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Co-simulation of an electrical distribution network and its supervision communication network

Benjamin Camus*, Anne Blavette†, Anne-Cécile Orgerie*, Jean-Baptiste Blanc-Rouchosse††

*Univ. Rennes, Inria, CNRS, IRISA, Rennes, France, Email: {benjamin.camus, anne-cecile.orgerie}@irisa.com
†Univ. Rennes, CNRS, SATIE, Rennes, France, Email: anne.blavette@ens-rennes.fr
‡Univ. Toulouse, CNRS, IRIT, Toulouse, France, Email: Jean-Baptiste.Blanc-Rouchosse@irit.fr

Abstract— Smart grids require the large-scale deployment of communication means to interconnect the electrical devices and to autonomously pilot their management. Hence, interconnected tools from both the communication and the power system communities are required in order to adequately simulate the mutual dependencies between these two infrastructures. In this paper, we propose an open-source co-simulation framework for evaluating the mutual impacts between an electrical distribution network and its supervision communication network. A case study dealing with line congestion mitigation is presented to illustrate the versatility of our tool.

Keywords: communication network, smart grids, co-simulation

I. INTRODUCTION

The electricity consumption keeps growing worldwide: the number of consumers increases along with the individual consumption. According to the International Energy Agency, in 2018, the worldwide electricity consumption increased by 4%, the fastest pace since 2010 [18]. The current challenges in the energy sector are multiple: integration of increasingly flexible distributed loads, such as electric vehicles, exploitation of renewable energy sources, often variable by nature and decentralized, and energy storage management. The digitization of electricity grids is expected to meet these challenges by jointly optimizing the production, distribution and consumption.

Smart grids involve the large-scale deployment of communication means to interconnect the electrical devices and to autonomously pilot their management. This digital infrastructure relies on computing, communication and storage resources to provide secure tools for processing, modeling, predicting and optimizing the electrical grid utilization.

Current telecommunication networks experience delays, jitter and bottlenecks. Such a dynamicty, ubiquitous in Internet networks, can greatly impact the management of smart grids as they require a guaranteed quality of service (QoS) [14], [23], [31]. Consequently, management solutions need to carefully take into account mutual impacts between the electrical network and its management telecommunication network.

Since real experiments require quite complex infrastructures and are limited in scale for security and economical reasons, simulation is commonly used in both the communication and energy fields, to assess the feasibility and validity of original solutions [2], [9]. Both communities have their own validated simulation frameworks providing accurate and reliable results. Yet, in the case of smart grid simulation, interconnected tools from both communities are required in order to adequately simulate the mutual dependencies of the management network and the electrical grid.

In this paper, we propose an open-source co-simulation framework for evaluating the mutual impacts between electrical network and its supervision communication network. This framework is available online [24]. It combines two validated and well-known simulators: SimGrid for the communication part and PowerFactory for the electricity part. Besides, our framework is extensible (e.g., any FMI-compliant power system simulator can be used) and scalable. We test it on a representative use-case taken from an electrical grid connecting 55 homes. A comparative analysis is then performed between a centralized and a decentralized management approach, with several values of communication latency, with respect to the impacts in terms of electrical load shedding and to the number of messages sent over the communication network.

The paper is organized as follows. Section II details the background and the related work. Section III presents our contribution. The validation use-case is presented in Section IV. Section V concludes this work and presents future work.

II. BACKGROUND AND STATE OF THE ART

Context. The emergence of smart grids renders more and more urgent the need for adopting a holistic view on the power system and on its supervision communication network. However, this combination renders the power system performances dependent on the design and management of its telecommunication supervising infrastructure. The performances of the former can of course be enhanced by using the additional amount of information provided by the latter [30]. Yet, recent works have shown that communication delays, inherent to the transmission of information, could also degrade dramatically the performances of a smart grid [14], [23], [31]. Hence, it is now recommended to take communication latencies into account when designing control algorithms requiring a significant share of information transmission, such as distributed energy management strategies.

Power systems simulators. Grid studies are usually performed using power system simulators that reproduce numerically the dynamics of an electrical network. Some of them are entirely dedicated to such simulations, such as PowerFactory [7], EMTP-RV, PSS®E and OpenDSS. Other less specific, and therefore more versatile, tools intended for the modeling and


the simulation of various complex systems are also applied to
electric power system simulations, such as Dymola [6].

