

Image Processing

Waleed A. Yousef
Faculty of Computers and Information,
Helwan University.

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Ch. 4

Inner product and orthogonalization

For vectors in p dimensional space, the inner product $\langle x, y \rangle$ is the dot product $x'y$,

$$\begin{aligned}\langle x, y \rangle &= x'y \\ &= \begin{pmatrix} x_1 & \dots & x_p \end{pmatrix} \cdot \begin{pmatrix} y_1 \\ \vdots \\ y_p \end{pmatrix} \\ &= \sum_{i=1}^p x_i y_i.\end{aligned}$$

when $x'y$ is zero we say they are orthogonal. It can be shown that, for $p \leq 3$,

$$\begin{aligned}\cos \theta &= \frac{x'y}{\|x\| \cdot \|y\|} \\ &= \frac{x'y}{\sqrt{x'x} \cdot \sqrt{y'y}}\end{aligned}$$

Then, we can generalize this definition in higher dimensions, and define the angle between two vectors for $p > 3$.

Simple example (a single projection): $\mathbf{X}_{p \times 1} = X$
if $X' = (1, 0, \dots, 0)$, we simply get the first component Y_1 .

if $X' = (1/\sqrt{2}, 1/\sqrt{2})$, we project on the $\pi/4$ direction; this will be the value of the new vector in the direction of X .

$$\begin{aligned}
X'Y &= \|X\| \|Y\| \cos(Y, X) \\
&= \|X\| \times \text{Projected Length}
\end{aligned}$$

If we need the projected length only, then project on a unit vector $\frac{X'}{\|X\|}$, then

$$\frac{X'}{\|X\|}Y = \|Y\| \cos(Y, X).$$

Multiply this scalar in the direction of the projection to get the new component in the direction X

$$\begin{aligned}
\widehat{Y} &= \left(\frac{X}{\|X\|} \right) \left(\frac{X'}{\|X\|} Y \right) \\
&= X \widehat{\beta} \\
&= \frac{XX'}{(X'X)} Y \\
&= P_{p \times p}^{(X)} Y_{p \times 1},
\end{aligned}$$

where we call P the projection matrix of the direction X .

For decomposition (not projection yet) on set of vectors constituting the columns of $\mathbf{X} = (X_1, \dots, X_n)$,

$$\begin{aligned}
&= X_1 \widehat{\beta}_1 + \dots + X_n \widehat{\beta}_n \\
&= \mathbf{X} \widehat{\beta}
\end{aligned}$$

We need to minimize the remaining error

$$\begin{aligned}
 e &= Y - \mathbf{X}\widehat{\beta}, \\
 \|e\|^2 &= \langle (Y - \mathbf{X}\widehat{\beta}), (Y - \mathbf{X}\widehat{\beta}) \rangle \\
 &= \langle Y, Y \rangle - 2\beta' \begin{pmatrix} \langle X_1, Y \rangle \\ \vdots \\ \langle X_n, Y \rangle \end{pmatrix} + \beta' \begin{pmatrix} \langle X_1, X_1 \rangle & \dots & \langle X_1, X_n \rangle \\ \vdots & \ddots & \vdots \\ \langle X_n, X_1 \rangle & \dots & \langle X_n, X_n \rangle \end{pmatrix} \beta \\
 &= Y'Y - 2\beta'\mathbf{X}'Y + \beta'\mathbf{X}'\mathbf{X}\beta
 \end{aligned}$$

$$\begin{aligned}
 \nabla \|e\|^2 &= -2\mathbf{X}'\mathbf{y} + 2\mathbf{X}'\mathbf{X}\beta \\
 &= -2\mathbf{X}'(\mathbf{y} - \mathbf{X}\beta),
 \end{aligned}$$

which means that the error is perpendicular to each component X_i ; i.e.,

$$\begin{aligned}
 \langle X_i, e \rangle &= 0, \text{ or} \\
 X_i' e &= 0
 \end{aligned}$$

