



BME I5000: Biomedical Imaging

Lecture FT Fourier Transform Review

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Blackboard: <http://cityonline.ccny.cuny.edu/>



Fourier Transform (1D)

The **Fourier Transform** (FT) is defined as*

$$H(k) = FT[h(x)] = \int_{-\infty}^{\infty} dx h(x) e^{-i2\pi kx}$$

The FT is an invertible transformation

$$h(x) = FT^{-1}[H(k)] = \int_{-\infty}^{\infty} dk H(k) e^{i2\pi kx}$$

We can show this using $\int_{-\infty}^{\infty} dk e^{-i2\pi kx} = \delta(x)$

$$\int_{-\infty}^{\infty} dk H(k) e^{i2\pi kx} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} dk dx' h(x') e^{-i2\pi k(x'-x)} = h(x)$$

* Notational convention: Use k for spacial, and ω for temporal frequency.



Fourier Transform – Frequency Domain

$H(k)$ is in general complex valued

$$H(k) = R(k) + i I(k) = A(k) e^{i\phi(k)}$$

with $R(k) = \text{Re}(H(k))$, $I(k) = \text{Im}(H(k))$ and

Amplitude: $A(k) = |H(k)| = \sqrt{R(k)^2 + I(k)^2}$

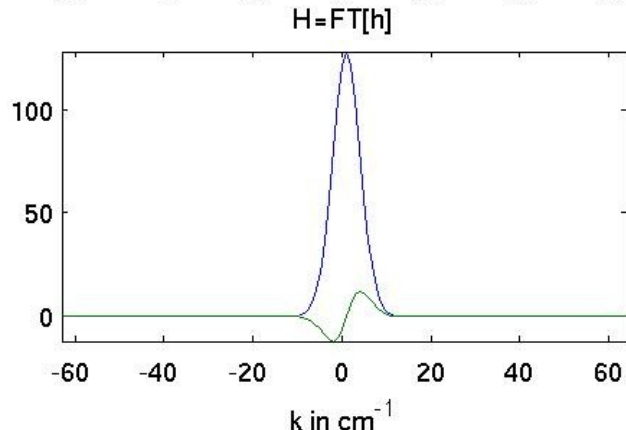
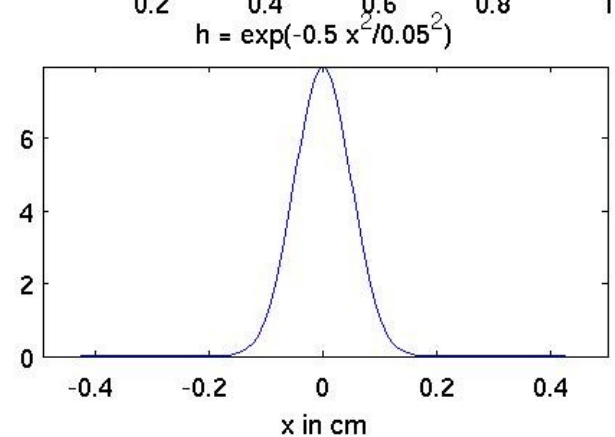
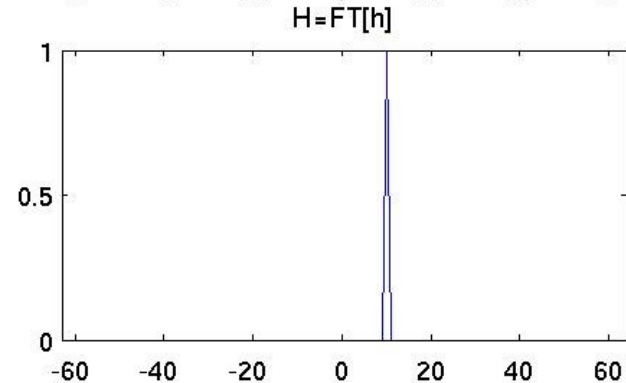
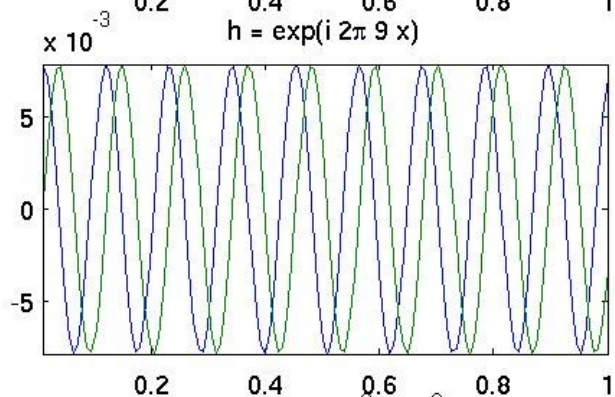
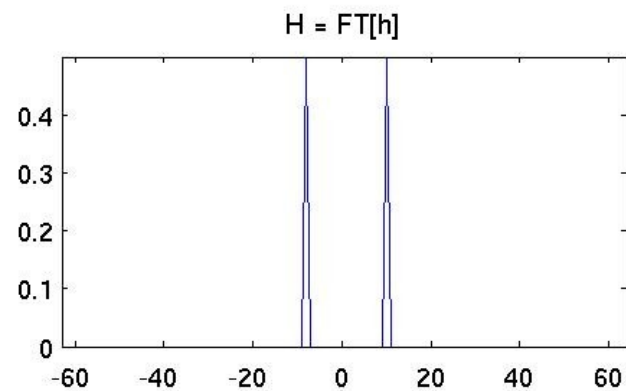
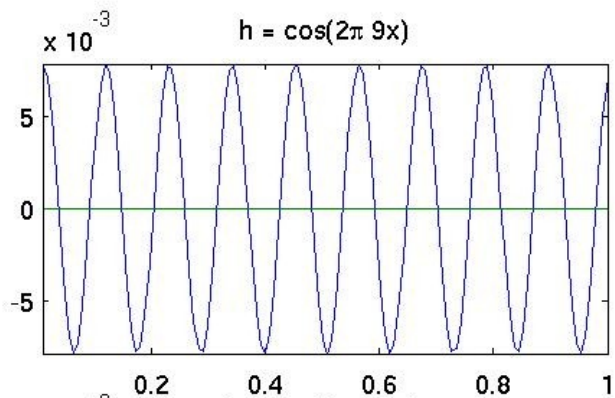
Phase: $\phi(k) = \arctan\left(\frac{I(k)}{R(k)}\right)$

