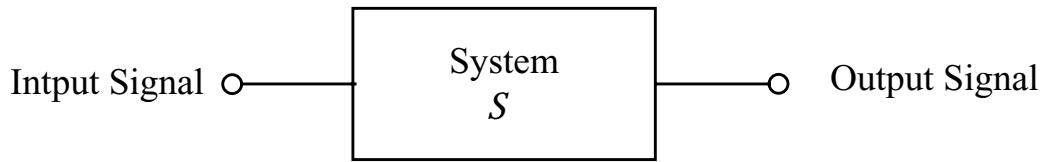


5. Response of Linear Time-Invariant Systems to Random Inputs

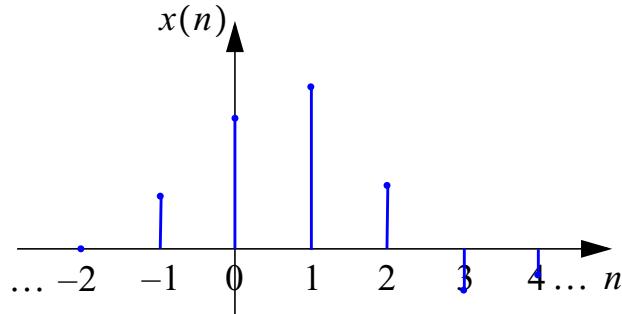
System:



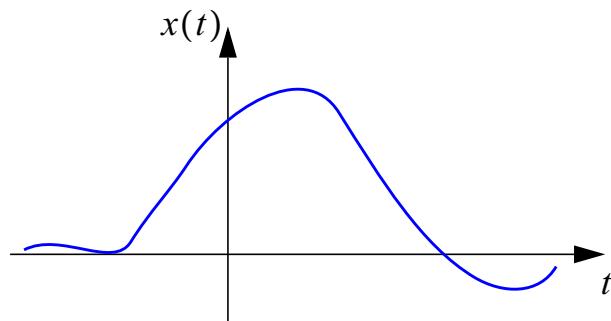
We look at a system as a black box which generates an output signal depending on the input signal and possibly some initial conditions.

We consider two types of signals:

- *Discrete-time signals or sequences*



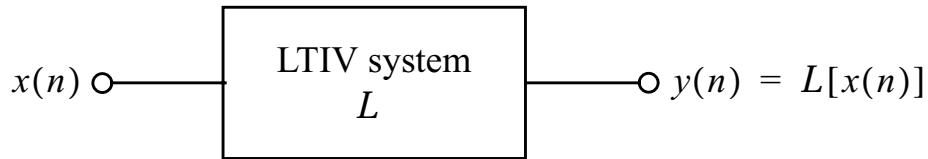
- *Continuous-time signals*



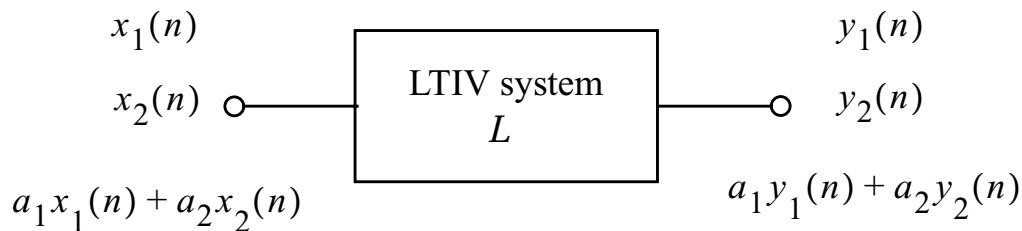
Discrete-time signals are obtained by sampling continuous-time signals.

5.1. Discrete-time linear time-invariant (LTIV) systems

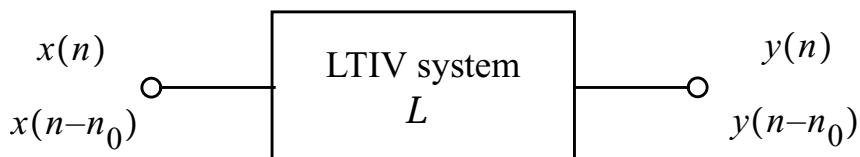
5.1.1. Discrete-time LTIV system



Linear:



Time-invariant:



5.1.2. Steady-state description of a LTIV system

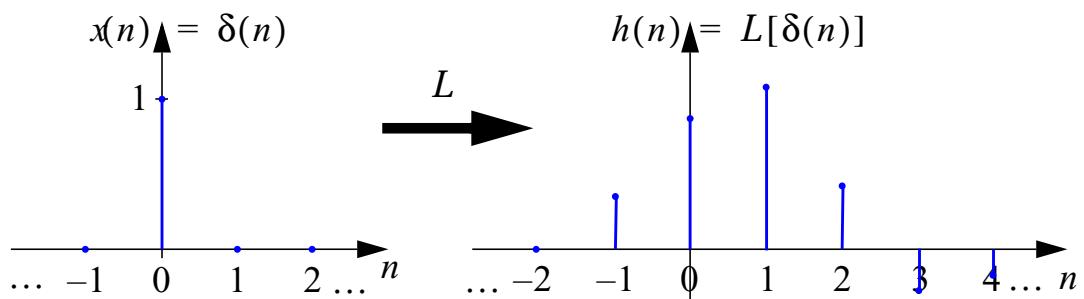
- **Impulse response:**

The impulse response (IR) $h(n)$ of L is the response of L to the unit pulse

$$\delta(n) \equiv \begin{cases} 1 & ; n = 0 \\ 0 & ; n \neq 0 \end{cases}$$

namely

$$h(n) = L[\delta(n)]$$



- **Stable LTIV system:**

A LTIV system is stable if its response to a bounded signal is bounded.

It can be shown [to this end we need (5.1)] that a LTIV system is stable if, and if,

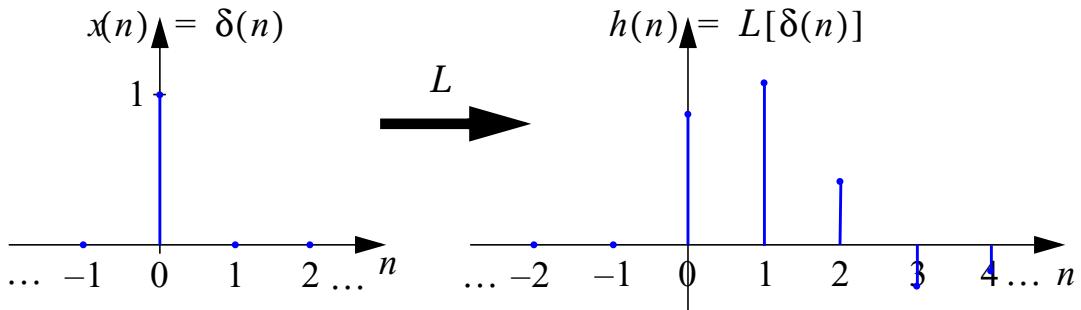
$$\sum_{m=-\infty}^{\infty} |h(m)| < \infty$$

Proof:

□

- **Causal LTIV system:**

$$h(n) = 0 \quad \text{for} \quad n < 0$$



- **Input-output (I-O) relationship of a LTIV system (time domain):**

$$y(n) = \sum_{m=-\infty}^{\infty} h(m)x(n-m)$$

$$= h(n)^*x(n)$$

(5.1)

The symbol * denotes the discrete convolution operation.