The solution will be

$$\widehat{\beta} = (\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}'Y$$

Therefore, the projection \widehat{Y} is

$$\begin{aligned}
 \widehat{Y} &= \mathbf{X}\widehat{\beta} \\
 &= \mathbf{X}(\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}'Y
 \end{aligned}$$

and the projection matrix \mathbf{P} is

$$\mathbf{P} = \mathbf{X} (\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}'$$

The very interesting thing is

$$\begin{aligned}\mathbf{X}'\mathbf{X} &= \begin{pmatrix} X'_1 \\ \vdots \\ X'_n \end{pmatrix} \begin{pmatrix} X_1 & \dots & X_n \end{pmatrix} \\ &= \begin{pmatrix} X'_1X_1 & X'_1X_2 & \dots & X'_1X_n \\ X'_2X_1 & X'_2X_2 & \dots & X'_2X_n \\ \vdots & \vdots & \vdots & \vdots \\ X'_nX_1 & X'_nX_2 & \dots & X'_nX_n \end{pmatrix}\end{aligned}$$

if we choose the basis X_i orthogonal

$$\begin{aligned}\mathbf{X}'\mathbf{X} &= \begin{pmatrix} X'_1X_1 & & & \\ & X'_2X_2 & & \\ & & \ddots & \\ & & & X'_nX_n \end{pmatrix} \\ &= \text{diag} \left(\|X_1\|^2, \dots, \|X_n\|^2 \right),\end{aligned}$$

$$\begin{aligned}
\mathbf{P} &= \begin{pmatrix} X_1 & \dots & X_n \end{pmatrix} \begin{pmatrix} \frac{1}{\|X_1\|^2} & & \\ & \ddots & \\ & & \frac{1}{\|X_n\|^2} \end{pmatrix} \begin{pmatrix} X'_1 \\ \vdots \\ X'_n \end{pmatrix} \\
&= \frac{X_1 X'_1}{X'_1 X_1} + \dots + \frac{X_n X'_n}{X'_n X_n} \\
&= \frac{X_1}{\|X_1\|} \frac{X'_1}{\|X_1\|} + \dots + \frac{X_n}{\|X_n\|} \frac{X'_n}{\|X_n\|} \\
&= \mathbf{P}_1 + \dots + \mathbf{P}_n
\end{aligned}$$

The projection will be

$$\widehat{Y} = \mathbf{P}_1 Y + \dots \mathbf{P}_n Y$$

The orthogonality of the basis made it possible to project on each and sum up the projections. If the set of basis X_i span the whole space, then they are complete and $\widehat{Y} = Y$ and the error is zero; which means we can express Y in terms of the new subspace \mathbf{X} . And for **orthonormal** basis, where $\|X_i\| = 1$

$$\begin{aligned}
\mathbf{P}_i &= X_i X'_i, \\
\widehat{\beta}_i &= X'_i Y
\end{aligned}$$

Function Spaces: nothing new from vectors. A function is a vector with infinite dimension

instead of vector Y , we have the function $f(x)$, and for two basis functions $X_1(x), X_2(x)$, their inner product is defined by

$$\begin{aligned}\langle X_1, X_2 \rangle &= \int_{-\infty}^{\infty} X_1(x) X_2(x) dx, \\ \|X_i\|^2 &= \langle X_i, X_i \rangle\end{aligned}$$

Therefore,

$$\begin{aligned}\hat{f}(x) &= X_1 \hat{\beta}_1 + \dots + X_n \hat{\beta}_n \\ &= \mathbf{X} \hat{\beta} \\ &= \mathbf{X} (\mathbf{X}' \mathbf{X})^{-1} \mathbf{X}' f \\ e &= f - \hat{f}.\end{aligned}$$

for a set of orthogonal functions

$$\begin{aligned}\hat{\beta}_i &= \left\langle \frac{X_i}{\langle X_i, X_i \rangle}, f \right\rangle \\ &= \int X_i(x) f(x) dx, \\ \hat{f} &= \sum_i \hat{\beta}_i X_i(x)\end{aligned}$$

