

# Modelling and Simulation

**Bionics, Biomimetics MSc.**

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# Organisation

Lectures weekly (Room 06 01 14)

Moodle ([moodle.hochschule-rhein-waal.de](https://moodle.hochschule-rhein-waal.de))

password: **ModSim1718**

Class starts at 08:15

## Content

- Introduction
- Modelling of Dynamic Systems  
Introduction to MATLAB / Simulink
- Solution of State Equations
- Numerical Methods for Dynamic Systems

# Introduction

## Target:

To understand and influence the behaviour of complex dynamic systems. For this purpose it is necessary to know and to understand the methods of modelling and analysis of this kind of systems along with their sub-systems and components.

## In this lecture we will deal with

- The mathematical description of dynamic systems
- Modelling techniques (with a focus on mechanical, electrical and biological systems)
- Numerical methods

# Mechanical Systems



# Mechanical Systems

Modelling and Simulation?  
What? Why?



# Mechanical Systems

## Modelling and Simulation?

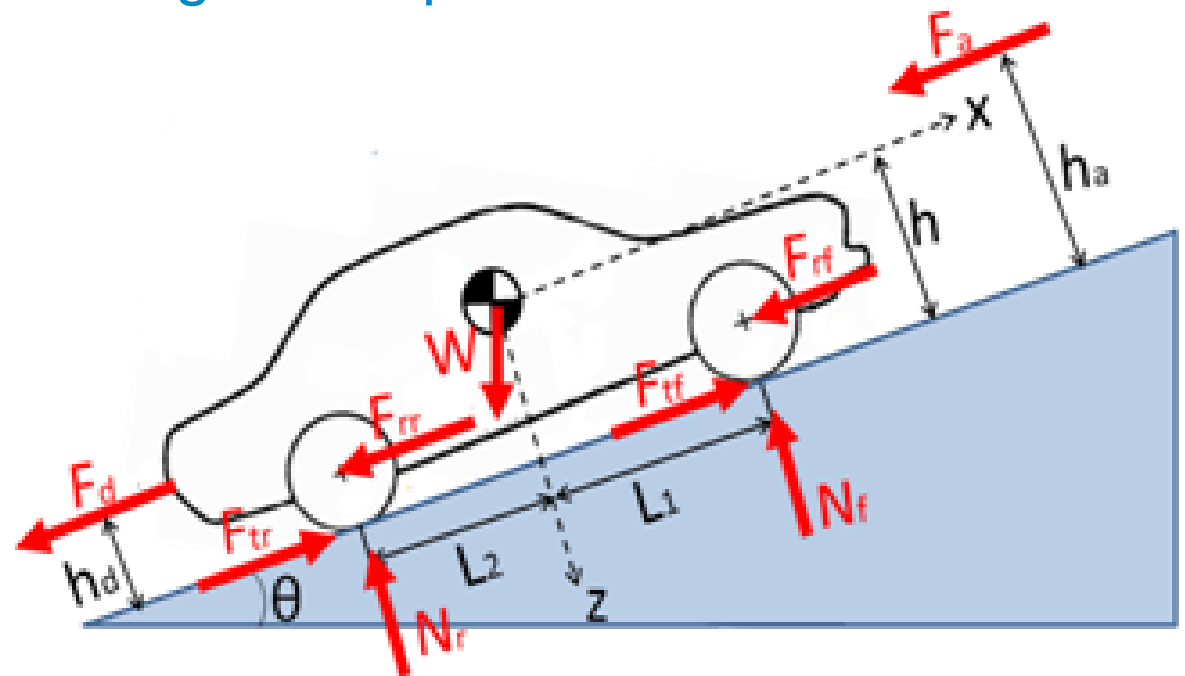
1. Identify effects and quantities that we are interested in.  
(e.g. forces and accelerations in vertical and longitudinal vehicle dynamics)



# Mechanical Systems

## Modelling and Simulation?

2. Modelling: Abstract from the concrete object and use a „domain specific“ modelling technique





# Mechanical Systems

## Modelling and Simulation?

3. Formulate the mathematical description of your model.  
(e.g. equations of motion)

$$\begin{aligned}\sum F_x &= m \cdot \ddot{x}_c \\ m \cdot \ddot{x}_c &= [K \cdot (\xi_1 + \xi_2) + c \cdot (\dot{\xi}_1 + \dot{\xi}_2)] \cdot \sin \alpha \\ &+ (F_{r1} + F_{r2}) \cdot \cos \alpha\end{aligned}$$

$$\begin{aligned}\sum F_y &= m \cdot \ddot{y}_c \\ m \cdot \ddot{y}_c &= -[K \cdot (\xi_1 + \xi_2) + c \cdot (\dot{\xi}_1 + \dot{\xi}_2)] \cdot \cos \alpha \\ &+ (F_{r1} + F_{r2}) \cdot \sin \alpha - P\end{aligned}$$

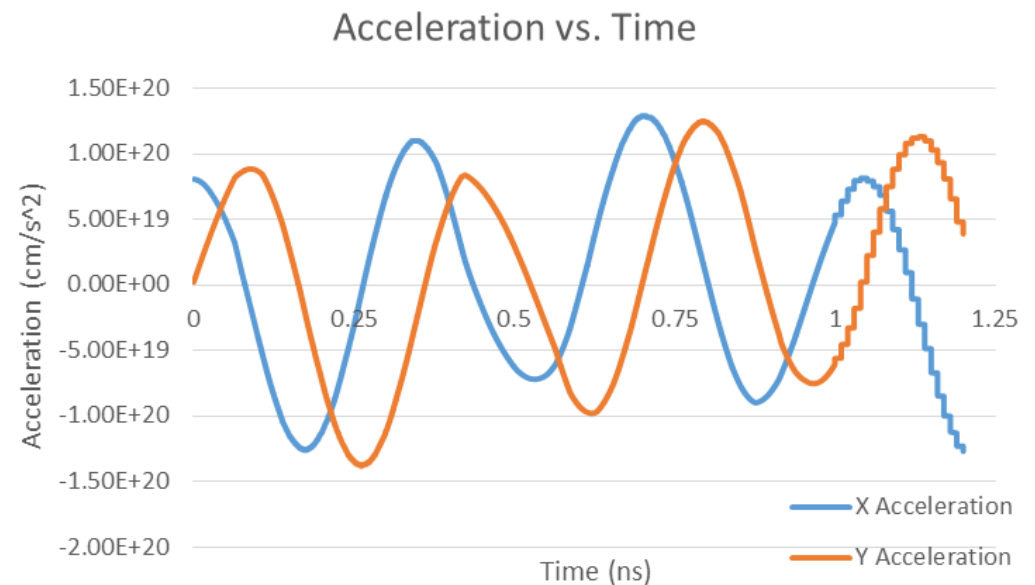
$$\begin{aligned}\sum M_{CM} &= I \cdot \ddot{\alpha} \\ I \cdot \ddot{\alpha} &= (K \cdot \xi_1 + c \cdot \dot{\xi}_1) \cdot L_1 - (K \cdot \xi_2 + c \cdot \dot{\xi}_2) \cdot L_2 \\ &+ F_{r1} \cdot (h_1 + \xi_1) + F_{r2} \cdot (h_2 + \xi_2)\end{aligned}$$

# Mechanical Systems

Modelling and Simulation?

