## DEALING WITH COMPLEX MODELS AND HOW TO USE THE IDEALIZATION OF PHYSICS TO OUR ADVANTAGE

Keynote at the 15th International Modelica Conference in Aachen
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What we regard as necessary:

- A necessary condition is:

The equations whose solution represent a physical system can be distributed among its components


## What we regard as necessary:

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The equations whose solution represent a physical system can be distributed among its components


## What we may regard as sufficient:

- A necessary condition is:

The equations whose solution represent a physical system can be distributed among its components

- But more than that, we like to state a sufficient condition:

Any valid combination of components (under rules of limited complexity) shall have a solution representing a physical system.

## What we may regard as sufficient:



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## What we may regard as sufficient:



Any valid combination of components (under rules of limited complexity) shall have a solution representing a physical system.


## WHAT IS NECESSARY?

## A very simple Example to begin with...

- Going from the actual physical system to the mathematical model contains many, many implicit assumption that are hardly ever discussed...



## Simulating the microverse:

- Maybe we can simulate the macroscopic system by applying the rules of quantum physics. Here we progress by a sum of quantum events (decoherences)

$$
\sum_{k}^{k=t / t_{d}}|\psi\rangle_{k} \rightarrow\left|\phi_{i}\right\rangle_{k}
$$

- There is an estimation formula for the average time between such events for macroscopic systems:

$$
t_{d}=t_{R} \frac{\hbar^{2}}{2 m k_{B} T(\Delta x)^{2}}
$$

- Plugging in our parameters, yields for $t=10$ :

$$
k=10^{48}
$$

## Why can we simulate macroscopic systems?

- Fortunately, quantum physics fulfills the principle of stationary action:

$$
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} T(\dot{q}(t))-V(q(t), \dot{q}(t)) d t=0
$$



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\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \underbrace{T(\dot{q}(t))}-\underbrace{V(q(t), \dot{q}(t))} d t=0 *
$$

Kinetic Energy Potential Energy

Lagrangian: L

$$
\begin{aligned}
& L(q, \dot{q})=T-V \\
& S=\int_{t_{a}}^{t_{b}} L d t
\end{aligned}
$$

Action: S

* for the conservative case



## Why can we simulate macroscopic systems?

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$$

> Kinetic Energy Potential Energy

Lagrangian: L

$$
L(q, \dot{q})=T-V
$$

Action: $S$

$$
S=\int_{t_{a}}^{t_{b}} L d t
$$



[^0]
## Can we solve this system reliably?

- It is hard to make any statement about the potential energy:

$$
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \underbrace{T(\dot{q}(t))}_{\text {Kinetic Energy }}-\underbrace{V(q(t), \dot{q}(t))}_{\text {Potential Energy }} d t=0
$$

- In Modelica we mostly state the gradient of the potential energy. Hence, at least $V$ is continuous.
- To illustrate the potential complexity, the example on the right is arbitrarily chosen.



## Can we solve this system reliably?

- Fortunately, kinetic energy has very beneficial properties:

$$
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} T(\dot{q}(t))-\underbrace{V(q(t), \dot{q}(t))}_{\text {Kinetic Energy Potential Energy }} d t=0
$$

- It ensures continuity of $q(t)$
- It promotes locality of $\dot{q}(t)$
- It cannot be overpowered by $V$ and is always inambiguous

The Kinetic Energy has a special role!

It is unlike all other forms of energy


## Can we solve this system reliably?

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\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} T(\dot{q}(t))-\underbrace{V(q(t), \dot{q}(t))}_{\text {Kinetic Energy Potential Energy }} d t=0
$$



## How do I solve principle of stationary action then?

- The direct method is to apply the Euler-Lagrange equations:

$$
\qquad \frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \underbrace{}_{\text {Kinetic Energy } \underbrace{T(\dot{q}(t))}_{\text {Potential Energy }}-\underbrace{V(q(t), \dot{q}(t))}_{L(q)} d t=0}
$$

