Fourier Series & Fourier Transforms

The fundamental idea is that any **periodic function** (i.e. one that repeats after a particular interval) can be expressed as an **infinite sum of sine and cosine waves** of different amplitude and frequency. This can be generalized to an *integral* (**The Fourier Transform**), which has wide ranging applications for the solution of **differential equations** which occur in the Physical sciences. (e.g. spectral analysis of sound and light, conduction of heat, diffraction of waves ...).

 $2n\pi x$

In essence, the **Fourier transform of a time varying signal will yield its harmonic content**, i.e. the frequencies which comprise it. A **Discrete Fourier Transform** (DFT). (which operates on signal amplitudes which are sampled at a fixed rate) can be efficiently implemented in a computer program via a **Fast Fourier Transform** (FFT).

Fourier Series

Consider a function which is periodic after interval L f(x+L) = f(x)

Define the infinite series:
$$f(x) = a_0 + \sum_{n=1}^{\infty} a_n \cos \frac{2n\pi x}{L} + \sum_{n=1}^{\infty} b_n \sin \frac{2n\pi x}{L}$$

How do we find out what the coefficients a_n and b_n are?

Note the following trigonometric identities:

$$2 \cos A \cos B = \cos(A+B) + \cos(A-B)$$

$$2 \sin A \cos B = \sin(A+B) + \sin(A-B)$$

$$2 \sin A \sin B = \cos(A-B) - \cos(A+B)$$

Hence if *n*, *m* are integers, and $n \neq m$

$$I = \int_0^L 2\cos\frac{2n\pi x}{L}\cos\frac{2m\pi x}{L}dx$$
$$I = \int_0^L \left(\cos\frac{2(n+m)\pi x}{L} + \cos\frac{2(n-m)\pi x}{L}\right)dx$$
$$I = \frac{L}{2\pi} \left[\frac{1}{n+m}\sin\frac{2(n+m)\pi x}{L} + \frac{1}{n-m}\sin\frac{2(n-m)\pi x}{L}\right]_0^L$$
$$I = 0$$

If
$$n = m$$
 $I = \int_{0}^{L} 2\cos^{2}\frac{2n\pi x}{L}dx = \int_{0}^{L} \left(1 + \cos\frac{4n\pi x}{L}\right)dx$
 $I = \left[x + \frac{L}{4n\pi}\sin\frac{4n\pi x}{L}\right]_{0}^{L} = L$
 $\cos^{2} x = \frac{1}{2}(1 + \cos 2x)$

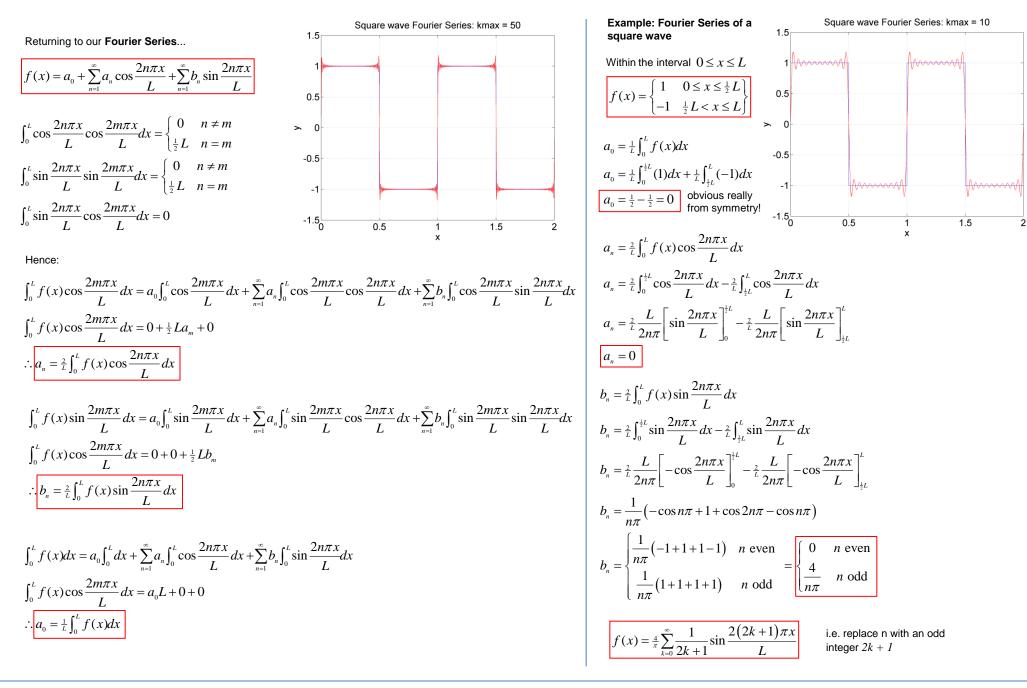
Hence:
$$\int_{0}^{L} \cos \frac{2n\pi x}{L} \cos \frac{2m\pi x}{L} dx = \begin{cases} 0 & n \neq m \\ \frac{1}{2}L & n = m \end{cases}$$

