

**Fourier Transform - Part I**

- Introduction to Fourier Transform
  - Image Transforms
  - Basis to Basis
  - Fourier Basis Functions
  - Fourier Coefficients
- Fourier Transform - 1D
- Fourier Transform - 2D

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## The Fourier Transform



Jean Baptiste Joseph Fourier

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## Efficient Data Representation

- Data can be represented in many ways.
- There is a great advantage using an appropriate representation.
- It is often appropriate to view images as combinations of waves.

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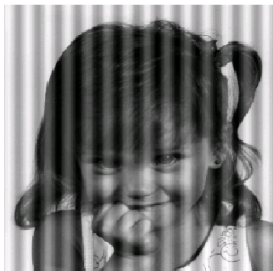
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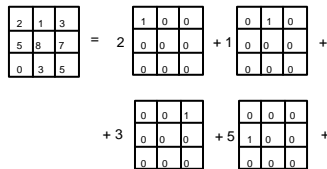
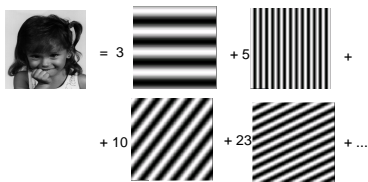
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How can we enhance such an image?



### Solution: Image Representation



### The inverse Fourier Transform

- For linear-systems we saw that it is convenient to represent a signal  $f(x)$  as a sum of scaled and shifted sinusoids.

$$f(x) = \int_{-\infty}^{\infty} F(w) e^{i2\pi wx} dw$$

How is this done?

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## Transforms: Change of Basis

Standard Basis

Grayscale Image

X Coordinate

New Basis

Fourier Image

Frequency Coordinate

Standard Basis:

$$[a_1 \ a_2 \ a_3 \ a_4] = a_1 [1 \ 0 \ 0 \ 0] + a_2 [0 \ 1 \ 0 \ 0] + a_3 [0 \ 0 \ 1 \ 0] + a_4 [0 \ 0 \ 0 \ 1]$$

Hadamard Transform:

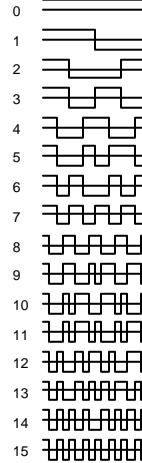
$$\begin{aligned} [2 \ 1 \ 0 \ 1] &= \\ &= 1 [1 \ 1 \ 1 \ 1] + 1/2 [1 \ 1 \ -1 \ -1] - 1/2 [-1 \ 1 \ 1 \ -1] + 0 [-1 \ 1 \ -1 \ 1] \\ &= [1 \ 1/2 \ -1/2 \ 0]_{\text{Hadamard}} \end{aligned}$$

1. Basis Functions.
2. Method for finding the image given the transform coefficients.
3. Method for finding the transform coefficients given the image.

## Hadamard Basis Functions - 1D

Wave Number

N = 16



## Finding the transform coefficients

Signal:  $X = [2 \ 1 \ 0 \ 1]_{\text{standard}}$

New Basis:

$$\begin{aligned} T_0 &= [1 \ 1 \ 1 \ 1] \\ T_1 &= [1 \ 1 \ -1 \ -1] \\ T_2 &= [-1 \ 1 \ 1 \ -1] \\ T_3 &= [-1 \ 1 \ -1 \ 1] \end{aligned}$$

New Coefficients:

$$\begin{aligned} a_0 &= \langle X, T_0 \rangle = \langle [2 \ 1 \ 0 \ 1], [1 \ 1 \ 1 \ 1] \rangle / 4 = 1 \\ a_1 &= \langle X, T_1 \rangle = \langle [2 \ 1 \ 0 \ 1], [1 \ 1 \ -1 \ -1] \rangle / 4 = 1/2 \\ a_2 &= \langle X, T_2 \rangle = \langle [2 \ 1 \ 0 \ 1], [-1 \ 1 \ 1 \ -1] \rangle / 4 = -1/2 \\ a_3 &= \langle X, T_3 \rangle = \langle [2 \ 1 \ 0 \ 1], [-1 \ 1 \ -1 \ 1] \rangle / 4 = 0 \end{aligned}$$

Signal:  $X = [1 \ 1/2 \ -1/2 \ 0]_{\text{new}}$



For continuous images/signals  $f(x)$ :

1) The number of Basis Elements  $B_i$  is  $\infty$ .

