



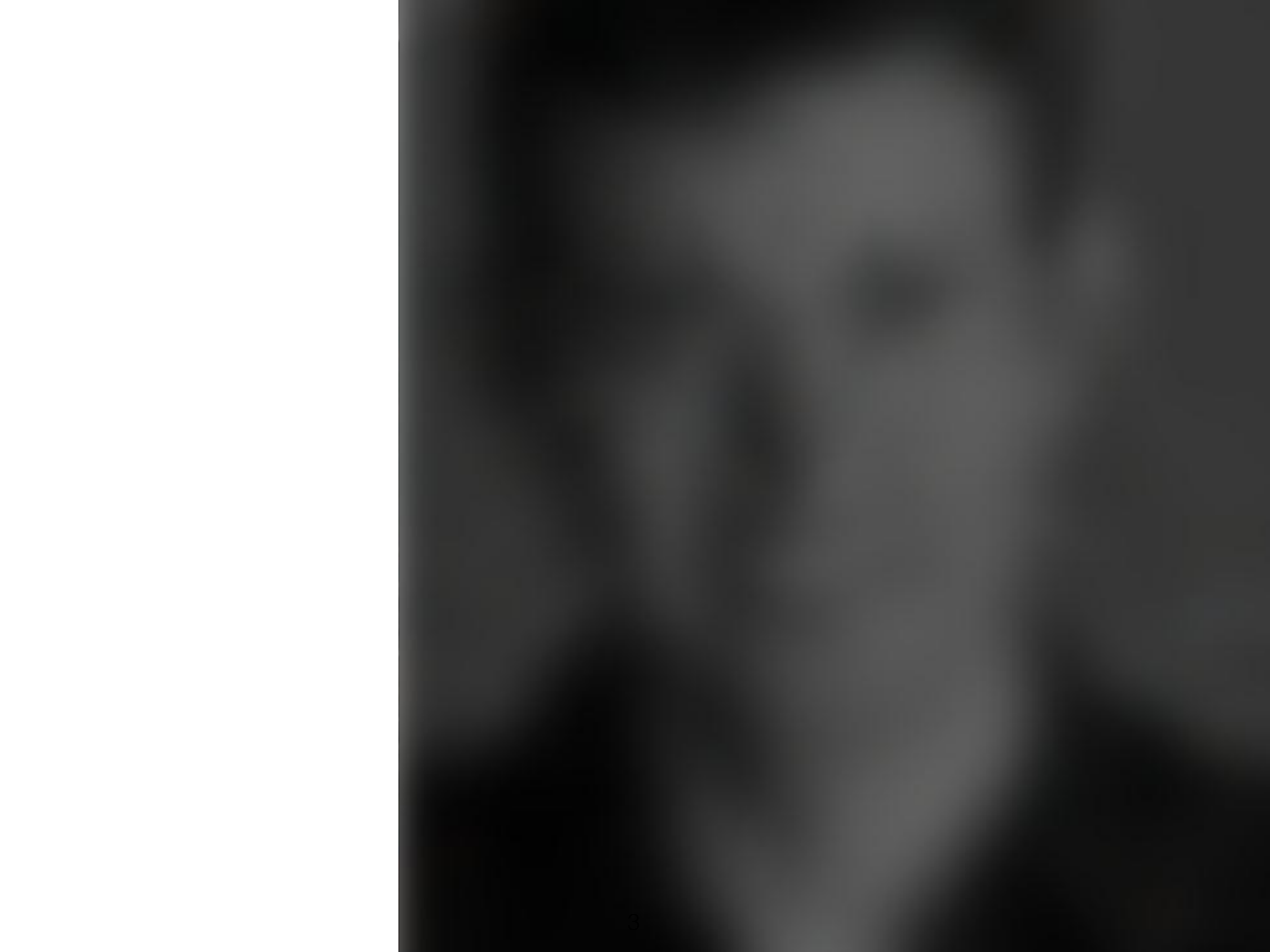
Frequency Analysis

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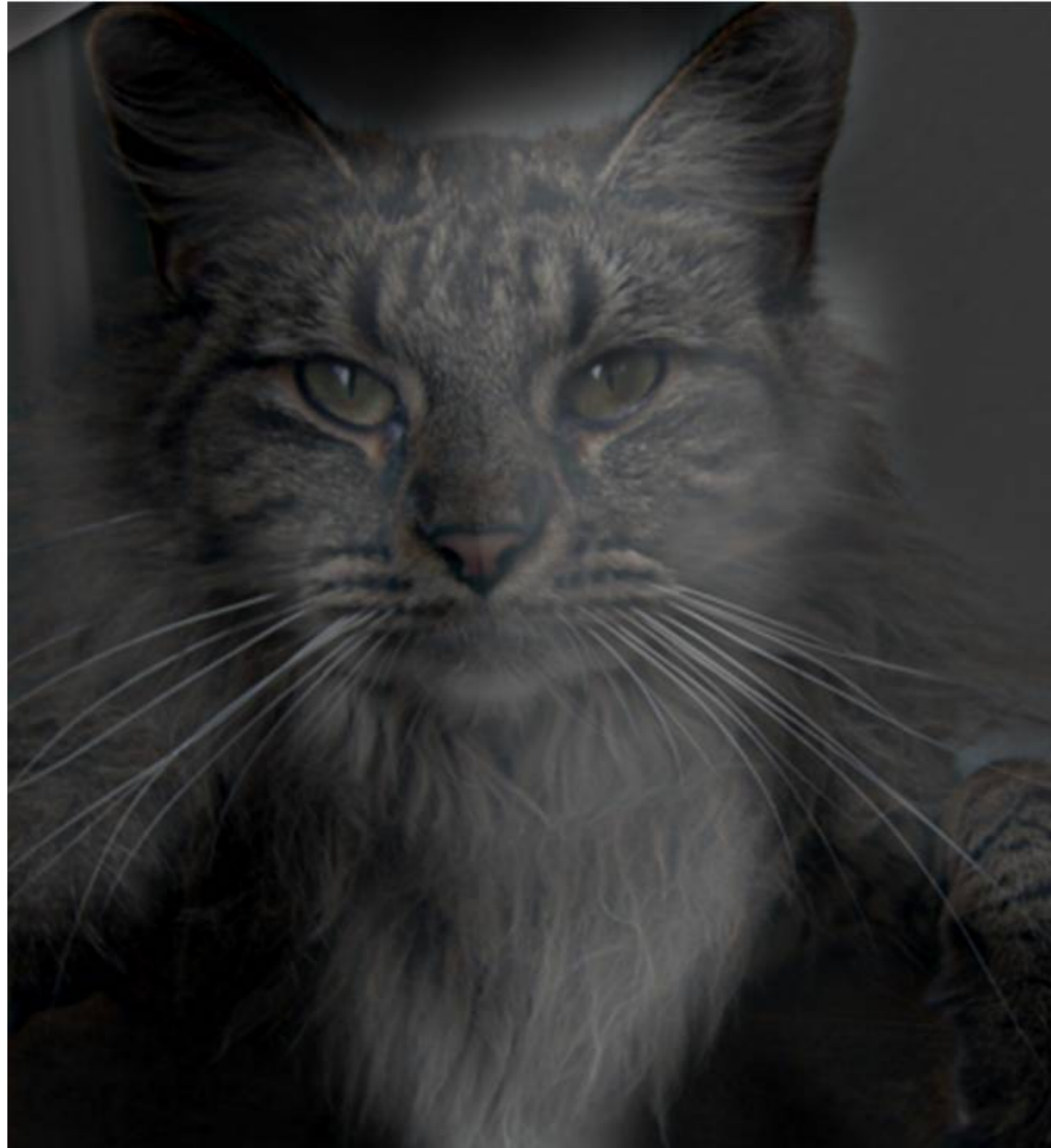
Salvador Dali

invented *Hybrid Images*





Why do we get two different distance dependent interpretations?



1



2



Why does lower resolution image still makes sense to us?



