Interoperable meta model for simulation-in-the-loop

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Abstract—In the context of industry 4.0, where factory components become more and more intelligent, the role of virtualization and simulation becomes central. This paper presents a flexible, modular, scalable, extensible and interoperable modelling language expressly designed to support multi-disciplinary simulations. In particular, the language is based on a meta-model that provides patterns of both hierarchical and graphical aggregations and imposes a modelling process based on the principle of separation of concerns using a layered model. An example of applying the model to a business use case is also reported.

Index Terms—Digital Continuity, Interoperability, Virtualization, Collaborative Simulation, Digital Twin

I. INTRODUCTION

In the manufacturing context, simulation refers to a broad cluster of methodologies, techniques, and resulting software applications aimed at modeling and analyzing, through simulated experiments, the behavior of real production systems [1], [2]. Simulation enables to test the effects of design choices by exploring what-if scenarios without having to involve the real shop floor and, potentially, to evaluate its performance even before it is actually deployed.

Historically, designers and managers have used these tools to perform specific analysis, targeting separated factory domains [3] such as process validation, optimization of production layouts, scheduling and purchasing forecast, to name a few. Today, the rising complexity of distributed value networks calls for systems that extend beyond the single company borders and disciplinary domains, envisioning the application of simulation techniques to be integrated into cooperative digital environments. It is only by abating those barriers that industrial companies will be able to leverage on the availability of Big Data, to enable collaborative decision-making processes and to align intra- and inter-company operations. To fully achieve these goals, simulation needs to grasp the benefits coming from the rising of the disruptive changes experienced by its technological enablers. As more and more factory smart objects, namely Cyber Physical Systems (CPS), become capable of establishing a capillary sensing of the shop floor and make it available in real-time, the distance between the simulation run-time and the real factory situation results shortened. This allows moving from what-if analysis of scenarios to the support in now-what decisions applying simulation to current production issues. Moreover, since computational and reasoning capabilities are distributed with CPSs, simulation can now be distributed and located closer to shop floor by exploiting emerging paradigms such as Edge Computing.

However, as [4] points out, there are still relevant gaps affecting the maturity level in the adoption of these enablers in simulation and forecasting. First, although high-performing computing services are available in the Cloud, simulation still under-exploits them, being almost always stuck in a centralized and localized vision that failed to take advantage of the opportunities offered by Cloud and, recently, by the Edge computing. Even the game-changing possibility to reliably collect and use in real-time data to mirror the factory is not currently exploited within simulation tools still struggling in the definition of virtual factory models [5]. Lack of flexible and extensible data models are hindering the emergence of multi-disciplinary simulation technologies to support the virtual investigation in a holistic perspective that promotes decision-making processes enriched by the analysis of the multiple domains interacting within the factory [6]. Ultimately, digital models need to evolve coherently with production systems along their life-cycles to leap over both punctual and paradigmatic transformations taking place in the shop floor and in the information systems that support them [7].

II. INDUSTRIE 4.0: FILLING THE GAPS

In the current context, dominated by the advent of the Industrie 4.0, production systems are characterized by a higher complexity of decision-making processes that are loosely distributed among the CPSs operating along the entire value network. It is time to envision interoperable simulation services capable to operate near the data sources to support first-time-right solutions to multi-disciplinary issues occurring in production, logistics and management processes.

This paper proposes four items that should lead the path of simulation technologies development to embrace and empower the digital transformation of industry. First, a virtual interoperable shop floor representation, relying on a shared modeling of core data that can be expanded to suit specific purposes, is required to sustain integration of different domains and provision of cloud-based simulation services. Especially SMEs, the least capable among industrial organizations to access the benefits of simulation, will profit from this democratizing approach that removes or reduces the adoption barriers while expanding the scope of simulation. Interoperability of data models will pave the grounds for achieving situation awareness, the second
item, that embodies the digital doppelganger (a.k.a. Digital Twin) concept to create live digital copies of the simulated environments and processes thus making it possible to oversee and control them on a factual basis. The achievement of the first two items will allow simulation to take place in the loop of factory operations. This third item will dramatically change simulation’s role from a predictive and testing technique that is used ex-ante to make decisions on possible designs, to a responsive and holistic system to support decision-making in real-time scenarios. Finally, simulation models will need to be flexible and reconfigurable to leverage the smart nature of CPSs pursuing the plug-and-simulate condition, that is our fourth and last item. Plug-and-simulate makes adding new resources or modifying their behavior to be automatically or semi-automatically mirrored in the corresponding factory digital twin. Supporting such ambitious objectives calls for a set of requirements that data models have to fulfill. To this end, our data model needs to be:

**Expressive** - to enable the description of possibly any resource or flow involved in production, logistics and management processes of different industrial domains.

**Extensible** - to build, upon a core representation of the factory environment, a growing model ecosystem capable to maintain the digital information available all along the factory life-cycle.

**Interoperable** - to provide users with a personalized view and simulation tools that present the virtual environment from their perspective supporting the decision-making processes, activity planning and operation controlling.

**Scalable** - that is the ability to function efficiently when the context is changed in size or volume featuring multi-level access features and suitable aggregation patterns.

**Modular** - so that it provides representation building blocks that can be rearranged and reconfigured to follow the continuous evolution of the real factory.

### III. Interoperable Virtualization and Simulation

For Industrie 4.0, the shop floor ceases to be a rigid environment, firmly regulated by time-based automation systems; the whole approach to manufacturing has to be redesigned to be composed of intelligent elements (from simple sensors to production machines and robots) that are often identified with the term CPS [8]. These elements are capable of interacting with each other, with the environment, and with products by sharing information on their state, making the entire production environment highly flexible. In order for this vision to be achievable, the elements of the digital factory need to be **smart** [9], [10], that is, programmable and equipped with sufficient computational capacity to react in the appropriate time frame to events. Moreover, they must also be connected through communication channels that guarantee reduced latency.

The role of communication in smart manufacturing environments is twofold: on the one hand it ensures that the actors at shop floor level can interact, by exchanging valuable information, and orchestrate production in a distributed way. On the other hand it enables the almost real-time monitoring, paving the way for the fully adoption of CPS digital twins. In fact, data collected in large quantities from CPSs can serve both to refine/learn their behavior in order to obtain always-accurate simulation models (**Real-to-Digital Synchronization**), and to feed the simulation with events and data from the factory in real-time (**Mirroring**). Both features prelude the integration of simulation in the production processes to exploit its potential not only in the factory design and planning phase but also in the operative one with a multitude of possible applications (**Simulation-in-the-loop**).

Achieving such ambitious objectives confronts us with a plethora of challenges, including the definition of a flexible, modular, extensible yet interoperable modeling language to describe the digital double of the factory to come. Solutions are available, such as Modelisar’s Functional Mockup Interface (FMI), that provide dynamic system of simulation environments in form of input/output black boxes to support simulation models exchange and co-simulation [11]. In this paper we present a data model purpose-designed to support simulation in the framework of Industrie 4.0. To provide the modeler with a powerful and flexible tool, we built the model atop a core meta-model proposing a mixed white-box/black-box paradigm, generic enough to shape the traits of any CPS using plain (like **attributes**) and opaque terms (**artifacts**).

Simulation in Industry is fundamentally multidisciplinary, covering different application domains. Accordingly, the same digital twins can be typified differently as distinct are the objectives to achieve. In order to better manage the situation where different simulators are involved, but also help modularize the work of people involved in the modeling, we have enforced the principle of **separation of concerns**. Moreover, in order to provide maximum flexibility in the modeling operations as well as to increase the possibility of the model being adopted by companies, the element of **artifact** was added. Both these features will be presented in full detail in the next section, while here we aim at briefly presenting the intended objectives. As for the **separation of concerns**, we have structured the model to enforce the digital representation of a production layout (from a single machine up to a whole factory) as a set of layers, on the assumption that each of them could be devoted to the description of a different point of view or domain. Consider, for instance, the case of a model to be processed by a Kinematic and a Discrete Event Simulator (DES). We picture it as organized in at least three levels, a first level containing kinematic and 3D representations, a second level featuring a set of logical models describing the behavior of entities in reaction to events, and the third level linking the physical description to the related logic.