4. Solve the equations of motion.

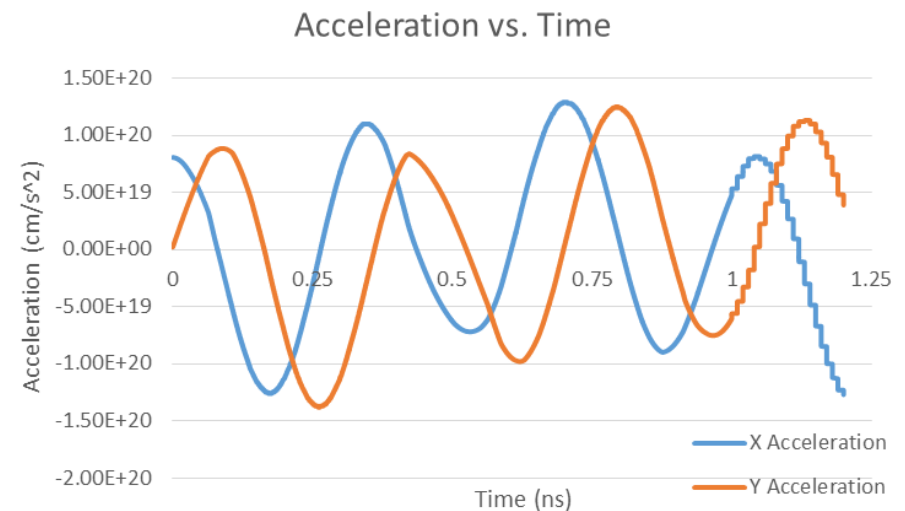
**Numerics!**



# Mechanical Systems

## Modelling and Simulation?

### 5. Interpretation and visualization of results, parameter studies, model optimization,...



## Definition (Hiller):

A system is a set of elements (parts, components, objects)

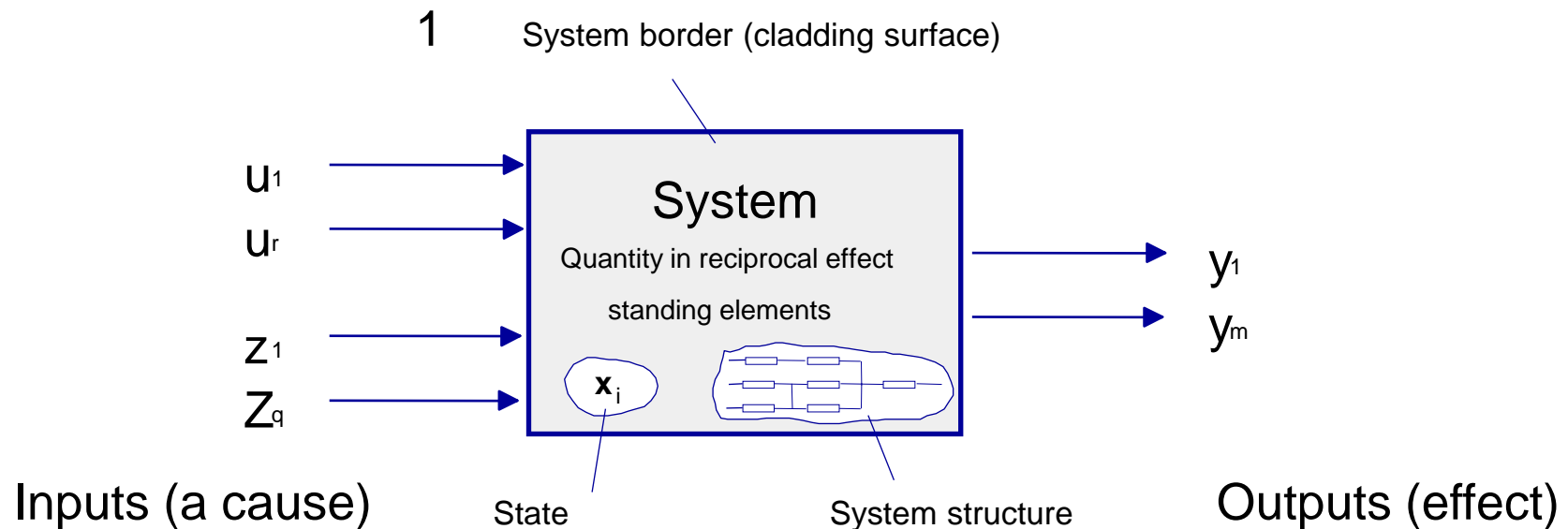
- which mutually influences each other (interaction)
- which are subject to external influence and affection (inputs)
- which effect to the external (outputs)

## Alternative definition:

A **system** is a defined arrangement of things influencing each other. This arrangement is separated from the environment by a thought surface. Connections of the system with its environment are cut by the surface. The characteristics and conditions transferred by means of these connections are the inputs and outputs of the system.

## Representation

Symbolically a system is represented by a box (block), whereby the cladding surface symbolizes the demarcation for environment.



The connections of the block with the environment are

- $r$  **Inputs**  $u_i$
- $m$  **Outputs**  $y_i$
- $q$  **Variable disturbances**  $z_i$

If no connections with the environment are present, then one speaks of a **closed system**.

**Definition:** *State variables* are those quantities of a system, whose knowledge is sufficient, in order to describe the system performance completely.

- The state variables are time-dependent.
- The states are summarized in the  $n \times 1$  state vector

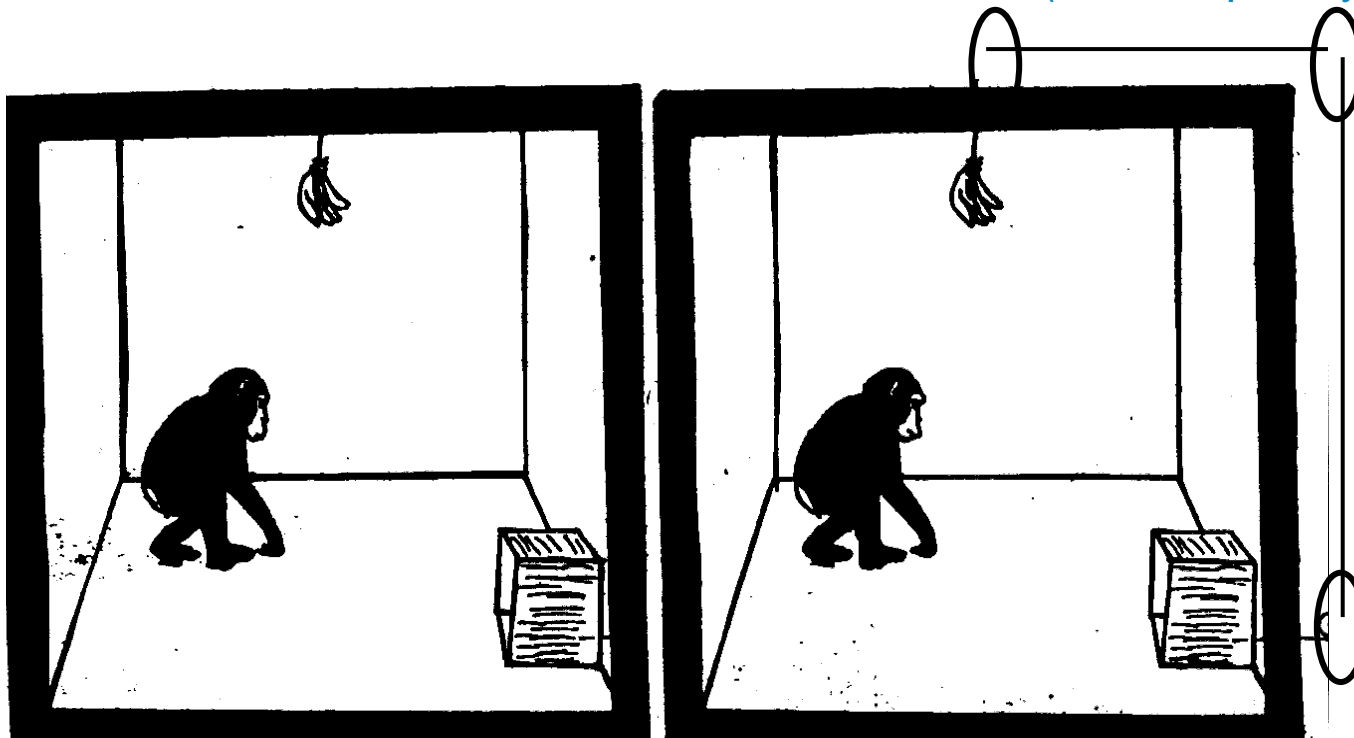
## Notes:

- The variables of state are to a certain extent "internal sizes" of the system.
- The choice of the variables of state is not unique.  
Variables of state must be independent and redundancy-free. From the requirement of being redundancy-free the number of state variables is fixed.
- Usually not all state variables are necessary to know for a particular application.  
Outputs of a system often follow from a combination of state variables.

