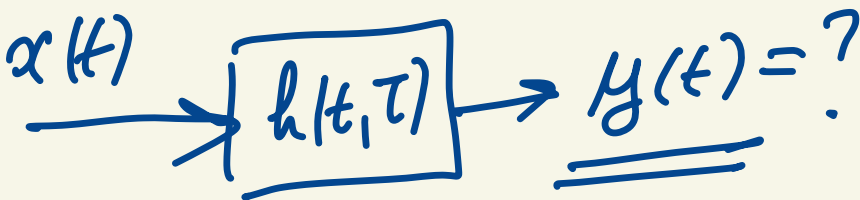
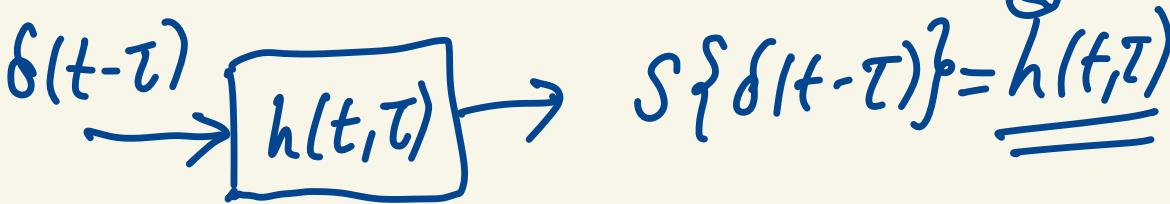
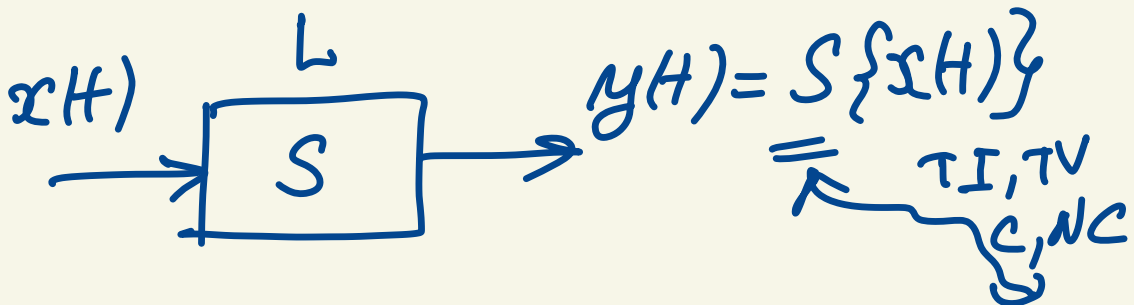



Lecture 5

Now, focus on Linear Systems



Generic representation of $x(t)$

$$x(t) = \int_{-\infty}^{+\infty} x(\tau) \delta(t - \tau) d\tau$$

$$y(t) = \int_{-\infty}^{+\infty} \{x(t)\}$$

$$= \int_{-\infty}^{+\infty} \left\{ \int_{-\infty}^{+\infty} x(\tau) \delta(t - \tau) d\tau \right\}$$

$$\rightarrow \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \{x(\tau) \delta(t - \tau)\} d\tau$$

$$\int_{-\infty}^{+\infty} x(\tau) \underbrace{\int_{-\infty}^{+\infty} \delta(t - \tau) d\tau}_{h(t, \tau)} d\tau$$

$$y(t) = \int_{-\infty}^{+\infty} x(\tau) \underline{h(t, \tau)} d\tau$$

Convolution Integral
for Linear Systems.

If S is $L + \underline{TI}$ $h(t)$

because of TI $h(t, \tau) = \underline{h(t - \tau)}$

$$y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t - \tau) d\tau$$

conv. int. for LTI systems

If S is $L + \underline{TI} + \underline{C}$

$$h(t, \tau) = h(t - \tau) \cdot u(t - \tau)$$

$$y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t - \tau) u(t - \tau) d\tau$$

$$y(t) = \int_{-\infty}^t x(\tau) h(t - \tau) d\tau$$

Conv. Int. for LTI/C sys.