Jean Baptiste Joseph Fourier

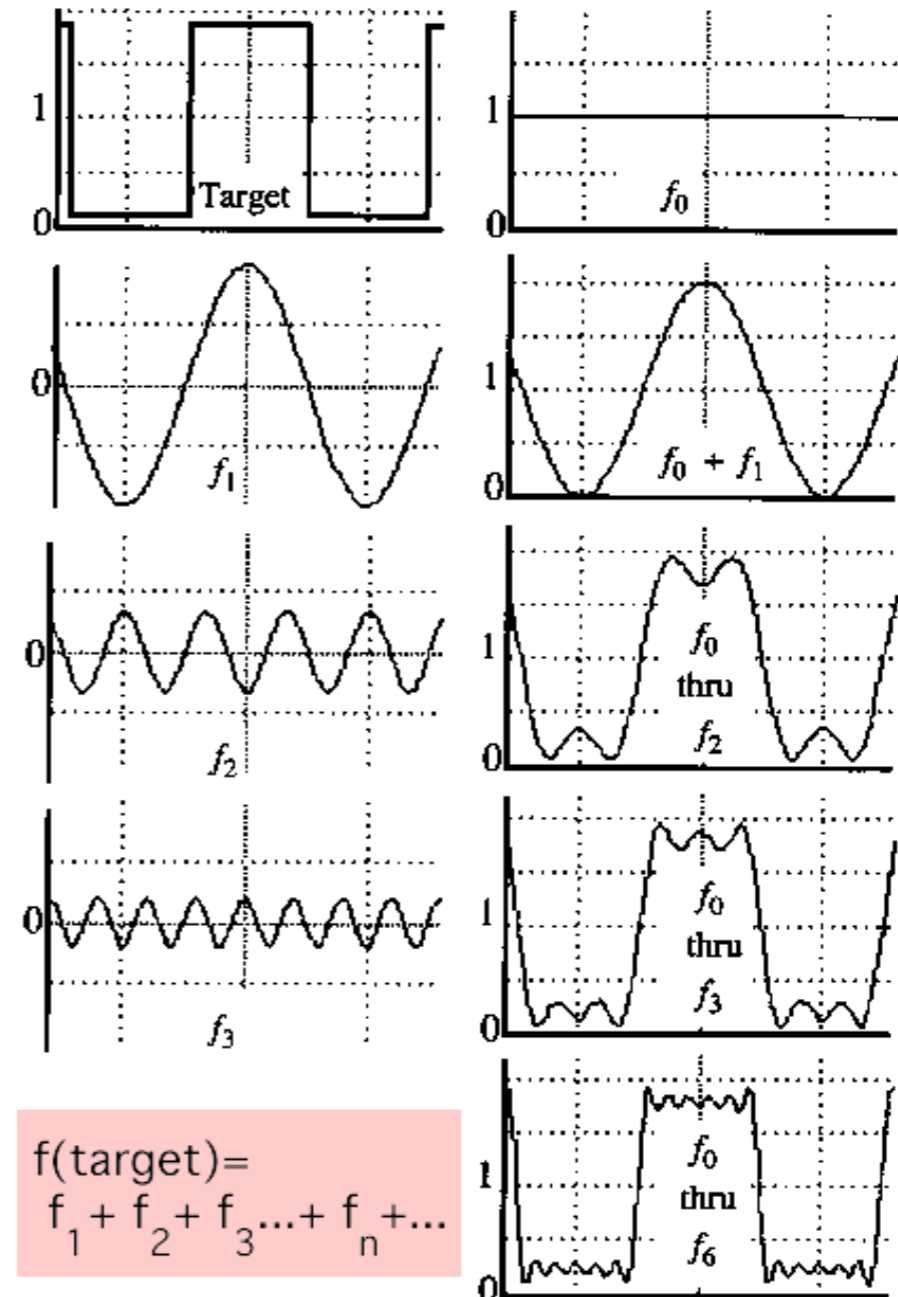
1786 — 1830

- Any univariate function can be re-written as a sum of sines and cosines of different frequencies (1807)
 - No one believed him
 - Not translated into English until 1878
- It's true
 - Called *Fourier Series*

A sum of sines

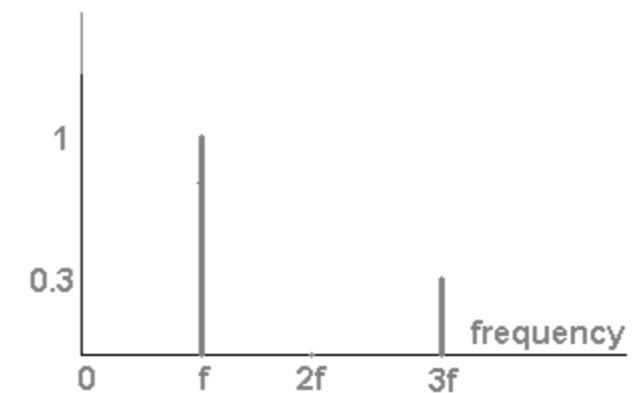
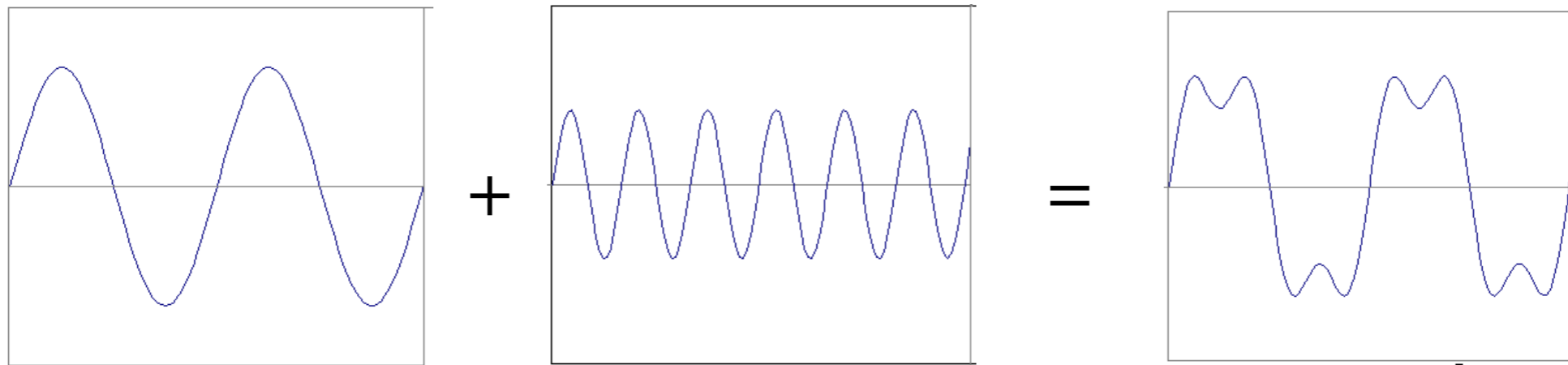
- Add enough of them to get any signal you want
- Basic building block

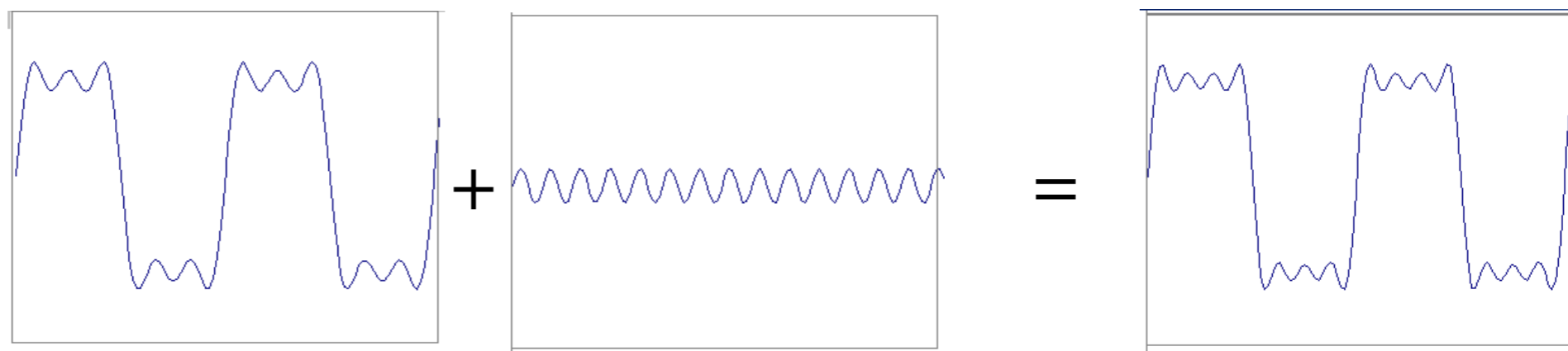
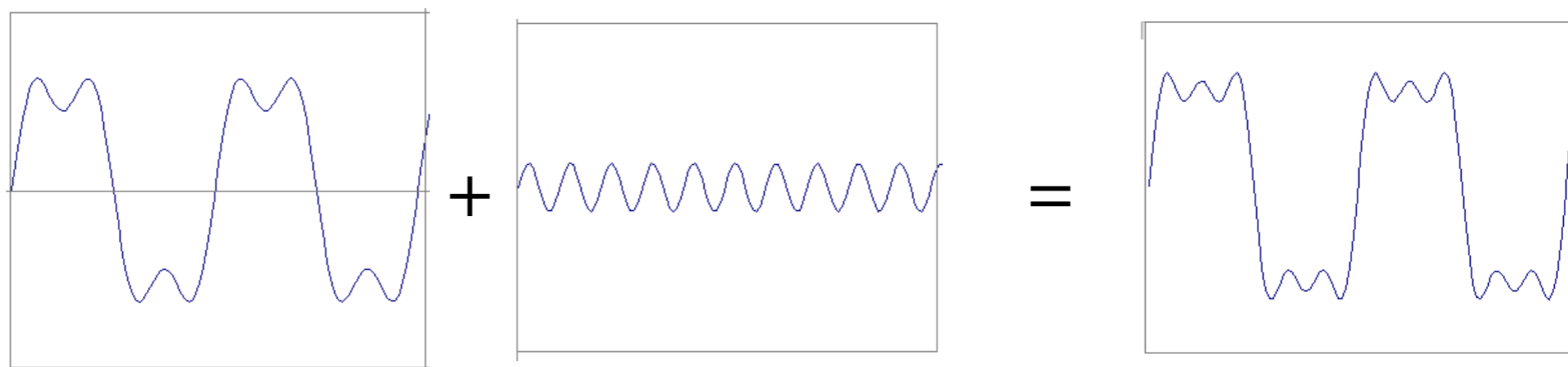
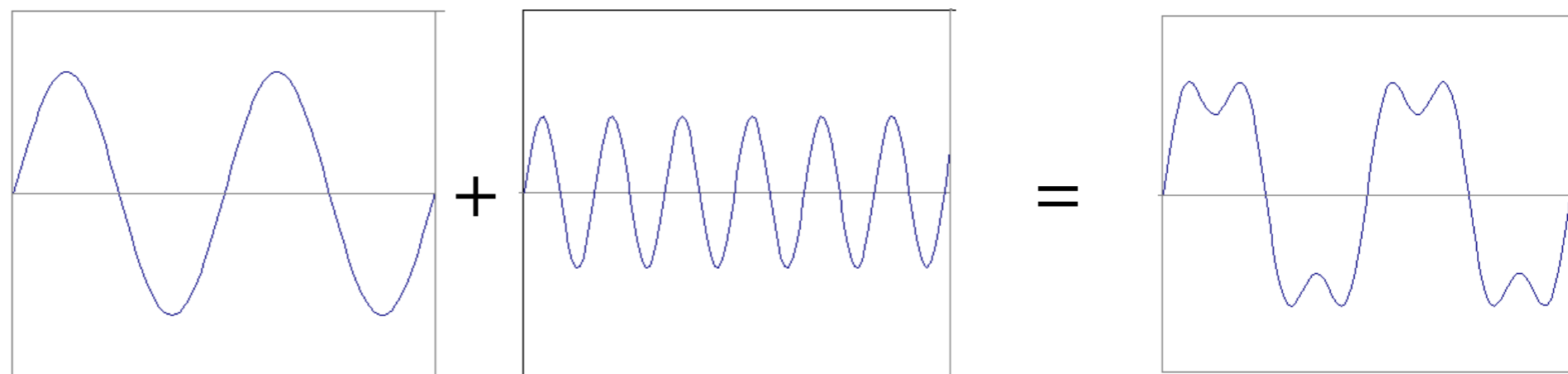
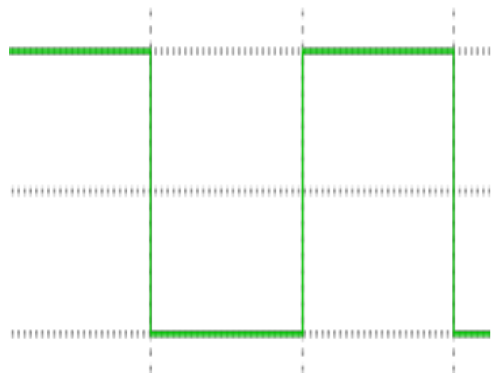
$$A \sin(\omega x + \phi)$$

