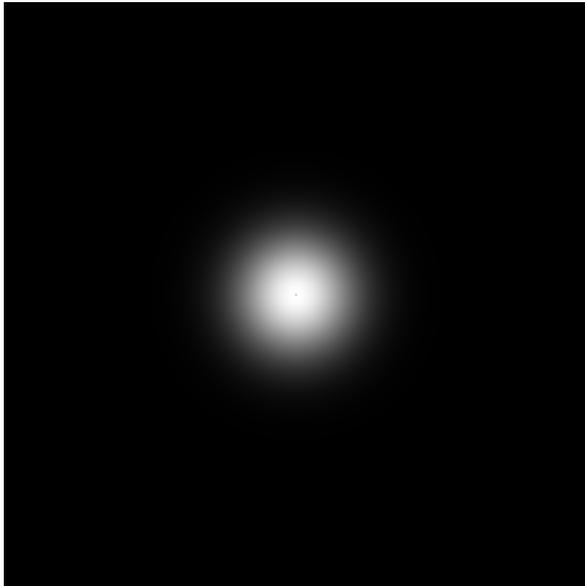
# Point spread function



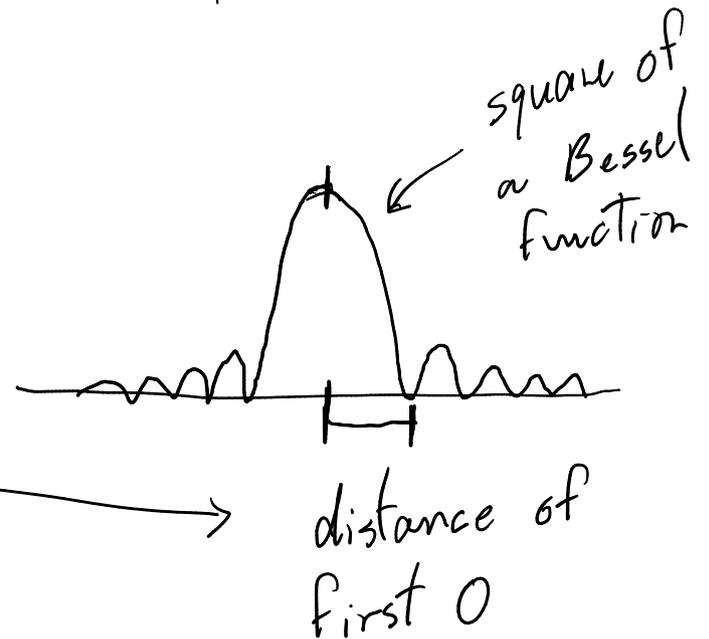
# PSF and resolution

\* Common way to describe (quantify)

resolution from PSF is "Full Width at Half Maximum" (FWHM)



\* Another definition of resolution is Rayleigh criterion



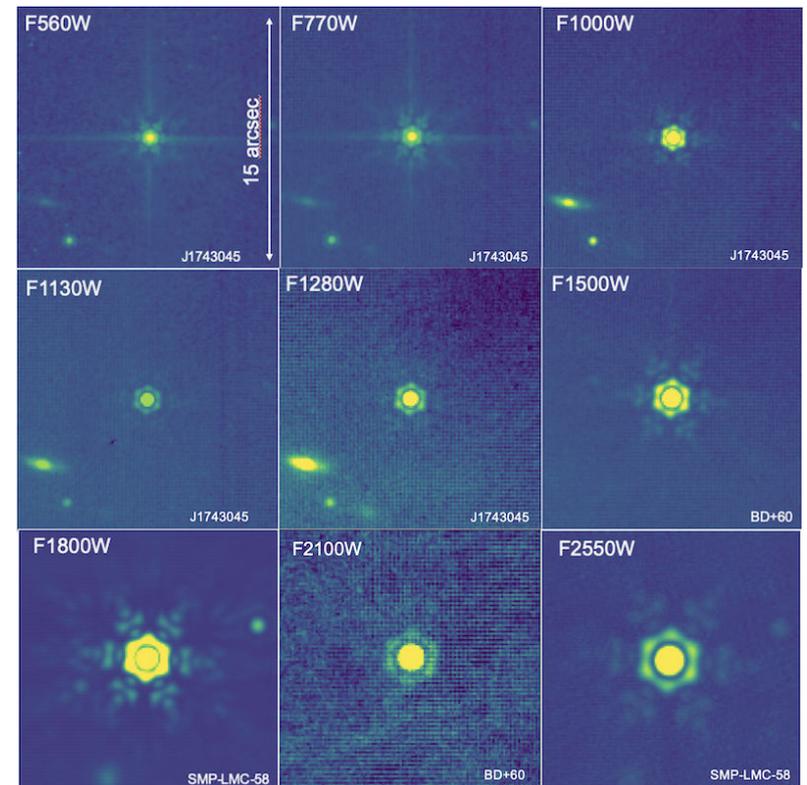
# Measurement of the PSF

- Direct measurement from impulse

*Most direct way: image a sharp point feature*

*astronomy: just pick a star!*

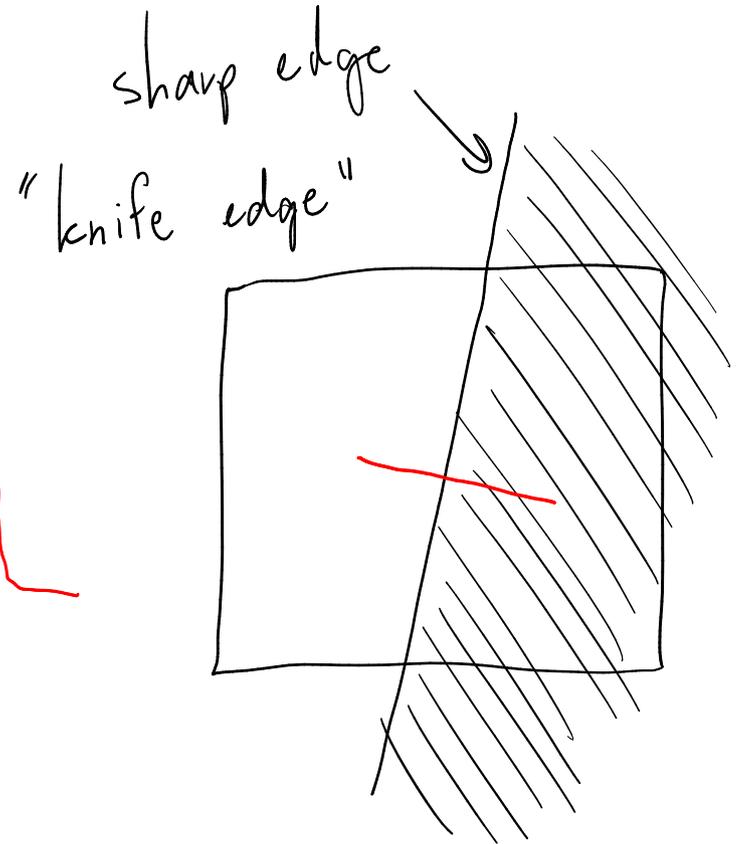
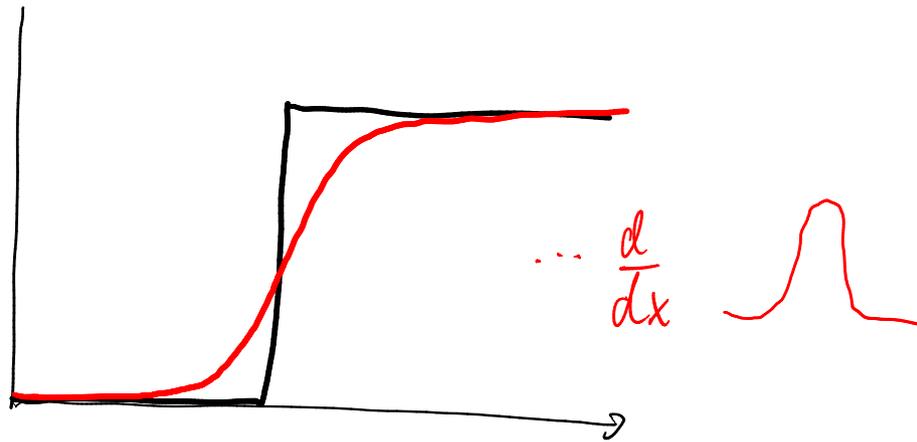
Example of measured PSF  
(James Web Spatial Telescope)



# Measurement of the PSF

Edge

- ~~Line~~-spread function

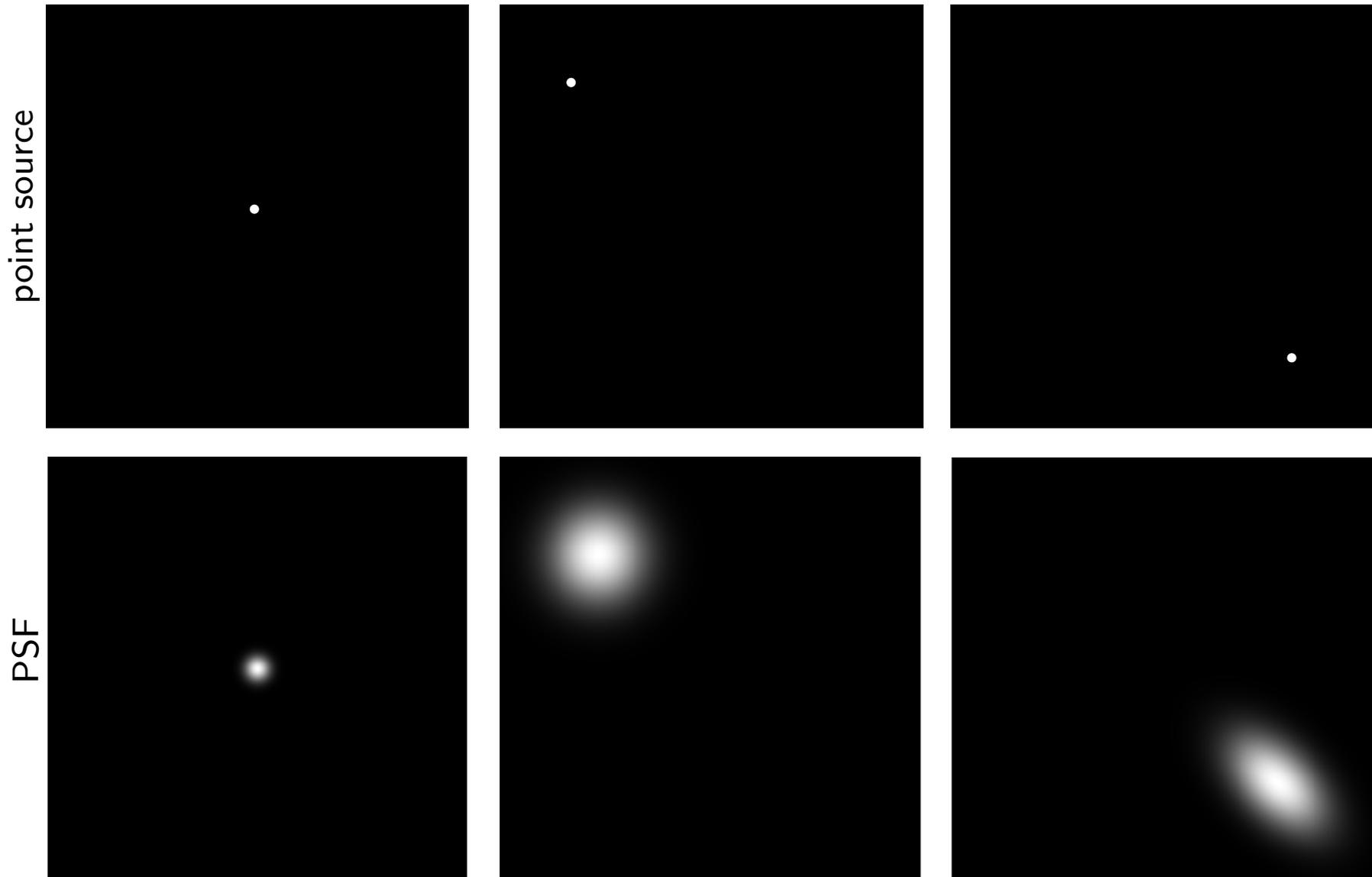


input:  $H(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$

$$\frac{dH}{dx} = \delta(x)$$

PSF = derivative of edge-spread function

# PSF and translation invariance



- Not translation invariant  $\rightarrow$  PSF depends on position  $\rightarrow$  not a convolution
- Useful to model system imperfections, lens aberrations, ...

LTI

# The Fourier picture

$$\mathcal{F}\{f * h\} = F(u) * H(u)$$

Fourier transform of PSF

||

"optical transfer function"

OTF

Consider a single spatial frequency  $u_0$

$$f(x) = A e^{2\pi i u_0 x}$$

Imaging system

$$F(u) = A \delta(u - u_0)$$

original amplitude

$$H(u_0) A \delta(u - u_0)$$

↓

$$H(u_0) A e^{2\pi i u_0 x}$$

modulated amplitude

\* pure oscillations of the form  $e^{2\pi i u_0 x}$  are a LTI imaging system

eigenfunctions of

# Optical transfer function

Response of a system to an oscillating signal with well-defined frequency

$$H(u) \rightarrow \text{OTF} = \mathcal{F}\{\text{PSF}\}$$

↑ complex valued in general

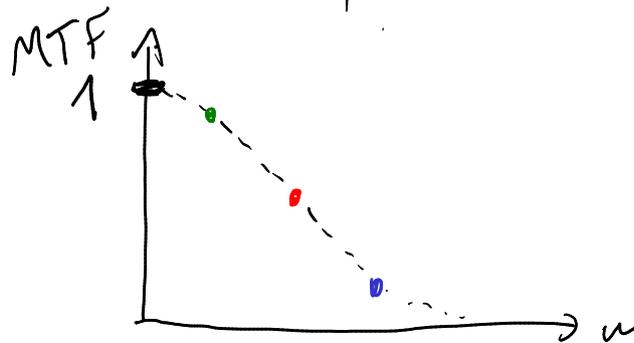
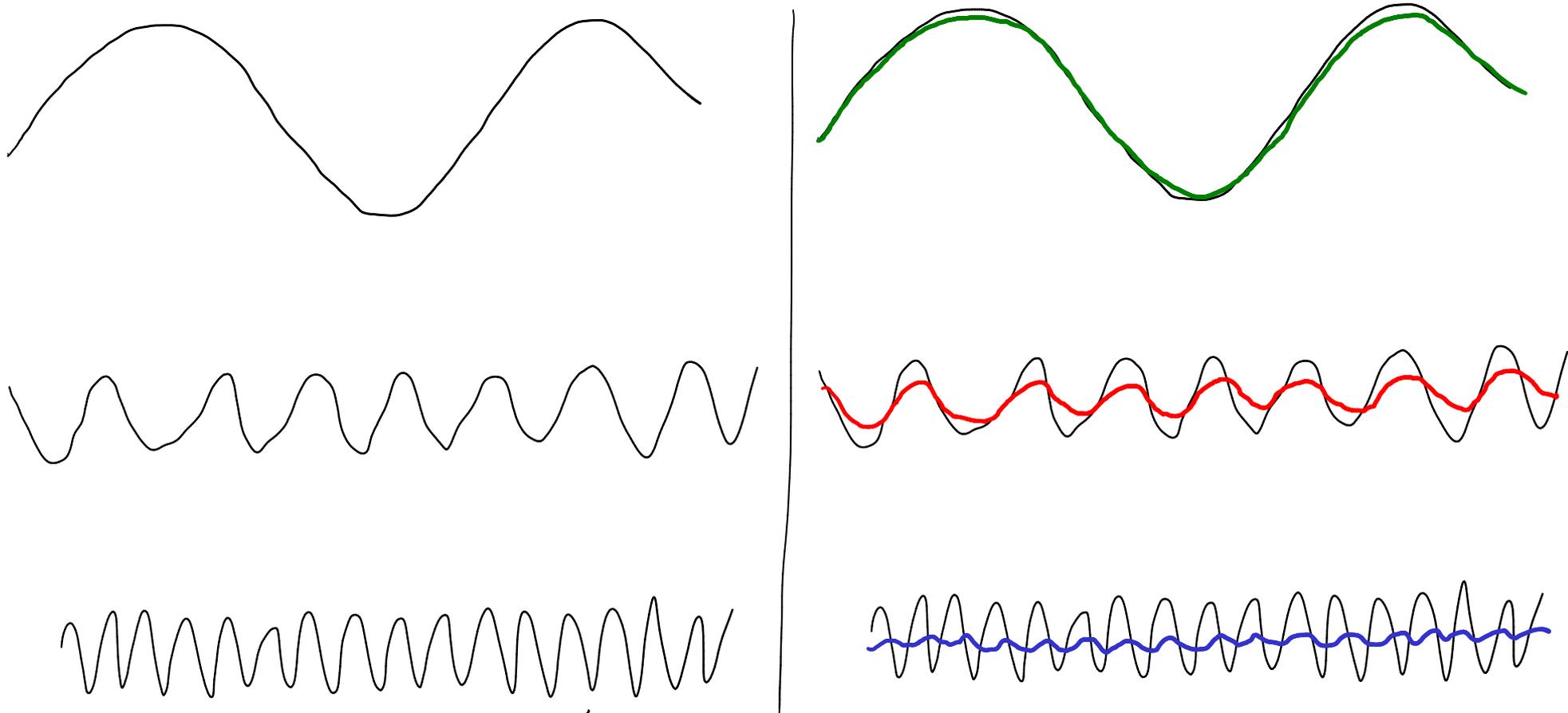
Amplitude  $|\text{OTF}| = \text{MTF}$  modulation transfer function"