- Unfortunately, the Lagrangian view results in a solution that cannot be distributed among its components.

$$
\text { that } \frac{\partial L}{\partial q}-\frac{d}{d t} \frac{\partial L}{\partial \dot{q}}=0
$$

## How to solve in distributable form?

- Fortunately, Sir William Hamilton created another formulation that doubles the (resulting) dimension :

■

$$
L(q, \dot{q}) \rightarrow H(q, p)
$$

- ... by introducing the generalized potential:

$$
p_{i}=\frac{\partial L}{\partial \dot{q}^{i}}
$$

- The Hamiltonian is then expresses the total energy*:

$$
H=T+V
$$

[^1]
## How to solve in distributable form?

- This leads to a reformulation of the action and a solution based on the Hamiltonian:

$$
\left.\begin{array}{l}
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \underbrace{\sum_{i} p_{i} q^{i}}_{i=1}-(T(p(t))+\underbrace{V(q(t), p(t))}_{\text {Kinetic Energy Potential Energy }}) d t=0 \\
\text { Hamiltonian: } H \quad H(q, p)=T+V
\end{array}\right] \quad \begin{aligned}
& \frac{d q}{d t}=\frac{\partial H}{\partial p} \\
& \frac{d p}{d t}=-\frac{\partial H}{\partial q}
\end{aligned}
$$

## How to solve in distributable form?

Let us go through one example: The pressure wave in a pipe

- Our path is expressed by the integral of volume flow: $q(t)=Q$
- Hence also: $\dot{q}(t)=\dot{Q}$
- Which means for the kinetic energy: $T=\frac{I \rho}{2} \dot{Q}^{2} \quad\left(\right.$ with $\left.I=\int \frac{d s}{A}\right)$
- The potential energy is: $V=\frac{K}{2 Q_{r e f}} Q^{2} \quad$ (if $Q$ and $\ell_{\text {ref }}$ are close and fluid incompressibe)

$$
\left.\begin{array}{l}
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \underbrace{\sum_{i} p_{i} q^{i}}_{\text {2x Kinetic Energy }}-\underbrace{(T(p(t))}_{\text {Kinetic Energy }}+\underbrace{V(q(t), p(t))}_{\text {Potential Energy }}) d t=0 \\
\text { Hamiltonian: } H \quad H(q, p)=T+V
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\end{aligned}
$$

## How to solve in distributable form?

- The Lagrangian $L=\frac{I \rho}{2} \dot{Q}^{2}-\frac{K}{2 Q_{r e f}} Q^{2}$
- Now for $p=\frac{\partial L}{\partial \dot{q}}=\frac{\partial\left(\frac{I \rho}{2} \dot{Q}^{2}-\frac{K}{2 Q_{r e f}} Q^{2}\right)}{\partial \dot{Q}}=I \rho \dot{Q}$
- Then: $H=\frac{1}{2} \frac{p^{2}}{I \rho}+\frac{K}{2 Q_{\text {ref }}} Q^{2}$

$$
\left.\begin{array}{l}
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \underbrace{\sum_{i} p_{i} q^{i}}_{\text {2x Kinetic Energy }}-\underbrace{(T(p(t))}_{\text {Kinetic Energy }}+\underbrace{V(q(t), p(t))}_{\text {Potential Energy }}) d t=0 \\
H(q, p)=T+V
\end{array}\right] \quad \begin{aligned}
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& \text { Hamiltonian: } H \quad \frac{d p}{d t}=-\frac{\partial H}{\partial q}
\end{aligned}
$$

