

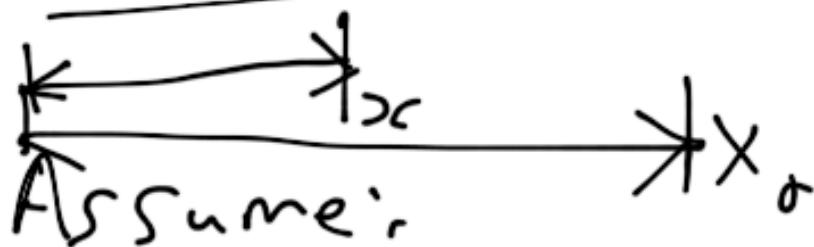
To-do:

- Sign up for pre-lab Jan 20.

Ch 1:

What is ctrl eng?

Ex: Automated highway.



Assume:

- car only moves in x -dir
- we can directly assign b .
- we have sensors to measure in front.

Car0 moves at the speed limit
 $v_0 > 0$

Obj: set the velo. of car1 &
maintain a safe dist $d_{safe} \rightarrow$
away from Car0





$$v = v_0 + k(d_{safe} - d)$$

in this example, any $k < 0$ works.

Ex: Queuing System.

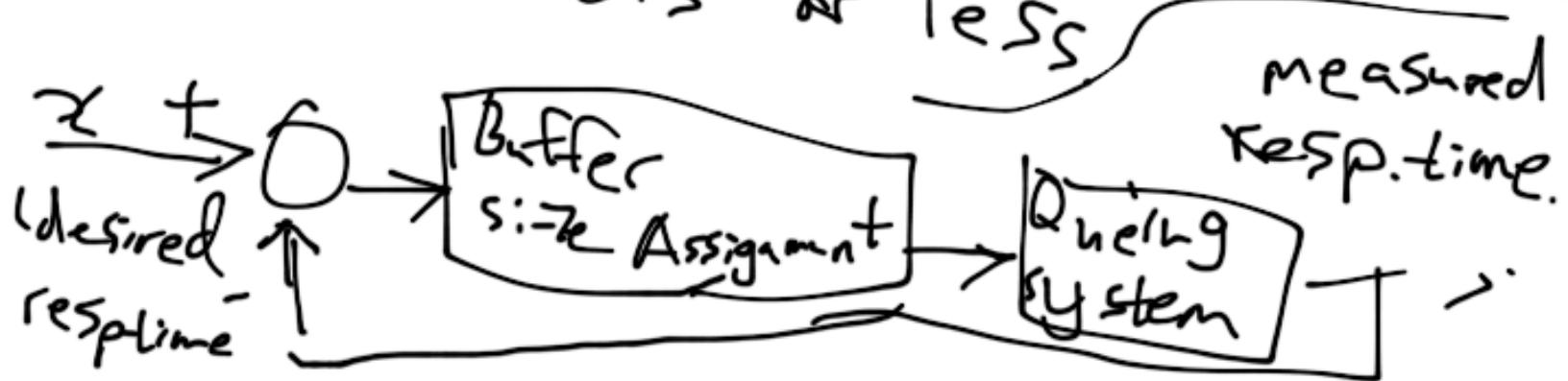


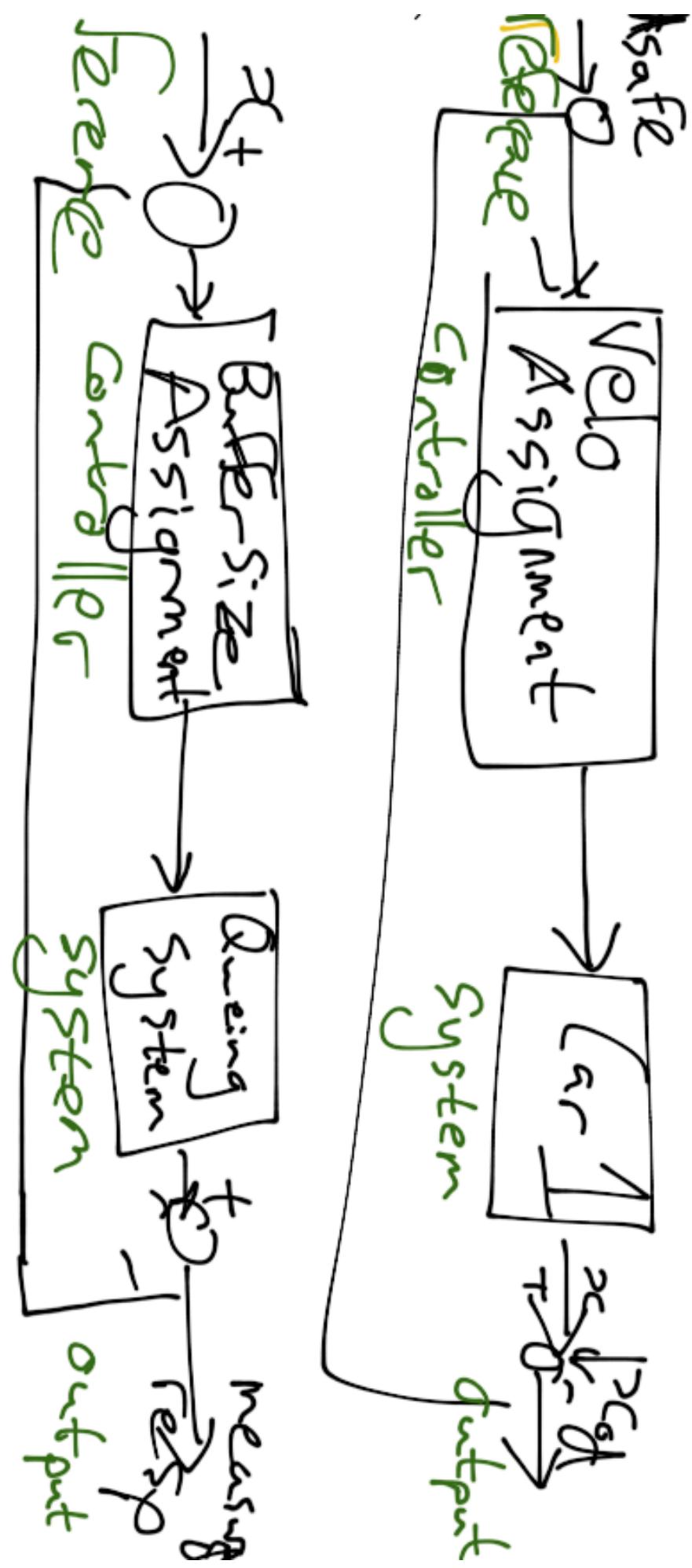
Assumption:

- Requests that don't fit buffer are redirected.
- two types of req's: buying & browsing.