Proof:

□

- **(Discrete) Fourier transform:**

Here, $z(n)$ denotes an arbitrary sequence.

$$Z(f) = F\{z(n)\} \equiv \sum_{n=-\infty}^{\infty} z(n) \exp(-j2\pi n f) \quad (|f| < 1/2)$$



$$z(n) = F^{-1}\{Z(f)\} = \int_{-1/2}^{1/2} Z(f) \exp(j2\pi n f) df$$

Useful property extensively used in the sequel:

$$z(n - n_0) \quad \circ \longrightarrow \bullet \quad \exp(-j2\pi n_0 f) Z(f)$$

Proof:

□

- **(Frequency) transfer function of a LTIV system:**

$$H(f) \equiv F\{h(n)\} \equiv \sum_{n=-\infty}^{\infty} h(n) \exp(-j2\pi n f)$$

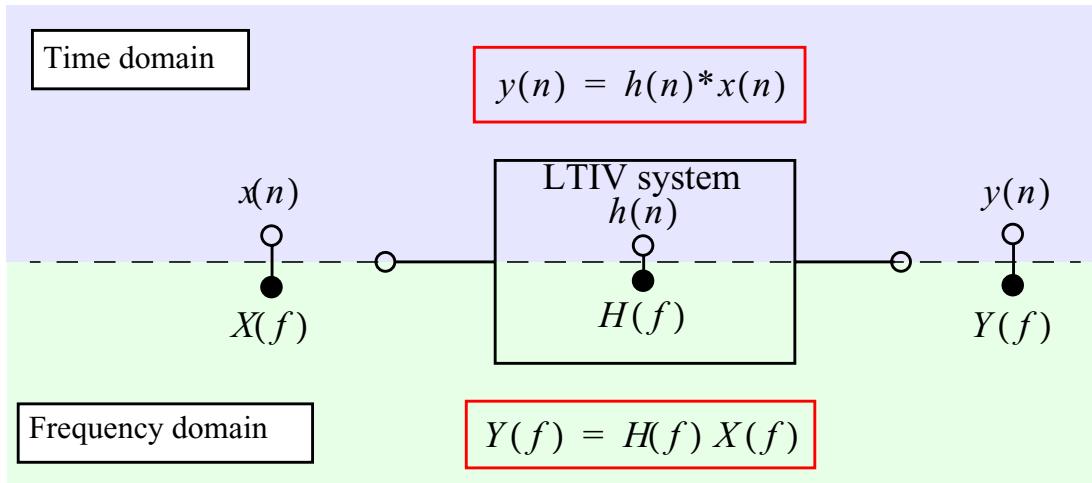
- **I-O relationship of a LTIV system (frequency domain):**

$$Y(f) = H(f)X(f)$$

Proof:

□

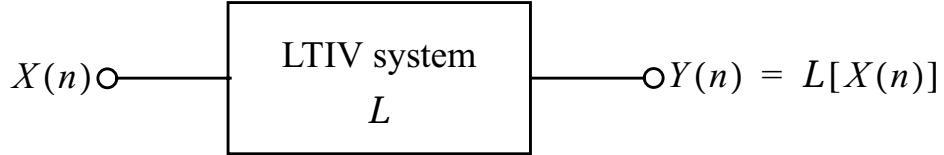
- **Summary: I-O relationship of a LTI system:**



5.1.3. First- and second-order characterization of a LTVI system

- **Random input and output sequences:**

If $X(n)$ is a random sequence (or process), so is $Y(n)$.



- **Second-order characterization of random sequences:**

Here, $Z(n)$ denotes an arbitrary random sequence.

- Expectation:

$$\mu_Z(n) \equiv \mathbf{E}[Z(n)]$$

- Autocorrelation function:

$$R_{ZZ}(n_1, n_2) \equiv \mathbf{E}[Z(n_1)Z(n_2)]$$

- **Second-order properties of the output sequence $Y(n)$:**

- Expectation:

$$\mu_Y(n) = h(n)*\mu_X(n)$$

- Autocorrelation function:

$$R_{YY}(n_1, n_2) = \sum_{m_1=-\infty}^{\infty} \sum_{m_2=-\infty}^{\infty} h(m_1)h(m_2)R_{XX}(n_1 - m_1, n_2 - m_2)$$

Proof:

□

5.1.4. Wide-sense-stationary (WSS) processes

- ***Definition:***

A random sequence $Z(n)$ is WSS if the following conditions are satisfied:

- Expectation:

$$\mu_Z(n) = \mathbf{E}[Z(n)] = \mu_Z$$

- Autocorrelation function:

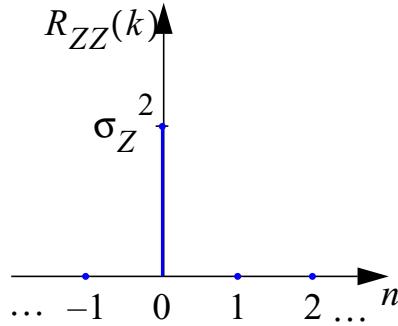
$$R_{ZZ}(n_1, n_1 + k) = \mathbf{E}[Z(n_1)Z(n_1 + k)] = R_{ZZ}(k)$$

- ***White process:***

$Z(n)$ is a white process if it satisfies the following conditions:

- $Z(n)$ is a random process
- $\mu_Z(n) = \mathbf{E}[Z(n)] = 0$

$$- R_{ZZ}(n, n+k) = \mathbf{E}[Z(n)Z(n+k)] = R_{ZZ}(k) = \sigma_Z^2 \delta(k)$$



- *Autocorrelation function of the impulse response:*

$$\begin{aligned} R_{hh}(k) &= \sum_{m=-\infty}^{\infty} h(m)h(m+k) \\ &= h(k)^*h(-k) \end{aligned}$$

Proof:

□

- *Second-order I-O relationship of a LTVI system (time domain):*

- Expectation:

$$\mu_Y = \left[\sum_{m=-\infty}^{\infty} h(m) \right] \mu_X = H(0) \mu_X$$

- Autocorrelation function:

$$R_{YY}(k) = R_{hh}(k)^* R_{XX}(k)$$

Proof:

□

- **Power spectrum of a WSS process:**

The power spectrum of the WSS process $Z(n)$ with the autocorrelation function $R_{ZZ}(k)$ is defined to be

$$S_{ZZ}(f) = F\{R_{ZZ}(k)\} = \sum_{n=-\infty}^{\infty} R_{ZZ}(k) \exp(-j2\pi kf)$$

Notice that from the inverse Fourier transformation

$$R_{ZZ}(k) = \frac{1}{2} \int_{-\frac{1}{2}}^{\frac{1}{2}} S_{ZZ}(f) \exp(j2\pi kf) df$$

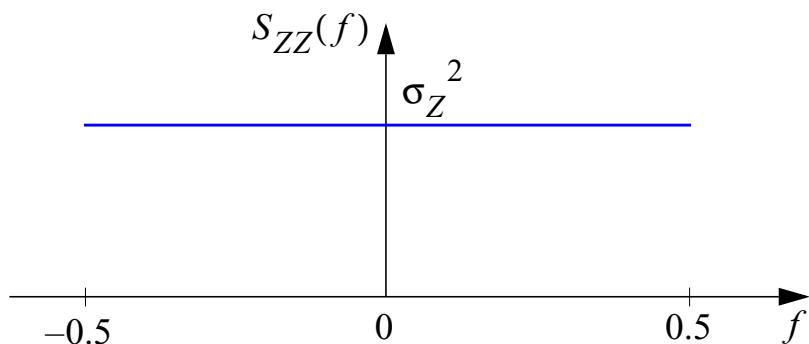
we conclude that

$$\mathbf{E}[Z(n)^2] = R_{ZZ}(0) = \int_{-\frac{1}{2}}^{\frac{1}{2}} S_{ZZ}(f) df$$

- **Spectrum of a white process:**

If $Z(n)$ is a white process:

$$S_{ZZ}(f) = \sigma_Z^2$$



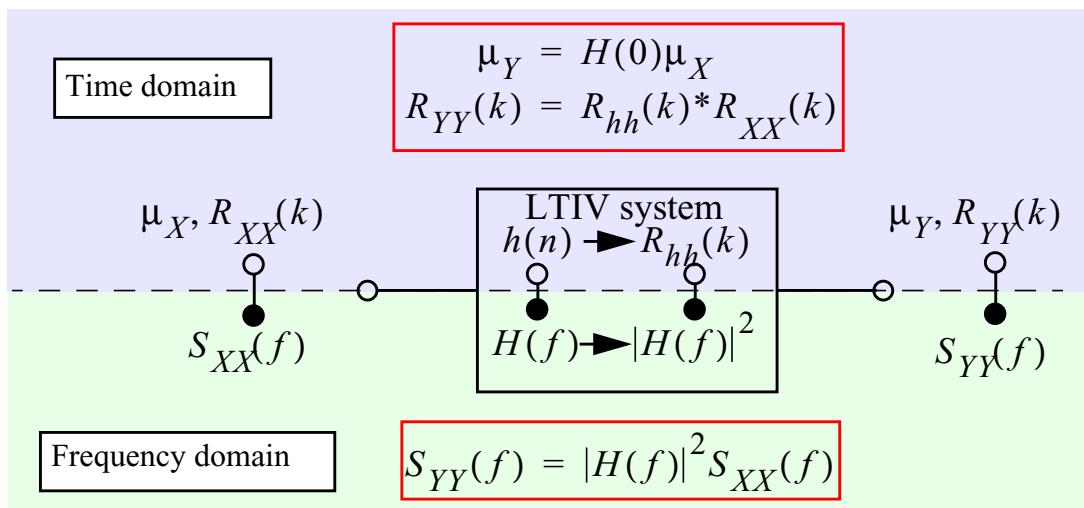
- **Second-order I-O relationship of a LTVI system (frequency domain):**

$$S_{YY}(f) = |H(f)|^2 S_{XX}(f)$$

Proof:

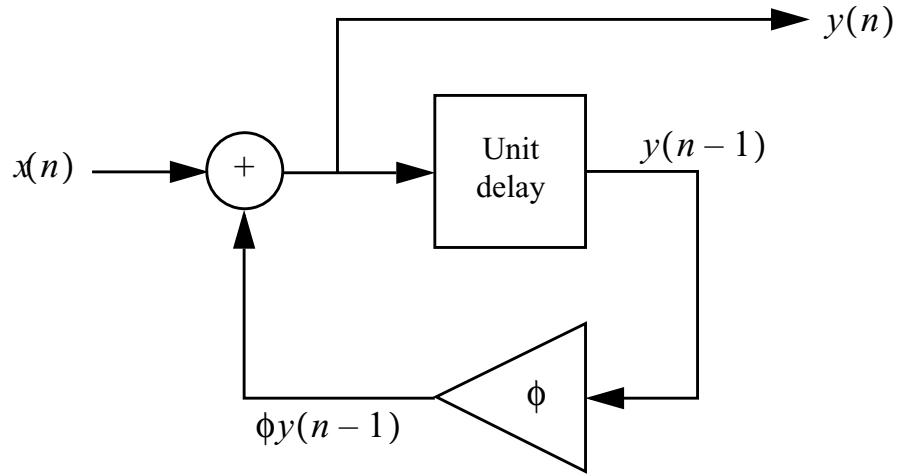
□

- **Summary: Second-order I-O relationship of a LTIV system:**



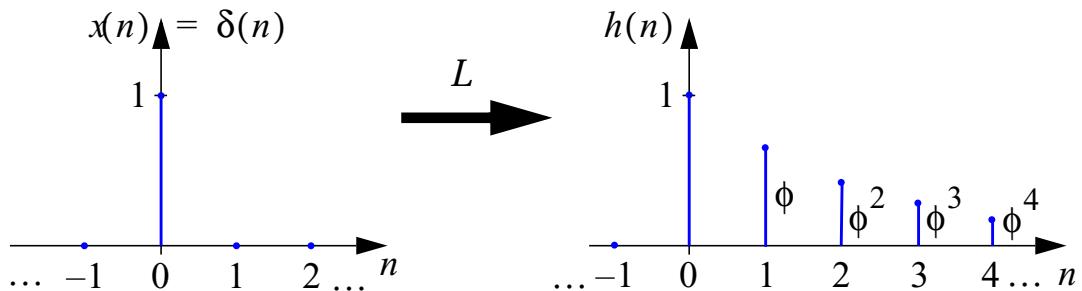
5.1.5. Example: First order recursive filter