## How to solve in distributable form?

- Hamiltonian: $H=\frac{1}{2 I \rho} p^{2}+\frac{K}{2 Q_{r e f}} Q^{2}$
- Plugging this into the Hamiltonian equation yields:

$$
\begin{aligned}
& \frac{d q}{d t}=\dot{Q}=\frac{\partial\left(\frac{1}{2} \frac{p^{2}}{I \rho}+\frac{K}{2 Q_{r e f}} Q^{2}\right)}{\partial p}=\frac{p}{I \rho} \\
& \frac{d p}{d t}=-\frac{\partial\left(\frac{1}{2} \frac{p^{2}}{I \rho}+\frac{K}{2 Q_{r e f}} Q^{2}\right)}{\partial q}=-\frac{Q}{\kappa Q_{r e f}}
\end{aligned}
$$

$$
\Rightarrow \quad \begin{aligned}
& \dot{Q}=\frac{p}{I \rho} \\
& \frac{d p}{d t}=-\frac{K}{Q_{r e f}} Q
\end{aligned}
$$

$$
\begin{aligned}
& \frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} \sum_{i} p_{i} \dot{q}^{i}-(T(p(t))+\underbrace{V(q(t), p(t)))}_{\text {2x Kinetic Energy }} d t=0 \\
& \text { Hamiltonian: H } H \quad \frac{d q}{d t}=\frac{\partial H}{\partial p} \\
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- Hamiltonian: $H=\frac{1}{2 I \rho} p^{2}+\frac{K}{2 Q_{r e f}} Q^{2}$
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$$

$$
\Rightarrow \quad \begin{gathered}
\ddot{Q} I \rho=\frac{d p}{d t} \\
\frac{d p}{d t}=-\frac{K}{Q_{r e f}} Q
\end{gathered}
$$

- It may help to define the pressure $P=\frac{d p}{d t}$
- This form may now be distributed.

$$
\frac{d \dot{\boldsymbol{Q}}}{d t} I \rho=P
$$

$$
\frac{d P}{d t} \frac{Q_{r e f}}{K}=-\dot{Q}
$$

## The pairs of Potential and Flow

- The equations are distributed to components:

$$
-\frac{d \dot{Q}}{d t} I \rho=\Delta P
$$

$$
\text { -T T } \quad \frac{d P}{d t} \frac{Q_{r e f}}{K}=-\dot{Q}
$$

- The doubling of the resulting dimensions by $H$ created our beloved pairs of potential and flow:
- $d p / d t$, here Pressure $\boldsymbol{P}$
- $q$ or sometimes $d q / d t$ as here with the Volume flow $\dot{\boldsymbol{Q}}$



## Modeling Additional Components

- The equations are distributed to components

$$
-\frac{d \dot{Q}}{d t} I \rho=\Delta P
$$



- Using this pair, we can model further components:

$$
\cdots \frac{\dot{Q}|\dot{Q}|}{\dot{Q}_{r e f}^{2}}=\Delta P
$$



## Composing the complete system

- Finally we can simulate the complete system



## Modeling is a lossy compression of reality

Quantum Events

## ODE Approach

$\sum_{k}^{k=t / t_{d}}|\psi\rangle_{k} \rightarrow\left|\phi_{i}\right\rangle_{k} \quad \mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}\right)$

$$
k>10^{48}
$$

$$
\begin{aligned}
& k=5 \cdot 10^{7} \\
& \operatorname{dim}(\mathbf{x})=9
\end{aligned}
$$

- Principle of stationary action
- Law of large numbers
- Homogeneity Assumptions
- Discretization of Space
- Continuity Assumption
- Floating Point Approximation
- Discretization of Time


## Recapitulation

- The principle of stationary is the basis for classic physics

$$
S=\int_{t_{a}}^{t_{b}} L d t
$$

- The presence of kinetic energy ensures solvability of this principle
- Being based on the derivative $\dot{q}$, ensures continuity.
- Its quadratic nature promotes locality of the solution
- Having a bijective and unbounded gradient field means that the kinetic energy can never be overpowered.
- The Hamiltonian view $(q, \dot{q}) \rightarrow(q, p)$ doubles the resulting dimension and enables the distribution of the equations and gives rise to the pairs of effort and flow for the components interface.


$$
L(q, \dot{q}) \rightarrow H(q, p)
$$



## Recombining the result

- There is an easy way to improve efficiency. Let us simply forget about modeling the pipe:



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## Recombining the result

- This is extremely effective. We are much faster now.
- But beware, we now have a non-linear equation system to solve:
- The pressure level below the valves needs to balance out the flows.