Phase  $\arg\{\text{OTF}\} = \text{PTF}$  phase transfer function

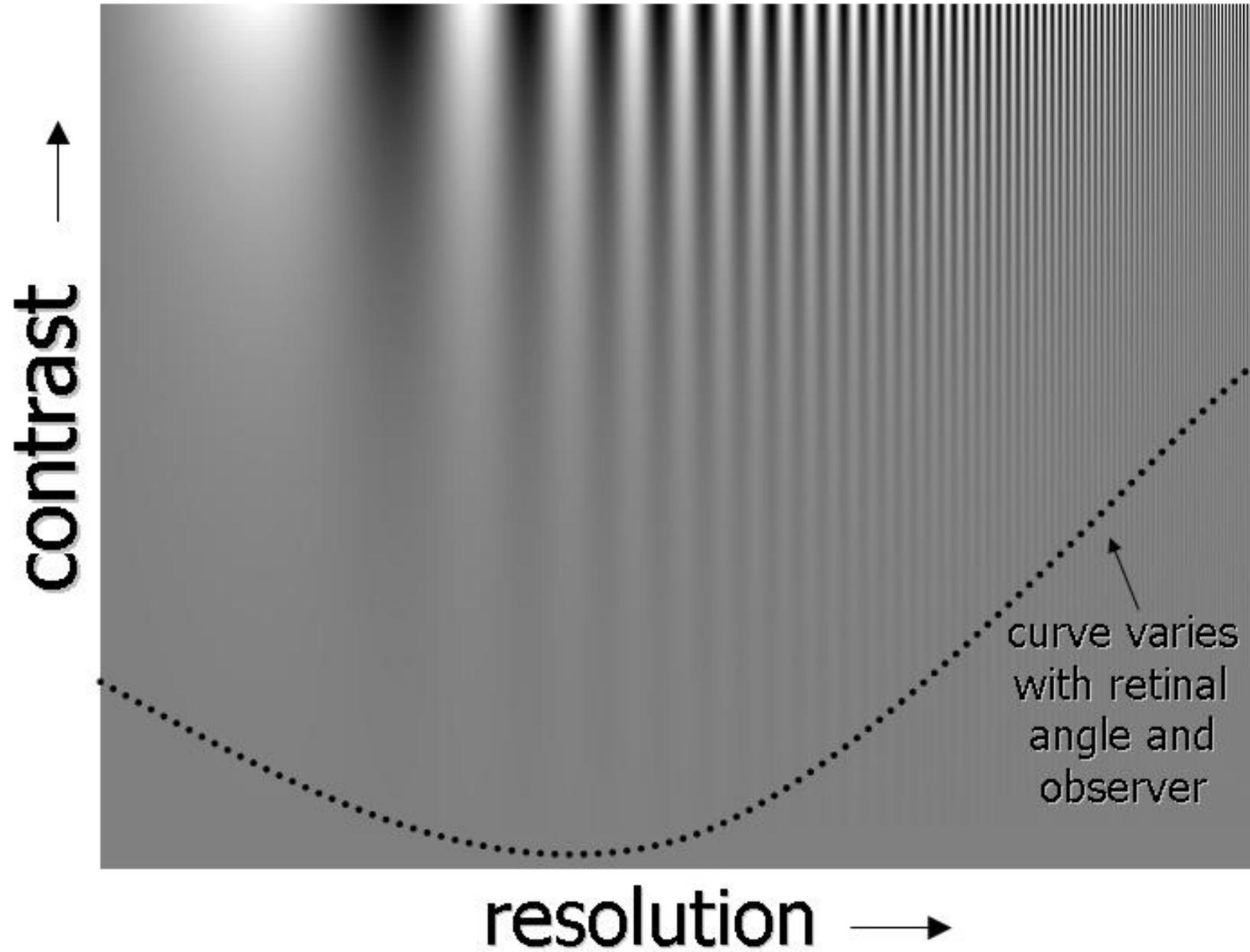
$$\mathcal{F}\{\text{PSF}\} = \text{OTF} = \text{MTF} e^{i\text{PTF}}$$

# Modulation transfer function

Amplitude change of an oscillating signal for a given frequency



# Eye MTF



# Campbell-Robson curve

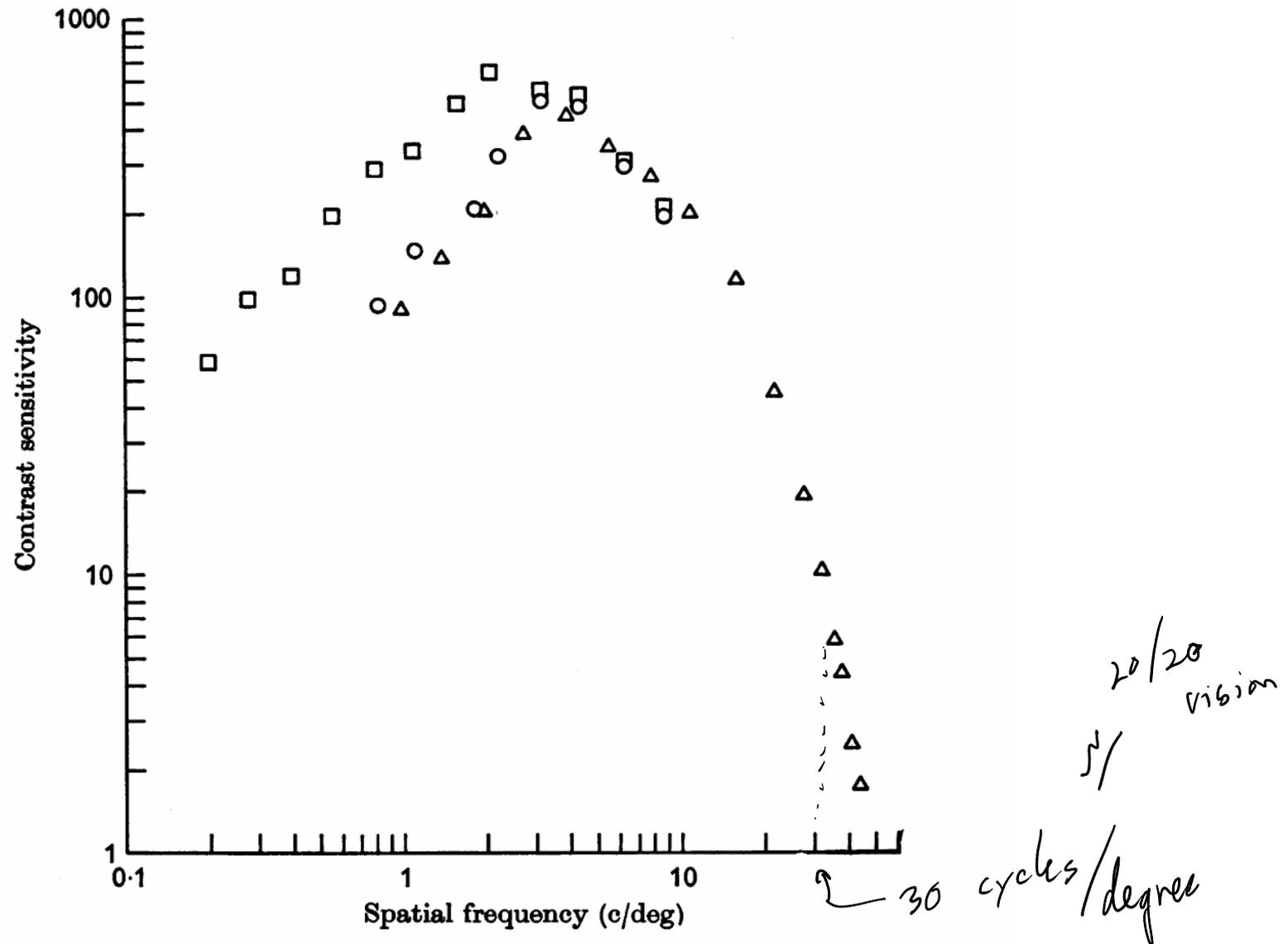
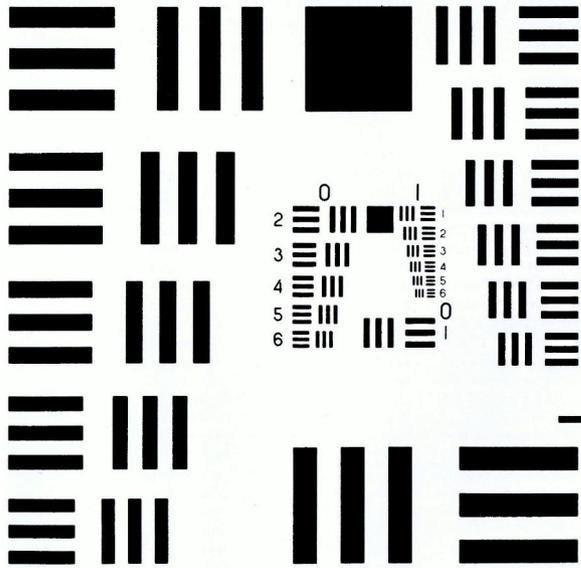
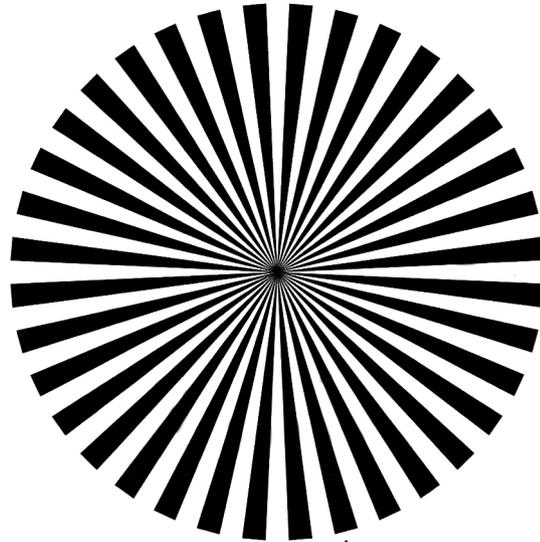


Fig. 2. Contrast sensitivity for sine-wave gratings. Subject F.W.C., luminance 500 cd/m<sup>2</sup>. Viewing distance 285 cm and aperture 2° × 2°, △; viewing distance 57 cm, aperture 10° × 10°, □; viewing distance 57 cm, aperture 2° × 2°, ○.

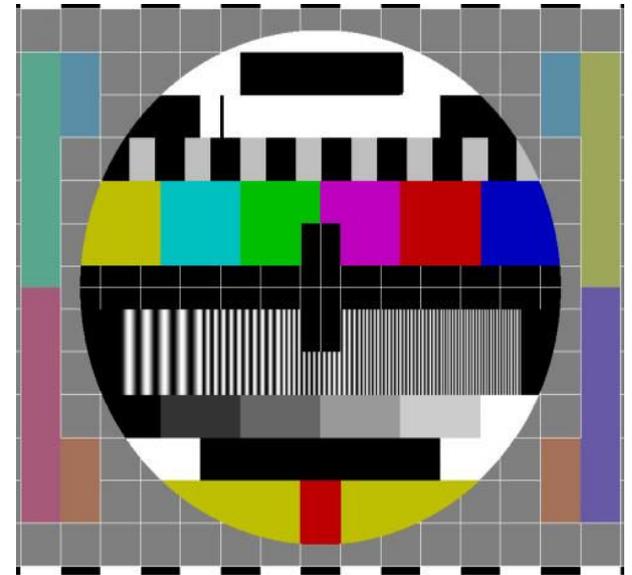
# Measurement of MTF



USAF  
resolution  
target



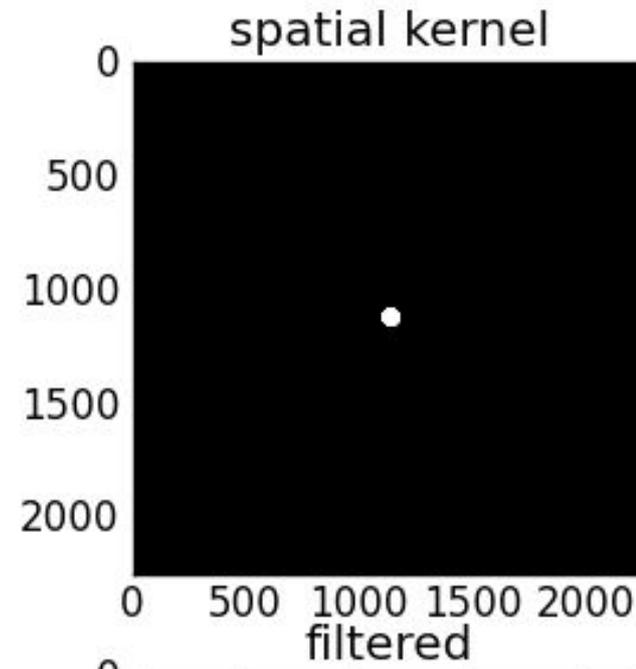
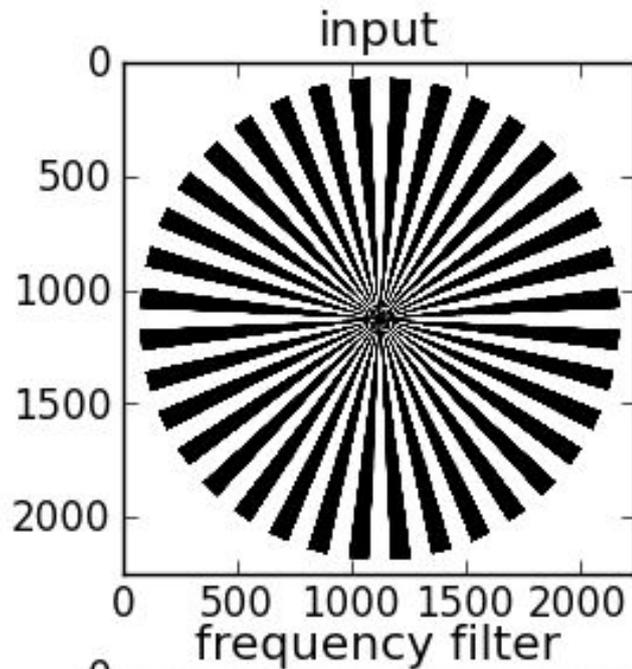
Siemens star



source: <http://fotomagazin.de>

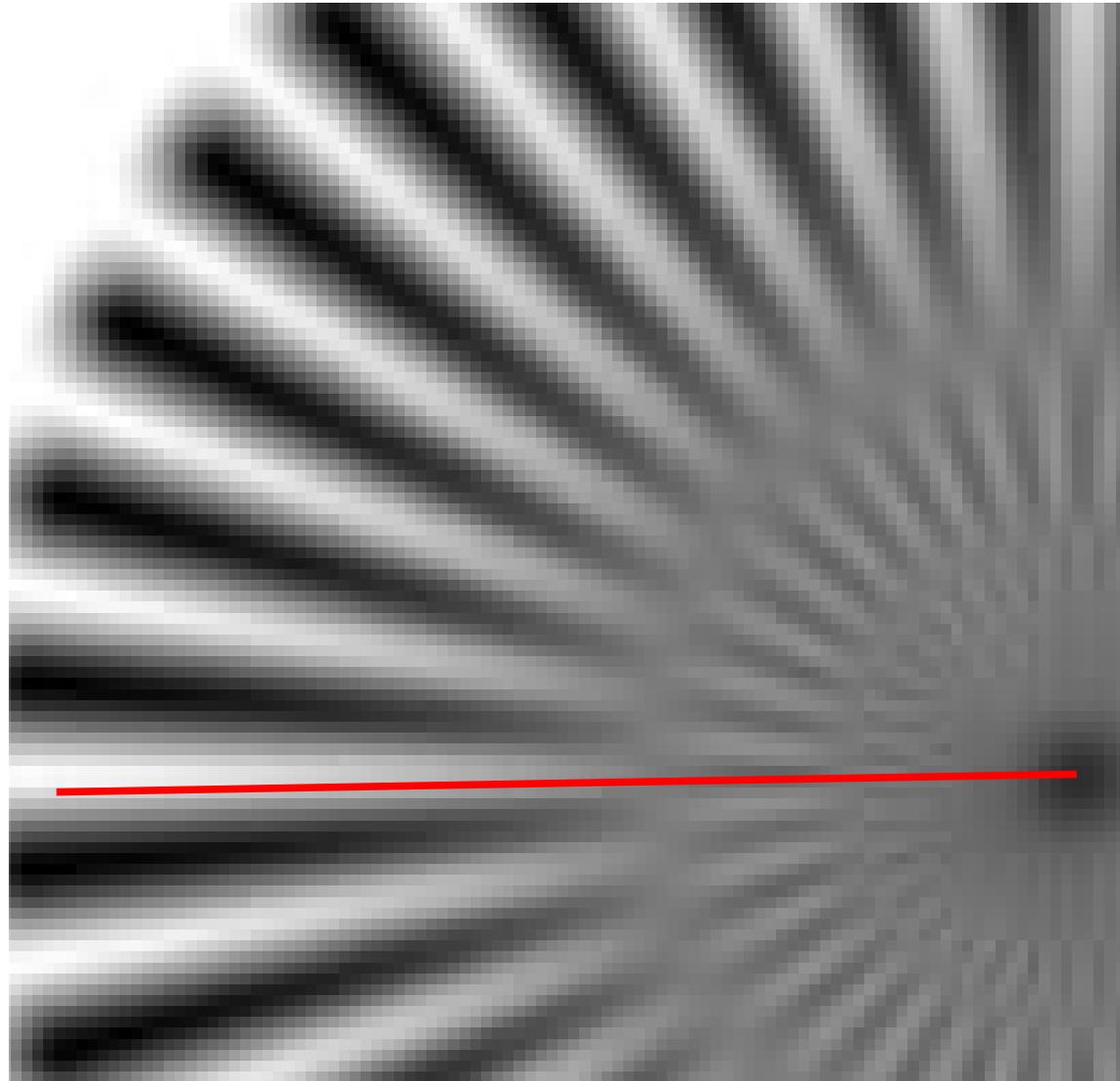
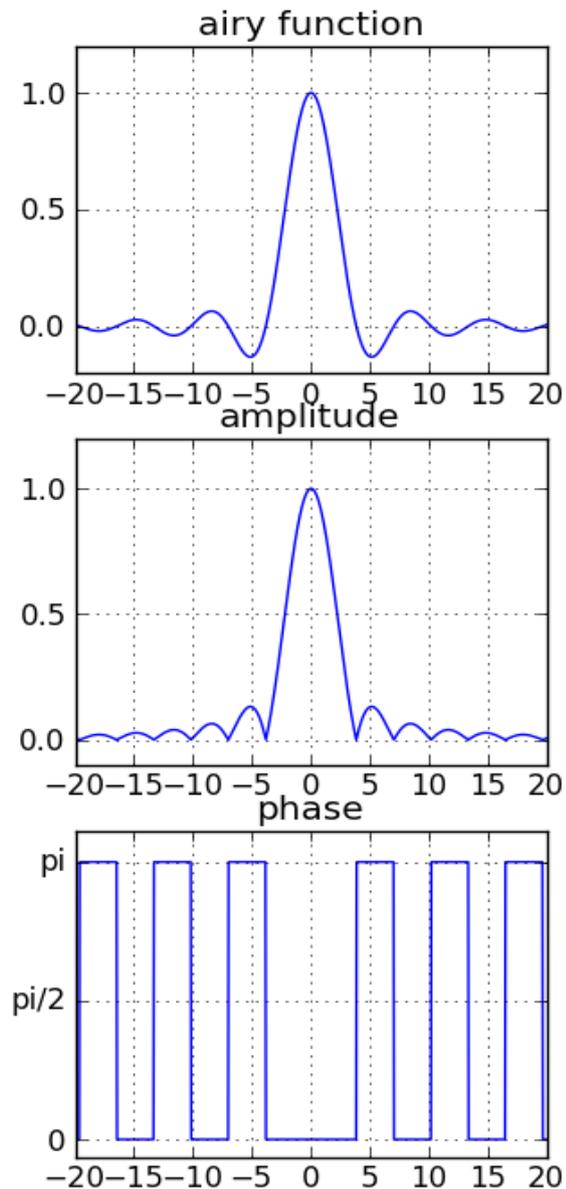
# Phase transfer function