Fourier expansion

choose the set X_i to be the trigonometric functions in the interval $[0, 2\pi]$, and approximate f by \hat{f} as

$$\hat{f} = \hat{\beta}_0 + \sum_{n=1}^N \hat{\beta}_n^{(e)} \cos(nt) + \sum_{n=1}^N \hat{\beta}_n^{(o)} \sin(nt),$$

Therefore, the basis is

$$1, \sin t, \cos t, \sin 2t, \cos 2t, \dots, \sin Nt, \cos Nt.$$

it is easy to show that

$$\begin{aligned}\langle \cos (nt) , \cos (mt) \rangle &= \pi \delta_{nm}, \\ \langle \sin (nt) , \sin (mt) \rangle &= \pi \delta_{nm}, \\ \langle \sin (nt) , \cos (mt) \rangle &= 0\end{aligned}$$

This means that all basis are orthogonal, and

$$\begin{aligned}\langle 1, 1 \rangle &= 2\pi, \\ \langle X_i, X_i \rangle &= \pi, \quad \forall i \neq 1\end{aligned}$$

Obviously, from before, we have

$$\begin{aligned}\widehat{\beta}_0 &= \frac{\langle 1, f \rangle}{\langle 1, 1 \rangle} \\ &= \frac{1}{2\pi} \int_0^{2\pi} f, \\ \widehat{\beta}_n^{(e)} &= \frac{1}{\pi} \int_0^{2\pi} f(t) \cos (nt) dt, \\ \widehat{\beta}_n^{(o)} &= \frac{1}{\pi} \int_0^{2\pi} f(t) \sin (nt) dt\end{aligned}$$

The restriction $t \in [0, 2\pi]$ can be immediately removed by setting the basis as

$$\begin{aligned}\widehat{f} &= \widehat{\beta}_0 + \sum_{n=1}^N \widehat{\beta}_n^{(e)} \cos\left(\frac{2\pi}{T}nt\right) + \sum_{n=1}^N \widehat{\beta}_n^{(o)} \sin\left(\frac{2\pi}{T}nt\right), \\ \widehat{\beta}_0 &= \frac{1}{T} \int_T f, \\ \widehat{\beta}_n^{(e)} &= \frac{2}{T} \int_T f(t) \cos\left(\frac{2\pi}{T}nt\right) dt, \\ \widehat{\beta}_n^{(o)} &= \frac{2}{T} \int_T f(t) \sin\left(\frac{2\pi}{T}nt\right) dt.\end{aligned}$$

Example

$$f(t) = 1, \quad -\pi < t < \pi$$

Then

$$\begin{aligned}\widehat{\beta}_0 &= \frac{1}{2\pi}, \\ \widehat{\beta}_n^{(e)} &= 0, \\ \widehat{\beta}_n^{(o)} &= 0,\end{aligned}$$

of course; but

$$f(t) = \begin{cases} 1 & -\pi/2 < t < \pi/2 \\ 0 & \text{otherwise} \end{cases}$$

It is more convinient to employ an expansion in terms of the complex notation

$$e^{i2\pi nt/T} = \cos(2\pi nt/T) + i \sin(2\pi nt/T)$$

In that sense we have to define the inner product in terms of the comlex notation to be

$$\langle X_1, X_2 \rangle = \int X_1^* X_2.$$

We can see directly that the basis $e^{i2\pi nt/T}$ are orthogonal. For $n \neq m$

$$\begin{aligned} \int_0^T e^{i2\pi nt/T} e^{-i2\pi mt/T} dt &= \frac{1}{i2\pi(n-m)/T} e^{i2\pi(n-m)t/T} \Big|_0^{2\pi} \\ &= 0; \end{aligned}$$

and for $n = m$, clearly

$$\int_0^T e^{i2\pi(n-m)t/T} dt = T$$

More compactly

$$\begin{aligned} \langle e^{i2\pi nt/T}, e^{i2\pi mt/T} \rangle &= T\delta_{nm}, \\ \widehat{f} &= \sum_{n=-N}^{n=N} \widehat{\beta}_n e^{i2\pi nt/T}, \\ \widehat{\beta}_n &= \frac{\langle f, e^{i2\pi nt/T} \rangle}{\langle e^{i2\pi nt/T}, e^{i2\pi nt/T} \rangle} \\ &= \frac{1}{T} \int_T f(t) e^{-i2\pi nt/T} dt. \end{aligned}$$

Take care of two things: (1) the ugly $\frac{2}{T}$ went away; (2) the different sign of the exponents above.

