

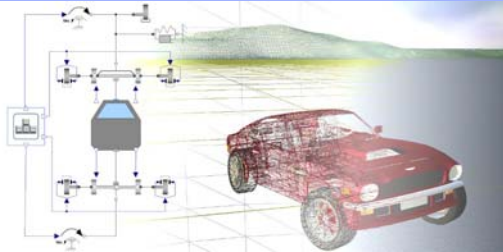
Virtual Physics Equation-Based Modeling

TUM, October 07, 2014

Equation-based modeling: first steps

```
equation
sx0 = cos(frame_a.phi)*sx_norm + ...
sy0 = -sin(frame_a.phi)*sx_norm + ...
vy = der(frame_a.y);
w_roll = der(flange_a.phi);
v_long = vx*sx0 + vy*sy0;
v_lat = -vx*sy0 + vy*sx0;
v_slip_lat = v_lat - 0;
v_slip_long = v_long - R*w_roll;

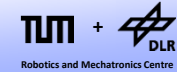
v_slip = sqrt(v_slip_long^2 + ...
-f_long*R + flange_a.tau);
frame_a.t = 0;
f = N* S_Func(vAdhesion,vSlide,...
f_long = f*v_slip_long/v_slip;
f_lat = f*v_slip_lat/v_slip;
f_long = frame_a.fx*sx0 + ...
f_lat = -frame_a.fy*sy0 + ...
```



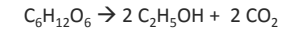
Dr. Dirk Zimmer

German Aerospace Center (DLR), Robotics and Mechatronics Centre

Modeling Example

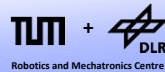


- Let us start with a simple modeling example:
Let us brew beer! (or ferment wine.. for the non-ba(rb/v)arians)
- In this example, we are going to model the fermentation of sugar into alcohol and the corresponding growth and decay of yeast.
- In the process of fermentation each molecule of sugar is transformed into a molecule of alcohol (plus CO₂)



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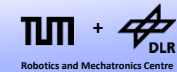
Variables and Parameters



- | | |
|--|--|
| <ul style="list-style-type: none"> These are our model variables: Population of yeast: p Birth-Rate: b Death-Rate: d Concentration of sugar: s Concentration of alcohol: a Consumption of sugar: f Current Temperature: T | <ul style="list-style-type: none"> These are our model parameters: Volume of vessel: $V = 1$ Initial concentration of sugar: $s_0 = 0.2$ Initial population of yeast: $p_0 = 0.001$ Feeding-Rate Coefficient: C_f Reproductivity: R Sensitivity to poison: S Reference Temperature: T_{ref} |
|--|--|

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Algebraic Equations



Let us start with the algebraic equations:

- The consumption of sugar (f) is proportional to concentration of sugar (s) multiplied by the population of yeast (p). The proportionality is determined by the feeding-rate (C_f) and the temperature (T)

$$f = s \cdot p \cdot C_f \cdot (T/T_{ref})$$

- Since each molecule of alcohol was transformed from one molecule of sugar at time 0 (s_0), the current concentration of alcohol (a) is:

$$a = s_0 - s$$

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Algebraic Equations

Let us continue with the algebraic equations:

- The Birth-Rate is proportional to concentration of sugar (s). The proportionality is determined by the reproduction (R):

$$b = R \cdot s$$

- The Death-Rate is dependent on the level of poisonous alcohol (a) and the sensitive (S) of the yeast.

$$d = S \cdot a$$

Algebraic Equations

The algebraic equations are:

$$f = s \cdot p \cdot C_f \cdot (T/T_{ref})$$

$$a = s_0 - s$$

$$b = R \cdot s$$

$$d = S \cdot a$$

T is determined from outside (input-variable)

Differential Equations

The differential equations describe the change over time:

- The change in population (dp/dt) equals the birth-rate (b) minus the death rate (d) and is proportional to the current population (p):

$$dp/dt = p \cdot (b-d)$$

- The change in concentration of sugar (ds/dt) multiplied by the Volume (V) equals the negative consumption rate (f) of sugar :

$$V \cdot ds/dt = -f$$

or

$$ds/dt = -f/V$$

Differential Equations

The differential equations are:

$$dp/dt = p \cdot (b-d)$$

$$ds/dt = -f/V$$

Differential Equations

Let us plug in the algebraic equations:

$$dp/dt = p \cdot (b-d)$$

$$ds/dt = -f/V$$

Differential Equations

Let us plug in the algebraic equations:

$$dp/dt = p \cdot (b-d)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot a)$$

$$ds/dt = -f/V$$

Differential Equations

Let us plug in the algebraic equations:

$$dp/dt = p \cdot (b-d)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot a)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot (s_0 - s))$$

$$ds/dt = -f/V$$

Differential Equations

Let us plug in the algebraic equations:

$$dp/dt = p \cdot (b-d)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot a)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot (s_0 - s))$$

$$\underline{dp/dt = p \cdot ((R+S) \cdot s - S \cdot s_0)}$$

$$ds/dt = -f/V$$

Differential Equations

Let us plug in the algebraic equations:

$$dp/dt = p \cdot (b-d)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot a)$$

$$dp/dt = p \cdot (R \cdot s - S \cdot (s_0 - s))$$

$$\underline{dp/dt = p \cdot ((R+S) \cdot s - S \cdot s_0)}$$

$$ds/dt = -f/V$$

$$\underline{ds/dt = -s \cdot p \cdot C_f \cdot (T/T_{ref}) \cdot 1/V}$$

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Differential Equations

Let us plug in the algebraic equations:

$$\underline{dp/dt = p \cdot ((R+S) \cdot s - S \cdot s_0)}$$

$$\underline{ds/dt = -s \cdot p \cdot C_f \cdot (T/T_{ref}) \cdot 1/V}$$

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Time Discretization

- Let us discretize the advance of time by the quantum h :

- Given x_t , we can compute x_{t+h} by using the Taylor-series expansion:

$$x_{t+h} = x_t + (dx/dt)_t \cdot h + (dx/dt^2)_t \cdot (h^2/2) + (dx/dt^3)_t \cdot (h^3/6) + \dots$$

- Let us drop all higher derivatives. We get:

$$x_{t+h} = x_t + (dx/dt)_t \cdot h$$

- This discretization scheme is called: **Forward Euler**

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Time Discretization

Let us apply Forward Euler to our differential equations:

$$p_{t+h} = p_t + (dp/dt)_t \cdot h$$

with

$$(dp/dt)_t = p_t \cdot ((R+S) \cdot s_t - S \cdot s_0)$$

$$s_{t+h} = s_t + (ds/dt)_t \cdot h$$

with

$$(ds/dt)_t = -s_t \cdot p_t \cdot C_f \cdot (T_t/T_{ref}) \cdot 1/V$$

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Simulation

- These four explicit equations are used to perform a simulation:

$$p_{t+h} = p_t + (dp/dt)_t \cdot h \text{ with } (dp/dt)_t = p_t \cdot ((R+S) \cdot s_t - S \cdot s_0)$$

$$s_{t+h} = s_t + (ds/dt)_t \cdot h \text{ with } (ds/dt)_t = -s_t \cdot p_t \cdot C_f \cdot (T_t/T_{ref}) \cdot 1/V$$

- We can simply punch them into a Python3 script:

```
while time < 10:
    dp_dt = p*((R+S)*s - S*s0)
    ds_dt = -s*p*C_f*(T/T_ref)*1/V
    p += h*dp_dt
    s += h*ds_dt
    a = s0-s
    time += h
    print(time, "\t", p, "\t", s, "\t", a)
```

- Here, there are computed within a loop. Each iteration represents one time-step: an advance of h in time.