describes how an oscillating signal changes in phase due to system

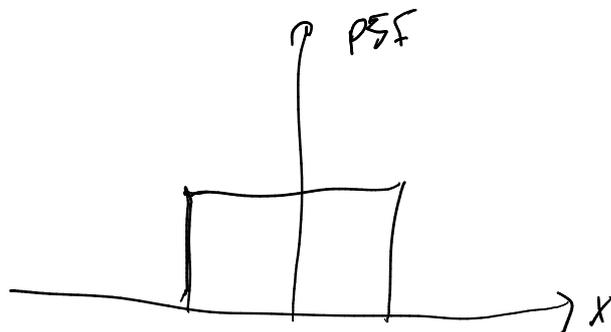
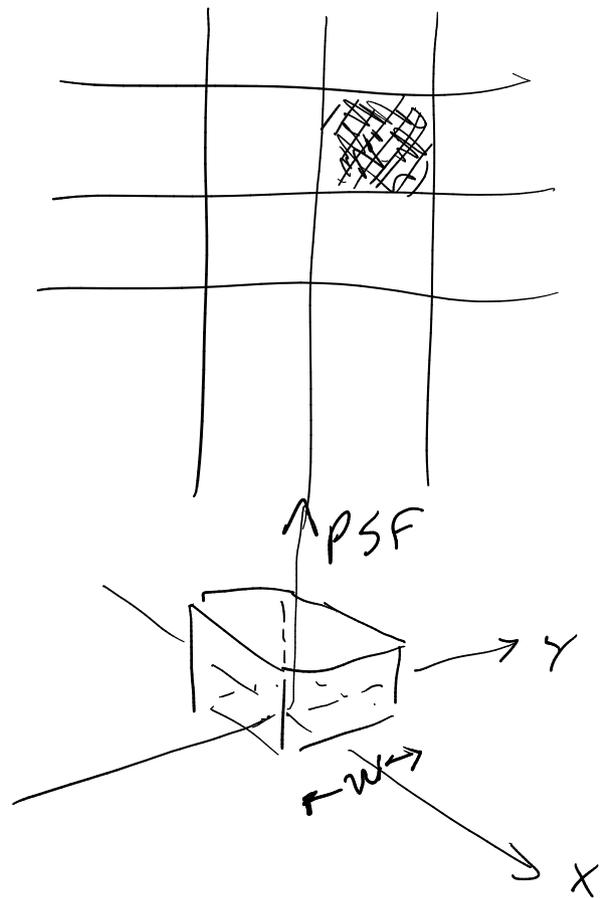


# Phase transfer function

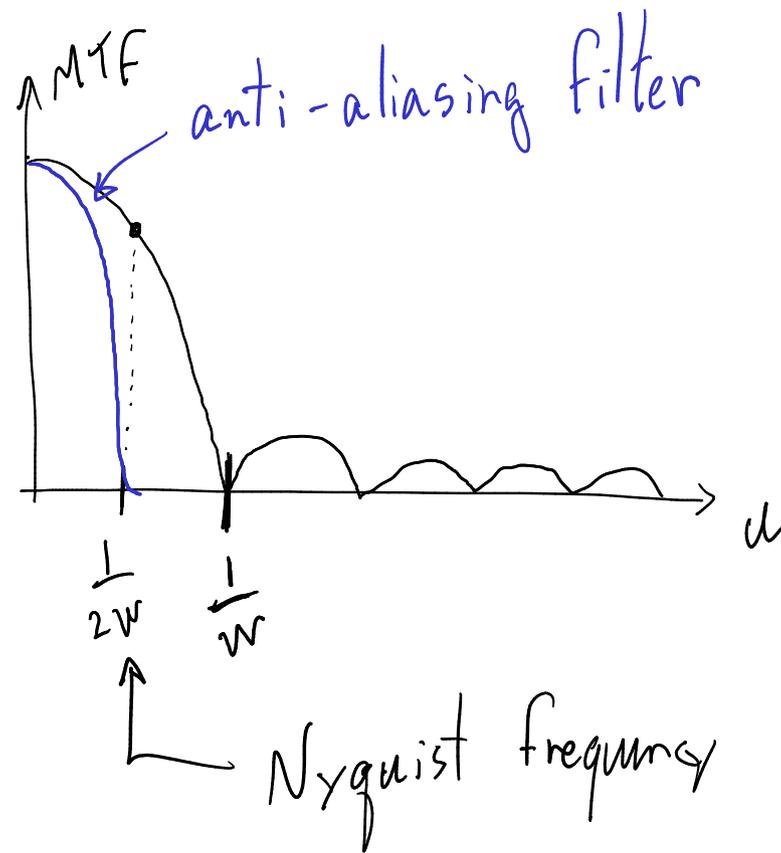
describes how an oscillating signal changes in phase due to system



# MTF of an ideal pixel

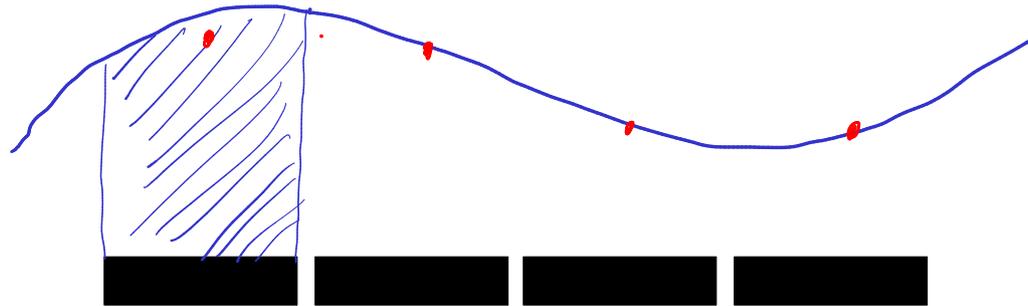


square

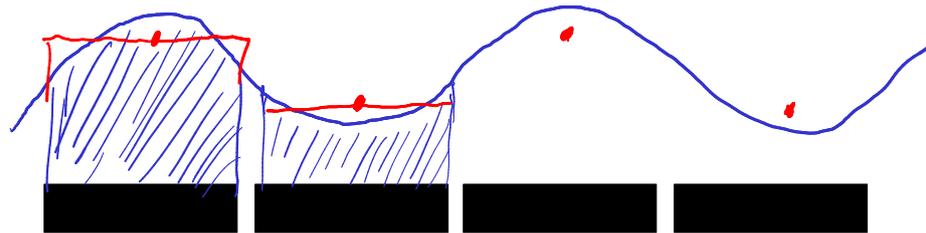


# Pixel MTF

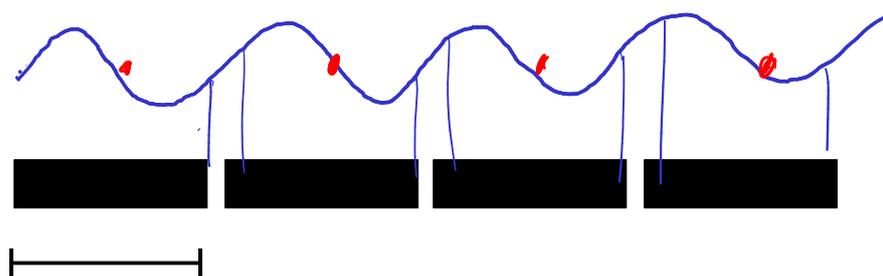
Modulation transfer function of a single detector pixel



$< N_{\text{Nyquist}}$



$N_{\text{Nyquist}}$

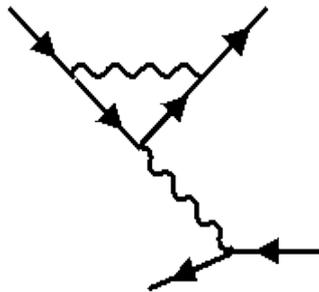
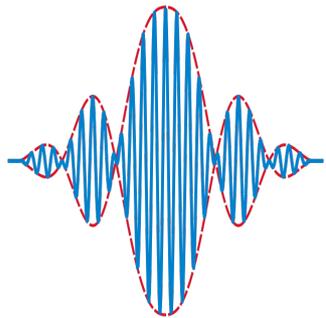


$2x N_{\text{Nyquist}}$

# Imaging as a linear filter

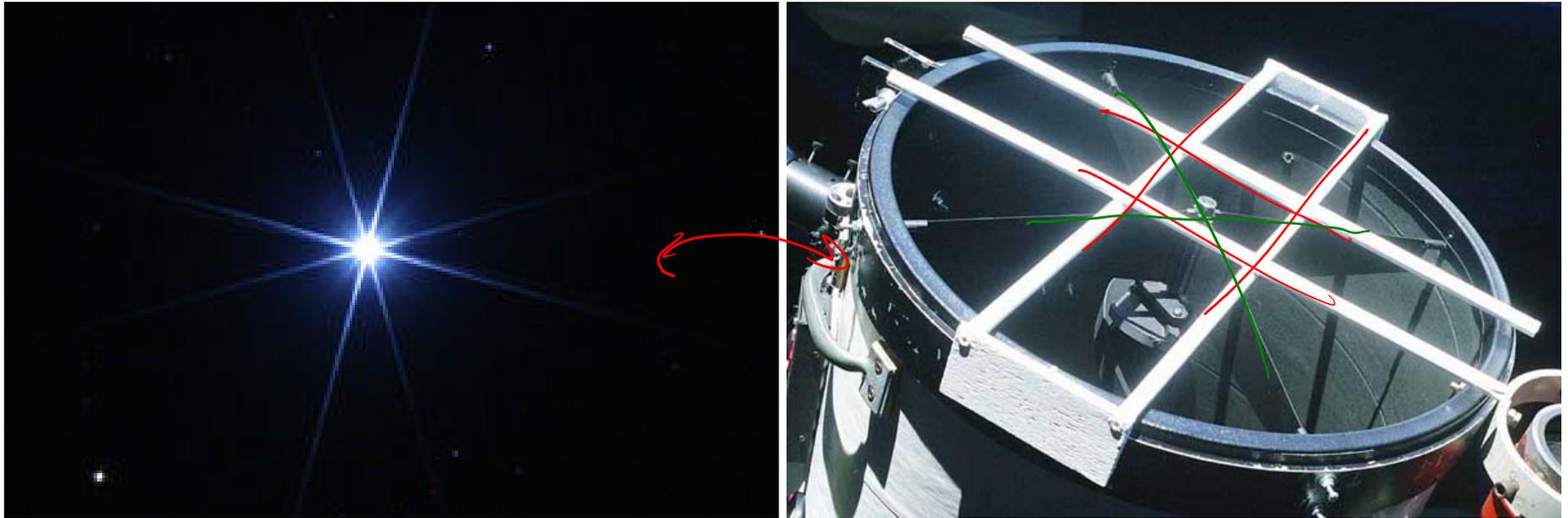
$$\text{Output}(u) = \text{Input}(u) \times \text{MTF}_{\text{optics}}(u) \times \text{MTF}_{\text{detector}}(u) \times \text{MTF}_{\text{algorithm}}(u) \times \dots$$

} effective MTF



# PSF examples

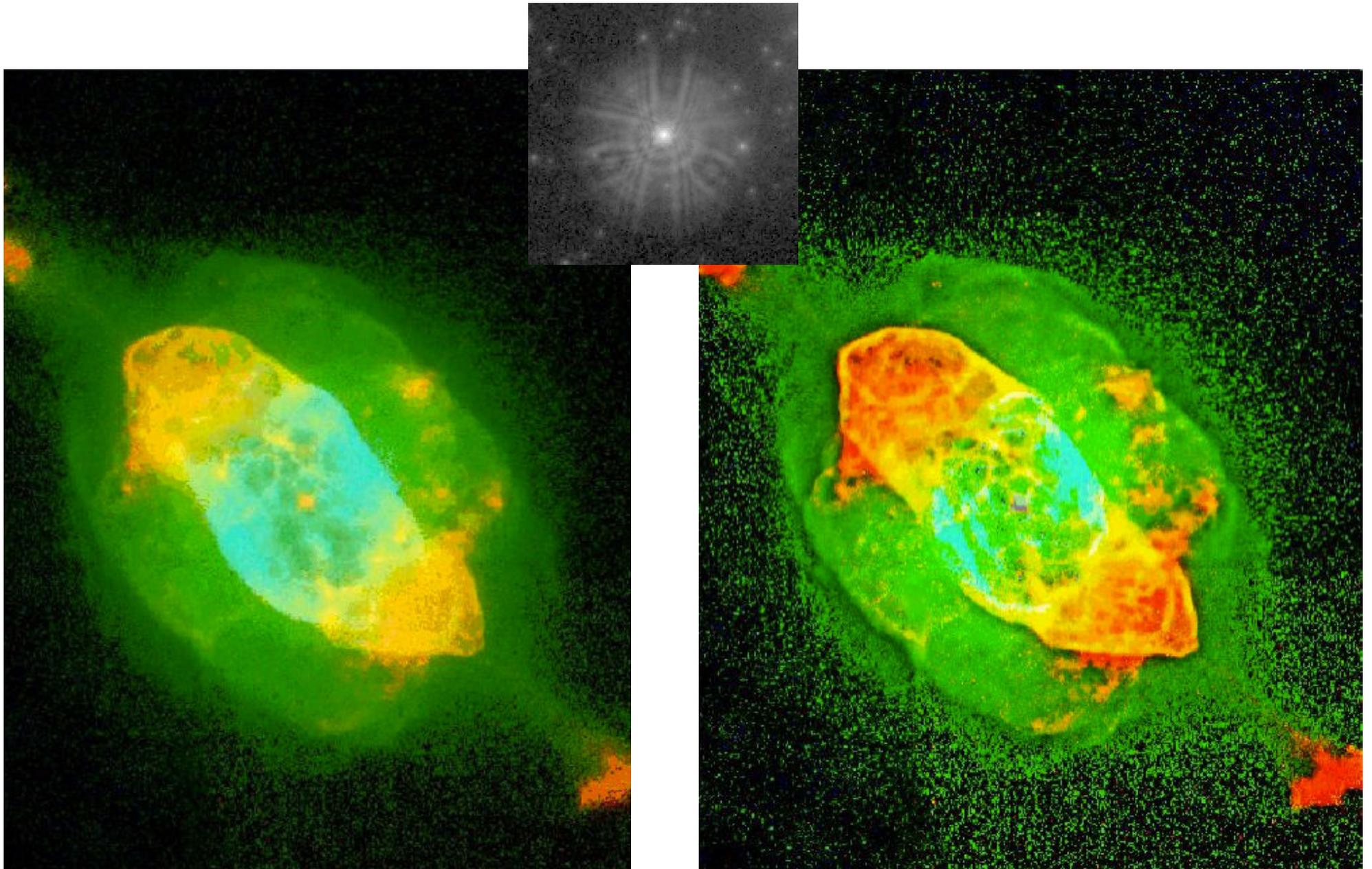
- isolated stars are essentially PSFs



source: [www.apod.nasa.gov](http://www.apod.nasa.gov)