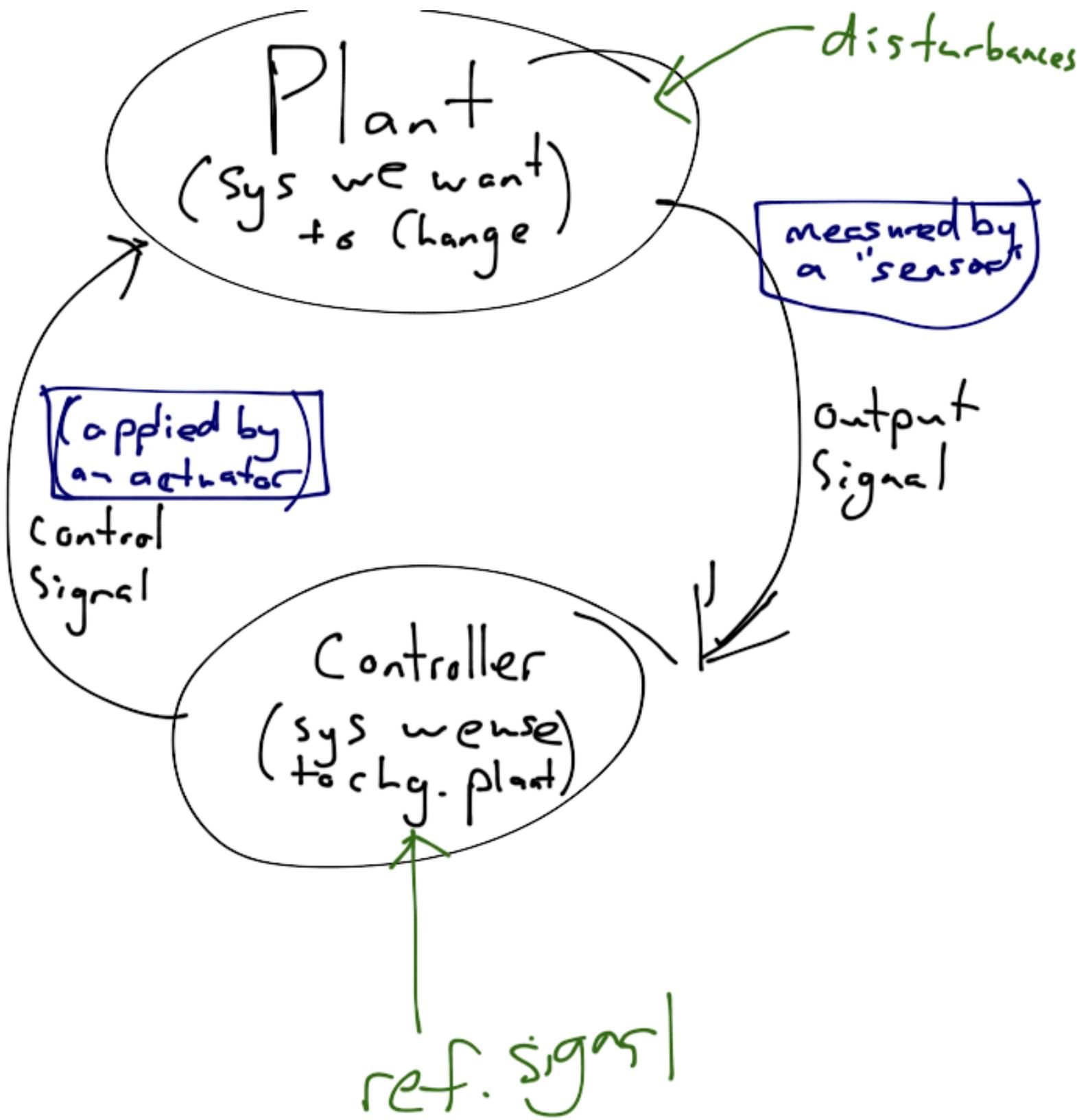
Obj: Assign the buffer size to max the # of completed requests while

ensuring "buying" req's are serviced in $x \geq 6$ seconds or less





System = plant | controller



Ctrl Eng Design Cycle:

1. get specs:

- closed-loop stability
- good steady-state tracking
- disturbance rejection
- transient performance.

2. Model the plant

- "model" = mathematical Model

- typically ≥ 1 differential eq'n's

$$\frac{dx(t)}{dt} = u(t)$$

- experiments are often used to determine the numerical values of plant params \rightarrow "system identification"

3. Obtain transfer function of plant

Classical control (this course, PID controllers) requires we have a transfer function for the plant.

e.g. car;

$$\mathcal{L}\left\{\frac{dx}{dt}\right\} = \mathcal{L}\left(\frac{U(s)}{s}\right) \Rightarrow sX(s)$$
$$\Rightarrow \frac{X(s)}{U(s)} = \frac{1}{s} = P(s)$$

$P(s)$ is the transfer function of the plant.

N.B.: For discrete time, use Z-transforms instead (not on course, but...)

NOT ON COURSE



4. Design the controller:

- Controller is also a transfer fn:
- It corresp. to the D.E. that relates the control signal to the output signal.

Can check PID controller:

$$C(s) = \frac{-2}{s} - \frac{3}{s^2} - \frac{s}{s+0.1}$$

5. Simulation

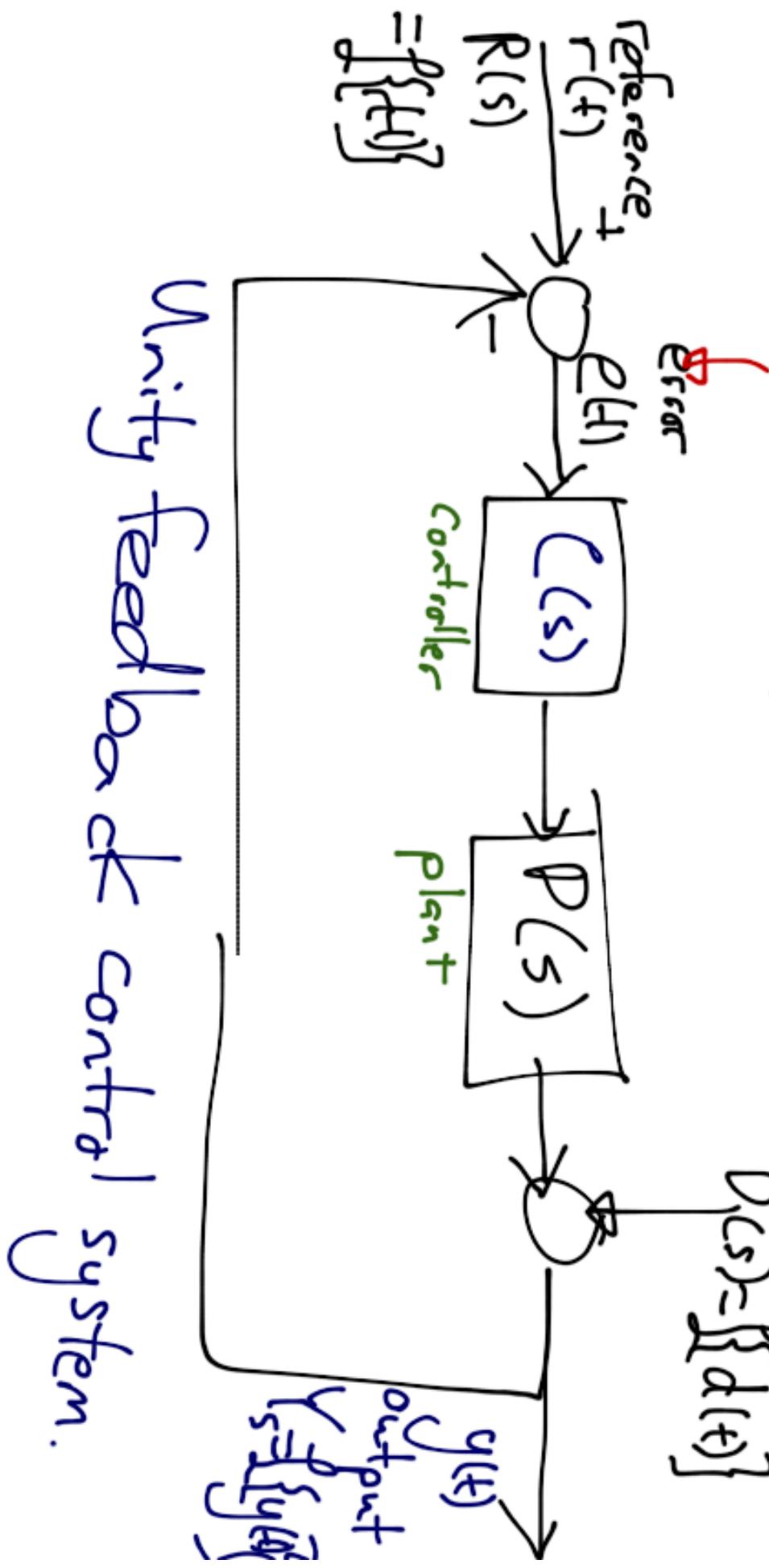
6. Implement Controller

- Can build a sys w/ transfer fn matches the one from step 4.
- Realistically the controller is implemented on a computer as a difference eq'n.
(ECE 481)

Basic Unity Feedback

$$E(s) = \{ \{ e(t) \} \}$$

$$D(s) = \{ \{ d(t) \} \}$$



Ch 2: Mathematical Models of Systems

For controller design we want
a "good" model \Rightarrow simple & effective

P.2: Comments on Modelling

- A model is a set of equations used to represent a physical system.
 \Rightarrow Never perfect
- models for design are usually simpler rather than accurate. But most are tradeoff.
- There are two modelling approaches:
 - \rightarrow statistical (experimental)
 - \rightarrow Analytical (science+theory)
- Use modular approach - break pblm into sub-sections



Apply "physics"

linearization

"operating" plant

Laplace

System of algebraic eqns.

Virtually all sys
are linear

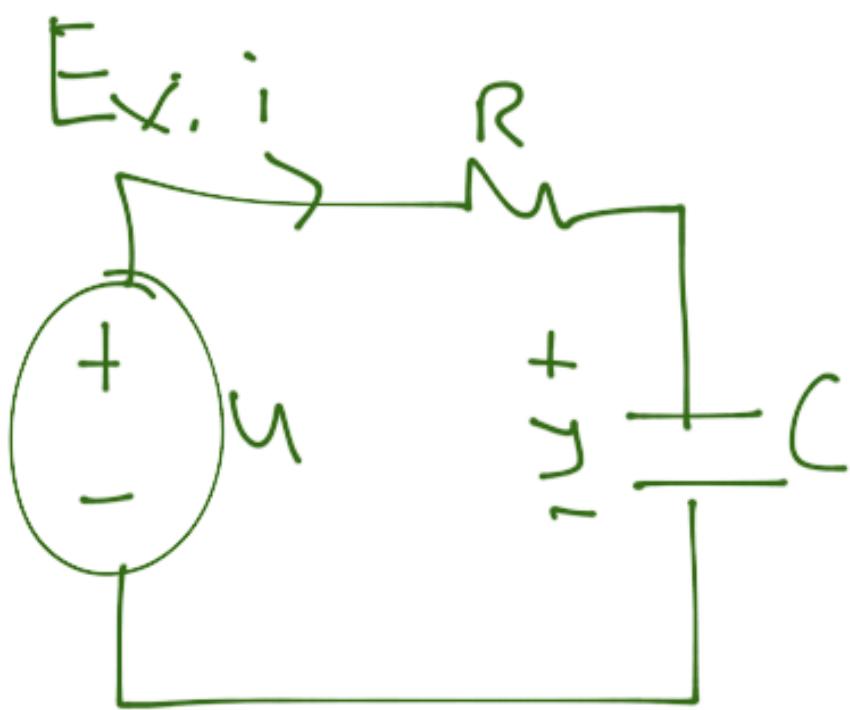
of linear
Odes

Solve for relationship
between T/θ

Experimentally
det. params in \sqrt{T}

F_h.

Faster



RC Circuit.

Ways to model Systems:

(a) Linear ODE:

$$-u + V_R + V_C = 0$$

$$V_R = R_i, \quad V_C = y, \quad i = C \frac{dy}{dt}$$

$$\Rightarrow -u + RC \frac{dy}{dt} + y = 0$$

$$\dot{y} = \frac{dy}{dt} \quad (\text{derp})$$

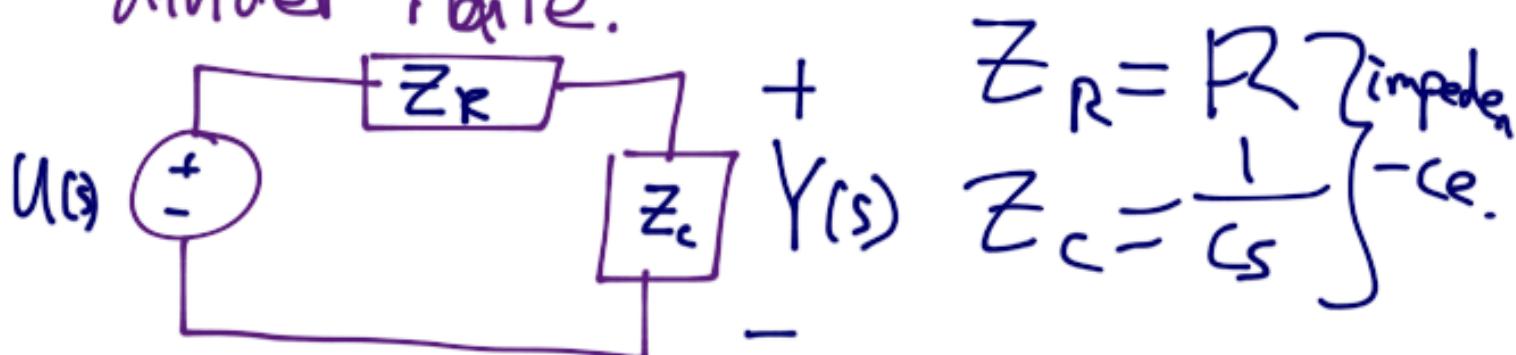
Transfer Fn:

i) L.T of CDE (assume 0; initial cond)

$$-U(s) + sRCY(s) + Y(s) = 0$$

$$\Rightarrow \frac{Y(s)}{U(s)} = \frac{1}{RCs + 1}$$

ii) use impedance and the voltage divider rule.



$$\left. \begin{aligned} Z_R &= R \\ Z_C &= \frac{1}{Cs} \end{aligned} \right\} \begin{array}{l} \text{impedance} \\ \text{-ce.} \end{array}$$

$$Y(s) = \frac{Z_C}{Z_C + Z_R} U(s) \quad (\text{voltage divider})$$

$$\Rightarrow Y(s) = \frac{1}{RCs + 1} U(s)$$

Circuits Review:

$$V = \frac{I}{R}$$

$$\text{KCL: } \sum V_{\text{loop}} = 0$$

$$\text{KVL: } \sum I_{\text{node}} = 0$$

	Voltage	Current	Laplace
 $\int I dt$	$v(t)$	$C \frac{dv}{dt}$	$V(s) = \frac{I(s)}{sC}$
 $\frac{d}{dt} \int v dt$		$I(t)$	$V_L(s) = sL I(s)$
	IR	$\frac{V}{R}$	$V_R(s) = R I(s)$

self-added

C) Convolution:

$$\text{Let } G(s) = \frac{1}{R(s+1)}$$

$$\text{then } g(t) = \mathcal{L}^{-1}\{G(s)\} = \frac{1}{RC} e^{\frac{-t}{RC}}$$

the whole sys output can be expressed as:

$$y(t) = g(t) * u(t) = \int_0^t g(t-\tau)u(\tau)d\tau$$

Convolution

(d) Bode Plot:

→ next chapter

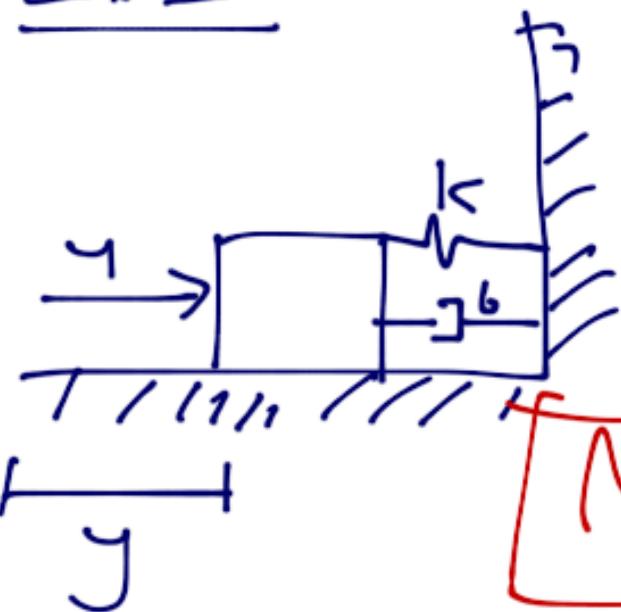
(e) state model:

→ 1+ 1st order ODEs.

$$\begin{aligned} \dot{x} &= \frac{-1}{RC}x + \frac{1}{RC}u \\ y &= x \end{aligned}$$

(x = voltage on capacitor)

Ex 1:



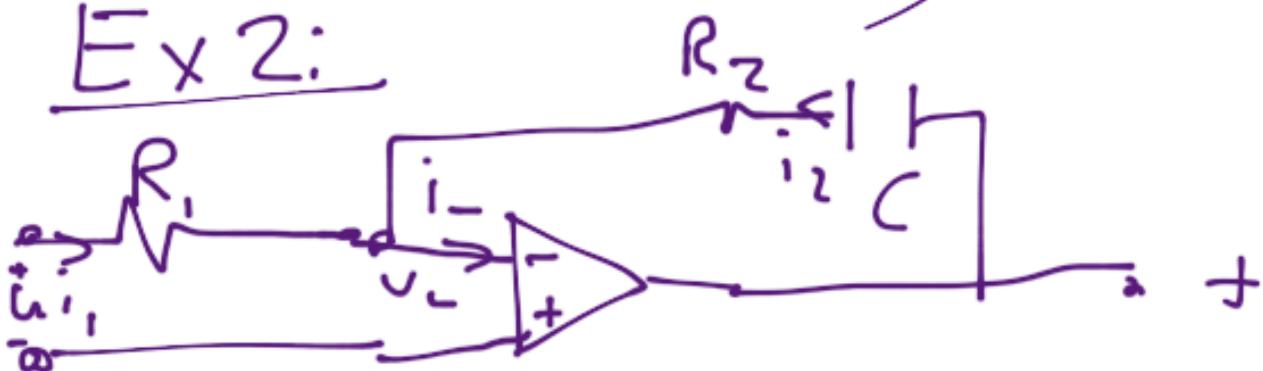
Newton's

$$M_{ij} \leq \sum \text{Forces}$$

$$= u - F_{\text{spring}} - F_{\text{damper}}$$

$$\boxed{M_{ij} = u - ky - by}$$

Ex 2:



Ideal op-amp

$$i_- = 0, v_- = v_+$$

KCL:

using impedances:

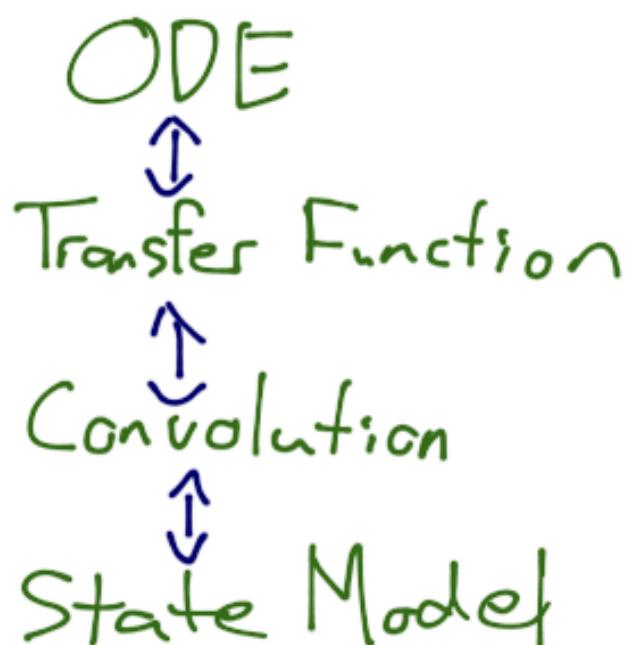
$$U(s) - V_-(s) = \frac{V_1 - V_s}{Z_{R_1} + Z_C}$$

$$\Rightarrow \frac{Y(s)}{U(s)} = -\frac{R_2}{R_1} - \frac{1}{R_C s}$$

Summary:

The main point of Modeling is to obtain an approx model of the system through **analysis & experiments**.

A system can be expressed in multiple ways:



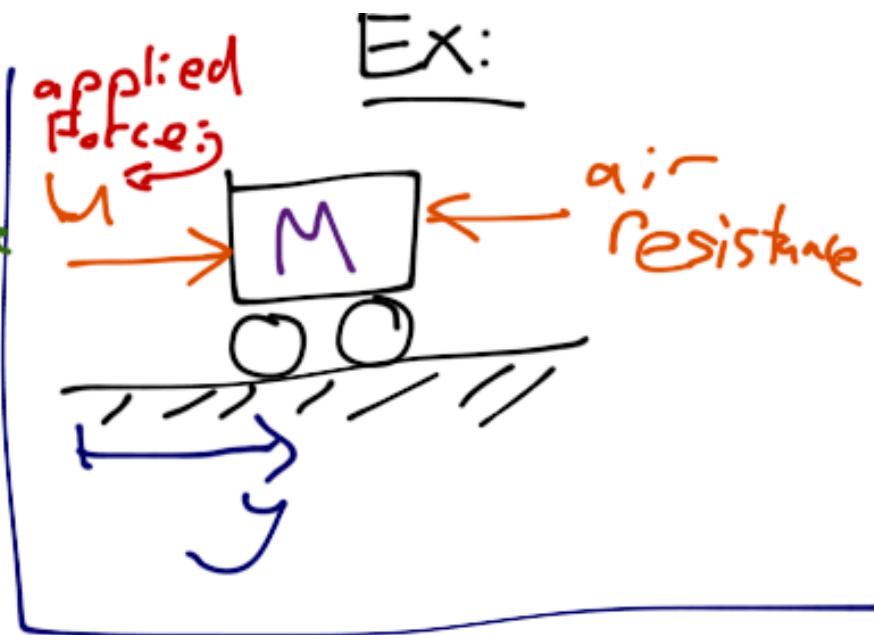
State x at time $t=t_0$, $x(t_0)$
encapsulates all the system
dynamics up to time t_0 .

