

Control theory: state-space approach for linear systems

Preliminary review

2 janvier 2022

Dynamical model and system

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- ▶ *linear*, if the *superposition principle* applies :

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- ▶ *stationnary* if the relations between the input and the output remain the same as time ellapses (no variation of system- parameters).
- ▶ *causal* if system output at any time instant t_0 , $y(t_0)$, does not depend on future values of input $u(t)$, $t > t_0$ (all the physical systems are causal).

Example : container handling gantry crane



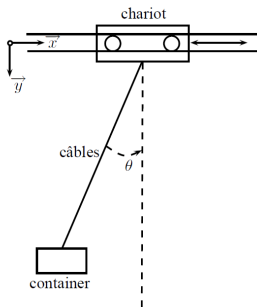
FIGURE – container handling gantry crane

We focus on horizontal shifting of set trolley-cables-container (lifting is stopped).

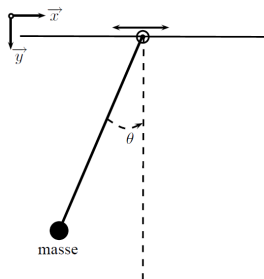
We aim at **regulating the container speed while avoiding the container to swing (dangling).**

Example : container handling gantry crane

- ▶ Set trolley-cables-container modeled as a pendulum with a mobile pivot
- ▶ Cables \equiv indeformable solid, in pivot-connection with the trolley
- ▶ Container \equiv punctual mass placed in the center of gravity
- ▶ Frictions are neglected

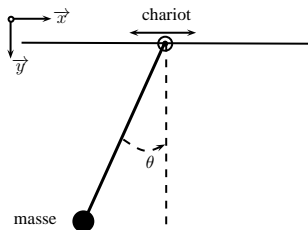


Schématisation de l'ensemble chariot-câbles-container



Abstraction sous la forme d'un pendule

Example : container handling gantry crane



m	mass of the container
L	cables length (\equiv indeformable solid)
g	gravity acceleration

p_m, v_m, a_m	position, speed and acceleration of the mass projected onto axis \vec{x}
p_c, v_c, a_c	position, speed and acceleration of the trolley projected onto axis \vec{x}
θ	angle between cables and the vertical axis

Example : container handling gantry crane

The goal is to regulate the mass-speed while avoiding it to swing. We consider that :

- ▶ the **input** of the system is the horizontal speed of the trolley v_c (at any time instant t , we assume to be able to set the value of $v_c(t)$);
- ▶ the **output** of the system is the horizontal speed v_m of the mass (container).

We neglect **disturbances (noises)** influencing the system : we could take into account the wind (uncontrolled exogeneous quantity influencing the container-dangling).

This system is **deterministic** : an input v_c leads to only one possible output v_m .

This system is **causal** : the output value at a time instant t_0 , $v_m(t_0)$, does not depend on any future value of the input $v_c(t)$ for $t > t_0$.

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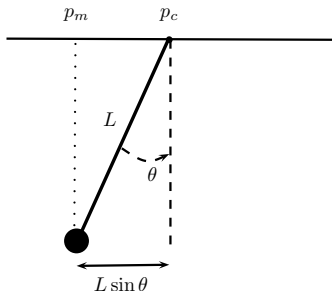
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We have $p_c = p_m + L \sin \theta$, hence the trolley speed is related to the mass-speed by :

$$v_c = v_m + L \frac{d \sin \theta}{dt}. \quad (1)$$



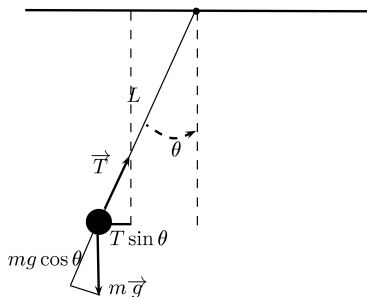
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Fundamental principle of the dynamics (Newton's second law) applied to the mass and projected onto \vec{x} gives : $ma_m = T \sin \theta$.

Cables considered as indeformables $\Rightarrow T = mg \cos \theta$.

We deduce that

$$ma_m = mg \sin \theta \cos \theta \quad (2)$$



Example : container handling gantry crane

System modeled by the equations

$$v_c = v_m + L \frac{d \sin \theta}{dt}$$
$$m a_m = m g \sin \theta \cos \theta$$

- ▶ **stationnary** since the parameters in these differential equations (that is m , g and L) are assumed to be constant as time ellapses.
- ▶ **not linear**.

