

# Lecture 5:

Two different classes of modeling tools

Modeling of complex systems – loudspeaker example (F4)

Introduction to simulation of continuous-time models  
(E14.2, E14.3.1)

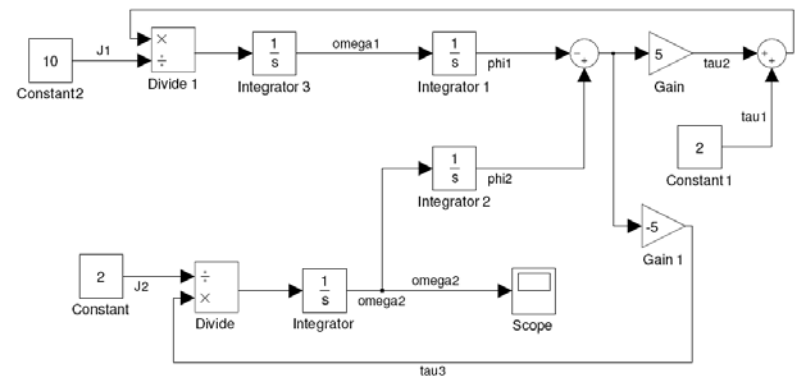
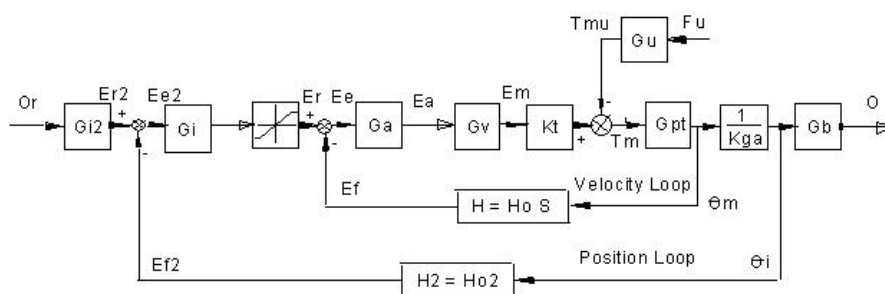
- Notation, computation errors
- Test system and stability function
- Euler's method

# Two different classes of modeling tools

# Signal-flow modelling

In cybernetics, we are often using *signal-flow* models

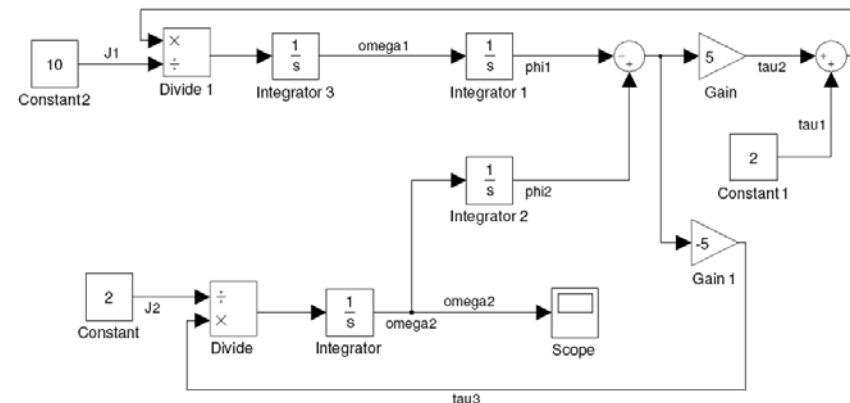
- Sub-models are 'blocks' with inputs and outputs
- Communication between blocks are signals with direction (arrows)
- Examples:
  - Block diagrams (signals are inputs/controls, outputs/measurements, references/set-points, etc.)
  - Simulink



- Other name: Causal modelling

# Signal-flow modelling

- Signal-flow models are well suited for control design and dynamic analysis
  - Intuitive(?) to implement, easy to simulate
  - Similarities with control methods taught in basic control courses
    - Transfer functions, Bode plots, Nyquist plots, etc.
    - In some respects: Passivity analysis, energy-based methods, ...
- Signal-flow models are not well suited for building "large"/"complex" simulation-models
  - Constructing large models can become complicated due to little physical structure
  - It can be difficult to make changes
  - It can be difficult to re-use models



