

ESTABLISHING SEMANTIC INTEROPERABILITY IN CONTEMPORARY ARCHITECTURAL PARADIGMS

Consolidating loosely coupled semantics

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Abstract

Background: *Access-and-Play semantic interoperability (sIOP) is the next glass ceiling in IT-based business collaboration. Current approaches towards sIOP still rely on conventions on the semantics of the exchanged terms, which can be considered both accepted folklore and an impediment to access-and-play sIOP. A breakthrough for this impasse requires consensus on the foundations of semantics and sIOP. In a previous paper we already conclude that in software, semantics cannot exist and is reduced to the reciprocity between data and software code.*

This reciprocity can be contained s and software are on odds with each other: software can only operate on a token-based machine whereas semantics require an interpretation outside the realm of tokens. We resolved this fundamental incompatibility by establishing that whereas semantics in software, its best and However, the effort of the inevitable human-in-the-loop can be reduced and her position improved. This is a matter of software architecture, and it should address how semantic interoperability can be consolidated by it.

Objective: *The objective of our study is to identify and formulate the fundamental demands towards access-and-play interoperability, to derive their supporting architectural principles, and its integration in contemporary architectural paradigms.*

Method: *We follow a design-science approach and address the business relevance of the problem, and identify six requirements on sIOP, two of which are concerned with a genuine understanding of semantics that demand for the human-in-the-loop. We assume that the collaborating agents have followed the architectural principles on semantics according to our preliminary study (Brandt et al. 2019), which results in an explicit representation of an atomic semantic monolith for each of the agents: two semantic anchorages. We subsequently develop*

four design principles in order to support interoperability between the semantic anchorages, and one design principle to cater for the semantic distinction between a formal semantic correspondence and the necessary data transcription during communication. We finally evaluate these principles by designing and formulating an ISO-42010 Architecture Viewpoint and View on sIOP.

Results: Semantics in software are the result of a reciprocity between data and the software code that operates on them, resulting in a local semantic monolith (Brandt et al. 2019). Data exchange breaks that semantic monolith and hence the aforementioned reciprocity. The main concern of sIOP is to re-establish a valid reciprocity between the internal data processing code from the receiving agent and the external data as received from the producing agent, without extending the semantic monolith from either agents. We show that loosely coupled semantics, semantic alignments and a shared ontological commitment of the applied modelling language can be considered the cornerstone to achieve sIOP. The supporting principles are: (i) assume responsibility for the semantics of one's data, (ii) maintain an explicit ontological commitment, (iii) abstract semantics from the communication syntax, (iv) align the internal and external semantic meaning of the exchanged data, and (v) encapsulate how agents exchange semantic meaning. This results in a loosely coupled semantics that is re-usable for every interoperating peer agent, even those that are not anticipated for during the agent's design. The resulting ISO-42010 Architecture Viewpoint and View on sIOP, including a semantic mediation capability, represents a pattern to consolidate sIOP in contemporary architectural paradigms.

Conclusions: The major shortcomings in architectural paradigms to account for an access-and-play sIOP are their negligence of a separation of concerns between the semantic representation and data communication syntax at the one hand and human-authored alignments and the automated mediation process at the other, and establishing the conditions in support of in advance. By their explicit inclusion, we show that loosely coupled semantics can be consolidated in contemporary architectural paradigms, stimulating access-and-play sIOP.

Chapter 1

Introduction

Never before, data were so ubiquitous, and managed access to external data was so easy. But *understanding precedes use*, and understanding the data requires a human-in-the-loop and, therefore, is time-consuming and hampers agility in business collaboration in all domains. For instance, consider the following (allegedly real) example of an interoperability failure.

A German steel producer upgraded its industrial process robot. Since the majority of the steel production process is dependent on time, from a safety point of view the decision was made to not rely on their own internal clocks but to use the German *Braunschweig Funkuhr* time radio signal as source for the exact time instead. At the end of April 1993, when Germany went on summer time, the computer clock of the steel producer went from 1:59 AM to 3:00 AM in one minute. This resulted in a production line allowing molten ingots to cool for one hour less than normal. When the process controller thought the cooling time had expired, its actions splattered still-molten steel, damaging part of the facility.¹

In this simple example a tiny difference in the meaning of `time` between the steel producer and the national time provider hampered interoperability to the extent of damaging the steel facility. This tiny difference rooted in the assumption by the steel producer that `time` expressed a continuous scale whilst for the Braunschweig Funkuhr, `time` denoted instant clock time for that time zone and therefore represented a non-continuous scale. In order to achieve that both collaborators, here the Braunschweig Funkuhr and the steel producer, can actually *use* their peers data, the need exists to design and implement wrappers that remove any inconsistency between the variations that may occur in terms, structures, dimensions and what have you. Many such variations exist, leading to a range of failures in so-called *semantic interoperability* (sIOP) and ?? provides for a short overview of sIOP-faults. Unfortunately, it is fundamentally impossible to automate the production of wrappers, because we need a genuine *understanding* upfront, which computers still cannot do. “Despite the notoriously difficult philosophical questions involved, semantic interoperability can be seen as an engineering problem, namely that of effectively constraining interpretations towards the ones that are considered allowable” (Kuhn 2009; in Scheider 2012).

The most disconcerting consequences of a lack of (automated) sIOP are time-to-deliver, flat interoperability failures, and as seen above, seemingly correct but quite invalid data analysis results leading to faulty system behaviour. Current sIOP implementations are essentially based on the (time-consuming) process of establishing a (local) convention on the semantics of the terms that are exchanged during collaboration, requiring custom solutions and collaboration-dependent software adaptations. Such conventions can be considered a semantic monolith, which makes dealing with data that originated outside the monolith impossible, unless again a time consuming (months) semantic adoption process is applied. Moreover, these semantic conventions consider semantic heterogeneity a bug instead of a feature necessary to achieve semantic accuracy. Nevertheless, this

¹ Source: <http://catless.ncl.ac.uk/Risks/14/57#subj1>, accessed May 20, 2018

conventions-based approach towards sIOP is considered accepted folklore, even state of the art in ICT. In view of the large uptake of the Internet, the Internet of Things (IoT), cloud computing and big data, and in view of economical pressure to intensify enterprise collaboration, we consider this approach “too little, too late”. Some form of automation is required to resolve these issues, and we place formal semantics at its core.

In comparison, scalability was a big architectural concern in the past, requiring custom solutions as well. In response to this concern, scalability was standardised in the form of architectural patterns, and finally totally embedded and hidden into the infrastructure. Similarly, sIOP can be considered the architectural concern of this decade. We first need to provide standardised solution patterns that address semantic concerns before we can embed it in a technological infrastructure. Only then we can claim that sIOP becomes transparent to the developer, and only then we can take down the tight coupling between semantics and the syntax of the shared data scheme. Where scalability resulted in a huge increase in performance-demanding applications against a fraction of the original costs and effort, business agility will emerge once the semantic monolith is removed and semantic services exist at the infrastructural level. Then sIOP becomes an access-and-play operation that can be achieved in due time with data not anticipated for during software design, and at any point in their life cycle. Metaphorically speaking, we consider sIOP as a *bridge* overarching a (semantic) gap: with *anchorages* (semantic concerns) on each side of the gap, with a *spanning* (semantic alignments) resting on them to structurally (semantically) support the interoperability bridge, and with a *roadway* (data mediation) enabling the crossing of the (data) traffic. Finally, architectural *principles* provide the necessary guidance to the architect for the various design decisions that effectively result in a particular bridge over a particular (semantic) gap. This has been depicted in Figure 1.1.