Example : container handling gantry crane

System modeled by

$$v_c = v_m + L \frac{d \sin \theta}{dt} \quad ma_m = mg \sin \theta \cos \theta$$

To have a linear model, we assume approximations on the behavior of the system (less precise model but easier to use since linear) : we consider that $\theta \approx 0$ and so $\cos \theta \approx 1$, $\sin \theta \approx \theta$. We deduce the following linear model :

$$\frac{d\theta}{dt} = -\frac{1}{L}v_m + \frac{1}{L}v_c \quad (3)$$

$$\frac{dv_m}{dt} (= a_m) = g\theta \quad (4)$$

Reminders about linear systems

	CONTINUOUS TIME
Convolution	$y(t) = (h * u)(t)$ $= \int_0^t h(\tau)u(t - \tau)d\tau$
Transfer F.	$H(s) = \frac{b_m s^m + b_{m-1} s^{m-1} + \dots + b_1 s + b_0}{s^n + a_{n-1} s^{n-1} + \dots + a_1 s + a_0}$ $\mathcal{L}[h(t)] = H(s)$ $Y(s) = H(s)U(s)$
Diff. Equ.	$y^{(n)} + a_{n-1}y^{(n-1)} + \dots + a_1\dot{y} + a_0y$ $= b_m u^{(m)} + b_{m-1}u^{(m-1)} + \dots + b_1\dot{u} + b_0u$ <p>C.I. $y(0^+), \dot{y}(0^+), \dots, y^{(n-1)}(0^+)$</p>

Example : container handling gantry crane

Laplace transform of equations

$$\frac{d\theta}{dt} = -\frac{1}{L}v_m + \frac{1}{L}v_c, \quad \frac{dv_m}{dt} (= a_m) = g\theta,$$

is

$$s\Theta(s) = -\frac{1}{L}V_m(s) + \frac{1}{L}V_c(s), \quad sV_m(s) = g\Theta(s),$$

assuming $\dot{\theta}(0) = 0$, $v_m(0) = 0$, $v_c(0) = 0$ for initial conditions.

A transfer function can then be deduced.

Example : container handling gantry crane

We have obtained :

$$\begin{cases} s\Theta(s) &= -\frac{1}{L}V_m(s) + \frac{1}{L}V_c(s) \\ sV_m(s) &= g\Theta(s) \iff \Theta(s) = \frac{s}{g}V_m(s) \end{cases}$$

This leads to

$$\begin{aligned} \frac{s^2}{g}V_m(s) &= -\frac{1}{L}V_m(s) + \frac{1}{L}V_c(s) \\ \iff \left(\frac{s^2}{g} + \frac{1}{L}\right)V_m(s) &= \frac{1}{L}V_c(s) \end{aligned}$$

Hence

$$\frac{V_m(s)}{V_c(s)} = \frac{1}{1 + \frac{L}{g}s^2} = \frac{Y(s)}{U(s)} = H(s).$$

Example : container handling gantry crane

The obtained transfer function

$$H(s) = \frac{Y(s)}{U(s)} = \frac{V_m(s)}{V_c(s)} = \frac{1}{1 + \frac{L}{g}s^2}$$

is a second-order ones whose standard form is

$$H(s) = K \frac{1}{1 + 2\xi \frac{s}{\omega_n} + (\frac{s}{\omega_n})^2}$$

By identification, we can deduce $\omega_n = \sqrt{\frac{g}{L}}$ and $\xi = 0$.

Stability

A linear system is said to be *stable* if for all bounded-range input, the output-response has also a bounded range.

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Theorem (Stability criterion regarding the transfer function)

Let a system be described by its transfer function $H(s)$. It is stable if, and only if, the poles (i.e. the roots of the denominator of $H(s)$) have all a strictly negative real part.

Example : container handling gantry crane

We can guess that the system corresponding to the set trolley-cables-container is not stable. For example, if the input is first set to a non-zero given value (the trolley-speed is set to a non-zero value), and then set to zero (the trolley is stopped) : in that case we can expect that the container will swing and these oscillations will continue indefinitely (frictions are neglected so that the range of the oscillations will not decrease) which means that the output-response will not tend to zero asymptotically.

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$$H(s) = \frac{1}{1 + \frac{L}{g}s^2}$$

The roots of $1 + \frac{L}{g}s^2$ (denominator of $H(s)$, the transfer function) are $\pm i\sqrt{\frac{g}{L}}$. Their real-part is not strictly negative. This demonstrates that the system is unstable.

Precision

A controlled-system with output $y(t)$ is all the more precise that the difference between the desired output $y_d(t)$ and the actual output $y(t)$ is low. The *precision* can be quantified by :

$$\varepsilon(t) \triangleq y_d(t) - y(t)$$

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Sometimes, the *stationary error* is considered. The stationary error of order n , denoted ε_n , is the steady-state error for an input of the form $U(s) = \frac{1}{s^n}$.