Computing systems simulators. Several tools are used in
the literature to simulate computing systems. Event-based
packet-level simulators such as NS-3 [15] and OMNeT++ [26]
are widely employed in the communication network com-
munity. These simulators are highly accurate due to their
full implementation of the whole communication protocol
stack. However, these tools do not scale up well because they
simulate each packet transfer on the network. Then, it may take
orders of magnitude longer than the simulated time to simulate
large and complex communication network topology [13].
Hence, packet-level simulators do not seem adapted to conduct
experiments on energy management in smart grids that may
consider timescales ranging from seconds to several years.

The SimGrid platform [4] is dedicated to the simulation of
large-scale distributed systems. It embeds a validated flow-
level TCP/IP model that is faster by orders of magnitude
than packet-level approaches [13]. Moreover, it has been
extensively demonstrated that this model is also reasonably
accurate [29]. Contrary to communication network simula-
tors, SimGrid also embeds theoretically and experimentally
assessed performance models to capture computing resources
sharing [28]. Thanks to these models, SimGrid can simulate
the execution of a complex large-scale distributed application
over a distributed infrastructure. Thus, SimGrid is perfectly
adapted to simulate a smart grid supervising system.

Co-simulation. For decades, the computing and the power
systems communities have developed independently complex
and sound simulation tools. However, with the advent of
smart grids, combining these tools in a co-simulation is
necessary [12]. Co-simulation consists in coupling different
stand-alone simulation tools, so that they simulate together
a whole system in a consistent way. This approach has the
advantage of (1) enabling to study the global behaviour of
the system, while (2) enforcing a clear separation of concerns in
a multidisciplinary modeling and simulation process, and (3)
reusing and factorizing efforts put into the development and
validation of preexisting simulation tools.

However, when applied in a multidisciplinary context like
smart grids, co-simulation raises several issues. First, the inter-
operability has to be managed so that the different simulation
software can be synchronized and exchange usable data. The
challenge is then to bridge the gap between software codes that
are written in different programming languages, implement
different APIs, and are compliant with different Operating
Systems (OS). In the smart grid context, co-simulation also im-
plies to rigorously integrate the different modeling paradigms
of the computing and power systems communities and their
simulation strategies. On one hand, power systems are mod-
eled by continuous differential-algebraic equations systems
and simulated by complex numerical solvers. On the other
hand, distributed computing systems applications are mostly
modelled with a programmatic interface as concurrent pro-
grams and simulated in a discrete event-based way. It is then
required to bridge the gap between continuous formal-based
simulation and discrete code-based simulation to perform an
hybrid co-simulation.

Several works in the literature propose co-simulation plat-
forms to study smart grids [1], [2], [10], [11], [16], [19],
[20], [22], [25], [27]. However, most of them are restricted
to a specific power system simulator, which strongly limits
the experiments that can be done with the co-simulation
tool [2], [10], [11], [16], [19], [20]. Some do not simulate
the computing system execution and neglect then its impact
on the smart grid operation [22], [25]. Others use packet-
level communication model and are then strongly limited by
simulation performances [1], [2], [10], [11], [16], [19], [20],
[27]. They also do not include computing resources (e.g. CPU)
sharing models, and neglect then the time taken by the control
system to compute decisions.

The Functional Mock-Up Interface (FMI) standard [21]
proposes a unified software interface to interact with multi-
physical equation-based systems and their solvers. The stan-
dard is directly or indirectly supported by more than one
hundred tools including several major power systems simu-
lators (e.g. EMTP-RV, PowerFactory, Dymola, TRNSYS).
Thanks to FMI, continuous models and their solvers can be
exported/imported from/to any of these tools as a standardized
shared library called FMUs (Functional Mock-up Units). Thus,
FMI offers a framework to solve the interoperability issues
between multi-physical simulation software. It becomes then
an key solution for co-simulation. Yet, the challenge is to
define a master algorithm to manage the co-evolution of FMUs
with their simulation environment. In particular, performing
hybrid co-simulation where FMU components interact with
discrete-event systems still remains an open issue [5].

III. CONTRIBUTION

To co-simulate smart grids systems, we propose to leverage
on the unique Information and Communication Technologies
(ICT) performance models of SimGrid (i.e. bandwidth and
CPU sharing), and on the federating power of FMI regarding
multi-physical systems. We developed a SimGrid plugin
called SimGrid-FMI to perform FMI-based co-simulations
with SimGrid. The C/C++ code of this plugin is available in
open-source at [24].

