

Matched Filter (3B)

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Gaussian Random Process

Thermal Noise

zero-mean white Gaussian random process

$n(t)$ random function

the value at time t is characterized by
Gaussian probability density function

$$p(n) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{1}{2}\left(\frac{n}{\sigma}\right)^2\right]$$

σ^2 variance of n

$\sigma = 1$ normalized (standardized)
Gaussian function

$$\rightarrow z(t) = a + n(t)$$

$$\rightarrow p(z) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{1}{2}\left(\frac{z-a}{\sigma}\right)^2\right]$$

Central Limit Theorem

sum of statistically independent random variables
approaches Gaussian distribution
regardless of individual distribution functions

White Gaussian Noise (1)

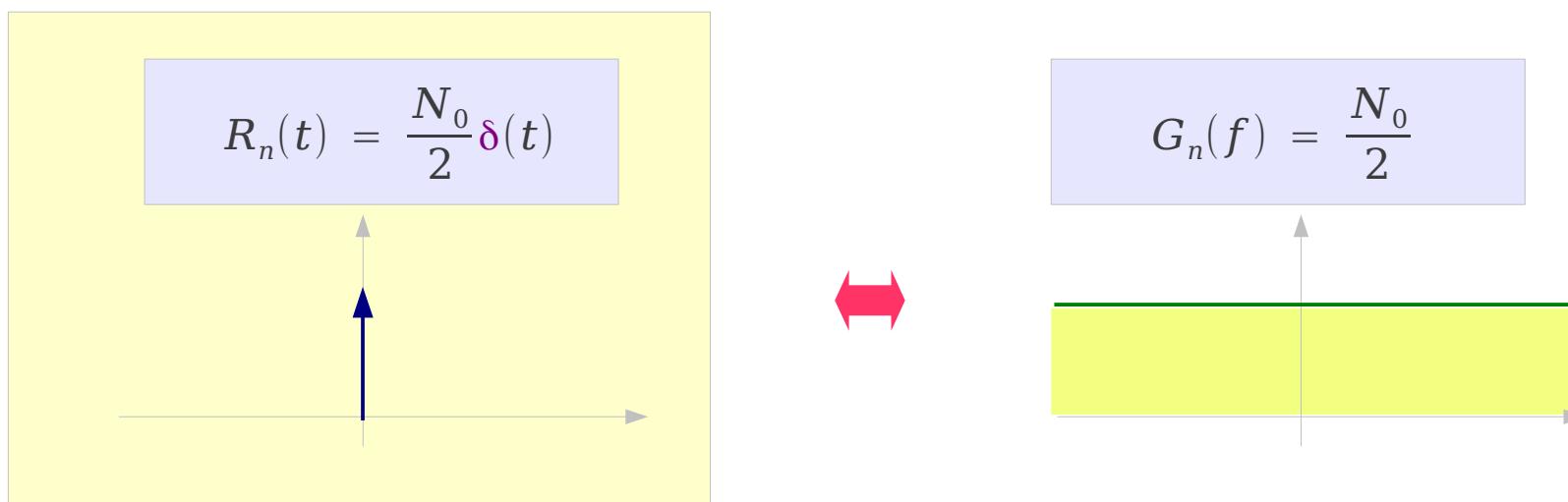
Thermal Noise

power spectral density is the same for all frequencies

$$G_n(f) = \frac{N_0}{2} \text{ watts / hertz}$$

equal amount of noise power
per unit bandwidth

uniform spectral density \rightarrow White Noise



$\delta(t)$ totally uncorrelated, noise samples are independent
memoryless channel

White Gaussian Noise (2)

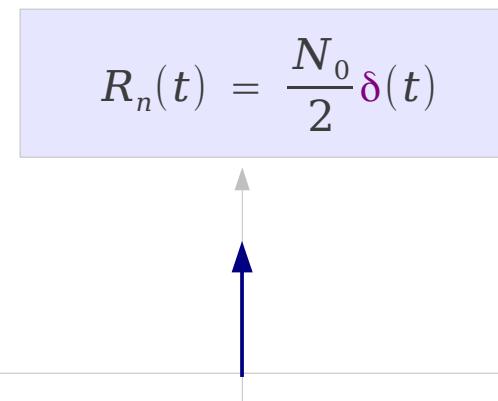
Thermal Noise

power spectral density is the same for all frequencies

$$G_n(f) = \frac{N_0}{2} \text{ watts / hertz}$$

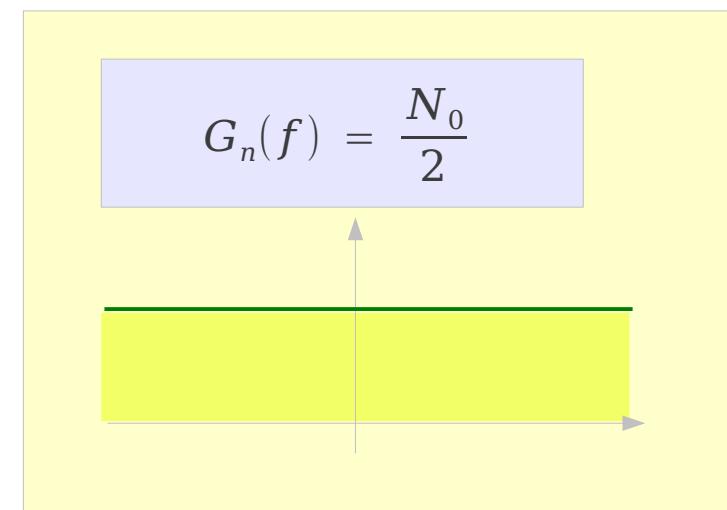
equal amount of noise power
per unit bandwidth

uniform spectral density \rightarrow White Noise



average power

$$P_x^T = \frac{1}{T} \int_{-T/2}^{+T/2} x^2(t) dt = \int_{-\infty}^{+\infty} G_x(f) df$$

