AEO: A Realism-Based Biomedical Ontology for the Representation of Adverse Events

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Abstract. The Adverse Event Ontology (AEO) is a realism-based biomedical ontology for adverse events. Currently AEO has 484 representational units annotated by means of terms including 369 AEO-specific terms and 115 terms from existing feeder-ontologies. In AEO, the term 'adverse event' is used exclusively to denote pathological bodily processes that are induced by a medical intervention. This requirement for a causal association between an adverse event and a medical intervention clearly distinguishes our approach from other approaches according to which any untoward phenomenon observed to have appeared in a mere temporal relation with some medical intervention becomes reported as an 'adverse event'. We label such phenomena as being the subject of 'adverse event hypotheses'.

Keywords: Adverse event, AEO, Adverse Event Ontology

1 Introduction

While medical interventions such as drug and nutritional product administrations, vaccinations, and use of medical devices are applied with the goal of producing positive effects, they might induce unwanted reactions which are typically described as 'adverse events' or 'side effects'. An ideal medical intervention should have high efficacy and no unwanted reactions. It is however well known that any substance (even water) might give rise to unwanted reactions, if administered at the wrong dose.

Adverse event related morbidity and mortality are a major public health issue. To better organize adverse event information, different sorts of systems such as COSTAR, MedDRA, the Common Terminology Criteria for Adverse Events (CTCAE), and the WHO's Adverse Reaction Terminology (WHO-ART) have been developed many years ago. These systems are typically constructed as controlled vocabularies, terminologies or classification systems. These older systems differ from various newer sorts of artifacts that are known as 'biomedical ontologies' and which in most cases are consensus-based controlled vocabularies of terms and relations with associated definitions, which are logically

formulated to promote automated reasoning. Bosquet *et al.*, for instance, have shown that terminological reasoning improves the performance of both data mining [1] and data access [2] in pharmacovigilance databases, and have done preliminary work toward the proposal of a categorial structure for adverse drug reactions (ADRs) [3]. However, although logically formulated definitions and axioms have the capacity to produce *valid* reasoning in deductive logic-based reasoning systems, they do not guarantee *sound* reasoning. Typical for prevailing paradigms in biomedical ontology design is concept-orientation which lacks a formal method to relate representational units to that in reality what they are representations of, and these representations are therefore more vulnerable for mistakes that lead to unsound reasoning [4]. Specifically in the context of what is called 'adverse event', there is much diversity in what is considered to be terminological practice good [5]and appropriate ontological analysis with the result that a variety of entities of totally different sorts with labels such as 'reaction', 'effect', 'event', 'problem', 'experience', 'injury', 'symptom', 'illness', 'occurrence', 'change', 'act', and even 'something', 'observation' and 'term', have been proposed as super-ordinate terms for 'adverse event' [6].

The Adverse Event Ontology (AEO), in contrast, is an ongoing realism-based effort that aims to reduce the confusion in adverse event terminology and representation using the framework offered by the OBO Foundry [7]. In this report, we present our current development of AEO, thereby distinguishing it in particular from another recent effort to generate an Adverse Event Reporting Ontology (AERO).

2 Methods

The development of AEO follows the OBO Foundry principles such as openness, collaboration, and use of a common shared syntax [7] in addition to the principles of Ontological Realism [8]. AEO is thus aligned with the Basic Formal Ontology (BFO) [9] and the Relation Ontology (RO) [10].

The AEO development method follows many guidelines provided by Ceusters *et al.* [6] in generating ontological representations of adverse events on the basis of inspecting the sorts of particulars that are involved when something that might be labeled as 'adverse event' comes into existence in some patient. These particulars are:

- (1) #1: a medical intervention (*e.g.*, vaccination, drug administration)
- (2) #2: a patient
- (3) t1: the time at which the medical intervention is given to the patient
- (4) #3: a clinically abnormal process (*e.g.*, a fever process)
- (5) t2: the time at which the clinically abnormal process happens

