# Analysis of Differential Algebraic Equation Systems for Connecting Energy Storages of Generally Valid Functional Mock-up Units

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Abstract: Functional Mock-up Units (FMU) refer to tool-independent models exported from their original simulation tools. They enable component manufacturers and system integrators to exchange models across entire production chains to validate solutions virtually. However, since system equations cannot be accessed or modified in an FMU, numerical challenges can arise, especially when coupling similar energy storages. In this paper, therefore, Differential Algebraic Systems of Equations are analyzed for their suitability for FMU couplings. It is shown how such systems of equations can be described in a general way and how suitable coupling constraints for FMUs are chosen. Subsequently, three solution approaches are presented and analyzed for their feasibility with FMUs.

# **1** INTRODUCTION

Due to the increasing complexity in mechatronic systems, a continuous simulation in the development is indispensable (Michael et al., 2016). In this way, partial solutions can already be virtually validated in domain-specific development. This reduces the construction of necessary prototypes and thus leads to increased cost and time efficiency.

However, a particular challenge lies in the large number of interacting domains. Specialized tools are often used for different domains. System integrators must therefore couple models from a heterogeneous tool landscape with each other in order to represent the overall system. The Functional Mock-up Interface (FMI) has proven to be a widely used way of coupling models in a tool-independent manner. Models are exported from their original modeling tools as compiled binary files. These are called Functional Mock-up Units (FMU). The models can then be interconnected via a standardized interface.

In addition to the tool-independent coupling of the models, the FMI standard also allows industrial know-how protection to be achieved. Since the model behavior is represented by binary files, the internal system equations can no longer be accessed or changed. Thus the FMI standard can be used for a modular model exchange over entire production chains. Manufacturers of individual components, e.g. from the electrical drive technology, have the opportunity to pass on models to customers without disclosing their know-how. This increases market visibility and enlarges the customer base. On the other hand, system integrators can test components from different manufacturers virtually in their overall solution.

The model boundaries of an FMU can be defined as small as desired. For example, an FMU can represent a physical component or an entire assembly. However, a component can also be divided further, so that an FMU can also be created at subcomponent level. For example, an industrial converter can be divided into a rectifier and an inverter. Similarly, individual FMUs can be created from software components, such as control algorithms. A large number of FMUs in the overall system ensures greater modularization. Individual submodels can be exchanged and reused more easily. For example, the user can assemble his own system model from a set of prefabricated FMUs.

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However, the tighter the model boundaries are chosen, the higher is the integration effort in the overall system and numerical difficulties can occur. This is especially the case when similar energy storages are interconnected across model boundaries. Energy storages are physical elements, such as masses, springs, capacitances or inductances, whose state is described by its stored energy. For example, if two masses are rigidly coupled together, they have the same state. However, if the two masses are arranged in two different models, each model calculates its own state for the masses, which are independent of each other.

In this article, therefore, methods are analyzed with which FMUs can be coupled to overall systems in a generally valid way. Special attention is paid to the representation of a total system as a differential algebraic equation (DAE) system. The couplings should be possible independently of the energy storage distribution in the system, so that two similar energy storages, which are arranged in different FMUs, can be rigidly coupled. In this paper, masses are considered as an example.

# 2 STATE OF THE ART

In the following, the required information about the Functional Mock-up Interface is given. Subsequently, methods for coupling two masses from conventional modeling are presented. An evaluation is made whether these methods can be implemented with the FMI standard.

#### 2.1 Functional Mock-up Interface

The Functional Mock-up Interface refers to a standard agreed upon by various vendors of modeling and simulation tools to export their models as binary files. With this standard a Co-Simulation or a model integration can be performed. The exported models are called Functional Mock-up Units. The standard was first published in 2011 and currently exists in versions 1.0 and 2.0. In addition, a pre-release of version 3.0 exists since 2021 (FMI Development Group, 2014).

An FMU consists of two files, a DLL file and an XML file. The DLL file is the binary file that represents the model behavior. For this purpose, it implements the system equations of a general nonlinear system, as follows.

$$\dot{x} = f(\underline{x}, \underline{u}, t) \tag{1}$$

$$y = w(\underline{x}, \underline{u}, t) \tag{2}$$

With

- x: System State
- *u*: System Input
- y: System Output
- t: System Time

The DLL file offers functions to read and write the variables. In addition, a single simulation step can be executed. The equations f and w, however, cannot be accessed.