Our contributions to consolidating semantic interoperability in software architectures are fivefold, and represented as architectural principles and concerns, as follows:

- *Semantic concerns (anchorage)*: Abstracting semantics from a tacit software implication into a tangible, computational and distinct artifact provides us with the potential to connect to it and to make comparisons with the semantic artifact of the peer software agent. Based on the disciplines of semiotics, philosophy, modelling and mathematics we elaborate in (Brandt et al. 2019) how to achieve a semantic anchorage. We formulate the principle of assuming responsibility on the semantics on data, and conclude what preparations about semantics are required for an agent before being able to engage in semantic interoperability (Chapter 2);
- *sIOP concerns (spanning)*: Since computers remain incapable of true understanding, sIOP remains in demand of human intervention in order to reconcile the semantic differences between collaborating software agents. However, human intervention is time consuming. We reduce the necessary human intervention to complement formal semantics to a task that suffices to achieve sIOP, viz. authoring semantic alignments only (Chapter 3);
- *Mediation concerns (roadway)*: We determine the demands for a generic component that allows for communication with the peer agent in one’s native vocabulary only, by considering both ontological models and the alignment. Such approach applies the principle *connectivity without dependency* at the semantic level. This consolidates the agent’s potential to collaborate to any unforeseen applications without the need to adopt external semantic definitions, and remain scalable in the process (Chapter 4);
- *Evaluation of semantic principles*: In order to consistently address the above concerns, their founding architectural principles have been derived. It is a matter of architectural hygiene to evaluate how these principles (??);
- *ISO42010 Architecture Viewpoint*: We verify the applicability of the above concerns and principles by formulating their architectural consequences as a specific ISO42010 sIOP Viewpoint, and we show their proper position in the total architecture as corresponding sIOP view. As ISO42010 is considered a set of best practises for architecture description, and therefore is used with architecture frameworks such as MoDAF, TOGAF, DoDAF, RM-ODP and so on, we conclude that our sIOP Viewpoint and View can be

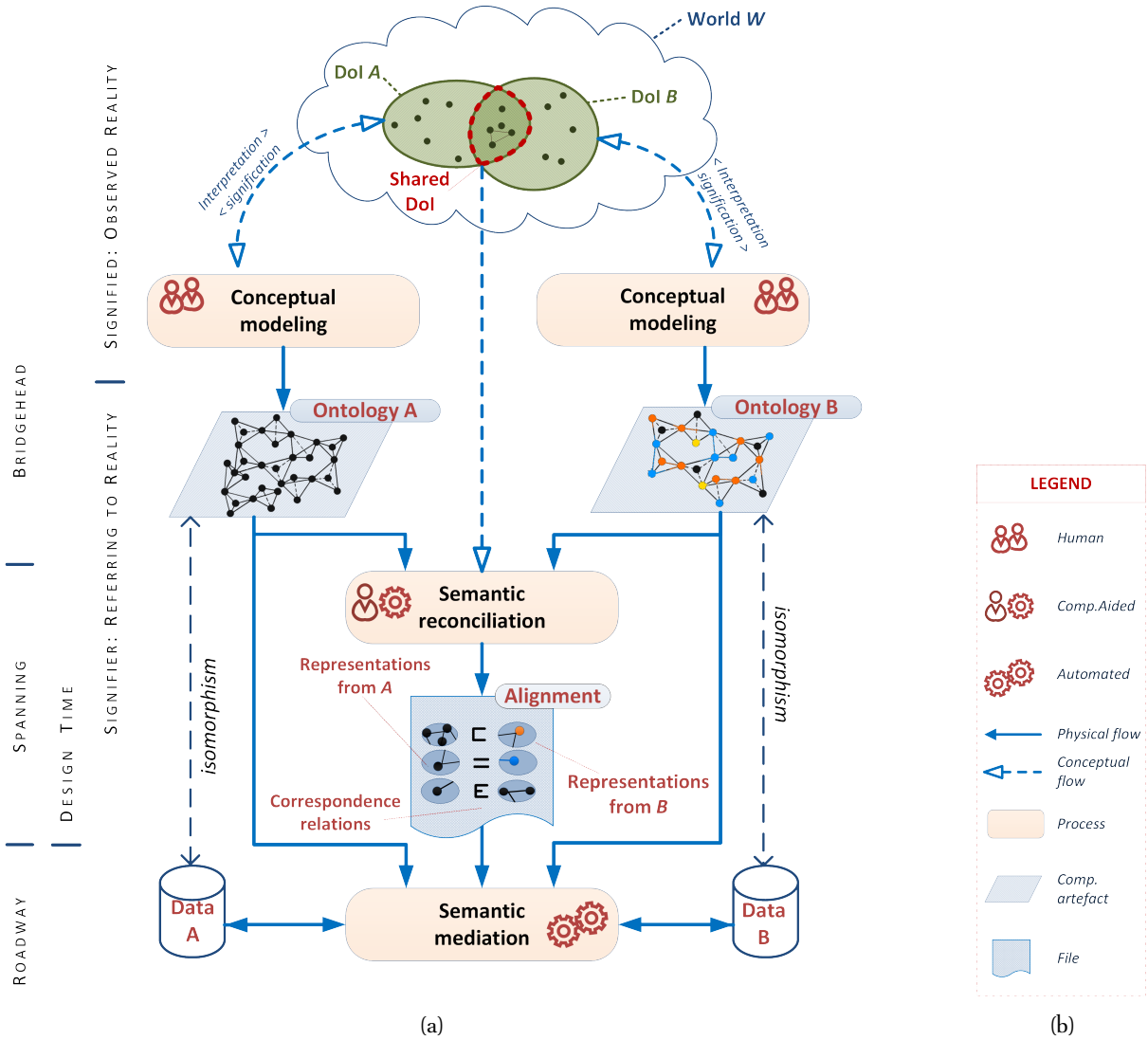


Figure 1.1: Conceptual overview of the relationships in sIOP between the anchorage (conceptual modelling), its spanning (semantic reconciliation) and roadway (semantic mediation), (a), and a legend explaining the used constructs (b).

considered to consolidate sIOP for contemporary architectural paradigms (Chapter 6).

This could imply just another day in the office, because for ICT architects and software engineers, refactoring the software agent in order to separate distinct functionality into a new, valuable component is one of their core skills. Application of architectural principles bring about the new component, including an API and standard for its access and use. Its adoption by IEEE, W3C, or a governing domain organisation suffices for its integration at the right infrastructural level. *Quod non est* for semantics, at least not that simple. “The successful standardisation of protocols made us believe that we should also *standardise meaning* on the Web. This is a fundamental *misconception*.” (Janowicz and Hitzler 2013). Because, this is the first time in the history of ICT that its discipline, viz. we the ICT community, are not speaking about a concern that belongs to the realm of information and communication technology but to that of the users instead. Consequently, we cannot control it, hence the existence of semantic heterogeneity: how users represent what they mean and what is meant with what is represented, differs between every so many stakeholders. Despite this observation, current viewpoints on semantics defy semantic heterogeneity and strive for semantic homogeneity: one single

agreed domain convention on how the syntactic representation and structure of the data or messages shall be semantically interpreted. Despite our, the ICT community, acceptance that semantics is a representation of some part of the world, viewed from a particular perspective of use, we don't acknowledge that this *particular* representation and *particular* perspective is fundamental to the domain users when building semantics and getting an understanding. And equally important, we don't acknowledge that this particular perspective is just one out of many equally legitimate ones. Some examples are given in Table 1.1.

Table 1.1: Semantics follows many alternative but equally legitimate points of view on reality, implying that no single one true representation exists. Hence, semantic heterogeneity is a feature that should be preserved, as opposed to a bug that should be sought to correct.

