

Announcements:

Problem session

Tonight 6pm MATX 1100.

(on Laplace Transforms).

The Convolution Integral

Theorem:

If $F(s) = \mathcal{L}\{f(t)\}$ & $G(s) = \mathcal{L}\{g(t)\}$

exist for some $s > a \geq 0$,

then

$$F(s) \cdot G(s) = \mathcal{L}\left\{\int_0^t f(t-\tau)g(\tau)d\tau\right\}$$

$$\underline{\text{and}} = \mathcal{L}\left\{\int_0^t f(\tau)g(t-\tau)d\tau\right\}.$$

We say

$$h(t) = \int_0^t f(t-\tau)g(\tau) d\tau \quad \text{and} \quad \int_0^t f(\tau)g(t-\tau) d\tau$$

is the "convolution integral"

of f & g .

Notation:

$$h(t) = \underline{f * g} = \int_0^t f(t-\tau)g(\tau) d\tau$$

and $\int_0^t \underline{g(\tau)f(t-\tau)} d\tau$.

Showing this: Let

$$F(s) = \mathcal{L}\{f(t)\} = \int_0^{\infty} f(u) e^{-su} du.$$

$$G(s) = \mathcal{L}\{g(t)\} = \int_0^{\infty} g(\tau) e^{-s\tau} d\tau$$

Then

$$F(s) \cdot G(s) = \left(\int_0^{\infty} f(u) e^{-su} du \right) \left(\int_0^{\infty} g(\tau) e^{-s\tau} d\tau \right)$$
$$= \int_0^{\infty} g(\tau) \left[\int_0^{\infty} f(u) e^{-s(u+\tau)} du \right] d\tau$$

For fixed τ ; inner integral is:

$$\int_0^{\infty} f(u) e^{-s(u+\tau)} du$$

Let $t = u + \tau$

so this integral becomes

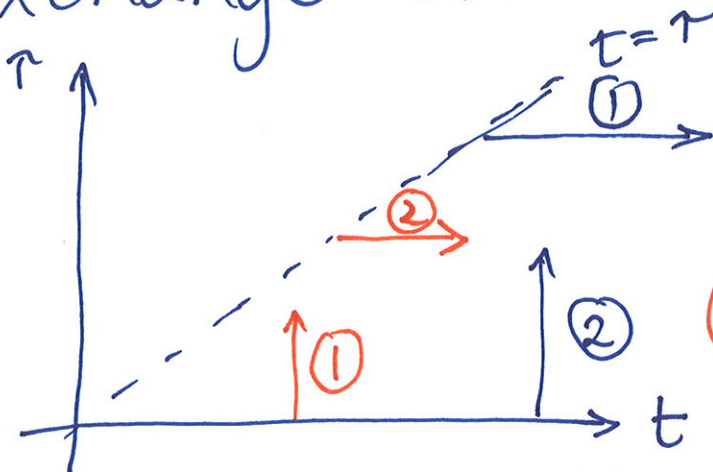
$$\int_{\tau}^{\infty} f(t-\tau) e^{-st} dt.$$

Putting back in:

$$F(s) \cdot G(s) = \int_0^{\infty} g(\tau) \left[\int_{\tau}^{\infty} f(t-\tau) e^{-st} dt \right] d\tau.$$

want: $\int_0^{\infty} (\quad) e^{-st} dt.$

Exchange order of integration.



integrate w.r.t.
 τ first:

① τ goes from 0 to t

then ② t goes from 0 to ∞

$$F(s) \cdot G(s) = \int_0^{\infty} \left(\int_0^t f(t-\tau) g(\tau) d\tau \right) e^{-st} dt.$$

(re-arranging).

$$= \mathcal{L} \left\{ \int_0^t f(t-\tau) g(\tau) d\tau \right\}$$

From def. of Laplace trans. □

This means if we want

$$\mathcal{L}^{-1} \{ F(s) G(s) \} = \int_0^t f(t-\tau) g(\tau) d\tau$$

where $\mathcal{L} \{ f \} = F(s)$, $\mathcal{L} \{ g \} = G(s)$.

Ex] Solve the IVP:

$$\begin{cases} y' = t^2 \\ y(0) = 0 \end{cases}$$

using L. transforms.

Still take transform of both sides: Let $Y(s) = \mathcal{L}\{y(t)\}$.

$$\mathcal{L}\{y'\} = \mathcal{L}\{t^2\} \quad (\text{our eq.})$$

$$\begin{aligned}\mathcal{L}\{y'\} &= sY(s) - y(0) \\ &= sY(s).\end{aligned}$$

$$\mathcal{L}\{t^2\} = \frac{2}{s^3}$$

$$\text{Then } sY(s) = \frac{2}{s^3}$$

$$Y(s) = \frac{2}{s^4}.$$

etc.