For any times $t_0 \leq t \leq t_1$, $t_0 \leq t_1$,
knowing $x(t_0)$ and knowing
the applied control

$\{u(t) : t_0 \leq t \leq t_1\}$ we can
compute $x(t_1)$ and hence $y(t_1)$

State Models:

Typically air resistance creates an air force that depends on \dot{y} . Let's say this force is a possibly non-linear function $D(\dot{y})$
eg: $D(\dot{y}) = \dot{y}^2$



Newton's 2nd Law: $M\ddot{y} = u - D(\dot{y})$

Can we find a Transfer Function?
No. Not if $D(\dot{y})$ is non-linear.

Let's Def the State Variables:

$$x_1 = y, \quad x_2 = \dot{y}, \quad \begin{cases} x = [x_1] \\ \dot{x}_2 \end{cases} = \begin{cases} \text{Position} \\ \text{velocity} \end{cases}$$

state of system

System Of equations:

$$\dot{x} = \begin{bmatrix} x_2 \\ \frac{1}{M}u - \frac{1}{M}D(x_2) \end{bmatrix} \quad \begin{cases} \text{state} \\ \text{equations} \end{cases}$$

Sometimes "u" too.

$y = x_1$

\{ output eq'n

of the form $\dot{x} = f(x, u), y = h(x)$

Ideas; in the case that $D(x_2)$ is not linear, we define a system of equations that we can solve.

System of eqns: $\begin{aligned} \dot{x} &= \begin{bmatrix} y \\ \dot{y} \end{bmatrix} = f(x, u) \\ y &= h(x, u) \end{aligned}$

↑
const
(not a
function)

useful for the case that

y^a (or others) have $a > 1$ L doesn't work
for these functions, and we need to
do it this way.

In the case that $D(x_2)$ is linear,
e.g.: $D(x_2) = dx_2$ (d is a const)

then,

$$\dot{x} = f(x, u) = \begin{bmatrix} 0 & 1 \\ 0 & \frac{-d}{M} \end{bmatrix} x + \begin{bmatrix} 0 \\ \frac{1}{M} \end{bmatrix} u$$

constant matrix "A"

constant matrix "B"

$$\text{so: } f(x, u) = Ax + Bu = \dot{x}$$

In general, we have:

$$\boxed{\begin{aligned}\dot{x} &= Ax + Bu \\ y &= Cx\end{aligned}}$$

$$C = [0 \ 1];$$

Linear, Time
invariant state
mode

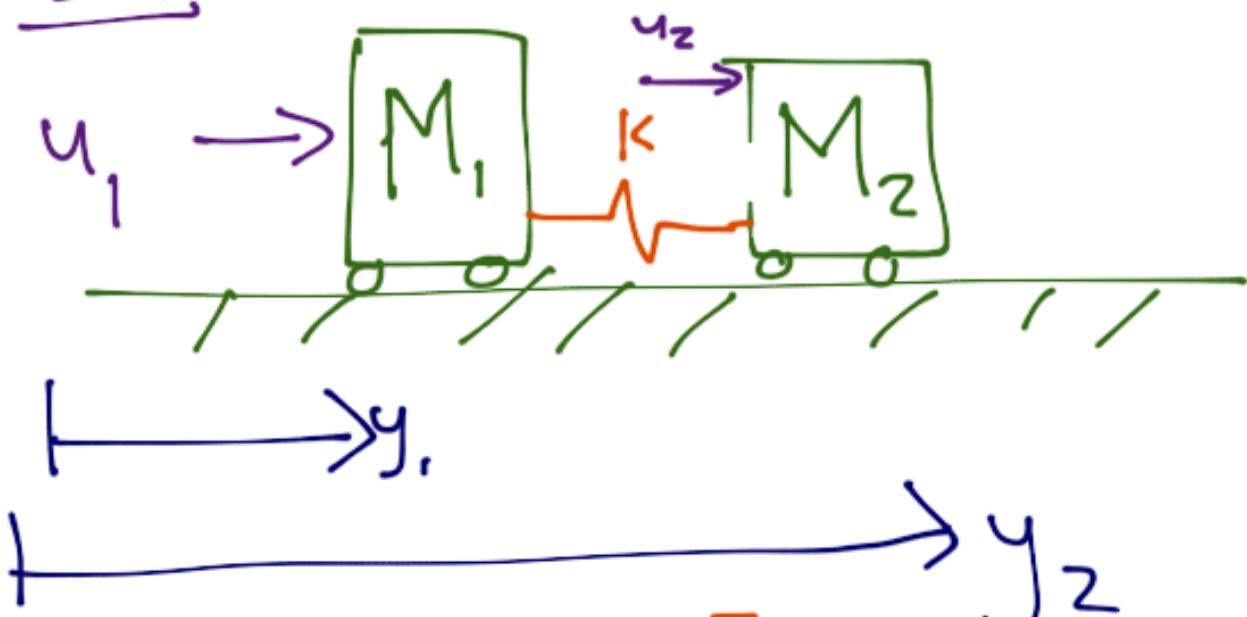
$$\text{if } \begin{bmatrix} a & b \\ c & d \end{bmatrix} x = \begin{bmatrix} ax_1 + bx_2 \\ cx_1 + dx_2 \end{bmatrix}$$

The idea is that many states have the model

$$\dot{x} = f(x, u), \quad f: \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n$$
$$y = h(x, u), \quad h: \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^p$$

output (y) dimension p
input (u) dimension m
state (x) dimension n

Ex:



$$u = (u_1, u_2) \in \mathbb{R}^2 \Rightarrow m = 2$$

$$y = [y_1, y_2] \in \mathbb{R}^2 \Rightarrow p = 2$$

$$x = [y_1, \dot{y}_1, y_2, \dot{y}_2] \in \mathbb{R}^4 \Rightarrow n = 4$$

the linear time-invariant special-case:

$$\dot{x} = Ax + Bu \quad A \in \mathbb{R}^{n \times n} \quad B \in \mathbb{R}^{n \times m}$$

$$y = Cx + Du \quad C \in \mathbb{R}^{p \times n} \quad D \in \mathbb{R}^{p \times m}$$

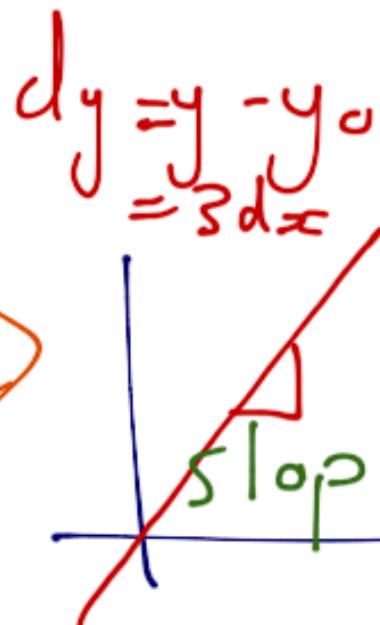
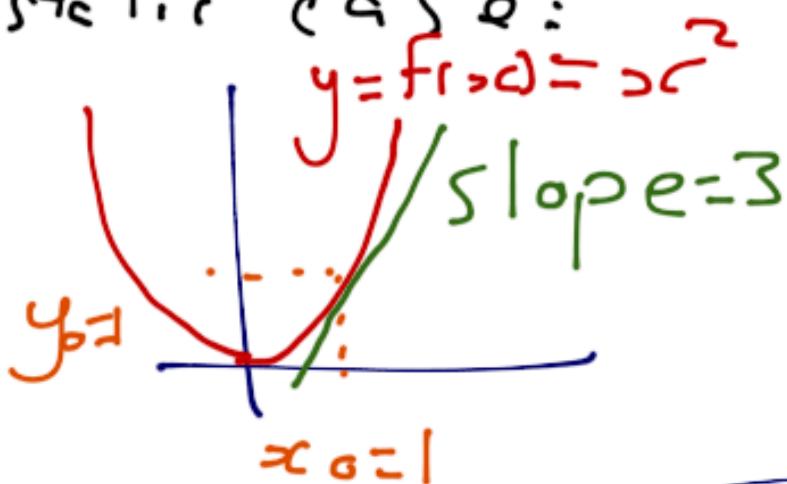
in this course we only deal w/
time-invariant systems.