- *Block diagram and recursive equation:*



$$y(n) = x(n) + \phi y(n-1) \quad (y(n) = 0 \quad n < 0)$$

- *Impulse response:*



$$h(n) = \begin{cases} 0 & ; \quad n < 0 \\ \phi^n & ; \quad n \geq 0 \end{cases}$$

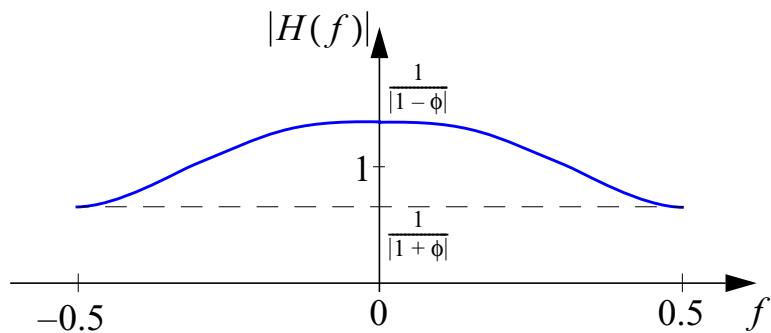
- *Stability condition:*

$$\sum_{n=-\infty}^{\infty} |h(n)| = \sum_{n=0}^{\infty} |\phi|^n = \lim_{N \rightarrow \infty} \frac{1 - |\phi|^N}{1 - |\phi|} < \infty \iff |\phi| < 1$$

$$\left[\sum_{n=0}^N a^n = \frac{1 - a^{N+1}}{1 - a} \right]$$

- **Transfer function:**

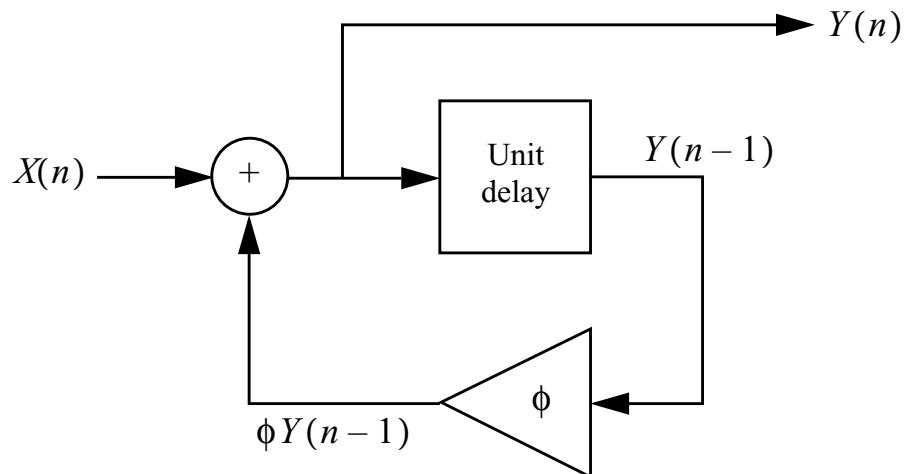
$$\begin{aligned}
 H(f) &= F\{h(n)\} = \sum_{n=0}^{\infty} \phi^n \exp(-j2\pi n f) \\
 &= \sum_{n=0}^{\infty} [\phi \exp(-j2\pi f)]^n \\
 &= \frac{1}{1 - \phi \exp(-j2\pi f)}
 \end{aligned}$$



Another more direct way to compute the transfer function:

$$\begin{aligned}
 y(n) &= x(n) + \phi y(n-1) \\
 Y(f) &= X(f) + \phi \exp(-j2\pi f) Y(f)
 \end{aligned}$$

- **Random input and output:**



$$Y(n) = X(n) + \phi Y(n-1)$$

- **Second-order I-O relationship:**

- Time domain:

$$\begin{aligned}\mu_Y &= H(0)\mu_X \\ &= \frac{1}{1-\phi}\mu_X\end{aligned}$$

$$\begin{aligned}R_{YY}(k) &= R_{hh}(k)*R_{XX}(k) \\ &= \frac{\phi^{|k|}}{1-\phi^2}*R_{XX}(k)\end{aligned}$$

$$\left[R_{hh}(k) = \sum_{m=0}^{\infty} \phi^m \phi^{m+|k|} = \phi^{|k|} \sum_{m=0}^{\infty} \phi^{2m} = \phi^{|k|} \frac{1}{1-\phi^2} \right]$$

- Frequency domain:

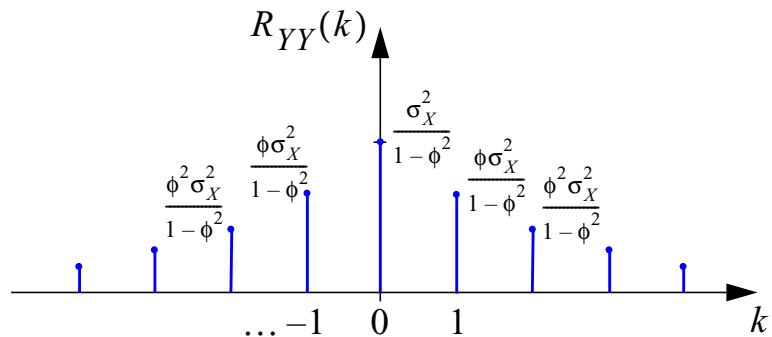
$$\begin{aligned}S_{YY}(f) &= |H(f)|^2 S_{XX}(f) \\ &= \frac{1}{|1-\phi \exp(-j2\pi f)|^2} S_{XX}(f)\end{aligned}$$