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Simulation Code

This is the complete Python3-Script:

```
#!/usr/bin/env python3
#Setting the input-value
T = 310

#Setting the parameters
V = 1 #volume of fermentation vessel
s0 = 0.2 #initial concentration of sugar
p0 = 1e-6 #initial population of yeast [m3]
C_f = 50 #feeding Coefficient [1/day]
R = 10 #reproductivity [1/day]
S = 15 #sensitivity w.r.t. alcohol [1/day]
T_ref = 300 #reference temperature [K]
h = 0.01 #time-step of forward Euler integration

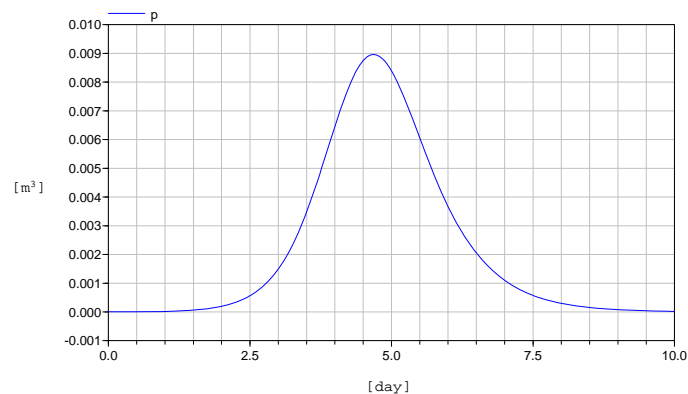
#Setting the initial values
p = p0
s = s0
a = s0 - s;
time = 0

#perform time-integration
while time < 10:
    dp_dt = p*((R+S)*s - S*s0)
    ds_dt = -s*p*C_f*(T/T_ref)*1/V
    p += h*dp_dt
    s += h*ds_dt
    a = s0-s
    time += h
    print(time, "\t", p, "\t", s, "\t", a)
```

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Simulation Results

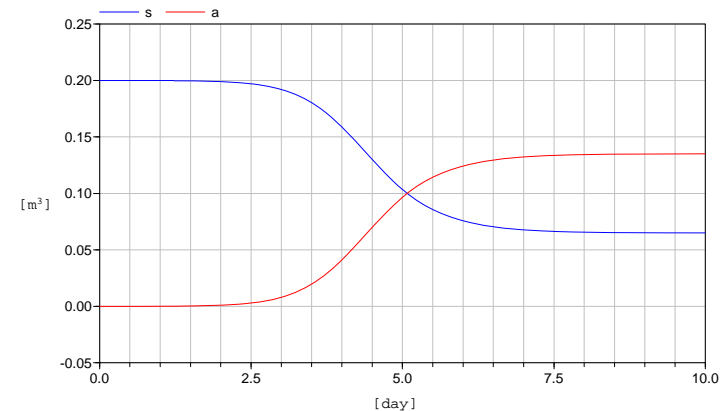
- And this is the result for the yeast population:



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Simulation Results

- Concentration of sugar and alcohol:



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Simulation Results

Interpretation of the simulation results:

- The population of yeast first grows exponentially. There seems to be an endless supply of sugar available.
- Then the population has reached a critical level and the concentration of sugar and alcohol are rapidly changing.
- Then, there is a sudden die-off due to the combination of starvation and self-poisoning.

State-Space Form

Let us look at the computational structure of our model. We can classify our variables into vectors of...

- Input Variables: $\mathbf{u} = (T)$
- State Variables: $\mathbf{x} = (p, s)$
- State Derivatives: $d\mathbf{x}/dt = (dp/dt, ds/dt)$
- Output Variables: $\mathbf{y} = (a)$
- The system was then transformed into two functions:

$$\begin{aligned} d\mathbf{x}/dt &= f(\mathbf{x}, \mathbf{u}, t) \\ \mathbf{y} &= g(\mathbf{x}, \mathbf{u}, t) \end{aligned}$$

- This specific form is called: **state-space form**

State-Space Form

- We form a row vector out of \mathbf{x}, \mathbf{u} , and t : (p, s, T, t)
- We form a column vector out of $d\mathbf{x}/dt$ and \mathbf{y} : $(dp/dt, ds/dt, a)$
- Now we can represent the dependences of our computational structure by a Boolean incidence matrix.