For LTI

$$y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t - \tau) d\tau$$

$$y(t) \triangleq \underline{x(t)} * \underline{h(t)}$$

↑
convolution

Property:

$$x(t) * h(t) = h(t) * x(t)$$

$$h(t) * x(t) = \int_{-\infty}^{+\infty} h(\tau) x(t-\tau) d\tau$$

|| -∞ +∞ || ? proof.

$$x(t) * h(t) = \int_{-\infty}^{+\infty} x(\tau) h(t-\tau) d\tau$$

Proof.

$$\int_{-\infty}^{+\infty} x(\tau) h(t-\tau) d\tau =$$

$$t - \tau = \sigma \quad d\tau = -d\sigma$$

$$\tau = t - \sigma$$

$\tau \rightarrow -\infty$	$\sigma \rightarrow +\infty$
$\tau \rightarrow +\infty$	$\sigma \rightarrow -\infty$

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

Back to system properties

- Linear

- TI

- C

+

1 more

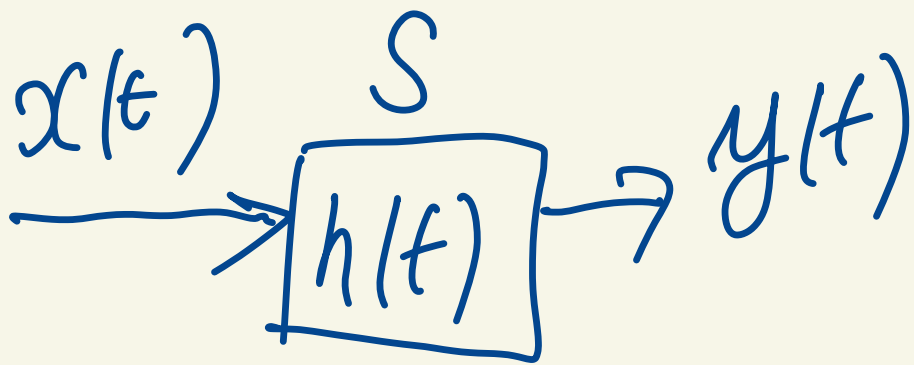
Bounded Input
Bounded Output
Stability

BIBO Stability

Bounded Input

$$|x(t)| < M < +\infty \quad \forall t$$

LTI



if $|x(t)| < M < +\infty$

results in $|y(t)| < K < +\infty$

$\Rightarrow S$ is BIBO stable.

$$y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t-\tau) d\tau$$

$$|y(t)| = \left| \int_{-\infty}^{\infty} h(\tau) x(t-\tau) d\tau \right|$$

reminder: $|a+b| \leq |a| + |b|$

$$|a \cdot b| = |a| \cdot |b|$$

$$\begin{aligned}
 |y(t)| &\leq \int_{-\infty}^{\infty} |h(\tau) x(t-\tau)| d\tau \\
 &= \int_{-\infty}^{\infty} |h(\tau)| \underbrace{|x(t-\tau)|}_{< M} d\tau
 \end{aligned}$$

$$\leq M \int_{-\infty}^{+\infty} |h(\tau)| d\tau$$

if BIBO stable
then

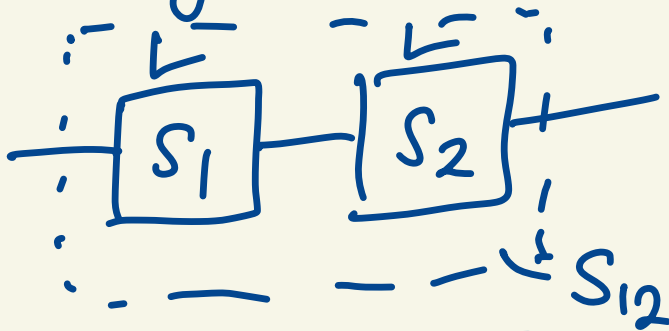
$$< +\infty$$

$$\Rightarrow \int_{-\infty}^{+\infty} |h(\tau)| d\tau < +\infty$$

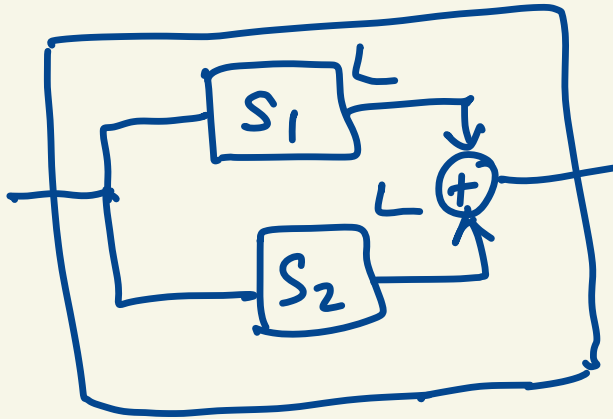
LT1 + BIBO stable

$h(t)$ is absolutely summable.