- **Special case: AR(1) process (see Section 6.2):**

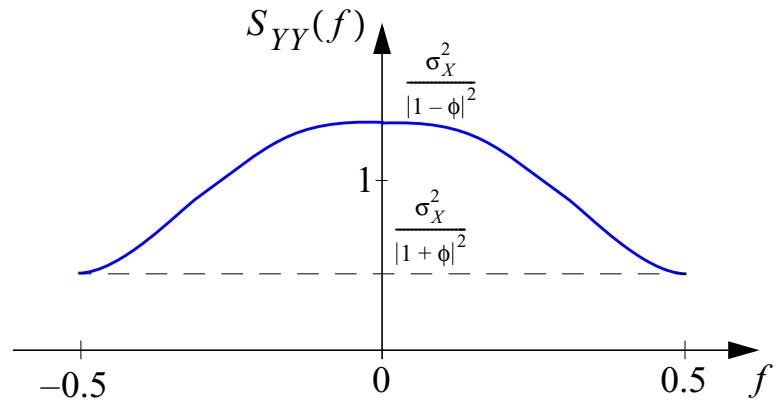
If $X(n)$ is a white Gaussian process,

$$\mu_Y = 0$$

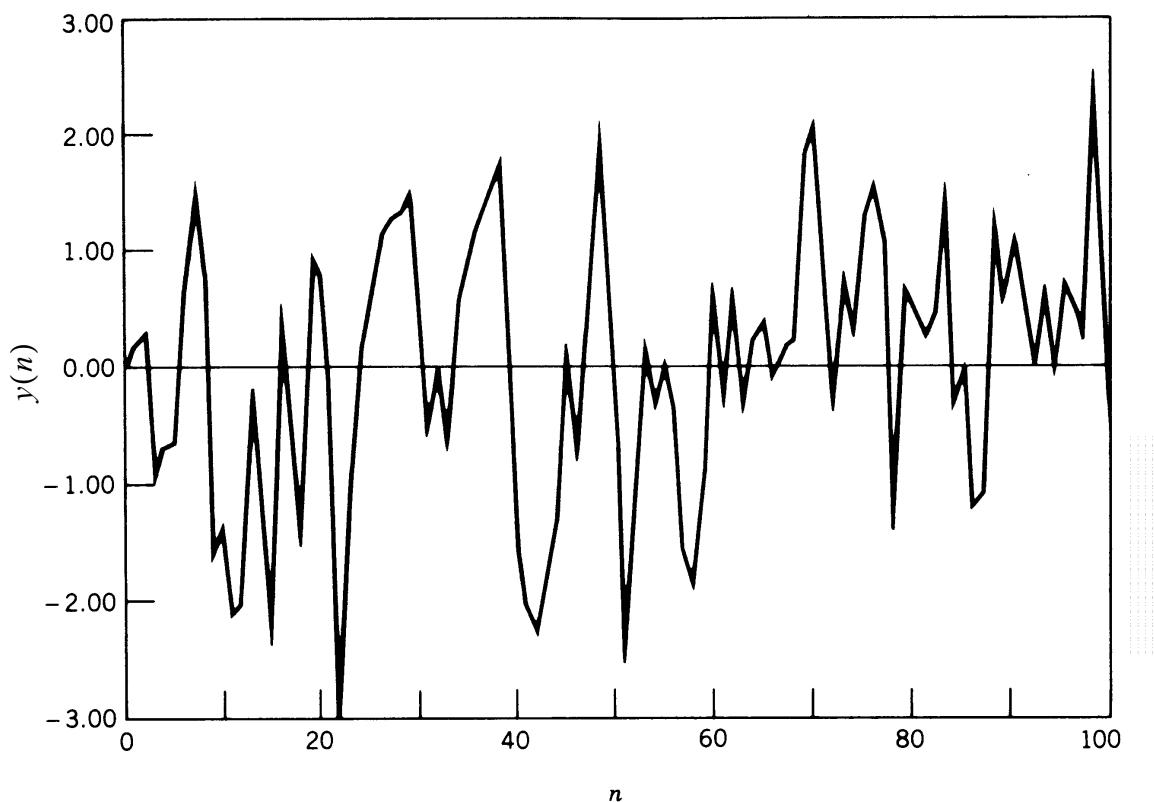
$$R_{YY}(k) = \frac{\phi^{|k|}}{1-\phi^2} \sigma_X^2$$



$$S_{YY}(f) = \frac{\sigma_X^2}{|1 - \phi \exp(-j2\pi f)|^2}$$

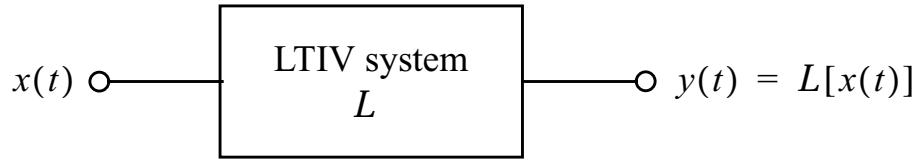


One realization of $Y(n)$:

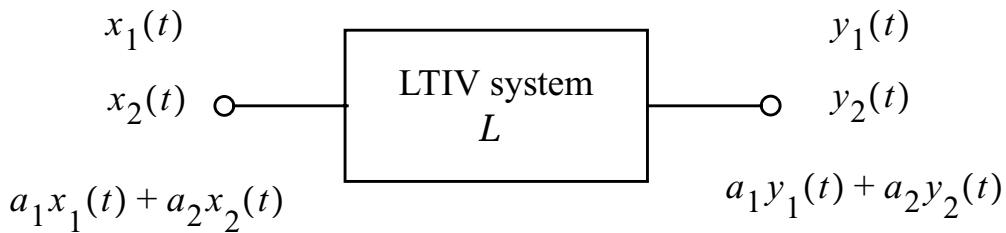


5.2. Continuous-time LTIV systems

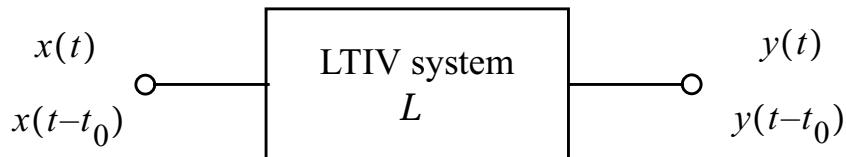
5.2.1. Continuous-time LTIV system



Linear:



Time-invariant:



5.2.2. Steady-state description of a LTIV system

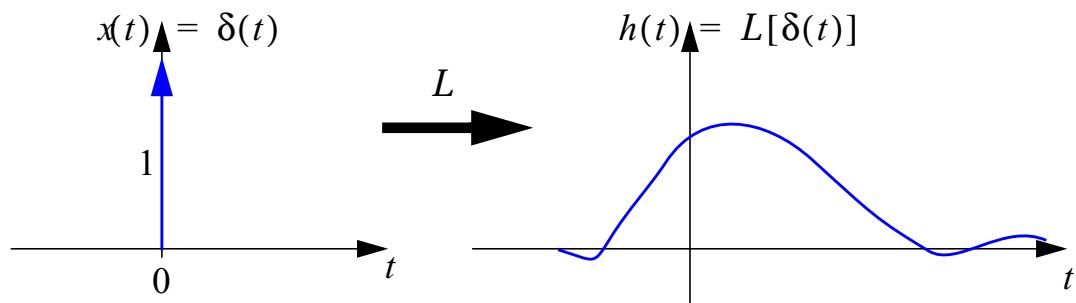
- **Impulse response:**

The impulse response $h(t)$ of L is the response of L to the Dirac impulse

$$\delta(t) \equiv \begin{cases} \infty ; & t = 0 \\ 0 ; & t \neq 0 \end{cases} \quad \int_{-\infty}^{\infty} \delta(t) dt = 1,$$

namely

$$h(t) = L[\delta(t)]$$



- **Stable LTIV system:**

A LTIV system is stable if its response to a bounded signal is bounded.