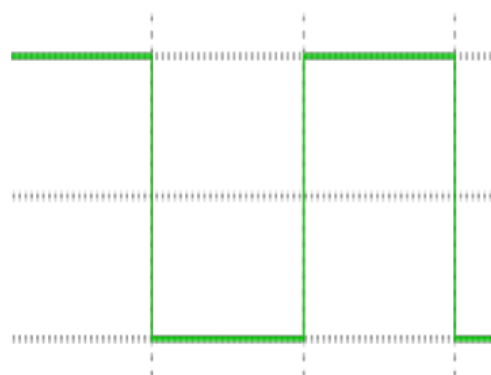
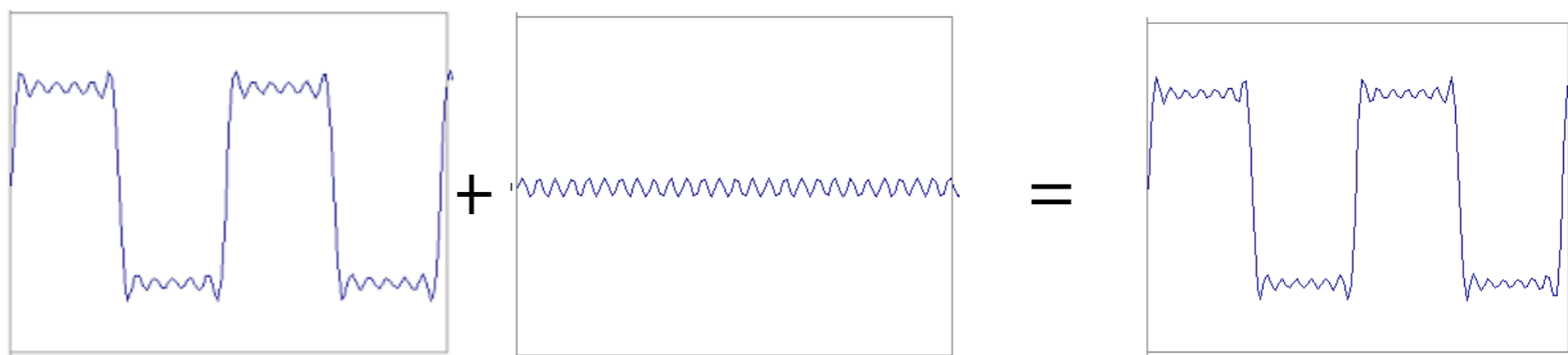
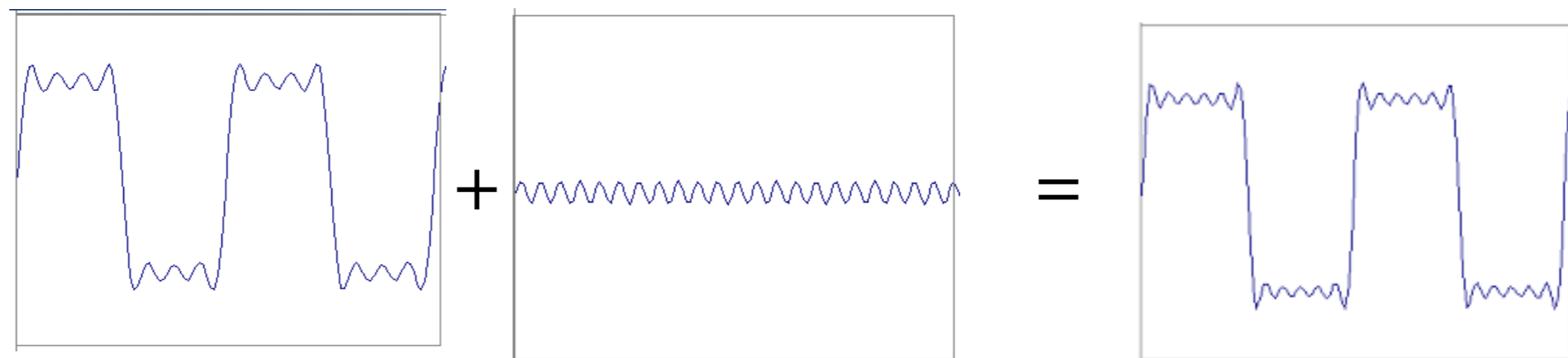
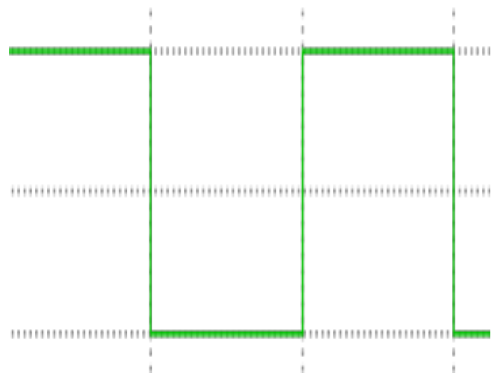


Frequency Spectra

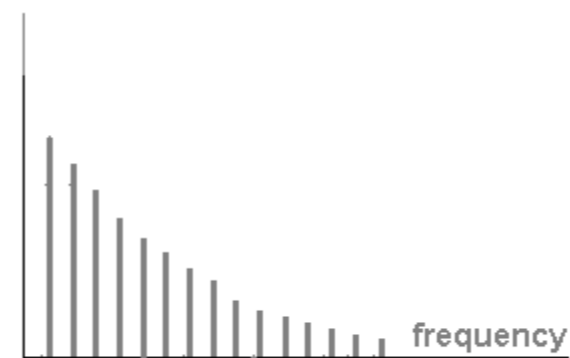
$$g(t) = \sin(2\pi ft) + \frac{1}{3} \sin(2\pi(3f)t)$$