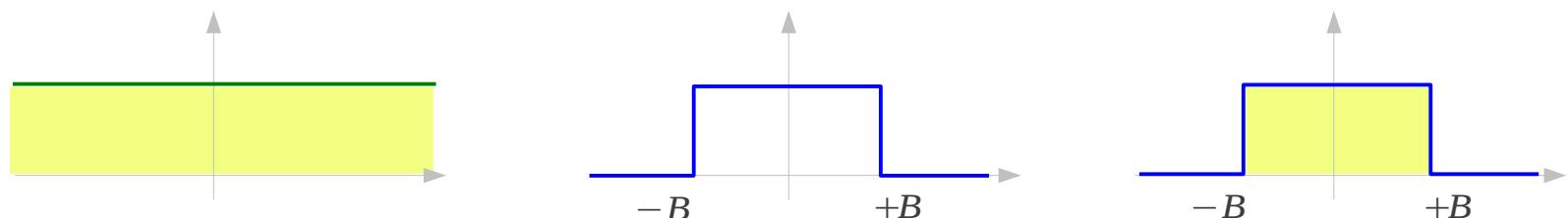
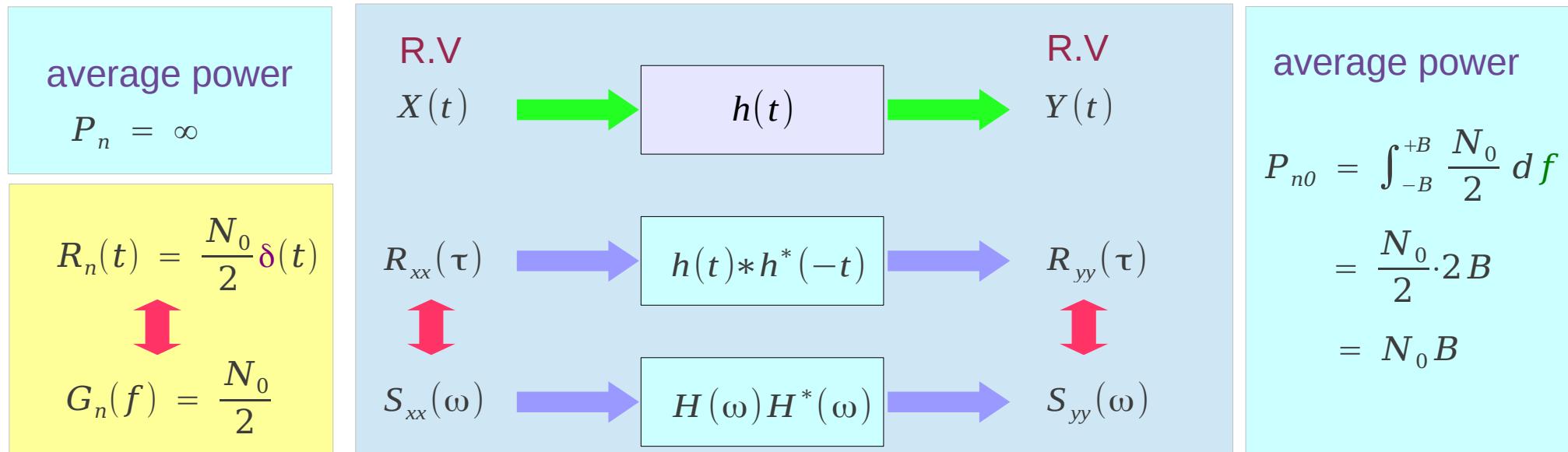


$$P_n = \int_{-\infty}^{+\infty} \frac{N_0}{2} df = \infty$$

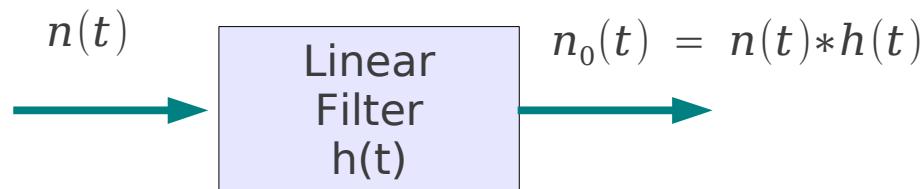
White Gaussian Noise (3)

Additive White Gaussian Noise (AWGN)

additive and no multiplicative mechanism



White Gaussian Noise (4)



$$G_n(f) = \frac{N_0}{2} \quad G_{n0}(f) = G_n(f) |H(f)|^2 = \begin{cases} \frac{N_0}{2} |H(f)|^2 & \text{for } |f| < f_u \\ 0 & \text{otherwise} \end{cases}$$