The XML file represents the model description, which contains all the required model information. It lists which variables the model contains. Value references are specified for these variables, which can be used for the model functions from the DLL file to reference a variable. Other attributes that a variable can have are *variability* and *causality*. *Causality* specifies whether a variable is an input, output or parameter. *Variability* specifies whether a variable may be changed during the simulation. Possible values here are *fixed* and *tunable*. For example, output variables cannot be written by the user, but are only calculated by the model (FMI Development Group, 2014).

In (Blochwitz et al., 2012) an approach was presented to couple masses in different FMUs. For this purpose, a function for calculating directional derivatives is used, which was introduced in the FMI 2.0 version. This can be used to compute a Jacobian matrix that can be used to solve algebraic loops created by the coupling. However, for this approach, the appropriate input and output variables for the FMUs must be provided. In this approach, one mass provides the position, velocity and acceleration as output. The second mass uses these quantities as input and calculates a counter torque as output. An example is provided by the Standard Modelica Library in the GenerationOfFMUs example. However, these masses have different interfaces and are not considered as generally valid in this article. Furthermore, the determination of directional derivatives refers only to the FMU inputs. Moreover, this function is optional and is only implemented by a few tools.

# 2.2 Coupling of Similar Energy Storages

In this section, first the problems of coupling similar energy storages are shown. Then, common approaches for solving the problem are presented. Table 1 summarizes the energy storages from the mechanical and electrical domains and shows the corresponding differential equations (Isermann, 2007). The input and output quantities given result from the assumption that the equations are solved by numerical integrations only.

Table 1: Equations of Energy Storages.

Energy Storages	Equations	Input	Output
Inductor	$\dot{\iota} = \frac{1}{L} \cdot u$	u	i
Capacitor	$\dot{u} = \frac{1}{C} \cdot i$	i	u
Mass	$\dot{v} = \frac{1}{m} \cdot F$	F	v
Spring	$\dot{F} = c \cdot \dot{x} \\ = c \cdot v$	ν	F

The following system variables appear in the equations from Table 1:

- u: voltage
- *i*: current
- *v*: velocity
- F: force

The parameters of the energy storages are given as:

- L: inductance
- C: capacity
  - m: mass
  - c: spring stiffness

In this paper masses are considered for coupling similar energy storages. In order to couple two masses, they are first considered individually. Figure 1 shows a free cut of two masses  $m_1$  and  $m_2$ .



Figure 1: Free cut of two masses.

Each mass is driven by a force F, resulting in a velocity v. The two masses are described by the differential equations from table 1:

$$\dot{v}_1 = \frac{1}{m_1} F_1$$
  $\dot{v}_2 = \frac{1}{m_2} F_2$  (3)

For each equation the input and output variables from table 1 are given. Thus, each mass has the force F as input and the velocity v as output. In this case the problem with the coupling of two masses is obvious, if they are located in different models. Both masses expect a force as input. However, the first mass provides a velocity as output, which cannot drive the second mass. Thus, there is an interface inconsistency. In (Ehlert et al., 2021) approaches were presented with which energy storages can be coupled, when all equations are known and adjustable. These approaches are presented in the following.

#### 2.2.1 Substitute Variables

One of the most common methods for coupling similar energy storages in conventional modeling is the creation of substitute quantities. In this case, the two masses are calculated to a total mass. This results in a differential equation for both masses.

$$\dot{\nu} = \frac{1}{m_{total}}F\tag{4}$$

With this method, exact simulation results with a fast computation time can be expected. However, the system equations must be changed for this, which is not possible with FMUs.

#### 2.2.2 Fictitious Coupling Elements

Another possibility is the dynamic coupling via fictitious coupling elements. In the case of two masses, a fictitious spring is placed between the masses. From this, the following differential equations are derived:

$$\dot{v}_1 = \frac{1}{m_1} \cdot (F_1 - F_c) \tag{5}$$

$$\dot{F}_c = c \cdot (v_1 - v_2) \tag{6}$$

$$\dot{\nu}_2 = \frac{1}{m_2} \cdot \left(F_2 - F_c\right) \tag{7}$$

Each mass is only dependent on its own input force and the fictitious spring force  $F_c$ . Thus, the states of the masses are decoupled from each other. The stiffer the spring is chosen, the more a rigid coupling is approximated.