These elements can be modeled in the adverse event design pattern of Fig. 1 which restricts the term 'adverse event' to those pathological bodily processes that are induced by a medical intervention.¹ Both adverse event and medical intervention are subclasses of processual_entity as defined in BFO. Instances

of these two processes occur each at a specific temporal region. The corresponding causal relation between the referents of these two process terms is represented using the term *induced* by in AEO. Such a relation term is not available in RO or any other ontologies. It is noted that the OBI term process is result of (OBI_1110060) is for direct causality and not indirect causality as required here. Fig. 1 introduces the basic adverse event at the class level. In clinical cases, instance level modeling can be generated. For example, a specific vaccination process carried out on a particular patient is an *instance* of a medical intervention. To illustrate this and other important points, an example is provided in the next section. It is also to be noted that the time at which a medical intervention is given to a patient is always earlier than the time at which an adverse event occurs, i.e., t1 earlier t2 (this can be made more precise in the context of some guideline, e.g., t1 less-than-4-days-earlier-than t2).



Figure 1. Basic AEO adverse event design pattern.

An OWL version of AEO is developed using Protégé 4. OntoFox [11] was used to extract terms from external ontologies and import them into AEO. For adverse event-specific terms, new identifiers, unique to AEO, were generated.

The latest AEO, although not completely curated in terms of the principles mentioned earlier, is available for public view and download at http://sourceforge.net/projects/aeo/. AEO has been submitted to the NCBO BioPortal for public visualization and querying. It is to be noted, however, that this version is a simplification brought about by the fact that OWL, and specifically OWL-DL, does not allow representing that continuants, in contrast to occurrents, exhibit relations in which time is

¹ Although we believe that this more specific meaning of 'adverse event' as used within AEO better captures what the entities denoted by this term objectively are and that it would be beneficial that this usage would be generally adopted, the goal of this communication is not to force such usage on the community.

one of the relata, and as a consequence is therefore inadequate for representations that follow these principles.

3 Results

3.1 AEO Statistics

Currently AEO has 484 representational units, annotated by means of 369 terms with specific AEO identifiers, and 115 terms imported from existing ontologies (Table 1). This ontology development design avoids regeneration of new ontology terms that are not in the scope of the adverse event domain and supports efficient ontology reuse on the condition that the feeder ontologies are based on the same principles.

Ontology Names	Classes	Object	Total
		properties	
AEO (Adverse Event	368	1	369
Ontology)			
BFO (Basic Formal	39	0	39
Ontology)			
RO (Relation Ontology)	6	25	31
IAO (Information	2	0	2
Artifact Ontology)			
OBI (Ontology for	8	3	11
Biomedical			
Investigations)			
OGMS (Ontology for	5	0	5
General Medical			
Science)			
VO (Vaccine Ontology)	19	3	22
NCBITaxon (NCBI	5	0	5
Taxonomy)			
Total	452	32	484

Table 1. Summary of ontology terms in AEO or imported from existing ontologies.

Existing ontologies are used in two different ways in AEO: one is to import the whole ontology (here BFO and RO), and the other is to import individual terms from existing ontologies. The OntoFox method is a newly developed approach to make individual term importing easy and standardized [11], although additional steps are required to make sure that the definitions for these terms in the feederontologies correspond to the intended referents in AEO.

Fig. 2 lists key terms in AEO. Based on the *adverse event* definition, AEO required the term *medical intervention*, which currently includes four subclasses: *vaccination* (imported

from VO), drug administration, medical device usage, and nutritional product usage. Each of these medical interventions can induce corresponding adverse events, e.g., vaccine adverse event.



Instances of adverse event can have outcomes of different types, for instance a sign (*e.g.*, fever, rash) as defined in the Ontology of General Medical Science (OGMS) or another process (*e.g.*, bacterial infection). AEO uses sign- and symptom-related terms (*e.g.*, fever generation) from other existing ontologies such as the Gene Ontology (GO).

3.2 Logical Definition of 'Adverse Event' in AEO

The term 'adverse event' may mean different things in different settings [6]. In AEO, the term 'adverse event' is reserved for those pathological bodily processes that are induced by a medical intervention. As defined in OGMS, a pathological bodily process (OGMS_0000060) is a bodily process that is clinically abnormal. This definition fits well with adverse event and thus is chosen as the parent term of *adverse event* in AEO.