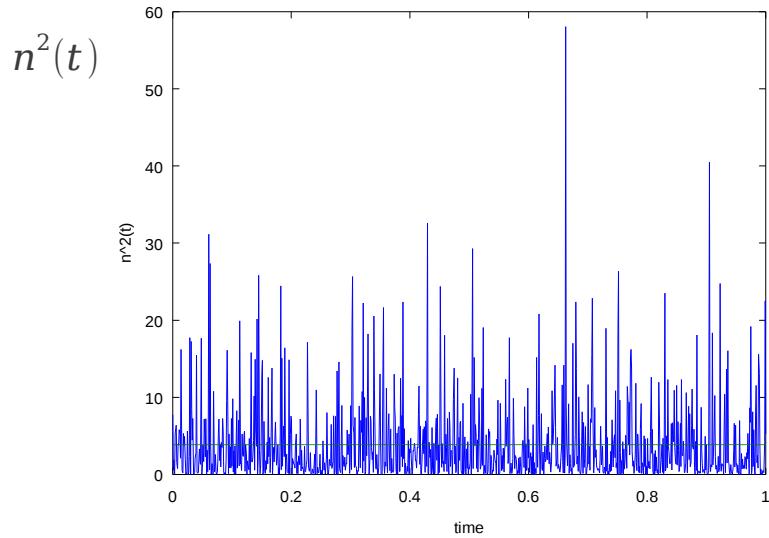
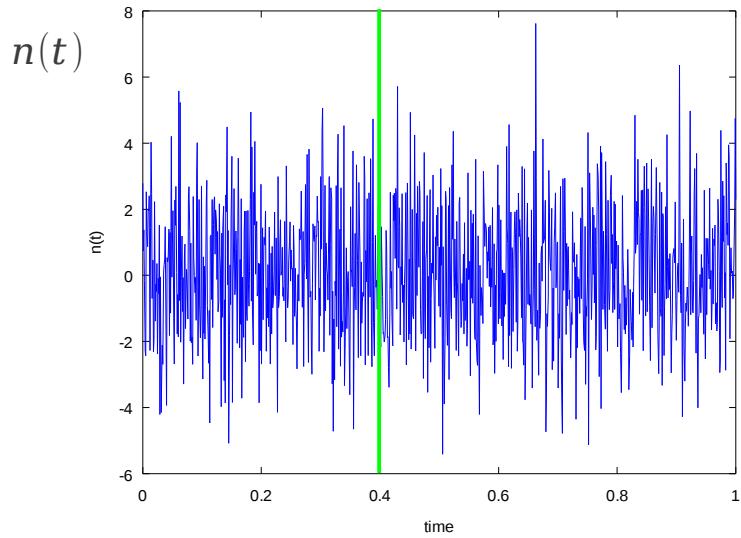
Average output noise power

$$\sigma_0^2 = \overline{n_0^2(t)} = \frac{N_0}{2} \int_{-\infty}^{+\infty} |H(f)|^2 d f$$

RMS

$$\sigma_0 = \sqrt{\overline{n_0^2(t)}} = \sqrt{\frac{1}{T} \int_{-T/2}^{+T/2} n_0^2(t) dt}$$

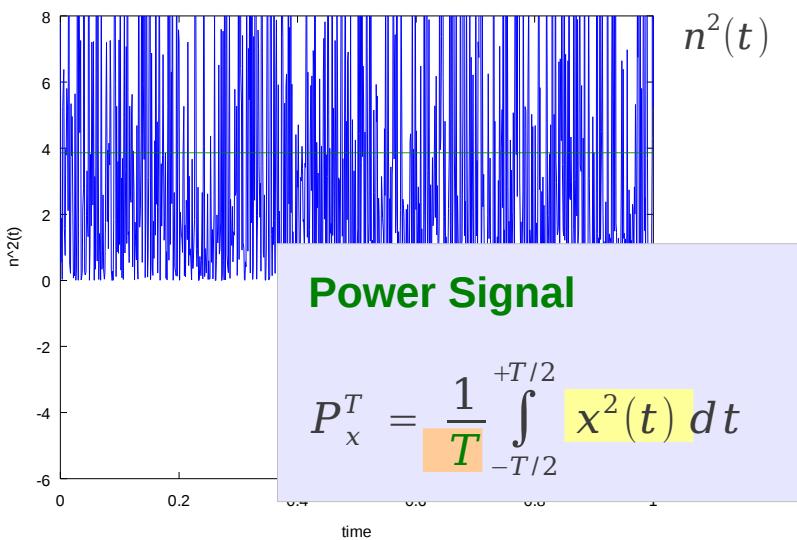
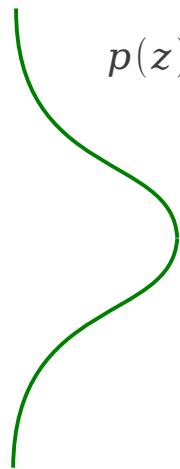
Gaussian Random Process



$$p(z) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{1}{2}\left(\frac{z}{\sigma}\right)^2\right]$$