However, this approach can lead to long computation times if the spring is chosen to be very stiff, since this results in small time constants in the system. On the other hand, if the stiffness is too small, the simulation results are distorted by fictitious dynamics.

One possibility to solve the coupled system with less computational effort is to perform an order reduction before the simulation. Equations (5) to (7) represent a third order differential equation system. Since a rigid coupling is approximated here, the difference of the independent states  $v_1$  and  $v_2$  will be very small. Here, an order reduction can compute a minimum order in which no independent states for the velocities would be considered. In the field of snapshot-based methods, there exist procedures for model order reduction that can also be implemented as a black box (Benner et al., 2021). It will be separately analyzed if these methods can be applied to FMUs.

#### 2.2.3 Definition of Coupling Constraints

A third way of coupling similar energy storages is to define constraints. Thereby, the ordinary differential equations of the masses are arranged independently in a state space model. This is completed by coupling constraints to a Differential Algebraic System (DAE). In (Najafi, 2018), an extension of the Functional Mock-up Interface is presented to solve DAE systems in FMUs. In this approach, however, the algebraic constraints are part of the FMUs. In this paper, the constraints are formed by the coupling. Thus, they are located outside the FMUs and have to be solved by a higher-level simulation algorithm considering the available model information.

Since the ordinary differential equations from the FMUs are considered independently, this approach will be further analyzed.

# 3 MODEL COUPLING USING DAE-SYSTEMS

For the model coupling using DAE systems, two masses are considered again. Each mass is arranged in its own FMU, as shown in Figure 2. Both FMUs have at least one force input and one velocity output. The coupling is realized via a constraint force  $F_z$ , which can be applied via the force inputs in the FMUs. In addition, the first mass is actuated via a driving force  $F_a$ .



Figure 2: Coupled masses using coupling constraints.

For the example shown, a DAE system can now be set up. DAE systems consist of a set of ordinary differential equations (ODE) and a set of algebraic constraints. The ordinary differential equations for the system shown above are:

$$\dot{v}_1 = \frac{1}{m_1}(F_a - F_z)$$
  $\dot{v}_2 = \frac{1}{m_z}F_z$  (8)

Whereas an algebraic equation could limit the velocities:

$$v_1 - v_2 = 0$$
 (9)

In general, DAE systems are defined as follows (Janschek, 2010):

$$\dot{\underline{x}} = f(\underline{x}, \underline{z}, \underline{u}, t) \tag{10}$$

$$0 = g(\underline{x}, \underline{z}, t) \tag{11}$$

Here, f represents the set of ordinary differential equations and g the set of algebraic constraints. The additional vector  $\underline{z}$  describes all algebraic variables that do not occur differentially. It should be noted that the DAE system represents the overall system and not the individual FMUs. Therefore, the DAE system has only the driving force  $F_a$  as input and not the other FMU inputs used for coupling. For the example above, this results in the following assignments:

• 
$$x = [v_1, v_2]^T$$
 (12)

• 
$$z = F_z$$
 (13)

$$\underline{u} = F_a \tag{14}$$

In order to be able to couple models using DAE systems, possible coupling constraints are analyzed first. Then, the index of a DAE system has to be determined in order to be able to select a solution approach based on it.

# 3.1 Choice of Coupling Constraints

Coupling constraints can be chosen differently for different systems. An example is provided by the Modelica modeling language (Elmquist et al., 1998). Here, the user creates a topology of a system. The modeling tool that implements the Modelica language uses this topology to create coupling constraints for the basic differential equations of the modeling modules and thus generates a system of equations. The coupling constraints are chosen in a way that the potential quantities of two elements, that have been connected with each other, are equal. The sum of all flux quantities in this connection must result in zero.

In (Woernle, 2016) coupling conditions for mechanical systems are classified. The constraints can be divided into holonomic and non-holonomic constraints. Holonomic constraints describe constraints on the position level. Non-holonomic constraints limit the velocities of the individual models. A further classification takes place in scleronome and rheonome constraints, whereby scleronome constraints are time independent and rheonome systems have a time dependence. In this paper, scleronomic bindings are chosen for coupling different masses, since it is a rigid coupling and no time dependence is necessary. Moreover, the bindings are defined at velocity level to keep the index of the DAE system small. Furthermore, possible solution approaches for solving the DAE system could be applied to other energy storages as well. This assumes that the initial states of the masses are equal. The resulting DAE system is classified as nonholonomic scleronomic. The algebraic constraint looks as follows:

$$g(\underline{x}) = \underline{x}_1 - \underline{x}_2 = v_1 - v_2 = 0$$
(15)

A detailed analysis of approaches for solving holonomic DAE systems with FMUs can be given as part of the future work.