Reality to refer to	Perspective #1	Perspective #2	Perspective #3	...	Perspective #n
Waves are	<i>objects</i> that I can point at	<i>forces</i> exerted on dikes, walls or oil rigs	an <i>accumulation</i> of other waves	...	a <i>relation</i> between water, current, wind and the seabed
How many terrorist attacks can be counted on 9/11?	1, the number of involved terrorist groups	2, the number of collapsed towers	4, the number of involved aircraft	...	2996, the number of casualties ²

We consider semantic homogeneity not only an illusion, but striving for it brings about several important disadvantages. For instance, implementation of semantic standards involves considerable expenses, and tends to solidify in software by implicit operations on data. Moreover, since semantics fixes representation, semantics becomes conflated with syntax, making the semantic standard dependent upon variations in syntax and, hence, the architectural paradigm, its implementation or applied technology. Together, this brings about a high impedance to change or evolution, which creates a significant barrier for new business opportunities, obstructing business agility. Finally, it is sheer impossible to create for each particular domain, or part of that domain, a semantic “standard”; even in the highly protocolised health domain that already addressed semantics in 1987 (Spronk 2014), adoption of a standard remains slow and cumbersome [REF]. This becomes all the more clear when comparing with the successful adoption of 2G GSM telecommunication standard, including its evolution to 2.5G, 3G, 3.5G, 4G and 5G standards during roughly the same era. We defend that software semantics should become explicitly specified “(...) not in order to know *what there is* [i.e., striving for semantic homogeneity], but in order to know what a given remark or doctrine, ours or someone else's, *says there is*” (Quine 1961). Once semantics becomes explicit and computational, semantic heterogeneity can become a feature as opposed to a bug (coined by (Janowicz and Hitzler 2013)).

Based on these contributions we conclude that there is a lot to gain towards access-and-play sIOP, despite the fundamental impediments in ICT to create an automated genuine understanding. By separating syntactical from semantic concerns and turning semiotic fundamentals into architectural principles, loosely coupled formal semantics emerges and can be consolidated in contemporary architectural paradigms. From that position it is only a small step towards embedding standard semantic services into the communication infrastructure. We first describe some background on the semantic foundations in the following section.

² as mentioned by Wikipedia, https://en.wikipedia.org/wiki/Casualties_of_the_September_11_attacks, accessed Dec 13, 2018

Chapter 2

Anchorage: Semantics

Despite the precise meaning of the term ‘semantics’ in semantic interoperability, it is clear that sIOP encompasses a communication between at least two actors. This brings a natural responsibility for both actors in the communication, described by (Grice 1975) as the particular purpose of communication, viz. to serve:

1. Quantity: Make your contributions as informative as is required (for the current purpose of the exchange), and not more than is required;
2. Quality: Do not say what you believe to be false, or for which you lack evidence;
3. Relation: Be relevant (to the immediate needs);
4. Manner: Avoid obscurity of expression, ambiguity, and be brief and orderly.

This leads to the definition of a design principle to its effect, applying the normative notation from (Greefhorst and Proper 2011):

Design Principle 2.1 (The responsibility for the semantic meaning of data lays with the source)

When it is reasonable to expect that the software agent will be engaged in collaboration or otherwise will interoperate with (an)other software agent(s), it is the responsibility of the software architect to serve the quantity, relation and manner of the potential interoperability by specifying the semantics of the data in advance.

Type of information: business, data

Quality attributes: semantics, semantic interoperability, usability, efficiency

Rationale:

1. Data represent the state of affairs of some part of the world, viewed from a particular perspective of use. Such view is just one particular perspective out of many equally legitimate ones;
2. Semantic heterogeneity, a direct consequence of the equally legitimate perspectives on reality, should not be considered a bug to resolve, but a feature to preserve and nurture in order to maximise semantic accuracy and relevancy;
3. Accepting semantic heterogeneity implies the probable uniqueness of the agents view on reality;
4. Computers are not capable of genuine understanding, hence cannot establish semantics from data and thus require the human-in-the-loop for that;
5. The responsibility for formulating the semantics that are expressed by the data can only lay by the software architect that has taken the particular perspective on reality when carving out the entities of interest to the software application;
6. On specifying semantics, Grice’s maxims on communication, and particularly on serving the quantity, relation and manner of communication, represent the natural constraints to respect;

7. Without adherence to this principle, the meaning of the data expressed by the software agent can be considered flawed, inaccurate, incomplete or otherwise insufficient in its support for semantic interoperability.

Implications:

1. The specification of the data semantics is only dependent on the agent's own perspective on the application domain, and can therefore be fulfilled before any interoperability with communication peers;
2. No matter the number of different communication peers, the software agent needs to specify the semantics of its data elements only once;
3. By providing an explicit semantic specification of the data, an agent facilitates other components and agents to connect to it and, consequently, grounds its semantic interoperability with them unequivocally;
4. Application of the same modelling language implies the use of the same ontological commitment, resulting in improved and comparable consistency between their models of reality. ◇

We argued in (Brandt et al. 2019) that since software is incapable of genuine understanding, semantics cannot exist in software. Nevertheless, the software agent acts as transport medium for semantics: for single-user software as the medium that transports the semantics as it was intended by the software engineer to the semantics as it is experienced by the end user at the human-machine interface; for multi-user software as the transport medium for the semantics as intended by one end user at the time of data insertion, to the semantics as experienced by another end user when retrieving the processed data. To act as valid transport medium for semantics, we further stated that the reciprocity between code and data does manifest itself as software semantics. This essential disposition discerns in semantics its *semantic meaning*, i.e., what is said and carried by data, and the *pragmatic meaning*, i.e., to connect with our frame of reference and carried by code. The latter implements comprehension as an inference process that starts from a set of premises and results in a set of conclusions that are warranted by them (Grice 1975). We explained that by observing that data and code are always tightly coupled and since their reciprocity emerges as software behaviour, software malfunction originates (amongst others) from a broken reciprocity, i.e., inconsistencies between data and code. Consequently, when the data and code are representations of the things and laws in the application domain and, hence, represents semantic meaning and pragmatic meaning, their reciprocity represents the degree with which the collective outcome of processing all potential data refers to the intended states of affairs in reality. Any incoherent reciprocity equates to unfaithfulness: semantics that are considered invalid in the application domain.

Despite the quality with which the data and the code are developed individually, we can maximise semantic validity by maximising their reciprocity, viz. demanding maximal coherence between code and data. We have called this the *semantic coherence principle*. The consequence of demanding high coherence between the data and its processing code is in its inevitably emerging monolith, which we denoted as the Atomic Semantic Monolith (ASM): a semantic monolith, for it refers to the monolith's reciprocity between data and its processing code that describe the affairs in the application domain; Atomicity refers to the level of granularity at which the entity that is referred to by the data token is considered a non-dividable whole in the application domain. Where it is the objective of sIOP to address this monolithic nature of the ASM, as we do in the next section, it is the objective of semantics to maximise and maintain the coherence of the ASM, as elaborated in (ibid.).

Regarding the quality of the data and code models we reasoned that the data model should have a backward-looking role (in contrast to forward-looking) (Gonzalez-Perez and Henderson-Sellers 2007), present an ontological mode of modelling as opposed to a linguistic mode (Atkinson and Kühne 2003), and demand a strong type-mode (as opposed to a token-mode) that result in non-transitivity and use the kind of abstraction known as classification (Henderson-Sellers 2012). From those demands, we concluded that for representing

semantics ontologies are best suited (Brandt et al. 2019). For example, the trueness of forward-looking models, i.e., all 42010:2011 models, is established against their meta-models, while the trueness of backward-looking models, i.e., ontologies, is established through the interpretation in the conceptualisation of reality.