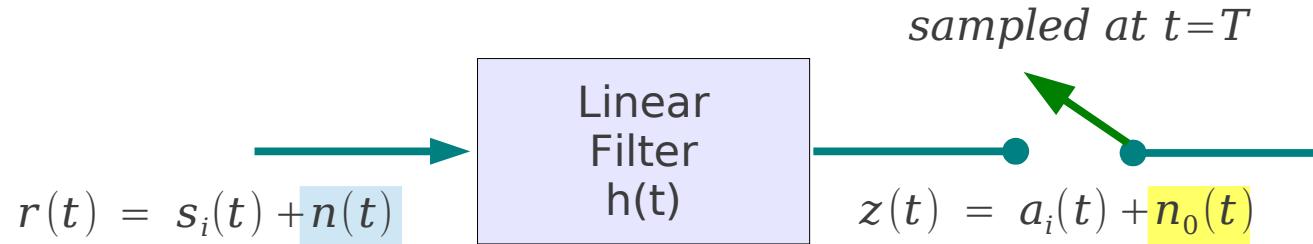
$$m = 0$$

$$\sigma^2 \neq 0$$



Matched Filter (1)

to find a filter $h(t)$ that gives **max** signal-to-noise ratio

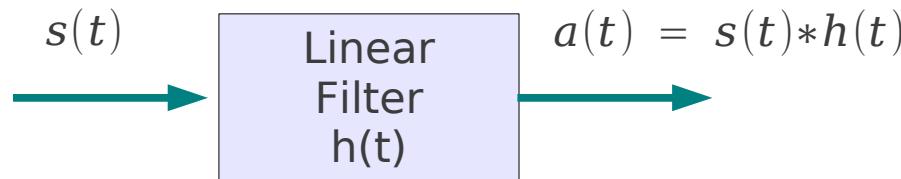


variance of $n_0(t)$ $\rightarrow \sigma_0^2$ avg noise power

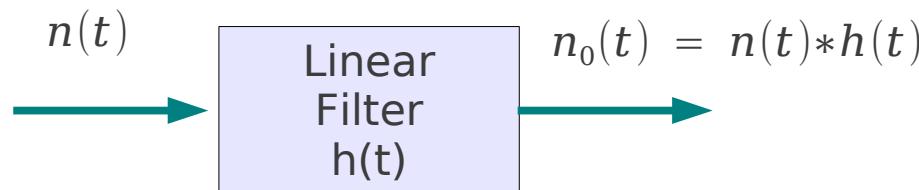
$$\frac{a_i^2(T)}{\overline{n_0^2(t)}} \xrightarrow{\text{instantaneous signal power}} \frac{\text{instantaneous signal power}}{\text{average noise power}} \rightarrow \left(\frac{S}{N} \right)_T = \frac{a_i^2}{\sigma_0^2}$$

assume $H_0(f)$ a filter transfer function that maximizes $\left(\frac{S}{N} \right)_T$

Matched Filter (2)



$$S(f) \quad A(f) = S(f)H(f) \quad \leftrightarrow \quad a(t) = \int_{-\infty}^{+\infty} S(f)H(f)e^{j2\pi f t} df$$



$$G_n(f) = \frac{N_0}{2} \quad G_{n0}(f) = G_n(f) |H(f)|^2 = \begin{cases} \frac{N_0}{2} |H(f)|^2 & \text{for } |f| < f_u \\ 0 & \text{otherwise} \end{cases}$$

Average output noise power $\sigma_0 = \frac{N_0}{2} \int_{-\infty}^{+\infty} |H(f)|^2 df$

Matched Filter (3)

instantaneous signal power

$$a_i^2$$



$$a(t) = \int_{-\infty}^{+\infty} S(f)H(f)e^{j2\pi f t} df$$

average output noise power

$$\sigma_0 = \frac{N_0}{2} \int_{-\infty}^{+\infty} |H(f)|^2 df$$

$$\left(\frac{S}{N}\right)_T = \frac{a_i^2}{\sigma_0^2} = \frac{\left| \int_{-\infty}^{+\infty} H(f)S(f)e^{+j2\pi f T} df \right|^2}{N_0/2 \int_{-\infty}^{+\infty} |H(f)|^2 df}$$

Does not depend on the particular shape of the waveform

Cauchy Schwarz's Inequality

$$\left| \int_{-\infty}^{+\infty} f_1(x)f_2(x) dx \right|^2 \leq \int_{-\infty}^{+\infty} |f_1(x)|^2 dx \int_{-\infty}^{+\infty} |f_2(x)|^2 dx \quad '=\text{'} \text{ holds when } f_1(x) = kf_2^*(x)$$

$$\left| \int_{-\infty}^{+\infty} H(f)S(f)e^{+j2\pi f t} df \right|^2 \leq \int_{-\infty}^{+\infty} |H(f)|^2 df \int_{-\infty}^{+\infty} |S(f)e^{+j2\pi f T}|^2 df \quad |e^{+j2\pi f T}| = 1$$

$$\left(\frac{S}{N}\right)_T = \frac{a_i^2}{\sigma_0^2} = \frac{\left| \int_{-\infty}^{+\infty} H(f)S(f)e^{+j2\pi f T} df \right|^2}{N_0/2 \int_{-\infty}^{+\infty} |H(f)|^2 df} \leq \frac{\left| \int_{-\infty}^{+\infty} |H(f)|^2 df \right| \left| \int_{-\infty}^{+\infty} |S(f)e^{+j2\pi f T}|^2 df \right|}{N_0/2 \left| \int_{-\infty}^{+\infty} |H(f)|^2 df \right|} = \frac{2}{N_0} \int_{-\infty}^{+\infty} |S(f)|^2 df$$

Matched Filter (4)

Two-sided power spectral density of input noise



$$\frac{N_0}{2}$$

Average noise power

$$\sigma_0 = \frac{N_0}{2} \int_{-\infty}^{+\infty} |H(f)|^2 df$$

$$\left(\frac{S}{N}\right)_T = \frac{a_i^2}{\sigma_0^2} = \frac{\left| \int_{-\infty}^{+\infty} H(f) S(f) e^{+j2\pi f T} df \right|^2}{N_0/2 \int_{-\infty}^{+\infty} |H(f)|^2 df}$$

