

▷ **8: Correlation**

Cross-Correlation

Signal Matching

**Cross-corr as
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Normalized Cross-corr

Autocorrelation

**Autocorrelation
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The *cross-correlation* between two signals $u(t)$ and $v(t)$ is

$$\begin{aligned}w(t) = u(t) \otimes v(t) &\triangleq \int_{-\infty}^{\infty} u^*(\tau)v(\tau + t)d\tau \\ &= \int_{-\infty}^{\infty} u^*(\tau - t)v(\tau)d\tau \quad \text{[sub: } \tau \rightarrow \tau - t \text{]}\end{aligned}$$

The complex conjugate, $u^*(\tau)$ makes no difference if $u(t)$ is real-valued but makes the definition work even if $u(t)$ is complex-valued.

Correlation versus Convolution:

$$u(t) \otimes v(t) = \int_{-\infty}^{\infty} u^*(\tau)v(\tau + t)d\tau \quad \text{[correlation]}$$

$$u(t) * v(t) = \int_{-\infty}^{\infty} u(\tau)v(t - \tau)d\tau \quad \text{[convolution]}$$

Unlike convolution, the integration variable, τ , has the **same sign** in the arguments of $u(\dots)$ and $v(\dots)$ so the arguments have a **constant difference** instead of a constant sum (i.e. $v(t)$ is not time-flipped).

- Notes:
- (a) The argument of $w(t)$ is called the “lag” (= delay of u versus v).
 - (b) Some people write $u(t) \star v(t)$ instead of $u(t) \otimes v(t)$.
 - (c) Some swap u and v and/or negate t in the integral.

It is all rather inconsistent 😞.

Signal Matching

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Cross correlation is used to find where two signals match: $u(t)$ is the test waveform.

Example 1:

$v(t)$ contains $u(t)$ with an unknown delay and added noise.

$w(t) = u(t) \otimes v(t)$
 $= \int u^*(\tau - t)v(\tau)dt$ gives a peak at the time lag where $u(\tau - t)$ best matches $v(\tau)$; in this case at $t = 450$

Example 2:

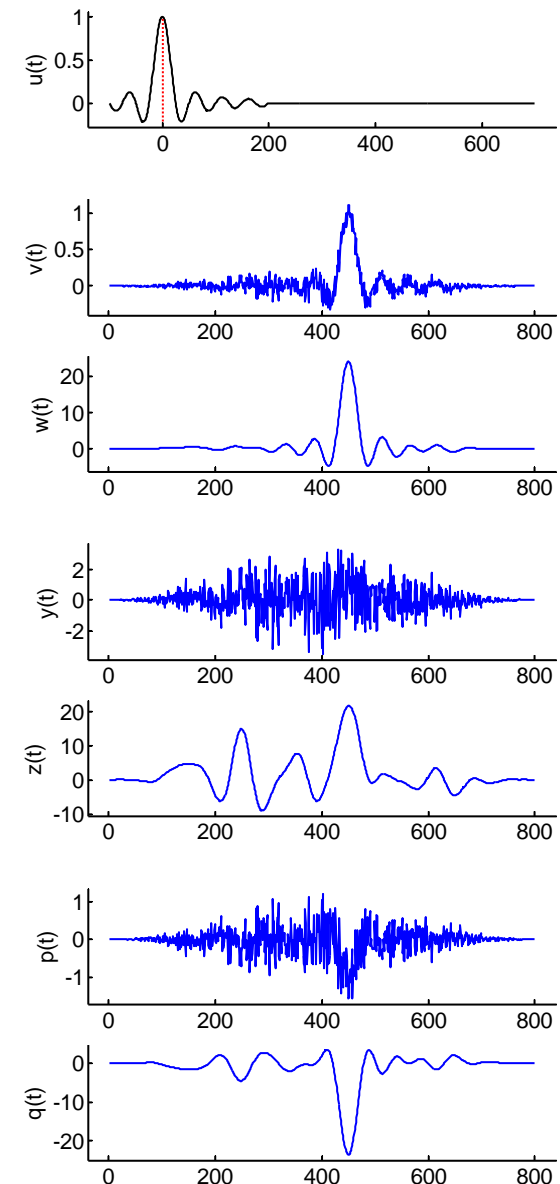
$y(t)$ is the same as $v(t)$ with more noise