		x		(u, t)		
		p	s	T	t	
dx/dt	dp/dt	X	X			f(...)
	ds/dt	X	X	X		
y	a		X			g(...)

$$\begin{aligned} dp/dt &= p \cdot ((R+S) \cdot s - S \cdot s0) \\ ds/dt &= -s \cdot p \cdot C_f \cdot (T/T_{ref}) \cdot 1/V \\ a &= s0 - s \end{aligned}$$

State-Space Form

- The incidence matrix can be decomposed into four blocks: **A, B, C, D**.
- If $g(\dots)$ and $f(\dots)$ represent linear functions (not the case here!), the system can indeed be expressed by real-valued matrices:

$$\begin{aligned} d\mathbf{x}/dt &= \mathbf{A}\mathbf{x} + \mathbf{B}(\mathbf{u}, t) \\ \mathbf{y} &= \mathbf{C}\mathbf{x} + \mathbf{D}(\mathbf{u}, t) \end{aligned}$$

	p	s	T	t
dp/dt	X	X		
ds/dt	X	X	X	
a		X		

Summary

Let us summarize the development process of our simulation:

- First, we had to analyze our model and select the variables of interest.
- Then, we formulated a set of *differential-algebraic equations* (DAEs).
- Next, we had to transform this set of expressions into a computable/solvable form (*state-space form*).
- Finally, a time-discretization scheme was applied and a numerical integration could be performed (*numerical ODE-solver*).

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Deficiencies

Even for this small and rather trivial example, this development process was rather laborious.

- Larger models cause much more work.
- Also there are more complicated models that are difficult to transform into state-space form.
- If we change the model, the complete process has to be redone.
- Programming a simulation manually turns out to be very inconvenient and is also very error-prone.
- For these reasons, a number of computer languages have been developed that aim to automate this process.
- Let us take a look back in history...

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MIMIC (History)

- The language MIMIC was developed mainly for the Control Data super-computers in 1964.
- The listing presents the MIMIC code for the simulation of a swinging pendulum.
- Successors of these language were CSMP and ACSL. They prevailed up to the 80s.

```
CON(G)      Declaration of constants
PAR(1X0,X0) Declaration of parameters
DT 0.05     Definition of time step

1X INT(-G*Z,1X0) Integration
X INT(1X,X0)
Y 1.-COS(X)  Equation for y position
Z SIN(X)     Equation for z position

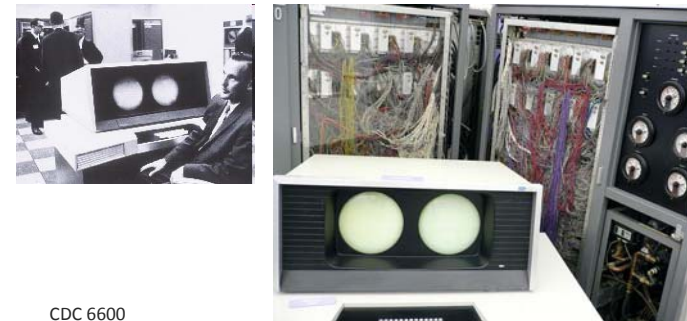
FIN(T,4.9)  Command for integration

PLO(T,X,Y,Z) Commands for plotting
ZER(0.,-5,0.,-1)
SCA(5.,5.,2.,1.)

END End of program
```

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MIMIC



CDC 6600

- 40 MHz, roughly 1MFLOPS, 64K 60-bit words of memory
- Roughly 400'000 transistors, over 100 miles of wiring
- A predecessor of the RISC-Architecture. Developed by Seymour Cray
- Prize: 7 – 10 Million \$ (and by that time, the dollar was worth something)

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MIMIC (Advantages)