## Modeling is a lossy compression of reality

## Quantum Events

$\sum_{k}^{k=t / t_{d}}|\psi\rangle_{k} \rightarrow\left|\phi_{i}\right\rangle_{k} \quad \mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}\right)$

DAE Approach

$$
k=5 \cdot 10^{7}
$$

$$
\operatorname{dim}(\mathbf{x})=9
$$

$$
\begin{gathered}
\mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}, \mathbf{s}_{k, i}\right) \\
\mathbf{s}_{k, i} \in\left\{\mathbf{s}_{k} \mid 0=g_{k}\left(\mathbf{s}_{k}\right)\right\} \\
k=10^{1} \\
\operatorname{dim}(\mathbf{x})=3
\end{gathered}
$$

$$
\xrightarrow{\eta>10^{40}}
$$

- Principle of stationary action
- Law of large numbers
- Homogeneity Assumptions
- Discretization of Space
- Continuity Assumption
- Floating Point Approximation
- Discretization of Time
- 


## What can we say about the solvability?

- Can we make a general statement about the solvability?

$$
\frac{\partial}{\partial q(t)} \int_{t_{a}}^{t_{b}} T<-\underbrace{V(q(t), \dot{q}(t))}_{\text {Kinetic Energy }} d t=0
$$

- The new system does not contain any model for the kinetic energy anymore.
- We have removed exactly the part that helped to ensure solvability.
- There is actually not a meaningful Lagrangian anymore! Does this still represent a physical system?


## Practical Application Example 1: Inertia of Turbine Shaft



- This thermodynamic process yields a non-linear system that can be very difficult to solve.
- We need to find the power balance between compressor, fan and turbine
- Adding an inertia to the shaft connecting turbine and compressor helps enormously

Practical Application Example 2: Constrained Mechanics


## Practical Application Example 2: Constrained Mechanics

- At the point of maximum extension the kinetic energy is constrained.
- The kinetic energy cannot be transferred since all bodies are assumed to be rigid.
- Elasticity needs to be added


## Preliminary Conclusions

- To ensure solvability, kinetic energy plays a vital role
- The Hamiltonian Form which is the basis for our standard interfaces is indifferent to kinetic energy and may mislead us.

$$
\begin{gathered}
n \text { dimensional result } \\
L(q, \dot{q})=T-V \\
\text { kinetic energy is special }
\end{gathered}
$$

```
2n dimensional result
    H(q,p) =T+V
kinetic energy is one among many
```

- The Hamiltonian offers a very general expression for the conservation of total energy but it is the action that tells us how to solve the system.



## WHAT IS SUFFICIENT?

## Modeling is a lossy compression of reality

## Quantum Events

$\sum_{k}^{k=t / t_{d}}|\psi\rangle_{k} \rightarrow\left|\phi_{i}\right\rangle_{k} \quad \mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}\right)$

DAE Approach

$$
\begin{gathered}
\mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}, \mathbf{s}_{k, i}\right) \\
\mathbf{s}_{k, i} \in\left\{\mathbf{s}_{k} \mid 0=g_{k}\left(\mathbf{s}_{k}\right)\right\} \\
k=10^{1} \\
\operatorname{dim}(\mathbf{x})=3
\end{gathered}
$$

$$
k=5 \cdot 10^{7}
$$

$$
\operatorname{dim}(\mathbf{x})=9
$$

$$
\xrightarrow{\eta>10^{40}}
$$

- Principle of extremal action
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## Modeling is a lossy compression of reality

## Quantum Events

$$
\sum_{k}^{k=t / t_{d}}|\psi\rangle_{k} \rightarrow\left|\phi_{i}\right\rangle_{k} \quad \mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}\right)
$$