# **3.2** Determination of the Index

The index of a DAE system is an indicator for the degree of difficulty to solve the system. It describes how often the algebraic constraints have to be differentiated in time to transform the DAE system into an ODE system. This procedure is also called index reduction. If an ODE system could be formed, it can be solved with ordinary numerical integration methods.

To determine the index, the algebraic constraint is first differentiated and checked whether it subsequently represents another ODE. For this purpose, the algebraic variables  $\underline{z}$  must occur in a differentiated form.

$$\frac{d}{dt}g(\underline{x}) = \frac{\partial g}{\partial \underline{x}}\underline{\dot{x}} = 0 \tag{16}$$

$$\rightarrow \frac{d}{dt}g(\underline{x}) = \frac{\partial g}{\partial \underline{x}}f(\underline{x}, \underline{z}, \underline{u}) = 0$$
(17)

Since  $\underline{\dot{z}}$  does not appear in equation (17), there is no further ODE for the constraint. Therefore, it is not an index 1 system and the constraint must be differentiated further.

$$\frac{d}{dt} \left[ \frac{\partial g}{\partial \underline{x}} f(\underline{x}, \underline{z}, \underline{u}) \right] = 0 \tag{18}$$

$$\rightarrow \frac{\partial g}{\partial x} \frac{\partial f}{\partial x} \frac{\dot{x}}{\dot{x}} + \frac{\partial g}{\partial x} \frac{\partial f}{\partial u} \frac{\dot{u}}{\dot{u}} + \frac{\partial g}{\partial x} \frac{\partial f}{\partial z} \frac{\dot{z}}{\dot{z}} = 0$$
(19)

To form another ODE, it must be possible to resolve to  $\underline{\dot{z}}$ . This leads to the following index 2 condition:

$$det\left(\frac{\partial g}{\partial x}\frac{\partial f}{\partial z}\right) \neq 0 \tag{20}$$

$$det \begin{bmatrix} \frac{\partial g_1}{\partial \underline{x}_1} & \cdots & \frac{\partial g_1}{\partial \underline{x}_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial g_m}{\partial \underline{x}_1} & \cdots & \frac{\partial g_m}{\partial \underline{x}_n} \end{bmatrix} \begin{bmatrix} \frac{\partial f_1}{\partial \underline{z}_1} & \cdots & \frac{\partial f_1}{\partial \underline{z}_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial \underline{z}_1} & \cdots & \frac{\partial f_m}{\partial \underline{z}_n} \end{bmatrix} \neq 0 \quad (21)$$

With the ODEs from equation (8), the assignment from (12) - (14) and the constraint from (15) this results in

$$det \left( \begin{bmatrix} \frac{\partial g}{\partial v_1} & \frac{\partial g}{\partial v_2} \end{bmatrix} \begin{bmatrix} \frac{\partial f_1}{\partial F_z} \\ \frac{\partial f_2}{\partial F_z} \end{bmatrix} \right) \neq 0$$
(22)  
$$\rightarrow det \left( \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} -\frac{1}{m_1} \\ \frac{1}{m_2} \end{bmatrix} \right) \neq 0$$
(23)

For  $m_1 \neq m_2$ .

Thus, DAE systems for coupling different masses are index 2 systems. In the following, it is explained how these can be solved (Janschek, 2010).

### 3.3 Solving the DAE-system

The following section presents procedures that can be used to solve the DAE systems described above. However, in addition to the general conception, it must also be evaluated whether these procedures can be implemented with the information that are provided by the FMUs.

DAE systems of index 2 can be solved directly with implicit integration methods. Otherwise, index reduction is necessary to reduce it further. With that explicit integration methods can be used for simulating an index 1 system or a full index reduction to an ODE system is performed (Janschek, 2010). All three possibilities are evaluated in the following.

#### **3.3.1 Index Reduction**

An index reduction has already been performed in the determination of the index. Thereby, it can be seen in equation (16) that a single differentiation of the constraints is independent of the system equations in the FMUs. However, since the coupling of the masses are constraints of index 2, a second differentiation is necessary. From equation (19) it can be seen that partial derivatives of the FMU system equations are necessary in this process. However, since these equations are not known, the second differentiation cannot be performed analytically with FMUs.