Positive frequencies $k > 0$: counterclockwise rotation

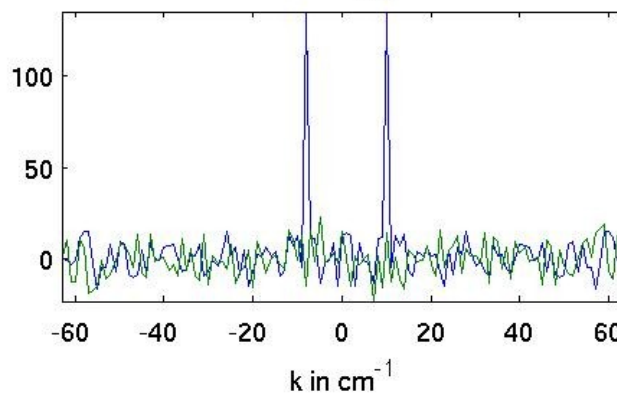
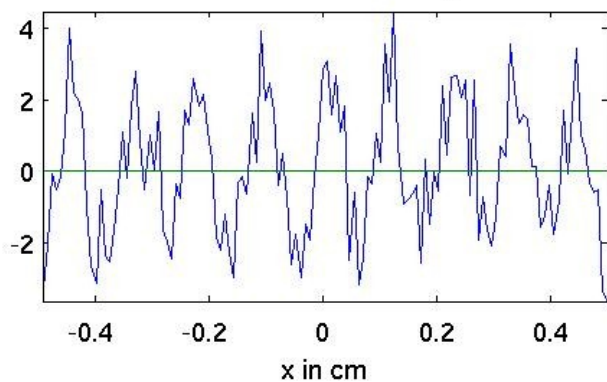
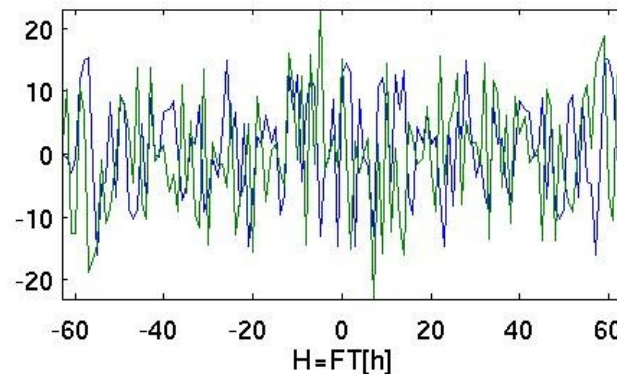
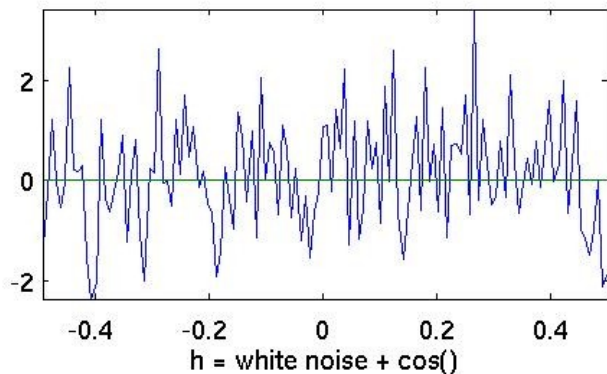
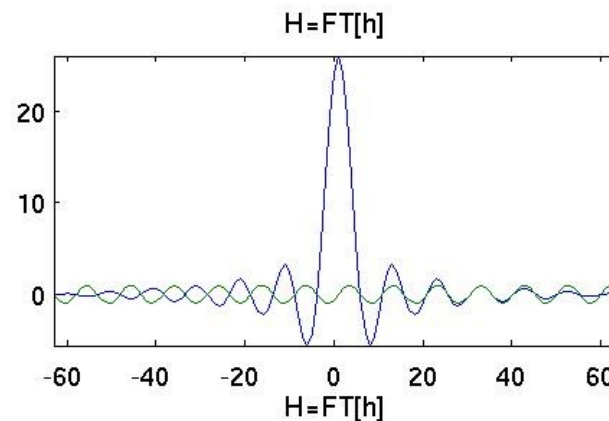
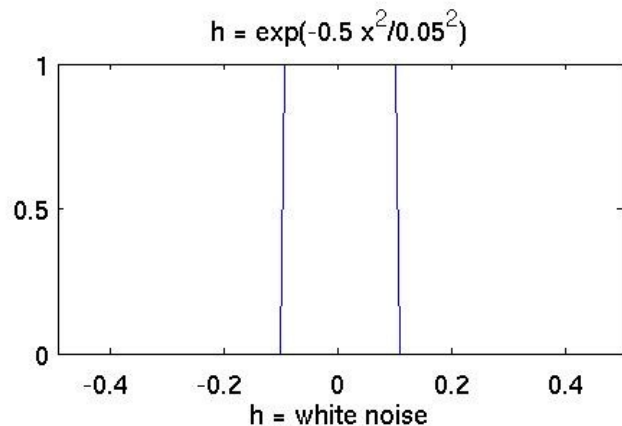
Negative frequencies $k < 0$: clockwise rotation