$z(t) = u(t) \otimes y(t)$ can still detect the correct time delay (hard for humans)

Example 3:

$p(t)$ contains $-u(t)$ so that

$q(t) = u(t) \otimes p(t)$ has a negative peak



Cross-correlation as Convolution

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Correlation: $w(t) = u(t) \otimes v(t) = \int_{-\infty}^{\infty} u^*(\tau - t)v(\tau)d\tau$

If we define $x(t) = u^*(-t)$ then

$$\begin{aligned}x(t) * v(t) &\triangleq \int_{-\infty}^{\infty} x(t - \tau)v(\tau)d\tau = \int_{-\infty}^{\infty} u^*(\tau - t)v(\tau)d\tau \\ &= u(t) \otimes v(t)\end{aligned}$$

Fourier Transform of $x(t)$:

$$\begin{aligned}X(f) &= \int_{-\infty}^{\infty} x(t)e^{-i2\pi ft}dt = \int_{-\infty}^{\infty} u^*(-t)e^{-i2\pi ft}dt \\ &= \int_{-\infty}^{\infty} u^*(t)e^{i2\pi ft}dt = \left(\int_{-\infty}^{\infty} u(t)e^{-i2\pi ft}dt \right)^* \\ &= U^*(f)\end{aligned}$$

$$\text{So } w(t) = x(t) * v(t) \Rightarrow W(f) = X(f)V(f) = U^*(f)V(f)$$

Hence the **Cross-correlation theorem**:

$$\begin{aligned}w(t) = u(t) \otimes v(t) &\Leftrightarrow W(f) = U^*(f)V(f) \\ &= u^*(-t) * v(t)\end{aligned}$$

Note that, unlike convolution, **correlation is not associative or commutative**:

$$v(t) \otimes u(t) = v^*(-t) * u(t) = u(t) * v^*(-t) = w^*(-t)$$

Normalized Cross-correlation

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Correlation: $w(t) = u(t) \otimes v(t) = \int_{-\infty}^{\infty} u^*(\tau - t)v(\tau)d\tau$

If we define $y(t) = u(t - t_0)$ for some fixed t_0 , then $E_y = E_u$:

$$\begin{aligned} E_y &= \int_{-\infty}^{\infty} |y(t)|^2 dt = \int_{-\infty}^{\infty} |u(t - t_0)|^2 dt \\ &= \int_{-\infty}^{\infty} |u(\tau)|^2 d\tau = E_u \end{aligned} \quad [t \rightarrow \tau + t_0]$$

Cauchy-Schwarz inequality: $\left| \int_{-\infty}^{\infty} y^*(\tau)v(\tau)d\tau \right|^2 \leq E_y E_v$

$$\Rightarrow |w(t_0)|^2 = \left| \int_{-\infty}^{\infty} u^*(\tau - t_0)v(\tau)d\tau \right|^2 \leq E_y E_v = E_u E_v$$

but t_0 was arbitrary, so we must have $|w(t)| \leq \sqrt{E_u E_v}$ for all t

We can define the *normalized cross-correlation*

$$z(t) = \frac{u(t) \otimes v(t)}{\sqrt{E_u E_v}}$$

with properties: (1) $|z(t)| \leq 1$ for all t

(2) $|z(t_0)| = 1 \Leftrightarrow v(\tau) = \alpha u(\tau - t_0)$ with α constant

[Cauchy-Schwarz Inequality Proof]

You do not need to memorize this proof

We want to prove the Cauchy-Schwarz Inequality: $\left| \int_{-\infty}^{\infty} u^*(t)v(t)dt \right|^2 \leq E_u E_v$
where $E_u \triangleq \int_{-\infty}^{\infty} |u(t)|^2 dt$.