System connections.



cascaded system.

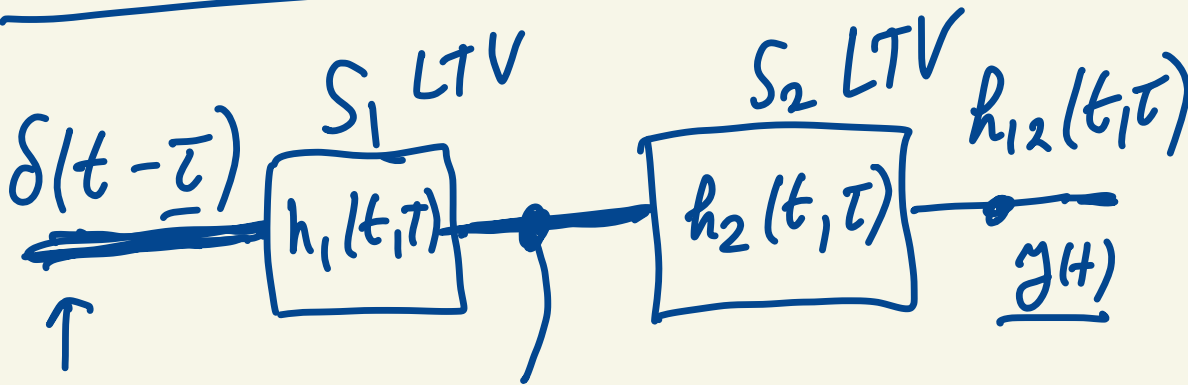
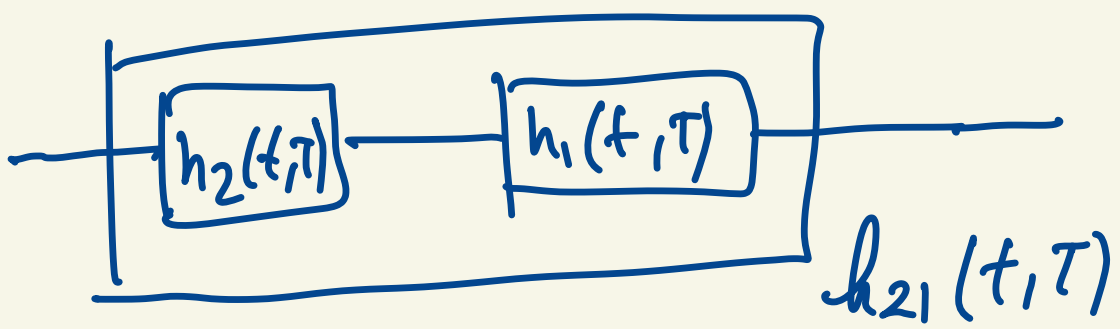


parallel system connection.