Cauchy Schwarz's Inequality

$$\left(\frac{S}{N}\right)_T \leq \frac{2}{N_0} \int_{-\infty}^{+\infty} |S(f)|^2 df$$

$$\max \left(\frac{S}{N}\right)_T = \frac{2}{N_0} \int_{-\infty}^{+\infty} |S(f)|^2 df = \frac{2E}{N_0}$$

input signal energy
power spectral density
of input noise

does not depend on the particular shape of the waveform

Matched Filter (5)

$$\left| \int_{-\infty}^{+\infty} H(f) S(f) e^{+j2\pi f t} dx \right|^2 df \leq \int_{-\infty}^{+\infty} |H(f)|^2 df \int_{-\infty}^{+\infty} |S(f) e^{+j2\pi f T}|^2 df$$
$$\left(\frac{S}{N} \right)_T \leq \frac{2}{N_0} \int_{-\infty}^{+\infty} |S(f)|^2 df$$


$$\max \left(\frac{S}{N} \right)_T = \frac{2}{N_0} \int_{-\infty}^{+\infty} |S(f)|^2 df = \frac{2E}{N_0}$$

when complex conjugate relationship exists

$$H(f) = H_0(f) = k S^*(f) e^{-j2\pi f T}$$



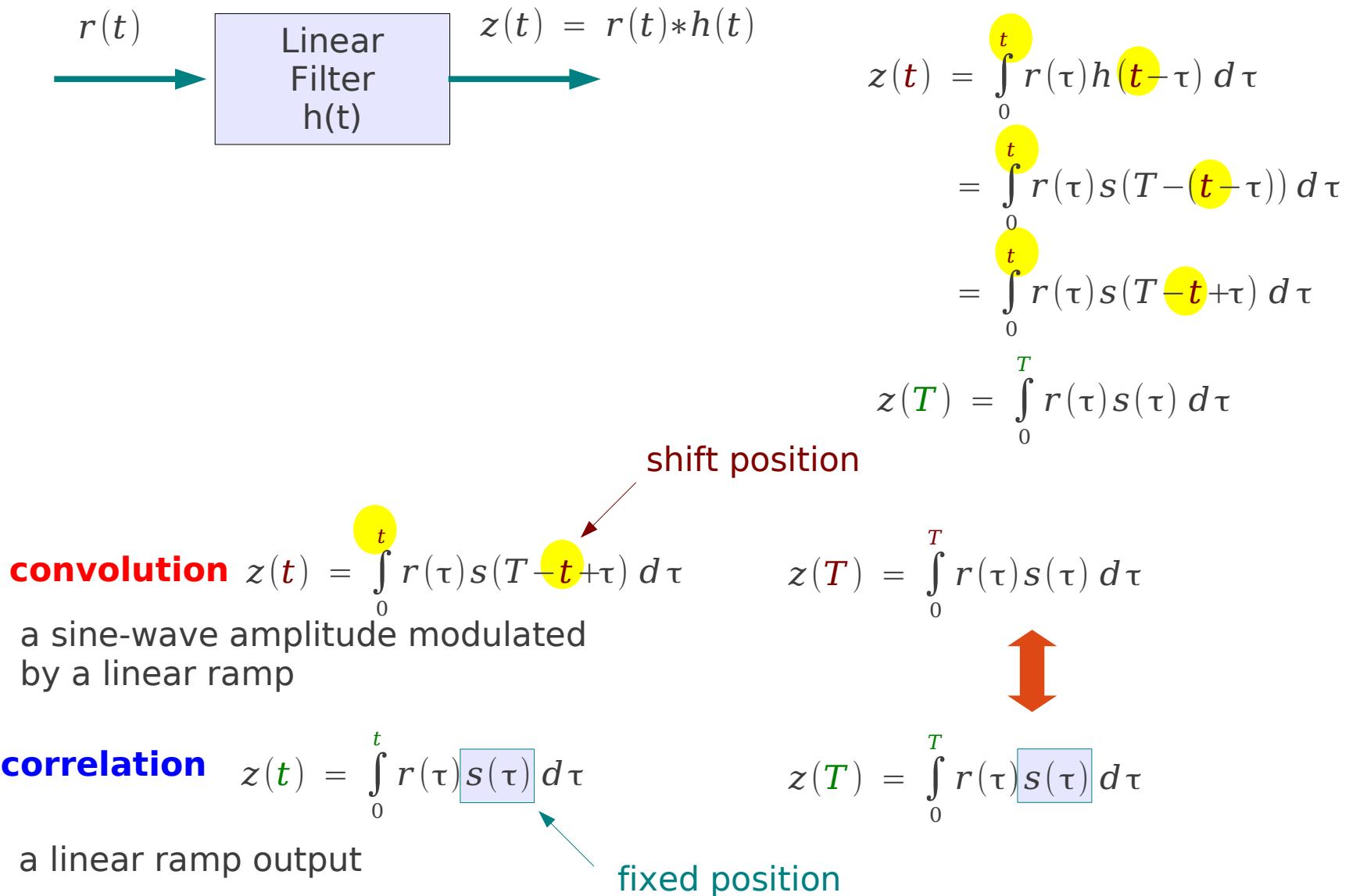
$$h(t) = h_0(t) = \begin{cases} ks(T-t) & 0 \leq t \leq T \\ 0 & elsewhere \end{cases}$$