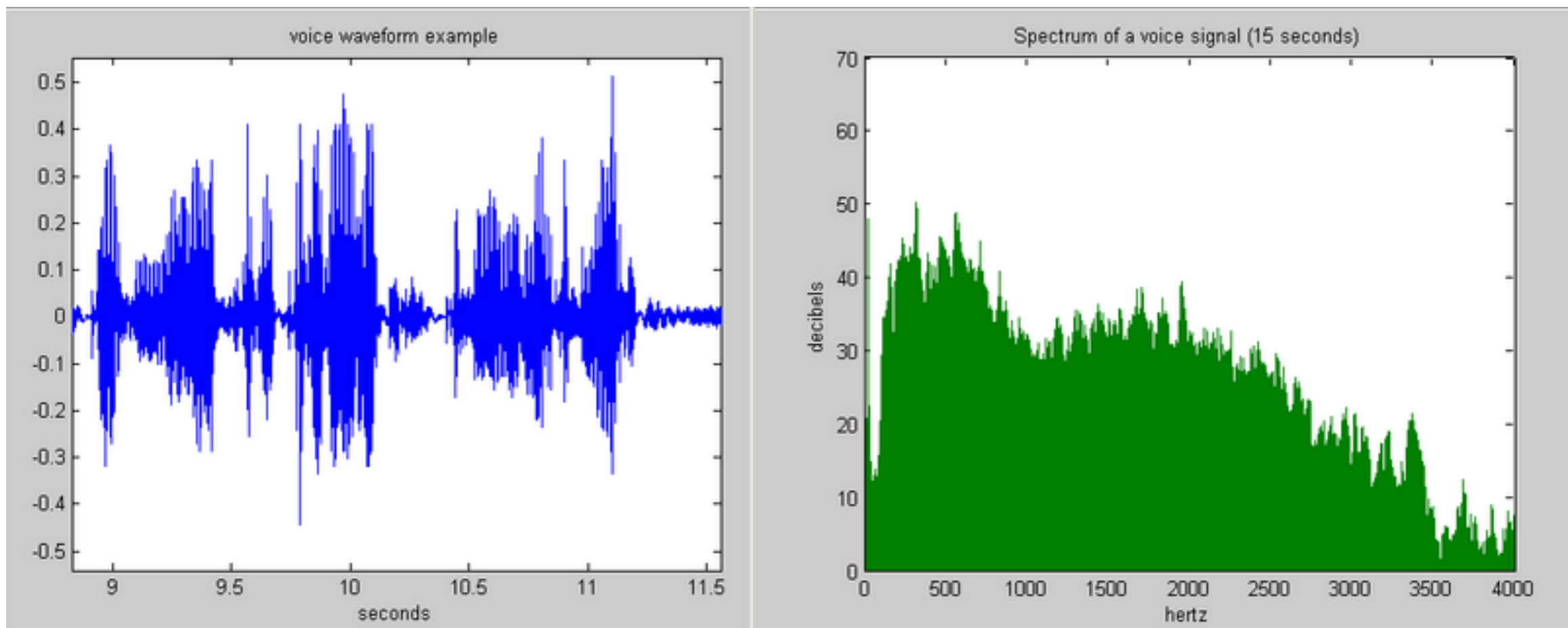


$$= A \sum_{k=1}^{\infty} \frac{1}{k} \sin(2\pi kt)$$



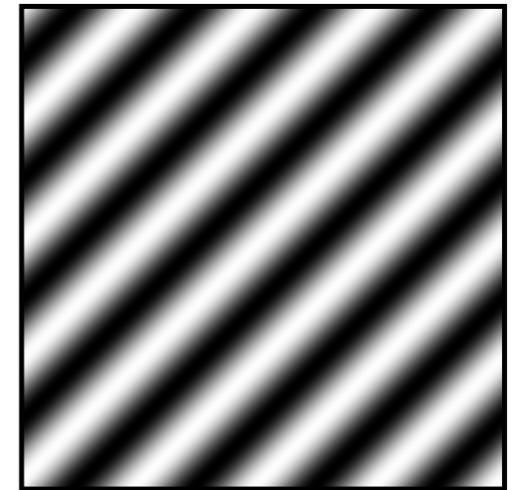
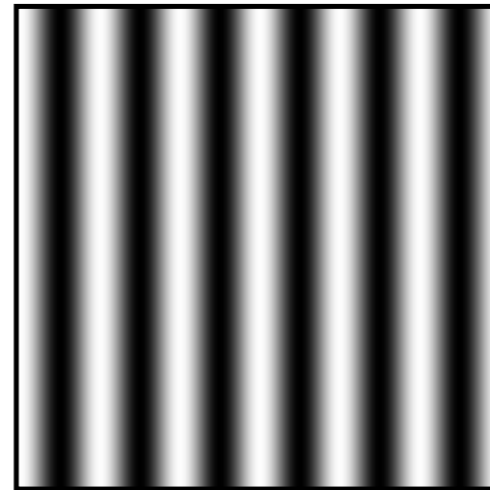
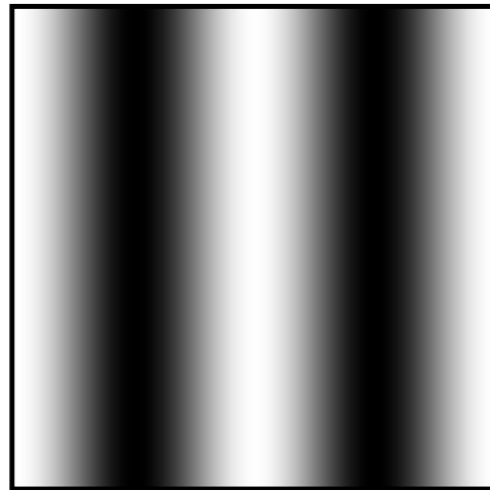
Example: Music

- We think of music in terms of frequencies at different magnitudes

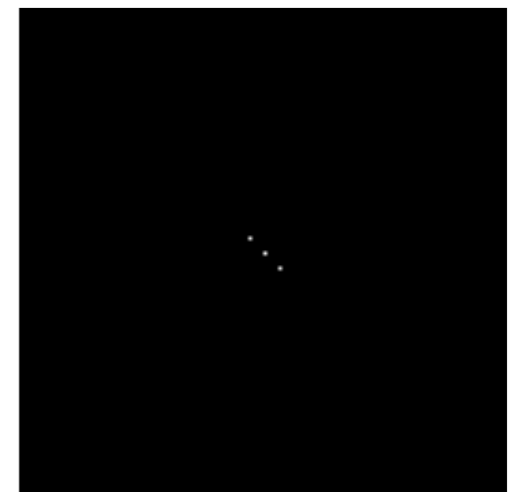
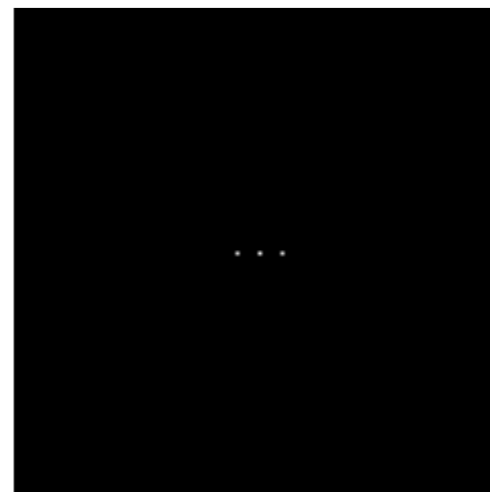
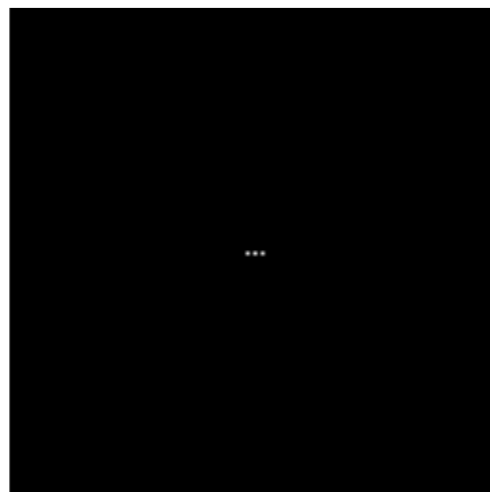


Fourier Analysis in Images

Intensity image



Fourier image



<http://sharp.bu.edu/~slehar/fourier/fourier.html#filtering>

Fourier Transform

- Fourier transform stores the magnitude and phase at each frequency

- Magnitude encodes how much signal is at a particular frequency

$$A = \pm \sqrt{R(\omega)^2 + I(\omega)^2}$$

- Phase encodes spatial information (indirectly)

$$\phi = \arctan \frac{I(\omega)}{R(\omega)}$$

- For mathematical convenience, these are often represented as real and complex numbers

Fourier Transform

- Fourier transform

$$\mathcal{F}\{f(x)\} = \hat{f}(w) := \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-iwx} dx$$

- Inverse Fourier transform

$$f(x) = \mathcal{F}^{-1}\{\hat{f}(w)\} := \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \hat{f}(w) e^{iwx} dw$$

Discrete Fourier Transform

- A sequence of N complex numbers

$$x_0, x_1, x_2, \dots, x_{N-1}$$

can be transformed into an N -periodic sequence of complex numbers

$$X_k := \sum_{n=0}^{N-1} x_n e^{-2\pi i kn/N}$$

- Fast Fourier Transform (FFT) is $N \log N$

Convolution Theorem

- Fourier transform of the convolution of two functions is the product of their Fourier transforms

$$\mathcal{F}\{g * h\} = \mathcal{F}\{g\}\mathcal{F}\{h\}$$

- The inverse Fourier transform of the product of two Fourier transform is the convolution of the two inverse Fourier transforms

$$\mathcal{F}^{-1}\{gh\} = \mathcal{F}^{-1}\{g\} * \mathcal{F}^{-1}\{h\}$$

Convolution in *spatial domain* is equivalent to multiplication in *frequency domain*

Properties of Fourier Transforms

- Linearity

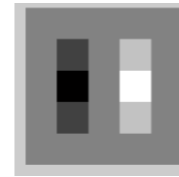
$$\mathcal{F}\{ag(x) + bh(x)\} = a\mathcal{F}\{g(x)\} + b\mathcal{F}\{h(x)\}$$

- Fourier transform of a real signal is symmetric around origin
- The energy of the signal is the same as the energy of its Fourier transform

