

Fourier Transform - Part I

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

The Fourier Transform



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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.

How can we enhance such an image?



Solution: Image Representation

$$\begin{array}{c} \text{Image} = 3 \begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline & & \\ \hline \end{array} + 5 \begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline & & \\ \hline \end{array} + \\ + 10 \begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline & & \\ \hline \end{array} + 23 \begin{array}{|c|c|c|} \hline & & \\ \hline & & \\ \hline & & \\ \hline \end{array} + \dots \end{array}$$

$$\begin{array}{c} \begin{array}{|c|c|c|} \hline 2 & 1 & 3 \\ \hline 5 & 8 & 7 \\ \hline 0 & 3 & 5 \\ \hline \end{array} = 2 \begin{array}{|c|c|c|} \hline 1 & 0 & 0 \\ \hline 0 & 0 & 0 \\ \hline 0 & 0 & 0 \\ \hline \end{array} + 1 \begin{array}{|c|c|c|} \hline 0 & 1 & 0 \\ \hline 0 & 0 & 0 \\ \hline 0 & 0 & 0 \\ \hline \end{array} + \\ + 3 \begin{array}{|c|c|c|} \hline 0 & 0 & 1 \\ \hline 0 & 0 & 0 \\ \hline 0 & 0 & 0 \\ \hline \end{array} + 5 \begin{array}{|c|c|c|} \hline 0 & 0 & 0 \\ \hline 1 & 0 & 0 \\ \hline 0 & 0 & 0 \\ \hline \end{array} + \dots \end{array}$$

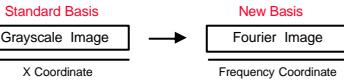
The inverse Fourier Transform

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

$$f(x) = \int_{\mathbf{w}} F(\mathbf{w}) e^{i 2 \pi \mathbf{w} \cdot \mathbf{x}} d\mathbf{w}$$

How is this done?

Transforms: Change of Basis



Standard Basis:

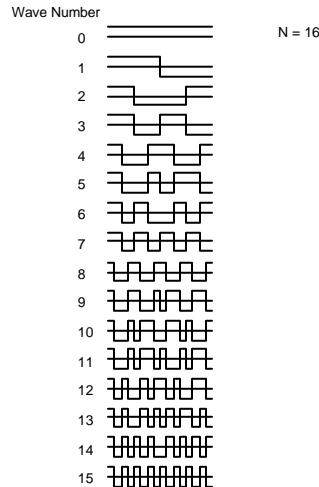
$$[a_1 \ a_2 \ a_3 \ a_4] = a_1 [1 \ 0 \ 0 \ 0] + a_2 [0 \ 1 \ 0 \ 0] + a_3 [0 \ 0 \ 1 \ 0] + a_4 [0 \ 0 \ 0 \ 1]$$

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



Finding the transform coefficients

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

New Basis: $T_0 = [1 \ 1 \ 1 \ 1]$

$T_1 = [1 \ 1 \ -1 \ -1]$

$T_2 = [-1 \ 1 \ 1 \ -1]$

$T_3 = [-1 \ 1 \ -1 \ 1]$

New Coefficients:

$$a_0 = \langle X, T_0 \rangle = \langle [2 \ 1 \ 0 \ 1], [1 \ 1 \ 1 \ 1] \rangle / 4 = 1$$

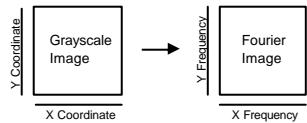
$$a_1 = \langle X, T_1 \rangle = \langle [2 \ 1 \ 0 \ 1], [1 \ 1 \ -1 \ -1] \rangle / 4 = 1/2$$

$$a_2 = \langle X, T_2 \rangle = \langle [2 \ 1 \ 0 \ 1], [-1 \ 1 \ 1 \ -1] \rangle / 4 = -1/2$$

$$a_3 = \langle X, T_3 \rangle = \langle [2 \ 1 \ 0 \ 1], [-1 \ 1 \ -1 \ 1] \rangle / 4 = 0$$

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

Transforms: Change of Basis - 2D



Standard Basis:

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = a_{11} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + a_{12} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + a_{21} \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + a_{22} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

Hadamard Transform:

$$\begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} = 1 \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} + 1/2 \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} - 1/2 \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} + 0 \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1/2 \\ -1/2 & 0 \end{bmatrix}_{\text{Hadamard}}$$

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

Standard Basis:

$$\begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} = 2 \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + 1 \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + 1 \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + 0 \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