Moreover, we showed (ibid.) that the predominant purpose of a model is to describe reality by distinguishing the entities of interest. These distinctions can be modelled, but the (modelling) language itself is also used to convey distinctions. The distinctions that are already articulated by the elementary language constructs define the expressiveness of that language; the more distinctions the language elements can convey, the more differences can be represented by that language. The (fundamental) categories that the language elements can discern apply during modelling as a *commitment*, e.g., the language that explicitly differentiate substantials into objects and amounts of matter commits to the intuitive difference between the cup and the coffee that it holds. When the modelling language does not commit to such distinctions, the user of the language is forced to specify these distinctions in the model itself. This so-called *ontological commitment* of the language (Bricker 2016; Guarino et al. 1994, Guarino:1998wq) lays the foundations for the model and its data and, consequently, for the code to process the data. Interoperating peer agents that apply different ontological commitments will therefore show major differences in the construction and internals of their respective ASM's. This observation leads to the following design principle:

Design Principle 2.2 (Maintain an explicit ontological commitment)

The language constructs that are used to formulate a model always represent an ontological commitment, explicitly or implicitly.

Type of information: *business, application, data*

Quality attributes: *semantics, semantic interoperability, reliability*

Rationale:

1. *The particular reciprocity that emerges in the ASM is influenced by the ontological commitment of the modelling language;*
2. *The purpose of sIOP is to re-establish a coherent reciprocity between the external semantic meaning and the internal pragmatic meaning;*
3. *Incompatibility between the ontological commitments of both interoperating agents creates a sIOP concern on the modelling language level;*
4. *sIOP cannot be established without having addressed this language concern;*
5. *This language concern and its related resolution is independent from any particular sIOP case;*
6. *By maintaining an explicit ontological commitment, its incompatibility with other ontological commitments can be addressed in a generic manner.*

Implications:

1. *Since the choice for a specific ontological commitment is only dependent on its applicability to the agent's semantics, its specification therefore is independent from any specific interoperability case;*
2. *No matter the number of different communication peers, the software agent needs to specify its ontological commitment only once;*
3. *By specifying its ontological commitment explicitly, an agent enables the emergence of a standard and related infrastructural components to address this concern and to provide for reconciliation of differences between ontological commitments.* ◇

Based on the principles and arguments in this section, we defend that software agents that might engage in interoperability should provide for a semantic anchorage in the form of a domain ontology and a foundational

ontology; the latter to explicate the ontological commitment and the former to specify the semantics of the data. Such ontology provides the ability to connect to the semantics of the agent in a computational manner, consolidating the semantic concerns for semantic interoperability.

Chapter 3

Spanning: semantic interoperability

3.1 Objectives of semantic interoperability

Semantic interoperability is about two software agents that share a particular reality in their domains of application, and exchange data that represent a certain state of affairs from that shared reality. Despite the (different) reasons that both agents might have for sharing the data, the only demand that is put on the exchange is to serve what was coined by Grice as the communication's quality: "Do not say what you believe to be false, or for which you lack evidence". Subsequent to the exchange the data will be processed by the receiving agent and it stands to reason that understanding the data precedes their faithful use. In conclusion, semantic interoperability discloses the capability between two software agents to faithfully use exchanged data that accurately represent the state of affairs about a particular shared reality.

We have seen that software semantics is necessarily reduced to the reciprocity that exists between data and code in the so-called atomic semantic monolith. Such ASM guarantees the coherence between the data and their processing code. Unfortunately, when communicating data they are necessarily separated from the ASM they belong to. (Why it is useless to exchange the complete semantic monolith in order to establish sIOP, is left as an exercise to the reader.¹) The consequence of data exchange on the semantic meaning (data), therefore, is twofold: it loses its coherence with its original pragmatic meaning (data processing code), and, a new reciprocity with the pragmatic meaning belonging to the receiving agent emerges. Unless it can be guaranteed that this new emerging reciprocity is as coherent as it needs to be, semantic interoperability cannot emerge from the data exchange and phantom semantics will emerge in stead. From this we conclude that the main task about establishing sIOP is to re-establish coherence between the external semantic meaning and the internal pragmatic meaning. Note that the resulting semantic monolith of the receiving agent will be different from the original semantic monolith, although they share the same semantic meaning. For example, by exchanging a heartbeat both agents share the a semantic meaning about the number of beats per second, however the pragmatic meaning can vary between an indication of health for an health-care application or an indication of performance potential in a sports application.

The **prime objective** from the perspective of the receiving agent is to assure that the reciprocity between data and code remains truthful to the state of affairs in reality. This relates for both the external data and the data that can be inferred from them. The two possible approaches are, (i) to modify the pragmatic meaning such that it can operate in a valid way on the external semantic meaning, or (ii) to modify the semantic meaning such that it can be operated on in valid way by the existing pragmatic meaning. The first approach clearly breaks one of the fundamental principles of software engineering, *low coupling, high cohesion* (e.g., Hitz and Montazeri 1995), by allowing external definitions to influence internal workings. The second approach only adapts that what was already external to the receiver, and thereby doesn't breach the integrity of its own software. By pursuing the second approach we also maintain a **second objective** for sIOP: ensure

¹ Answer: the semantic monolith would result in both agents to perform the exact same functionality with regards to the data.

that the ASM's from both agents remain independent from each other, viz. establishing a semantical loose coupling between both agents. Such loose coupling does not require one single homogeneous view on reality, which aligns neatly with the **third objective** for sIOP to allow for semantic heterogeneity: distinct agents will probably maintain alternative but equally legitimate points of view on reality, implying that semantic heterogeneity is a feature to preserve necessarily. The **fourth objective** of sIOP is to strive for access-and-play sIOP: ideally, sIOP between agents can be achieved instantaneously, also for unforeseen collaborations. This objective is very hard to achieve because when we accept that software is incapable of genuine understanding, and when we accept that correct use of data is to be preceded by its understanding, a human-in-the-loop to provide that understanding becomes a necessary condition for sIOP. The **fifth objective** of sIOP is to allow for semantic evolution and, consequently, the maintainability of sIOP. Finally, we consider that semantic heterogeneity brings about an issue of scalability, since semantics won't be a centrally coordinated anymore; in stead, semantic definitions will be distributed all over the place. We therefore include as **sixth objective** of sIOP to allow for scalable semantics.

We conclude that from these six objectives, only the first (re-establish reciprocity) and fourth (access-and-play) are concerned with genuine understanding of semantics, leaving the others as engineering challenges. Our focus in the subsequent section is only on achieving the re-establishing reciprocity and access-and-play objectives. We address the latter four as evaluation of the principles that have been introduced to achieve the two core objectives.

3.2 Explicit sIOP by alignment

We maintain the position that computers lack the capability of genuine understanding. We subsequently defend that with the current state of the art in AI the human-in-the-loop remains a necessary element in sIOP in order to cater for the understanding of the semantic differences between the interoperating agents. In its pure sense, an access-and-play capability can therefore not be established. However, current solutions on semantic standards solidify the understanding in the syntax of the data. In this way, semantics are carried by a data schema that is primarily designed to serialise data and to support data transfer by message construction and exchange. We consider this a significant neglect of the principle on separation of concerns, conflating the semantic interoperability concerns with the data communication concerns. The consequence of conflating these concerns is that source code which should be concerned primarily with message construction, parsing, storage, and other data communication related tasks, becomes dependent on how semantics influence the syntax. In a message-oriented paradigm, for instance, any difference in structure in order to reflect the local perspective on semantic structure will have a significant impact on how to (de)compose the message. And any new data source to connect to will proliferate into a new software release. We thus observe that the current approach to data understanding results in an architecture which imposes a significant complication on interoperability (and other -ilities as well), impeding access-and-play. And despite the current limitations of AI-software to genuinely understand, a significant gain towards the software agent's access-and-play capabilities can be achieved by untangling the syntax and semantics through separation of the sIOP concerns from the data communication concerns. We propose the following design principle to its effect:

Design Principle 3.1 (Abstract semantics from communication syntax)

When a software agent engages in interoperation with (an)other software agent(s), resolve their semantic differences independently from the syntax of the exchanged data.