## System demarcation

The system must be defined against its environment. This demarcation results not from the physical borders, but from the question, which is the basis for the view of system.

How important a correct demarcation is, this cartoon shows (not completely seriously meant):



# Systems

Examples of state variables for various disciplines

Electrical engineering	Mechanics	Chemical processes	Vital statistics
Current	Position	Temperature	Number of inhabitants
Voltage	Velocity	Mass proportion	State of environment
Electric charge	Acceleration		CO <sub>2</sub> Pollution
	Kinetic Energy		Ozon value
	Potential Energy		



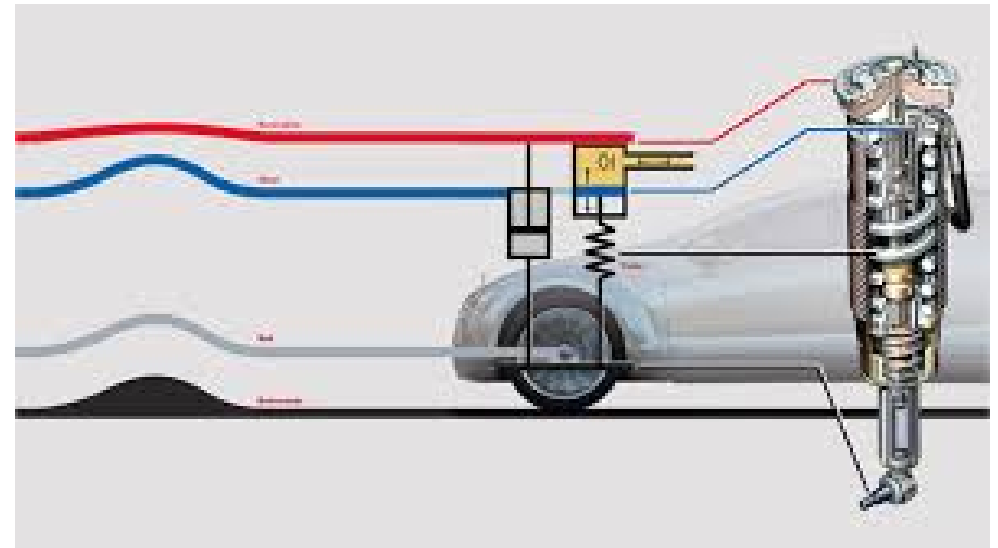
# Systems

Examples for system elements from various disciplines

<b>Electrical engineering</b>	<b>Mechanics</b>	<b>Chemical processes</b>	<b>Vital statistics</b>
Resistance	Mass	Reservoir	Birth
Capacitor	Spring	Valve	Death
Coil	Damper	Pipe	Disease
Transistor	Beam	Filter	Consumption
Amplifier	Bearing	Reactor	
Filter	Guidance		
	Force actuator		

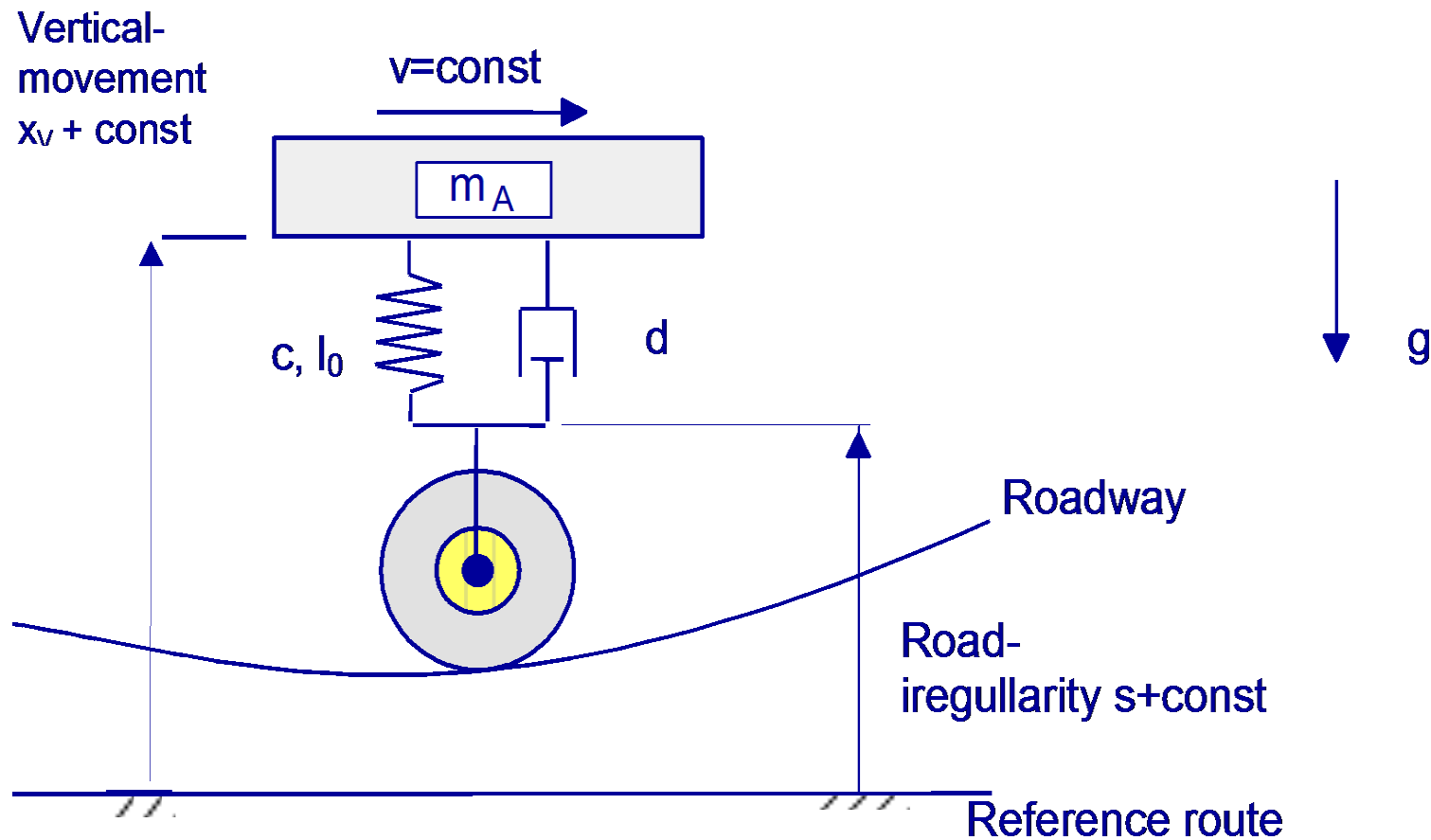
# Systems

## Example: Wheel suspension



Quelle: Auto Motor Sport

## Example: Wheel suspension



## Motion equations (mechanics):

$$m_A \ddot{x}_V = -c(x_V - s - l_0) - d(\dot{x}_V - \dot{s}) - m_A g$$

$$m_A \ddot{x}_V + d\dot{x}_V + cx_V = \underbrace{cs + d\dot{s}}_{h(t)} + cl_0 - m_A g$$

$$\Rightarrow \ddot{x}_V = \frac{1}{m_A} (h - mg - d\dot{x}_V - cx_V + cl_0)$$

$$0 = -cx_V + cl_0 - m_A g \Rightarrow x_{V0} = l_0 - \frac{m_A}{c} g$$

$$x_{VA} = x_V - x_{V0} = x_V - l_0 + \frac{m_A}{c} g$$

$$\ddot{x}_{VA} = \frac{1}{m_A} (h - d\dot{x}_{VA} - cx_{VA})$$

## Equations of state

**Set:**

$$x_1 = x_{V_A}$$

$$x_2 = \dot{x}_{V_A}$$



$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = \ddot{x}_{V_A} = \frac{1}{m_A}(h - dx_2 - cx_1)$$