It can be shown that the bases $e^{i2\pi nt/T}$ are complete for the piecewise continuous functions $f(t)$; i.e.,

$$f(t) = \sum_{n=-\infty}^{n=\infty} \widehat{\beta}_n e^{i2\pi nt/T}$$

Now, we can see Fourier series in two different ways: (1) a summation of sinusoidal components representing a finite support function $f(t)$, $0 < t < T$; (2) since $e^{i2\pi nt/T} = e^{i2\pi n(t+T)/T}$, we have

$$f(t) = f(t + T),$$

which is a representation of a periodic function of a period T . Said differently, the reconstruction $\sum_{n=-\infty}^{n=\infty} \widehat{\beta}_n e^{i2\pi nt/T}$ will produce our original function $f(t)$ repeated infinitely with a period T .

Delta Function

We wish to have the identity operator (for many mathematical and physical convenience shown later) to have an operator $\delta(t)$ such that at $t = 0$, we can pick up only the value $f(0)$, of the function f at its argument t .

Such a approach can be pursued by four routes:

1- Elementray and very non-rigorous

$$\delta(t) = \begin{cases} 0 & \text{if } t \neq 0 \\ \infty & \text{if } t = 0 \end{cases}, \text{ or}$$

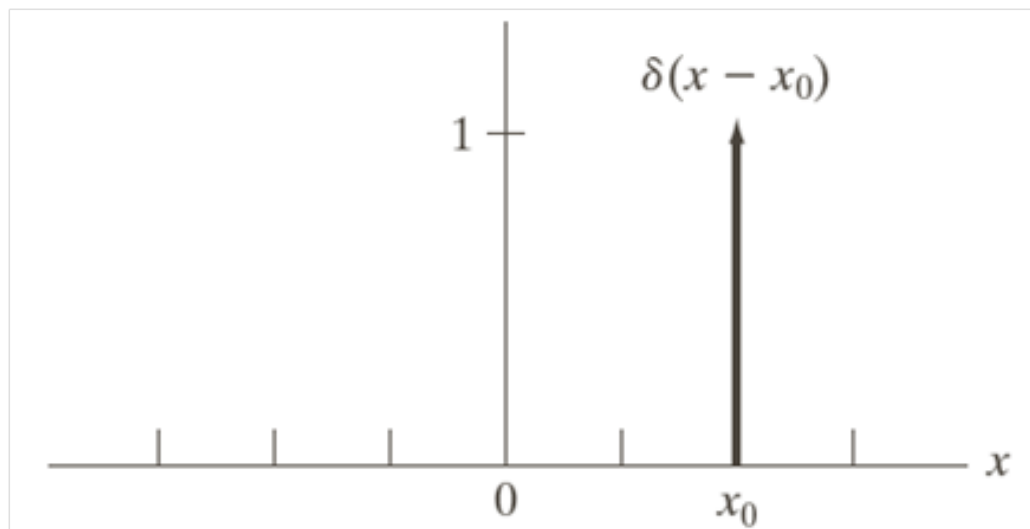
$$\delta(t - t_0) = \begin{cases} 0 & \text{if } t \neq t_0 \\ \infty & \text{if } t = t_0 \end{cases},$$

$$\int_{-\infty}^{\infty} \delta(t) dt = 1.$$

Therefore, for any smooth (continuous and differentiable) function, it is straightforward to show

$$\int_{-\infty}^{\infty} \delta(t) f(t) dt = f(0)$$

$$\int_{-\infty}^{\infty} \delta(t - t_0) f(t) dt = f(t_0)$$



Corollary

$$\sum_{x=-\infty}^{\infty} f(x) \delta(x - x_0) = f(x_0)$$

2- Rigorous approach for most practical purposes (limiting process):