Type of information: data, technology

Quality attributes: semantic interoperability, portability, maintainability, efficiency, usability (reuse), reliabil-

ity, functionality

Rationale:

1. Data schemata are defined to support the (de)serialisation processes that consolidate the data communications concern;
2. Neglecting the principle of separation of concerns solidifies dependency between otherwise disjoint concerns, here the semantic level and the syntactic level of data communication;
3. Access-and-play capabilities are supported by assuring minimal impact on software code when introducing semantic modifications;
4. Minimising impact on software code that is concerned with data communication is realised by abstracting semantics away from the data schemata.

Implications:

1. Separation of concerns has a strong positive effect on software quality, including but not limited to sIOP;
2. Removing any dependency between semantics and data syntax enables to support multiple communication paradigms without the need to modify the semantic abstraction;
3. Similarly, modifications in the semantic representation, or supporting multiple semantic representations become possible without the need to modify the communication layer;
4. Align semantics, not data schemata: Semantic reconciliation is applied at a higher conceptual level and abstracts away from data communication schemata;
5. Heterogeneous semantics from multiple data sources are more easily supported;
6. Semantic alignments imply the need for a mediation capability between the semantic representations of the communicating agents.

◇

This principle is in clear contrast with principle A.20 , Data that are exchanged adhere to a canonical data model, as determined in the Principles Catalogue from (Greefhorst and Proper 2011). Indeed principle A.20 reflects the current practises to achieve interoperability. It is not necessarily wrong, since following it results in achieving interoperability, as justified by the many if not all interoperable systems that exist today. However, its application impedes access-and-play interoperability since such capability require more specification and less implementation, more generic infrastructural solutions based on meta-standards as opposed to local solutions that rely on data standards.

brandtp, 18-11-2019 Move this paragraph to section Related Work?

Current sIOP practises already require humans-in-the-loop to reconcile the semantic differences that occur. Often, the subject of reconciliation is the differences in data schemata, and the result of the reconciliation is laid down as a canonical data model. By applying semantic reconciliation on the conceptual level, the dependency on the (data) syntax, and vice-versa, is minimised. Moreover, by representing the result of the reconciliation as an alignment (between ontologies) as opposed to a canonical semantic model (core ontology), the influence of the peer agent's semantics on one's own semantics is minimised as well. An alignment, thus, functions as an interface that enforces loosely coupled semantics by enabling semantic transparency between communicating peers. Reducing the human-in-the-loop to author an alignment only, (i) accelerates the deployment of sIOP by removing all human effort that is concerned with implementation activities, and (ii) decouples the sIOP scope to bilateral alignments only. This process has been depicted in Figure 3.1.

From the perspective of the receiving agent the first objective of sIOP is to guarantee that the reciprocity between the external semantic meaning and (our) internal pragmatic meaning remains faithful to reality. We've determined that we can only convert the external semantic meaning, and we need to do it such that

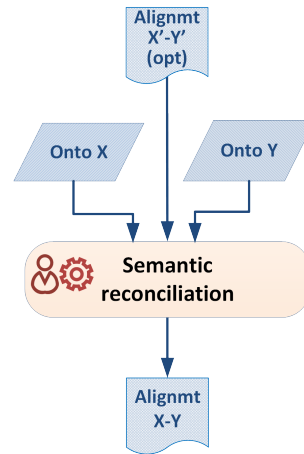


Figure 3.1: Semantic reconciliation results in an alignment between the semantic representations of two ontologies. We defend that semantic reconciliation is a computer-aided but ultimately human-authored task.

the result becomes coherent with the internal semantic meaning. We can assume that the quality of our own agent software is such that the internal semantic meaning is in coherence with the internal pragmatic meaning. Contrarily, we cannot assume that any extension of the internal semantic meaning, no matter how small, will remain in coherence with the internal pragmatic meaning. Founded on this essential disposition we conclude that aligning the external semantic meaning with the internal semantic meaning such that the former does not overlap the latter, is the only approach that can guarantee sIOP. We reflect this with the following principle:

Design Principle 3.2 (Align the internal and external semantic meaning of the exchanged data)

When a software agent engages in interoperation with (an)other software agent(s), establish for the exchanged data a maximal coherence between external semantic meaning and internal pragmatic meaning by formalising the alignment between the external and internal semantic meaning.

Type of information: application, data

Quality attributes: semantics, semantic interoperability

Rationale:

1. On processing external data, semantics manifest as the reciprocity between data and processing code;
2. Data are considered isomorphic to the semantic meaning as specified by their source agent;
3. Formalising a correspondence relation between the semantic meanings of collaborating peers connects the external semantic meaning with the internal pragmatic meaning;
4. Assuring that the internal semantic meaning encompasses the external semantic meaning, or assuring that the semantic consequences of the latter extending the former are insignificant, assures the semantic validity of the correspondence relation.

Implications:

1. The conversion from external to internal semantic meaning is specified by a correspondence;
2. The collection of all correspondences specify the semantic alignment that holds between a pair of interoperating agents;
3. Software agents that are unable to align their semantic meaning with the external semantic meaning cannot engage in sIOP without introducing phantom semantics, with unforeseen consequences in their data processing.

◇

A *correspondence* specifies as accurately as possible the semantic difference (out of those listed in Appendix A) that exists between a pair of related concepts, i.e., it aligns between the semantic meanings of interoperating agents. By exhaustively addressing all semantic differences that exist between both agents, the set of correspondences collectively specify the *alignment* that holds between two agents. The purpose of the alignment is to establish how the truth of expressions that are formulated in terms of agent A, can be established by using formulations in terms of agent A', and to capture their potential difference as a relation. To that end we differentiate between two categories of semantic differences:

1. *Conceptual differences*: variations that can be specified as logical relation between (constructions of) concepts from both ontologies, e.g., naming conventions or structural engineering variations;
2. *Value differences*: variations in conventions on how to represent values with or without magnitudes, e.g., differences in value scales, units of measure or nominal properties.

The language used to specify the correspondences must be expressive enough to identify the atomic elements of the ontologies, to combine them into logical combinations as well as to formulate the relationship that holds between them. In (Euzenat et al. 2007; Scharffe et al. 2011), an investigation has been reported towards the requirements for such an alignment language, summarised as follows. A *correspondence* denotes a single particular inter-ontological relation, prescribed, and assumed to represent a semantically valid relation between both concepts, as:

$$\mu = \langle e, e', \theta \rangle$$

with:

- $\theta \in \{=, \sqsubset, \sqsupset, \perp, \tilde{\cap}\}$ specifying the *correspondence relation* that holds between entity constructions from the source, e , and the target, e' . The basic correspondence relations denote $=$: semantic equivalence, \sqsubset : subsumption of, \sqsupset : subsumes, \perp : disjointness, and $\tilde{\cap}$: overlap. Although more relations can be required to include for a particular use case, such does not invalidate the general principle. Further note the correspondence relation is a directed relationship.
- The source and target *entity constructions*, e , are build on the atomic elements of the ontology language. An entity construction connects concepts by applying:
 - conceptual connectors:
 - * logical connectors *AND*, *OR*, and *NOT*;
 - * a path connector as a sequence of zero or more Object Relations, R , optionally ending with an Object Property P , summarised as follows: $R^*[P]$;
 - * property construction operators (inverse, composition, reflexive, transitive and symmetric closures);
 - * constraints (through domain, range, cardinality and value restrictions);
 - value functions:
 - * mathematical calculations operating on one or more values having a magnitude for, e.g., unit conversion;
 - * transcriptors operating on one or more nominal values without magnitude, e.g., ISO two-letter country code, or blood type.

