

Concepts and Algorithms of Scientific and Visual Computing

–Signal Analysis–



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Review: Fourier Transform

For functions $f \in L^1(\mathbb{R}^n)$, the **Fourier transform** of f is given by

$$\hat{f}(\xi) = \int_{\mathbb{R}^n} f(x) e^{-2\pi i \xi \cdot x} dx$$

and the corresponding **inverse Fourier transform** of \hat{f} is given by

$$f(x) = \int_{\mathbb{R}^n} \hat{f}(\xi) e^{2\pi i \xi \cdot x} d\xi$$

where x, ξ are **vectors** in \mathbb{R}^n .

Recall: $f \in L^1(\mathbb{R}^n) \iff$

$$\int_{\mathbb{R}^n} |f(x)| dx < \infty$$

This matters—signal vanishes at extreme frequencies (**Riemann-Lebesgue Lemma**).

Riemann-Lebesgue Lemma

Lemma (Riemann-Lebesgue): If f is L^1 integrable on \mathbb{R}^n , then

$$\hat{f}(\xi) = \int_{\mathbb{R}^n} f(x) e^{-2\pi i \xi \cdot x} dx \rightarrow 0 \text{ as } |\xi| \rightarrow \infty.$$

Proof (n=1): Suppose f is smooth and has compact support. First apply integration by parts to $|\hat{f}|$, yielding

$$\begin{aligned} \left| \int_{\mathbb{R}} f(x) e^{-2\pi i \xi x} dx \right| &= \left| \int_{\mathbb{R}} \frac{1}{2\pi i \xi} f'(x) e^{-2\pi i \xi x} dx \right| \\ &\leq \frac{1}{|\xi|} \int_{\mathbb{R}} |f'(x)| dx \rightarrow 0 \text{ as } |\xi| \rightarrow \infty \end{aligned}$$

So $|\hat{f}| \rightarrow 0 \implies \hat{f} \rightarrow 0$ as $\xi \rightarrow \pm\infty$. (continued on next slide)

Riemann-Lebesgue Lemma

If f does not have compact support, then since it is L^1 , it can be approximated arbitrarily closely by a smooth function g that does have compact support. So pick a g such that $\|f - g\|_{L^1} < \epsilon$. Then

$$\begin{aligned} |\hat{f}| &= \left| \int_{\mathbb{R}} (f - g)e^{-2\pi i \xi x} dx + \int_{\mathbb{R}} ge^{-2\pi i \xi x} dx \right| \\ &\leq \left| \int_{\mathbb{R}} (f - g)e^{-2\pi i \xi x} dx \right| + \left| \int_{\mathbb{R}} ge^{-2\pi i \xi x} dx \right| \end{aligned}$$

The right-hand term goes to zero by the previous slide, and the left-hand term can be bounded by ϵ (L^1 assumption). Therefore, the whole term goes to zero. \square

So as long as f is L^1 integrable, its Fourier modes vanish at $\pm\infty$. (Multi-dimensional proof is basically identical)

Fourier Transform Examples

The Fourier transform of a Gaussian $f(x) = e^{-ax^2}$ is also a Gaussian:

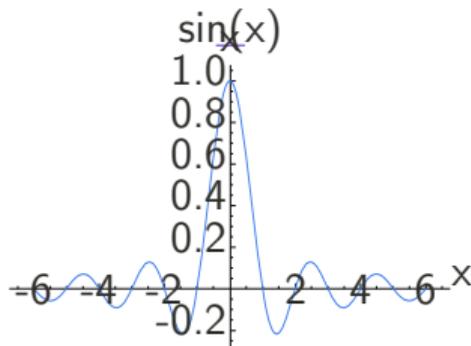
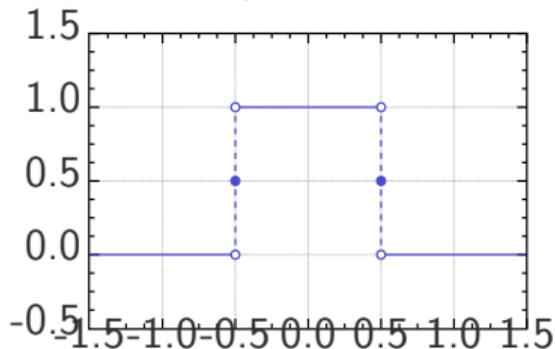
$$\mathcal{F}(f) = \int_{-\infty}^{\infty} e^{-ax^2} e^{-2\pi i x \xi} dx = \int_{-\infty}^{\infty} e^{-(ax^2 + 2\pi i \xi x)} dx$$