**Input sizes:**

$$u_1 = h(t)$$

Road irregularity

**Variables of state:**

$$x_1, x_2$$

Position and speed

**Output values:**

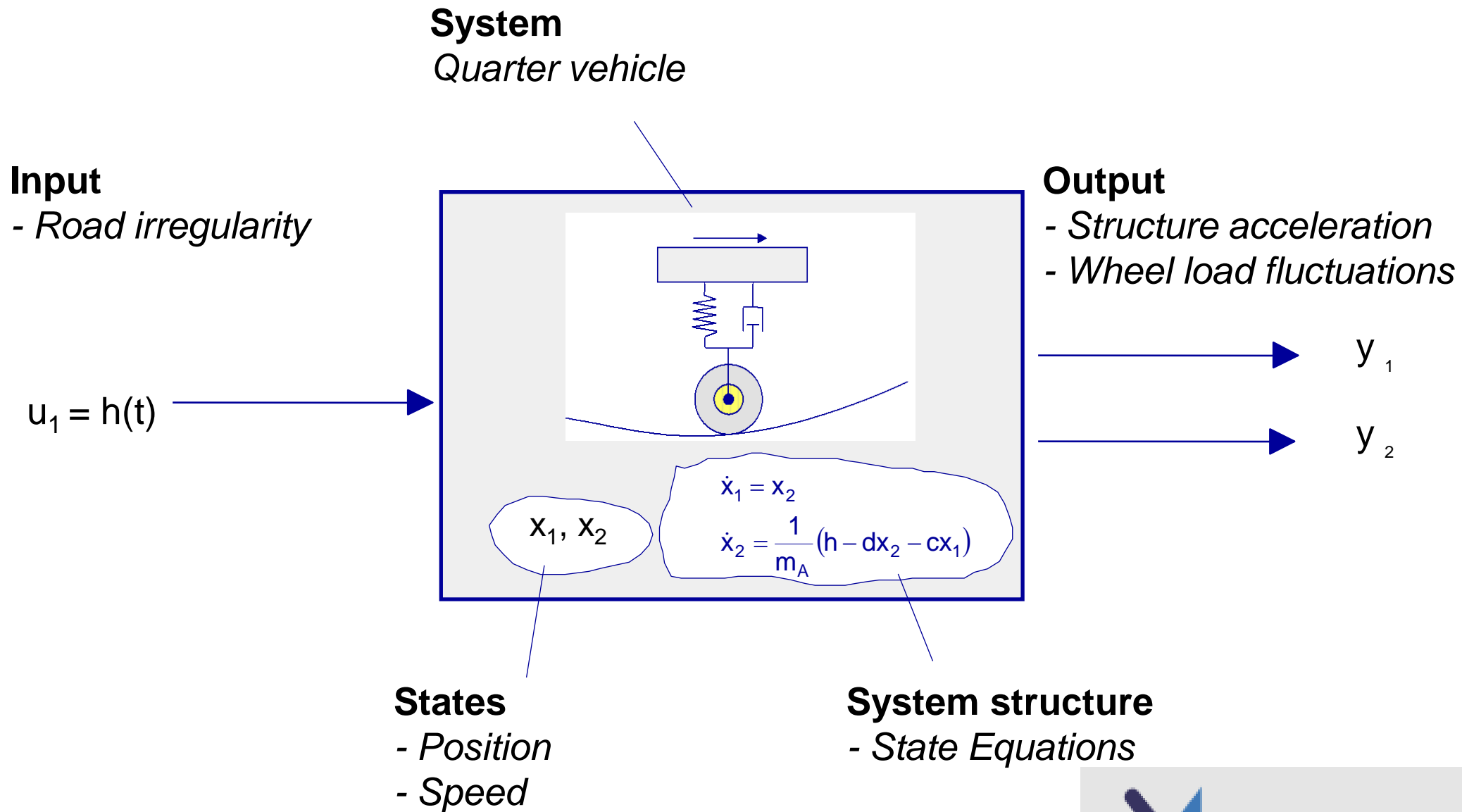
$$y_1 = \dot{x}_2$$

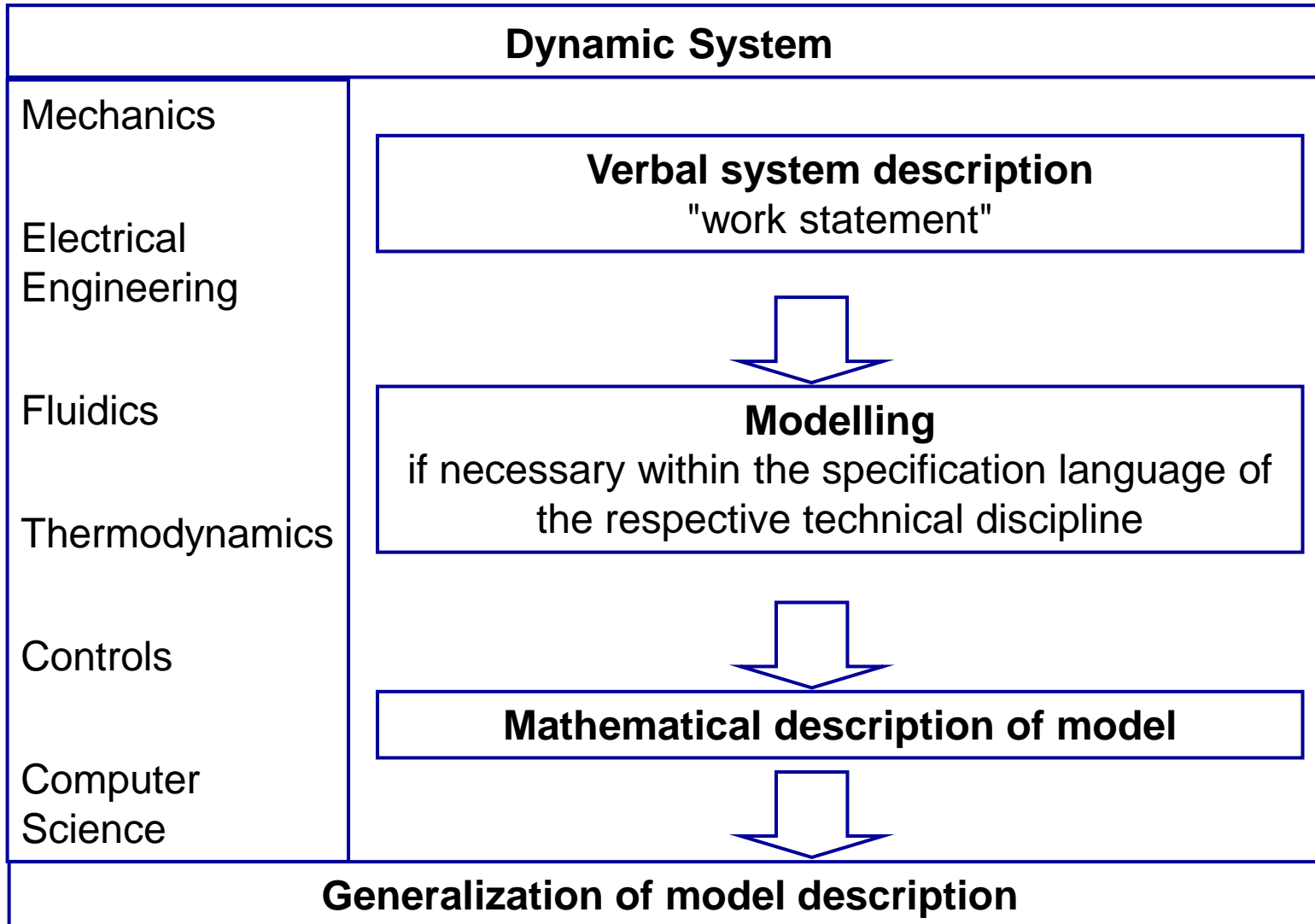
Structure acceleration

$$y_2 = h - dx_2 - cx_1$$

Wheel load fluctuation

# Systems





For the understanding of complex systems a concept, which describes the dynamics behavior of the system sufficiently exact is often necessary. This concept takes place with domain-specific systems of a certain discipline predominantly within the common "model language".

**Mechanics: mechanical components**

**Electro-technics: Connection diagrams, etc.**

In case of multi-domain systems however a common modeling is necessary. Often simplification of the subsystems is required.



## Model concepts

Models are simplified images of the reality, in order to be able to analyze certain functions and behaviours. Models are e.g. the basis for simulation of the system. For different investigations, in particular with different investigation depth, different models are needed.