Without the conceptual connectors it is only possible to address a single concept or individual as defined by the ontology, representing an aggregation level that is relevant for the software agent but might not be relevant in terms of the interoperating agent, and hence, for their mutual SIOP. By application of conceptual connectors the architecture gains the capability to address a specific compound of individuals in either the source or target ontology that relate to the semantic difference at hand. Similarly, with the application of value functions the architecture gains the capability to specify transformations between conventions on value representations and nominal properties.

Chapter 4

Roadway: Mediation

The mediation pattern has already been described in (Gamma et al. 1994), albeit in the context of object-orientation as opposed to sIOP. It is described as “an object that encapsulates how a set of objects interact”, and it promotes loose coupling “by keeping objects from referring to each other explicitly” and by enabling you to “vary their interaction independently”. In this way, the mediator turns a many-to-many interaction into a many-to-one interaction, each of which is easier to understand, maintain and evolve. The fundamental idea behind the pattern, viz. trading the complexity of the interactions with the complexity in the mediator, can also be applied on a semantic level, and we formulate the following principle to its effect:

Design Principle 4.1 (Encapsulate how agents exchange semantic meaning)

When software agents engage in interoperation, encapsulate how the representation of their semantic meaning should be transcribed without inducing phantom semantics.

Type of information: *business, data*

Quality attributes: *semantic interoperability*

Rationale:

1. *The semantic meaning is codified in (onto)logical representations;*
2. *Keeping agents from referring to each others representation therefore requires transcription between representations;*
3. *A solution where each agent needs to implement one transcription component between its own representation and each of its interoperating peer, increases complexity;*
4. *Encapsulating the transcription into an alignment-based intermediary component results in less communication complexity and relieves the agents from development and maintenance of local wrappers;*

Implications:

1. *A mediator creates representational transparency between communicating agents, keeping agents from using each others representations;*
2. *This enables independent development of the individual agent's semantic meaning;*
3. *The need to enforce a canonical semantic representation, viz. semantic homogeneity, expires, allowing semantic heterogeneity to become the norm;*
4. *Each agent can reuse its semantic anchorage in any other interoperation;*
5. *Data transcription logic can become a generic service provided by the communication infrastructure;*
6. *Each agent can communicate with any other agent in its own native semantic representation.* ◊

In this reading a mediator encapsulates how a pair of agents represent their semantic meaning and provides for a generic transcription logic to mediate between native semantic representations. However, one paramount issue must be resolved by the transcription logic of the mediator, as follows.

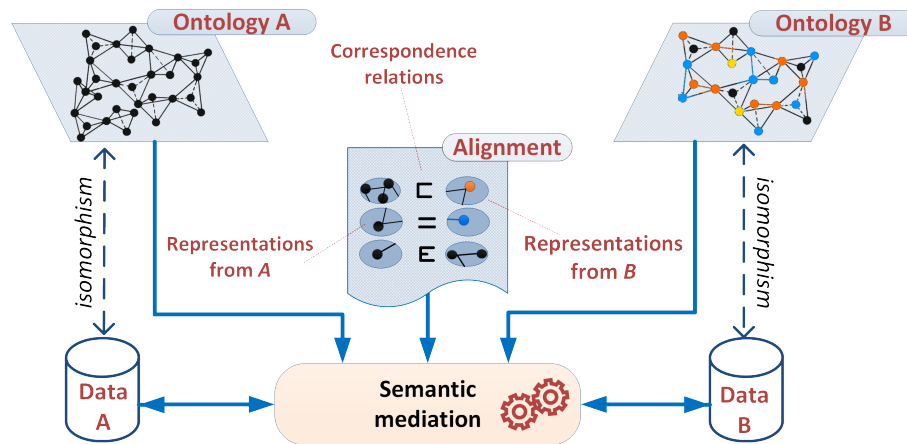


Figure 4.1: Semantic mediation encapsulates how agents exchange semantic meaning. It implements a generic transcription service between the particular representations of data between both agents. It depends only on the representations of the semantic meaning from both agents (ontologies) and the alignment that holds between them.

We have seen in the previous sections that a correspondence assures the semantic validity between the different semantic meanings of both agents. It can do so by token of the broad correspondence relations that can be specified, both for the conceptual and the value differences that may exist between them. Unfortunately, this is in sheer contrast with the fact that a transcription can only replace one term for another, which basically implies that an equivalence relation holds between both terms. As can be seen from the correspondence relation, this is not necessarily true. In (Brandt et al. 2018), we make a distinction between a naive transcription, which ignores this inconsistency, and a semantic valid transcription that can establish under what conditions this inconsistency does not produce invalid semantics. We show that these conditions rely on logical constructs only and, consequently, are independent on the actual ontologies that apply. We have formulated these logical constructs as rules that generically apply for any transcription. Unfortunately, a set of conditions remain for which a semantic valid transcription cannot be guaranteed, either due to an incomplete alignment or to (onto)logical incorrect correspondences. It follows that a semantic mediation service must provide for options about the resolution for these remaining issues. We suggest at least the following options: Firstly, as run-time options we consider (i) a means to fall-back to the application of naive transcriptions and continue the data exchange, or (ii) to raise a Transcription Error and refrain from data exchange. Both represent necessary service implementations that might satisfy scenarios that have either less or more stringent semantic demands. Secondly, as design-time option, it serves as a necessary tool to generate an exhaustive list of shortcomings of the transcription as a result of the current ontologies and alignment. Such list will provide the semantics engineer with sufficient information to adapt the alignment, or introduce modifications to one or both of the semantic anchorages in order to remove transcription issues that are considered vital to the business. Finally, the third option is to try to resolve the transcription issue by starting a dialogue-based semantic reconciliation process in run-time with the aim to improve upon the shortcomings of the current alignment. Despite our insistence on the need for a human to author the genuine understanding, it might still be worthwhile to have the system meticulously, consistently and methodologically negotiate logical and ontological constructs in order to find additional correspondences (Diggelen 2007).

Chapter 5

Evaluation of sIOP principles

The main (business) requirement is to achieve sIOP as quickly as possible, with as minimal effort as possible, for collaborations that had not been foreseen and consequently could not be anticipated for during design time of the (two or more) software agents. We have introduced some new principles to its support in the previous sections, and now evaluate their consequences on objectives two (loose coupling at semantic level), three (allow for semantic heterogeneity), five (support to semantic evolution) and six (scalable semantics).

5.1 Loosely coupled semantics

Consolidating sIOP demands the emergence of *loosely coupled semantics*. As analogy, consider a vehicle with its cargo. Logistics rely on external transfer services that allow the cargo to be transported over different vehicles, from a truck into an aircraft into ship into a truck again, without ruining the cargo. This requires the cargo to be firmly connected to the vehicle, but at the same time to be completely independent from any particular vehicle in order to complete the transport. “Connected but as independent as possible”, also known as loosely coupled, is a need for sIOP as well. When software agents interact they exchange meaning. Loosely coupled semantics implies that the semantic meaning (the cargo) remains as independent from the representation of semantics (the vehicle) as possible. Similarly to logistics, sIOP should rely on infrastructural services that can transcribe the semantic representation from its native form into a foreign form without invalidating the semantics that it bears. Loose coupling is known as a strong characteristic which brings many advantages. This applies to its semantic variety as well: agents can now communicate in their own native representations without the need to learn or integrate foreign representations; define semantics once and achieve sIOP many times with many different peers; development of the agents’ semantic representations can be locally isolated to fit their particular domain and application; and it enables local re-use which on its turn increases its quality.