- The model could be formulated by assignments and integrators.
- These model "equations" could be arbitrarily ordered.
- The appropriate order for the state-space form is automatically derived.
- The time-discretization is not part of the model anymore. Different numerical ODE-solvers can be applied (better than FE)

```

CON ( G )      Declaration of constants
PAR ( 1X0 , X0 ) Declaration of parameters
DT 0.05       Definition of time step

1X INT(-G*Z,1X0) Integration
X INT(1X,X0)
Y 1.-COS(X)   Equation for y position
Z SIN(X)      Equation for z position

FIN(T,4.9)    Command for integration

PLO(T,X,Y,Z)  Commands for plotting
ZER(0.,-5,0.,-1)
SCA(5.,5.,2.,1.)

END End of program
    
```

MIMIC (Deficiencies)

- MIMIC could not handle real equations, only causal assignments.
- There were hardly any means to structure the program. The language was almost completely flat and there is only one global namespace.

```

CON ( G )      Declaration of constants
PAR ( 1X0 , X0 ) Declaration of parameters
DT 0.05       Definition of time step

1X INT(-G*Z,1X0) Integration
X INT(1X,X0)
Y 1.-COS(X)   Equation for y position
Z SIN(X)      Equation for z position

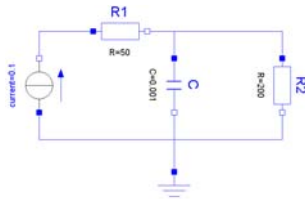
FIN(T,4.9)    Command for integration

PLO(T,X,Y,Z)  Commands for plotting
ZER(0.,-5,0.,-1)
SCA(5.,5.,2.,1.)

END End of program
    
```

Dymola

- The **Dynamic Modeling Language** was developed by Hilding Elmquist in 1978.
- The listing on the left displays the code of an assembled electric circuit and of its capacitor component.



```

model type capacitor
cut A (Va / I) B (Vb / -I)
main cut C [A B]
main path P <A - B>
local V
parameter C
V = Va -Vb
C*der(V) = I
end

model Network
submodel ( resistor ) R1 R2
submodel ( capacitor ) C
submodel ( current ) F
submodel Common
input i
output y
connect Common to F to R1 to (C par R2)
to Common
E.I = i
y = R2.Va
end
    
```

Dymola

- Dymola is a **declarative** language. It only contains code for the model-equations. The simulation is completely decoupled from the model description.
- This language enabled the formulation of hierarchic elements such as sub-components.
- These components could be automatically connected.

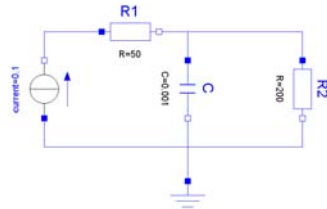
```

model type capacitor
cut A (Va / I) B (Vb / -I)
main cut C [A B]
main path P <A - B>
local V
parameter C
V = Va -Vb
C*der(V) = I
end

model Network
submodel ( resistor ) R1 R2
submodel ( capacitor ) C
submodel ( current ) F
submodel Common
input i
output y
connect Common to F to R1 to (C par R2)
to Common
E.I = i
y = R2.Va
end
    
```


Dymola

- Dymola can handle non-causal equations such as $u = R \cdot i$



- In R1, the causality is: $u := R \cdot i$
- In R2, the causality is: $i := u/R$
- In Dymola, one can use the same, non-causal equations for both resistor components.

```

model type capacitor
cut A (Va / I) B (Vb / -I)
main cut C [A B]
main path P <A - B>
local V
parameter C
V = Va - Vb
C*der(V) = I
end

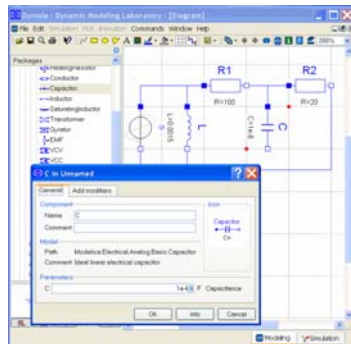
model Network
submodel ( resistor ) R1 R2
submodel ( capacitor ) C
submodel ( current ) F
submodel Common
input i
output y
connect Common to F to R1 to (C par R2)
to Common
E.I = i
y = R2.Va
end
    
```