- **Scaled physical models**
- **Mathematical-physical models.**

In this lecture we will deal exclusively with mathematical-physical models.

## Block diagrams

*Block diagrams* represent a descriptive representation of the model equations, i.e. block diagrams provide a representation of the interconnections between the single parts of a system.

<b>Signal</b>	-	<b>Arrow</b>
<b>Operation</b>	-	<b>Block</b>

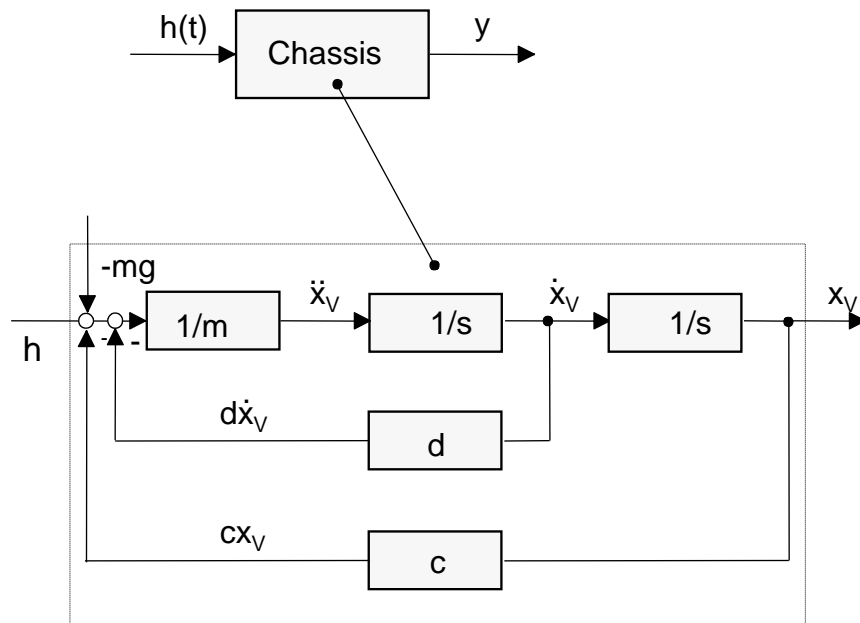
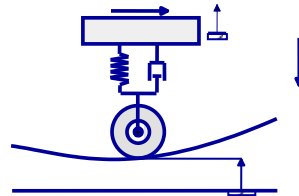
A block diagram thus always consists of

- The blocks
- Arrows for in and outgoing signals at the blocks, whereby by a head of the arrow in the respective direction of action it is indicated whether it concerns the output signal or one of the input signals.
- Definition of the input and output values
- Summations, represented as small circles

Further details to the structure and the elements of block diagrams are standardized in *DIN 19226*.

## Examples of block diagrams

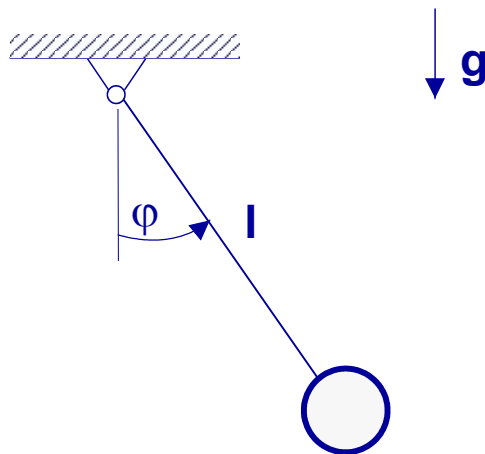
### Wheel suspension



$$\ddot{x}_V = \frac{1}{m_A} (h - mg - d\dot{x}_V - cx_V)$$

**Definition:** An equation is called *ordinary differential equation* (ODE), if it contains apart from the variable also at least one derivative of this variable.  
*Systems with concentrated parameters* (e.g. *multi-body systems*) are described by ODE.

**Example:** ODE (*equation of motion*) for a mathematical pendulum



$$ml^2\ddot{\varphi} = -mgl \sin \varphi$$

# Modelling

**Definition:** A function of several independent variable and their derivatives is called *partial differential equation* (PDE).

*Systems with (locally) distributed parameters* (e.g. continuous mechanical systems) are described by PDE.

**Example:** Longitudinal oscillations of a continuous bar

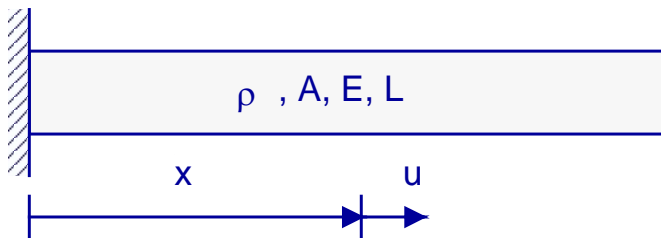
## Designations:

$\rho$ : Density

E: Elasticity modulus

A: Bar cross section

L: Bar length

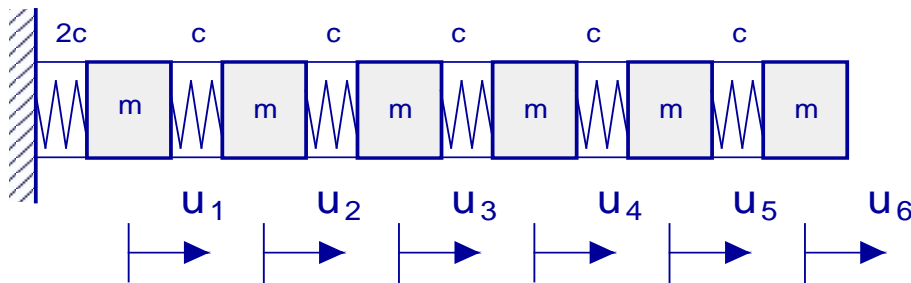


$$\rho A \ddot{u}(x,t) = \frac{EA}{L^2} \frac{\partial^2 u}{\partial x^2}(x,t)$$

Here derivatives with respect to the local coordinate  $x$  and time arise.

# Modelling

**Note:** A system with distributed parameters can frequently be approximated by a spatial discretisation by a system with concentrated parameters. In this case it receives a system instead of ODE instead of one PDE:

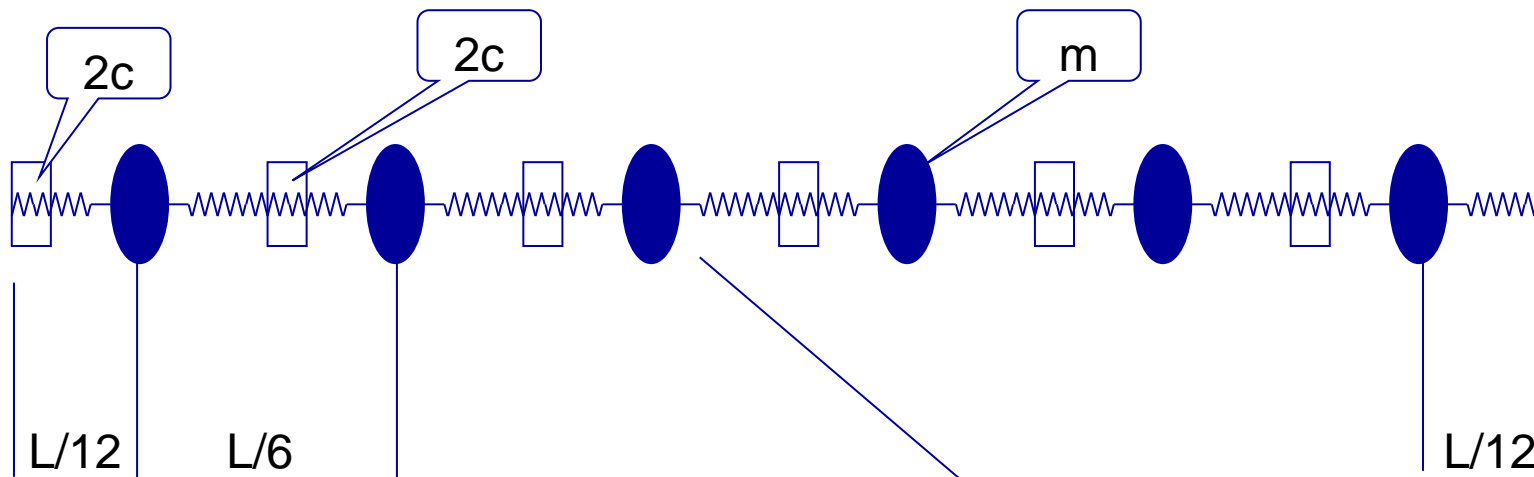


$$\begin{aligned} m\ddot{u}_1 &= -3cu_1 + cu_2 \\ m\ddot{u}_2 &= cu_1 - 2cu_2 + cu_3 \\ m\ddot{u}_3 &= cu_2 - 2cu_3 + cu_4 \\ m\ddot{u}_4 &= cu_3 - 2cu_4 + cu_5 \\ m\ddot{u}_5 &= cu_4 - 2cu_5 + cu_6 \\ m\ddot{u}_6 &= cu_5 - cu_6 \end{aligned}$$