$$\delta(t - t_0) = \lim_k \psi_k(t - t_0)$$

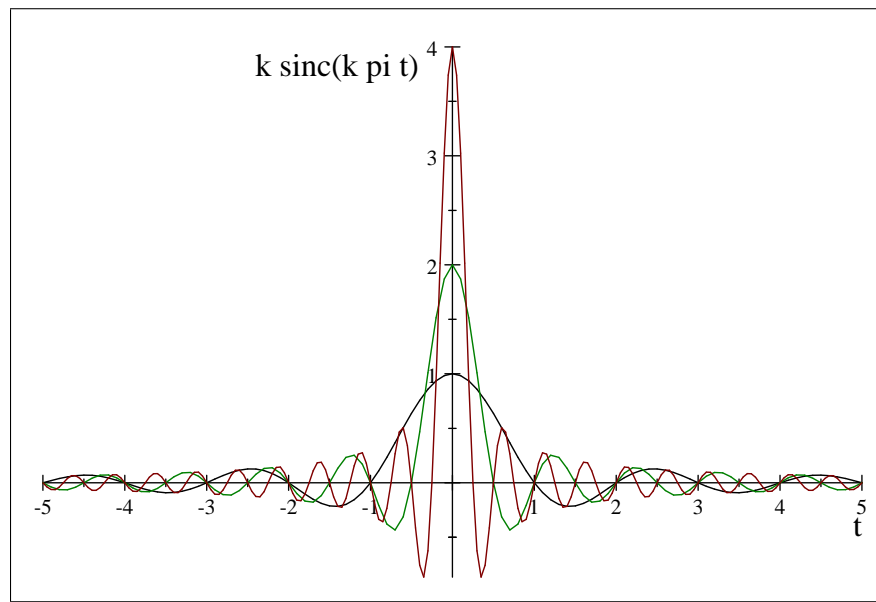
For example:

$$\delta(t) = \lim_{\epsilon \rightarrow 0} \frac{1}{\epsilon} \text{rect}\left(\frac{t}{\epsilon}\right),$$

$$\text{rect}(t) = \begin{cases} 1 & -1/2 \leq t \leq 1/2 \\ 0 & \text{otherwise} \end{cases}$$

Also,

$$\begin{aligned} \delta(t) &= \lim_{k \rightarrow \infty} k \text{sinc}(kt) \\ &= \lim_{k \rightarrow \infty} k \left[\frac{\sin(\pi kt)}{\pi kt} \right] \\ &= \lim_{k \rightarrow \infty} \frac{\sin(\pi kt)}{\pi t} \end{aligned}$$



The zero-crossing always occurs at

$$\begin{aligned}\pi k t &= m\pi, \\ t &= m/k\end{aligned}$$

The figure above is for $k = 1, 2, 4$.

It can be proven that for any of the delta representation above

$$\begin{aligned}\lim_{\epsilon \rightarrow 0} \int f(t) \frac{1}{\epsilon} \text{rect}\left(\frac{t}{\epsilon}\right) dt &= f(0), \\ \lim_{k \rightarrow \infty} \int f(t) \frac{\sin(\pi k t)}{\pi t} dt &= f(0).\end{aligned}$$

Therefore, both are valid representation of delta function.