S_3



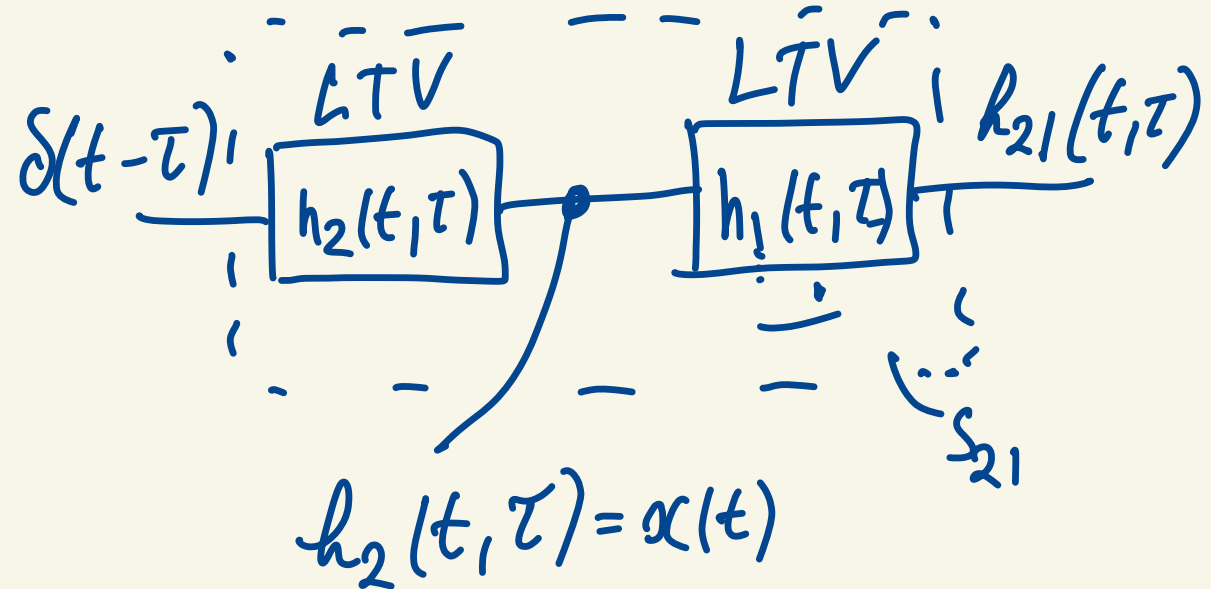
$h_{12}(t, T)$ ✓



$$\underline{x(t)} = \underline{h_1(t, \underline{\tau})}$$

$$S_2: y(t) = \int_{-\infty}^{+\infty} x(\sigma) h_2(t, \sigma) d\sigma$$

$$S_{12}: h_{12}(t, \underline{\tau}) = \int_{-\infty}^{+\infty} h_1(\sigma, \tau) h_2(t, \sigma) d\sigma$$



$$h_{21}(t, \tau) = \int_{-\infty}^{+\infty} x(\sigma) h_1(t, \sigma) d\sigma$$

$$s_{21} h_{21}(t, \tau) = \int_{-\infty}^{+\infty} h_2(\sigma, \tau) h_1(t, \sigma) d\sigma$$

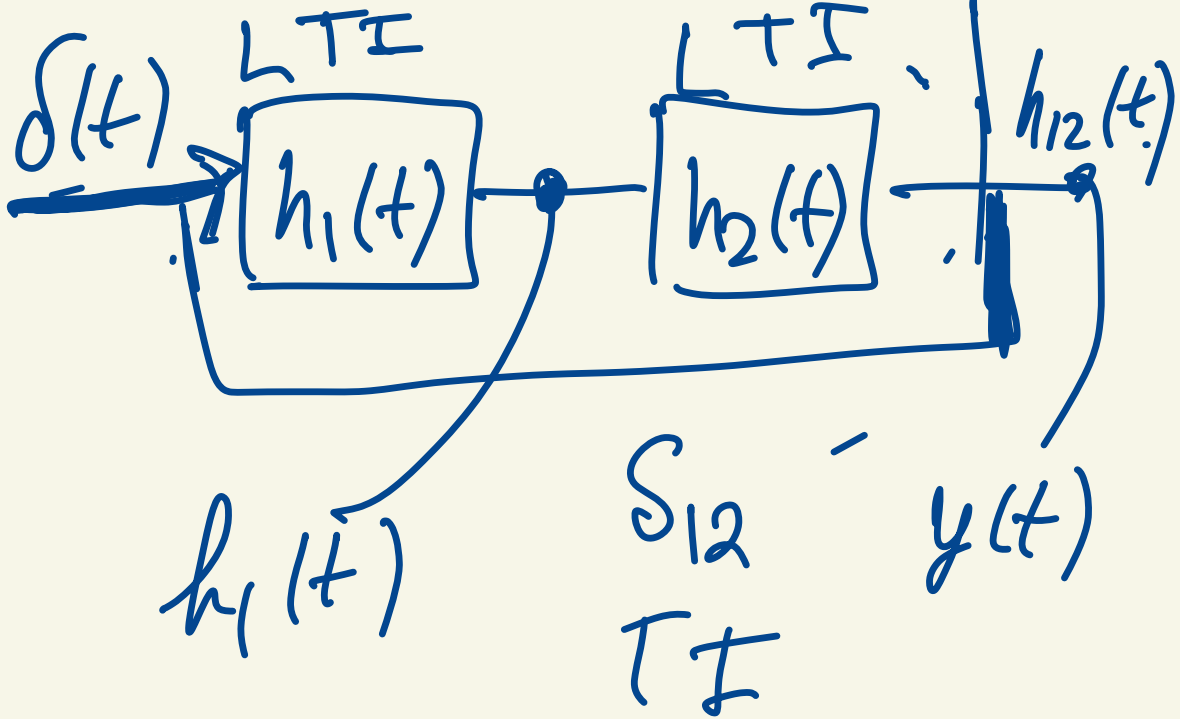
$$\underline{h_{12}(t, \tau)} \neq \underline{h_{21}(t, \tau)}$$