Instead we'll leave $\mathcal{L}\{t^2\}$ alone.

$$\underline{\text{So}} \quad sY(s) = \mathcal{L}\{t^2\}.$$

Solving for $Y(s)$:

$$Y(s) = \frac{1}{s} \cdot \mathcal{L}\{t^2\}.$$

Now we want to invert to get $y(t)$. $\mathcal{L}^{-1}\{Y\} = y(t)$.

$$\mathcal{L}^{-1}\left\{\frac{1}{s} \cdot \mathcal{L}\{t^2\}\right\}$$

Let $F(s) = \frac{1}{s}$, $G(s) = \mathcal{L}\{t^2\}$.

So we have $\mathcal{L}^{-1}\{F(s)G(s)\}$

which, using convolution integrals, is

$$\int_0^t f(t-\tau)g(\tau)d\tau.$$

where $f(t) = \mathcal{L}^{-1}\{F(s)\}$

$g(t) = \mathcal{L}^{-1}\{G(s)\}$

$$\text{Well, } f(t) = \mathcal{L}^{-1}\{F(s)\} = \mathcal{L}^{-1}\left\{\frac{1}{s}\right\}$$
$$= 1.$$

$$g(t) = \mathcal{L}^{-1}\{G(s)\} =$$
$$= t^2.$$

Then since $y(t) = \mathcal{L}^{-1}\{Y(s)\} = \mathcal{L}^{-1}\{F(s)G(s)\}$,

$$y(t) = \int_0^t \underbrace{f(t-\tau)}_{=1} \underbrace{g(\tau)}_{\tau^2} d\tau = \int_0^t \tau^2 d\tau$$
$$= \frac{\tau^3}{3} \Big|_0^t$$

$$\boxed{y(t) = \frac{t^3}{3}}$$

(OR $y(t) = \int_0^t \underbrace{f(\tau)}_{=1} \underbrace{g(t-\tau)}_{(t-\tau)^2} d\tau$)

$$= \int_0^t (t-\tau)^2 d\tau$$
$$= \frac{t^3}{3}.$$

Motivational Eg:

We had

$$Q(s) = \underbrace{\mathcal{L}\left\{\frac{1}{\beta} \sin \beta t\right\}}_{F(s)} \cdot \underbrace{\mathcal{L}\left\{\sin(\omega t)(u_{\pi}(t) - u_{2\pi}(t))\right\}}_{G(s)}.$$

Using convolution integrals,

$$q(t) = \mathcal{L}^{-1}\{Q(s)\} = \int_0^t \left(\frac{1}{\beta} \sin[\beta(t-\tau)]\right) \cdot \sin(\omega \tau) [u_{\pi}(\tau) - u_{2\pi}(\tau)] d\tau.$$

Much nicer than partial fractions expansion!

(Done out \rightarrow handout online)

Linear Systems (7.5-9).

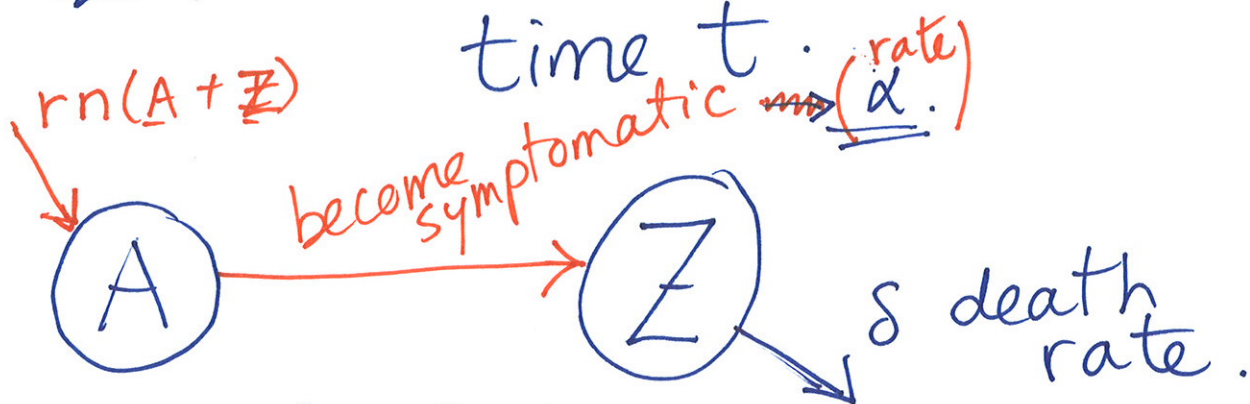
Motivating example.

(springs, complicated circuits in text).

Early Epidemic model.

Let $A(t)$ = # asymptomatics at time t .

$Z(t)$ = # ~~infectious~~ zombies at time t .



r = rate of infection

n = approx. const. population.

$$\begin{cases} \frac{dA}{dt} = rn(A+Z) - \alpha A \\ \frac{dZ}{dt} = \alpha A - \delta Z \end{cases}$$

System of 1st order ODEs
to understand zombie plague.

What can happen?

$(A, Z) = (0, 0)$ is

As $t \rightarrow \infty$, $A \text{ \& } Z \rightarrow \infty$ - unstable ~~node~~

OR $A \text{ \& } Z \rightarrow 0$ - stable.

Note: as A, I get large,
 $n \approx \text{const.}$ assumption ~~fails~~
invalid (nonlinear regime)