$$\ddot{\mathbf{u}} = \begin{bmatrix} \ddot{u}_1 \\ \ddot{u}_2 \\ \ddot{u}_3 \\ \ddot{u}_4 \\ \ddot{u}_5 \\ \ddot{u}_6 \end{bmatrix} = \frac{36E}{\rho L^2} \begin{bmatrix} -3 & 1 & 0 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 & 0 \\ 0 & 1 & -2 & 1 & 0 & 0 \\ 0 & 0 & 1 & -2 & 1 & 0 \\ 0 & 0 & 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \\ u_5 \\ u_6 \end{bmatrix} = \mathbf{A}\mathbf{u}$$

with  $c = \frac{AE}{L/6}$   
and  $m = \frac{\rho AL}{6}$

# Modelling

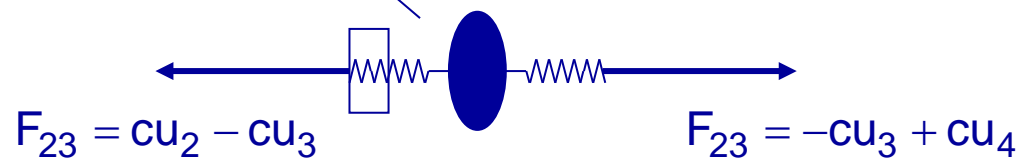


spring constant

$$\sigma = \frac{F}{A} = E\varepsilon = \frac{u_i}{L/6} \Rightarrow F = \underbrace{AE \frac{1}{L/6}}_c u_i$$

mass

$$m = \frac{\rho AL}{6}$$



Newton's Law

$$m\ddot{u}_3 = cu_2 - cu_3 - cu_3 + cu_4 = cu_2 - 2cu_3 + cu_4$$

In systems with concentrated parameters the dynamic system can be described by ODE as follows:

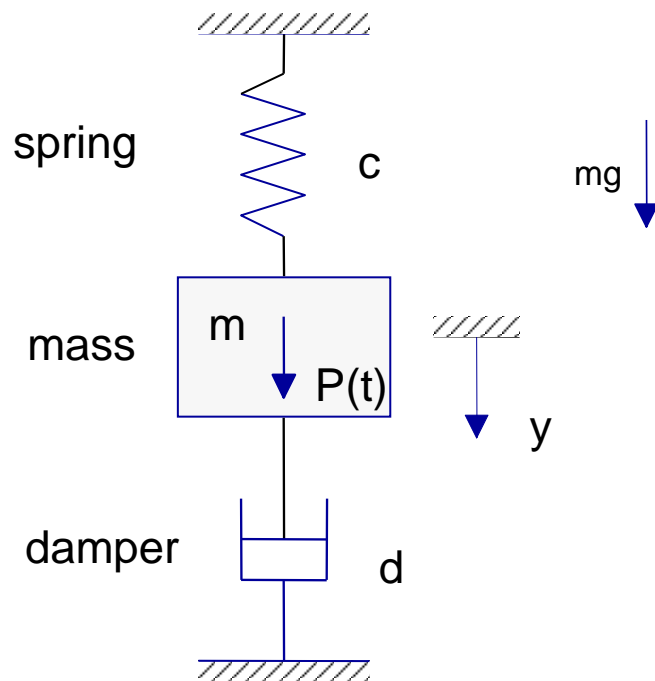
or                      *implicit Form*       $\mathbf{F}^i(t, \mathbf{y}(t), \mathbf{y}'(t), \dots, \mathbf{y}^{(n)}(t)) = \mathbf{0}$   
                         *explicit Form*       $\mathbf{y}^{(n)}(t) = \mathbf{F}^e(t, \mathbf{y}(t), \mathbf{y}'(t), \dots, \mathbf{y}^{(n-1)}(t))$

## Notes:

- In some cases the transfer of the implicit form into the explicit equation form is *analytically* not possible. In the system dynamics this is however usually not the case.
- In practice often systems arise, which can be described by coupled differential and algebraic equations (DAE = differential Algebraic Equations). An example of it are multi-body systems with kinematic loops.
- The number of the highest arising derivative is called order of the ODE.
- Since the function  $y$  in system dynamics usually depends on the time, we mark the derivatives of it by a dot on top of it.



## Example: oscillator with viscous damper



**implicit equation:**

$$F^i(t, y, \dot{y}, \ddot{y}) = m\ddot{y} + d\dot{y} + cy - P(t) = 0$$

**explicit equation:**

$$\ddot{y}(t) = F^e(t, y, \dot{y}) = \left[ -\frac{1}{m}(d\dot{y} + cy - P(t)) \right]$$

By introduction the state variables  $x_1, \dots, x_n$

and the substitutions  $x_1 = y \quad x_2 = y' \quad x_n = y^{(n-1)}$   
 $x_1 = y \quad x_2 = \dot{x} \quad x_n = \dot{x}_{n-1}$

$$y^{(n)}(t) = F^e(t, y(t), y'(t), \dots, y^{(n-1)}(t))$$
$$\dot{x}_n(t) = F^e(t, x(t), \dot{x}(t), \dots, \dot{x}_{n-1}(t))$$

can be transferred into a system of 1th order differential equations.

## Example

Linear oscillator  $\ddot{y} + 2\delta\dot{y} + v_0^2 y = h(t)$

Substitution  $x_1 = y \quad x_2 = \dot{y}$

Equations of state  $\dot{x}_1 = x_2$   
 $\dot{x}_2 = h(t) - 2\delta x_2 - v_0^2 x_1$

# State Equations

## State equations

$$\dot{\mathbf{x}} = \frac{d\mathbf{x}}{dt} = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t)$$

$$\mathbf{y}(t) = \mathbf{g}(\mathbf{x}(t), \mathbf{u}(t), t)$$

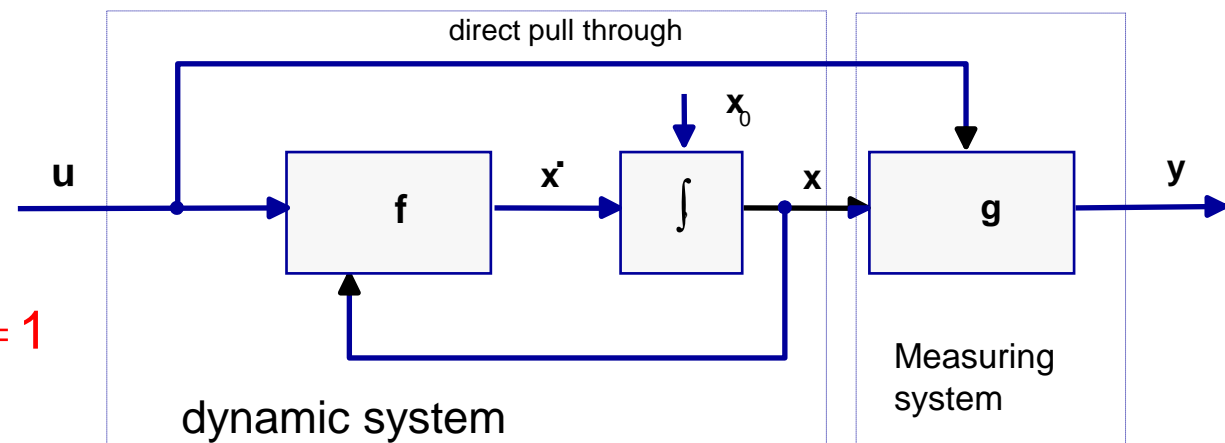
$\mathbf{x} : n \times 1$  – state vector