S_{12} is different S_{21}
from

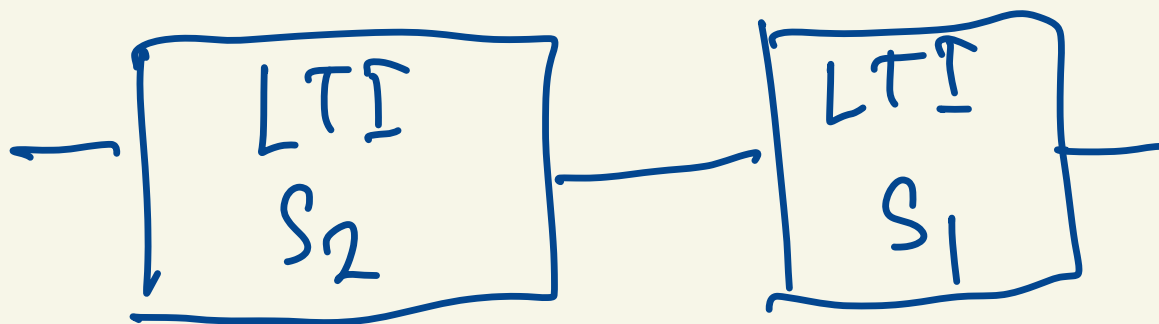
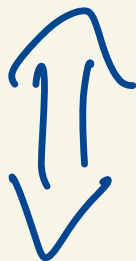
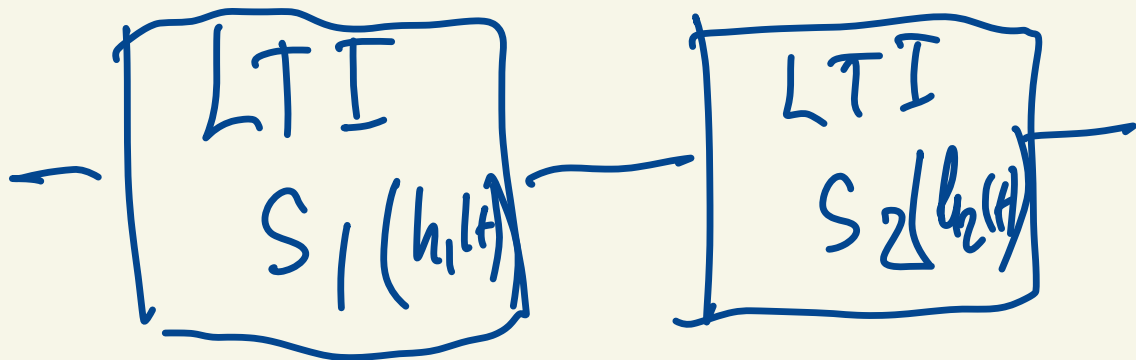
Order of cascading
LTV systems matters

What if S_1 and S_2
are LTI?

$$h(t) = S\{S(t)\}$$

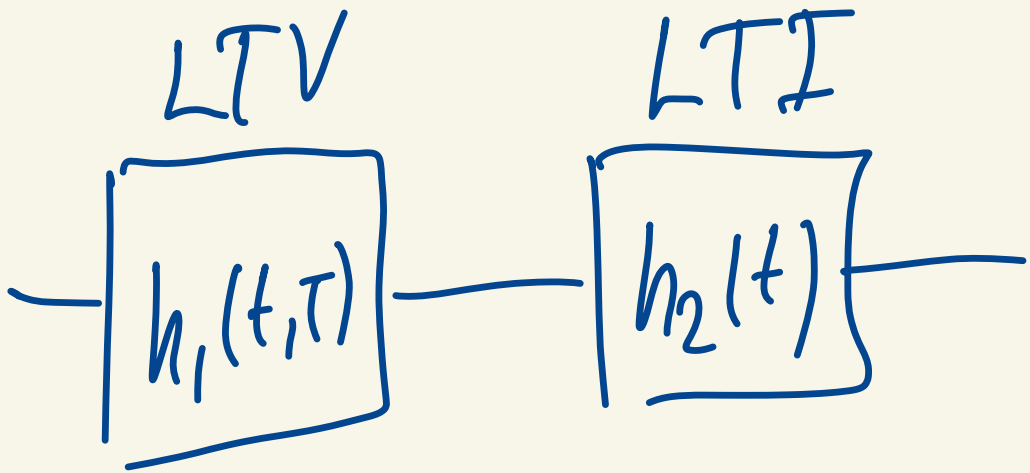


$$\begin{aligned}
 y(t) &= \underline{h_{12}(t)} = h_1(t) * h_2(t) \\
 &= h_2(t) * h_1(t) \\
 &= h_{21}(t)
 \end{aligned}$$



if S_1 and S_2 are LTI
order of cascading does

not matter.



use LTV cascade.