It can be shown [to this end we need (5.2)] that a LTIV system is stable if, and if,

$$\int_{-\infty}^{\infty} |h(t)| dt < \infty$$

- **Causal LTIV system:**

$$h(t) = 0 \quad t < 0$$

- **Input-output relationship of a LTIV system (time domain):**

$$\begin{aligned} y(t) &= \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau \\ &= h(t)^*x(t) \end{aligned} \tag{5.2}$$

Here, the symbol * denotes the continuous convolution operation.

- **(Continuous) Fourier transform:**

$$\begin{aligned} Z(f) &= F\{z(t)\} \equiv \int_{-\infty}^{\infty} z(t) \exp(-j2\pi ft) d\tau \\ z(t) &= F^{-1}\{Z(f)\} = \int_{-\infty}^{\infty} Z(f) \exp(j2\pi ft) df \end{aligned}$$

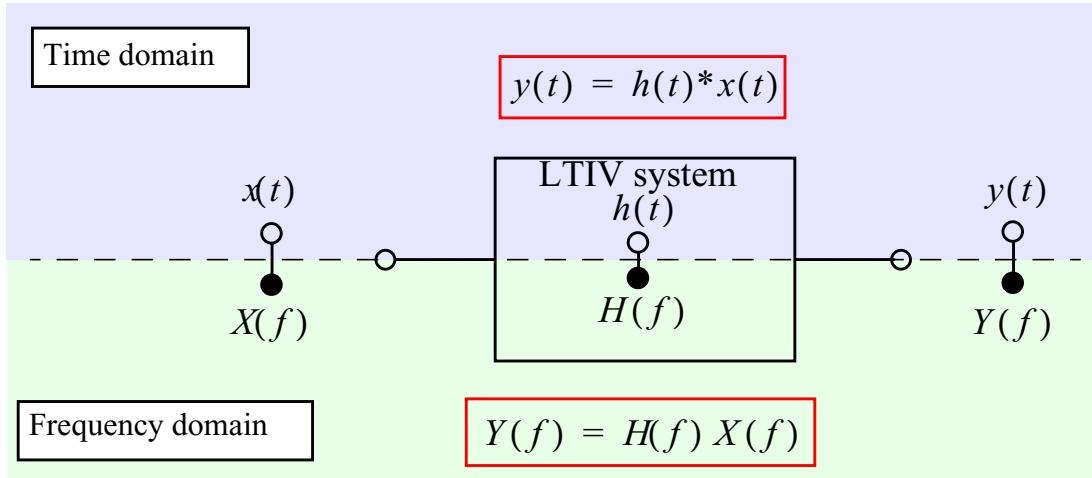
- **(Frequency) transfer function of a LTIV system**

$$H(f) = F\{h(t)\} = \int_{-\infty}^{\infty} h(t) \exp(-j2\pi ft) dt$$

- **I-O relation ship of a LTIV system (frequency domain):**

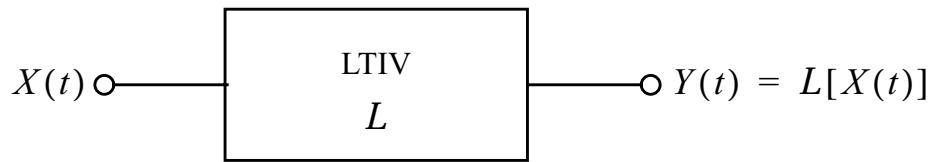
$$Y(f) = H(f)X(f)$$

- **Summary: I-O relationship of a LTI system:**



5.2.3. Second-order characterization of LTI

- **Random input and output processes:**



- **Second-order characterization of random processes:**

Here, $Z(t)$ denotes an arbitrary random process.

- Expectation:

$$\mu_Z(t) \equiv \mathbf{E}[Z(t)]$$

- Autocorrelation function:

$$R_{ZZ}(t_1, t_2) \equiv \mathbf{E}[Z(t_1)Z(t_2)]$$

- **Second-order properties of the output process $Y(t)$:**

- Expectation:

$$\mu_Y(t) = h(t)*\mu_X(t)$$

- Autocorrelation function:

$$R_{YY}(t_1, t_2) \equiv \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(u_1)h(u_2)R_{XX}(t_1-u_1, t_2-u_2)du_1 du_2$$

5.2.4. Wide-sense-stationary (WSS) processes

- **Wide-sense stationarity:**

A random process $Z(t)$ is WSS if the following conditions are satisfied:

- Expectation:

$$\mu_Z(t) \equiv \mu_Z$$

- Autocorrelation function:

$$R_{ZZ}(t_1, t_1 + \tau) \equiv R_{ZZ}(\tau)$$

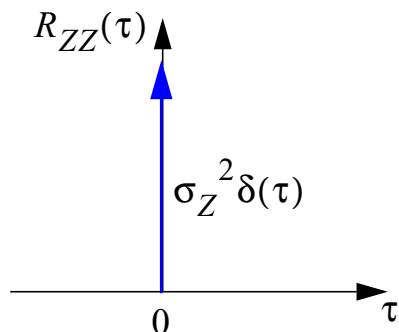
- **White process:**

$Z(t)$ is a white process if it satisfies the following conditions:

- $Z(t)$ is a random process

- $\mu_Z(t) = \mathbf{E}[Z(t)] = 0$

- $R_{ZZ}(t, t + \tau) = \mathbf{E}[Z(t)Z(t + \tau)] = R_{ZZ}(\tau) = \sigma_Z^2 \delta(\tau)$



- **Autocorrelation function of the impulse response:**

$$\begin{aligned} R_{hh}(\tau) &= \int_{-\infty}^{\infty} h(u)h(u + \tau)du \\ &= h(\tau)^*h(-\tau) \end{aligned}$$

- **Second-moment I-O relationship of a LTVI system (time domain):**

- Expectation:

$$\mu_Y = \left[\int_{-\infty}^{\infty} h(u)du \right] \mu_X = H(0)\mu_X$$

- Autocorrelation function:

$$R_{YY}(\tau) = R_{hh}(\tau)*R_{XX}(\tau)$$

- **Power spectrum of a WSS process:**

$Z(t)$ is WSS with the autocorrelation function $R_{ZZ}(\tau)$.