$H_0(f)$ a filter transfer function that maximizes

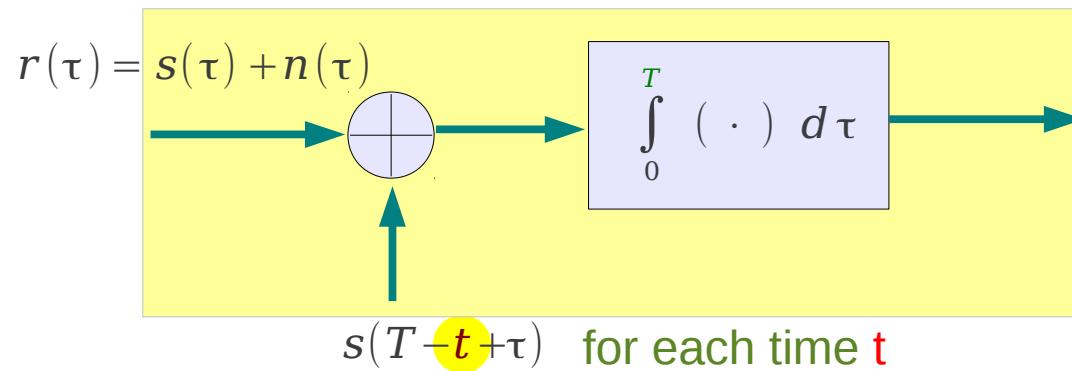
$$\left(\frac{S}{N} \right)_T$$

impulse response : delayed version of
the mirror image of the signal waveform

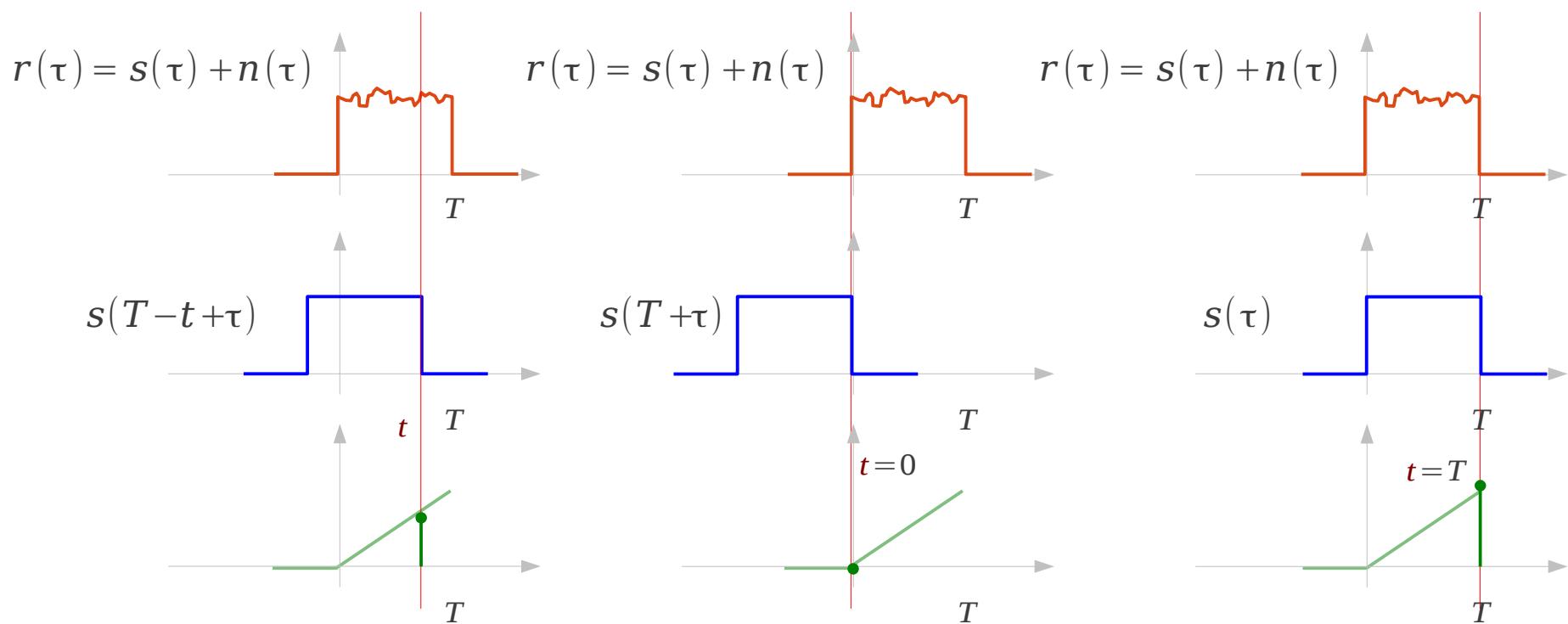
Convolution vs. Correlation Realization



Convolution Realization



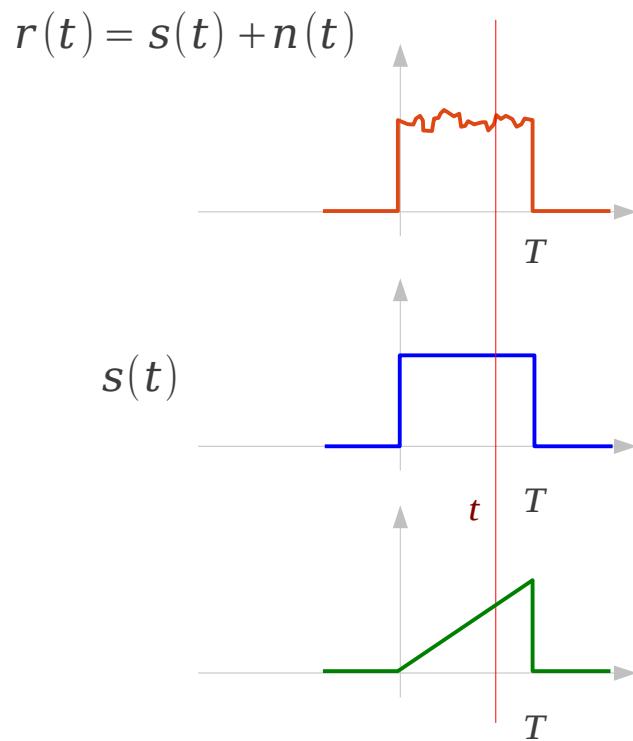
$$\begin{aligned} z(t) &= \int_0^t r(\tau)h(t-\tau) d\tau \\ &= \int_0^t r(\tau)s(T-(t-\tau)) d\tau \\ &= \int_0^t r(\tau)s(T-t+\tau) d\tau \end{aligned}$$



Correlation Realization (1)