## DAE Approach

$$
\begin{array}{ll}
k>10^{47} & k=5 \cdot 10^{7} \\
& \operatorname{dim}(\mathbf{x})=9
\end{array}
$$

$$
\mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}, \mathbf{s}_{k, i}\right)
$$

$$
\mathbf{s}_{k, i} \in\left\{\mathbf{s}_{k} \mid 0=g_{k}\left(\mathbf{s}_{k}\right)\right\}
$$

$$
k=10^{1}
$$

$$
\operatorname{dim}(\mathbf{x})=3
$$

$\eta>10^{7}$

- Principle of extremal action
- Law of large numbers
- Homogeneity Assumptions
- Discretization of Space
- Continuity Assumption
- Floating Point Approximation
- Discretization of Time
- ...
- Simultaneity
- Instantaneity
- Conservation laws
$\eta>10^{40}$


## About Simultaneity.

- Simultaneity is a non-physical concept.

It wrongfully assumes the existence of a global time

- However, our computing devices have their own clock Simultaneity is thereby a very powerful concession; pleasing the way we compute.
- In our example we could gain a factor of $10^{7}$ by replacing propagation of energy through a wave by the direct instantaneous transfer of energy between potentials.
- It is a powerful but a little dangerous concept



## Making different Use of Simultaneity

But here is another way to achieve the result:



And the efficiency is almost as good as working with the potentials

## Modeling is a lossy compression of reality

| ODE Approach | LIED Approach | DAE Approach |
| :---: | :---: | :---: |
| $\mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}\right)$ | $\begin{gathered} \mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{\substack{k=1 \\ \mathbf{A}_{\mathrm{k}} \mathbf{s}_{k}=\mathbf{b}_{\mathrm{k}}}}^{k=t / h} h f\left(\mathbf{x}_{k-1}, \mathbf{s}_{k}\right) \\ \hline \end{gathered}$ | $\begin{gathered} \mathbf{x}_{t}=\mathbf{x}_{0}+\sum_{k=1}^{k=t / h} h f\left(\mathbf{x}_{k-1}, \mathbf{s}_{k, i}\right) \\ \mathbf{s}_{k, i} \in\left\{\mathbf{s}_{k} \mid 0=g_{k}\left(\mathbf{s}_{k}\right)\right\} \end{gathered}$ |
| $\begin{aligned} & k=5 \cdot 10^{7} \\ & \operatorname{dim}(\mathbf{x})=9 \end{aligned}$ | $\begin{aligned} & k=1 \cdot 10^{1} \\ & \operatorname{dim}(\mathbf{x})=5 \\ & \operatorname{dim}(\mathbf{s})=2 \end{aligned}$ | $\begin{gathered} k=10^{1} \\ \operatorname{dim}(\mathbf{x})=3 \\ \operatorname{dim}(\mathbf{s})=1 \end{gathered}$ |
| - Simultaneity on -Simultaneity on <br> kinetic energy <br> potential energy <br> - Instantaneity Conservation laws <br> - ...  |  |  |

## Making different Use of Simultaneity

## One idea to get the best of both worlds:

- Let us apply the concept of simultaneity but
- avoid the direct exchange of potential energy
- instead just apply it to kinetic energy.
- To implement this, we need to re-establish the special
 role of kinetic energy in our connector:
- This needs another dimension and we have to split up the potential variable:
- For hydraulics: we decompose the pressure:

$$
p=\hat{p}+r
$$

with $r$ being the inertial pressure resulting from the kinetic energy of the flow.


## Implementation in DLR ThermoFluid Stream / HEXHEX

- For a general solution, we need a special interface
- The pair $(r, \dot{m})$ represents the kinetic part and builds up a linear equation systems
- The signal © represents the thermodynamic state based on $\hat{p}=p-r$
- With a partial base model we ensure that the kinetic energy is always present.
- We can then choose to either directly couple the kinetic parts or whether they shall interact with the potentials (boundaries, volumes). We thereby limit the spatial frequency of interaction.
end Inlet;


## Implementation in DLR ThermoFluid Stream / HEXHEX

Robust component models
$\rightarrow$
Robust system model

- All components take kinetic energy into account.
- The connector enables us to use one state for each branch of mass-flow rate
- The connector enables us to limit the choose a different spatial resolution for $r$ than for $\hat{p}$.
small non-linear blocks for complex components


This method scales...