$\mathbf{y} : m \times 1$  – output vector

$\mathbf{u} : r \times 1$  – control vector

$\mathbf{f}, \mathbf{g} : \text{non-linear } n \times 1 - / m \times 1$  – vector-function

## Block diagram

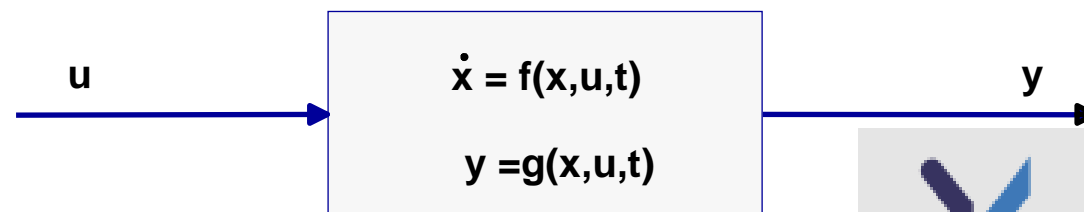


example :  $n = 2, m = 1, r = 1$

$$\dot{x}_1 = f_1(x_1, x_2, u_1, t)$$

$$\dot{x}_2 = f_2(x_1, x_2, u_1, t)$$

$$y_1 = g_1(x_1, x_2, u_1, t)$$



# State Equations

## Example: Mass Point

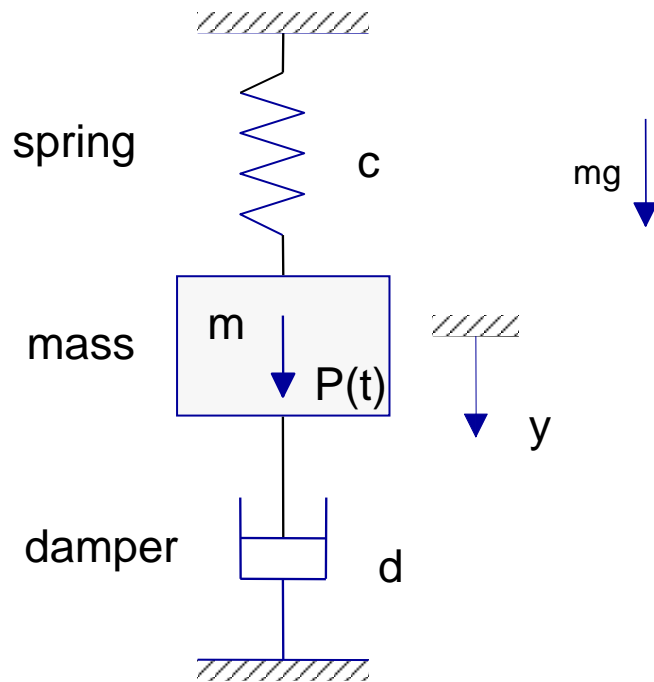
Newton's law:  $m\ddot{y} = F$

Substitution:  $x_1 = y \quad x_2 = \dot{y}$

State equations:  $\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = f(x_1, x_2, u, t) = \begin{bmatrix} x_2 \\ \frac{F}{m} \end{bmatrix}$  with  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad u = \frac{F}{m}$

## Example: oscillator with viscous damper

# State Equations



**Implicit equation:**

$$F^i(t, y, \dot{y}, \ddot{y}) = m\ddot{y} + d\dot{y} + cy - P(t) = 0$$

**Explicit equation:**

$$\ddot{y}(t) = F^e(t, y, \dot{y}) = \left[ -\frac{1}{m_1} (d\dot{y} + cy - P(t)) \right]$$

## State equations

$$\begin{aligned} x_1 &= y & \dot{x}_1 &= \dot{y} = f_1(x_1, x_2, t) = x_2 \\ x_2 &= \dot{y} & \dot{x}_2 &= \ddot{y} = f_2(x_1, x_2, t) = -\frac{c}{m}x_1 - \frac{d}{m}x_2 + \frac{1}{m}P(t) \end{aligned} \Rightarrow \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -\frac{c}{m} & -\frac{d}{m} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} P(t)$$

$$u_1 = P(t) \quad y_1 = g(x_1, x_2, u_1, t) = x_1 \Rightarrow [y_1] = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

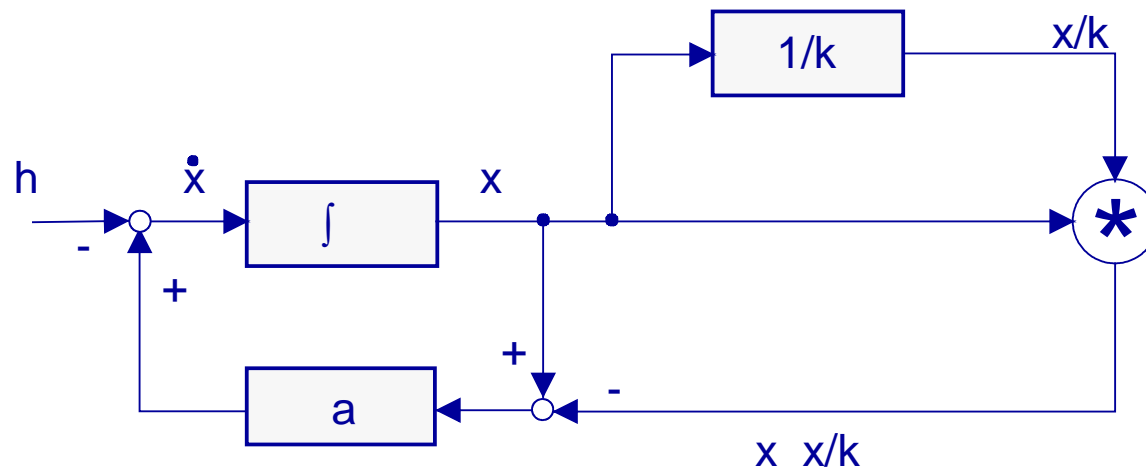
# State Equations

## Example logistic growth

state variable (asset e.g. fishes) :  $x$

control variable (e.g. fishing):  $u = h = \text{const}$

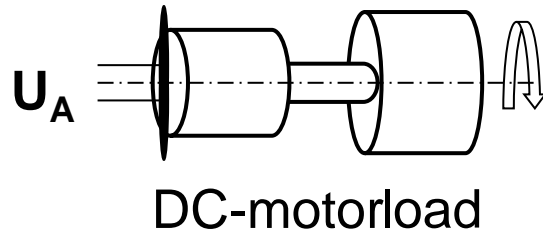
State equations:  $\dot{x} = a \left( 1 - \frac{x}{k} \right) x - h$