Complete the square inside the exponential. Want $(\sqrt{a}x + b)^2$ inside the exponent
 $\implies b = \pi i \xi / \sqrt{a}$. So multiply both sides of equation by $e^{\pi^2 \xi^2 / a}$:

$$\begin{aligned} e^{\pi^2 \xi^2 / a} \hat{f}(\xi) &= \int_{-\infty}^{\infty} e^{-(\sqrt{a}x + b)^2} dx = \int_{-\infty}^{\infty} e^{-z^2} dz = \sqrt{\pi} \\ \implies \hat{f}(\xi) &= \sqrt{\pi} e^{-\pi^2 \xi^2 / a} \end{aligned}$$

Fourier Transform Examples

Not all Fourier transforms are invertible! Consider rectangular function (L^1 integrable), whose Fourier transform the sinc function, which is not Lebesgue integrable. (Image credits: Wikipedia. Display errors: mine.)



Fourier Transform Examples

Demo time... (musical analysis + harmonics)

Real-time Fourier transform visualization

Some Fourier Transform Properties

We can see that differentiation in the time domain corresponds to multiplication in the Fourier domain:

$$\mathcal{F}(f'(x))(\xi) = \int_{-\infty}^{\infty} f'(x)e^{-2\pi i \xi x} dx$$

Integration by parts \implies

$$\begin{aligned} &= f(x)e^{-2\pi i \xi x} \Big|_{-\infty}^{\infty} + \int_{-\infty}^{\infty} 2\pi i \xi f(x)e^{-2\pi i \xi x} dx \\ &= 2\pi i \xi \mathcal{F}(f(x))(\xi) \end{aligned}$$

\implies The Fourier transform can make our lives easier!

Some Fourier Transform Properties

For two functions $f, g \in L^1(\mathbb{R}^n)$, their **convolution** is defined to be the function

$$h(z) = \int_{\mathbb{R}^n} f(x)g(z-x)dx$$

and the **pointwise product** is just the function given by $f(x)g(x)$.

The Fourier transform of a convolution is a pointwise product in frequency space:

$$\hat{h}(\xi) = F(\xi)G(\xi)$$

where F and G are the respective Fourier transforms of f and g . Let's derive this...

Some Fourier Transform Properties

First: does h even have a Fourier transform? Check:

$$\begin{aligned}\|h\|_{L^1} &= \int_{\mathbb{R}^n} |h(x)| dx = \int_{\mathbb{R}^n} \left| \int_{\mathbb{R}^n} f(x)g(z-x) dz \right| dx \\ &\leq \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} |f(x)g(z-x)| dz dx = \int_{\mathbb{R}^n} |f(x)| \int_{\mathbb{R}^n} |g(z-x)| dz dx \\ &= \int_{\mathbb{R}^n} |f(x)| \|g\|_{L^1} dx = \|f\|_{L^1} \|g\|_{L^1} < \infty\end{aligned}$$

since $f, g \in L^1(\mathbb{R}^n)$. So $h \in L^1(\mathbb{R}^n)$, which means that h does indeed have a valid Fourier transform. Now let's show that the Fourier transform is what we claim...