This method scales...


## This method scales...



- The ENERGIZE model describes a more electric aircraft with 220 passengers
- Combination of thermal and electrical power management.
- Complete aircraft missions through different environmental conditions can be simulated.
- > 18,000 Equations
- > 300 States


## Implementation in Dialectic Mechanics

- In Mechanics we can go for a similar solution:
- The pair $v_{\text {kin }}, f$ cares about the kinetic energy
- The position signal $r$ describes the current configuration
- Again we can ensure the presence of kinetic

```
connector Flange
    [...]
    SI.Velocity v_kin;
    flow SI.Force f;
    input SI.Position r;
``` energy in all dimensions
- We can limit the temporal interaction frequency between elastic potentials and kinetic energy at the joint elements by stating
- \(\frac{d v_{k i n}}{d t} T_{D}=\frac{d r}{d t}-v_{k i n}\)
end Inlet;


\section*{Implementation in Dialectic Mechanics}

- With only linear equations and bounded eigenvalues, Dialectic Mechanics is very attractive for real-time applications
- Gripping
- Variable Structure Systems
- Even stiff system can be computed in hard real-time.

\section*{Linear Implicit Equilibrium Dynamics}
- The way of modeling that we derived leads to a special class of DAE systems: Linear Implicit Equilibrium Dynamics.
\[
0=F\left(\mathbf{x}_{P}, \dot{\mathbf{x}}_{P}, \mathbf{u}, t\right)
\]
\[
\begin{aligned}
& \mathbf{L} \dot{\mathbf{x}}_{L}=g\left(\mathbf{x}_{L}, \mathbf{x}_{N}, \mathbf{u}, t\right) \\
& \dot{\mathbf{x}}_{N}=f\left(\mathbf{x}_{L}, \mathbf{x}_{N}, \mathbf{u}, t\right)
\end{aligned}
\]
...where \(g\) and \(f\) can be constructed just by sorting the equations of \(F\). Given \(\mathbf{x}_{L} \cup \mathbf{x}_{N}=\mathbf{x}_{P}\) with \(\mathbf{x}_{L} \cap \mathbf{x}_{N}=\{ \}\)
- Hence the implicit part may only expressed using the matrix \(\mathbf{L}\)
- What looks like a very restrictive class of models is actually more powerful than what we might think.

\section*{Linear Implicit Equilibrium Dynamics and Pre-Compilation}
- It turns out that pre-compilation of Components is feasible for LIED systems.

\section*{- A component then yields several blocks of code:}
```

model PressureDrop
TFSPlug inlet;
TFSPlug outlet;
parameter VolumeFlowRate v_ref;
parameter Pressure dp_ref;
VolumeFlowRate v_norm;
SI.Pressure dp;
SI.MassFlowRate m;
equation
v=inlet.m.flow/rho(inlet.state);
v_norm = v/v_ref;
dp*2 = dp_ref**(v_norm+v_norm^2);
inlet.m + outlet.m = 0;
v = inlet.v;
inlet.p - dp = outlet.p;
end PipeFrictionNL;

```
```

void PressureDrop::evalState() {
const double v = inlet.m.flow/rho(inlet.state);
const double v_norm = v/v_ref;
const double dp = 0.5*dp_ref*(v_norm + v_norm*v_norm);
outlet.state.h = inlet.state.h;
outlet.state.p = inlet.state.p - dp;
}

```
void PressureDrop: :evalFlow() \{outlet.m = -inlet.m; \}
```

void PressureDrop::evalInertial() {
inlet.inertial.r = outlet.inertial.r
+ L*inlet.m.flow_der;
}