As far as the concept of **artifact** is concerned, the proposed model features generic files as first-class citizens in the description of the digital twins. This choice was deemed necessary as simulators often use very different behavioral models, and they export them using incompatible (often proprietary) formats. In other words, the modeler can decide to define digital twins using exclusively the syntax and constructs...
provided by the language (white-box approach), or else use specific and opaque formats whenever (s)he deems it necessary (black-box approach). It is important to note that our modeling language offers syntaxes, concepts and types of data characterized by different degrees of granularity, which, also thanks to different patterns of aggregation, guarantee the greatest degree of freedom to the modeller. Moreover, we emphasize that the choice to conceive a model as the set of independent layers, possibly related to different domains, guarantees great flexibility and allows to create a single representation model yet usable by different simulators (interoperability).

Figure 1 displays the general architecture of the simulation-in-the-loop system as conceived within the project motivating our work. The model is created and managed via an open API. This can be used both by modelers to create the digital representation of the factory at issue and by simulators. The model is saved in a repository and continuously updated through a module in charge of implementing Real-to-Digital synchronization. As regards the Open API for virtualization they are structured in two levels. The lower level API allows the management of the core elements while the upper level features endpoints specific for each simulation class. In this level all the logic that enforces a semantic coherence of the domain (e.g. integrity check, validation) is implemented.

The Real-to-Digital Synchronization can be defined as the process of continuously updating of CPS models stored in the Model Repository, tracking the evolution of the shop floor. Having up-to-date models is especially important in simulation since CPSs along their life cycle are subject to aging, straining, and reconfiguration processes, which can change their behavior and performance. In this situation the simulation outcomes may result substantially different from reality and, therefore, of very limited utility. In order to make decisions based on reliable simulation models, it is of paramount importance to detect changes in the CPSs automatically and continuously, and to adapt parameters and scenarios accordingly. The core of the synchronization grounds in the processing of data gathered at shop floor level. Synchronization with shop floor, while massively involving the digital representation of the factory, goes beyond the objectives of this document. An extended discussion of this subject is deferred to future publications.

IV. A Model for Simulation

One of the most common features of manufacturing plant design is to use standardized components (machine tools, robots, etc.) composed in a modular way. Through a good organization of modules, in fact, it is possible to speed up engineering as well as the simulation setups, maximizing the reuse of components. Similarly, simulation software solutions often provides libraries of models that can be aggregated and extended to assemble full plant layouts. In this work we aspire to provide the same efficient re-use approach implementing classes to describe resources, called Archetypes and Elements. The Archetype-Element relationship is similar to the one that exists in Object Oriented Programming (OOP) [12] between a Class and an Instance (Object) of that class.

Another important requirements is that the resulting data model shall support semantically meaningful collection of relations (referred to as Layers). We deem particularly important to provide the modelers with a tool to gather links of the same type. The idea is to simplify the modeling process by dividing the overall model in several layers. By way of example, a first layer may contain the definition of the plant topology with its hierarchies of production resources; the other are graphs that can be used to express relations between resources. Logical, electrical, and pneumatic layers are just a few examples of layers. Figure 2 illustrates graphically this multi-layer modeling approach.

This remainder of section documents the resulting data model for simulation, which is organized into 9 areas shortly described below:

- **Core Model** - documents the core classes used for the definition of entities and relations used in other sections.
- **Archetype Model** - introduces the concepts of Archetypes as the basis of the model reuse paradigm.
- **Element Model** - this model presents the concept of Element as instantiation of an Archetype according to the OOP.
- **Connection Model** - the connection model is devoted to the definition of the links between two model entities.
- **Layer Model** - presents the concept of layer as a collection of items that can be further specified using attributes.
- **Attribute Model** - specifies the concept of property to be used to annotate other elements of the model, namely Library, Archetype, Artifact, and Endpoint.
Role Model - presents a set of classes that defined the concept of Role as semantic description to further specify the constituent elements of the model.