So far we were
able to solve

system analysis

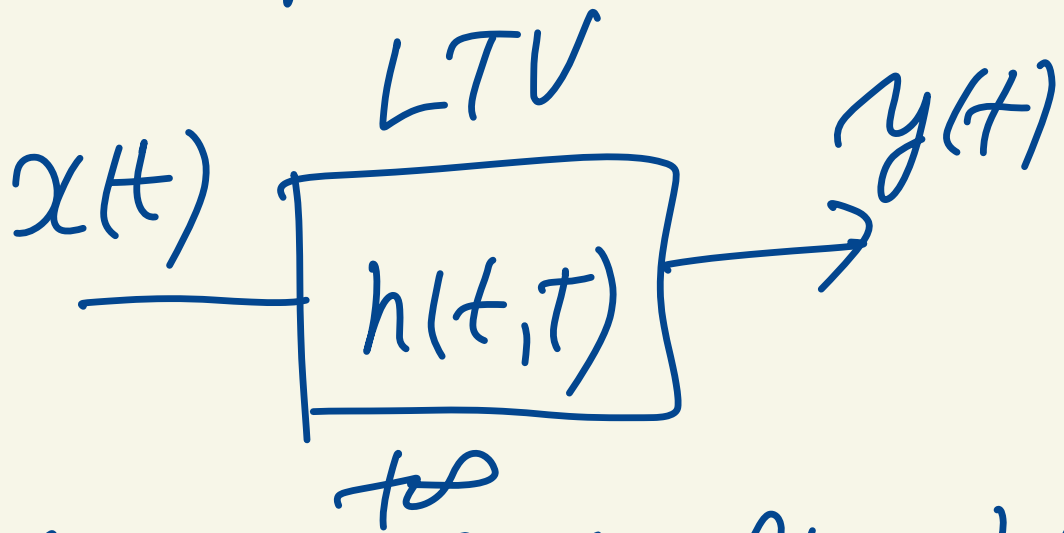
questions using

convolution
integral

given $x(t)$ input

and $S(L)$ $h(t, \tau)$
IRF of S .

we could predict $y(t)$ output.