# State Equations

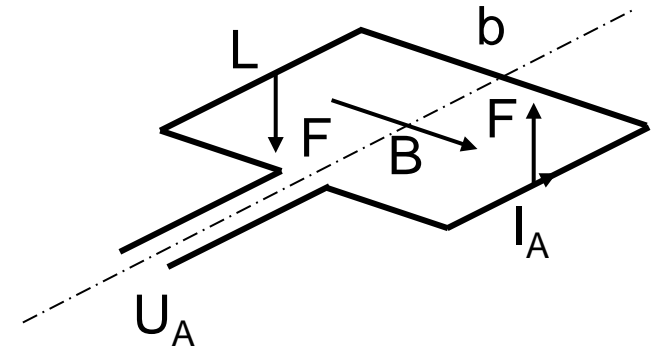
## Example: DC Motor

electrical  
input



$\omega$

mechanical  
output



torque:  $M = \alpha n L b (B I_A) = k_1 (B I_A)$

where

n: number of conductor loops

B: magnetic field strength

L, b: dimensions of conductor loop

$I_A$ : anchor current

$\alpha$ : compensation factor

induced voltage (EMF)  $U_b = k_2 B \omega$

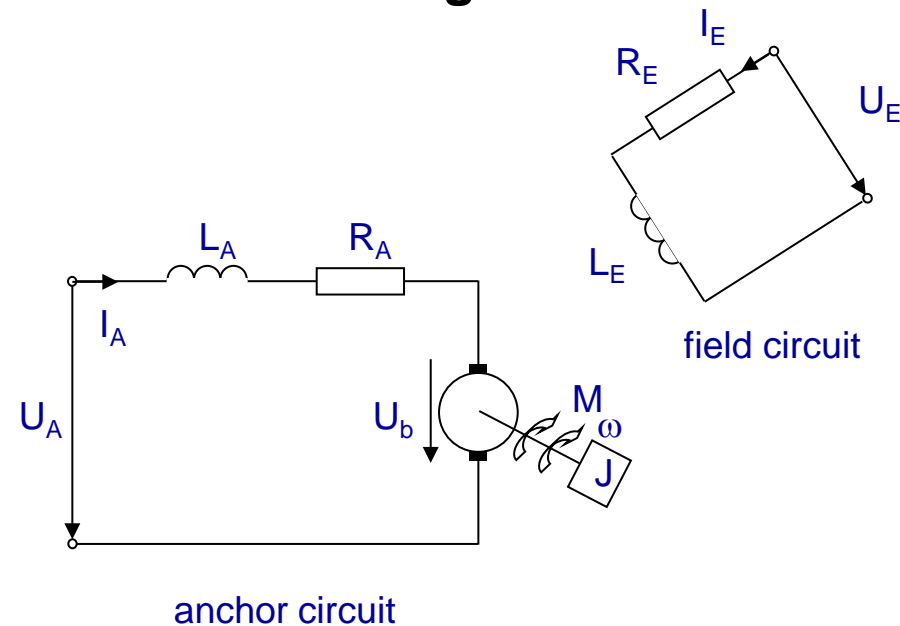
where

$k_2$ : constant

$\omega$ : anchor rotation velocity

# State Equations

## model plan circuit drawing



for a anchor controlled motor  
 $I_E = \text{const}$  and hence  $B = \text{const}$

$$U_b = k_2 B \omega = k_3 \omega$$

anchor voltage

$$U_A$$

induction law for anchor circuit

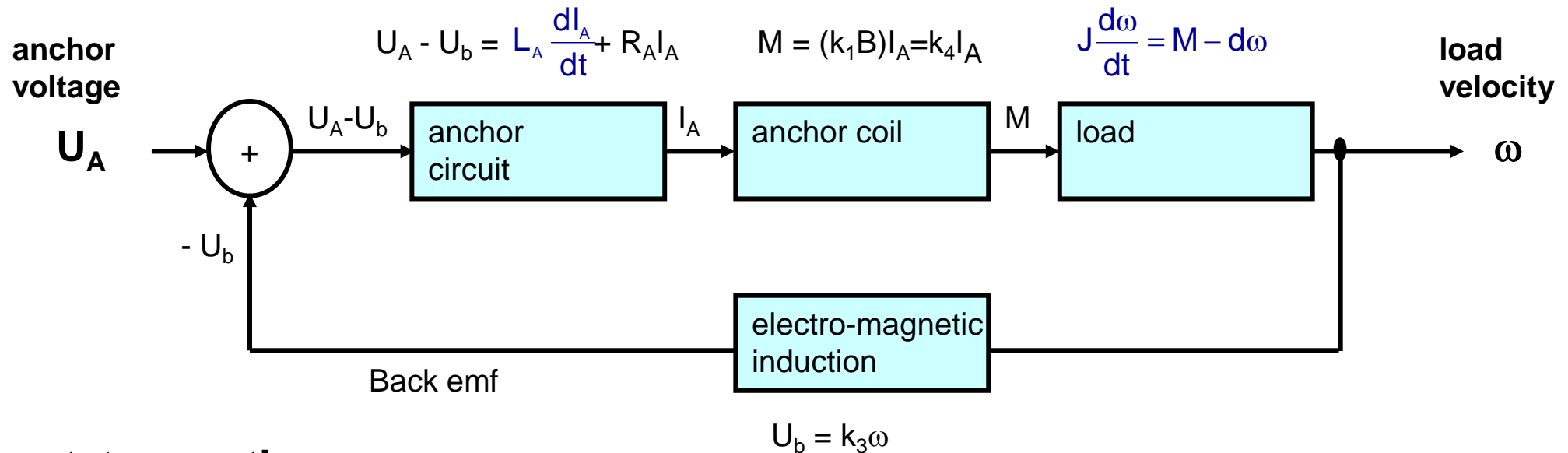
$$U_A - U_b - R_A I_A - L_A \frac{dI_A}{dt} = 0$$

$$U_A - U_b = L_A \frac{dI_A}{dt} + R_A I_A$$



# State Equations

## anchor controlled motor



## state equations

$$x_1 = I_A \quad \dot{x}_1 = \dot{I}_A = f_1(x_1, x_2, t) = -\frac{R_A}{L_A} x_1 - \frac{k_3}{L_A} x_2 + \frac{1}{L_A} U_A$$

$$x_2 = \omega \quad \dot{x}_2 = \dot{\omega} = f_2(x_1, x_2, t) = \frac{k_4}{J} x_1 - \frac{d}{J} x_2$$

$$u_1 = U_A \quad y_1 = g(x_1, x_2, u_1, t) = \omega = x_2$$

$$\underbrace{\frac{L_A}{R_A}}_{\text{time constant } T_A} \dot{x}_1 + x_1 = \frac{k_3}{R_A} x_2 + \frac{1}{R_A} U_A$$

$$\underbrace{\frac{J}{d}}_{\text{time constant } T_J} \dot{x}_2 + x_2 = \frac{k_4}{d} x_1$$

# Steady State Solutions

## Definition:

The solution  $x_0(t)$  of a dynamic system is called ***steady state*** if for a given load  $u_0(t)$

$$F(x_0(t), u_0(t), t) = 0 \quad \text{and hence} \quad \dot{x}(t) = 0$$

If  $u_0(t)=0$  then  $x_0(t)$  is an equilibrium solution for the free not controlled system.

**Remark:** This equation is generally non-linear and therefore normally numerical methods have to be used to find a solution.

# Steady State Solutions

**Example 1:** single mass point

$$\mathbf{f}(\mathbf{x}, \mathbf{u}, t) = \begin{bmatrix} \dot{x}_2 \\ \frac{F}{m} \end{bmatrix} = \mathbf{0} \Rightarrow x_2 = 0 \quad \text{and} \quad F = 0$$

**Example 2:** spring-mass-oscillator with viscous damper

$$\mathbf{f}(\mathbf{x}, \mathbf{u}, t) = \begin{bmatrix} x_2 \\ -\frac{1}{m}(cx_1 + dx_2 - mg - P(t)) \end{bmatrix} = \mathbf{0} \quad \Rightarrow x_2 = 0 \Rightarrow x_1 = \frac{mg}{c}$$

$\frac{mg}{c}$  is the displacement of the mass  $m$  due to gravity until equilibrium is reached

# Steady State Solutions

**Example 3:** single mass pendulum

$$\mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{t}) = \begin{bmatrix} x_2 \\ -\frac{g}{l} \sin x_1 \end{bmatrix} = \mathbf{0} \quad \Rightarrow x_2 = 0 \quad \sin x_1 = 0$$

$$\text{hence } x_1 = 0, \pi, 2\pi, \dots$$

**Remark:** these equilibrium points show completely different stability behaviour

**Example 4:** logistic growth with constant harvest

$$f(x, u, t) = a \left( 1 - \frac{x}{k} \right) x - h = 0 \Rightarrow ax - \frac{a}{k} x^2 - h = 0$$
$$\Rightarrow x_{1/2} = \frac{1}{2} \left( k \pm \sqrt{k^2 - 4 \frac{kh}{a}} \right)$$

For  $h=0$  we get in particular  $x_1 = 0$   $x_2 = k$