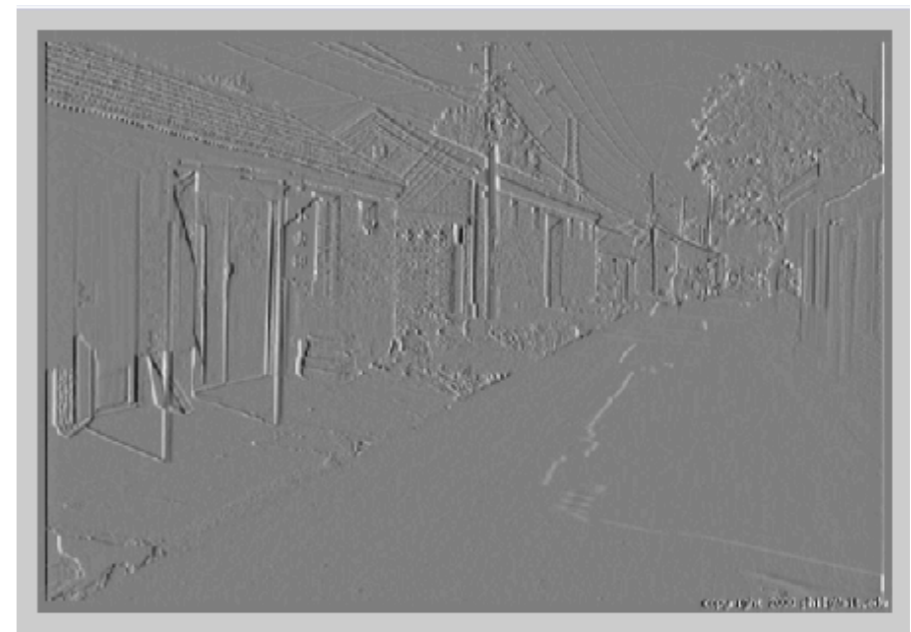
Filtering in Spatial Domain



*



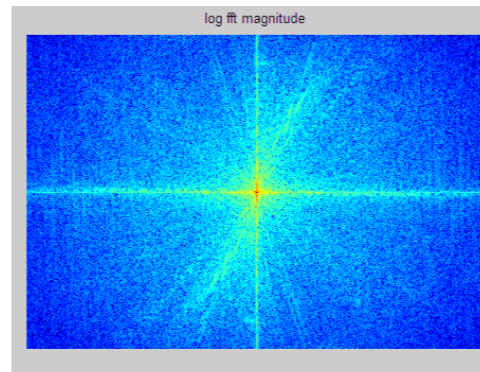
=



Filtering in Frequency Domain



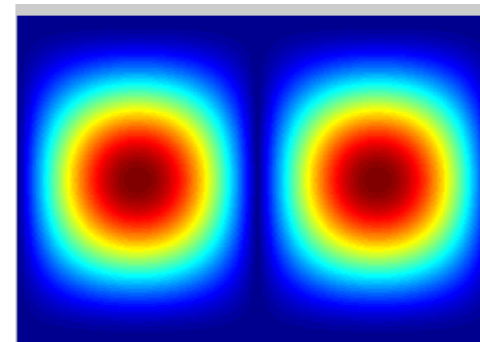
FFT
→



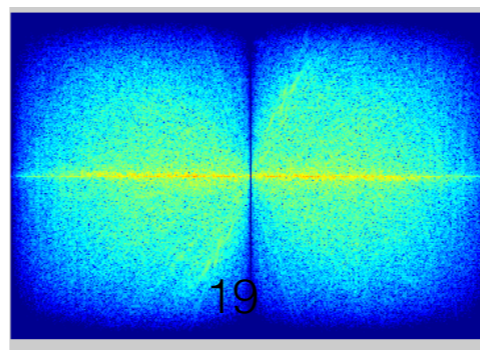
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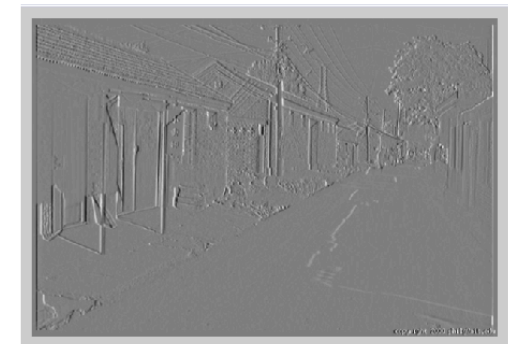
FFT
→



=

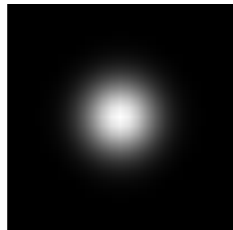


Inverse
FFT
→

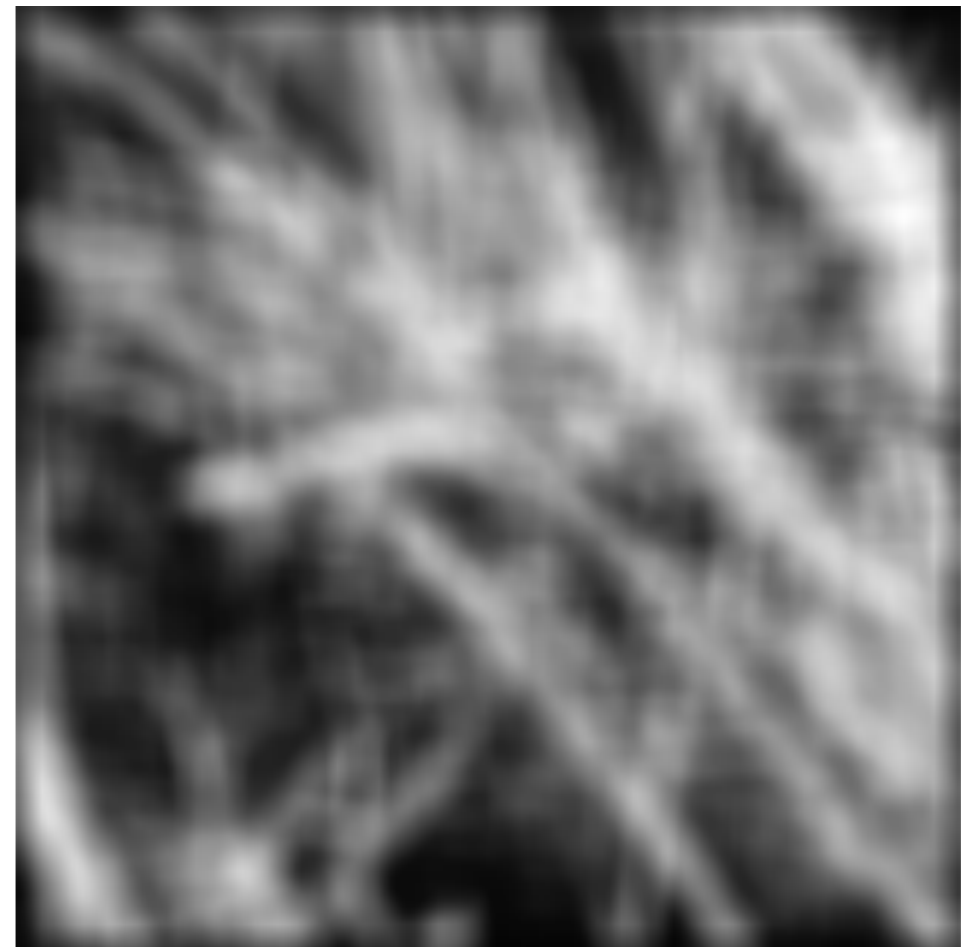
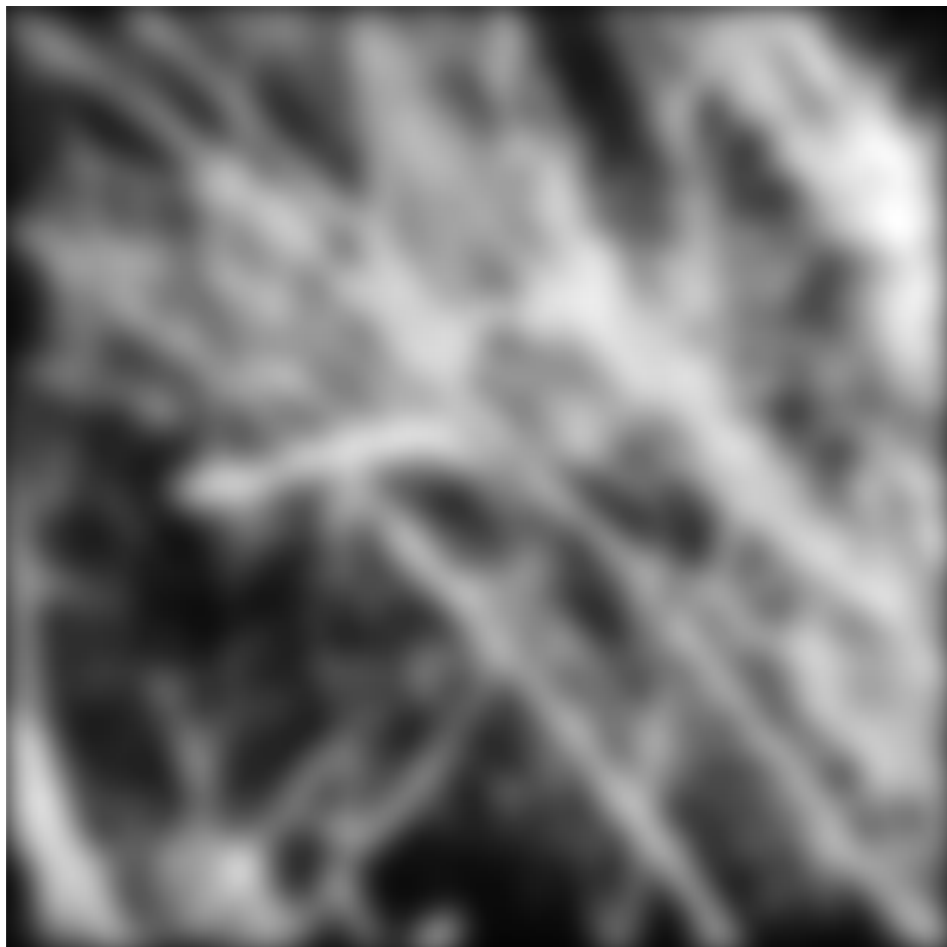


Why does Gaussian filter gives a smooth image, where as the square filter gives edgy artifacts?