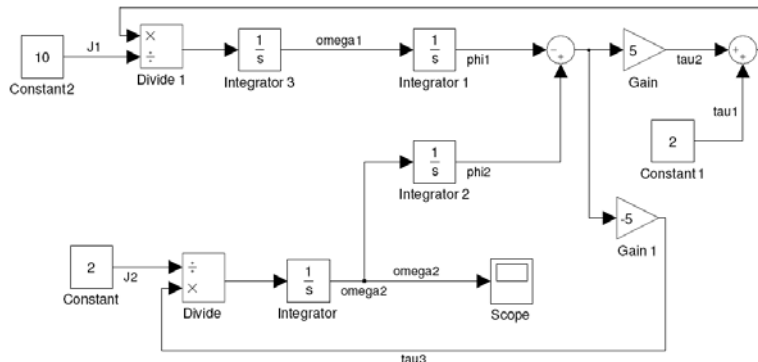
# The other alternative: Object-oriented modelling

- Variants (also known as):
  - Equation-based modelling
  - Network description (from electrical circuit theory)
  - Energy-flow modeling (used in the book E)
  - Component based modelling, Acausal modelling, ...
- Technologies
  - Bond graphs
  - Modeling languages & tools
    - Modelica (Dymola, OpenModelica, MapleSim, ...), gPROMS, Ascend, ...
- Sub-systems interconnected via physically motivated interfaces
  - Easy to replace sub-systems
  - Easy to re-use models (build model libraries)
  - Graphical user interfaces
    - Can build models by drag-and-drop
    - Models are (to some extent) self-documenting

# Tools for simulation

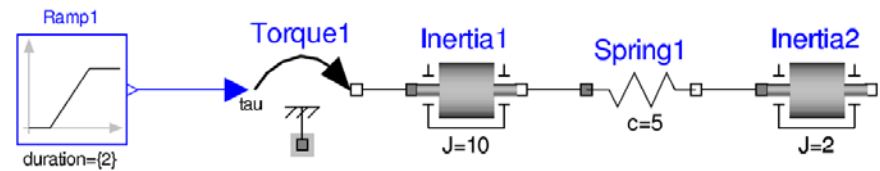
## Signal-flow modelling

- [Simulink](#)
- Simulink "clones"
  - e.g. [SciCos](#), [Xcos](#)
- [SystemBuild](#)
  - (NI/Labview, MATRIXx)
- ...



## Object-oriented modelling

- [Modelica](#) (language)
  - Tools: [Dymola](#), [MapleSim](#), [SimulationX](#), [OpenModelica](#), [Wolfram SystemModeler](#), ...



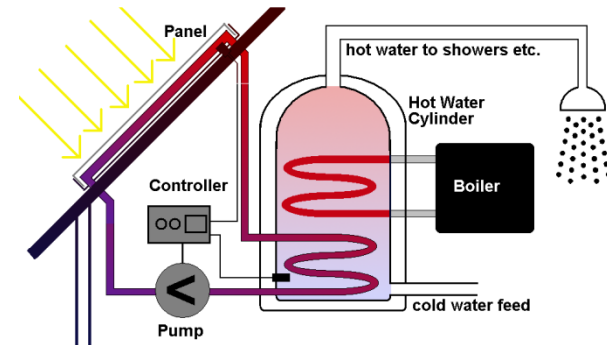
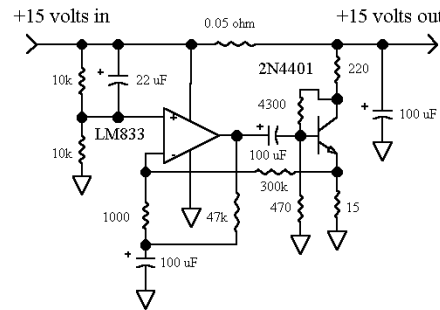
## Others

- [gPROMS](#)
- [ASCEND](#)
- [Aspen HYSYS Dynamics](#)
- [Matlab Simscape](#)
- ...



# How should sub-systems be connected?

- How does physical sub-systems interact?
  - DC motor and load
  - Resistor and a capacitor
  - Heating water
  - Hydraulics on an excavator



- Change requires energy/power!
- Need information about two variables to decide energy/power transfer:
  - Force and velocity, or torque and rotational velocity (mechanical systems)
  - Voltage and current (electrical systems)
  - Enthalpy/temperature and mass flow (thermal systems)
  - Pressure and volume flow (hydraulic/flow systems)

# Potential and flow variables

- Two variables needed to model **flow of energy** between two sub-systems
  - Potential variables: Variables that should be equal at interconnection (ex: voltage)
  - Flow variables: Variables that should sum to zero at interconnection (ex: current)

Domain	Potential	Flow
Translational mechanics	Velocity [m/s]	Force [N]
Rotational mechanics	Angular velocity [1/s]	Torque [Nm]
Electrical	Voltage [V]	Current [A]
Magnetical	Magnetomotive Force [A-turn]	Magnetic flux rate [Wb/s]
Hydraulical	Pressure [Pa]	Volume flow rate [m <sup>3</sup> /s]
Thermal	Temperature [K]	Heat flow rate [J/Ks]
Chemical	Chemical potential [J/mol]	Molar flow rate [mol/s]

- Product of Potential and Flow variables is power transfer!
  - Other choices of potential and flow variables are possible, as long as power can be inferred
- Note: Book E uses other terminology and call the variables effort and flow (from bond graph-theory). But essentially the same concept.