Suppose we define $w \triangleq \int_{-\infty}^{\infty} u^*(t)v(t)dt$. Then,

$$\begin{aligned} 0 &\leq \int |E_v u(t) - w^* v(t)|^2 dt && [|\dots|^2 \text{ always } \geq 0] \\ &= \int (E_v u^*(t) - w v^*(t)) (E_v u(t) - w^* v(t)) dt && [|z|^2 = z^* z] \\ &= E_v^2 \int u^*(t)u(t)dt + |w|^2 \int v^*(t)v(t)dt - w^* E_v \int u^*(t)v(t)dt - w E_v \int u(t)v^*(t)dt \\ &= E_v^2 \int |u(t)|^2 dt + |w|^2 \int |v(t)|^2 dt - E_v w^* w - E_v w w^* && [\text{definition of } w] \\ &= E_v^2 E_u + |w|^2 E_v - 2 |w|^2 E_v = E_v (E_u E_v - |w|^2) && [|z|^2 = z^* z] \end{aligned}$$

Unless $E_v = 0$ (in which case, $v(t) \equiv 0$ and the C-S inequality is true), we must have $|w|^2 \leq E_u E_v$ which proves the C-S inequality.

Also, $E_u E_v = |w|^2$ only if we have equality in the first line,

that is, $\int |E_v u(t) - w^* v(t)|^2 dt = 0$ which implies that the integrand is zero for all t .

This implies that $u(t) = \frac{w^*}{E_v} v(t)$.

So we have shown that $E_u E_v = |w|^2$ if and only if $u(t)$ and $v(t)$ are proportional to each other.

Autocorrelation

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The correlation of a signal with itself is its *autocorrelation*:

$$w(t) = u(t) \otimes u(t) = \int_{-\infty}^{\infty} u^*(\tau - t)u(\tau)d\tau$$

The autocorrelation at zero lag:

$$\begin{aligned}w(0) &= \int_{-\infty}^{\infty} u^*(\tau - 0)u(\tau)d\tau \\ &= \int_{-\infty}^{\infty} u^*(\tau)u(\tau)d\tau \\ &= \int_{-\infty}^{\infty} |u(\tau)|^2 d\tau = E_u\end{aligned}$$

The autocorrelation at zero lag, $w(0)$, is the energy of the signal.

The *normalized autocorrelation*: $z(t) = \frac{u(t) \otimes u(t)}{E_u}$
satisfies $z(0) = 1$ and $|z(t)| \leq 1$ for any t .

Wiener-Khinchin Theorem: [Cross-correlation theorem when $v(t) = u(t)$]

$$w(t) = u(t) \otimes u(t) \quad \Leftrightarrow \quad W(f) = U^*(f)U(f) = |U(f)|^2$$

The Fourier transform of the autocorrelation is the energy spectrum.

Autocorrelation example

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Cross-correlation is used to find when two different signals are similar.
Autocorrelation is used to find when a signal is similar to itself delayed.