Plant Model - introduces the Plant as an aggregation of layers representing among other aspects its physical and logical nature.

Project Model - documents all the classes that represent multi-plant simulation projects, and enable simulation tools to share plant models and results.

In this paper, nonetheless, due to the limited space available, only the main concepts and elements of the meta-model will be presented.

A. Core Model

Our modeling language is entirely upheld by a meta-model featuring the ground concepts and the syntax. Nine entities are contained in the core model. Figures 3 and 4 illustrate graphically such entities along with their relations, using the UML class diagram notation. We have chosen to divide the core model into two separate diagrams to improve the fruition. The central components of the model are: the Archetype and the Element. These have a symmetrical structure: both are characterized by one or more roles that identify them semantically, by a unique identifier within the model, by attributes that can be complex at will, by endpoints and artifacts. An element can inherit from an archetype but unlike OOP this is not strictly necessary. This means that an element can be defined as a stand-alone component or that it is possible to add or modify the characteristics inherited from an archetype.

The other elements of the model are:

Semantics, identification, and grouping of language elements, namely SemanticRole, AbsIdentifiedSemanticClass, and Library.

Artifacts - references to external relevant models treated as third-party black-box components (e.g., Geometry Models).

Attributes - properties used by the modeler to specify Archetypes and Elements.

Endpoints - semantically defined extremes of links connecting Archetypes or Elements. Links, and the rules by which they can be created, are defined in the Connection Model.

B. Artifact Model

One of the cornerstone requirements that our data model has to fulfill is to provide support also for black-box attributes, referred to as Artifacts. It will allow managing elements that represent information that is processable only by a specific software solution. An artifact, in our model, can be defined as an informative fiche containing, among other pieces of information, a URI that points to a particular file. Four further specifications are provided off-the-shelf (see Figure 5):

KinematicModel - this class represents a reference to an external model carrying information about the kinematic behavior of the artifact the model is referred. For instance a file in COLLADA [13] format also carries information about the kinematics and could be a candidate for this kind of model. Nonetheless, specific simulators might use their own binary format to express kinematic properties.

GeometryModel - this class embodies the concept of a black-box model holding geometry-related information about the particular plant resource. 3D models as COLLADA or JT [14] belong to this particular category.

LogicalModel - this class serves as a reference for models that define the behavior of plant resources. Such models, expressed for instance in PLCOpen XML [15], fully describe the actions of the actors involved in the plant as machines, CPS, sensors, personnel or software. While other models describe the static nature of a plant, Logical Models (or Behavioral models) describe the way the elements react to events (or clock ticks).

SynchModel - this class represents the synchronization model. In a nutshell, such models are executed by an edge-computing platform to carry out the real-to-digital synchronization process. This is the mechanism in charge of updating the digital representation of a plant resource through the processing of data collected on the shop floor.
A comprehensive description of the synchronization model, the abstractions on which it is based and its execution mode is beyond the scope of this document.

C. Connection Model

The connection model, within our model for simulation, is the part devoted to the definition of the connection mechanisms between two entities. To this end, the Endpoint class (see Section IV-A) is extended and the concepts of Link and ConnectionLayer are introduced. A link is described by means of an abstract class (AbsLink) then extended into two concrete classes (UndirectedLink and DirectedLink). The UndirectedLink joins together two instances of the Endpoint class and represents a connection between the elements to which the endpoints belong. This link does not provide any information about the direction of such a connection. To express the concept of direction, a DirectedLink has to be used instead. Figure 5 shows the components shaping the connection model along with the relationships that exists among them. The reader should notice that links are collected within either a DirectedConnectionLayer or an UndirectedConnectionLayer, both instantiating the template class GenericLayer(T). Finally, the Endpoint, Link and ConnectionLayer, in their directed or undirected form, are subject to the constraint that all the actors must share the same role to be semantically meaningful.