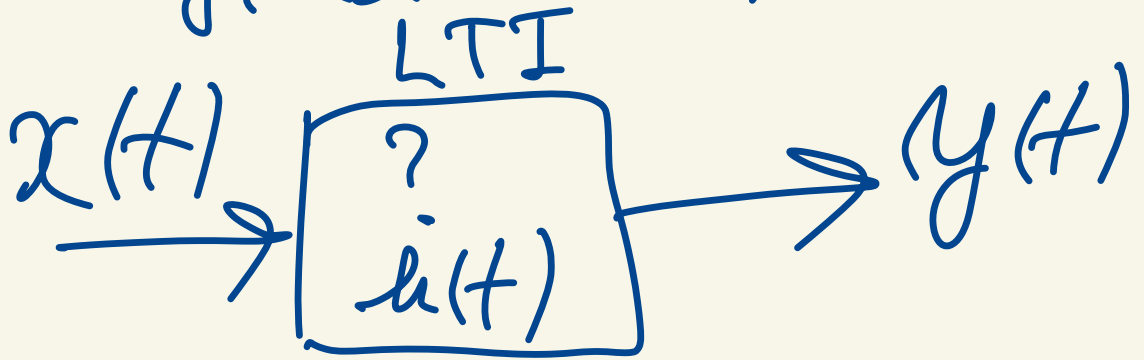


$$y(t) = \int_{-\infty}^{\infty} x(\tau) h(t, \tau) d\tau$$

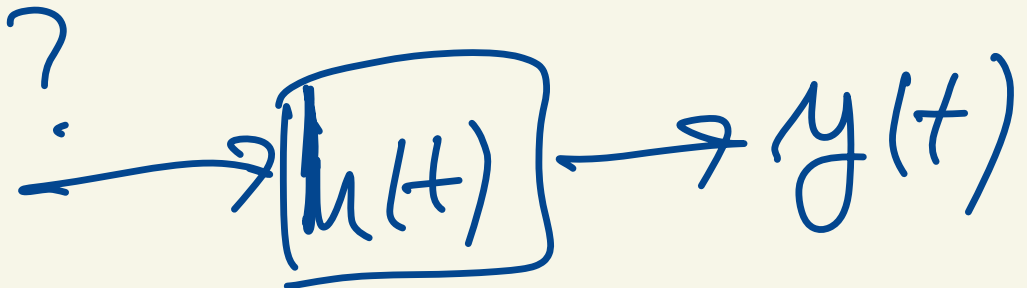
LTI

$$\underline{\underline{y(t) = \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau}} =$$

How about
given $x(t)$ and $y(t)$
find $h(t)$



not easy! deconvolution



what is $x(t)$?

We will learn
transformations
to solve these
types of problems.

Laplace Transform.

$$x(t) \longrightarrow X(s)$$

s-domain

Fourier Transforms

$$x(t) \rightarrow X(f)$$
$$X(\omega)$$

f (or ω) domain

$X(s) \rightarrow$ polynomials

operations

factoring polynomials

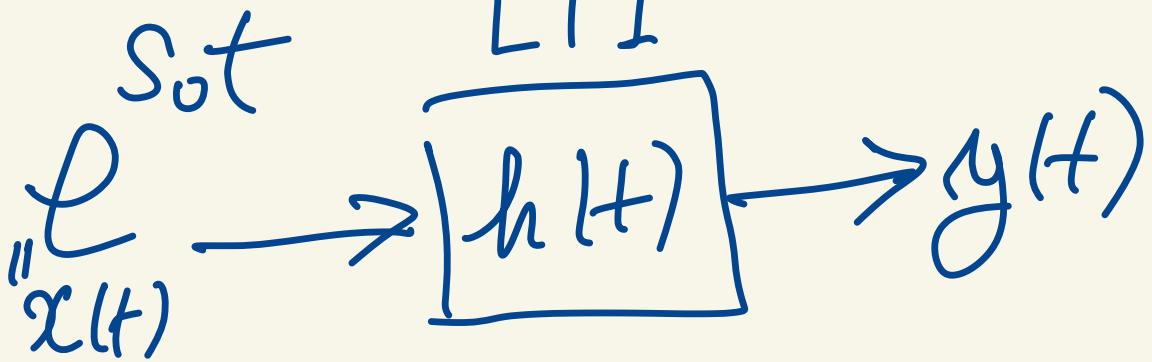
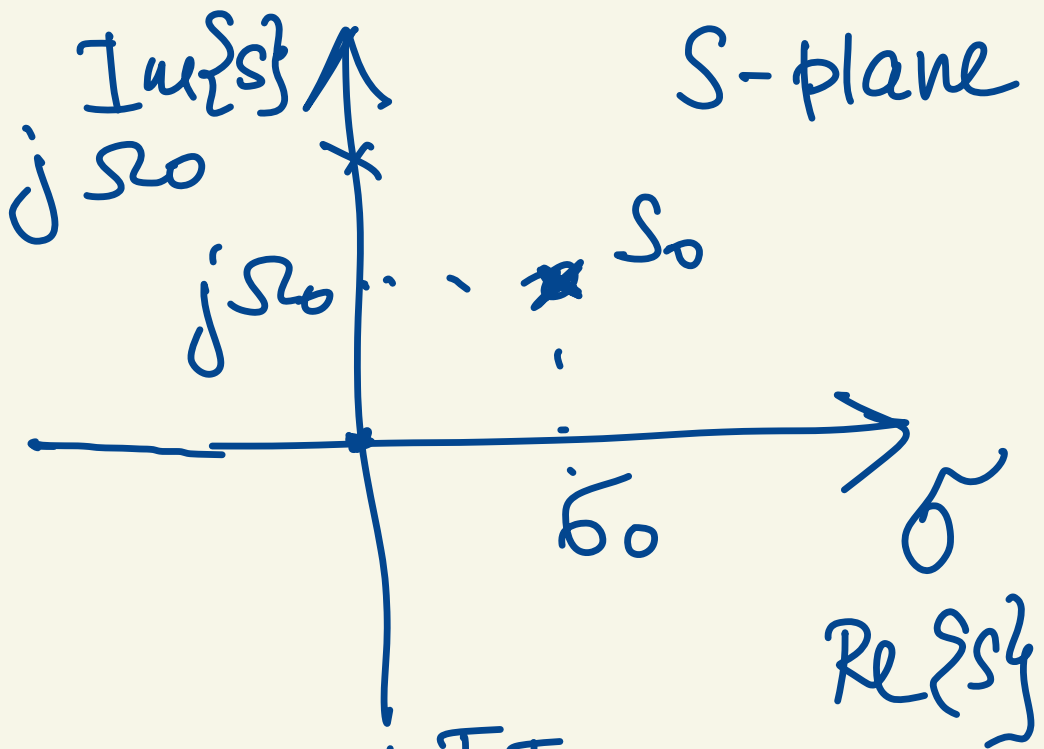
$$y(t) = \overset{\swarrow}{x(t)} * h(t)$$