Therefor a numerical approach to calculate partial derivatives is discussed in a later section.

At this point it should be mentioned that in the FMI 2.0 standard there is the possibility to calculate directional derivatives from FMUs. However, this function is optional in the standard specification and is only supported by a few tools. Furthermore, these directional derivations refer to the input variables of an FMU and not to general algebraic variables.

#### 3.3.2 Explicit Integration Methods

Explicit integration methods perform a calculation of a new state value based solely on past state values. For DAE systems, this means that the ordinary differential equations and the algebraic constraints can be solved sequentially. This can be illustrated by a simple Euler method applied to an index 1 system.

$$\underline{x}_{k+1} = \underline{x}_k + h \cdot f(\underline{x}_k, \underline{z}_k) \to \underline{x}_{k+1}$$
(24)

$$0 = g(\underline{x}_{k+1}, \underline{z}_{k+1}) \to \underline{z}_{k+1}$$
(25)

Here k describes the previous number of integration steps and h the step size. For already known  $\underline{x}_k$  and  $\underline{z}_k$ ,  $\underline{x}_{k+1}$  can be determined from the ordinary differential equations. This  $\underline{x}_{k+1}$  can then be used to calculate  $\underline{z}_{k+1}$  from the constraints. For the initial values,  $\underline{x}_0$  is arbitrary. This implies that  $\underline{z}_0$  can be derived directly from the constraint (Janschek, 2010).

However, explicit integration methods can only be used for systems up to index 1. As mentioned in section 3.3.1, the constraint can be differentiated once, which obtains an index 1 system for an FMU coupling. This results in a new constraint, which is shown in equation (17). This constraint now depends on the system equations f. If it is set equal to zero,  $\underline{z}_{k+1}$  can be determined. However, this is not directly possible with the FMU system equations because the output variables are not declared as *tunable*. Thus, the FMU output cannot be written and the internal parameters do not change.

One possibility to use this approach is to integrate the constraint equation in a control loop, as shown in Figure 3. In this case, only the input variables are written in the FMU, whereby  $\underline{z}$  is adjusted by a Pcontroller in such a way that the output is controlled to zero. However, several iterations per time step may be necessary until the output has reached its stationary final value. To reduce the number of iterations and the stationary error of the equation output, the gain factor K should be chosen very high.



Figure 3: Control loop for constraint equation.

#### 3.3.3 Implicit Integration Methods

Implicit integration methods are used when a system of equations cannot be solved sequentially. This is the case for DAE systems of index 2, as will be shown below using Euler's method.

$$\underline{x}_{k+1} = \underline{x}_k + h \cdot f(\underline{x}_k, \underline{z}_k)$$
(26)

$$0 = g(\underline{x}_{k+1}) \tag{27}$$

Here, the constraint is independent of  $\underline{z}_{k+1}$ . Thus, the algebraic variables cannot be determined for solving the ordinary differential equations in the next integration step. An implicit method must be chosen here, in which equations (26) and (27) are formulated into a root finding problem. The unknown quantities  $\underline{x}_{k+1}$  and  $\underline{z}_{k+1}$  can then be determined by a zero search. An implicit Euler method looks as follows (Janschek, 2010):

$$\underline{x}_{k+1} = \underline{x}_k + h \cdot f(\underline{x}_{k+1}, \underline{z}_{k+1})$$
(28)  
$$0 = g(\underline{x}_{k+1})$$
(29)

When formulating the root finding problem, a substitution can be made for simplicity, introducing the following new variables:

$$\Phi_{k+1}(\underline{P}_{k+1}) \coloneqq \begin{bmatrix} \varphi_{1,k+1} \\ \varphi_{2,k+1} \end{bmatrix}$$
(30)

With

$$\underline{P}_{k+1} \coloneqq \left(\underline{x}_{k+1}, \underline{z}_{k+1}\right)^T \tag{31}$$

and

$$\varphi_{1,k+1} = \underline{x}_{k+1} - \underline{x}_k - f(\underline{x}_{k+1}, \underline{z}_{k+1})$$
(32)

$$\varphi_{2,k+1} = g(\underline{x}_{k+1}) \tag{33}$$

The root finding problem can thus be described in a simple way.