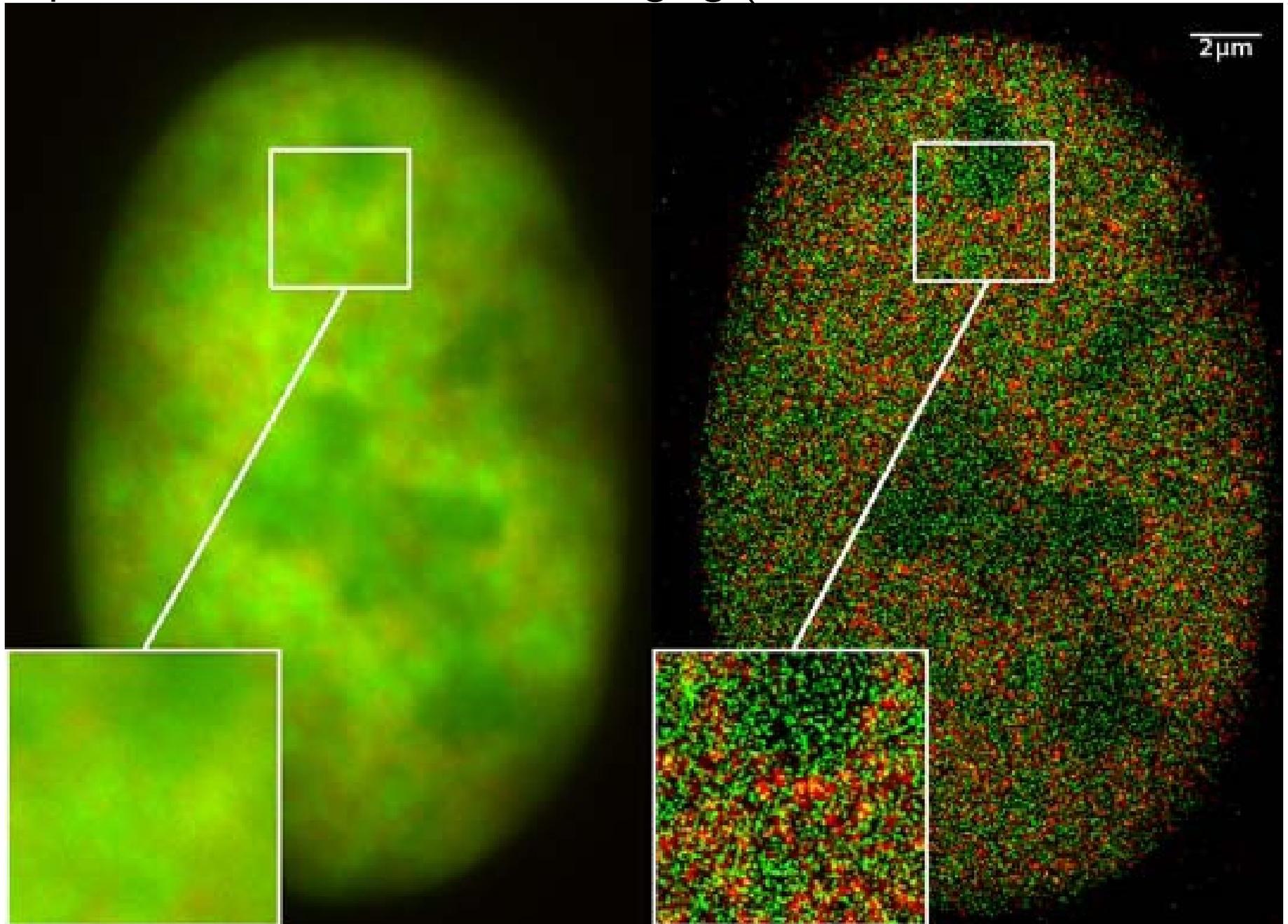
# PSF examples

Hubble flawed mirror deconvolution (correction for spherical aberration)



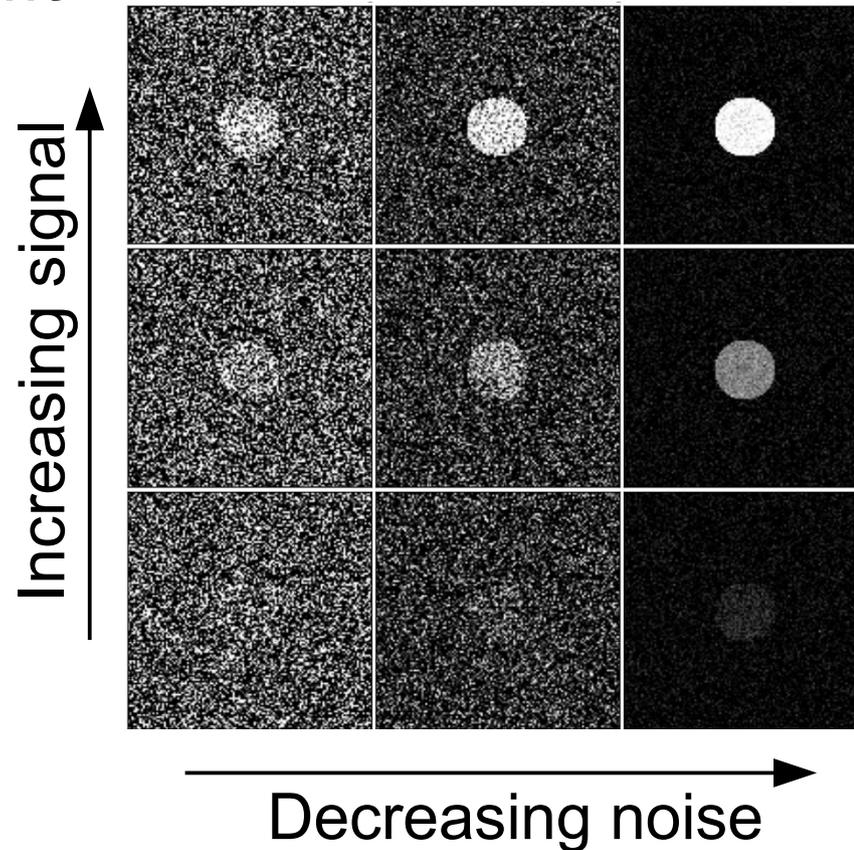
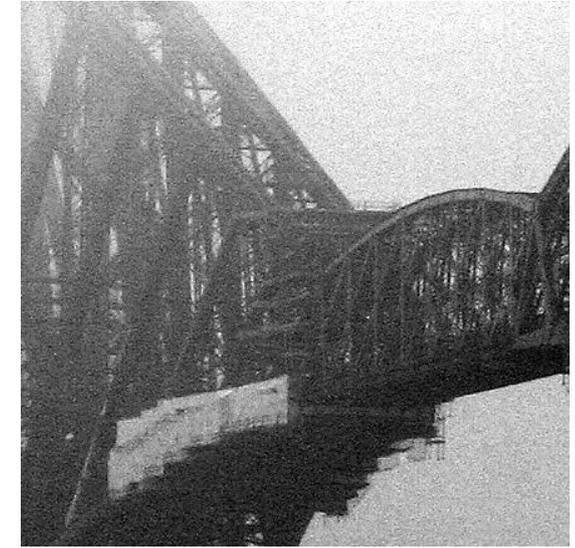
# PSF examples

Super-resolution fluorescence imaging (STORM, STED, PALM, ...)



# Contrast and noise

- Intensity operation:  
higher contrast,  
higher noise
- Contrast-to-noise  
remains constant



# Random variables

- random variable, sample space

$X$        $\Omega$

probability of measuring  $x$ :  $p(x)$

$$p(\Omega) = 1$$

- probability density function

$$p(a < x < b) = \int_a^b p(x) dx$$

↖ probability density

$$\int_{\Omega} p(x) dx = 1$$

- expectation value

$$\langle f \rangle = \int_{\Omega} f(x) p(x) dx$$

mean:  $\langle x \rangle = \int x p(x) dx$

- variance

$$\text{var}(x) = \langle (x - \langle x \rangle)^2 \rangle = \langle x^2 \rangle - \langle x \rangle^2$$

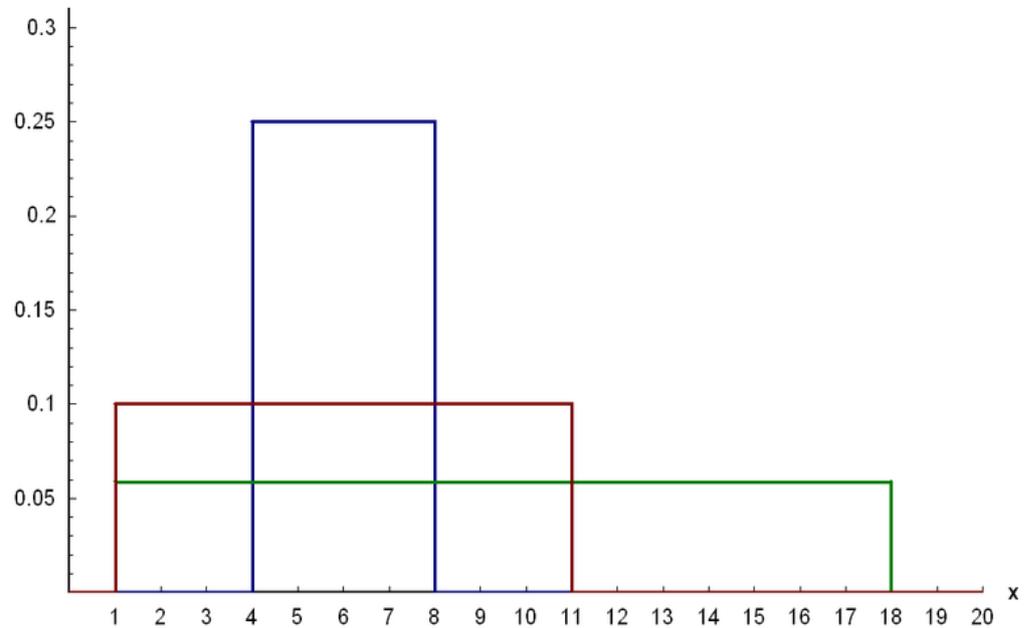
# Uniform distribution

- probability density function

$$f(x) = \begin{cases} \frac{1}{b-a} & a < x < b \\ 0 & \text{otherwise} \end{cases}$$

- expectation value

$$\langle x \rangle = \frac{1}{2}(a+b)$$

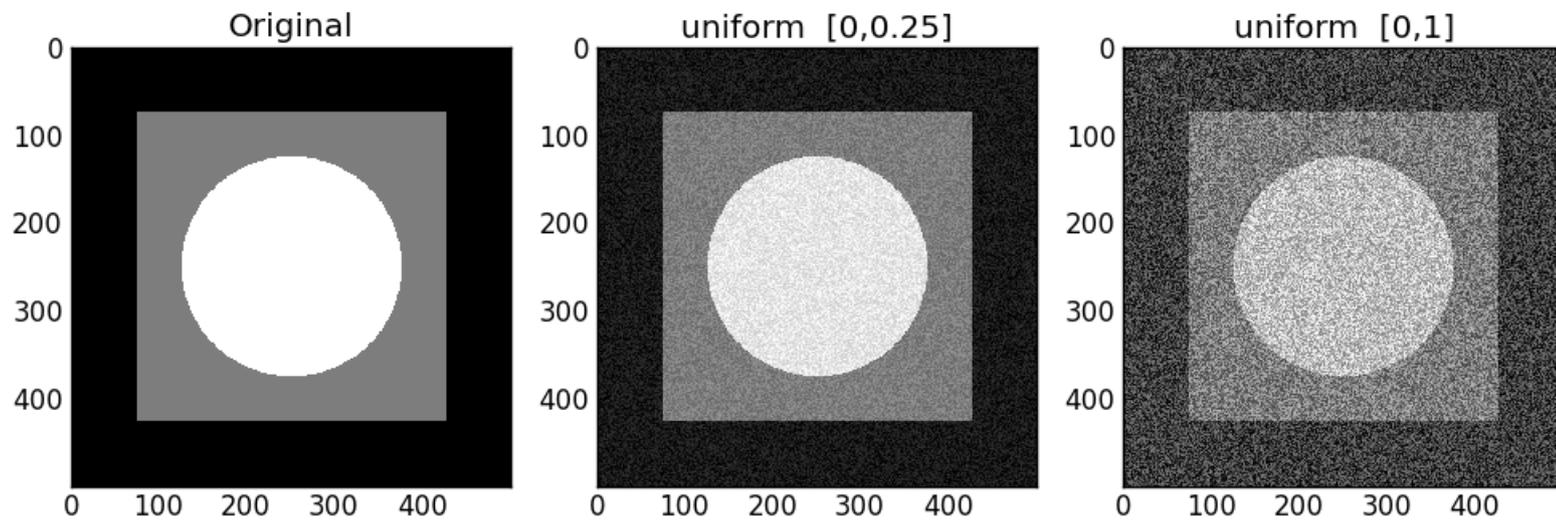


- variance

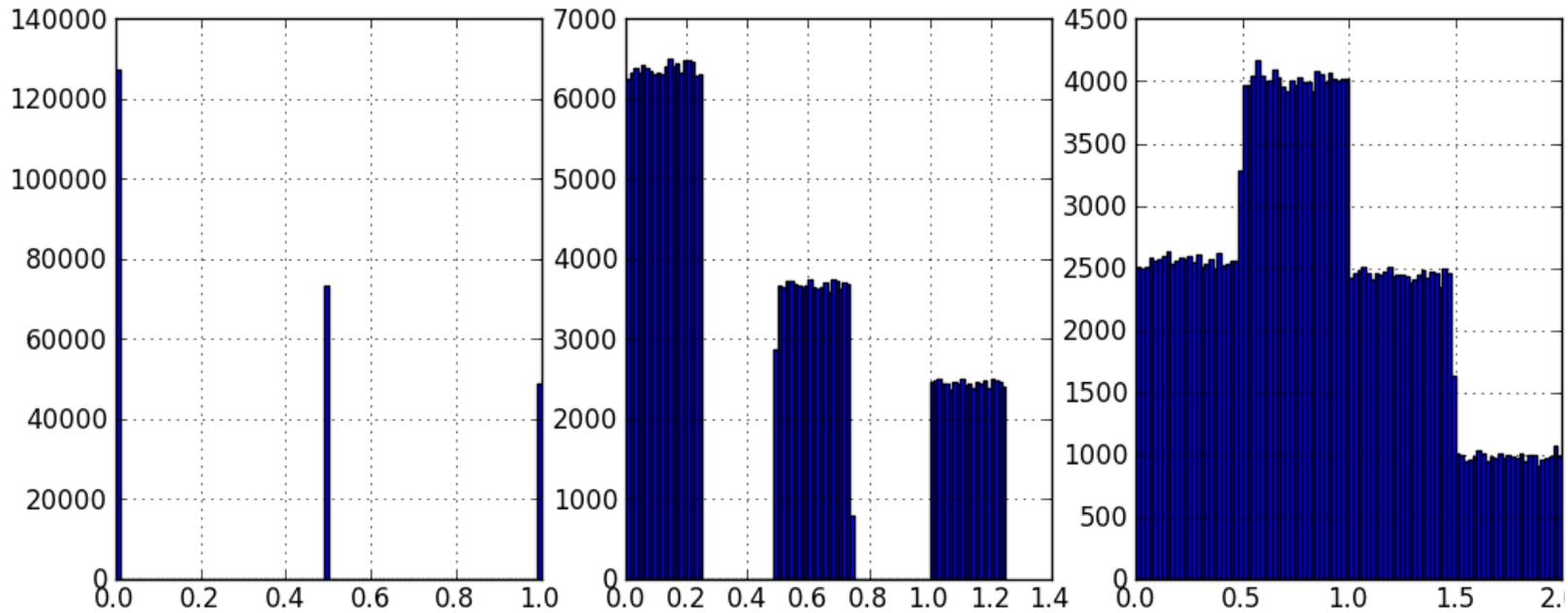
$$\text{var } x = \frac{(b-a)^2}{12}$$

- occurrence → not very common in imaging, but useful to build other distributions