```

\section*{Variable Structure Systems}
- For LIED systems, the code blocks would then be sorted. This can be done at run-time.
- In our example:
-evalState() and evalFlow () are both sorted downstream
- \(\square\) evalInertia() is sorted upstream.
- Then Variable Structure Systems suddenly become quite trivial


CONCLUSIONS
- Our current standard interfaces, ultimately result from Hamilton's trick to double the dimension. They are what is necessary for object-oriented modeling.
\begin{tabular}{|r|c|c|c|c|c|c|}
\hline Domain & \begin{tabular}{l} 
Translational \\
Mechanics
\end{tabular} & \begin{tabular}{l} 
Rotational \\
Mechanics
\end{tabular} & Hydraulics & Electrics & Thermal & \(\cdots\) \\
\hline Potential & \(r\) & \(\varphi\) & \(P\) & \(V\) & \(T\) & \\
\hline Flow & \(f\) & \(\tau\) & \(\dot{Q}\) & \(i\) & \(Q\) & \\
\hline
\end{tabular}
- We can find extended interfaces that offer a sufficient form.
(Unfortunately hardly anyone is looking for these forms)
\begin{tabular}{|r|c|c|c|c|c|c|}
\hline Domain & \begin{tabular}{c} 
Translational \\
Mechanics
\end{tabular} & \begin{tabular}{c} 
Rotational \\
Mechanics
\end{tabular} & \begin{tabular}{c} 
Thermo \\
Fluids
\end{tabular} & Electrics & \(?\) & \(\ldots\) \\
\hline Potential & \(v_{\text {kin }}\) & \(\omega_{\text {kin }}\) & r & \(?\) & \(\ldots\) & \\
\hline Flow & \(f\) & \(\tau\) & \(\dot{m}\) & \(?\) & \(\ldots\) & \\
\hline Signal & \(r\) & \(\varphi\) & \(\boldsymbol{\Theta}\) & \(?\) & \(\ldots\) & \\
\hline
\end{tabular}

\section*{The Value of a Sufficient Statement}
- Sufficient forms have rules for their components and their composition.
\(\checkmark\) 深 ThermofluidStream
i User's Guide
(i) Basic composition rules
i Nomenclature
¿ \(\Rightarrow\) Release notes

\section*{Important Claim/Observation:}

The complexity of these rules transfers to the complexity of code generation
- This is the key to pre-compilation, variable structure system, large systems etc.
- LIED Systems are one primary example.

\section*{A Dire Word of Warning}
- Insisting on solutions that work on arbitrary general hybrid non-linear DAE Systems is a sure way to stall progress...


Source: Homer the Great, The Simpsons, TV Series S6 Ep12
...it did so the last 20 years and will do so in the next 20 years.
- We should simple be pragmatic and support forms that have a rule set.

Libraries

- Our libraries of free and commercial models is our biggest asset.
- We form a wonderful community hosting and nurturing these models.
- We should support models of all kind. Those who profit from rules and those who want freedom.
- Sufficient forms may spur progress in many areas!
- The upcoming AI Revolution will swipe over everything that is high-level. Having a good understanding of the low-level is what will make us thrive and survive.

\section*{References to current work on sufficient forms}

\section*{On Thermofluid Streams:}
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\section*{On Dialectic Mechanics:}
- Zimmer, D. C. Oldemeyer (2023). "Introducing Dialectic Mechanics". Proceedings of the 15 \({ }^{\text {th }}\) International Modelica Conference, Aachen.
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\section*{On Linear Implicit Equilibrium Dynamics (LIED):}
- Zimmer, D. C. Oldemeyer (2023). "Object-Oriented Formulation and Simulation of Models using Linear Implicit Equilibrium Dynamics". Proceedings of the \(15^{\text {th }}\) International Modelica Conference, Aachen.

\section*{Imprint}

Topic: Dealing with complex models and how to use the idealization of physics to our advantage

Date: 2023-01-01 (YYYY-MM-DD)
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[^0]:    * for the conservative case

[^1]:    * under certain conditions