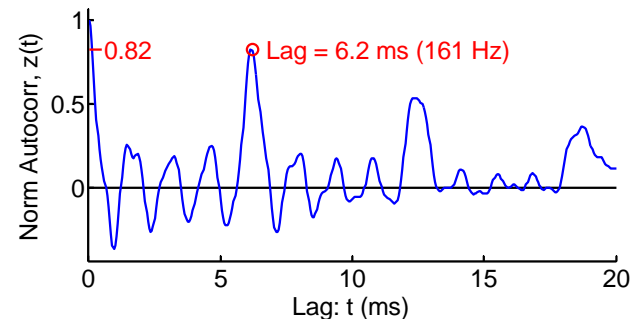
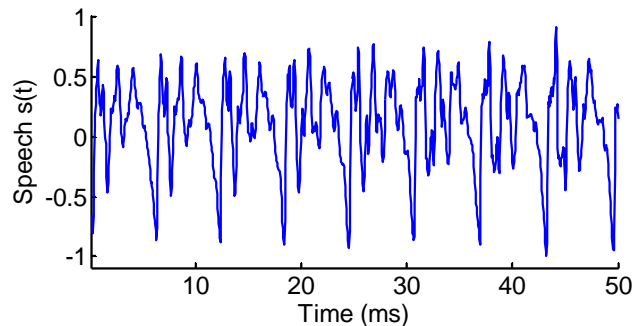
First graph shows $s(t)$ a segment of the microphone signal from the initial vowel of “early” spoken by me. The waveform is “quasi-periodic” = “almost periodic but not quite”.

Second graph shows normalized autocorrelation, $z(t) = \frac{s(t) \otimes s(t)}{E_s}$.

$z(0) = 1$ for $t = 0$ since a signal always matches itself exactly.

$z(t) = 0.82$ for $t = 6.2 \text{ ms}$ = one period lag (not an exact match).

$z(t) = 0.53$ for $t = 12.4 \text{ ms}$ = two periods lag (even worse match).



Fourier Transform Variants

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There are three different versions of the Fourier Transform in current use.

(1) Frequency version (we have used this in lectures)

$$U(f) = \int_{-\infty}^{\infty} u(t)e^{-i2\pi ft} dt \quad u(t) = \int_{-\infty}^{\infty} U(f)e^{i2\pi ft} df$$

- Used in the communications/broadcasting industry and textbooks.
- The formulae do not need scale factors of 2π anywhere. 😊😊😊

(2) Angular frequency version

$$\tilde{U}(\omega) = \int_{-\infty}^{\infty} u(t)e^{-i\omega t} dt \quad u(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tilde{U}(\omega)e^{i\omega t} d\omega$$

Continuous spectra are unchanged: $\tilde{U}(\omega) = U(f) = U(\frac{\omega}{2\pi})$

However **δ -function spectral components are multiplied by 2π** so that

$$U(f) = \delta(f - f_0) \quad \Rightarrow \quad \tilde{U}(\omega) = 2\pi \times \delta(\omega - 2\pi f_0)$$

- Used in most signal processing and control theory textbooks.

(3) Angular frequency + symmetrical scale factor

$$\hat{U}(\omega) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} u(t)e^{-i\omega t} dt \quad u(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \hat{U}(\omega)e^{i\omega t} d\omega$$

In all cases $\hat{U}(\omega) = \frac{1}{\sqrt{2\pi}} \tilde{U}(\omega)$

- Used in many Maths textbooks (mathematicians like symmetry)

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Fourier Transform using Angular Frequency:

$$\tilde{U}(\omega) = \int_{-\infty}^{\infty} u(t)e^{-i\omega t} dt \quad u(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tilde{U}(\omega)e^{i\omega t} d\omega$$

Any formula involving $\int df$ will change to $\frac{1}{2\pi} \int d\omega$ [since $d\omega = 2\pi df$]

Parseval's Theorem:

$$\int u^*(t)v(t)dt = \frac{1}{2\pi} \int \tilde{U}^*(\omega)\tilde{V}(\omega)d\omega$$

$$E_u = \int |u(t)|^2 dt = \frac{1}{2\pi} \int |\tilde{U}(\omega)|^2 d\omega$$

Waveform Multiplication: (convolution implicitly involves integration)

$$w(t) = u(t)v(t) \Rightarrow \tilde{W}(\omega) = \frac{1}{2\pi} \tilde{U}(\omega) * \tilde{V}(\omega)$$

Spectrum Multiplication: (multiplication \nRightarrow integration)

$$w(t) = u(t) * v(t) \Rightarrow \tilde{W}(\omega) = \tilde{U}(\omega)\tilde{V}(\omega)$$

To obtain formulae for version (3) of the Fourier Transform, $\hat{U}(\omega)$, substitute into the above formulae: $\tilde{U}(\omega) = \sqrt{2\pi}\hat{U}(\omega)$.