Gaussian

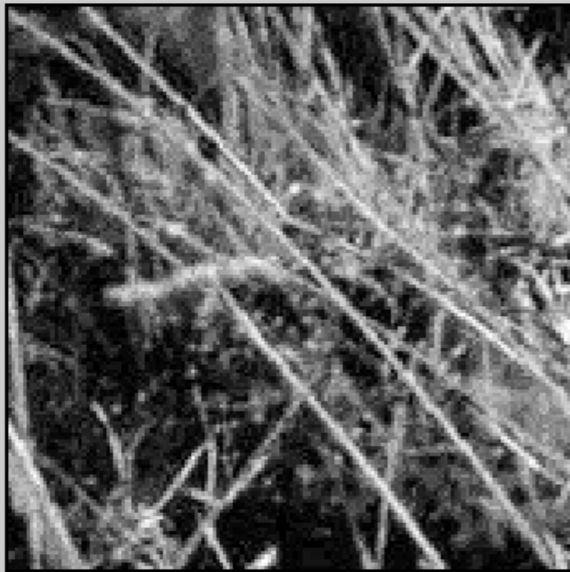


Box filter

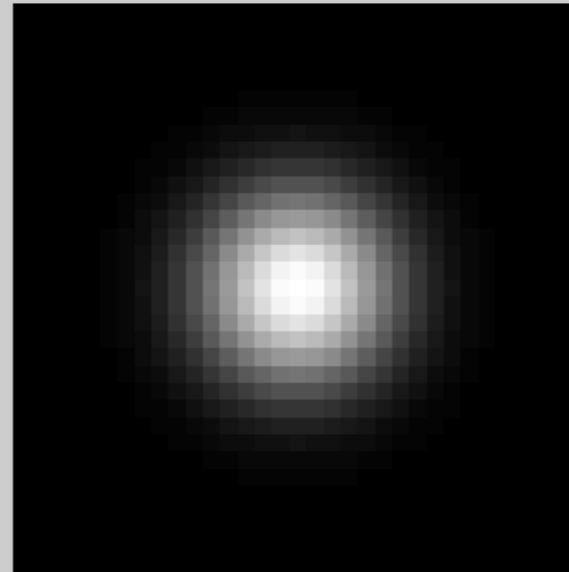


Gaussian

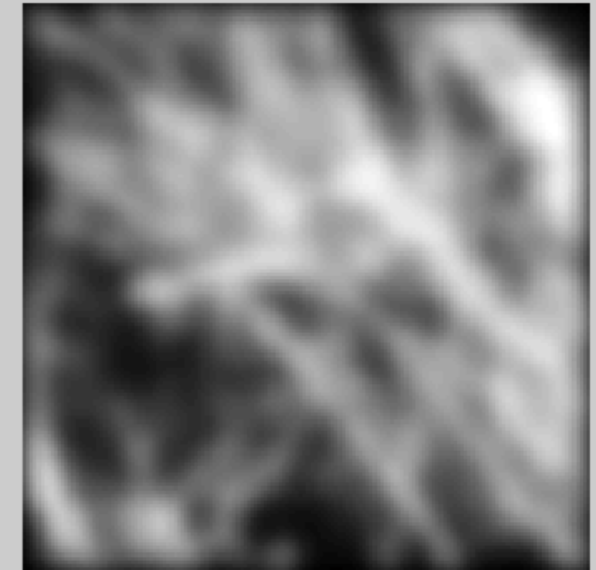
intensity image



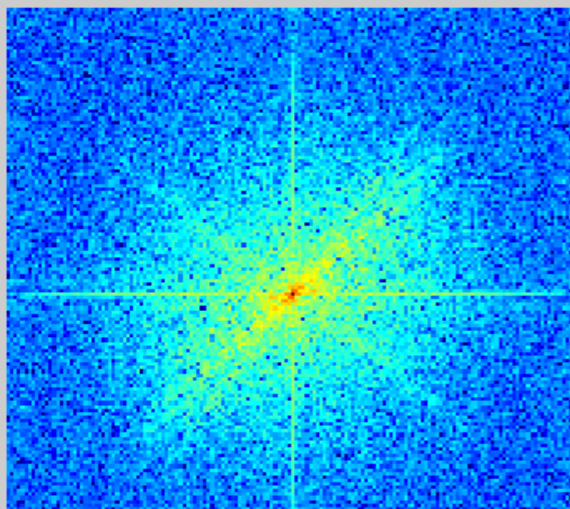
filter: gaussian



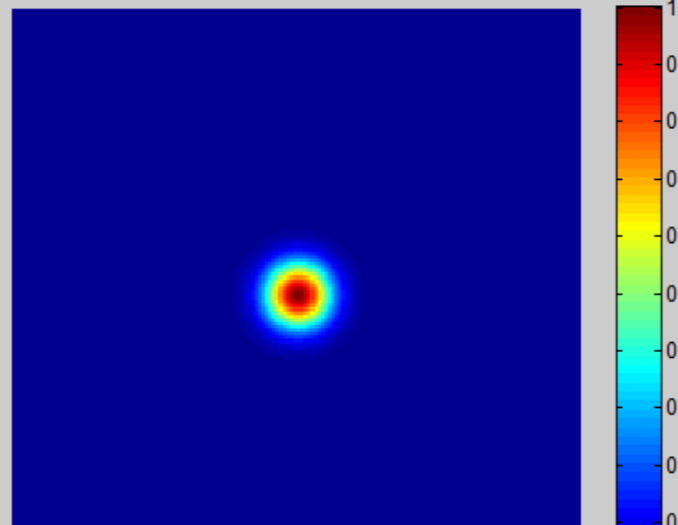
filtered image



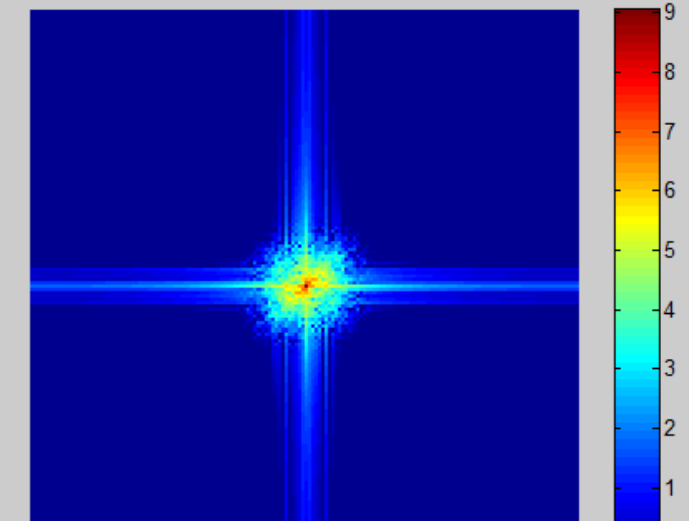
log fft magnitude of image



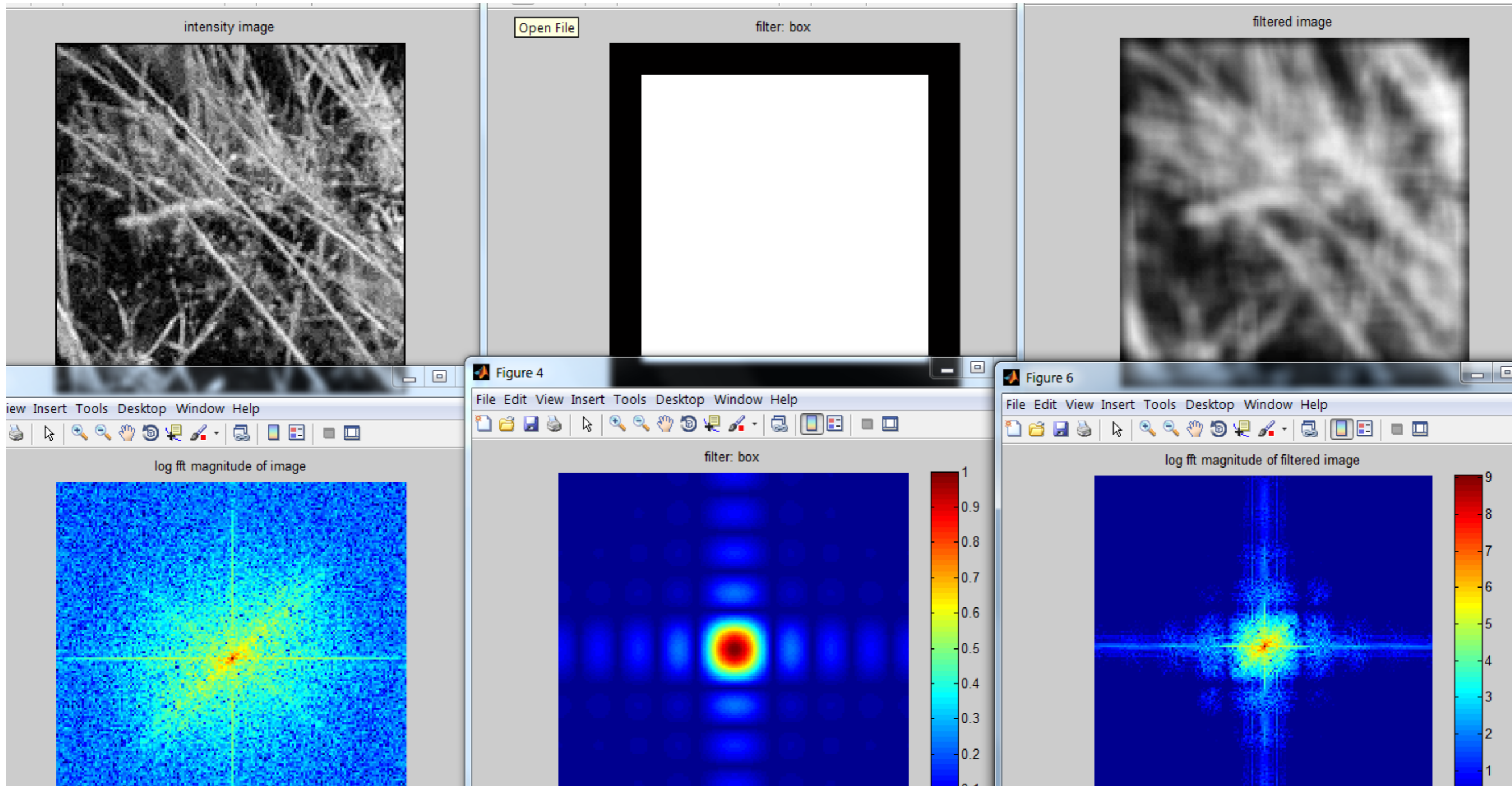
filter: gaussian



log fft magnitude of filtered image



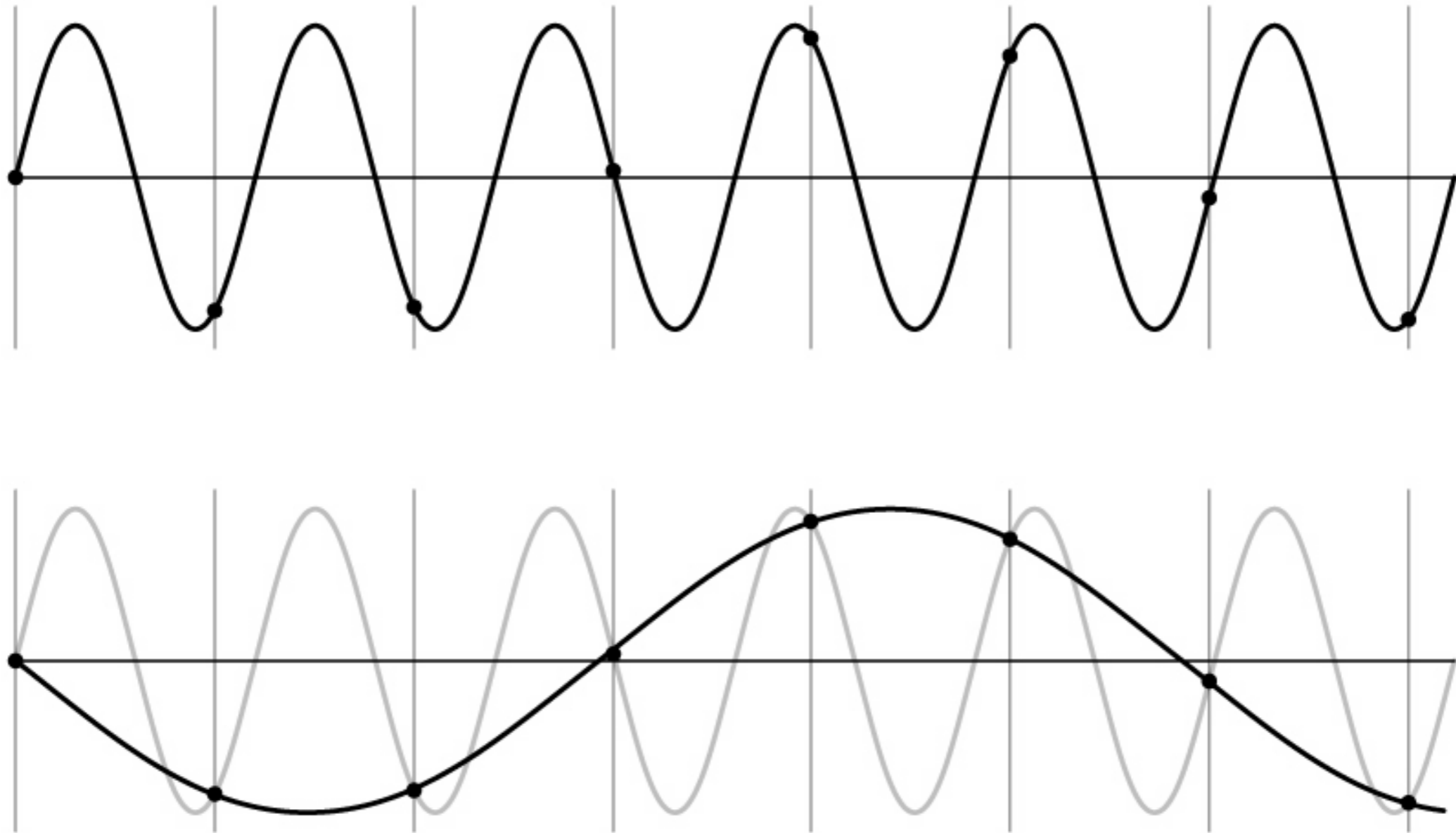
Box Filter



Why does lower resolution image still makes sense to us?