we are heading here.

$$\underline{Y(s)} = \underline{X(s)} \cdot \underline{H(s)}$$

Complex exponential
as a probing signal
to LTI systems

$$x(t) = e^{s_0 t} \quad s_0 \in \mathbb{C} \quad s_0 = \sigma_0 + j\Omega_0$$



$$y(t) = x(t) * h(t)$$

$$= \int_{-\infty}^{+\infty} h(\tau) x(t-\tau) d\tau$$

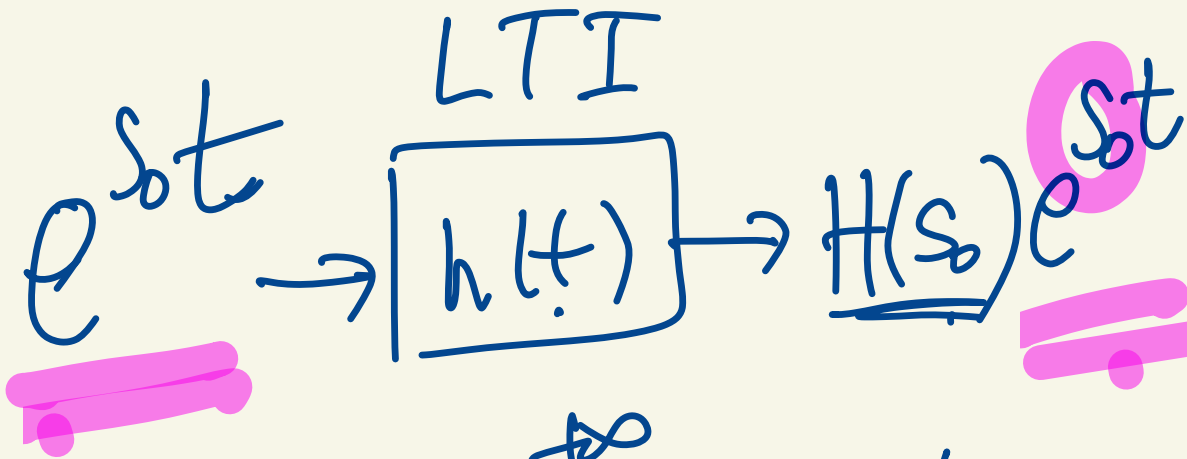
$$= \int_{-\infty}^{+\infty} h(\tau) e^{s_0(t-\tau)} d\tau$$

$$= \int_{-\infty}^{+\infty} h(\tau) e^{s_0 t} \cdot e^{-s_0 \tau} d\tau$$

$$= e^{s_0 t} \int_{-\infty}^{+\infty} h(\tau) e^{-s_0 \tau} d\tau$$

$$H(s_0)$$

$$= H(s_0) \cdot e^{s_0 t}$$



$$H(s_0) = \int_{-\infty}^{\infty} h(t) e^{-s_0 t} dt$$

$H(s_0)$ \rightarrow is scaling constant.

$e^{s_0 t}$ are eigenfunction
of LTI systems.

Why eigen?

Remember Linear
Algebra:

eigen vector

A is matrix

$$A \cdot \underbrace{x}_{\substack{\uparrow \\ \text{eigen vector}}} = \underbrace{\lambda}_{\substack{\uparrow \\ \text{eigen value}}} \underline{x}$$

$$\underbrace{x(t)} = \sum A_k \underbrace{e^{s_k t}}_{\rightarrow} \boxed{\begin{matrix} \text{LTI} \\ h(t) \end{matrix}} \rightarrow \dots$$

$$\rightarrow \sum A_k H(s_k) e^{s_k t}$$