$$S_{ZZ}(f) = F\{R_{ZZ}(\tau)\}$$

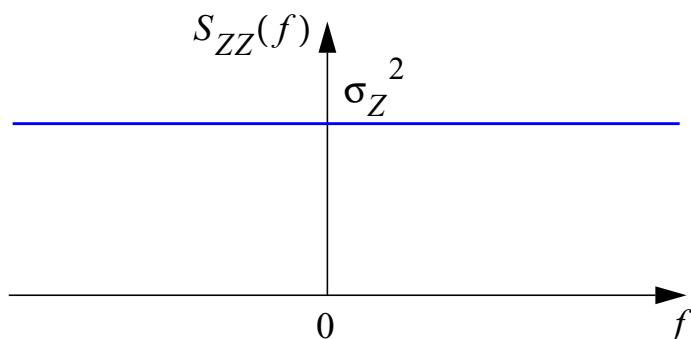
Notice the identities

$$\mathbf{E}[Z(t)^2] = R_{ZZ}(0) = \int_{-\infty}^{\infty} S_{ZZ}(f) df$$

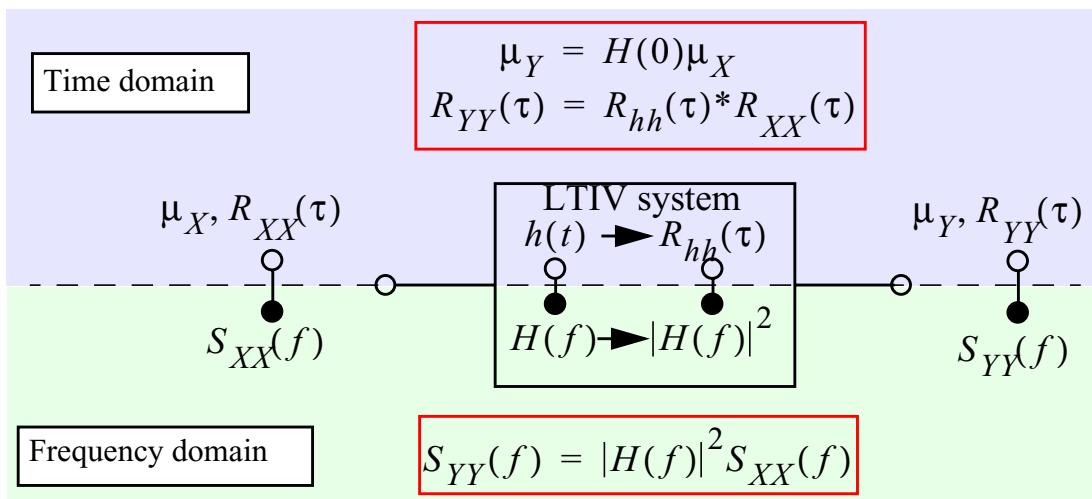
- **Spectrum of a white process:**

If $Z(t)$ is a white process:

$$S_{ZZ}(f) = \sigma_Z^2$$

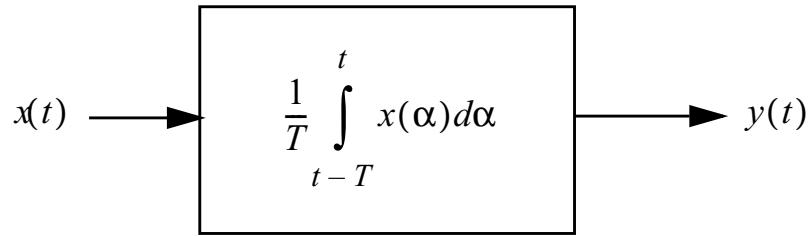


- **Summary: Second order I-O relationship of a LTIV system:**



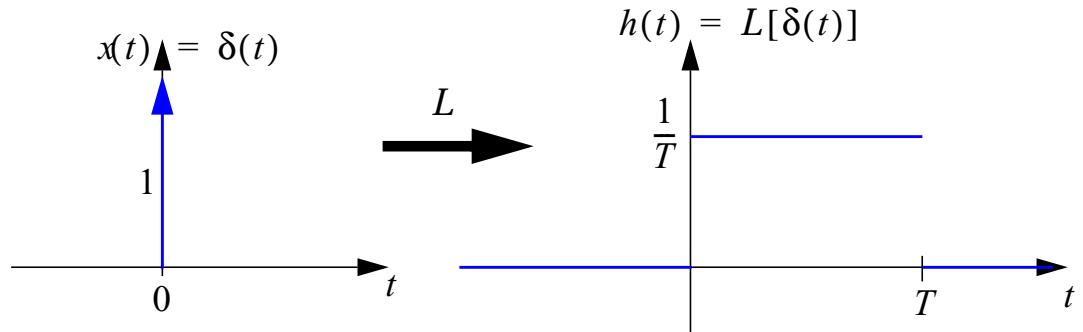
5.2.5. Example: Ideal integrator

- *Block diagram and input-output relationship:*



$$y(t) = \frac{1}{T} \int_{t-T}^t x(\alpha) d\alpha$$

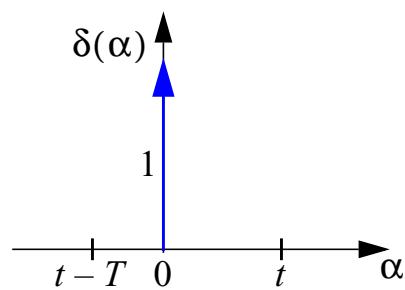
- *Impulse response:*



$$h(t) = \begin{cases} \frac{1}{T} ; & 0 < t < T \\ 0 ; & \text{elsewhere} \end{cases}$$

Proof:

$$\frac{1}{T} \int_{t-T}^t \delta(\alpha) d\alpha = \begin{cases} 1 ; & 0 \in (t, t-T) \\ 0 ; & \text{elsewhere} \end{cases}$$