# Uniform distribution



*histogram*



# Gaussian distribution

- probability density function

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

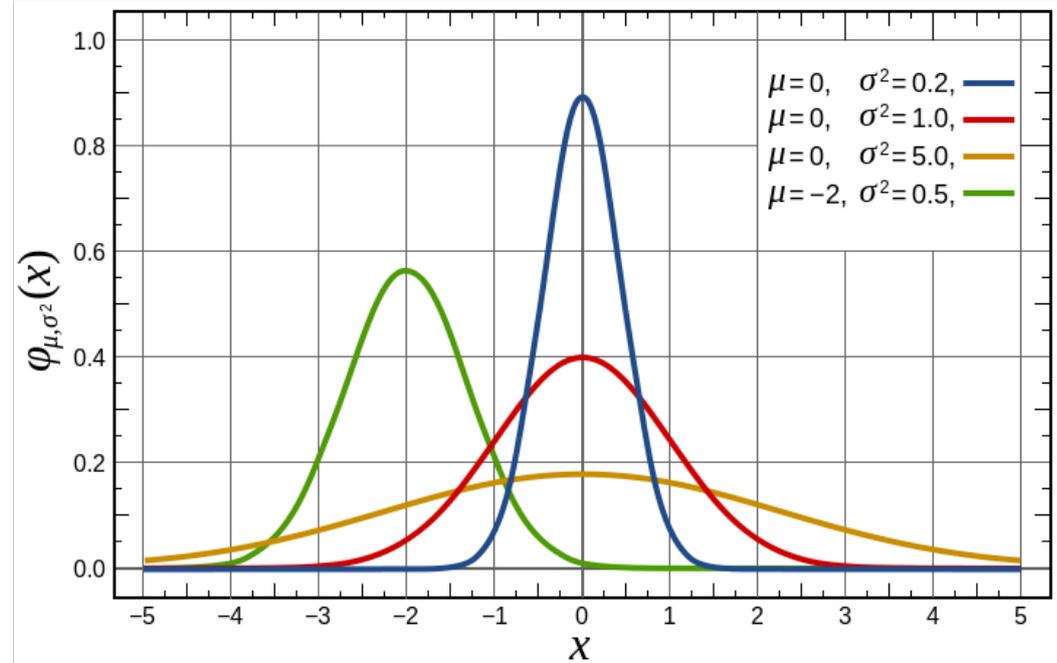
- expectation value

$$\langle x \rangle = \mu$$

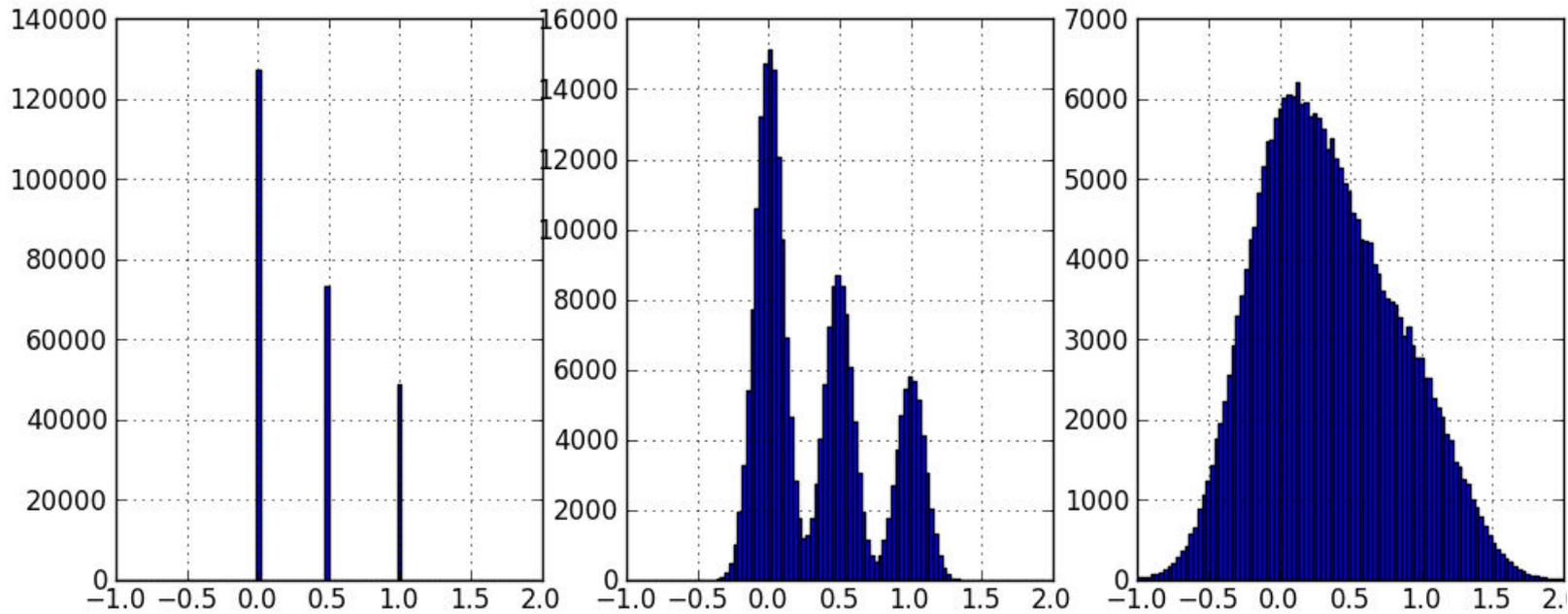
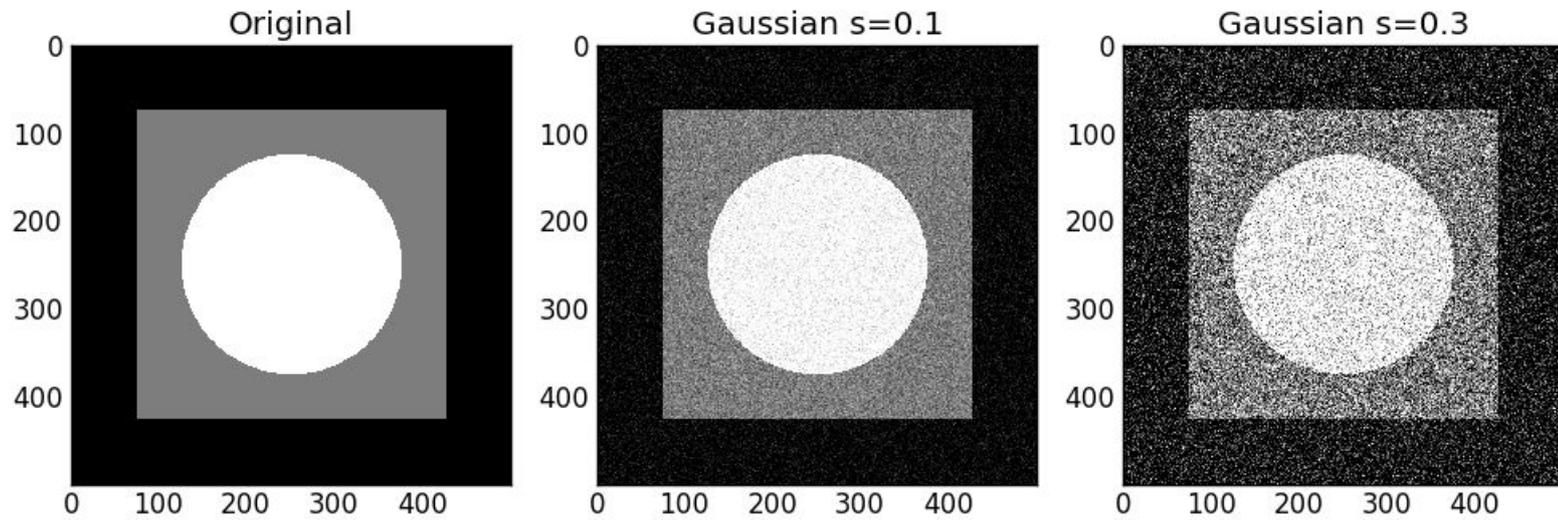
- variance

$$\text{Var } X = \sigma^2$$

- occurrence *Very common*



# Gaussian distribution



# Poisson distribution

- probability mass function

$$p(n) = \frac{1}{n!} \lambda^n e^{-\lambda}$$

- expectation value

$$\langle n \rangle = \lambda$$

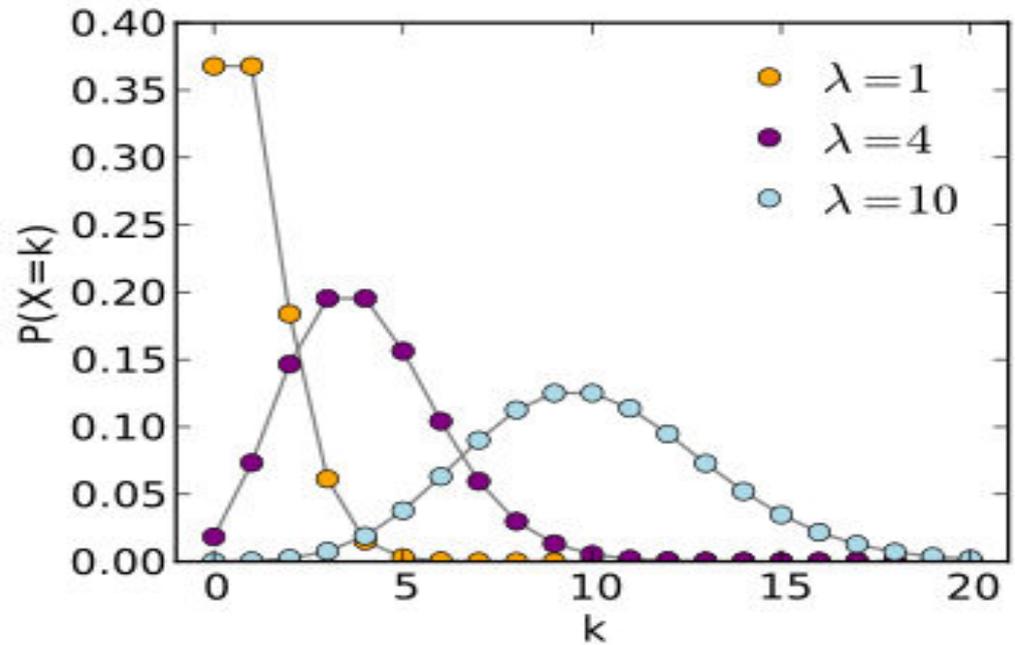
- variance

$$\text{var } n = \lambda$$

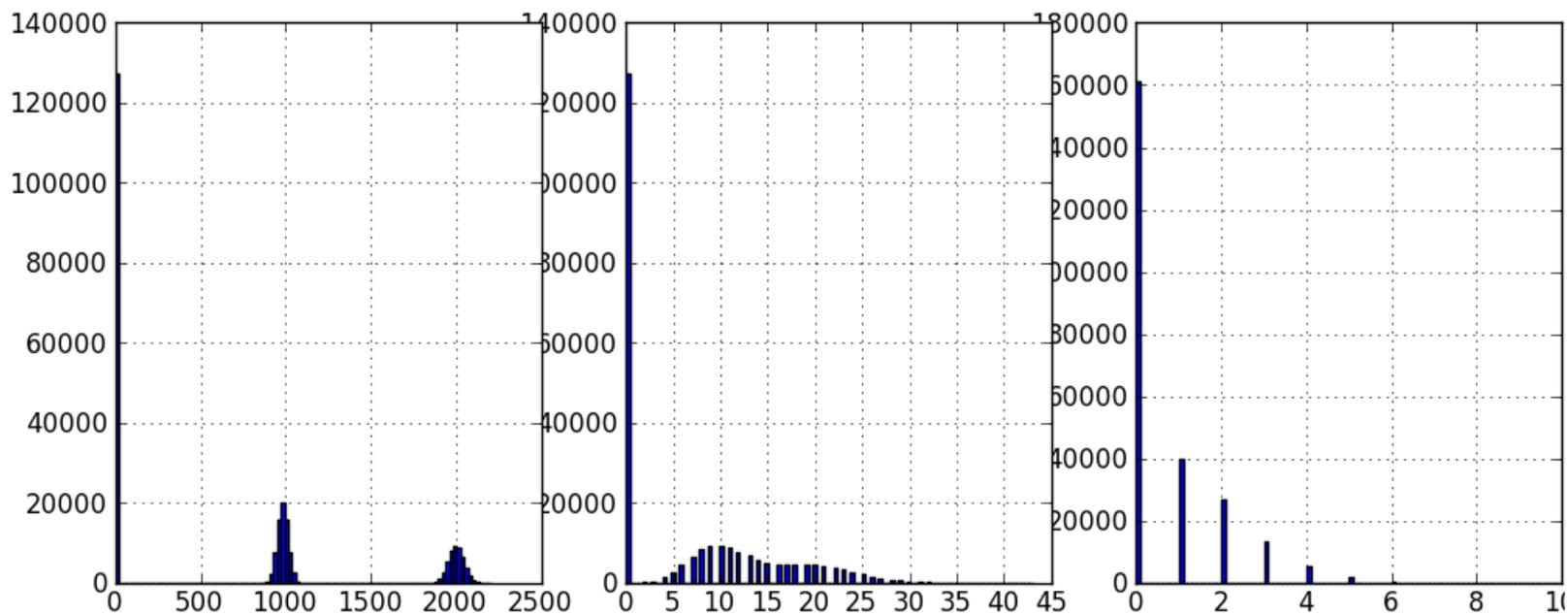
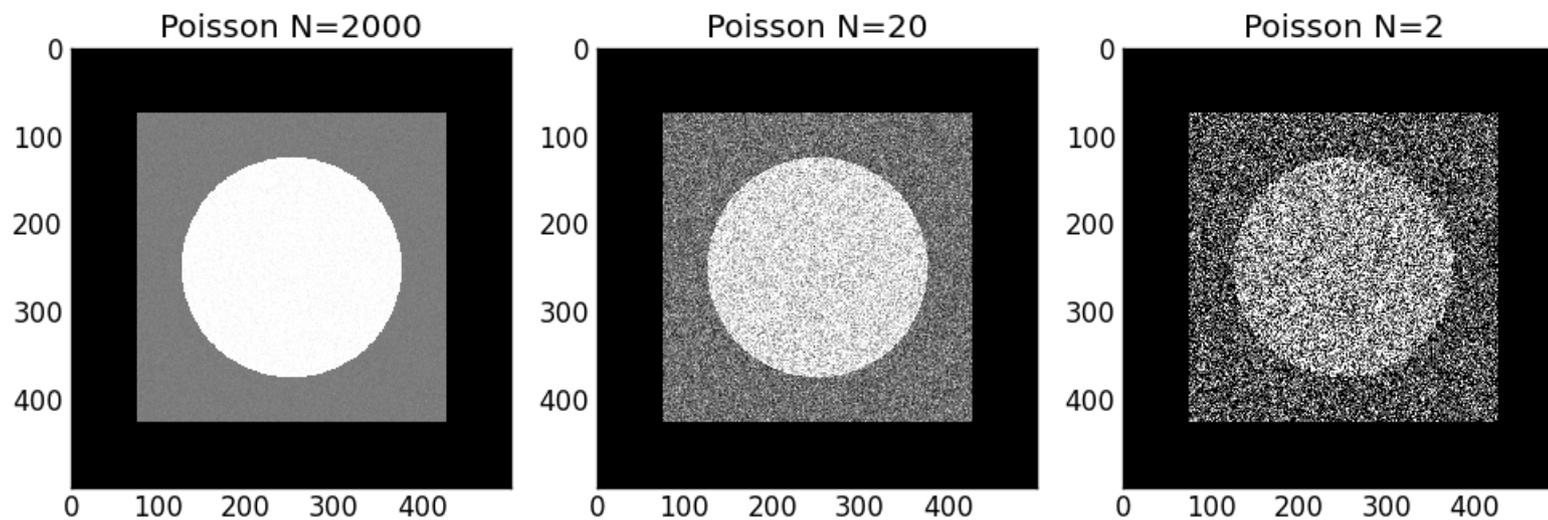
- occurrence

counting process (of independent events)

photons, electrons "shot noise"



# Poisson distribution

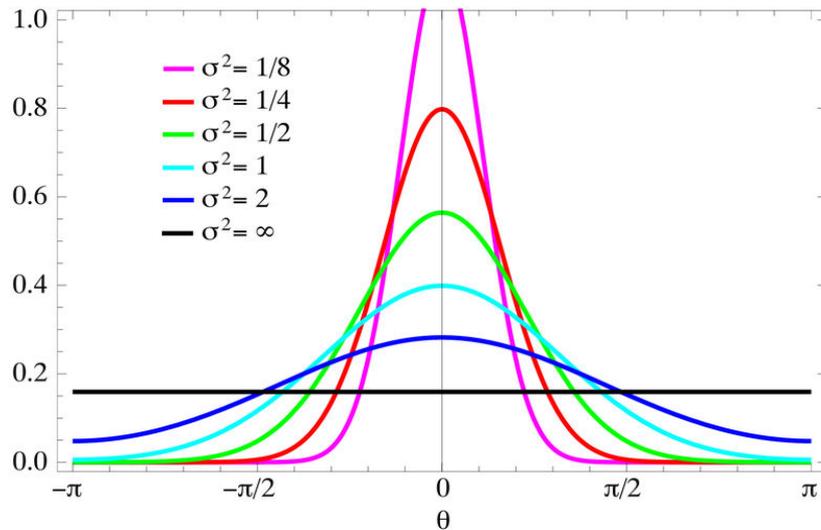


# Poisson distribution

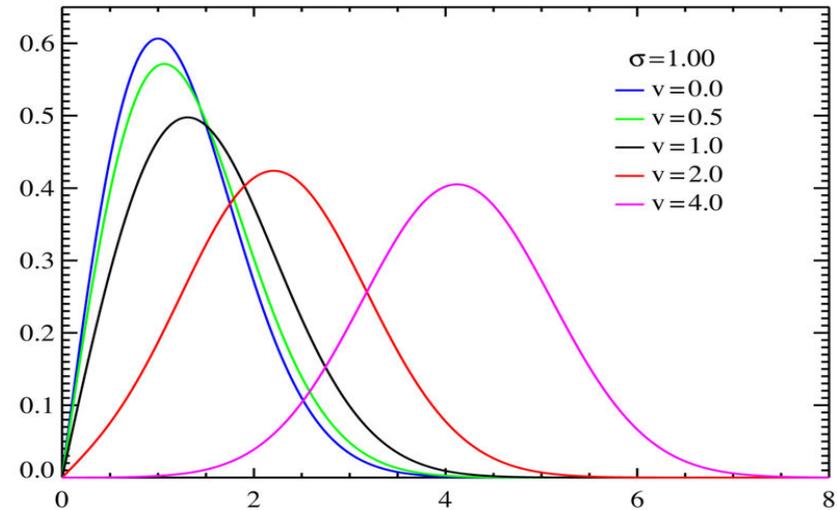


# Many other distributions

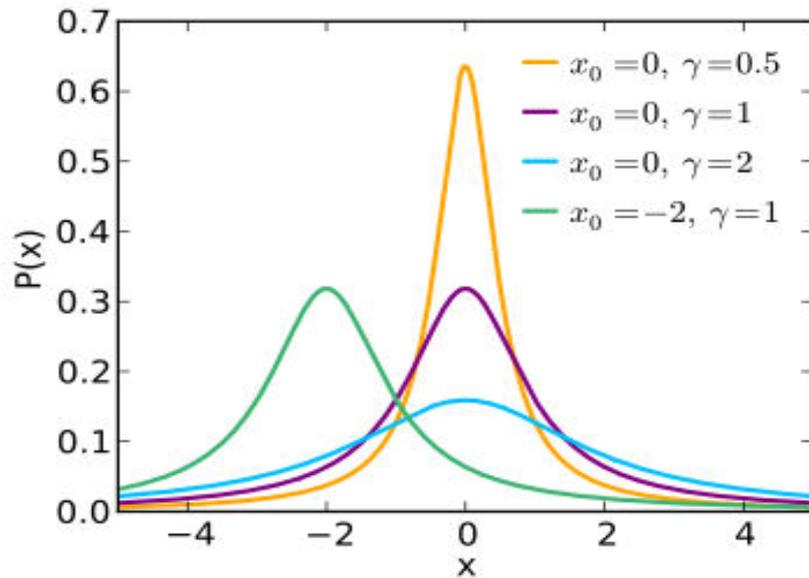
## Wrapped normal distribution



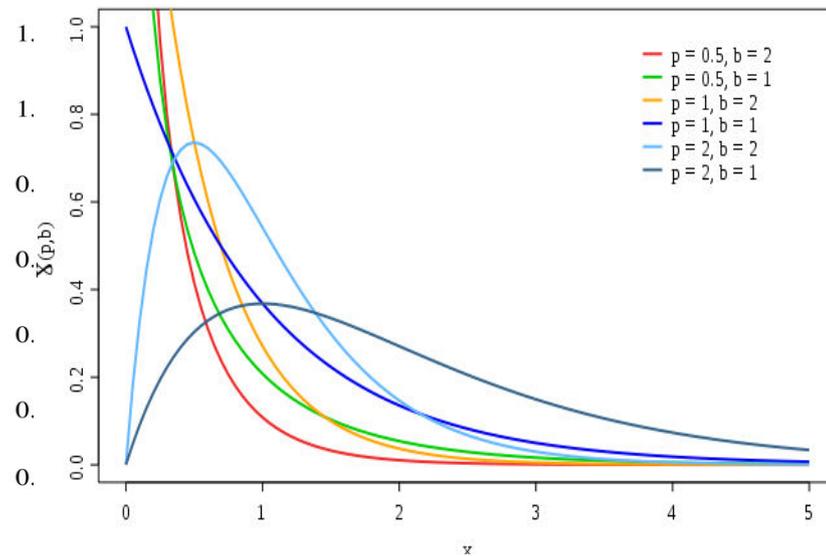
## Rice distribution



## Lorentz distribution



## Gamma distribution



# Detector noise (CCD)

- Various sources:
  - shot noise (photon statistics, Poisson)
  - dark current (thermal electronic fluctuations in semiconductor, Poisson)
  - readout noise (fluctuations during amplification and digitization, Gauss)
  - many other imperfections ...
- dark frame measures detector noise, hot pixels, dead pixels
- bright frame measures gain differences and imperfections (dust, etc)