Summary (Add JWei's notes here)

- States:

- linearization

- static case:



- vector case —

$$dy = \left. \frac{dF}{dx} \right|_{x=x_0} dx$$

- Apply this to: $\dot{x} = f(x, u)$

$$y = h(x, u)$$

First assume (x_0, u_0) is an equilibrium

$$f(c_0, u_0) = 0$$

$$f(x, u) = f(x_0, u_0) + \frac{df}{dx} \Big|_{x=x_0, u=u_0} (x - x_0) + \frac{df}{du} \Big|_{x=x_0, u=u_0} (u - u_0)$$

higher order terms.

Now consider a solution "near" the constant solution

$$x(t) = x_0 + \int x(t)$$

$$u(t) = u_0 + \int u(t)$$

$$\dot{x} = \frac{d(x(t))}{dt} = \frac{d x_c(t)}{dt} - \frac{d(x_o)}{dt} \\ = f(x, u) - 0$$

$$\approx f(x_0, u_0) + A\delta x + B\delta u$$

$$= Adx + B dy$$

$$\Rightarrow \delta x = A \delta x + B \delta u$$

We can linearize the output eq'n as in the previous example.

$$\delta y = \left. \frac{\delta h}{\delta x} \right|_{\substack{x=x_0 \\ u=u_0}} \delta x + \left. \frac{\delta h}{\delta u} \right|_{\substack{x=x_0 \\ u=u_0}} \delta u$$

C D

Summary: Linearization

1) Select, if one exists an equilibrium (x_0, u_0)

2) Compute A, B, C, D (Jacobians)

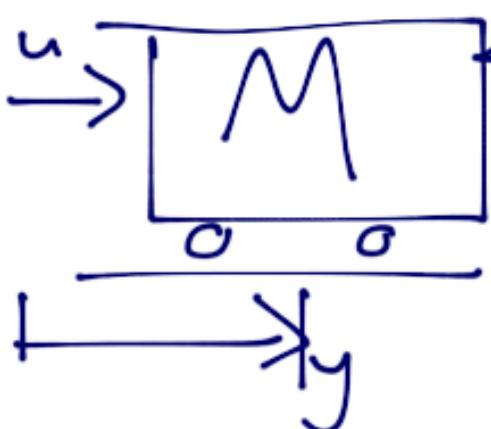
3) The linearized system is:

4) Under mild conditions
the linearization

$$\begin{aligned}\delta \dot{x} &= A \delta x + B \delta u \\ \delta y &= C \delta x + D \delta u\end{aligned}$$

is a good approx for "small" δx and δu .

Ex:



$$D(y)$$

Cart example @
position $y_0 = 10$

$$D(y) = Ky^2, K > 0$$

i) The model is $\dot{x} = f(x, u) = \begin{cases} x_2 \\ -Kx_2^2 + u \end{cases}$
 we need an operating point (x_0, u_0) corresponding to $y = x_1 = 10$.

$$\text{Solve: } \begin{bmatrix} 0 \\ 0 \end{bmatrix} = f(x_0, u_0) = \begin{bmatrix} x_{20} \\ -\frac{Kx_{20}^2}{M} + \frac{u_0}{M} \end{bmatrix}$$

$$\Rightarrow x_{20} = 0, u_0 = 0$$

$$\Rightarrow (x_0, u_0) = \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, 0 \right)$$

$$z) A = \frac{\delta f}{\delta x} \Big|_{\substack{x=x_0 \\ u=u_0}} = \begin{bmatrix} 0 & 1 \\ 0 & \frac{-2Kx_2}{M} \end{bmatrix} \Big|_{\substack{x=x_0 \\ u=u_0}} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$B = \frac{\delta f}{\delta u} \Big|_{\substack{x=x_0 \\ u=u_0}} = \begin{bmatrix} \frac{\delta f_1}{\delta u} \\ \frac{\delta f_2}{\delta u} \end{bmatrix} \Big|_{\substack{x=x_0 \\ u=u_0}} = \begin{bmatrix} 0 \\ \frac{1}{M} \end{bmatrix}$$

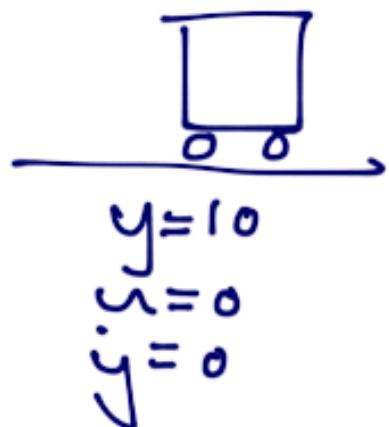
$$C = \frac{\delta h}{\delta x} \Big|_{\substack{x=x_0 \\ u=u_0}} = \begin{bmatrix} \frac{\delta h}{\delta x_1} & \frac{\delta h}{\delta x_2} \end{bmatrix} \Big|_{\substack{x=x_0 \\ u=u_0}} = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