$$H(k) = \delta(k - k_0), \quad h(x) = e^{i2\pi k_0 x}$$

Fourier Transform - Examples



Fourier Transform - Examples





Fourier Transform – Convolution Theorem

Convolution is defined as

$$b(x) = \int_{-\infty}^{\infty} dx' h(x') g(x-x') = h(x) * g(x)$$

Convolution Theorem states that

$$B(k) = FT[h(x) * g(x)] = H(k)G(k)$$

Note that with the convolution theorem we can implement convolution as a multiplication in the frequency domain.

$$\begin{array}{c}
 h(x) \xrightarrow{FT} H(k) \\
 g(x) \xrightarrow{FT} G(k)
 \end{array}
 \begin{array}{c}
 \searrow \\
 \nearrow
 \end{array}
 \begin{array}{c}
 \textcircled{\times} \\
 \rightarrow B(k)
 \end{array}
 \xrightarrow{FT^{-1}} b(x)$$



Fourier Transform – Convolution Theorem

Proof

$$\begin{aligned} FT[h(x)*g(x)] &= \int_{-\infty}^{\infty} dx h(x)*g(x) e^{-i2\pi kx} = \\ &= \int_{-\infty}^{\infty} dx \int_{-\infty}^{\infty} dx' h(x') g(x-x') e^{-i2\pi kx} \\ &= \int_{-\infty}^{\infty} dx \int_{-\infty}^{\infty} dx' h(x') g(x) e^{-i2\pi k(x+x')} \\ &= \int_{-\infty}^{\infty} dx' h(x') e^{-i2\pi kx'} \int_{-\infty}^{\infty} dx g(x) e^{-i2\pi kx} \\ &= H(k)G(k) \end{aligned}$$



Fourier Transform - Inverse Filter

With the Convolution Theorem we can derive the inverse convolution (or inverse filter)

$$b(x) = g(x) * h(x) \Leftrightarrow B(k) = G(k) H(k)$$

Therefore

$$G(k) = \frac{B(k)}{H(k)}$$

And the inverse filter is given by the inverse FT of $H^{-1}(k)$:

$$g(x) = FT^{-1} \left[\frac{1}{H(k)} \right] * b(x)$$



Fourier Transform – 2D and higher

The Fourier transform in 2D is defined as

$$H(k_x, k_y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} dx dy h(x, y) e^{-i2\pi(k_x x + k_y y)}$$

The inverse transform is defined correspondingly.