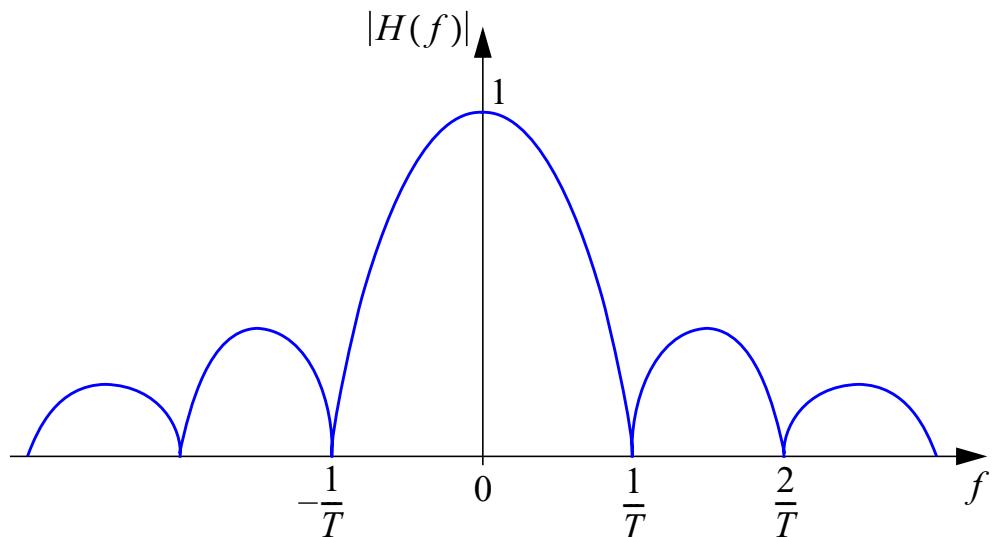
□

- **Stability condition:**

$$\int_{-\infty}^{\infty} |h(t)| dt = 1$$

- **Transfer function:**

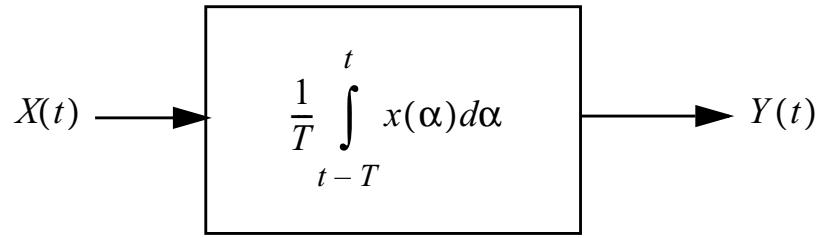
$$\begin{aligned}
 H(f) &= F\{h(t)\} = \int_{-\infty}^{\infty} h(t) \exp(-j2\pi ft) dt \\
 &= \frac{1}{T} \int_0^T \exp(-j2\pi ft) dt = \exp(-j\pi fT) \frac{\sin(\pi fT)}{\pi fT} \\
 &= \exp(-j\pi fT) \operatorname{sinc}(fT)
 \end{aligned}$$



Proof:

□

- **Random input and output:**



$$Y(t) = \frac{1}{T} \int_{t-T}^t X(\alpha) d\alpha$$

- **Second-order I-O relationship:**

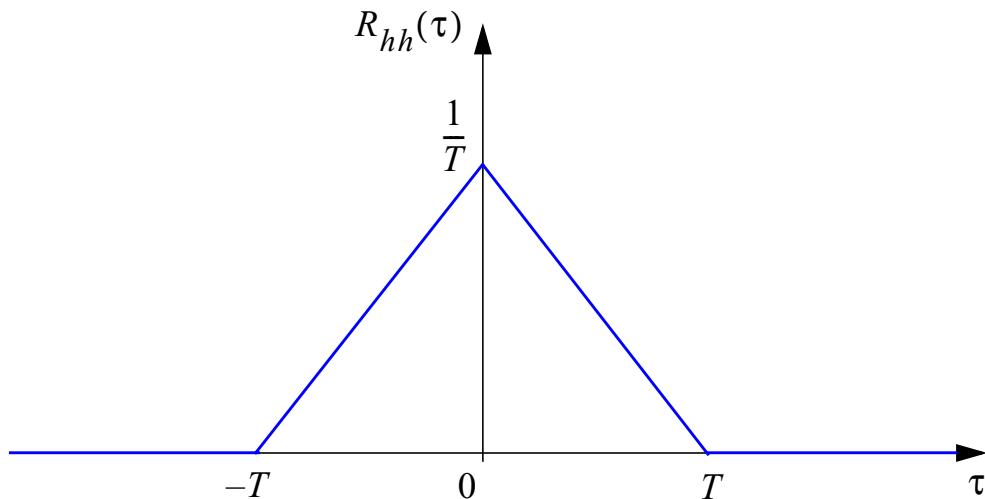
- Time domain:

$$\begin{aligned}\mu_Y &= H(0)\mu_X \\ &= \mu_X\end{aligned}$$

$$R_{YY}(\tau) = R_{hh}(\tau)^* R_{XX}(\tau)$$

with

$$R_{hh}(\tau) = \begin{cases} \frac{1}{T} \left(1 - \frac{|\tau|}{T}\right) ; & -T < \tau < T \\ 0 ; & \text{elsewhere} \end{cases}$$



Proof:

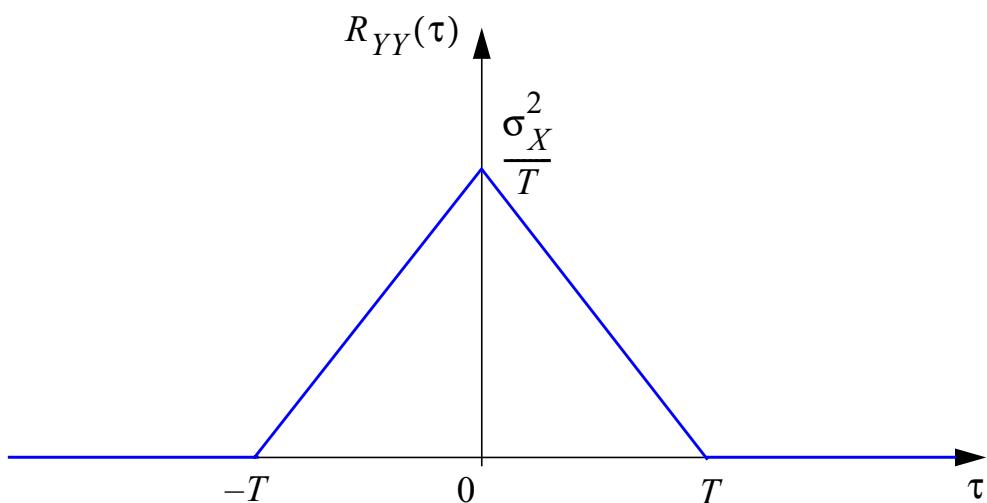
□

- Frequency domain:

$$\begin{aligned} S_{YY}(f) &= |H(f)|^2 S_{XX}(f) \\ &= \text{sinc}(fT)^2 S_{XX}(f) \end{aligned}$$

- **Special case:** $X(t)$ is a white process,

$$\mu_Y = 0 \quad R_{YY}(\tau) = \begin{cases} \frac{\sigma_X^2}{T} \left(1 - \frac{|\tau|}{T}\right); & -T < \tau < T \\ 0; & \text{elsewhere} \end{cases}$$



$$S_{YY}(f) = \sigma_X^2 \operatorname{sinc}(fT)^2$$