To mediate interactions between code-based models of
distributed applications and equation-based models of power
systems, SimGrid-FMI leverages the SimGrid architecture that
is highly inspired from the classical design of an Operat-
ing System. Indeed, SimGrid natively compartmentalizes the
distributed application processes using dedicated execution
contexts. Then, a simulation kernel strictly manages the co-
evolution and interactions of these processes with the ICT
performance models. SimGrid-FMI adds to this architecture a
new SimGrid model that implements an FMI master algorithm.
The SimGrid kernel manages then the co-evolution of this
master with the distributed application processes in a transpar-
ent way like any other ICT performance model. The plug-in
also extends the SimGrid API for the distributed application
processes in order to be able to interact with FMUs (e.g. set/get
input/output). To be fully compliant with the requirements of hybrid discrete-continuous co-simulations, our API also introduces the concept of state-events in SimGrid. They constitute a special class of discrete-events whose occurrences depend on specific conditions on the continuous state (usually when a continuous variable crosses a given threshold). During the simulation, the distributed application processes can then use our API to define state-events and register callbacks to their occurrences.

The master can manage a set of coupled FMUs thanks to an Euler fixed step-size integration algorithm [3]. Algebraic dependencies between FMUs are handled with a brute-force algorithm that iteratively solves the dependencies until the system state reaches a stable point. This algorithm assumes nonetheless that there is no algebraic loop in the FMUs couplings. The master detects state-events occurrences thanks to a fixed-size look-ahead research strategy. We deliberately choose simple numerical algorithms to make as few assumptions as possible on the FMUs capabilities, and therefore to be compliant with the maximum number of multi-physical simulation tools. To transparently use simulation tools that are compliant with different OS, we design a proxy FMU that uses network sockets to mediate interactions between the master and a remote FMU deployed on another environment.

Thanks to the versatility of SimGrid and FMI, our plugin can be used to simulate a larger class of systems where distributed computing infrastructures and multi-physical systems interact. In a previous work [3], we have validated SimGrid-FMI with the co-simulation of a data-center and its air-cooling system. In this paper, we demonstrate that our tool can also be used to co-simulate smart grids.

IV. USE-CASE

A. Scenario

1) Power system modeling: The scenario considered here is based on the publicly available “European Low Voltage Test Feeder” electrical network model [17]. To the best of the authors knowledge, it is the sole IEEE benchmark (available publicly) of a low-voltage European distribution network that also provides power consumption time series for each electrical load it contains. They represent the consumption of 55 electrical loads in the same district, each seemingly representing a single household. The electrical grid between the homes exhibits a tree topology that is connected to the district power substation through an electrical line named Line1 in the rest of the paper.

An arbitrarily-chosen number of 15 electric heaters (direct-acting) were added to this electrical network. The power consumption of each of these heaters was modeled as a cyclic profile alternating between a typical value of 2 kW and 0 kW. The power profile of a single radiator, observed as part of an experiment, was used to model the power consumption profile of all the radiators modeled in this article, to which random time delays were applied to model the aggregation effect.

It must be noted that, in the absence of additional experimental data, a simple heater model was considered. In this model, the post-shedding rebound effect on the power consumption was not included. However, in the case where the shedding duration is relatively short, it may be assumed that the corresponding rebound effect is also small.

2) Case study description - Power system perspective: It was assumed that congestion would occur during the peak hour of the IEEE model. Line congestion occurs when the current flowing through a line exceeds its rated value, and it should be avoided. In the past, this type of issue used to be prevented by a sufficient over-sizing of the electric network. However, grid upgrading is extremely costly and time-consuming. Hence, distribution system operators (DSOs) seek now to maximise their assets usage by deploying smart energy management strategies. Short-term load shedding may be one of them. This strategy consists in suspending temporarily the power supply to given electric loads. In particular, the shedding of heaters over a sufficiently short period of time may have a negligible influence on the consumer’s thermal comfort. However, the successive and repeated shedding (called “cascado-cyclic shedding” [27]) of a sufficiently important number of radiators may solve a line congestion issue. It is important to note that automated shedding, as opposed to consumer-activated shedding, is necessary to harness this flexibility potential. This strategy requires indeed a short reaction time and may potentially need to be repeated a significant number of times.

3) Computing system modeling: To support automated shedding in the electrical network, we consider the following TCP/IP communication network. The households and Line1 are equipped with computing devices that are controlled and monitored. All these devices are connected through Ethernet links on the same local network. For the sake of simplicity, we consider a star topology with homogeneous bandwidth and latencies between the nodes, similarly to the connection of each house of the district to the district DSLAM (digital subscriber line access multiplexer).

4) Case study description - Computing system perspective: We consider the following cascado-cyclic policy to automate the shedding of the household heaters in the electrical network. A shedding sequence is initiated when the current in Line1 reaches a given upper threshold. Several households are then selected for shedding. After a short time, a new iteration of the process starts: a new group of households is selected and the shedding process switches to these households. This process is repeated until the current in Line1 falls below a lower threshold. At this point, the shedding process is suspended. To ensure that all the households receive the same amount of shedding commands, they are selected in a cyclic way.