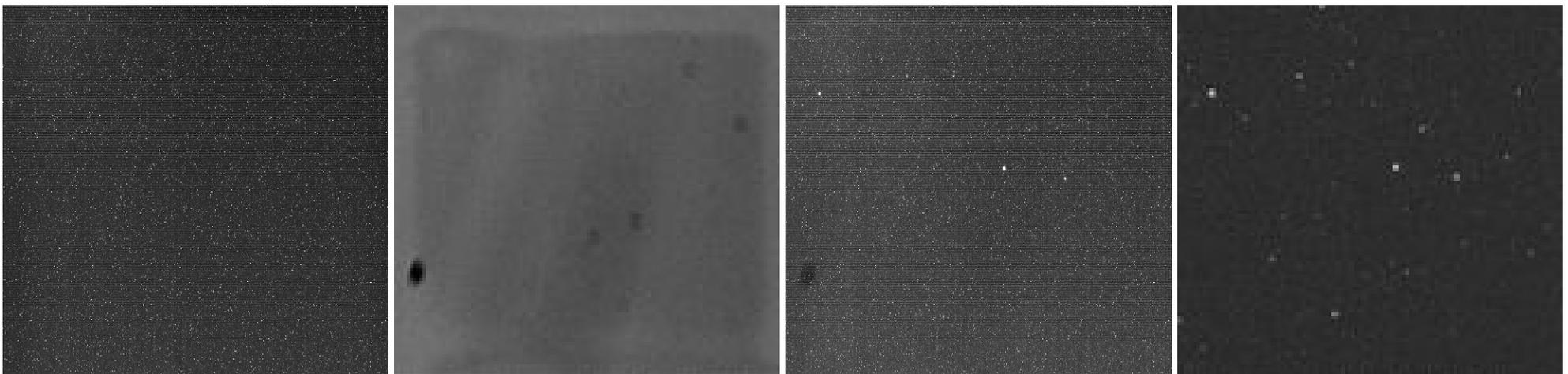
"flat field"

*dark frame*

*bright frame*

*raw image*

*calibrated image*



# Correlation & Convolution

\* Convolution:  $f * g = \int_{-\infty}^{\infty} f(x') g(x-x') dx'$

Fourier:  $F\{f * g\} = F \cdot G$

\* Correlation:  $f \otimes g = \int_{-\infty}^{\infty} f^*(x') g(x+x') dx'$

$$f \otimes g = \int_{-\infty}^{\infty} dx' \int_{-\infty}^{\infty} F^*(u) e^{-2\pi i u x'} du \int_{-\infty}^{\infty} G(u') e^{2\pi i u' (x+x')} du'$$

$$= \iint_{-\infty}^{\infty} du du' F^*(u) G(u') e^{2\pi i u' x} \underbrace{\int_{-\infty}^{\infty} dx' e^{2\pi i x' (u-u')}}_{\delta(u-u')}$$

$$\int_{-\infty}^{\infty} du F^*(u) G(u) e^{2\pi i u x} = F^{-1}\{F^* G\}$$

# Noise power spectrum

- power spectrum of pure noise image

$$NPS = \left\langle \left| \mathcal{F} \{ n(x,y) \} \right|^2 \right\rangle$$

ensemble average

random variable (multivariate)

$$N(u,v) = \mathcal{F} \{ n(x,y) \}$$

- connection to auto-correlation

$$|N(u,v)|^2 = N^* N$$

$$\mathcal{F}^{-1} \{ NPS \} = \mathcal{F}^{-1} \{ \langle N^* N \rangle \} = \langle n \otimes n \rangle$$

noise autocorrelation

# Measuring NPS

1) measure multiple realizations of random variable  $n$   
 $\{n_i\}$

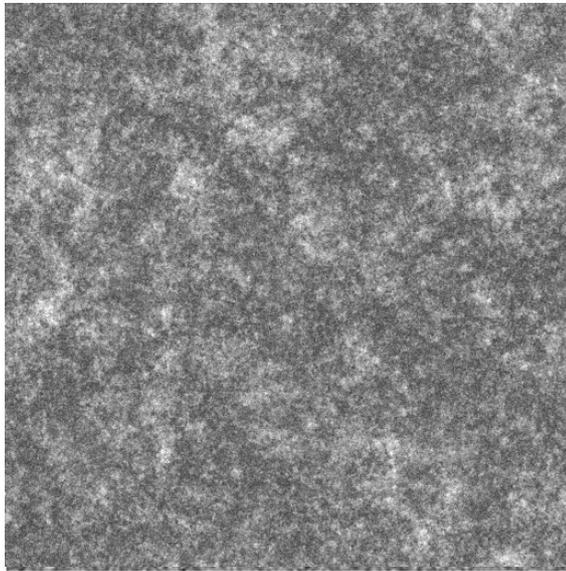
$$2) N_i(u, v) = \mathcal{F}\{n_i(x, y)\}$$

$$3) \langle |N(u, v)|^2 \rangle = \frac{1}{M} \sum_i |N_i(u, v)|^2 = \text{NPS}$$

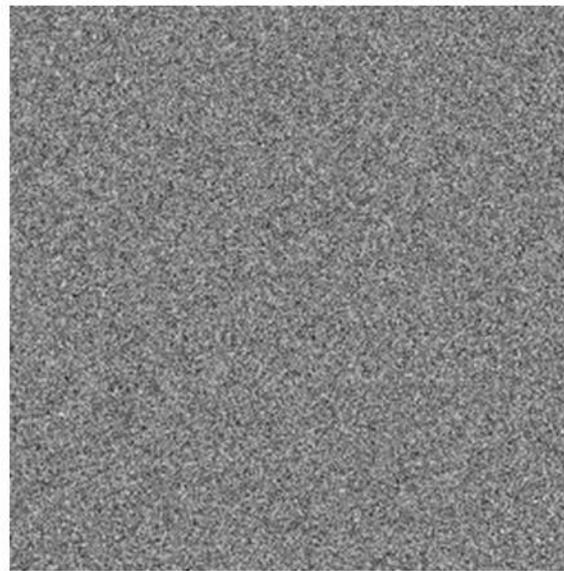
4)  $\mathcal{F}^{-1}\{\text{NPS}\}$ : estimate of noise autocorrelation

# Noise power spectrum

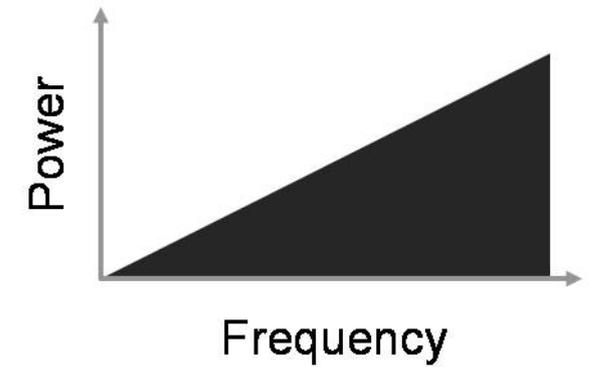
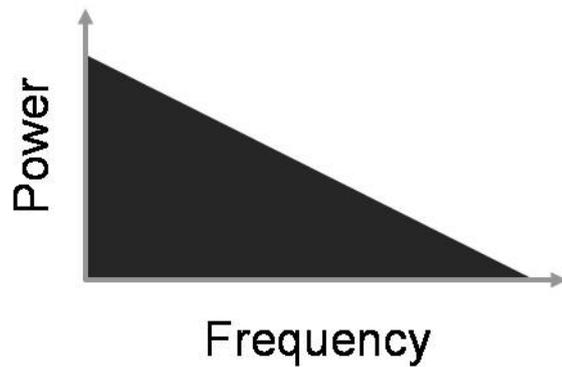
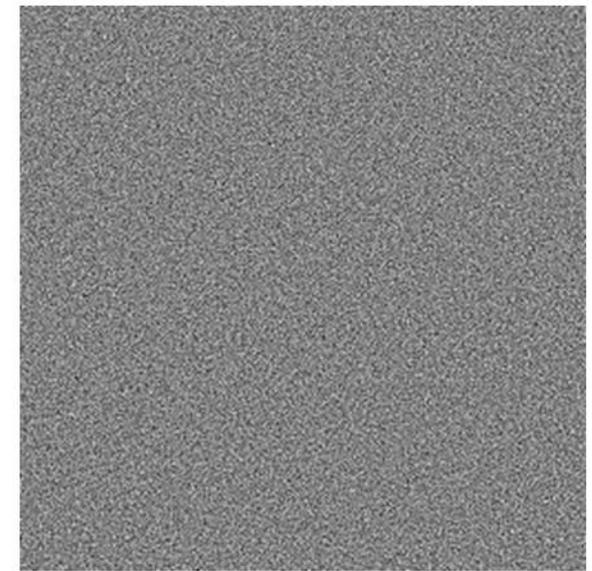
Red noise



White noise

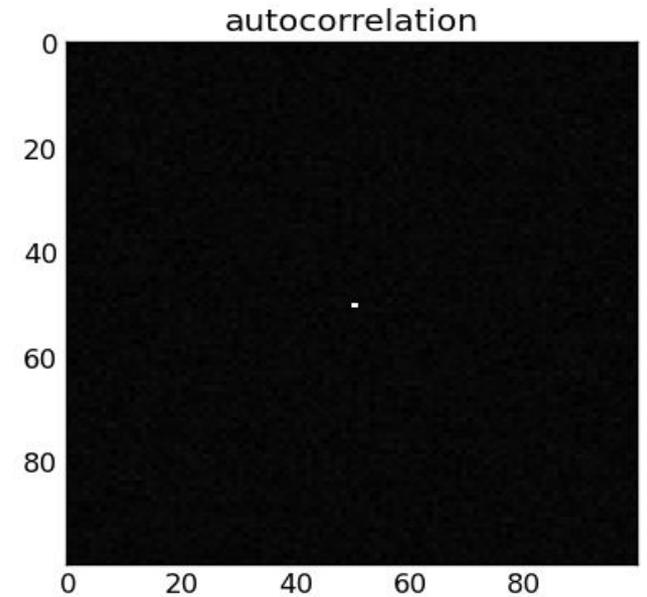
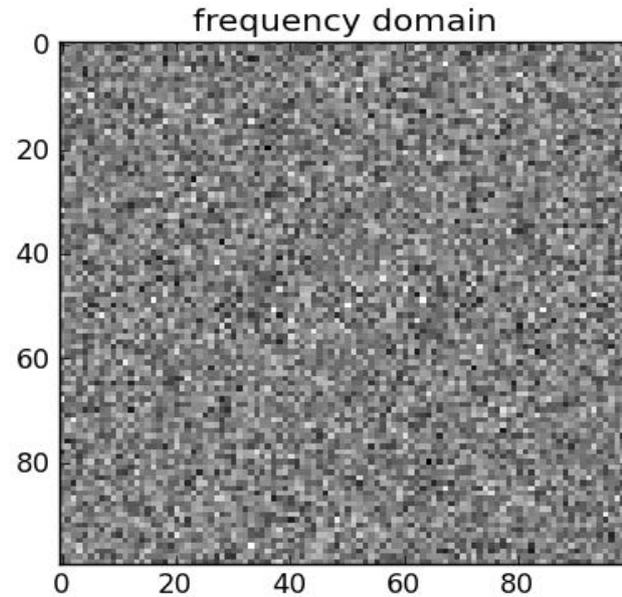
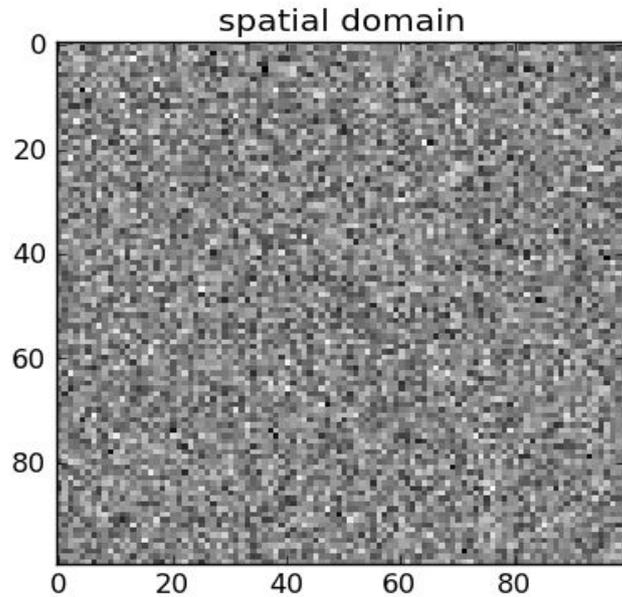


Blue noise



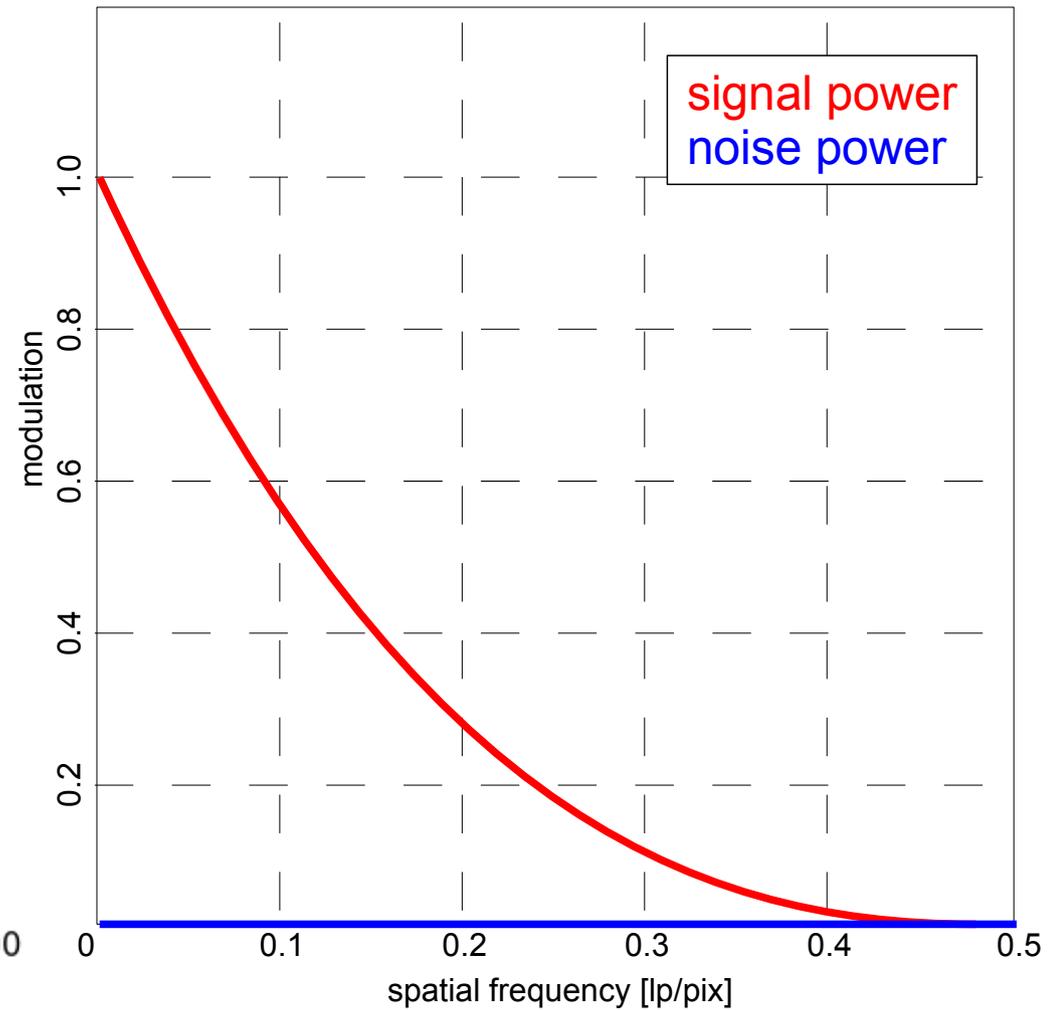
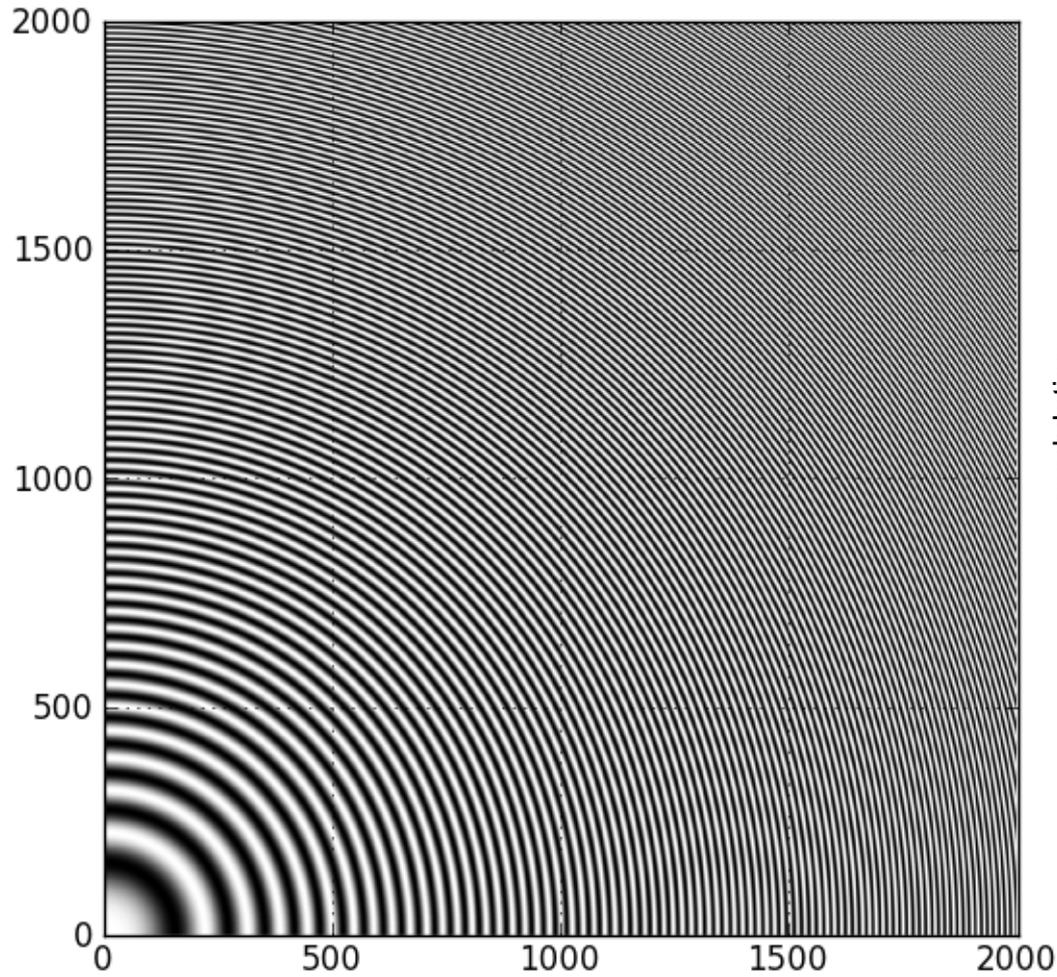
source: [http://scien.stanford.edu/pages/labsite/2008/psych221/projects/08/AdamWang/project\\_report.htm](http://scien.stanford.edu/pages/labsite/2008/psych221/projects/08/AdamWang/project_report.htm)