Loose coupling in the classical sense is realised through the principles of separation of concerns and transparency. In its original reading separation of concerns turns complex functionalities into simple, atomic and complementary functional capabilities. In a semantic reading separation of concerns is not about maintaining complementary semantics; in fact, the domain of interest of the agents are required to overlap, since interoperation would be completely useless otherwise. Instead, semantic separation of concerns refer to enforce an explicit division between syntax and semantics, as discussed in Section 3.2. Additionally, it refers to keeping each other’s representations of the semantic meaning strictly separated. We described how this can be achieved in Chapter 4. The classical results of applying the principle is the emergence of unique functions that are implemented only once and used many times. In its semantic application this results in every software agent to maintain its own semantics. Collectively, all agents make that semantics become distributed all over the place which seems counterintuitive or even plain wrong. We come back to that intuition when we talk about scalability, Section 5.4, but we can already see that this is an indirect

consequence of the demand to enable sIOP with agents that were not anticipated for during software design; this, on its turn, requires that independent development of semantics shall be possible without disabling sIOP.

The classical reading on transparency separates access to the unique functions from the particular design and implementation of the functions. Remaining agnostic to *how* its function are achieved makes it possible to communicate with minimal mutual dependency. Semantic transparency remains agnostic to how semantics are *represented*, which makes it possible to communicate with minimal syntactic dependency and without prior mutual agreements on semantic representation. In its classical reading, transparency requires the introduction of standards in the components' API. Semantic transparency, contradictory, requires the total absence of any standard on representation. In stead, semantic transparency, too, requires to separate semantics from syntax. Furthermore, a need emerges for a semantic oracle that knows how to align distinct representations and to translate between them subsequently. The latter has already been discussed in Chapter 4 while the former is directly related to the human-in-the-loop as authoring authority as discussed in Chapter 3.

From the above discussion we draw the conclusion that all demands necessary to allow for semantic separation of concerns and semantic transparency are met by the sIOP principles **??**. Therefore, loosely coupled semantics should emerge between agents that comply to these principles.

5.2 Semantic heterogeneity

We have already determined that the principle to align semantics as opposed to data schemata, Design Principle 3.1, breaks the conflation of semantics and syntax, enabling to consider semantics on its own terms¹. Abstaining from a canonical model by introducing a semantic alignment between pairs of semantic meanings, Design Principle 3.2 introduces the capability for each agent to develop its semantic representation in a way that fits its local perspective optimally. By also encapsulating the particular means to provide valid semantic transcriptions only and refrain from naive transcriptions between communicating agents, Design Principle 4.1, the necessary elements to support semantic heterogeneity are present.

5.3 Semantic evolution

Consider an agent who's local semantics are in demand for a change. Assume that the agent has modified its internal pragmatic meaning to accommodate the evolved semantics. This implies a change in what we called the semantic anchorage, which acts as semantic interface to any sIOP concerns. The semantic change can be considered an altered or additional difference with the semantic anchorage from interoperating peer agents. In order to not destroy the sIOP that had been established before, it is necessary to reflect the new difference as modifications to the alignments that apply. By having performed the necessary modifications, and by relying on the semantic validity of the own ontology, the mediator is now capable of transcribing the data in accordance to the evolved semantics.

The consequence of allowing semantic heterogeneity is therefore a sufficient condition to also enable local semantic evolution and remaining semantic interoperable with existing counterparts in the process.

¹ Pun not intended

5.4 Scalable sIOP

Many definitions exist to constrain the semantics of *scalability*, academic (e.g., Neuman 1994) and popular^[http://www.linio.org/scalable.html, accessed Nov. 2019]^[https://en.wikipedia.org/wiki/Scalability, accessed Nov. 2019] alike. Our summary refers to increasing the demand that is placed on a system, and/or adding resources to a system, without experiencing loss of performance or increase in management to an extent that defeats its primary objective. We consider scalable sIOP as the capability to allow for increase in number of communicating agents as well as in the level of semantic heterogeneity without degrading the agent's communication performance or its ability to manage and control the semantic differences with its interoperating peers.

If we consider the agent's communicating performance degradation, we argue that since complexity of the connections have been traded with the complexity in the mediator (Chapter 4), the agent will only experience communication degradation when the mediator experiences transcription latencies that exceed communication parameters. In other words, the performance bottleneck is with the mediator, not with the agent. Transcription latency will result from complexity in the transcription algorithm, or from the number of transcription request that exceeds the capacity of a single mediator. In case of the latter, no particular mechanism in a mediator impedes horizontal scaling to increase the collective processing capacity to match the transcription demands. In case of the former

Regarding the ability to manage and control the semantic differences with all interoperating peers, the root cause for potential scalability issues are laid in the need to establish an alignment with each peer agent an agent engages in semantic interoperation with. This might become impractical due to our insistence on the need for a human-in-the-loop to author the alignment. We discern different solutions for different semantic topologies:

- i. Star alignments (core domain ontology, aligned to local ontologies) for relative stable and homogeneous domain semantics
 - Good: semantic governance remains controllable independent from the number of actors;
 - Bad: very big semantic monolith, hence, low agility in dynamic environments. Moreover, the more actors involved, the higher the need for semantic compromises, and the lower the overall semantic accuracy.
- ii. Mesh alignments (bilateral alignments) for very dynamic and heterogeneous (domain) semantics, or low number of peers
 - Good: quickly established bilateral sIOP; granularity-on-demand, viz. intricate where necessary, coarse-grained where possible;
 - Bad: semantic governance may become an issue to the level where it becomes impractical.
- iii. Mix-n-Match (coarse-grained star-alignment with intricate bilateral alignments as specialisations to the core domain ontology) for the 70% bulk
 - Good: controllable semantic governance; after central alignment, quickly established bilateral sIOP;
 - Bad: slightly more complicated mediation due to double alignment support.
- iv. Daisy-chained alignments (when A is aligned to B, and B is aligned to C, A and C are indirectly aligned as well)
 - Good: self-organised alignments emerge, and an instantaneous access-and-play becomes possible;
 - Bad: No guarantees can be given on the completeness of the indirect alignment. Furthermore, more intermediate alignments will increase the chance of impossible end-to-end transcriptions that would not occur with a direct alignment.

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In conclusion, scalable sIOP can be guaranteed when considering the communication performance. With respect to the ability of a single agent to manage and control the number of alignments with increasing number of interoperating agents, several options exist to support scalable sIOP although no guarantees can be given.