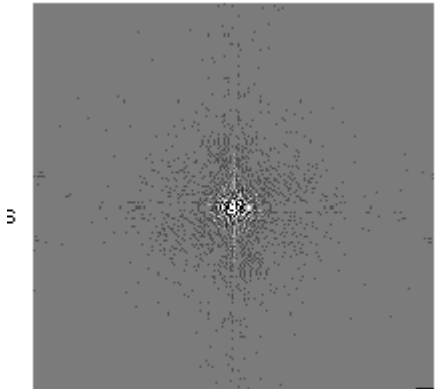
Note that the exponent is **separable** and therefor the Fourier transform can be applied **sequentially in each dimension!**

$$H(k_x, k_y) = \int_{-\infty}^{\infty} dy e^{-i2\pi k_y y} \int_{-\infty}^{\infty} dx h(x, y) e^{-i2\pi k_x x}$$

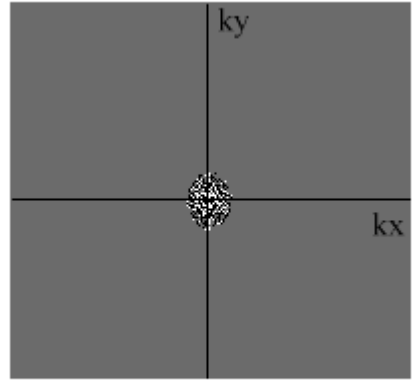
In multiple dimensions with $\mathbf{k} = [k_x, k_y, k_z, \dots]^T$, $\mathbf{r} = [x, y, z, \dots]^T$

$$H(\mathbf{k}) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} d\mathbf{r} h(\mathbf{r}) e^{-i2\pi \mathbf{k} \cdot \mathbf{r}}$$