$$D = \frac{\delta h}{\delta u} \Big|_{\substack{x=x_0 \\ u=u_0}} = 0$$

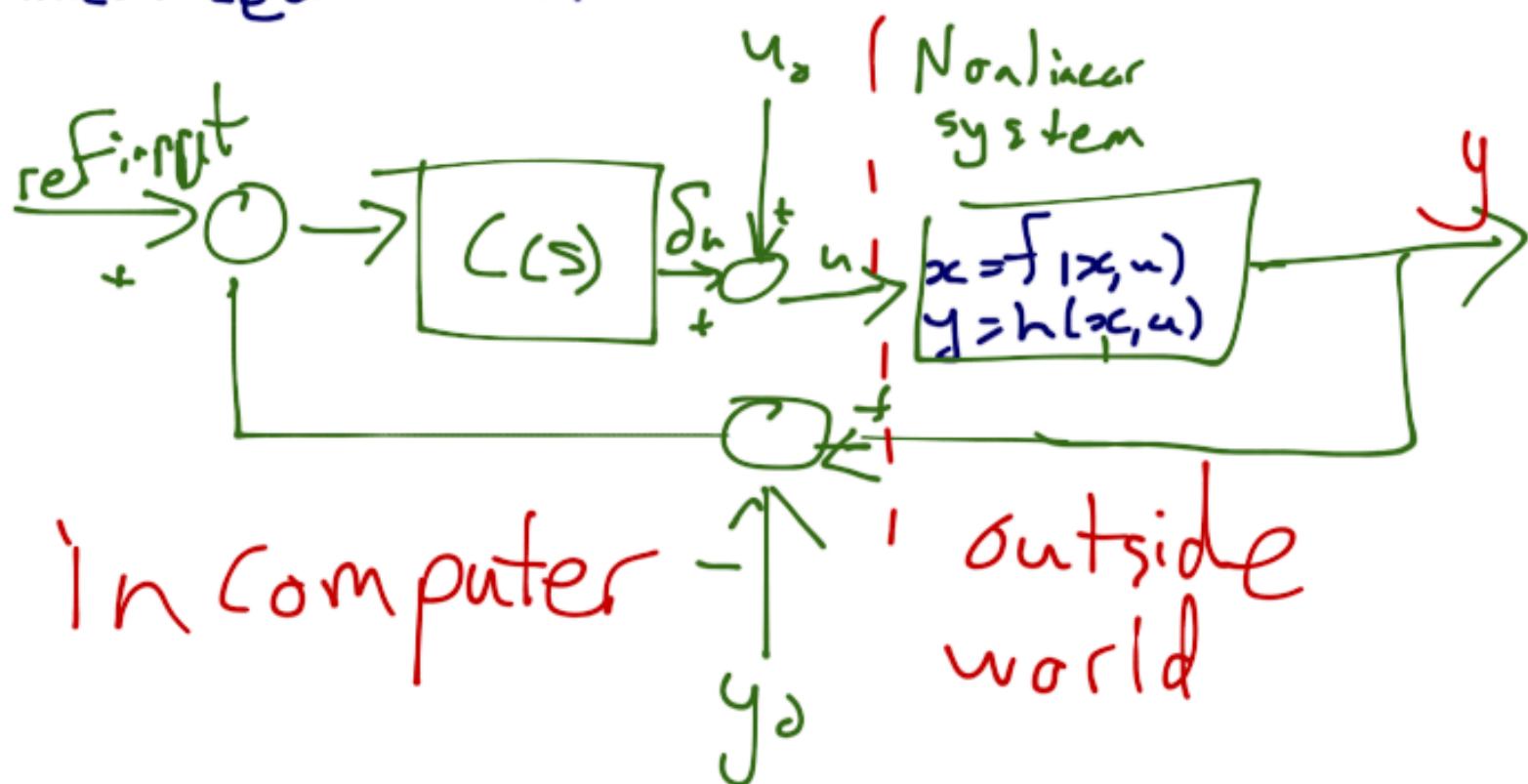
3) Linear Model

$$\delta \dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \delta x + \begin{bmatrix} 0 \\ \frac{1}{M} \end{bmatrix} \delta u$$

$$\delta y = \begin{bmatrix} 1 & 0 \end{bmatrix} \delta x$$



Implementing a controller designed using a linearized model:



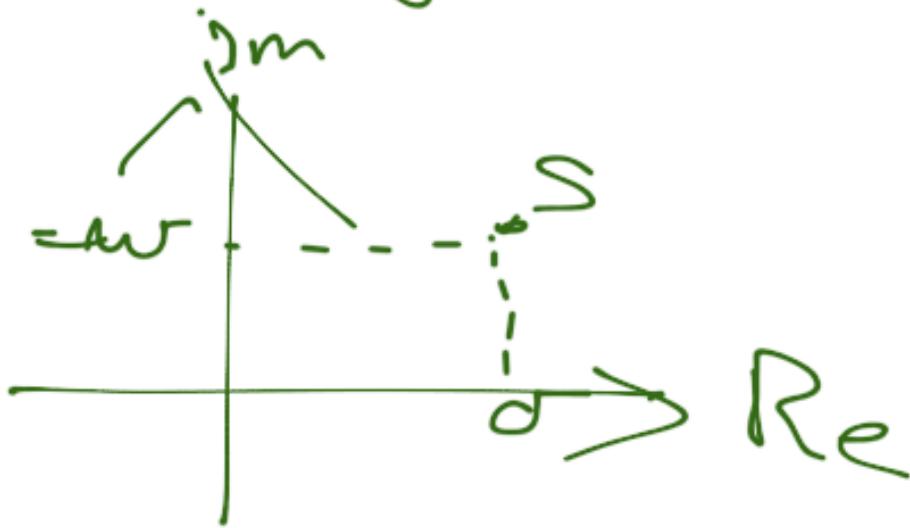
2.8 Laplace transform

Let $f(t)$ be a signal (function)

def'd for $t \geq 0$

$$F(s) = \int \{f(t)\} = \int_0^{\infty} f(t) e^{-st} dt$$

here, $s = \sigma + j\omega$



Intuition:

$F(s)$ is a decomposition of $f(t)$ into a weighted sum of complex exponentials

$$e^{-st} = e^{-\sigma t} (\cos \omega t - j \sin \omega t)$$

Summary

- linearize state models

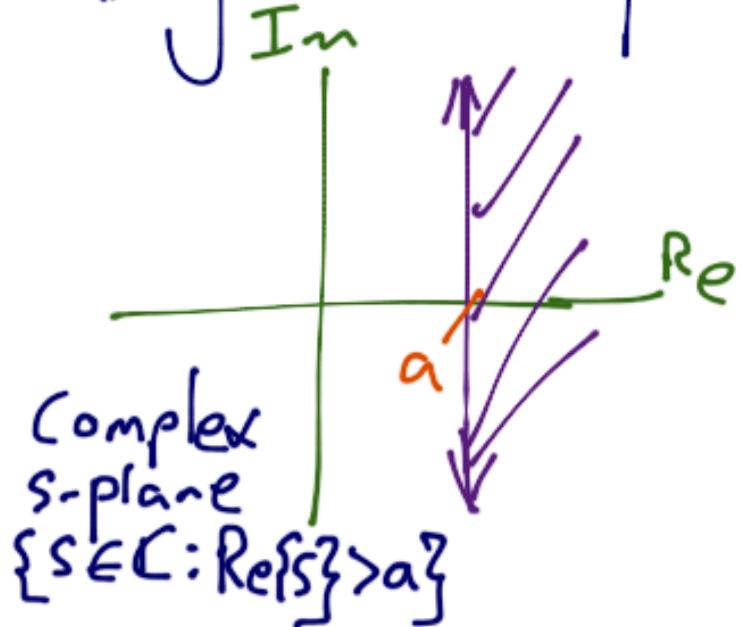
$$\begin{aligned}\dot{x} &= f(x, u) \\ y &= h(x, u)\end{aligned}\rightarrow \begin{aligned}\delta \dot{x} &= A\delta x + B\delta u \\ \delta y &= C\delta x + D\delta u\end{aligned}$$

- Approximation is valid near the operating points (x_o, u_o)

$$f(x_o, u_o) = 0$$

- Laplace Transform: $F(s) = \mathcal{L}\{f(t)\} = \int f(t)e^{-st} dt$
 There's a region of convergence for some s . Where the region converges.
 R.O.C.