Chapter 6

ISO42010 viewpoint on sIOP

Chapter 7

Related work

Discuss the following papers:

1. Zhou, L., Cheatham, M., Krisnadhi, A., & Hitzler, P. (2018). A Complex Alignment Benchmark: GeoLink Dataset. In D. Vrandečić, K. Bontcheva, M. C. Suárez-Figueroa, V. Presutti, I. Celino, M. Sabou, . . . E. Simperl (Eds.), *The 17th International Semantic Web Conference, ISWC 2018* (pp. 273–288). Monterey, CA, USA: Springer Nature Switzerland AG. https://doi.org/10.1007/978-3-030-00668-6_17
2. Wilkinson, M. D., Dumontier, M., e.a. (2016). Comment: The FAIR Guiding Principles for scientific data management and stewardship. *Scientific Data*, 3, 1–9. <https://doi.org/10.1038/sdata.2016.18>
3. Pagano, P., Candela, L., Castelli, D., & Paolucci, M. (2013). Data Interoperability. In *Data Science Journal* (Vol. 12, pp. GRDI19–GRDI25). <https://doi.org/10.2481/dsj.GRDI-004>
4. Fahad, M., Moalla, N., & Bouras, A. (2012). Detection and resolution of semantic inconsistency and redundancy in an automatic ontology merging system. *Journal of Intelligent Information Systems*, 39(2), 535–557. <https://doi.org/10.1007/s10844-012-0202-y>
5. Götz, S., Beckel, C., Heer, T., & Wehrle, K. (2008). ADAPT: A semantics-oriented protocol architecture. *Lecture Notes in Computer Science (Including Subseries Lecture Notes in Artificial Intelligence and Lecture Notes in Bioinformatics)*, 5343 LNCS, 287–292. <https://doi.org/10.1007/978-3-540-92157-8-27>
6. Kühne, T. (2018). Unifying nominal and structural typing. *Software and Systems Modeling*, 1–15. <https://doi.org/10.1007/s10270-018-0660-y>
7. Renner, S. A., Scarano, J. G., & Rosenthal, A. S. (1996). Data interoperability: Standardization or Mediation. *1st IEEE Metadata Conference*, 1–8.
8. Naudet, Y., Latour, T., Guedria, W., & Chen, D. (2010). Towards a systemic formalisation of interoperability. *Computers in Industry*, 61(2), 176–185.
9. “Semantic heterogeneity is a major problem in realizing interoperability”, in: A.P. Sheth. Changing focus on interoperability in information systems: From system, syntax, structure to semantics. In R. Fegeas, M.F. Goodchild, M.J. Egenhofer and C.A. Kottman, editors, *Interoperating Geographic Information Systems*, pages 5–30. Kluwer, Norwell, MA, USA, 1999.

In (Horsch et al. 2020) the authors discuss the ongoing work on establishing a European Virtual Marketplace Framework, into which diverse platforms can be integrated. It addresses common challenges that arise when marketplace-level domain ontologies are combined with a top-level ontology like the European Materials and Modelling Ontology (EMMO) by ontology alignment. A multi-tier system of ontologies is established with the EMMO at the top and all others subsumed by it. The authors show that with such a setup the top-level ontology is crucial in the creation of the alignments between the (domain)ontologies that are subsumed by it. At the one hand this shows support to our conclusion that semantic alignments and an explicit use of ontological commitment can be considered cornerstones to achieve semantic interoperability. At the other hand their particular approach is a centralised one that does not scale well in large distributed environments due to its dependency on one single ontological commitment. Moreover, the top-level ontology, here EMMO,

necessarily conflates its function as ontological commitment with a function to construct alignments from. Although this is of great help to resolving the (automated) ontology matching problem, it creates a semantic monolith that extends to all communicating peers which, as we have seen in Chapter 1, impedes not only access-and-play sIOP but semantic scalability, evolvability, maintainability and other qualities as well.

The automatic tabular document exchange (DocEx) framework proposed by **Yang2017** divides semantic interoperability into two stages: *interpretation*, described as automatic unambiguous information understanding, and *employment*, understood as the capability to automatically operate on the information according to the interpreted semantics. The interpretation phase is dependent on a global vocabulary that “provides uniquely coded and unambiguous concepts across different domains”. Essentially, this is a clear example of the *semantic standard fallacy* described by **Janowicz:2013ui**: “The successful standardisation of protocols made us believe that we should also standardise meaning on the Web. This is a fundamental misconception”, particularly since it defies semantic heterogeneity and different but equally legitimate perspectives on the same thing. The authors remind us of three limitations of the ontology alignment approach; firstly, it cannot guarantee complete semantic interoperability for situations where terms are not aligned; secondly, creating alignments are time consuming; thirdly, ontologies are often local and their point-to-point alignments limits the semantic consistency on a more global scheme. While we do not deny any of these we consider that (i) alignments exist to facilitate interoperability, hence, lacuna are to be corrected; (ii) their creation, despite ontology matching algorithms and other automation, will take time but allow for local semantic qualities and independence that are impossible to achieve with a global semantic standard; (iii) local applications are not seeking for global interoperability but business network interoperability only.

The DocEx framework can be considered a simplified version of the openEHR framework¹ as introduced by **Beale:2001vz**, further elaborated in (Beale 2007; Beale and Heard 2007, 2008; Beale et al. 2007a, 2007b, 2008) and incorporated into CEN 13606 as a European and ISO standard. Its founding key paradigm is to model generic knowledge apart from the specific information structures, and let the former constrain the latter: knowledge is expressed as “statements which say how instances of a reference model should be constrained to form a valid business entity of some kind”. Those statements are embodied by *archetypes*. They introduce a Reference Model (RM) that can be semantically constrained by an Archetype Model (AM). The latter can be considered a meta-model or modelling language to express archetypes, i.e., a particular semantic model representing knowledge. The former is provided to each stakeholder as a unified software implementation (“the run-time platform”), providing invariant patterns of information structures. This separation makes what the RM is to the AM similar to what the JVM is to the java program. Any information item created by a user is registered as an instance of RM-specified invariant patterns. At the same time that information item is conforming to an archetype that expresses (constrains) its semantics in terms of the AM. Such approach follow Design Principles 2.1 and 2.2 but the application of a central definition of archetypes maintain a tight coupling and thus defies semantic heterogeneity and truly independent semantic representations.

The authors in (Haller et al. 2005) propose the Web Service Execution Environment (WSMX) that enables the execution of Semantic Web Services based on a Web Service Modelling Ontology (WSMO), and consider it a reference implementation for WSMO. It is meant as a means for automated discovery, composition and execution of Web Services which are based on logical inference-mechanisms, and in this way similar to our objective. Another similarity is in their acceptance of semantic heterogeneity and the need for a generic data mediator to overcome semantic differences, thereby following Design Principle 3.1. Despite these similarities, we consider two main differences with our approach. Firstly, the goals of WSMO are of another, broader, dimension than our goal to consolidate semantic interoperability and for which we have introduced details that are out of scope of WSMO. Secondly,

¹ <https://www.openehr.org/>, accessed Jan 24, 2020

Recently the international data spaces (IDS) association² forms the basis for a data marketplace as a strategic link between the creation of data in the internet of things and applying this data in machine learning (ML) and artificial intelligence (AI) algorithms. The proposed architecture is much alike the WSMX in the sense that a well-defined connector provides infrastructural services concerning security and trust, sovereignty, interoperability, ease of adoption and use, and more. In fact, if we conceive WSMX as an abstracted version of Web Services with a particular attention to semantics, IDS can be conceived as an abstracted version of the REST framework that considers data resources (spaces) the central assets in an ecosystem, with a particular attention to technology transparency when it comes to asset discovery and disclosure, identity, their secure, managed and accountable use, and interoperability. IDS considers data as assets, and provides many if not all necessary components for its managed exchange. Contrarily, we observe a clear absence of any considerations similar to those we bring forward in this paper towards the consolidation of sIOP. Still, and like WSMX's and our objectives, IDS clearly intends to put forward an architectural design with the aim to solve the concerns generically into a transparent infrastructure. In conclusions, we consider IDS an interesting complement to our approach.

² <https://www.internationaldataspaces.org/the-principles/#overview>, accessed Jan 28, 2020

Chapter 8

Discussion & future work

Complement weak AI with human brain:

- use AI where possible (computational semantics for software agent; supporting semantic reconciliation)
- use human brain where necessary (but not more): ontology engineering @ design time; alignment authoring @ pre-runtime

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