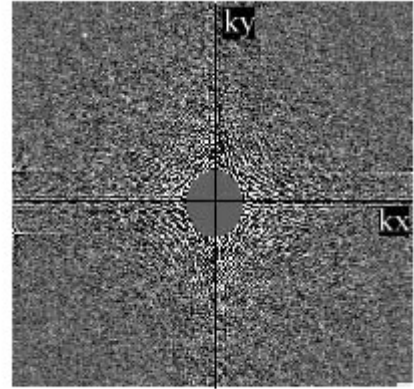
Fourie Transform k-space



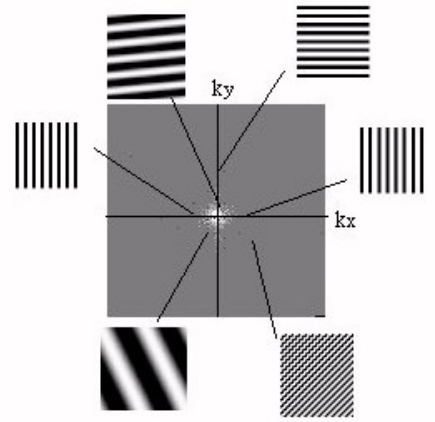
The Mona Lisa in k-Space



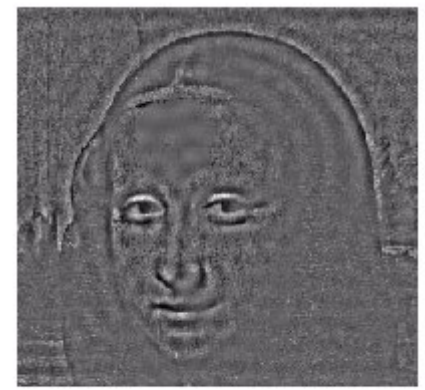
k-Space



k-Space



Low Frequency Mona



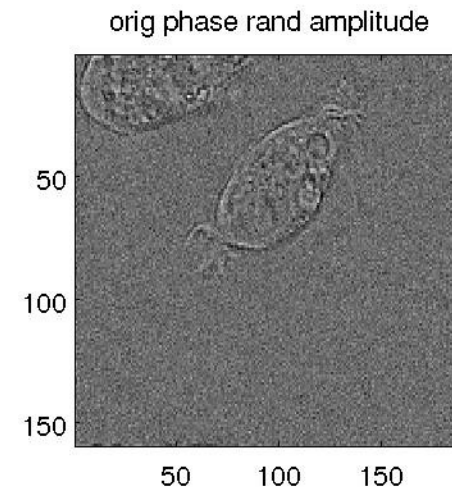
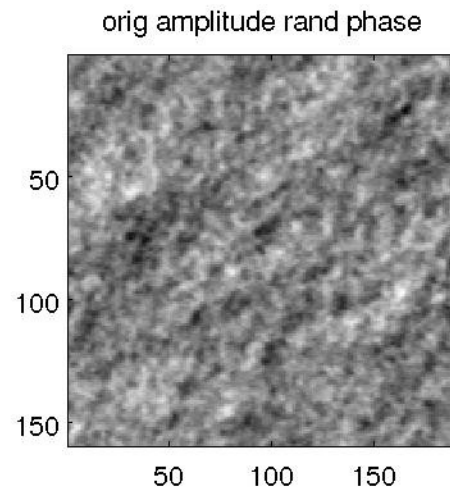
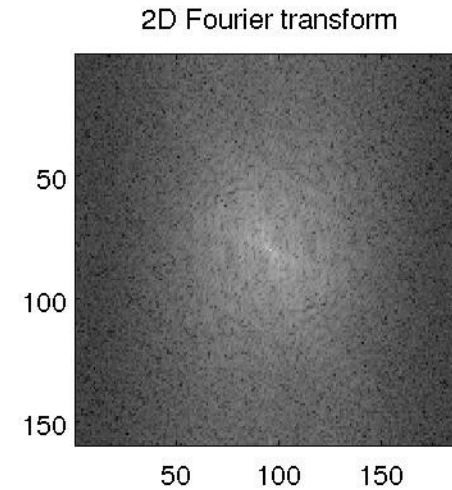
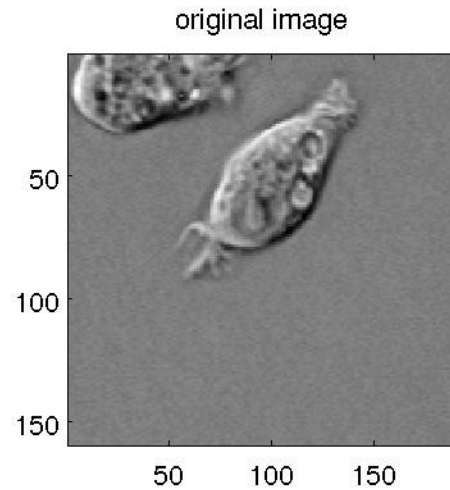
High Frequency Mona

Source: (C.A. Mistretta)



2D Fourier Transform – Phase and Amplitude

Note that in the images the phase carries most of the information

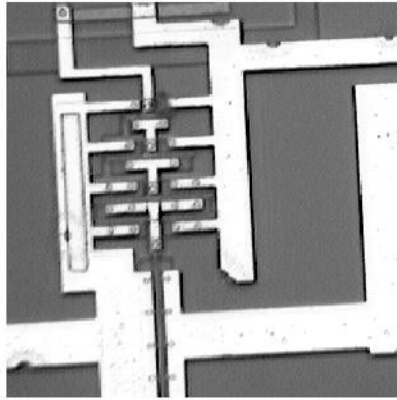




2D Fourier Transform – Phase and Amplitude

Note that in images the phase carries most of the information

original image 1



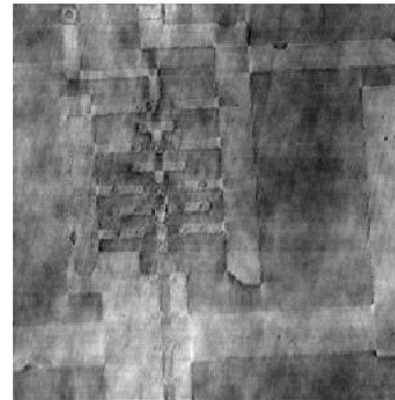
original image 2



Amp of image 1 and Phi of image 2



Phi of image 1 and Amp of image 2





Fourier Transform – 2D Convolution

The 2D convolution with a PSF $h(x,y)$ is

$$\begin{aligned} b(x, y) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} dx' dy' h(x', y') g(x-x', y-y') \\ &= h(x, y) * g(x, y) \end{aligned}$$

The Convolution Theorem applies in higher dimensions as well.

$$B(k_x, k_y) = G(k_x, k_y) H(k_x, k_y)$$

Even though $h(x,y)$ may not be separable ($h(x,y) \neq h(x)*h(y)$) with the convolution theorem we never have to truly compute a 2D convolution.



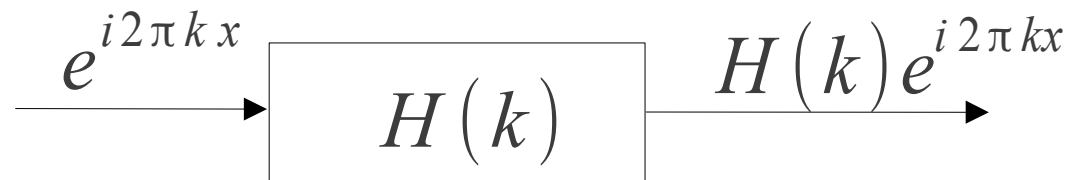
Fourier Domain System Response

Consider stationary oscillatory input to a LSI system $h(x)$:

$$g(x) = e^{i2\pi kx}$$

$$b(x) = h(x) * g(x) = \int_{-\infty}^{\infty} dx' h(x') e^{i2\pi k(x-x')} = H(k) e^{i2\pi kx}$$