We can repeat this analysis for other combinations of sine and cosine terms
If
$$n, m$$
 are integers, and $n \neq m$
If $n = m$



Joseph Fourier

1768-1830



We can extend the ideas developed for the Fourier Series to define a **Fourier Transform**

$$f(x) = \sum_{n=-\infty}^{\infty} a_n \left(\cos \frac{2n\pi x}{L} + i \sin \frac{2n\pi x}{L} \right)$$
 Define this *complex* series

$$f(x) = \sum_{n=-\infty}^{\infty} a_n e^{i\frac{2n\pi x}{L}} \quad \text{de-Moivre's Theorem } e^{ix} = \cos x + i \sin x$$

$$\int_{-\frac{1}{2}L}^{\frac{1}{2}L} f(x) e^{-i\frac{2m\pi x}{L}} dx = \sum_{n=-\infty}^{\infty} a_n \int_{-\frac{1}{2}L}^{\frac{1}{2}L} e^{i\frac{2(n-m)\pi x}{L}} dx$$

$$\int_{-\frac{1}{2}L}^{\frac{1}{2}L} e^{i\frac{2(n-m)\pi x}{L}} dx = \begin{cases} 0 & n \neq m \\ L & n = m \end{cases}$$
Complex Fourier Series

$$\therefore \int_{-\frac{1}{2}L}^{\frac{1}{2}L} f(x) e^{-i\frac{2m\pi x}{L}} dx = La_n$$

$$\therefore a_n = \frac{1}{L} \int_{-\frac{1}{2}L}^{\frac{1}{2}L} f(x) e^{-i\frac{2\pi\pi x}{L}} dx$$
Hence:

$$f(x) = \sum_{n=-\infty}^{\infty} a_n e^{i\frac{2\pi\pi x}{L}}$$

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$$f(x) = \sum_{n=-\infty}^{\infty} \frac{1}{L} \int_{-\frac{1}{2}L}^{\frac{1}{2}L} f(x) e^{-i\frac{2\pi\pi x}{L}} dx \times e^{i\frac{2\pi\pi x}{L}}$$

$$k = \frac{2\pi n}{L}, \quad \Delta k = \frac{2\pi}{L}$$

$$f(x) = \frac{1}{2\pi} \sum_{n=-\infty}^{\infty} \Delta k \int_{-\frac{1}{2}L}^{\frac{1}{2}L} f(x) e^{-ikx} dx \times e^{ikx}$$

$$L \to \infty$$
In this limit we can turn the infinite sum into an integral f(x) \to \frac{1}{2\pi} \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} f(x) e^{-ikx} dx\right) e^{ikx} dk

For analysis of a

spatial waveform

 $k = \frac{2\pi}{\lambda}$

we can associate k

with the wavenumber

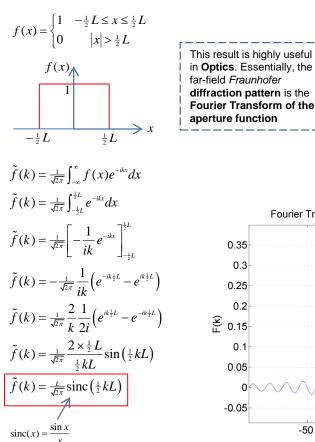
Define the **Fourier Transform** of f(x) $\tilde{f}(k) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-ikx} dx$ $\therefore f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \tilde{f}(k) e^{ikx} dk$

i.e. the Inverse Fourier Transform

For a time signal

$$\tilde{f}(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(t) e^{-i\omega x} dx$$
$$\therefore f(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \tilde{f}(\omega) e^{i\omega x} dk$$
$$\omega = \frac{2\pi}{T}$$
Angular frequency Period

Fourier Transform of a 'top hat' function

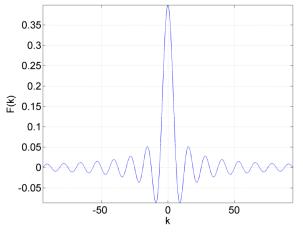


A **diffraction grating** illuminated by a laser will result in a pattern of bright lines angularly separated. The main line spacings are defined by the *separation of the thin slits* which comprise the grating.