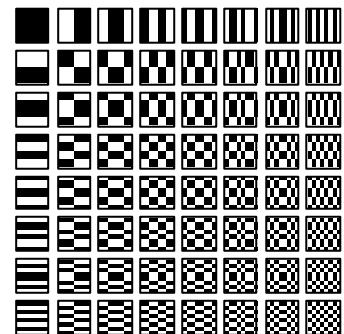
$$\begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix} \xleftrightarrow{\text{coefficients}} \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \xleftrightarrow{\text{Basis Elements}} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix}$$

Hadamard Transform:

$$\begin{bmatrix} 1 & 1/2 \\ -1/2 & 0 \end{bmatrix} = 1 \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} + 1/2 \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} - 1/2 \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} + 0 \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1/2 \\ -1/2 & 0 \end{bmatrix} \xleftrightarrow{\text{coefficients}} \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & -1 \\ 1 & -1 \end{bmatrix} \xleftrightarrow{\text{Basis Elements}} \begin{bmatrix} 1 & 1 & 1 & -1 \\ 1 & 1 & -1 & 1 \\ 1 & -1 & 1 & 1 \\ 1 & -1 & -1 & -1 \end{bmatrix}$$

Hadamard Basis Functions



size = 8x8 Black = +1 White = -1

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

1) The number of Basis Elements B_i is ∞ .

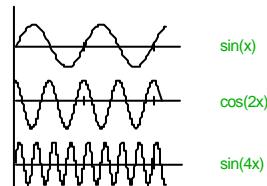
$$f(x) = \int_i a_i B_i(x) dx$$

2) The dot product:

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

Fourier Transform

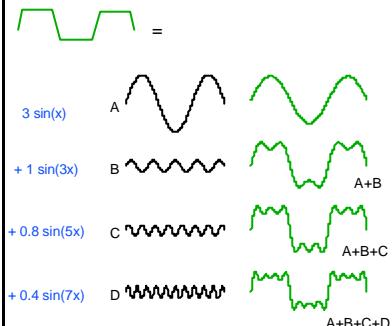
Basis Functions are sines and cosines



The transform coefficients determine the amplitude:



Every function equals a sum of sines and cosines



The Fourier Transform

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

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

- 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_ω tells us how much of frequency ω is needed.
- θ_ω tells us the shift of the Sine wave with frequency ω .

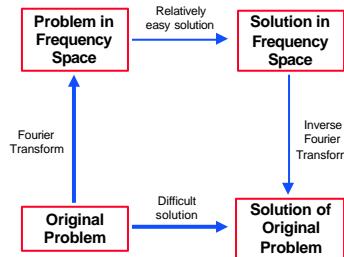
- Alternatively:

$$F(\omega) = a_\omega + i b_\omega$$

- a_ω tells us how much of cos with frequency ω is needed.
- b_ω tells us how much of sin with frequency ω is needed.

- R_ω - is the amplitude of $F(\omega)$.
- θ_ω - 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 ω .
- If the signal $f(x)$ is "smooth" $F(\omega)$ will be low for high ω .

Why do we need representation in the frequency domain?

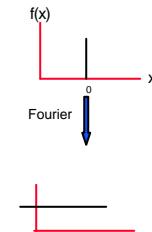


Examples:

The Delta Function:

- Let $f(x) = \delta(x)$

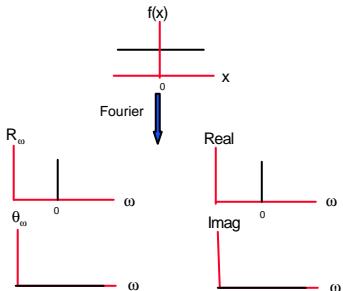
$$F(w) = \int_{-\infty}^{\infty} \delta(x) \cdot e^{-i2\pi w x} dx = 1$$



The Constant Function:

- Let $f(x) = 1$

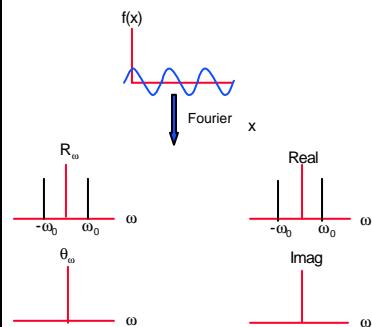
$$F(w) = \int_{-\infty}^{\infty} e^{-i2\pi wx} dx = \delta(w)$$



The Cosine wave:

- Let $f(x) = \cos(2\pi w_0 x)$

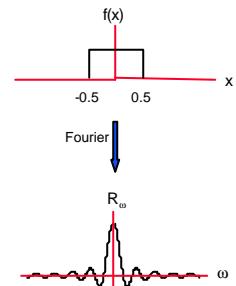
$$\begin{aligned} F(w) &= \int_{-\infty}^{\infty} \frac{1}{2} (e^{i2\pi w_0 x} + e^{-i2\pi w_0 x}) e^{-i2\pi wx} dx = \\ &= \frac{1}{2} [\delta(w-w_0) + \delta(w+w_0)] \end{aligned}$$