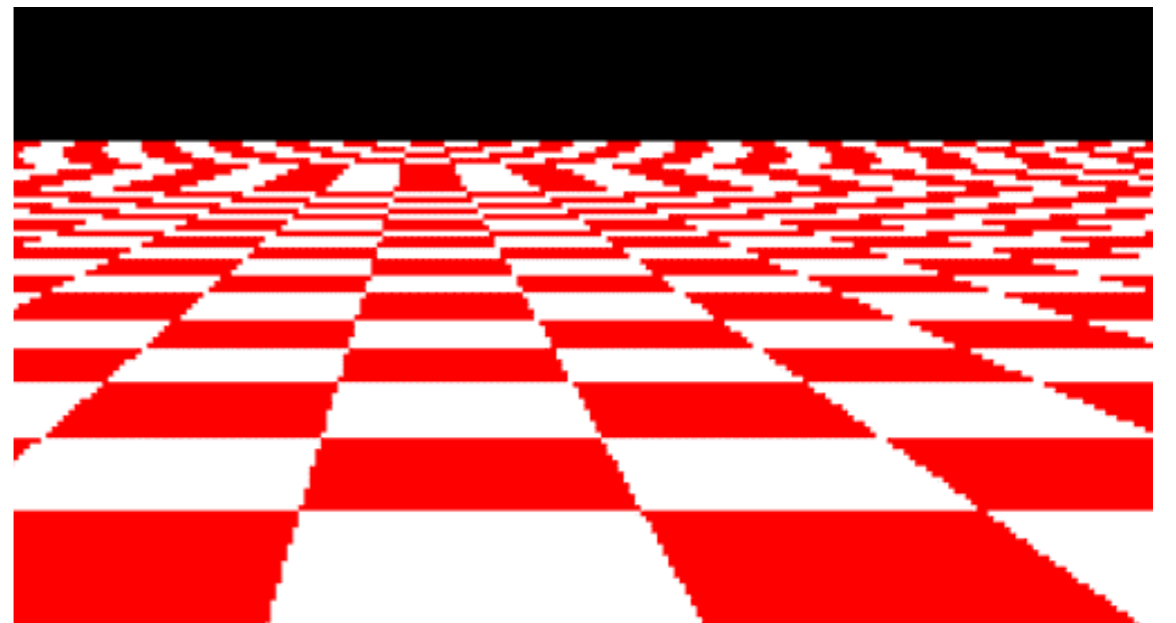


Aliasing Problem



Aliasing Problem

- Sub-sampling can result in aliasing artifacts
 - Wagon wheels rolling the wrong way in movies
 - Checkerboards disintegrate in ray tracing
 - Striped shirts look funny



Sampling and Aliasing

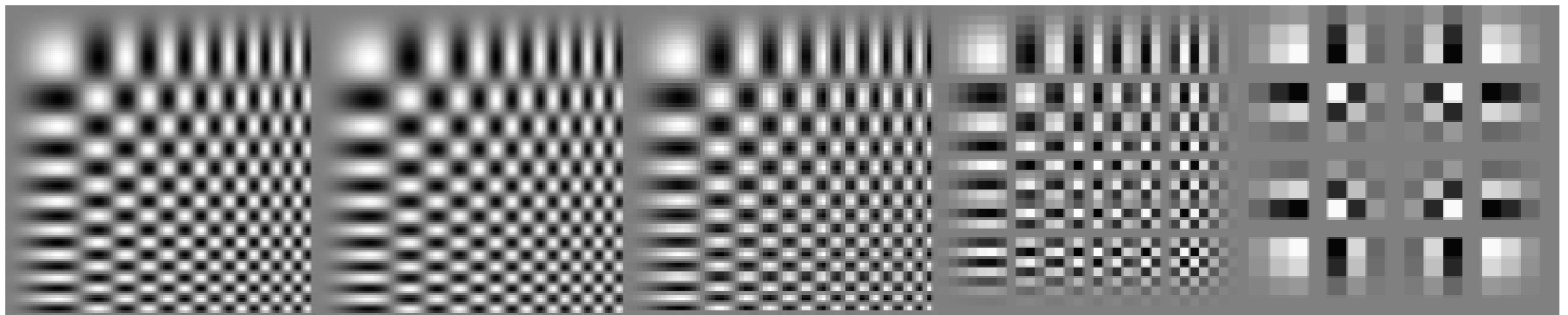
256x256

128x128

64x64

32x32

16x16



Nyquist-Shannon Sampling Theorem

- When sampling a signal at discrete intervals, the sampling frequency must be twice the maximum frequency of the signal
- This will allow us to perfectly reconstruct the original signal from its samples