# In exercise 1:

- Attempt to use physically motivated interfaces in Simulink!
  - Advantage: You get Simulink blocks that are easy (easier) to reuse
  - Disadvantage (with Simulink):
    - No "language constructs" forcing/helping you to use "right" port variables
    - Signal flow direction must be decided beforehand
- Then do the same with Modelica/Dymola

# **MODELING OF COMPLEX SYSTEMS – LOUDSPEAKER EXAMPLE**

# Modeling of complex systems

- Many (modern) technical systems are
  - complex (composed of many components)
  - multi-domain (electrical, mechanical, . . . )
- Often non-trivial to go from physical process to mathematical model
  - what to model and what to not
  - appropriate abstractions & simplifications, relevant physics
- Need a modeling methodology!

# Multidomain modeling example: Loudspeaker



**Task:** Model the relationship between applied voltage and resulting sound pressure

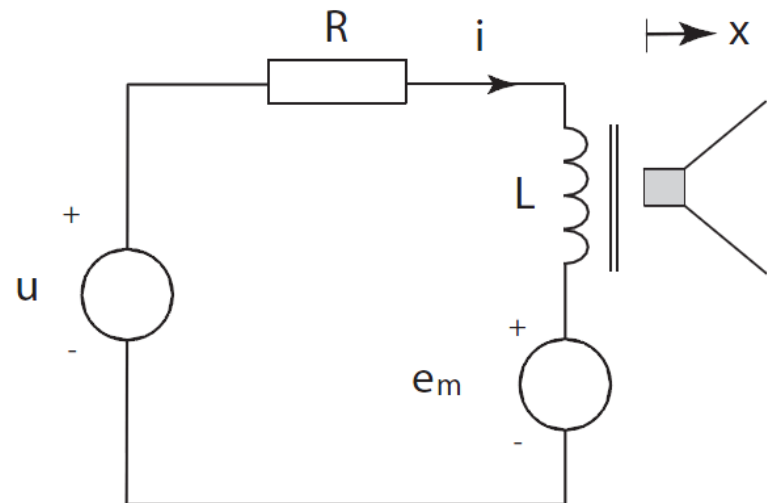
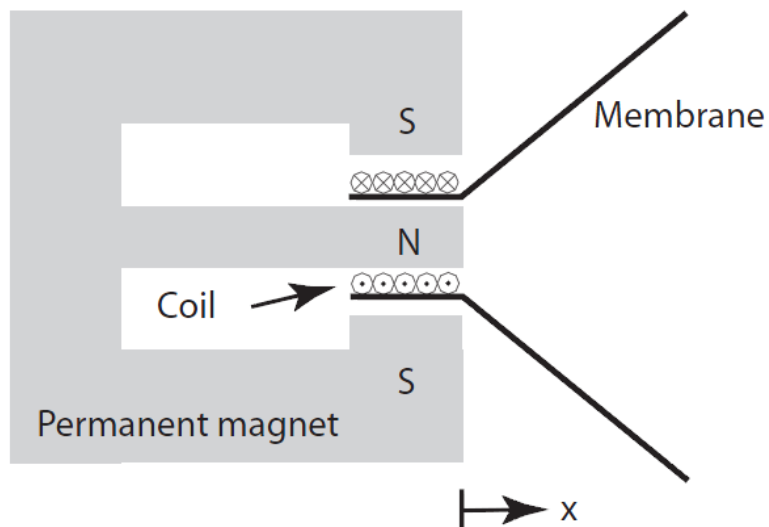
(Example taken from H. Hjalmarsson, KTH)

# Modeling methodology

- Structure the problem
  - Understand how the system works!
  - What is the purpose of the model?
  - Decompose the model hierarchically into sub-systems (if the system is «large»)
- Define models for sub-systems, and interactions
  - Based on physics and domain knowledge
- Organize equations
  - State-space form (ODE, DAE or PDE)
- Validate and simplify