The entire pattern will also have an *amplitude envelope* (given by a sinc function) which results from the finite width of the slit. If a slit is uniformly illuminated by the laser, we can model its aperture function by a 'top hat.'



Fourier Transform of a Top Hat function. L = 1



Fourier Transform of a Gaussian function

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

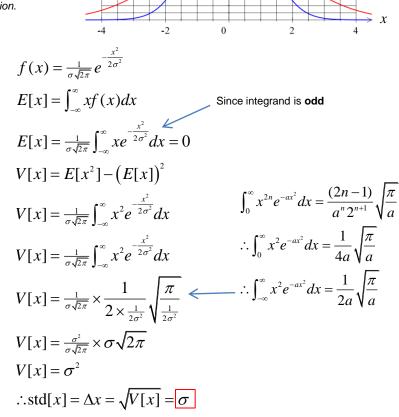
$$\int_{-\infty}^{\infty} f(x) dx = \frac{1}{\sigma\sqrt{2\pi}} 2\int_{0}^{\infty} e^{-\frac{x^2}{2\sigma^2}} dx = \frac{2}{\sigma\sqrt{2\pi}} \times \frac{1}{2} \times \sqrt{\frac{\pi}{\frac{1}{2\sigma^2}}} = 1$$

$$\begin{split} \tilde{f}(k) &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-ikx} dx & \text{This normal the Gaussic can be use probability} \\ \tilde{f}(k) &= \frac{1}{2\sigma\pi} \int_{-\infty}^{\infty} e^{-\frac{1}{2\sigma^2}x^2} e^{-ikx} dx & \text{probability} \\ \tilde{f}(k) &= \frac{1}{2\sigma\pi} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2\sigma^2}x^2 - ikx\right) dx \\ \tilde{f}(k) &= \frac{1}{2\sigma\pi} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2\sigma^2}\left\{x^2 + 2i\sigma^2kx\right\}\right) dx \\ \tilde{f}(k) &= \frac{1}{2\sigma\pi} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2\sigma^2}\left\{\left(x + i\sigma^2k\right)^2 - i^2\sigma^2k^2\right\}\right) dx \\ \tilde{f}(k) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\sigma^2k^2} \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp\left(-\frac{1}{2\sigma^2}\left(x + i\sigma^2k\right)^2\right) dx \\ z &= x + i\sigma^2k, \ dz &= dx \\ \tilde{f}(k) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\sigma^2k^2} \underbrace{\frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp\left(-\frac{z^2}{2\sigma^2}\right) dz}_{=1} & \stackrel{*}{\tilde{f}(k) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\sigma^2k^2}} \end{split}$$

So the Fourier Transform of Gaussian function is *also* a Gaussian function.

Note the **standard deviation** of the Gaussian and its Fourier Transform are *inversely related*. This property underpins the **Uncertainty Principle** in **Quantum Mechanics**, where the *wavefunction* of quantities such as **momentum** are related via a Fourier Transform to the wavefunction of **position**.

This **normalization** means the Gaussian function can be used as a *probability distribution*.



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

 $\sigma = 1$

 $\sigma = 2$

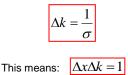
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So if the **standard deviation** of a Gaussian function is:



The standard deviation of the Fourier Transform of the Gaussian (which is *also* a Gaussian) is



In **Quantum Mechanics**, the uncertainties in position and momentum are related to the standard deviations of the probability density functions associated with their wavefunctions. (Actually the *modulus square* of the wavefunction – the **Born Interpretation**).

Now for a particle of wavevector k, its momentum p is:

$$p = \frac{h}{\lambda} = \frac{\hbar}{\lambda/2\pi}$$
 de-Broglie
relation
$$\therefore p = \hbar k$$

$$\therefore \Delta p \Delta x = \hbar \Delta k \Delta x = \hbar$$

$$\therefore \Delta p \Delta x = \hbar$$

The wavefunctions are not always Gaussian, so the **Uncertainty Principle** is actually an *inequality*



*Strictly speaking we ought to perform a contour integral here, but as in Riley, Hobson, Bence; Mathematical Methods for Physics and Engineering, pp436, we shall be cavalier and assume the high frequency oscillations caused by the imaginary terms are negligible!