The output is the input times the FT of the impulse response



The oscillation with frequency k has been modified in phase ϕ and amplitude A

$$A = |H(k)| \quad \phi = \arg(H(k))$$

$$H(k) = A e^{i\phi}$$



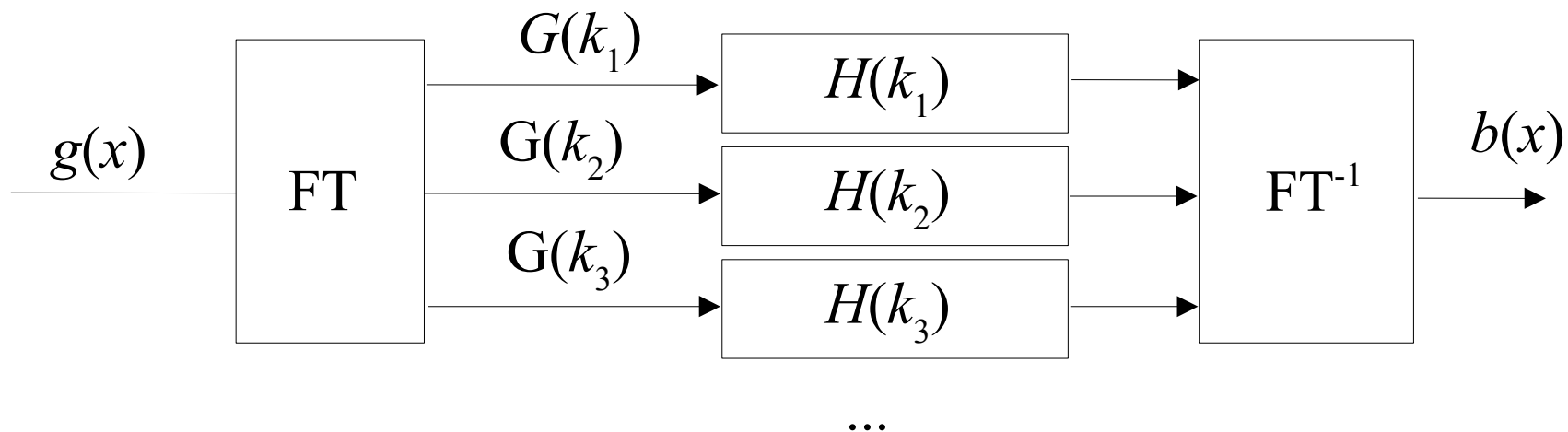
Fourier Domain System Response

FT inversion formula tells us that arbitrary input $g(x)$ can be decomposed into sum of oscillations weighted by $G(k)$