The Region of convergence is in the right-half plane:



$$f(t) = e^{-at} \Rightarrow \frac{1}{s+a}$$

$$\Rightarrow \text{ROC} = \text{Re}\{s\} > a$$

$$f(t) = te^{-at} \Rightarrow \frac{1}{(s+a)^2}$$

$$\Rightarrow \text{ROC} = \text{Re}\{s\} > a$$

$f(t) = e^{t^2} \Rightarrow \text{ROC is empty} \Leftrightarrow \text{there is no } \mathcal{L}\{f\}$

L + L{f} a is const, f, g are fns.

i) $\mathcal{L}\{f+g\} = \mathcal{L}\{f\} + \mathcal{L}\{g\}$ } linear

ii) $\mathcal{L}\{af\} = a\mathcal{L}\{f\}$

iii) $\mathcal{L}\left\{\frac{df}{dt}\right\} = s\mathcal{L}\{f\} - f'(0)$ } derivative

iv) $\mathcal{L}\{f * g\} = \mathcal{L}\{f\} \mathcal{L}\{g\}$ } convolution

v) $\mathcal{L}\left\{\int_0^t f(\tau) d\tau\right\} = \frac{1}{s} \mathcal{L}\{f\}$ } integral

vi) if $\lim_{t \rightarrow \infty} f$ exists & is finite, then

$$\lim_{t \rightarrow \infty} f = \lim_{s \rightarrow 0} sF(s)$$

Final Value
Theorem (FVT)

Ex: 

$\frac{f(t)}{e^{-t}}$	$\lim_{t \rightarrow \infty} f(t)$	$\{f\}$	$\lim_{s \rightarrow 0} sF(s)$	FVT Applies?
e^{-t} \downarrow $f(t)$ heavy side	0	$\frac{1}{s+1}$	0	✓
$t e^{-t}$	1	$\frac{1}{s}$	1	✓
e^t const \downarrow coswt	∞	$\frac{1}{s-1}$	0	✗
coswt	DNE	$\frac{s}{s^2 + \omega^2}$	0	✗

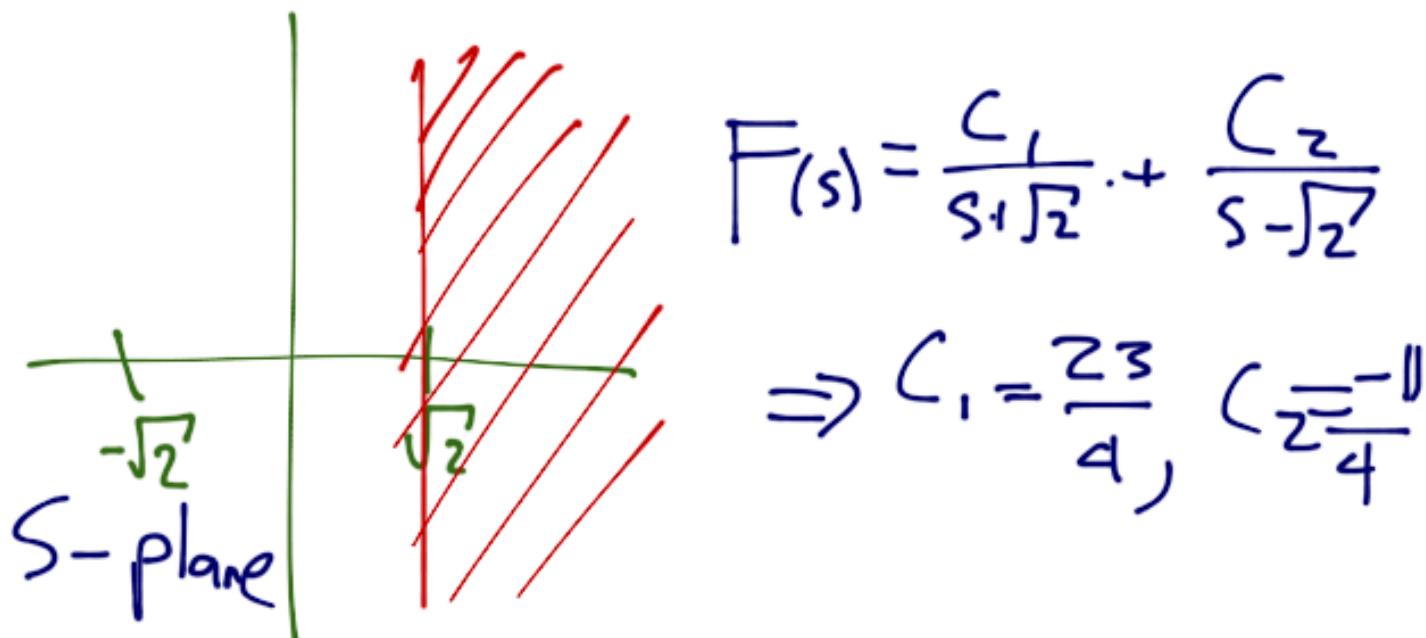
The main idea is that if $sF(s)$ is defined (no div by 0) for any s with $\text{Re}\{s\} > 0$, the FVT applies.

If $s = -j\omega$, the $\{f\}$ dne, and we cannot apply the final value theorem.

We can use PFE (partial fraction expansion) when

$$F(s) = \frac{r(s)}{d(s)} \quad \left\{ \begin{array}{l} \text{Polynoms} \\ \rightarrow \deg(d) > \deg(r) \end{array} \right.$$

Eg: $F(s) = \frac{3s+17}{s^2-2}$ find $f(t)$



$$f(t) = \left\{ \left\{ \frac{C_1}{s+\sqrt{2}} \right\} \right\} + \left\{ \left\{ \frac{C_2}{s-\sqrt{2}} \right\} \right\}$$

$$= \frac{23}{4} e^{-\sqrt{2}t} - \frac{11}{4} e^{\sqrt{2}t}, \quad t \geq 0$$

Ey: $F(s) = \frac{s+1}{s(s+2)^2} \Rightarrow f(t) = ?$ [2.3]

$$F(s) = \frac{C_1}{s} + \frac{C_2}{s+2} + \frac{C_3}{(s+2)^2} = \frac{s+1}{s(s+1)^2}$$

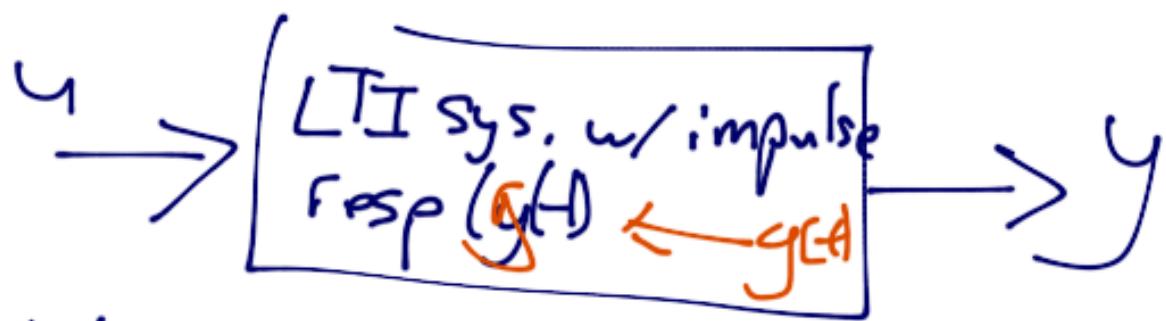
$$\left. \begin{array}{l} C_1 = \frac{1}{4} \\ C_2 = -\frac{1}{4} \\ C_3 = \frac{1}{2} \end{array} \right\} f(t) = \frac{1}{4} + -\frac{1}{4}e^{-2t} + \frac{1}{2}te^{-2t}$$

The idea is that L⁻¹ are useful for solving initial value functions, like

$$\dot{y} - 2y = t, y(0) = 1$$

2.4 Transfer Functions:

Consider a linear time-invariant system (LTI) with response $y(t)$



You should have seen

$$y(t) = g(t) * u(t)$$

$X(s) \rightarrow$ the height from the
ground. \rightarrow input!

5

V