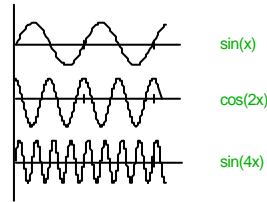
$$f(x) = \int_i a_i B_i(x) dx$$

2) The dot product:

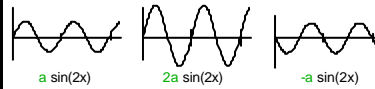
$$\langle f(x), B_i(x) \rangle = \int_x f(x) B_i(x) dx$$

### Fourier Transform

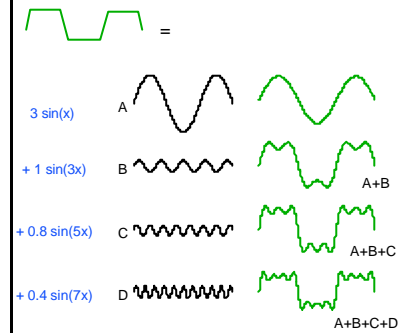
Basis Functions are sines and cosines



The transform coefficients determine the amplitude:



### Every function equals a sum of sines and cosines




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## The Fourier Transform

- The **inverse Fourier Transform** composes a signal  $f(x)$  given  $F(\omega)$

$$f(x) = \int_{\omega} F(\omega) e^{i2\pi\omega x} d\omega$$

- The **Fourier Transform** finds the  $F(\omega)$  given the signal  $f(x)$ :

$$F(\omega) = \int_x f(x) e^{-i2\pi\omega x} dx$$

- $F(\omega)$  is the Fourier transform of  $f(x)$ :

$$\tilde{F}\{f(x)\} = F(\omega)$$

- $f(x)$  is the inverse Fourier transform of  $F(\omega)$ :

$$\tilde{F}^{-1}\{F(\omega)\} = f(x)$$

- $f(x)$  and  $F(\omega)$  are a Fourier transform pair.

- The Fourier transform  $F(\omega)$  is a function over the complex numbers:

$$F(\omega) = R_{\omega} e^{i\theta_{\omega}}$$

- $R_{\omega}$  tells us how much of frequency  $\omega$  is needed.
- $\theta_{\omega}$  tells us the shift of the Sine wave with frequency  $\omega$ .

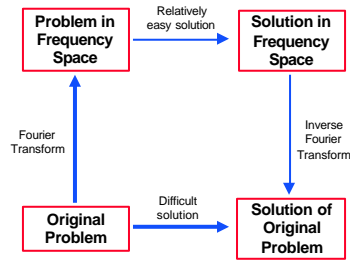
- Alternatively:

$$F(\omega) = a_{\omega} + ib_{\omega}$$

- $a_{\omega}$  tells us how much of cos with frequency  $\omega$  is needed.
- $b_{\omega}$  tells us how much of sin with frequency  $\omega$  is needed.

- $R_\omega$  - is the amplitude of  $F(\omega)$ .
- $\theta_\omega$  - is the phase of  $F(\omega)$ .
- $|R_\omega|^2 = F^*(\omega) F(\omega)$  - is the power spectrum of  $F(\omega)$ .
- If a signal  $f(x)$  has a lot of fine details  $F(\omega)$  will be high for high  $\omega$ .
- If the signal  $f(x)$  is "smooth"  $F(\omega)$  will be low for high  $\omega$ .

Why do we need representation in the frequency domain?

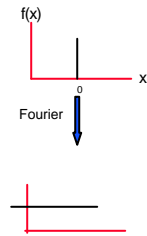


Examples:

The Delta Function:

• Let  $f(x) = d(x)$

$$F(\omega) = \int_{-\infty}^{\infty} d(x) \cdot e^{-i2\pi\omega x} = 1$$




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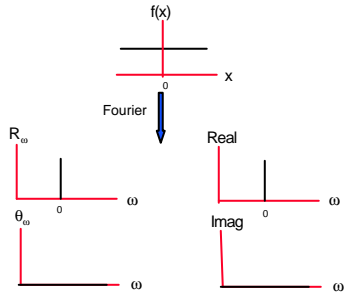
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The Constant Function:

- Let  $f(x)=1$

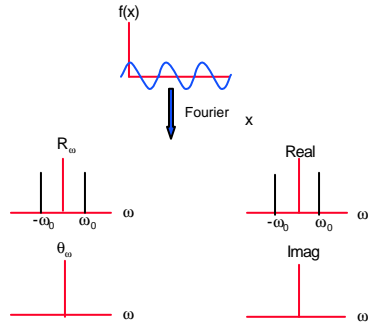
$$F(\omega) = \int_{-\infty}^{\infty} e^{-i2p\omega x} dx = d(\omega)$$