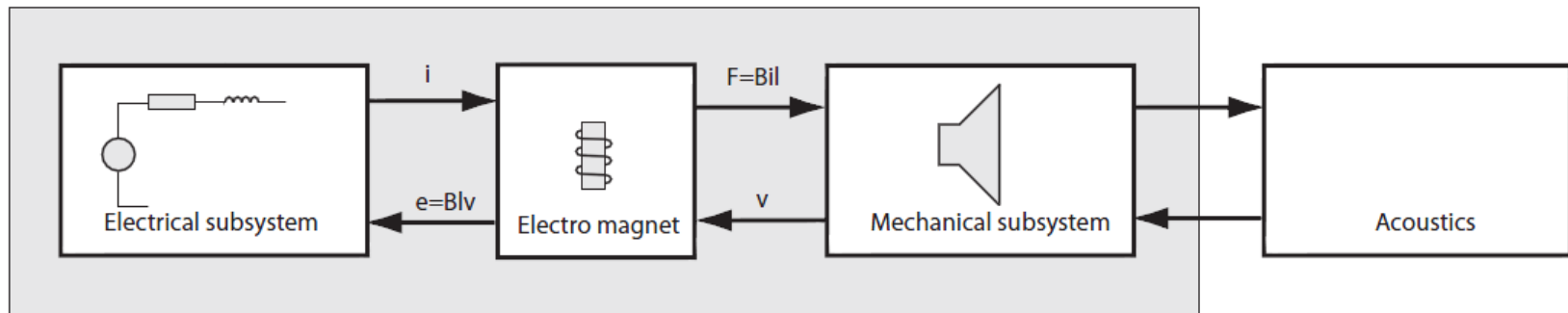
# Structuring the problem

- How does a loudspeaker work?



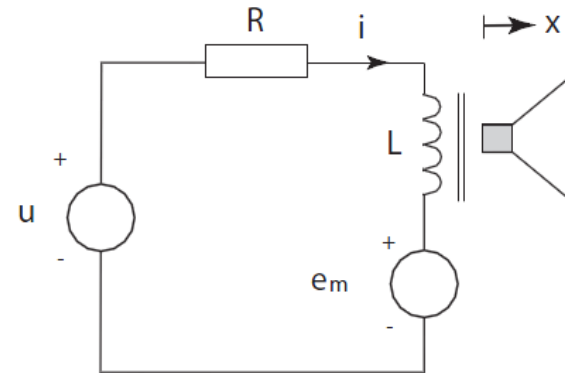
# Structuring the problem

- Draw simple pictures and diagrams



- Try to reflect logical relations, physical principles, cause-and-effect

# Modeling of subsystems



Electrical subsystem:

$$u - Ri - L \frac{di}{dt} - e_m = 0$$

Electromagnetic subsystem:

$$e_m = Blv$$

$$F_m = Bil$$

Mechanical subsystem:

$$m \frac{d^2x}{dt^2} = F_m - dv - kx$$



# Organizing the equations

Goal: Have the model equations in useful form (for analysis, simulation, ... )