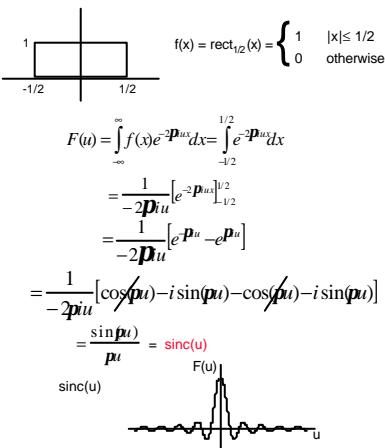
The Window Function ($\text{rect}(x)$):

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

$$F(w) = \int_{-0.5}^{0.5} e^{-i2\pi wx} dx = \frac{\sin(\pi w)}{\pi w} = \text{sinc}(pw)$$



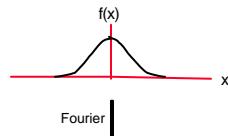
Proof:



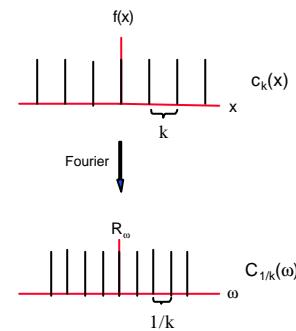
The Gaussian:

- Let $f(x) = e^{-px^2}$

$$F(u) = e^{-pw^2}$$



The bed of nails function:



Fourier Transform - 2D

Given a continuous real function $f(x,y)$, its Fourier transform $F(u,v)$ is defined as:

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

The Inverse Fourier Transform:

$$F^{-1}\{F(u,v)\} = f(x,y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} F(u,v) e^{2\pi i(ux+vy)} du dv$$

$$F(u,v) = a(u,v) + ib(u,v) = |F(u,v)| e^{i\phi(u,v)}$$

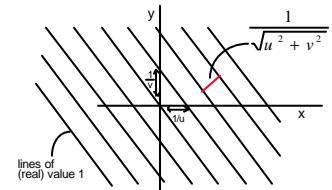
$$\text{Phase} = \phi(u,v) = \operatorname{tg}^{-1}(b(u,v)/a(u,v))$$

$$\text{Spectrum (Amplitude)} = |F(u,v)| = \sqrt{a^2(u,v) + b^2(u,v)}$$

$$\text{Power Spectrum} = |F(u,v)|^2 = a^2(u,v) + b^2(u,v)$$

Fourier Wave Functions - 2D

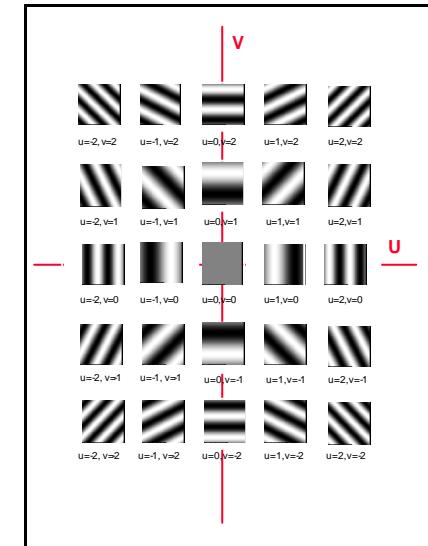
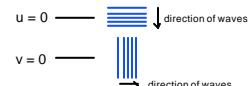
$F(u,v)$ is the coefficient of the sine wave $e^{2\pi i(ux+vy)}$



$$e^{2\pi i(ux+vy)} = \cos(2\pi(ux+vy)) + i\sin(2\pi(ux+vy))$$

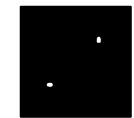
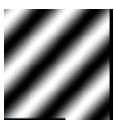
The ratio $\frac{u}{v}$ determines the Direction.

The size of u,v determines the Frequency.



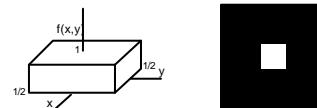
Fourier Transform 2D - Example

2D Function



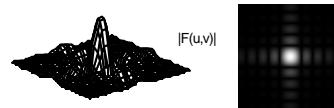
2D Fourier Transform

Fourier Transform 2D - Example



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

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



Proof of Fourier of Rect = sinc in 2D

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

$$= \int_{-1/2}^{1/2} e^{-2\pi i ux} dx \int_{-1/2}^{1/2} e^{-2\pi i uy} dy$$

$$= \frac{\sin(\pi u)}{\pi u} \frac{\sin(\pi v)}{\pi v} = Sinc(u,v)$$

Fourier Transform Examples

Image Domain Frequency Domain