This policy can be implemented by several algorithms. Here, we propose two of them, that are representative of respectively centralized and decentralized approaches. The decentralized approach avoids that a single actor gathers information on all the consumers at very short spatio-temporal scales. Thus, it may be considered by the general public as less intrusive. At the opposite, with the centralized approach, the DSO manages directly all the system. It conserves then a higher level of confidence regarding its own ability to operate safely and
reliably its local network.

In the centralized version (shown in Figure 1a), a master application is in charge of coordinating the whole cascado-cyclic process. It sends commands through network packets to each household to start and stop the shedding of its heaters. It selects in priority the households that have the highest average power consumption over the last δ seconds (and that have not been already selected in the current shedding cycle). The master relies on the information it receives periodically through network packets from power and current probes that are deployed over the network.

In the decentralized version (shown in Figure 1b), the households are sorted in an arbitrary and predefined order. Due to the decentralization of the information, we cannot directly compare the power consumption of the households anymore. A probe located on Line1 regularly samples the current evolution. When it detects an overcurrent, it sends in a network packet the shedding commands to the first household. When a household receives such a command, it decides if it handles it or not (the decision process is described later). If it does, it sheds its heaters for a given duration, and forwards the command packet to the next household afterwards. However, if it does not comply with the shedding command, the latter is immediately forwarded to the next household. Thus, the shedding commands are propagated in the network and handled by the available households on the way. The last household forwards the command packet to the Line1 probe to signal that a cycle has been completed. This command is then forwarded by the probe to the first household to start a new cycle. To ensure that all the households handle the same number of commands, the probe counts the number of performed cycles and includes this information in the shedding commands. Thus, a household only handles the commands that have a cycle number equal to the number of times it has already been shed. When the probe detects that the current falls below the lower threshold, it sends a command packet to the first household to stop the shedding. This stop-shedding packet is propagated in the network in a similar way than described previously. This process insures that when the packet reaches back the probe, shedding has stopped for all the heaters. When the probe restarts the cascado-cyclic process, the ongoing cycle is automatically resumed to the point at which it stopped.

B. Results

To compare the centralized and decentralized versions of the cascado-cyclic process, we model the pilot network using SimGrid v3.23.2, and the electric network using PowerFactory 2018. We use the FMI++ PowerFactory FMU export utility [8] and our SimGrid-FMI plug-in to co-simulate the entire smart-grid. Our experiments run on a single machine. PowerFactory runs on Windows 10 whereas SimGrid is executed in Windows Subsystem for Linux. We use our FMU proxy to communicate between the two environments. During the co-simulation, SimGrid activates and deactivates household heaters shedding in PowerFactory. It reads the power and current consumption values from PowerFactory to simulate the probes behavior. We also perform a monolithic simulation with PowerFactory (i.e. without SimGrid) to observe the system trajectory when no shedding is performed. We use this trajectory as a baseline to reflect the impact of the control system on the smart grid operation, and to validate the co-simulation.

The approaches are compared according to three metrics. The first one is the cumulative overcurrent duration –i.e. the cumulative duration during which the current in Line1 is above the allowed limit. It is computed by PowerFactory and it measures the control system efficiency from a DSO perspective. The second metric is the cumulative household shedding duration required to solve the congestion problem. It is computed by SimGrid and indicates the efficiency of the control system from the smart grid consumers’ perspective. The last metric is the number of messages sent on the communication network. It is computed by SimGrid and indicates the impact of the control system on the communication network.

We simulate the smart grid execution during the peak hour –i.e. from 5:30pm to 6:30pm. To see the impact of the communication network on the cascado-cyclic process efficiency, we vary the communication network latency between 1 ms and 20 ms. Table I sums up the parameters values set in the experiments.

| TABLE I: Co-simulation parameters values. |
| Network bandwidth | 10Gbps |
| Probe and command messages size | 1 KB |
| Power and current probes sampling frequency | 1 sec |
| Duration of a single shedding (if not interrupted) | 60 sec |
| Number of households jointly selected for shedding | 4 |
| Cascado-cyclic process activation threshold | 160 A |
| Cascado-cyclic process deactivation threshold | 120 A |
| Duration δ to average households consumption | 300 sec |

Figure 2 shows the evolution of the current in Line1 over time with the centralized approach and a communication network latency of 1 ms. From this graph, we see a nominal behavior of the co-simulation. The cascado-cyclic process starts when the current in Line1 reaches the upper threshold. Then, the current decreases due to the shedding of the heaters.
When the current falls below the lower threshold, the cascado-cyclic process stops. Then, the trajectories of the co-simulation and the monolithic PowerFactory simulation coincide perfectly due to the absence of shedding. We get similar validity results with the decentralized version.