The purpose of this model is to provide the tools (links and layers) to define aggregations of elements in the form of direct or indirect graphs. Supporting this aggregation pattern is one of the model requirements that, together with the hierarchical aggregation pattern, allows to shape one or more plants by superimposing different layers. The advantage of this approach is that it enables distributed, flexible and independent creation and evolution of the model. Different experts, in fact, can work in the definition of different elements and connect them by means of a graph of semantically defined links.

D. Plant Model

The concepts presented so far define a meta-model containing core items and tools to extend them. Starting from those core classes, specific manufacturing and simulation concepts have been devised. The Plant component is one of those, it is a specification of the Element class featuring set of resources (physical, logical, and product) as well as their connections. Furthermore, since one of our goals is to enable a layer-based modeling approach wherein different actors (experts in different domains) can easily collaborate in defining a complete simulation model (see Figure 6), the fact that the structure proposed to represent a plant also includes several layers should not surprise. Some of them are mandatory and specific as the physical, logical and product layers, which model hierarchies of physical, logical and product entities, respectively. Other are optional like one or more GenericLayers, which model graph-shaped relations among the elements of the mandatory layers. Furthermore, Endpoints can be defined to model connections among plants (via Endpoints). Plants in fact can be aggregated into their own layer that, together with other pieces of information, defines a Simulation Project.

V. USE CASE: WHITE-GOODS PRODUCTION

The rationale of this section is to present a use case scenario, highlighting the way it has been modeled using the proposed data model. The use case describes the hob handling and palletizing operations carried out within a real production plant in Italy. The plant produces over a hundred different configurations of cooking hobs, belonging to three categories: gas, electric or pyro-ceramic. The production is organized in batches; hobs of the same type are aggregated taking into account different orders. Each batch (and every hob in it) is identified by a code to ensure a high level of traceability. For the same reason, as soon as a single hob is released, an automated reading system records its associated event. The cooking hobs are realized by different production stations; each station produces one hob at a time and it is specialized in producing items of only one category. Seven palletizers are placed at the end of the stations; they have the task of aggregating in a pallet elements belonging to the same lot, assigning a unique 2D barcode to the pallet and placing them on a conveyor. The length of the conveyor, its average speed, as well as the shelf life of the pallets on it are known. On the end of the conveyor there is a 2D barcode reader and an automated sorter system. The sorter has the task of dispatching the pallets out of the conveyor to different shipping lines. The plant features 18 shipping lines freely assignable to any type of product.
decisions made. In what follows some highlights are reported: explained the needs the model addresses and motivated the within the frame of Industrie 4.0 is presented. We also built to support multi-disciplinary virtualization and simulation: specifically:

- a Physical layer, defining the main elements of the plant along with their artifacts as geometries, representing 3D models of the components (conveyors, sorter and bays) of the graphical simulation model;
- a Logical layer, defining the internal automation logic of all the components involved in the simulation;
- a Product layer, describing the product stacks and the single product units composing them;
- a DirectedConnectionLayer named SimulationConnectionLayer, which connects the physical elements and the logical ones.

VI. CONCLUSIONS

In this paper a flexible and innovative data model purpose-built to support multi-disciplinary virtualization and simulation within the frame of Industrie 4.0 is presented. We also explained the needs the model addresses and motivated the decisions made. In what follows some highlights are reported:

1) In order to grant a high degree of re-usability the Object Oriented Paradigm has been taken as reference to develop the Archetype-Element duality. In this way the model enables the creation of libraries of components to be reused in different simulation/virtualization scenarios.

2) To achieve a high level of abstraction, flexibility, and compatibility among different simulators a meta-model featuring a white-box/black-box approach has been proposed.

3) To streamline the process of modeling in a multidisciplinary scenario, a layered pattern is implemented and enforced.

Future work will concentrate on providing the language with concepts pertaining to time, and CPS state. In addition, the language will provide specific constructs for the description of data flows generated by CPS with the aim of facilitating and automating the real-to-digital synchronization process.

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