$$g(x) = \int_{-\infty}^{\infty} dk G(k) e^{i2\pi kx}$$

The system response to that is given by the convolution theorem

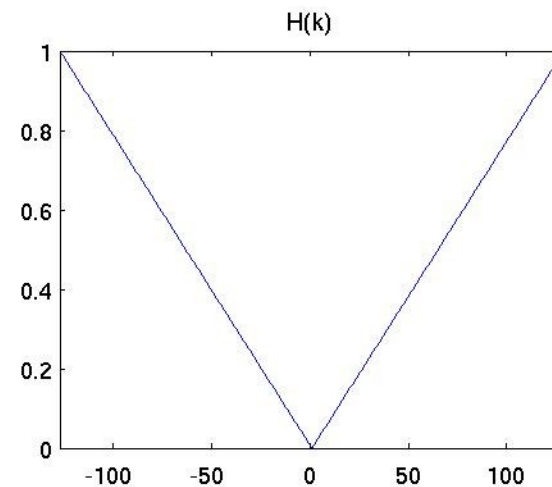
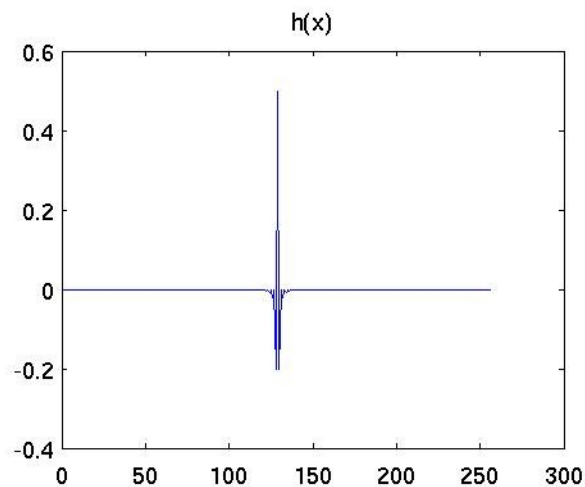
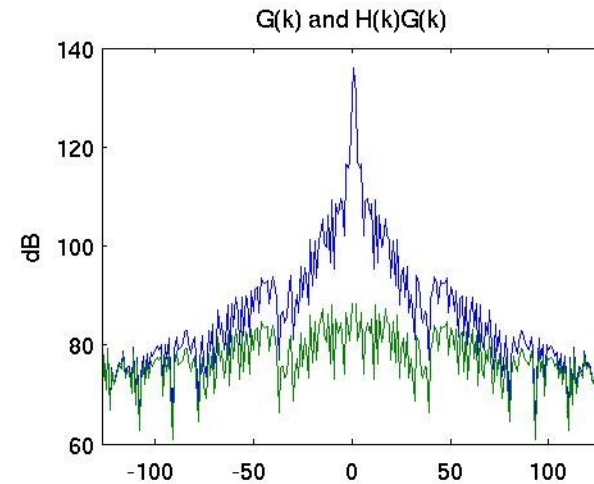
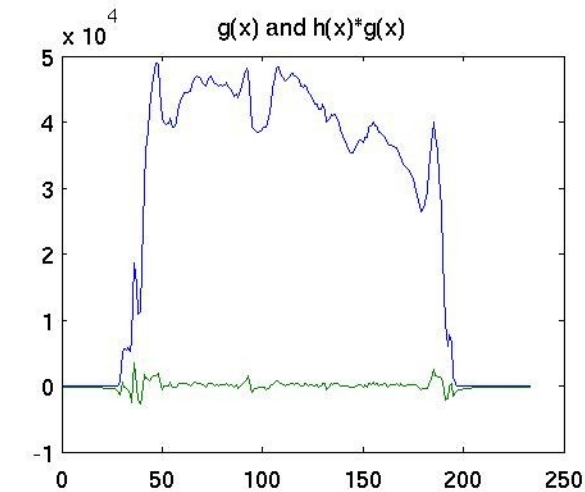
$$B(k) = H(k)G(k)$$





Fourier Domain System Response

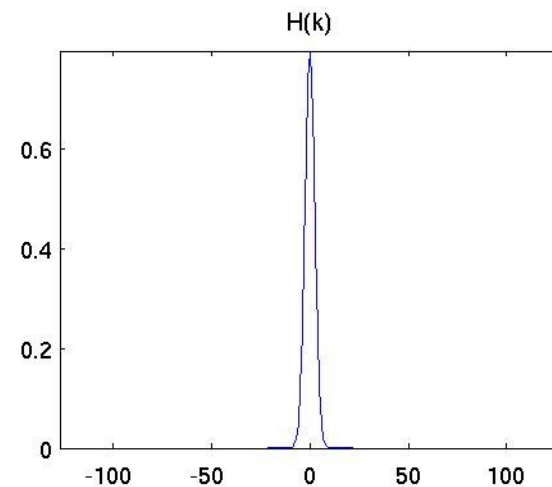
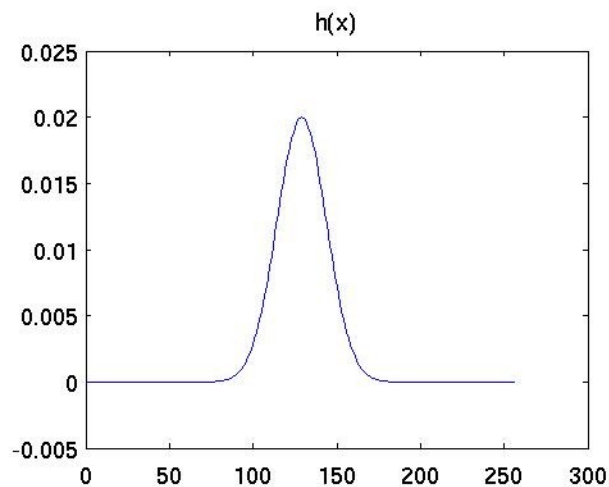
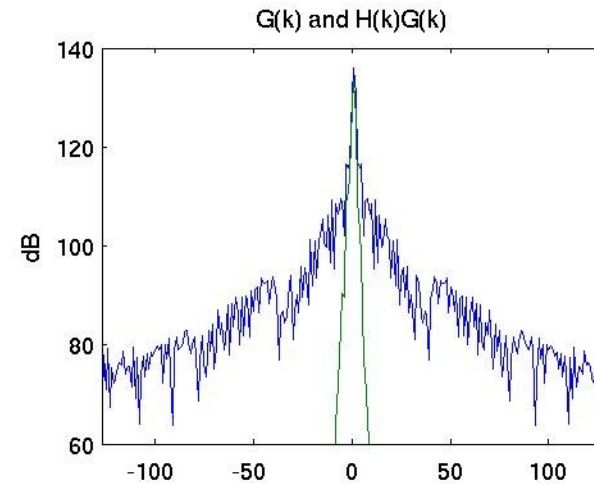
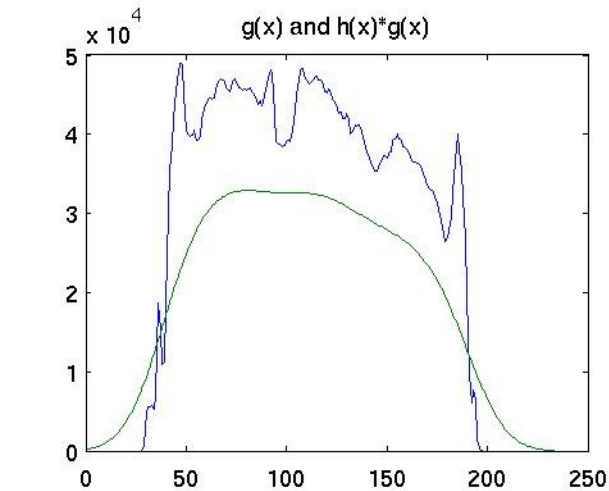
Radon inverse filter kernel, $H(k) = |k|$ (high pass filter)