Anti-Aliasing

- Sample more often
- Get rid of all frequency that are greater than half the sampling frequency
 - We will lose information
 - But still better than aliasing artifacts

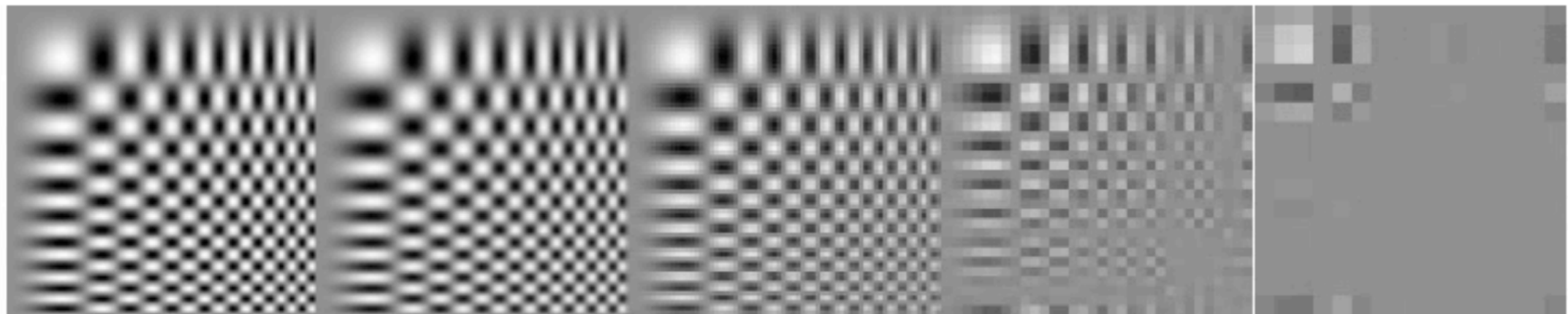
256x256

128x128

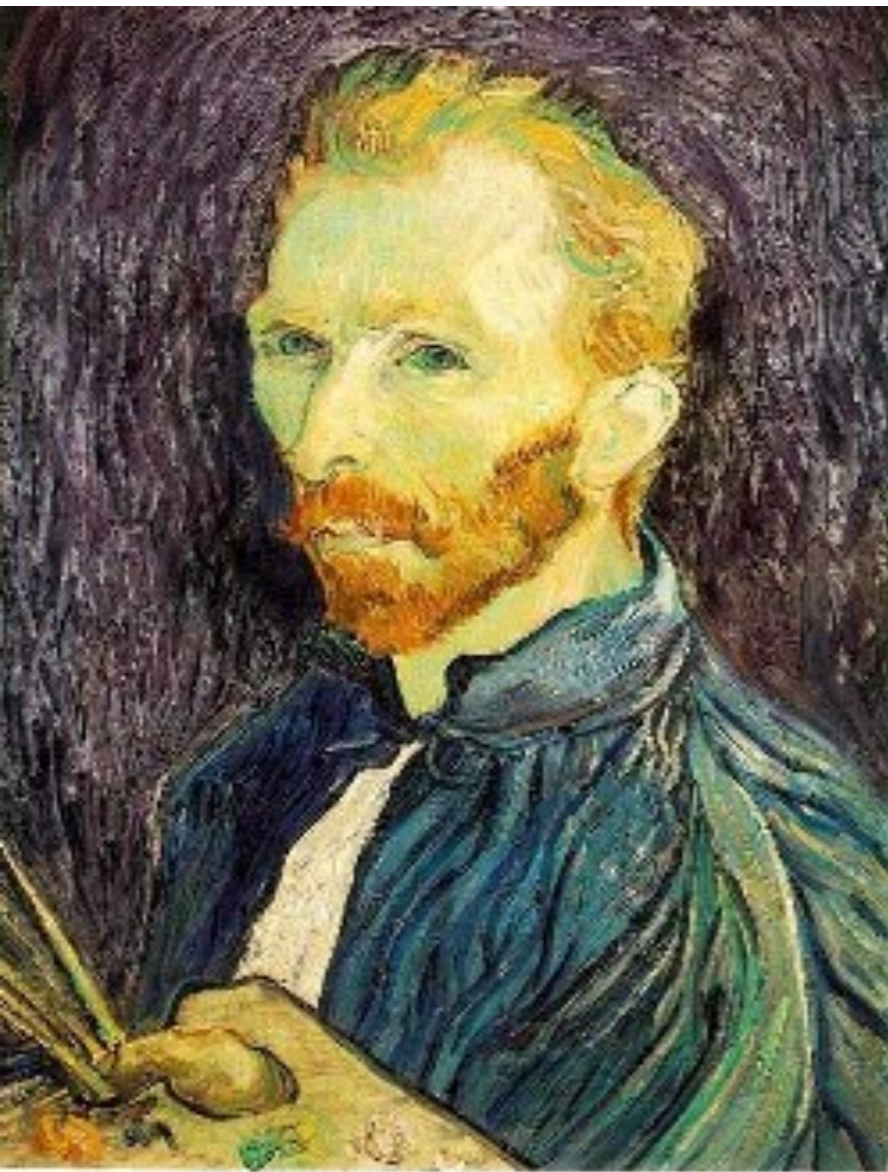
64x64

32x32

16x16



Subsampling without Pre-Filtering



1/2

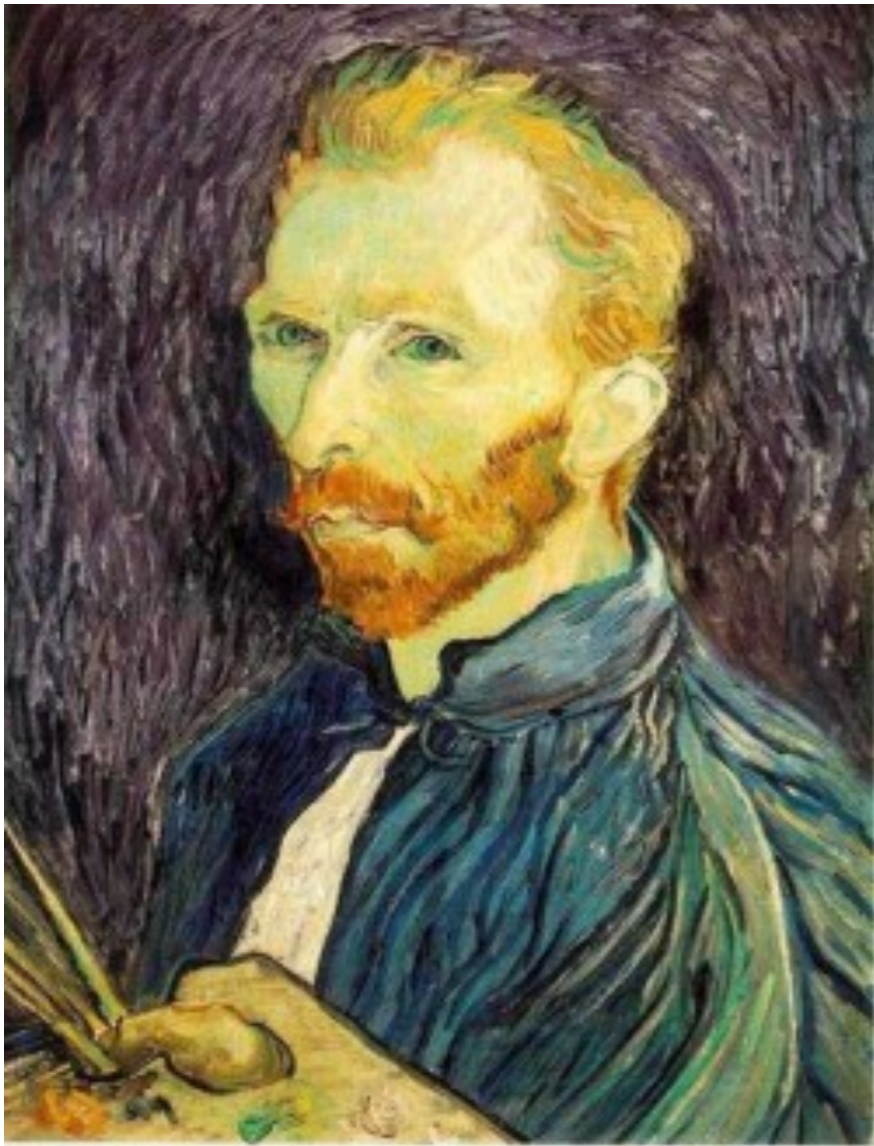


1/4
2x zoom



1/8
4x zoom

Subsampling with Gaussian Pre-Filtering



1/2



1/4
2x zoom



1/8
4x zoom

Why do we get two different distance dependent interpretations?



Summary

- Image filtering in frequency domain
 - Fourier analysis
- Image filtering in frequency domain is vs for auto correlation
- Images are smooth — image compression
- Low-pass filter before sampling