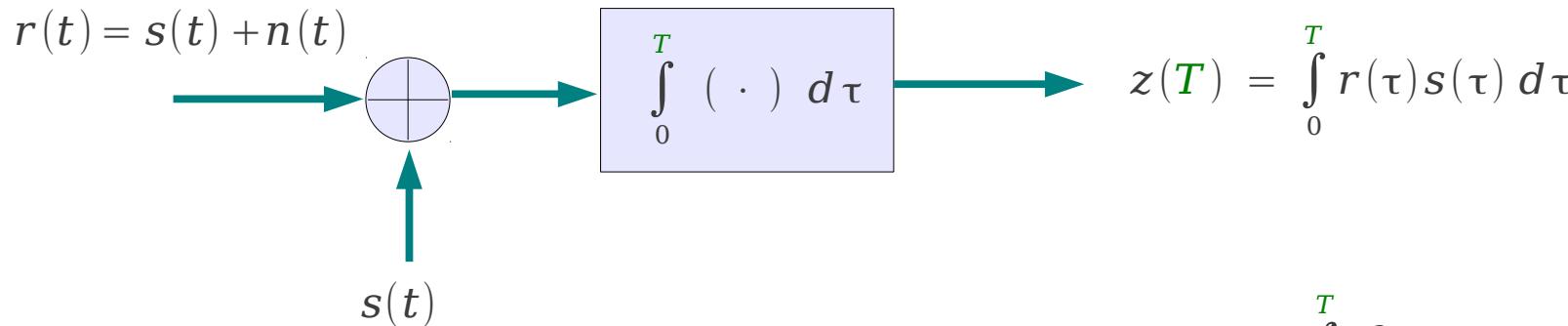
$$r(t) = s(t) + n(t)$$

A block diagram showing the realization of correlation. An input signal $r(t) = s(t) + n(t)$ enters a circular summing junction. A feedback signal $s(t)$ is fed into the same summing junction from below. The output of the summing junction goes to a rectangular integrator block. The integrator block contains the expression $\int_0^T (\cdot) dt$. The output of the integrator is $z(T) = \int_0^T r(t)s(t) dt$.



$$z(t) = \int_0^t r(\tau)s(\tau) d\tau$$

Correlation Realization (2)