...typically, in state space form

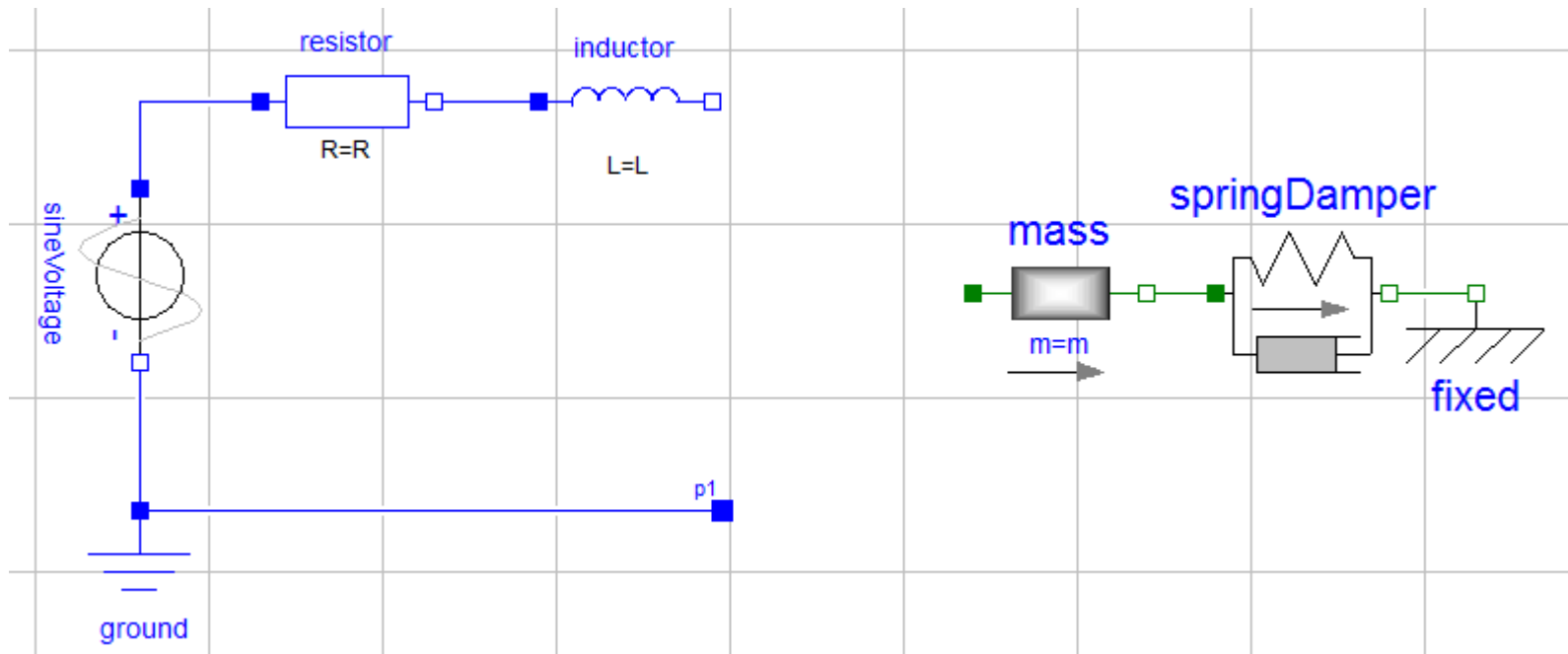
- Can be tedious and time-consuming, but good software support exists!
  - For example, Dymola does this automatically for overall system
- Transforming model into state-space form
  1. Choose state variables
  2. Express time derivatives of states in terms of states and inputs
  3. Express outputs in terms of states and inputs

# Implement model in computer

- Choose software
  - Low-level, high-level: C, Fortran, Matlab, Simulink, Modelica, ...
  - Higher-level software increases reusability, correctness, implementability
- To save time and avoid unnecessary errors: Always re-use models as much as you can
  - Use existing (verified) library model components, or
  - Modify and/or improve library model components
- Develop new model components only when this is not feasible
  - And consider making this a new library model (do the extra effort)

# Loudspeaker model in Modelica

- Most of the model can be constructed using components from Modelica Standard Library



- We need only to construct a component for the electro magnet

# Modelica model of electro magnet

- Need both electrical and mechanical (translational) connectors

```
connector Pin
```

```
  Voltage      v;
```

```
  flow Current i;
```

```
end Pin;
```

```
connector Flange
```

```
  Position     s "Absolute position of flange";
```

```
  flow Force f "cut force directed into flange";
```

```
end Flange;
```