Summary

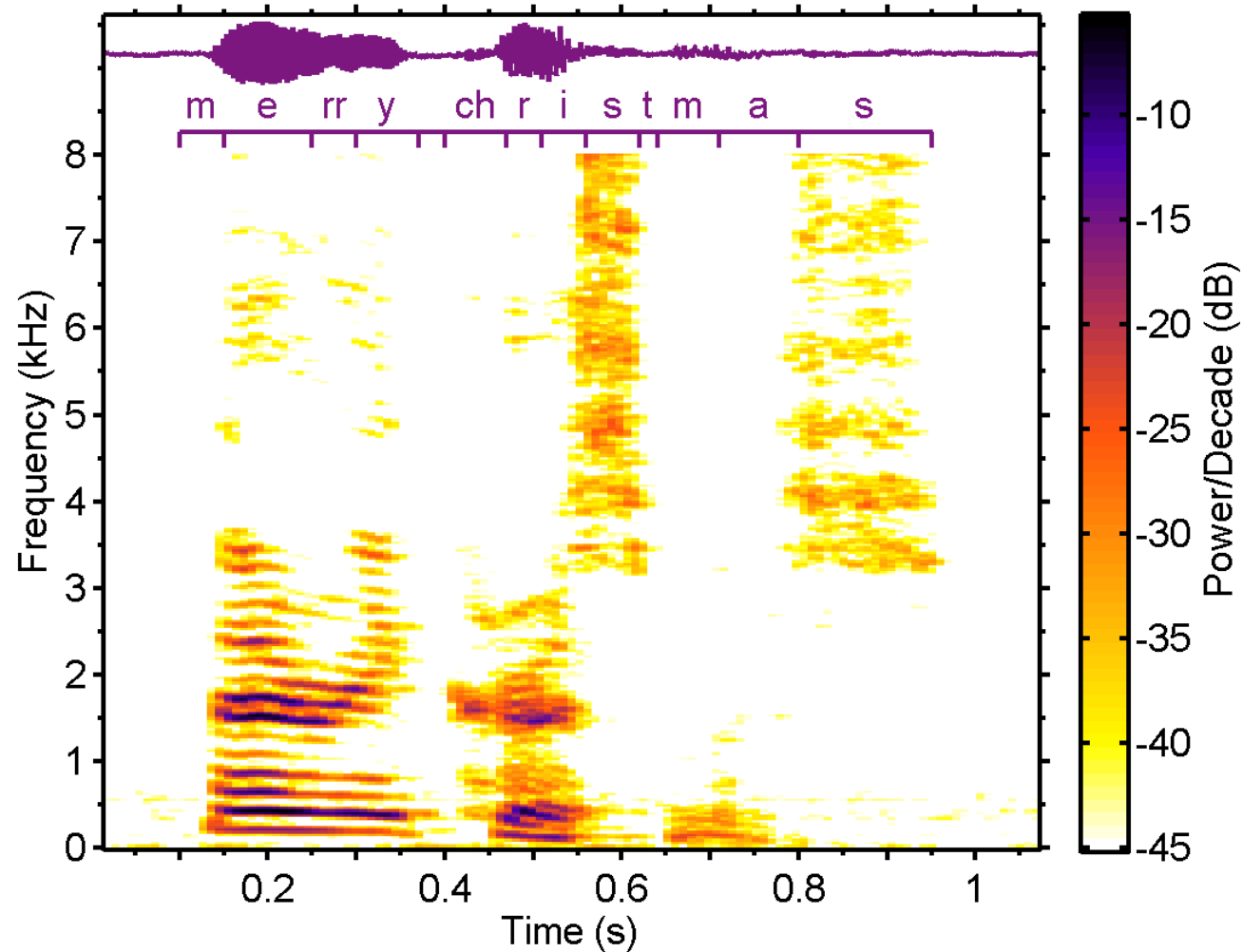
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- **Cross-Correlation:** $w(t) = u(t) \otimes v(t) = \int_{-\infty}^{\infty} u^*(\tau - t)v(\tau)d\tau$
 - **Used to find similarities** between $v(t)$ and a delayed $u(t)$
 - Cross-correlation theorem: $W(f) = U^*(f)V(f)$
 - Cauchy-Schwarz Inequality: $|u(t) \otimes v(t)| \leq \sqrt{E_u E_v}$
 - ▷ Normalized cross-correlation: $\left| \frac{u(t) \otimes v(t)}{\sqrt{E_u E_v}} \right| \leq 1$
- **Autocorrelation:** $x(t) = u(t) \otimes u(t) = \int_{-\infty}^{\infty} u^*(\tau - t)u(\tau)d\tau \leq E_u$
 - **Wiener-Khinchin:** $X(f) =$ energy spectral density, $|U(f)|^2$
 - **Used to find periodicity** in $u(t)$
- **Fourier Transform using ω :**
 - Continuous spectra unchanged; spectral impulses multiplied by 2π
 - In formulae: $\int df \rightarrow \frac{1}{2\pi} \int d\omega$; ω -convolution involves an integral