Figure 3 shows that with a communication network latency of 1ms, the two approaches greatly reduce the cumulative overcurrent duration of approximately 97%. In this context, the latency is so low that the impact of the communication network becomes negligible and the two cascado-cyclic implementations offer similar results. Yet, the centralized approach achieves this performance with a cumulative shedding duration 9% smaller, because it can determine and select in priority the households that have the highest average power consumption. Yet, this information centralization greatly increases the load of the communication network with a traffic almost 30 times higher than with the decentralized approach. Thus, compared to the decentralized approach, the centralized one has a smaller impact on grid consumers at the cost of a greater impact on the communication network.

Figure 3 also shows that the decentralized implementation is more sensitive to the communication network latency. When the latency increases to 10ms (resp. 20ms), the cumulative over-current duration of the decentralized implementation is 2.2 (resp 2.6) times higher than with the centralized approach. Indeed, the shedding command packet may be forwarded several times from household to household before it is received by schedulable one. At the opposite, the shedding commands are directly sent to the right households with the centralized implementation. Thus, latency has more impact on the responsiveness of the decentralized architecture. Yet, even with a latency of 20 ms, the decentralized (resp. centralized) approach still reduces overcurrent duration by 77% (resp. 91%).

C. Discussion

As mentioned previously, the heaters post-shedding rebound effect is not modeled here in the absence of related experimental data. However, it can be observed that the maximum shedding duration per heater is relatively short (60 s maximum, as shown in Table I) and the cumulative duration does not exceed 7 minutes per heater over the considered hour. Hence, the impact of not modeling the rebound effect in this study is reasonably small. Yet, an advantage of SimGrid-FMI is that we benefit from the versatility of FMI. Thus, to capture the rebound effect, as we intend to do in future work, we can use another dedicated FMI-compliant multi-physical simulator to build a thermal model of the households and to add it directly to our SimGrid-FMI co-simulation framework. Integrating such thermal models in SimGrid-FMI has already been realized in a previous work using OpenModelica [3].

SimGrid-FMI benefits from the simulation performance of the flow-based communication model of SimGrid. The drawback of SimGrid compared to packet-level simulators like NS-3 is that currently, its model is limited to TCP/IP and does not include wireless communications. Yet, if other communication protocols are required, we can use the ad-hoc coupling between SimGrid and NS-3 that has been already developed [13]. Thanks to this coupling, SimGrid can use any NS-3 model instead of its own flow-based model. Thus, our framework is not limited to the flow-based model of SimGrid.

V. Conclusion

The worldwide increase in electricity consumption pushes traditional electrical distribution networks to their limits, while smart grids allow for a more flexible grid management, using a piloting communication network. In this context, we proposed a co-simulation tool named SimGrid-FMI to jointly simulate electrical grids and their management communication network using PowerFactory on the electrical side (or any FMI-compliant power system simulator) and SimGrid on the communication side. This framework is open-source and available online and can be directly used by the smart grid community to obtain sound co-simulations [24].

Thanks to this framework, we studied the influence of the communication network QoS on the management of an electrical grid. This study is conducted through co-simulations of a representative use-case comprising an electrical grid with 55 homes and a star communication network between them. In this use-case, we observe the number of messages for, and the impact of the communication latency on two management strategies of the electrical grid: centralized and decentralized management. One of the main outcomes of the proposed case study consists in the quantification of the electric network performances degradation observed with a 20 ms latency between each home and the central point on the star communication network. In this case, the decentralized approach leads to an overcurrent duration 2.6 times higher than with the centralized approach. Observations based on other metrics relevant to the power system community, such as cumulative shedding duration per heater, etc. are also provided thanks to our framework.

This work constitutes a first step towards a comprehensive framework able to co-simulate the interactions between the electrical grid and its management communication network. The simple use-case taken here does not highlight the scalability and versatility of the proposed tool, this analysis will be included in future work. The current use-case focuses on the impact of network latency, but other important factors can be studied, such as packet delivery reliability. Our future work includes studying optimized smart grid energy management strategies with SimGrid-FMI, and evaluating the impact of wireless communication QoS on these strategies.

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Fig. 2: Evolution of the current in Line1 over time with the centralized implementation.

Fig. 3: Cumulative overcurrent duration vs. latency.