# White noise

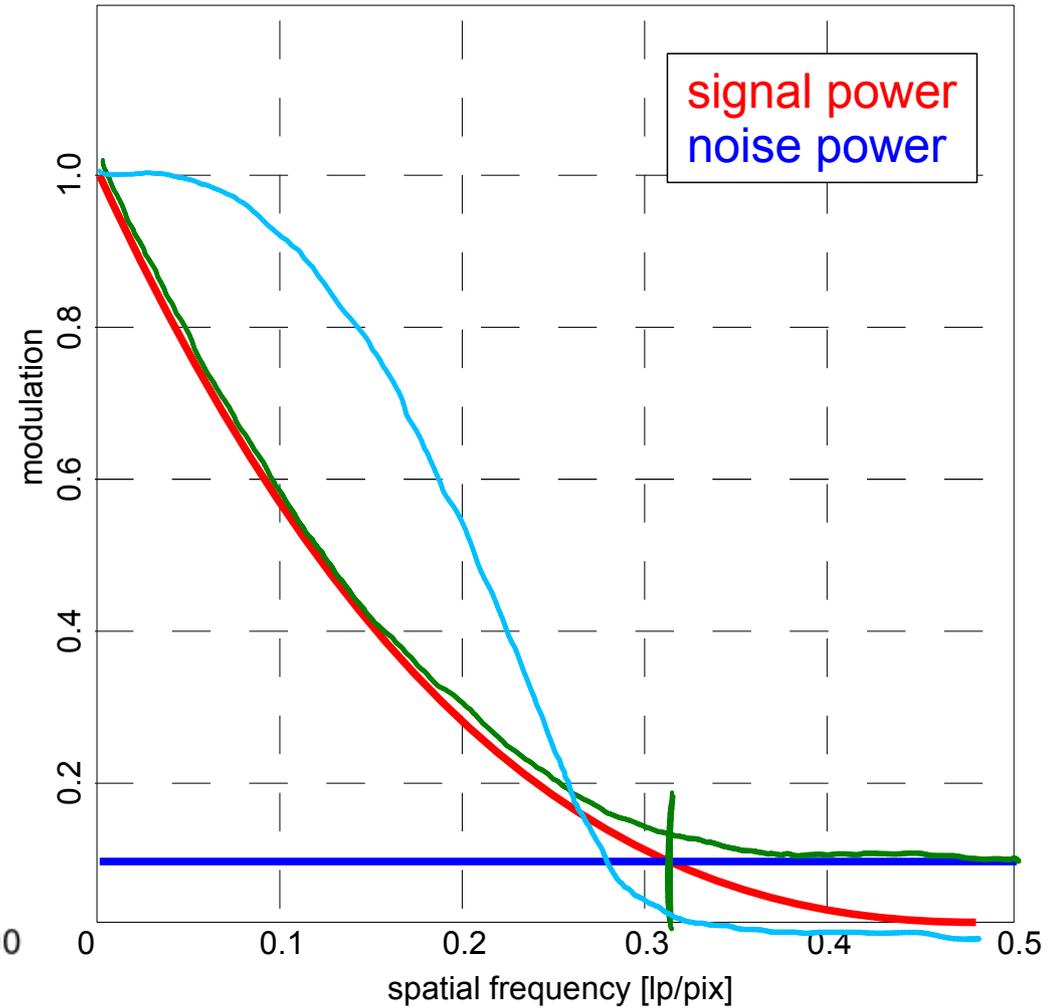
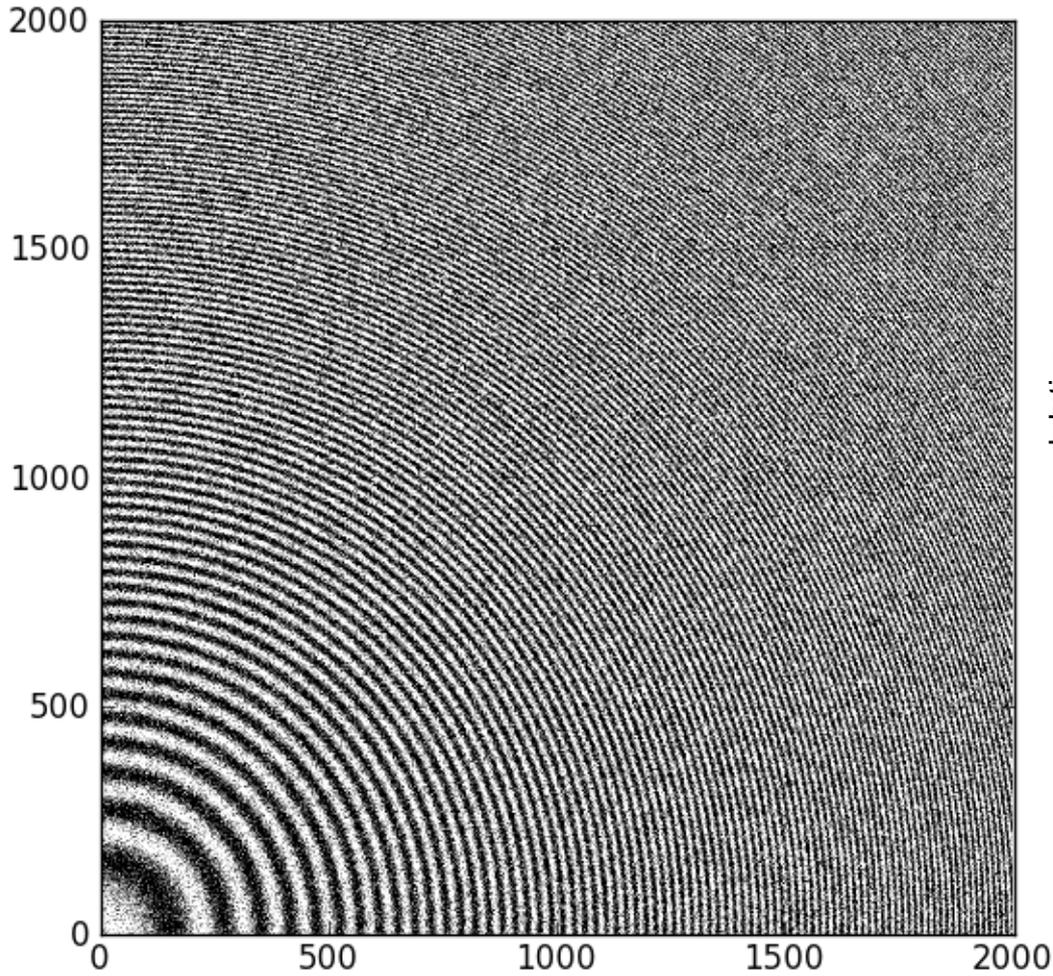


- white noise in spatial domain equals white noise in frequency domain
- white noise is perfectly uncorrelated
- all other types of noise are correlated to some degree
- white noise is an idealization

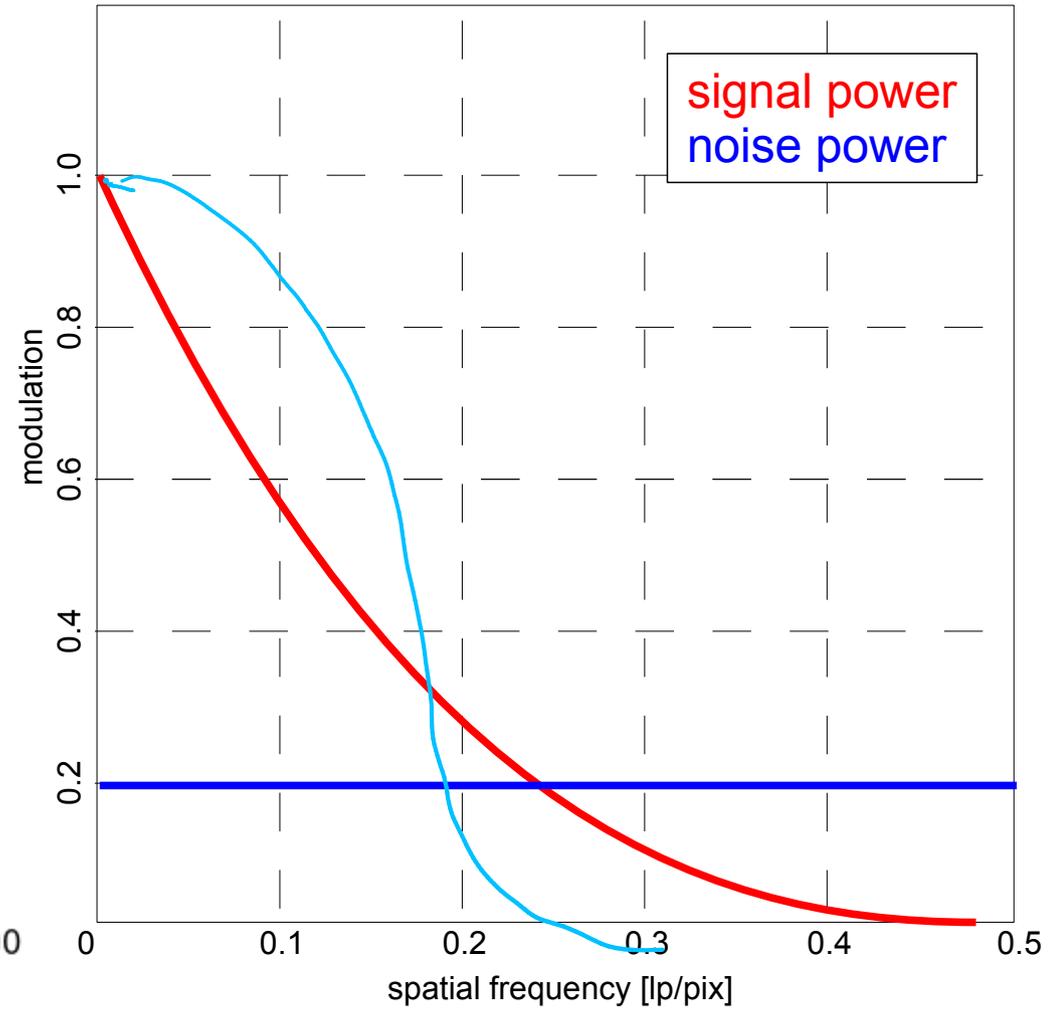
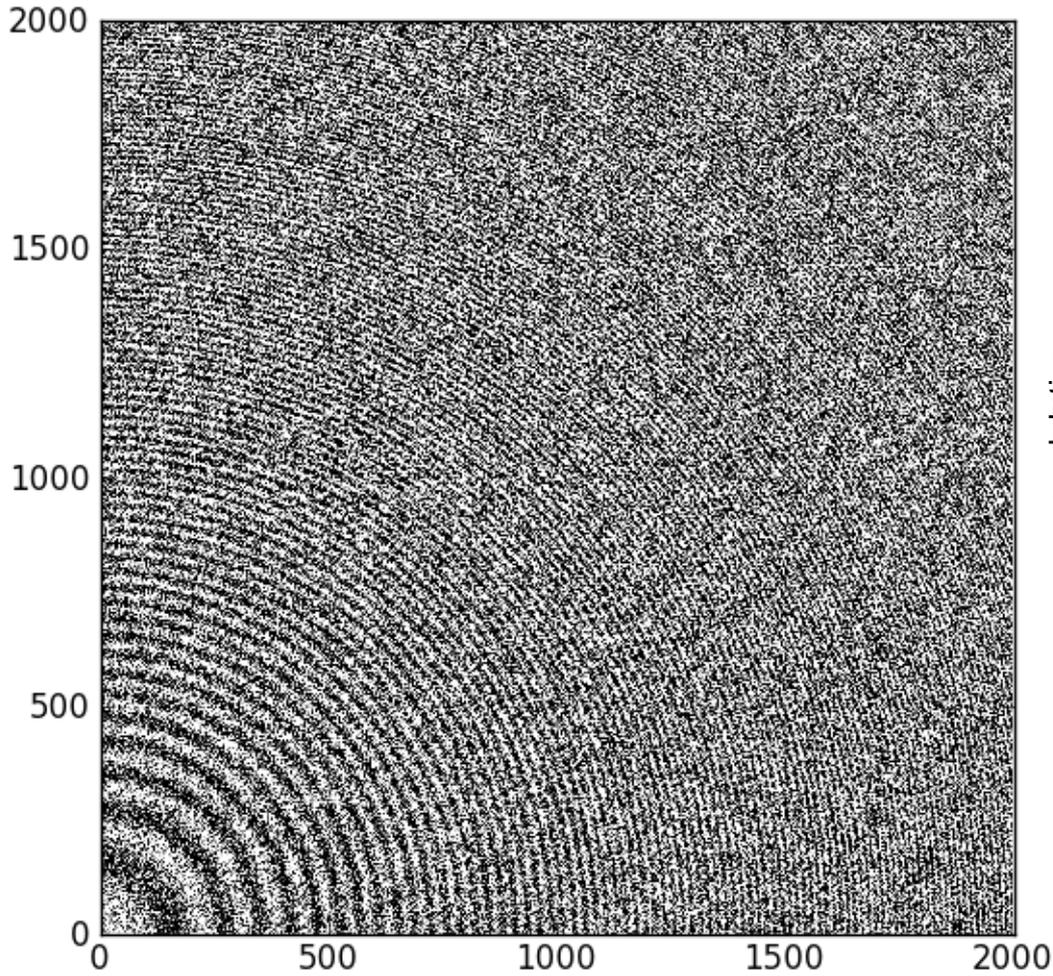
# Signal power vs. noise power