For further details see RHB Chapter 13.1

Spectrogram

Spectrogram of “Merry Christmas” spoken by Mike Brookes



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[Complex Fourier Series]

All waveforms have period $T = 1$. $\delta_{condition}$ is 1 whenever “condition” is true and otherwise 0.

Waveform	$x(t)$ for $ t < 0.5$	X_n
Square wave	$2\delta_{ t < 0.25} - 1$	$\frac{2 \sin 0.5\pi n}{\pi n} \times \delta_{n \neq 0}$
Pulse of width d	$\delta_{ t < 0.5d}$	$\frac{\sin \pi d n}{\pi n}$
Sawtooth wave	$2t$	$\frac{i(-1)^n}{\pi n} \times \delta_{n \neq 0}$
Triangle wave	$1 - 4 t $	$\frac{2(1 - (-1)^n)}{\pi^2 n^2}$

[Fourier Transform Properties A]

You need not memorize these properties. All integrals are $\int_{-\infty}^{\infty}$

Property	$x(t)$	$X(f)$
Forward	$x(t)$	$\int x(t)e^{-i2\pi ft} dt$
Inverse	$\int X(f)e^{i2\pi ft} df$	$X(f)$
Spectral Zero	$\int x(t)dt$	$= X(0)$
Temporal Zero	$x(0)$	$= \int X(f)df$
Duality	$X(t)$	$x(-f)$
Reversal	$x(-t)$	$X(-f)$
conjugate	$x^*(t)$	$X^*(-f)$
Temporal Derivative	$\frac{d^n}{dt^n} x(t)$	$(i2\pi f)^n X(f)$
Spectral Derivative	$(-i2\pi t)^n x(t)$	$\frac{d^n}{df^n} X(f)$
Integral	$\int_{-\infty}^t x(\tau)d\tau$	$\frac{1}{i2\pi f} X(f) + \frac{1}{2} X(0)\delta(f)$
Scaling	$x(\alpha t + \beta)$	$\frac{1}{ \alpha } e^{\frac{2i\pi f\beta}{\alpha}} X\left(\frac{f}{\alpha}\right)$
Time Shift	$x(t - T)$	$X(f)e^{-i2\pi fT}$
Frequency Shift	$x(t)e^{i2\pi Ft}$	$X(f - F)$