# Modelica model of electro magnet

```
model ElectroMagnet
  extends Modelica.Electrical.Analog.Interfaces.OnePort;
    // OnePort defines variables v and i
  Modelica.Mechanics.Translational.Interfaces.Flange_b flange;

  parameter Modelica.SIunits.MagneticFluxDensity B = 1;
  parameter Modelica.SIunits.Length l = 1;

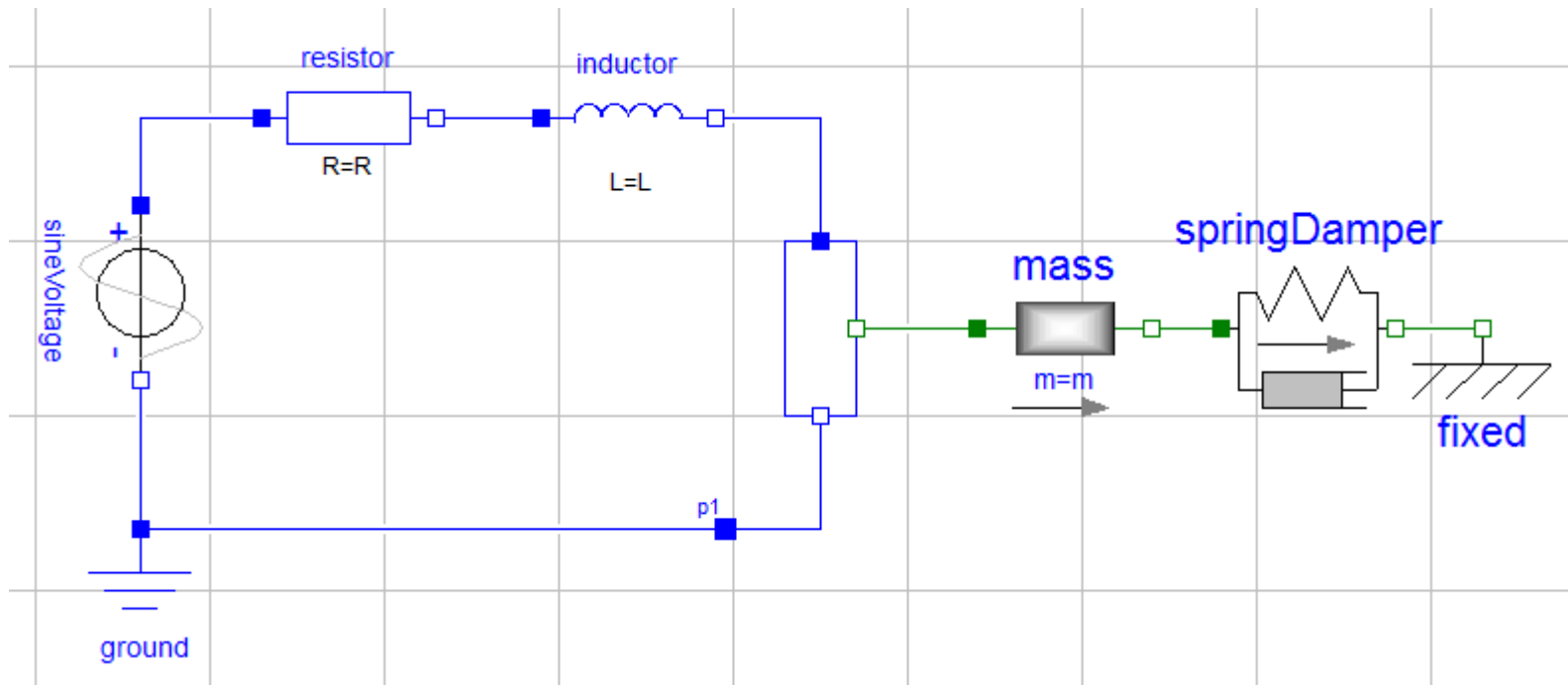
  Modelica.SIunits.Velocity velocity;
  Modelica.SIunits.Force F;

equation
  // Model
  v = B*l*velocity; // Voltage
  F = B*i*l; // Force

  // Set variables in connectors
  velocity = der(flange.s);
  flange.f = F;
end ElectroMagnet;
```

Note use of  
Modelica.SIunits!

# Complete loudspeaker model



# Validation of model

- Best: validate against real data (experiment)
- Soft validation also useful:
  - Verify that equations are dimensionally correct

**Example** Bernoullis law  $v = \sqrt{2gh}$ .

$$[v/\sqrt{gh}] = [ms^{-1}(ms^{-2}m)^{-1/2}] = m^0 s^0$$

- verify that equations describe qualitatively correct behavior
  - stationary points, static relations
  - linearized dynamics (time constants)

# Modeling purpose and complexity

- The purpose of the model determines its complexity
  - simple models often enough for order-of-magnitude estimates or “adequate” control performance
  - more detailed models may be required for critical design decisions or high performance designs
- Remember:
  - all models are approximate (in practice)
  - a highly accurate sub-model may be of limited value if other sub-models are inaccurate
  - a good model is simple, yet captures the essentials!
    - Einstein: “...as simple as possible, but not simpler”



# Simplification in modeling

Many approaches, for example

- Neglect minor effects
- Use idealized relationships
  - e.g., ideal gas law, non-compressible fluids, ...
- Aggregation of state variables
  - Approximate infinite-dimensional variables (systems of PDEs) by one or more «lumped» variables (systems of ODEs)
  - Example
    - The water temperature in a tank described by a single (average) temperature
    - The pressure in a long pipeline described by a single pressure (or a few pressures along the line)
- **Separate time constants**
  - Perhaps especially important for control applications

# Separation of time constants

Focus modeling effort on dynamics whose time constants are relevant for the intended purpose of the model

- subsystems with fast dynamics are approximated as static relations
- variables that vary slowly are approximated by constants

This gives models that are easier to manipulate and simulate

For example, in control applications:

- To verify stability (and performance) of a design, dynamics in bandwidth frequency range most important (remember Bode)
- Slow dynamics are not important for control *if its impact can be measured* (then integral control takes care of it anyway)