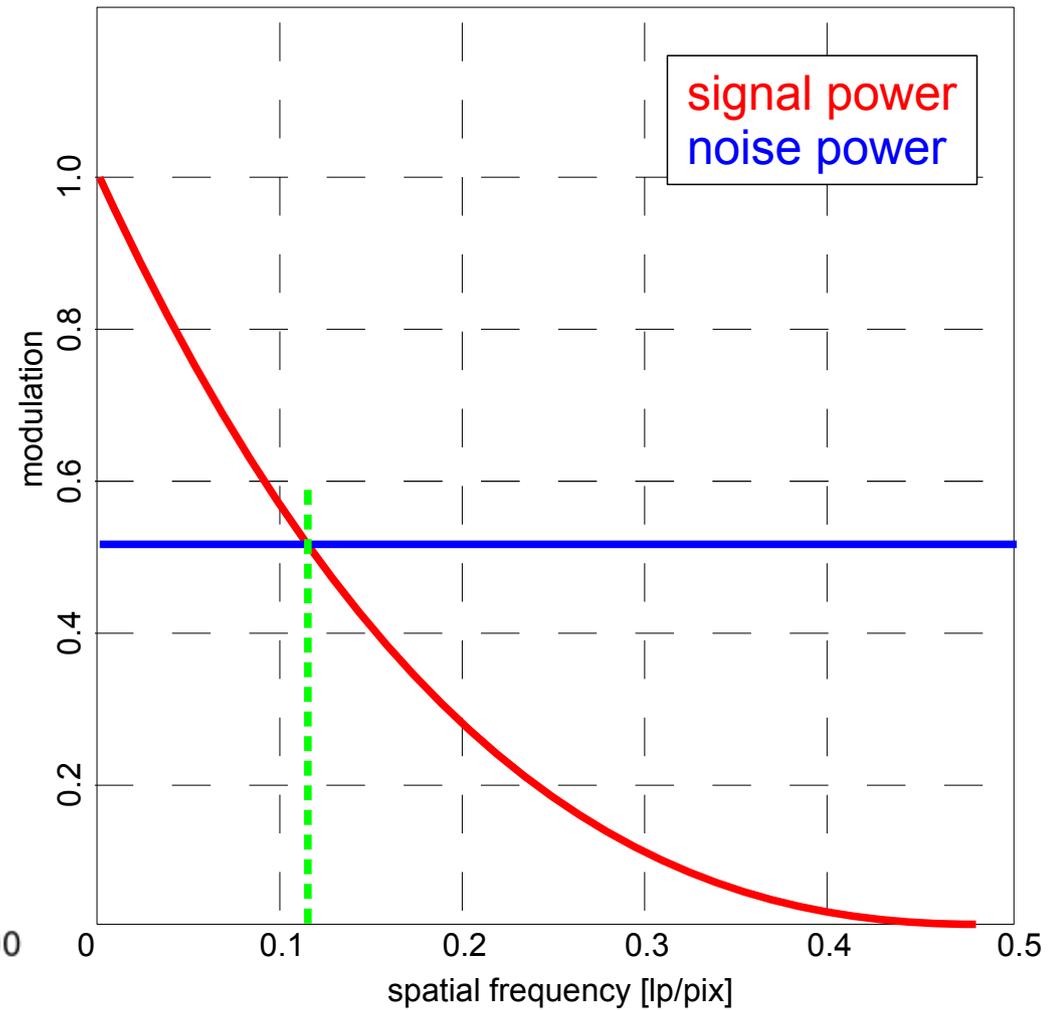
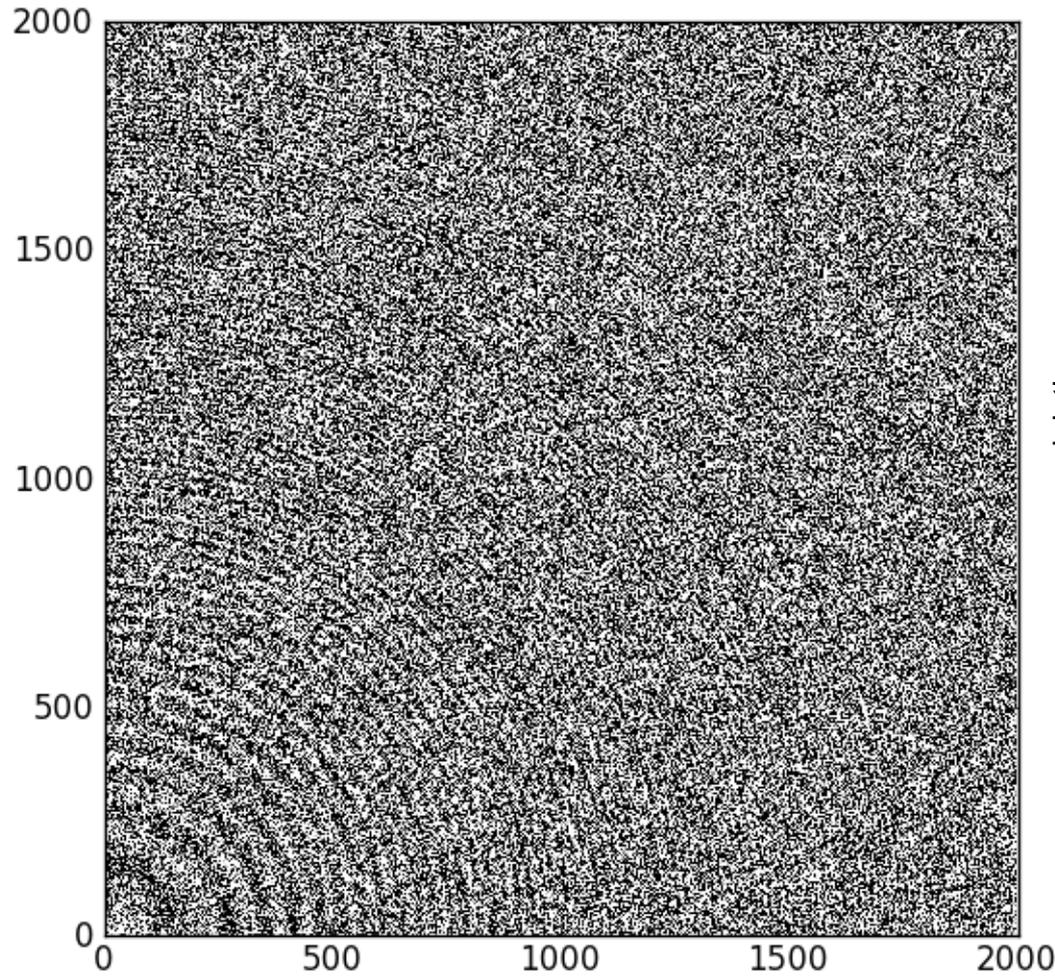
# Signal power vs. noise power



# Signal power vs. noise power

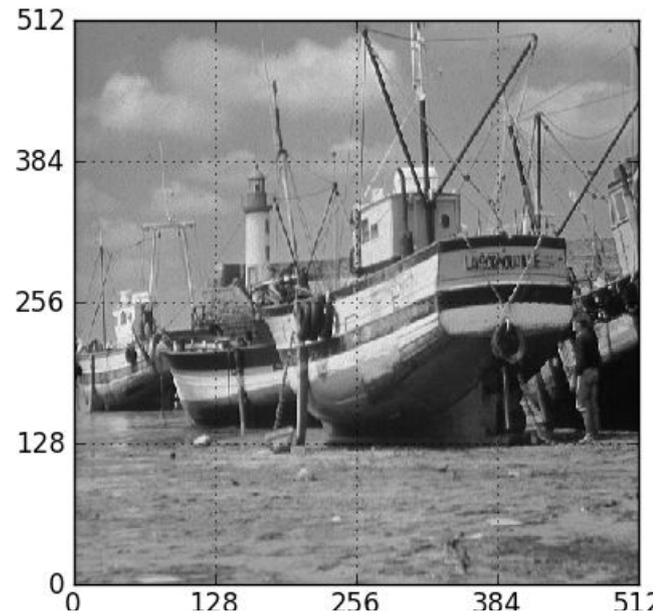
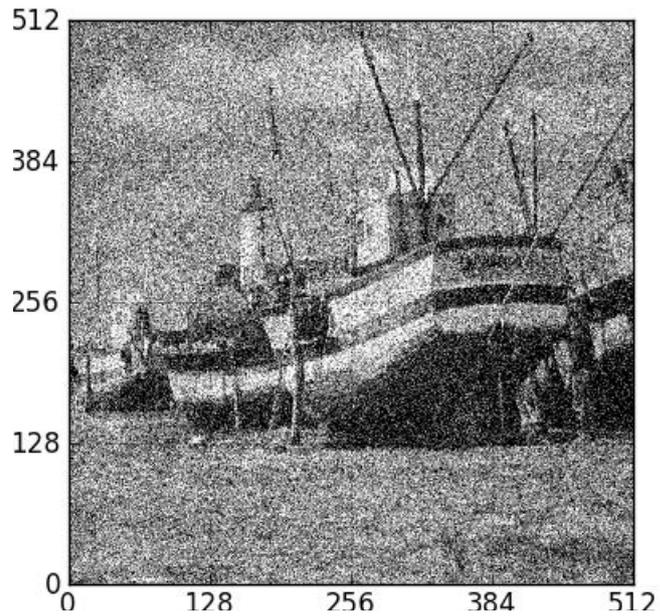
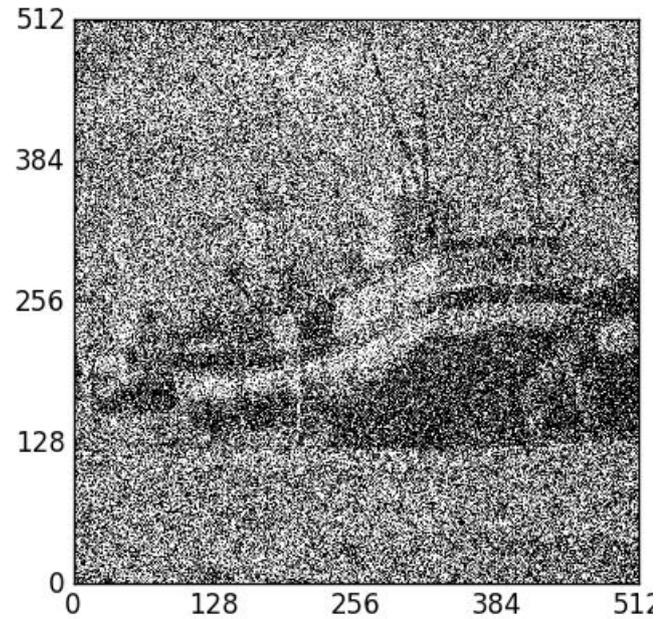
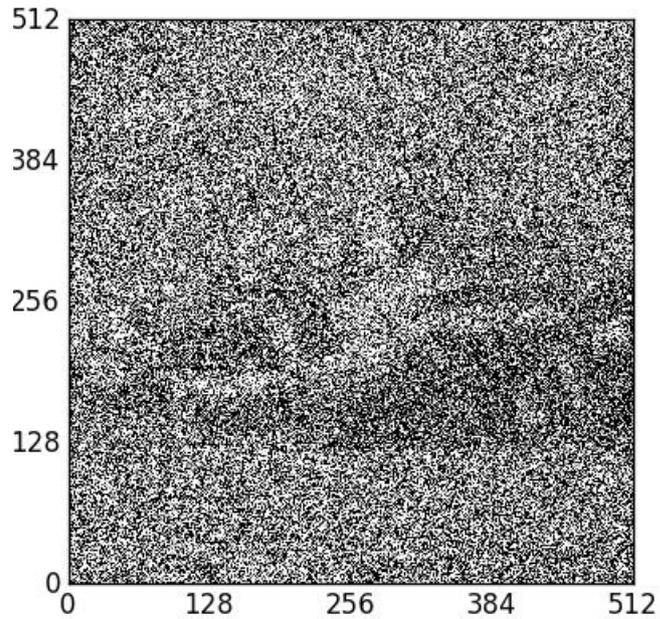


# Signal power vs. noise power



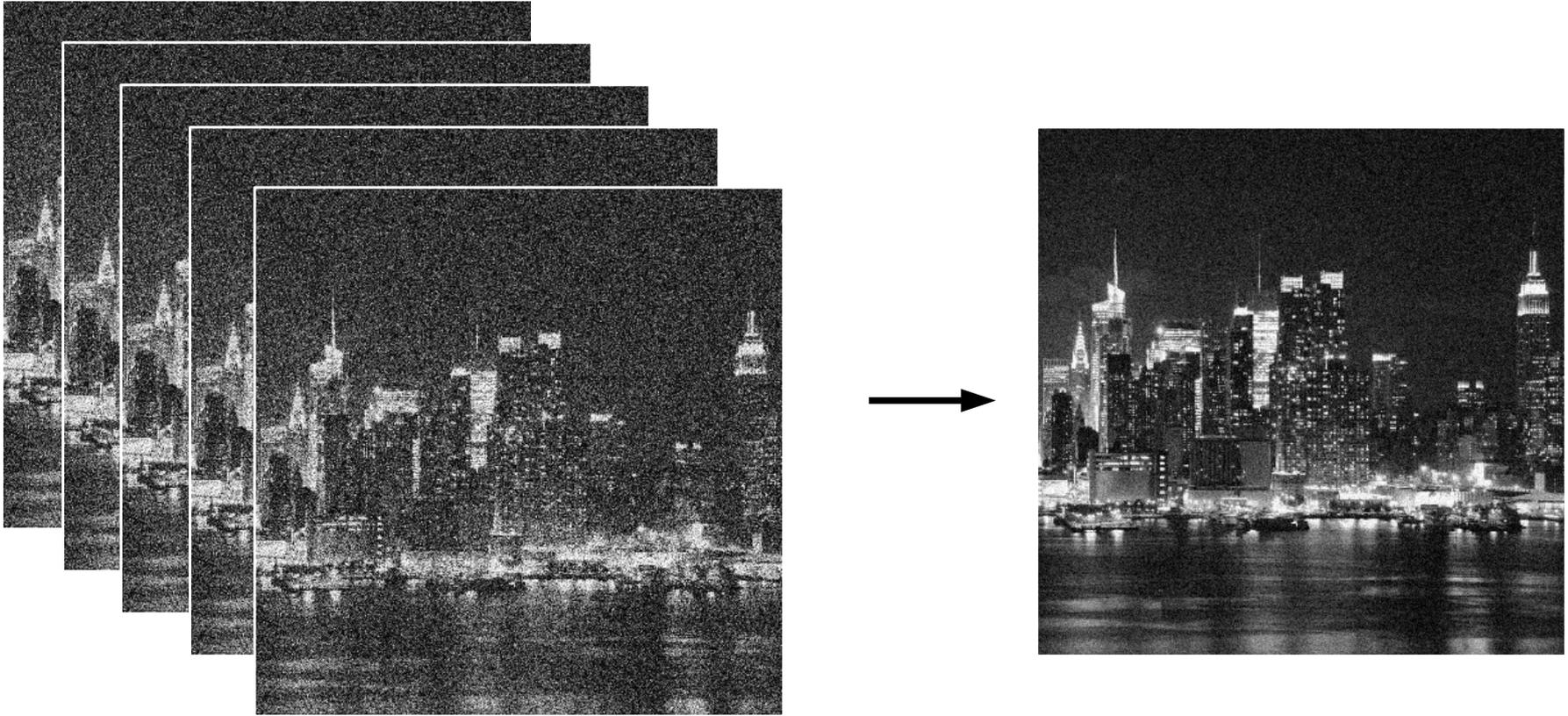
- Noise power exceeds signal power for high frequencies
- Small scale image details are lost in noise first

# Signal power vs. noise power



# Noise reduction by averaging

- Average multiple images



- requirement: additive noise, zero mean

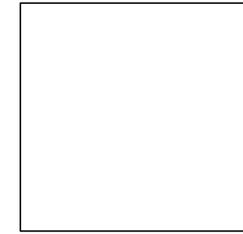
# Denoising by linear filtering

- use spatial convolution or frequency filtering to reduce noise
- noise reduction possible, but at cost of sharpness
- trade-off between noise reduction and resolution
- need fancier methods

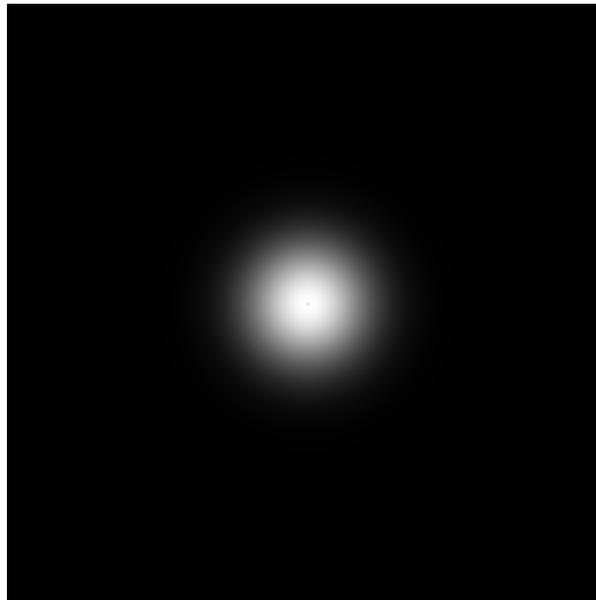
original



convolution kernel



frequency filter

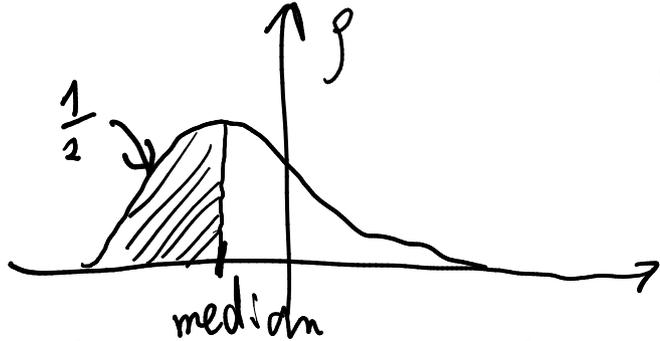


Resulting image



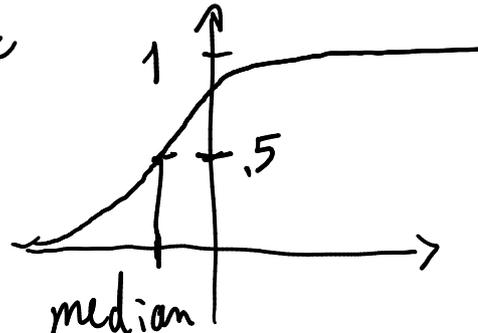
# Median filtering

- Use median as estimator for fat tail distributions



cumulative

$\rightarrow \int$



- less sensitive to outliers in pixel ensemble, better edge preservation

Salt and pepper noise



Gauss sigma=1 pixel



Median 1 pixel



# Median filtering

1x Gauss



2x Gauss



5x Gauss



1x Median



2x Median



5x Median



*edge-preserving to some extent*

# Common abbreviations

Abbreviation	Name	Definition
IRF	Impulse response function	Linear operator map of delta function
PSF	Point spread function	Image of point object (optical IRF)
OTF	Optical transfer function	Fourier transform of PSF
PTF	Phase transfer function	Phase part of OTF
MTF	Modulation transfer function	Amplitude of OTF
CTF	Contrast transfer function	MTF for non-sinusoidal objects
PDF	Probability density function	Probability distribution for a given random variable
SPS	Signal power spectrum	Amplitude squared of signal F.T.
NPS	Noise power spectrum	Amplitude squared of noise F.T.
SNR	Signal to noise ratio	Mean signal / mean noise
CNR	Contrast to noise ratio	Mean contrast / mean noise