[Fourier Transform Properties B]

You need not memorize these properties. All integrals are $\int_{-\infty}^{\infty}$

Property	$x(t)$	$X(f)$
Linearity	$\alpha x(t) + \beta y(t)$	$\alpha X(f) + \beta Y(f)$
Multiplication	$x(t)y(t)$	$X(f) * Y(f)$
Convolution	$x(t) * y(t)$	$X(f)Y(f)$
Correlation	$x(t) \otimes y(t)$	$X^*(f)Y(f)$
Autocorrelation	$x(t) \otimes x(t)$	$ X(f) ^2$
Parseval or Plancherel	$\int x^*(t)y(t)dt$ $E_x = \int x(t) ^2 dt$	$= \int X^*(f)Y(f)df$ $= \int X(f) ^2 df$
Repetition	$\sum_n x(t - nT)$	$ \frac{1}{T} \sum_k X\left(\frac{k}{T}\right) \delta\left(f - \frac{k}{T}\right)$
Sampling	$\sum_n x(nT)\delta(t - nT)$	$ \frac{1}{T} \sum_k X\left(f - \frac{k}{T}\right)$
Modulation	$x(t) \cos(2\pi Ft)$	$\frac{1}{2}X(f - F) + \frac{1}{2}X(f + F)$

Convolution: $x(t) * y(t) = \int x(\tau)y(t - \tau)d\tau$

Cross-correlation: $x(t) \otimes y(t) = \int x^*(\tau)y(\tau + t)d\tau = \int x^*(\tau - t)y(\tau)d\tau$

[Fourier Transform Pairs]

You need not memorize these pairs.

$x(t)$	$X(f)$	$x(t)$	$X(f)$
$\delta(t)$	1	1	$\delta(f)$
$\text{rect}(t)$	$\frac{\sin(\pi f)}{\pi f}$	$\frac{\sin(t)}{t}$	$\pi \text{rect}(\pi f)$
$\text{tri}(t)$	$\frac{\sin^2(\pi f)}{\pi^2 f^2}$	$\frac{\sin^2(t)}{t^2}$	$\pi \text{tri}(\pi f)$
$\cos(2\pi\alpha t)$	$\frac{1}{2}\delta(f + \alpha) + \frac{1}{2}\delta(f - \alpha)$	$\sin(2\pi\alpha t)$	$\frac{i}{2}\delta(f + \alpha) - \frac{i}{2}\delta(f - \alpha)$
$e^{-\alpha t}u(t)$	$\frac{1}{\alpha + 2\pi i f}$	$te^{-\alpha t}u(t)$	$\frac{1}{(\alpha + 2\pi i f)^2}$
$e^{-\alpha t }$	$\frac{2\alpha}{\alpha^2 + 4\pi^2 f^2}$	$e^{-\pi t^2}$	$e^{-\pi f^2}$
$\text{sgn}(t)$	$\frac{1}{i\pi f}$	$u(t)$	$\frac{1}{2}\delta(f) + \frac{1}{2\pi i f}$
$\sum_{n=-\infty}^{\infty} \delta(t - nT)$	$ \frac{1}{T} \sum_{k=-\infty}^{\infty} \delta(f - \frac{k}{T})$		

Elementary Functions:

$$\text{rect}(t) = \begin{cases} 1, & |t| < 0.5 \\ 0, & \text{elsewhere} \end{cases}$$

$$\text{tri}(t) = \begin{cases} 1 - |t|, & |t| < 1 \\ 0, & \text{elsewhere} \end{cases}$$

$$\text{sgn}(t) = \begin{cases} -1, & t < 0 \\ 0, & t = 0 \\ 1, & t > 0 \end{cases}$$

$$u(t) = \frac{1}{2}(1 + \text{sgn}(t)) = \begin{cases} 0, & x < 0 \\ 0.5, & x = 0 \\ 1, & x > 0 \end{cases}$$