Fourier Domain System Response

Gaussian low-pass filter kernel, $H(k) = \exp\left(\frac{-k^2}{2\sigma^2}\right)$, $\sigma = 1.27$



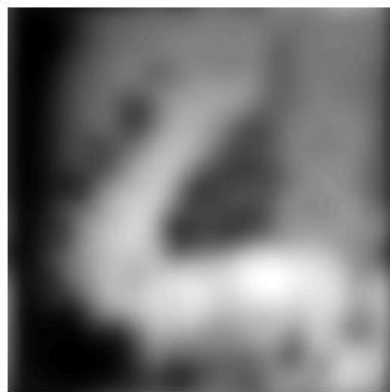


Fourier Domain System Response 2D

Original image



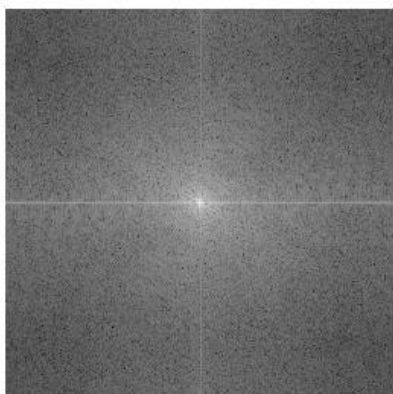
Low pass filtered image



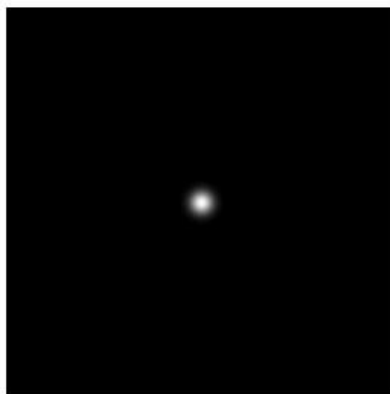
High pass filtered image



Original Spectrum (dB)



$|H(k)|$ (low pass)



$|H(k)|$ (high pass)





Fourier Transform – FFT

The numerical implementation of the FT is discrete in x and k is referred to as Discrete Fourier Transform (DFT).

$$G[k] = \sum_{x=0}^{N-1} g[x] e^{-j2\pi kx/N}$$
$$g[x] = \frac{1}{N} \sum_{k=0}^{N-1} G[k] e^{j2\pi kx/N}$$

- A fast algorithm is available (FFT) to compute the DFT in only $N \log_2 N$ operations instead of N^2 .
- Convolution with long filters is therefore often implemented using the FFT.
- FFT requires N to be a power of 2.



Fourier Transform – Filter with FFT

To filter signals with lengths not a power of 2:

- Pad zeros at the end up to next power of 2
- FFT
- multiply each frequency
- inverse FFT
- keep only first L values.

Matlab $b(x,y) = h(x,y) * g(x,y)$ using fft (assumes h smaller than g):

```
L = size(g);  
N = 2.^ceil(log2(L));  
b = ifft2(fft2(g,N(1),N(2)).*fft2(h,N(1),N(2)));  
if isreal([g;h]) b = real(b); end  
b = b(1:L(1),1:L(2));
```

Demonstrate: `fftshift`, inverse filtering, separability.

Assignment FT: Generate image with randomized phase and original amplitude, and image with random amplitude and original phase.