Some Fourier Transform Properties

$$\begin{aligned}H(\xi) &= \int_{\mathbb{R}^n} h(z)e^{-2\pi i\xi \cdot z} dz = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} f(x)g(z-x)dx e^{-2\pi i\xi \cdot z} dz \\&= \int_{\mathbb{R}^n} f(x) \left(\int_{\mathbb{R}^n} g(z-x)e^{-2\pi i\xi \cdot z} dz \right) dx \text{ by Fubini's Theorem} \\&= \int_{\mathbb{R}^n} f(x) \left(\int_{\mathbb{R}^n} g(y)e^{-2\pi i\xi \cdot (y+x)} dy \right) dx \text{ substituting } y = z - x \\&= \int_{\mathbb{R}^n} f(x)e^{-2\pi i\xi \cdot x} \left(\int_{\mathbb{R}^n} g(y)e^{-2\pi i\xi \cdot y} dy \right) dx = \int_{\mathbb{R}^n} f(x)e^{-2\pi i\xi \cdot x} dx \int_{\mathbb{R}^n} g(y)e^{-2\pi i\xi \cdot y} dy\end{aligned}$$

which is exactly $F(\xi)G(\xi)$. So the Fourier transform of a convolution is a pointwise product. \square

Some Fourier Transform Properties

Similar properties hold with the convolution / pointwise product. **Convolution Theorem** says

$$H(\xi) = F(\xi)G(\xi) \text{ (just proved)}$$

$$\widehat{fg}(\xi) = F(\xi) * G(\xi) \text{ (convolution)}$$

$$(f * g)(x) = \mathcal{F}^{-1}(F(\xi)G(\xi))$$

$$(fg)(x) = \mathcal{F}^{-1}(F(\xi) * G(\xi))$$

Sanity check: why do we care about all these properties we've been deriving?

Solving a Simple Differential Equation with Fourier Transforms

The ODE

$$\frac{d^2y}{dt^2} + 2r\frac{dy}{dt} + \omega_0^2y = \delta(t - t^*)$$

is the equation of motion for a driven, damped simple harmonic oscillator (think: spring system with friction) with “damping constant” r and natural frequency ω_0 , with some impulse instantaneously applied at positive time $t = t^*$.

How do we solve this ODE? Not totally obvious... Fourier transforms to the rescue!

$$\begin{aligned} &(-4\pi^2\xi^2 + 4r\pi i\xi + \omega_0^2)\hat{y} = e^{-2\pi i\xi t^*} \\ \implies y(t) &= \int_{-\infty}^{\infty} \frac{e^{-2\pi i\xi t^*}}{-4\pi^2\xi^2 + 4r\pi i\xi + \omega_0^2} e^{2\pi i\xi t} d\xi \end{aligned}$$

Still a hard (this one not impossible) integral, but this is a better situation to be in

Some Applications of Convolutions

Convolving an image with a Gaussian function yields a smoothed version of the image (**Gaussian Blur**) (image credit Wikipedia):



(Will discuss signal filtering like this more on Thursday...)

Some Applications of Convolutions

Convolution reverb: convolve arbitrary audio recording with an **impulse response**, an audio sample that tells how an (approximately) ideal impulse reverberates in an environment. Result is the input audio recording sounding like it's reverberating in that environment.

[Demo](#)

Uncertainty Principle

Compare the time-domain and frequency-domain plots of various Gaussians. What do you see?

[Interactive Demo](#)

Uncertainty Principle

The **dispersion about zero** $D_0(f)$ of a function f is given by the second moment of $|f(x)|$:

$$D_0(f) = \int_{-\infty}^{\infty} x^2 |f(x)|^2 dx$$

(assume f is normalized in the L^2 sense). Under proper smoothness assumptions, the Uncertainty Principle states that

$$D_0(f)D_0(\hat{f}) \geq \frac{1}{16\pi^2}$$

(Pinsky, Mark (2002), Introduction to Fourier Analysis and Wavelets).

So the more spread out / concentrated f becomes, the more concentrated / spread out \hat{f} becomes.

Uncertainty Principle

What function is “most certain?”

Recall: the Fourier transform of a Gaussian is a Gaussian, with possibly different normalization constants.

For

$$f(x) = \frac{2^{1/4}}{\sqrt{\sigma}} e^{-\pi x^2/\sigma^2},$$

$$\hat{f}(\xi) = \sigma \frac{2^{1/4}}{\sqrt{\sigma}} e^{-\pi \sigma^2 \xi^2}.$$

One can plug into the definition of dispersion about zero to see that for this Fourier transform pair,

$$D_0(f)D_0(\hat{f}) = \frac{1}{16\pi^2}$$

which means that this function is close as you can get to beating the Uncertainty Principle.

Next time: Heisenberg uncertainty principle in particular