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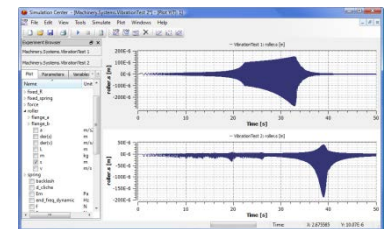
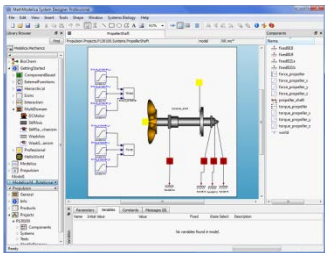
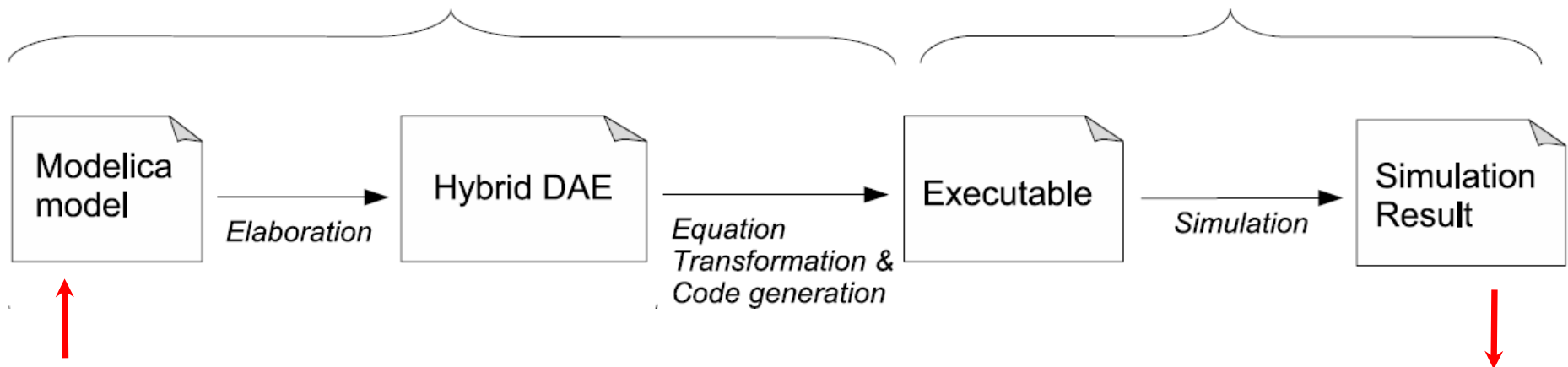
# Why learn about simulation methods?

1. You will need to implement your own solvers
  - What solver fits my problem, what time-step should I choose?
  - Primarily: Explicit solvers
  
2. You will need to make qualified choices of solvers when using advanced modeling software
  - What solver fits my problem, choice of accuracy?
  - Typically: Implicit solvers with varying time-steps
  
  - Examples:
    - Simulink: Three-body problem, satellite in combined moon and earth gravity field (orbit.mdl, ode45 vs ode1 (Euler))
    - Dymola

# Modelica/Dymola Simulation Process

“Static” semantics / compile time

“Dynamic” semantics / run time



# Linearization

(14.2.4)

- System  $\dot{y} = f(y, t)$ ,  $y = (y_1, \dots, y_d)^\top$
- Linearize around operating point  $y^*$ :  $\Delta\dot{y} = J\Delta y$ ,  $J = \left. \frac{\partial f}{\partial y} \right|_{y=y^*}$
- Diagonalize:  $Jm_i = \lambda_i m_i$ , where  $\begin{cases} m_i : \text{eigenvectors of } J \\ \lambda_i : \text{eigenvalues of } J \end{cases}$
- Define  $q = M^{-1}\Delta y$ :  

$$\dot{q} = M^{-1}J\Delta y = M^{-1}JMq = \Lambda q, \quad \Lambda = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_d \end{pmatrix}$$
- That is,  $\dot{q}_i = \lambda_i q_i$  from which we can find  $\Delta y(t) = Mq = \sum_{i=1}^d q_i(t)m_i$

We can study properties of a method used to simulate the system  $\Delta\dot{y} = J\Delta y$ , by study properties of the method for the systems  $\dot{q}_i = \lambda_i q_i$ ,  $i = 1, \dots, d$ .