Summary of Fourier Transform theorems

$$\tilde{f}(k) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-ikx} dx$$
$$f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \tilde{f}(k) e^{ikx} dk$$

Definition of the Fourier Transform and Inverse Fourier Transform

$$\Im[f(x)] = \tilde{f}(k)$$

'Operator' notation for 'find the Fourier Transform of the function
$$f(x)$$
'

$$\Im [f(x-a)] = e^{-iak} \tilde{f}(k)$$

$$\Im [e^{iax} f(x)] = \tilde{f}(k-a)$$

Fourier Shift Theorem. This is the algorithmic basis for many 'pitch shifting' signal processing effects

$$\Im[f(ax)] = \frac{1}{|a|} \tilde{f}(\frac{1}{a}k)$$

$$\Im\left[\frac{d^{n}}{dx^{n}}f(x)\right] = \left(ik\right)^{n}\tilde{f}(k)$$
$$\Im\left[x^{n}f(x)\right] = i^{n}\frac{d^{n}}{dk^{n}}\tilde{f}(k)$$

These theorems involving derivatives provide a powerful tool for solving differential equations. Applying a Fourier Transform will turn a differential equation into an algebraic one. Once solved, application of the Inverse Fourier Transform will yield the solution.

$$\Im \left[f(x) * g(x) \right] = \sqrt{2\pi} \, \tilde{f}(k) \tilde{g}(k)$$
$$\Im \left[f(x)g(x) \right] = \frac{1}{\sqrt{2\pi}} \, \tilde{f}(k) * \tilde{g}(k)$$

Convolution Theorem. This is very useful in signal processing. e.g. 'Pulse Compression' in Radar applications.

$$f(x) * g(x) = \int_{-\infty}^{\infty} f(z)g(x-z)dz$$

 $\int_{-\infty}^{\infty} f(x)g^{*}(x)dx = \int_{-\infty}^{\infty} \tilde{f}(k)\tilde{g}^{*}(k)dk$ Parseval relat $\int_{-\infty}^{\infty} |f(x)|^{2}dx = \int_{-\infty}^{\infty} |\tilde{f}(k)|^{2}dk \qquad f(x) = g(x)$

i.e. the area under the **Power Spectrum** is the same as the integral of signal power. This must be true since energy is conserved!

Note the Fourier Transform is a special case of a general class of Integral Transforms. Another, which is popular in signal processing, is the Laplace Transform.

$$F(s) = \int_0^\infty f(t) e^{-st} dt$$

The Discrete Fourier transform and spectral analysis

If a signal x(t) is sampled at rate f_s , we can define a **digital filter**, which can be used to extract useful features, while perhaps inhibiting others.

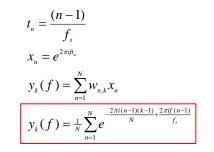
$$y = \sum_{n=1}^{N} w_n x_n$$

y is the filter output, which combines *n* subsequent *weighted* samples of the signal.

A Digital Fourier Transform (DFT) is a filter bank with filter weights

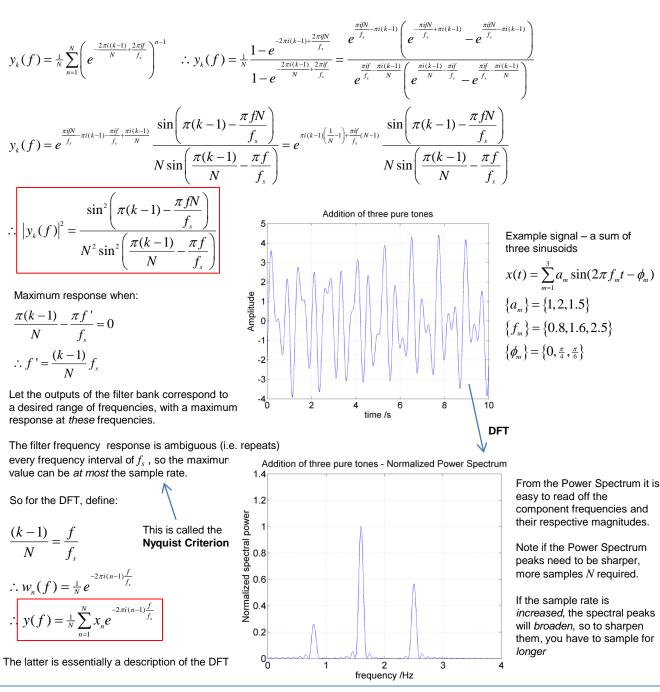
$$W_{n,k} = \frac{1}{N} e^{-\frac{2\pi i (n-1)(k-1)}{N}}$$

The filter has a **frequency response** (i.e. the output vs frequency due to N samples of a pure sinusoid)



This can be simplified by considering the sum of a geometric progression

$$a + ar + ar^{2} + \dots + ar^{N-1} = \sum_{n=1}^{N} ar^{n-1}$$
$$a + ar + ar^{2} + \dots + ar^{N-1} = a \frac{1 - r^{N}}{1 - r}$$
$$\therefore \sum_{n=1}^{N} ar^{n-1} = a \frac{1 - r^{N}}{1 - r}$$



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