It is also beneficial to notice that

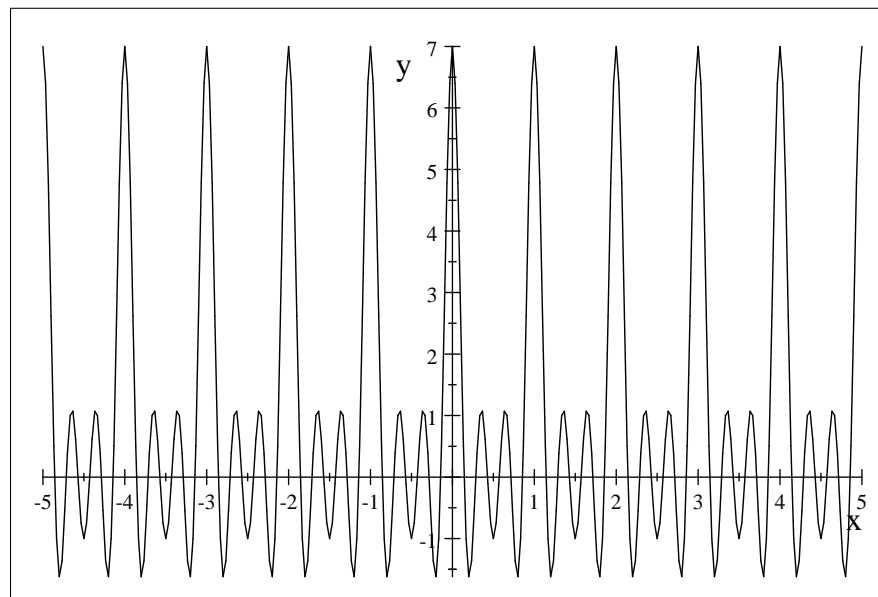
$$\begin{aligned}\int_{-k/2}^{k/2} e^{2\pi i \nu t} d\nu &= \frac{e^{2\pi i \nu t} \Big|_{-k/2}^{k/2}}{2\pi i t} \\ &= \frac{2i \sin(\pi k t)}{2\pi i t} \\ &= k \operatorname{sinc}(k t)\end{aligned}$$

Therefore,

$$\begin{aligned}\delta(t) &= \lim_{k \rightarrow \infty} \int_{-k/2}^{k/2} e^{2\pi i \nu t} d\nu \\ &= \int_{-\infty}^{\infty} e^{2\pi i \nu t} d\nu\end{aligned}$$

Another helpful form is

$$\operatorname{comb}(t) = \lim_{k \rightarrow \infty} \frac{\sin(\pi(2k+1)t)}{\sin(\pi t)}$$



It is a series of delta functions. It is easy to see that at small t , $\sin(\pi t) \approx \pi t$ and therefore $\text{comb}(t) \approx \lim_{k \rightarrow \infty} k \text{sinc}(kt)$, which is $\delta(t)$ for large k . Also, since $\text{comb}(t)$ is periodic with period 1, $\delta(t)$ is repeated.

Another helpful form is to notice that (it would be nice, but not mandatory, if you try to prove)

$$\sum_{n=-k}^{n=k} e^{2\pi i n t} = \frac{\sin(\pi(2k+1)t)}{\sin(\pi t)}$$

Therefore

$$\begin{aligned} \text{comb}(t) &= \sum_{n=-\infty}^{n=\infty} e^{2\pi i n t} \\ &= \sum_{n=-\infty}^{n=\infty} \delta(t - n) \end{aligned}$$

Of course,

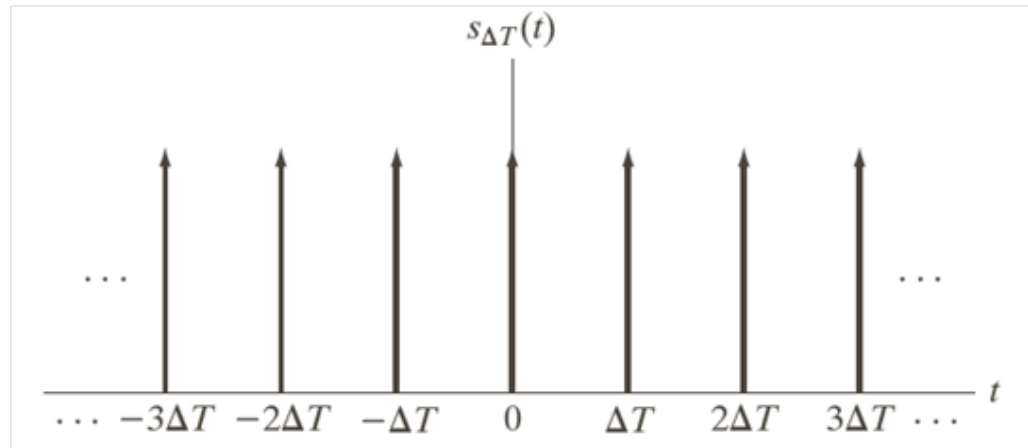
$$\begin{aligned} \text{comb}\left(\frac{t}{\Delta T}\right) &= \sum_{n=-\infty}^{n=\infty} e^{2\pi i n \frac{t}{\Delta T}} \\ &= \sum_{n=-\infty}^{n=\infty} \delta\left(\frac{t}{\Delta T} - n\right) \\ &= \Delta t \sum_{n=-\infty}^{n=\infty} \delta(t - n\Delta T). \end{aligned}$$