$$\Phi_{k+1}(\underline{P}_{k+1}) = 0 \tag{34}$$

For the solution of this equation different iterative methods for the root finding are possible. In the following, the Newton-Raphson iteration from (Schwarz et al., 2009) is considered. For this method the following recursion rule results:

 $\underline{P}_{k+1,i+1} = \underline{P}_{k+1,i} - J(\underline{P}_{k+1,i})^{-1} \Phi_{k+1}(\underline{P}_{k+1,i}) \quad (35)$ The variable *i* denotes the number of previous iterations of the Newton-Raphson method. Several iterations are required per time step until  $\underline{P}_{k+1,i+1}$  is close enough to  $\underline{P}_{k+1,i}$ . The Jacobian matrix *J* is defined as follows:

$$J(\underline{P}) = \begin{bmatrix} \frac{\partial \varphi_1}{\partial \underline{x}_{i+1}} & \frac{\partial \varphi_1}{\partial \underline{z}_{i+1}} \\ \frac{\partial \varphi_2}{\partial \underline{x}_{i+1}} & \frac{\partial \varphi_2}{\partial \underline{x}_{i+1}} \end{bmatrix}$$
(36)

Now it has to be evaluated whether this approach is feasible with FMUs. Equation (36) shows that the partial derivatives of the equations  $\varphi_1$  and  $\varphi_2$  are needed to form the Jacobian matrix. However, these equations depend on the FMU system equations f. Since f cannot be accessed from the FMUs, these partial derivatives cannot be formed analytically. Therefore, in the following section, a possibility of numerical calculation of partial derivatives is discussed.

#### 3.4 Numerical Partial Differentiation

Both the index reduction and the implicit integration methods depend on partial derivatives of the system equations. Since these are not known, an analytical solution is not possible. Here, a numerical solution using difference quotients can be considered. A numerical partial differentiation of the system equation f to  $\underline{x}$  looks as follows (Schwarz et al., 2009):

$$\frac{\partial f}{\partial x} \approx \frac{f(\underline{x} + \Delta \underline{x}, \underline{z}, \underline{u}) - f(\underline{x}, \underline{z}, \underline{u})}{\Delta x}$$
(37)

A differentiation according to z results in

$$\frac{\partial f}{\partial \underline{z}} \approx \frac{f(\underline{x}, \underline{z} + \Delta \underline{z}, \underline{u}) - f(\underline{x}, \underline{z}, \underline{u})}{\Delta \underline{z}}$$
(38)

With this approach, the system equations only have to be evaluated and not changed. Thus, the difference quotients offer an opportunity to couple and simulate FMUs using DAE systems.

# 4 CONCLUSIONS

In this paper, methods for coupling masses from different models were investigated. Thereby, the

couplings using DAE systems were dealt with in more detail. It was shown how FMUs can be arranged in such systems and how coupling constraints have to be chosen. Subsequently, various possible solutions were presented. A full index reduction or the direct simulation via an implicit or explicit integration method are suitable for solving the DAE system. For index reduction and implicit solvers a numerical partial differentiation is necessary. Explicit integration methods could use a control algorithm to solve the constraint equations of the DAE system. A concrete implementation of a FMU coupling remains to be validated afterwards. For this purpose one of the presented solutions has to be chosen.

# **5 FUTURE WORK**

As an outlook, the concrete implementation of an FMU coupling via DAE systems can be given. For this, first one of the presented solution methods must be chosen. Thereby an index reduction up to an ODE system or the solving with an implicit or explicit integration method is suitable. The choice can strongly depend on the application. For example, an iterative method would not be suitable for real-time applications.

Furthermore, the couplings of other energy storages still need to be analyzed. These could lead to changed coupling constraints. At the position level further coupling constraints could be analyzed for masses as well. Once coupling constraints are defined for all relevant energy storages, the interaction with the user of a simulation has to be defined. FMUs are signal flow oriented models. However, the couplings described here do not take place on the basis of signals. An input must be found with which the user determines the FMUs to be interconnected. From this input, the coupling constraints have to be derived in an automated way in order to build a DAE system. Here, it is particularly important to analyze the interfaces of the FMUs.

Besides the FMU coupling via DAE systems, the dynamic coupling with fictious coupling elements can be analyzed further. Here, a model order reduction for black box models can be investigated to reduce the computation time of the simulation.

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