Omola

- Dymola* never had any real impact in industry, it remained within academia.
- There, its main ideas were preserved and extended by *Omola*. This language enables a truly object-oriented modeling, featuring inheritance, wrapping etc.
- Modeling in *Omola* was also performed graphically. Only the fundamental equations are entered in textual form. All higher-level model are assembled graphically.
- Also *Omola* remained within academia. Things started to change as *Modelica* was born in 1997.

Modelica

Demonstration



Modelica

- As you see: *Dymola* is still alive, but not as modeling language but as an M&S environment for *Modelica*.
- In *Modelica*, we can directly punch in our model equations.
- There is no need anymore to derive the state-space form by paper and pencil.

```

model Yeast
parameter Real V = 1 "volume of fermentation vessel";
parameter Real s0 = 0.2 "initial concentration of sugar";
parameter Real p0 = 1e-6 "initial population of yeast";
parameter Real C_f = 50 "Feeding Coefficient [1/day]";
parameter Real R = 10 "Reproductivity [1/day]";
parameter Real S = 15 "Sensitivity w.r.t. alcohol [1/day]";
parameter Real T_ref = 300 "reference temperature";
Real p "population of yeast";
Real b "birth rate";
Real d "death rate";
Real s "concentration of sugar";
Real a "concentration of alcohol";
Real f "consumption of sugar (feeding)";
Real T "current temperature";

initial equation
p = p0;
Capacitor s = s0;
equation
f = s * p * C_f * (T/T_ref);
a = s0 - s;
b = R * s;
d = S * a;
T = 310;
der(p) = p*(b-d);
V*der(s) = -f;
end Yeast;
    
```

Modeling and Simulation

- In the field of programming languages, there are high-level languages (Python, C++) and low-level languages (Assembler)
- The same is true for modeling languages.
- The state-space form is a common target of their compilation scheme (the Assembler language of a modeler).

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Modeling and Simulation

- The first (and larger) part of the lecture concerns the modeling side.
- You will learn to model in Modelica using the software Dymola.

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Modeling and Simulation

The way up:

- To this end, we have to learn how to formulate the laws of physics in an object-oriented way.

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Modeling and Simulation

The way up:

- To this end, we have to learn how to formulate the laws of physics in an object-oriented way.
- This is a sole matter of physics. It has nothing to do with computer science.

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Modeling and Simulation

The way down:

- Then we have to learn how the languages are compiled, and how the state-space form is automatically derived.

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Modeling and Simulation

The way down:

- Then we have to learn how the languages are compiled, and how the state-space form is automatically derived.
- This is a sole matter of computer science. It has nothing to do with physics.

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Modeling and Simulation

The way down:

- Then we have to learn how the languages are compiled, and how the state-space form is automatically derived.
- This is a sole matter of computer science. It has nothing to do with physics.
- (And by the way, we are going to model a lot of cool systems...)

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Modeling and Simulation

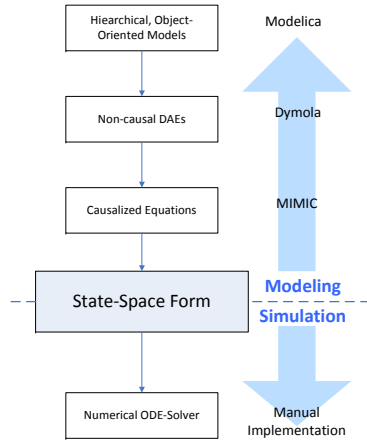
- The second (and smaller) part of this lecture series concerns simulation.
- You will learn different techniques how to implement numerical ODE solvers, and how they influence the simulation result.
- In addition, the handling of events will be discussed.

Numerical ODE-Solver

Manual Implementation

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Modeling and Simulation



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Questions?