Observe that:

$$\begin{aligned}\int_{-\infty}^{\infty} \delta(at) dt &= \frac{1}{a} \int_{-\infty}^{\infty} \delta(at) d(at) \\ &= \frac{1}{a}\end{aligned}$$

Therefore

$$\delta(at) = \frac{1}{a} \delta(t)$$



3- Very Rigorous approach of mathematicians: (theory of distribution) which is out of scope.

Fourier Integraion (Fourier Transform)

There are many routes to pursue this topic; the simplest is what was pursued by Fourier himself, (and we will explain it with overlooking many of mathematical regularity conditions). But let's first review Riemann-sum integral. For a function $F(\xi)$, defined on $0 \leq \xi \leq L$, the area under this function can be approximated by

$$\begin{aligned} Area &\cong \sum_{n=0}^{N-1} F\left(n\frac{L}{N}\right) \left(\frac{L}{N}\right) \\ &= \sum_{n=0}^{N-1} F(n\Delta\xi) \Delta\xi \end{aligned}$$

and if the limit exists, this is the integration

$$\int_0^L F(\xi) d\xi = \lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} F(n\Delta\xi) \Delta\xi, \quad \Delta\xi = \frac{L}{N}$$

Now, consider our periodic function $f(t)$, whose period is T , and Fourier coefficients are F_n ; then

$$\begin{aligned} f(t) &= \sum_{n=-\infty}^{\infty} F_n e^{2\pi i n t / T} \\ &= \sum_{n=-\infty}^{\infty} \Delta\xi \frac{F_n}{\Delta\xi} e^{2\pi i t n \Delta\xi}, \quad \Delta\xi = \frac{1}{T} \\ &= \int_{-\infty}^{\infty} \left[\lim_{T \rightarrow \infty} \frac{F_n}{\Delta\xi} \right] e^{2\pi i t \xi} d\xi \end{aligned}$$

Now, if we allow $T \rightarrow \infty$, i.e., $\Delta\xi \rightarrow 0$,

$$\begin{aligned} F(\xi) &= \lim_{T \rightarrow \infty} \frac{F_n}{\Delta\xi} \\ &= \lim_{T \rightarrow \infty} \int_{-T/2}^{T/2} f(t) e^{-2\pi i n t / T} \\ &= \int_{-\infty}^{\infty} f(t) e^{-2\pi i t \xi} dt, \\ f(t) &= \int_{-\infty}^{\infty} F(\xi) e^{2\pi i t \xi} d\xi \end{aligned}$$

Convolution

$$\begin{aligned} [f * h](x) &= \int_{-\infty}^{\infty} f(x') h(x - x') dx' \\ &= \int_{-\infty}^{\infty} h(x') f(x - x') dx' \\ &= [h * f](x) \end{aligned}$$

Correlation

$$\begin{aligned} [f \star h](x) &= \int_{-\infty}^{\infty} f(x') h(x' - x) dx' \\ &= \int_{-\infty}^{\infty} h(x') f(x' + x) dx' \\ &\neq [h \star f](x) \end{aligned}$$

If the kernel h is symmetric, then convolution is the same as correlation (prove).

Meaning of Correlation. (Figure 12.9)

Fourier Transform of

Bibliography