The Cosine wave:

- Let  $f(x)=\cos(2p\omega_0 x)$

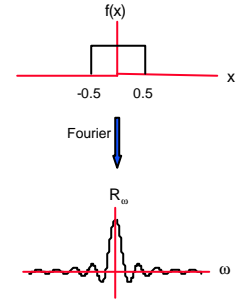
$$F(\omega) = \int_{-\infty}^{\infty} \frac{1}{2} (e^{i2p\omega_0 x} + e^{-i2p\omega_0 x}) e^{-i2p\omega x} dx = \frac{1}{2} [d(\omega - \omega_0) + d(\omega + \omega_0)]$$



The Window Function (rect):

- Let  $\text{rect}_{\frac{1}{2}}(x) = \begin{cases} 1 & \text{if } |x| < \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$

$$F(\omega) = \int_{-0.5}^{0.5} e^{-i2p\omega x} dx = \frac{\sin(p\omega)}{p\omega} = \text{sinc}(p\omega)$$




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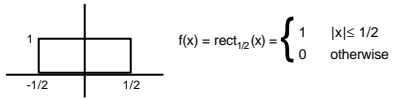
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**Proof:**



$$F(u) = \int_{-\infty}^{\infty} f(x)e^{-2\pi iux} dx = \int_{-1/2}^{1/2} e^{-2\pi iux} dx$$

$$= \frac{1}{-2\pi iu} [e^{-2\pi iux}]_{-1/2}^{1/2}$$

$$= \frac{1}{-2\pi iu} [e^{-\pi iu} - e^{\pi iu}]$$

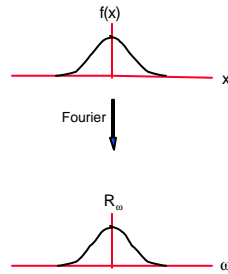
$$= \frac{1}{-2\pi iu} [\cos(\pi u) - i \sin(\pi u) - \cos(\pi u) - i \sin(\pi u)]$$

$$= \frac{\sin(\pi u)}{\pi u} = \text{sinc}(u)$$

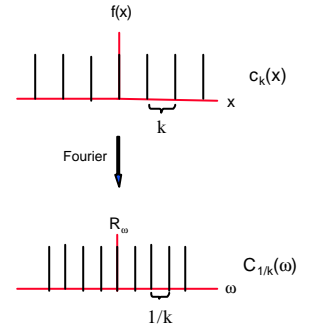
**The Gaussian:**

• Let  $f(x) = e^{-\pi x^2}$

$$F(\omega) = e^{-\pi \omega^2}$$



**The bed of nails function:**




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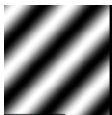
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Fourier Transform 2D - Example

2D Function



2D Fourier Transform

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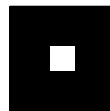
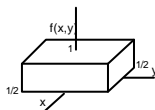
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Fourier Transform 2D - Example

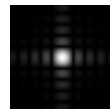


$$f(x,y) = \text{rect}(x,y) = \begin{cases} 1 & |x| \leq 1/2, |y| \leq 1/2 \\ 0 & \text{otherwise} \end{cases}$$

$$F(u,v) = \text{sinc}(u) \cdot \text{sinc}(v) = \text{sinc}(u,v)$$



|F(u,v)|




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Proof of Fourier of Rect = sinc in 2D

$$F(u) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) e^{-2\pi i(u,x+v,y)} dx dy = \int_{-1/2}^{1/2} \int_{-1/2}^{1/2} e^{-2\pi i(u,x+v,y)} dx dy$$

$$= \int_{-1/2}^{1/2} e^{-2\pi i u x} dx \int_{-1/2}^{1/2} e^{-2\pi i v y} dy$$

$$= \frac{\sin(\mathbf{p}u)}{\mathbf{p}u} \frac{\sin(\mathbf{p}v)}{\mathbf{p}v} = \text{Sinc}(u,v)$$

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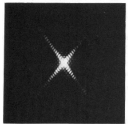
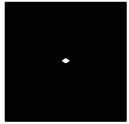
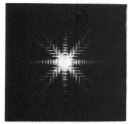
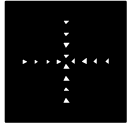
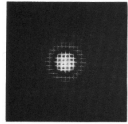
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Fourier Transform Examples

Image Domain

Frequency Domain



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