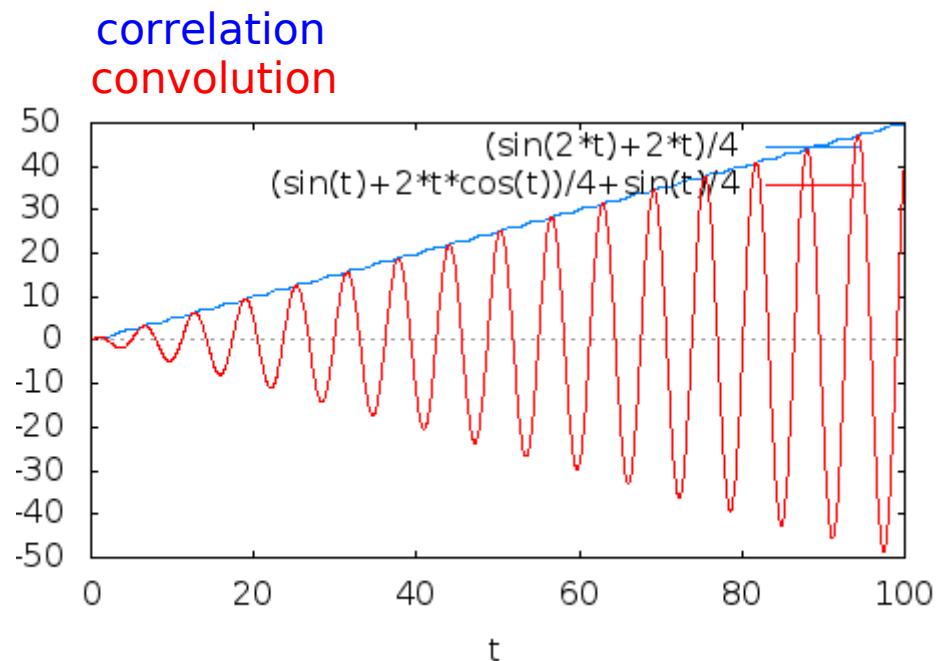
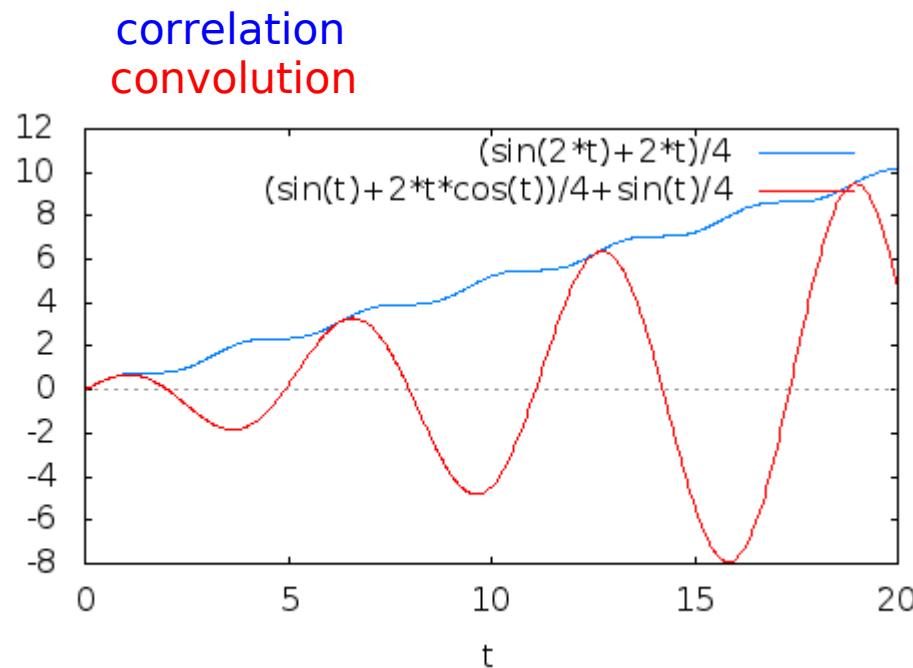


$$r(t) = s(t) \quad z(T) = \int_0^T s^2(\tau) d\tau = E$$

$$\begin{aligned} \sigma_0^2 &= E[n_o(t)] = E[\int_0^T n(t)s(t) dt] \int_0^T n(\tau)s(\tau) d\tau \\ &= E[\iint_0^T n(t)n(\tau) s(t)s(\tau) dt d\tau] \\ &= \iint_0^T E[n(t)n(\tau)] s(t)s(\tau) dt d\tau \\ &= \iint_0^T \frac{N_0}{2} \delta(t - \tau) s(t)s(\tau) dt d\tau \\ &= \frac{N_0}{2} \int_0^T s^2(t) dt = \frac{N_0}{2} E \end{aligned}$$

$\frac{a_i^2(T)}{n_0^2(t)}$	$\left(\frac{S}{N}\right)_T = \frac{a_i^2}{\sigma_0^2}$
$\max \left(\frac{S}{N}\right)_T = \frac{2}{N_0} \int_{-\infty}^{+\infty} S(f) ^2 df = \frac{2E}{N_0}$	

Correlation and Convolution Examples (1)



$z : \text{integrate}(\cos(x)*\cos(2*\%pi - t + x), x, 0, t);$ convolution

$(\sin(t) + 2*t*\cos(t))/4 + \sin(t)/4$

correlation

$z : \text{integrate}(\cos(x)*\cos(x), x, 0, t);$

$(\sin(2*t) + 2*t)/4$

Correlation and Convolution Examples (2)

$$\begin{aligned} s(t) &= A \cos(\omega_0 t) & 0 \leq t < T \\ &= 0 & \text{elsewhere} \end{aligned}$$

$$z(t) = \int_0^t r(\tau) s(T - \textcolor{red}{t} + \tau) d\tau$$

when $r(t) = s(t)$

$$\begin{aligned} z(t) &= \int_0^t s(\tau) s(T - \textcolor{red}{t} + \tau) d\tau \\ &= A^2 \int_0^t \cos(\omega_0 \tau) \cos(\omega_0(T - t + \tau)) d\tau \\ &= \frac{A^2}{2} \int_0^t [\cos(\omega_0(T - t)) + \cos(\omega_0(T - t + 2\tau))] d\tau \\ &= \frac{A^2}{2} \left[\cos(\omega_0(T - t)) - \frac{1}{2\omega_0} \sin(\omega_0(T - t + 2\tau)) \right]_0^T \\ &= \frac{A^2}{2} \left[\cos(\omega_0(T - t)) \textcolor{blue}{\tau} - \frac{1}{2\omega_0} \sin(\omega_0(T - t + 2\tau)) \right]_0^T \end{aligned}$$

$$\begin{aligned} z(\textcolor{red}{t}) &= \int_0^t r(\tau) h(\textcolor{yellow}{t} - \tau) d\tau \\ &= \int_0^t r(\tau) s(T - (\textcolor{red}{t} - \tau)) d\tau \\ &= \int_0^t r(\tau) s(T - \textcolor{red}{t} + \tau) d\tau \end{aligned}$$

References

- [1] <http://en.wikipedia.org/>
- [2] <http://planetmath.org/>
- [3] B. Sklar, "Digital Communications: Fundamentals and Applications"