# Example linearization

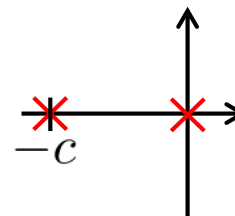
- System: Linearization about  $(y_1^*, y_2^*)^T$ :

$$\begin{aligned} \dot{y}_1 &= y_2 \\ \dot{y}_2 &= -y_1^3 - cy_2 \end{aligned} \quad \begin{pmatrix} \Delta \dot{y}_1 \\ \Delta \dot{y}_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -3(y_1^*)^2 & -c \end{pmatrix} \begin{pmatrix} \Delta y_1 \\ \Delta y_2 \end{pmatrix}$$

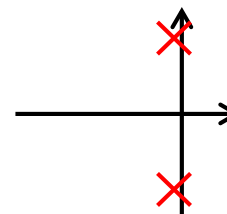
- Eigenvalues:

$$\lambda^2 + c\lambda + 3(y_1^*)^2 = 0 \quad \lambda_{1,2} = -\frac{c}{2} \pm \sqrt{\left(\frac{c}{2}\right)^2 - 3(y_1^*)^2}$$

$$y_1^* = 0 : \quad \lambda_1 = 0, \lambda_2 = -c$$



$$y_1^* = \text{large} : \quad \lambda_{1,2} \rightarrow \pm j\omega_0$$



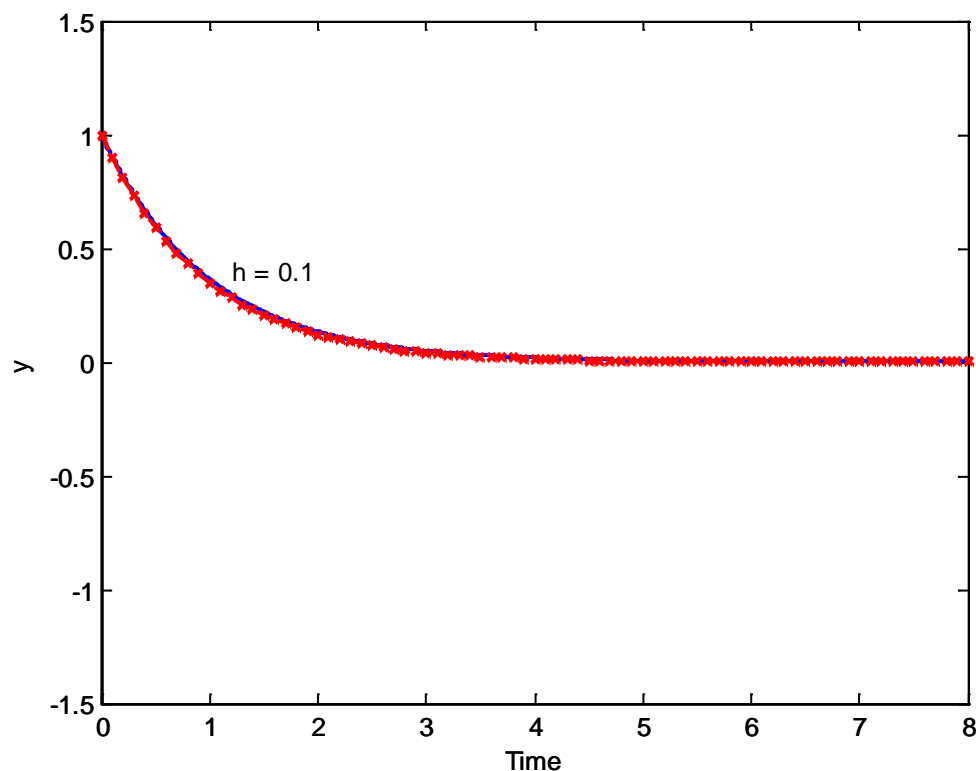
# Example Euler's method

ODE:  $\dot{y} = -y, \quad y(0) = 1$

Euler simulation:  $y_{n+1} = y_n + h(-y_n), \quad y_0 = 1$

Example,  $h = 0.1$ :

$n$	$t_n$	$y_n$
0	0	1
1	0.1	
2	0.2	
3	0.3	
4	0.4	
...	...	...





# Order (accuracy)

- Given IVP:

$$\dot{y} = f(y, t), \quad y(0) = y_0$$

- One-step method:

$$y_{n+1} = y_n + h\phi(y_n, t_n), \quad h = t_{n+1} - t_n$$

- If we can show that

$$y_{n+1} = y_n + hf(y_n, t) + \frac{h^2}{2} \frac{df(y_n, t)}{dt} + \dots + \frac{h^p}{p!} \frac{d^{p-1}f(y_n, t)}{dt^{p-1}} + O(h^{p+1})$$

- Then:

- Local error is  $O(h^{p+1})$
- Method is order  $p$

# Example Euler's method

ODE:  $\dot{y} = -y, \quad y(0) = 1$

Euler simulation:  $y_{n+1} = y_n + h(-y_n), \quad y_0 = 1$

Stability:  $|R(h\lambda)| = |1 - h| \leq 1 \Rightarrow 0 \leq h \leq 2$