The word 'induced' in the AEO 'adverse event' definition indicates the existence of a causal chain. A medical intervention is a process in which several independent continuants (*e.g.*, anatomical parts of human body) participate in a variety of ways and of which other processes are parts in which these or other independent continuants participate. Some independent continuants existed already before the intervention started (e.g., cells and molecules of the patient), others are created (e.g., molecular complexes formed by bodily molecules and drugs) or modified (e.g., opening and closing of membrane channels, folding of proteins) through processes that are part of the intervention or bodily processes that come into existence in response to the creation or modification of these continuants. After the intervention, there are still bodily processes going on in which at least one of the continuants independent just mentioned participates and further independent continuants are created. The term 'induced' means that there is at least one chain of processes that starts with some process that is part of the intervention and ends with a pathological bodily process, the chain being further such that for each process within it (except the first one) there is at least one independent continuant that participated or was created in the process immediately preceding it. Note that we are not saying that there is one such independent continuant that participates in the entire chain, but rather something like this:

P1: C1,	C2,	C3
P2: C2,	С4,	C5
P3: C5,	C6,	

Mere temporal precedence is not enough because that would allow for chains of processes in which there is a pair that does not 'share' at least one continuant.

An alternative definition for 'adverse event' would be to assign it as a child term of ogms:sign, which has the textual definition of "A quality of a patient, a material entity that is part of a patient, or a processual entity that a patient participates in, any one of which is observed in a physical examination and is deemed by the clinician to be of clinical significance." Although this appears to cover different adverse events, this ogms:sign definition is too broad since all adverse events are processes. At the same time, it is too narrow because there are adverse events that are not observed. The definition of sign in OGMS clearly states "is observed in a physical examination", instead of "CAN BE observed".

4 Discussion

Several adverse event representation systems have been proposed thus far while others are under development. For example, the EU-'Patient Safety funded project through Intelligent Procedures in medication' (PSIP) aims to develop innovative tools for generating providing relevant and knowledge to healthcare professionals and patients for ADE prevention. Another relevant project funded by EU is the European Public Warning System (EU-ALERT). The French VigiTermes project is an application that automates potential adverse event detection by identifying statistical and semantic links between drugs, treatments and induced pathologies or symptoms. The EU funded ReMINE project uses an adverse event ontology to manage patient safety risks in hospital settings [12]. These projects focus on using ontologies in order to facilitate identification of drug related adverse events, combining ontologies with information extraction and also applying ontologies to hospital data.

However, as shown in [6], there is a wide variation in opinions about what would count as adverse event and many definitions fall short in various aspects. Edwards et al for instance define an adverse drug reaction as 'An appreciably harmful or unpleasant reaction, resulting from an intervention related to the use of a medicinal product, which predicts hazard from future administration and warrants prevention or specific treatment, or alteration of the dosage regimen, or withdrawal of the product' [3, 13]. The problem with this definition is that it is not specified, for instance, whom the reaction is unpleasant for (appreciation can be different for the patient, his caregivers and his relatives) and that it is prone to, so we assume, unwanted interpretations. Imagine a patient that took an oral overdose of some medicinal product and therefore is subjected to gastric suction to remove what is left in the stomach. Due to erroneous manipulation of the suction device, the patient develops a gastric bleeding. Clearly, this intervention is related to the use of a medicinal product, but it would be wrong to state, although in line with Edwards' definition, that this gastric bleeding is an adverse drug reaction.

4.1 Adverse Events versus Adverse Event Hypotheses

AEO's requirement of a causal relation between an adverse event and a medical intervention is an important and novel point which removes a lot of ambiguity. The causal requirement is indeed the major aspect in which AEO differs from that of the concept of adverse event as used in existing adverse event reporting systems such as the Vaccine Adverse Event Reporting System (VAERS) and the new Adverse Event Ontology Reporting Ontology (AERO). The latter systems do not require a causal relation to be established between a reported side effect and a medical intervention. Since what is reported as 'adverse event' in these systems may not be truly induced by a medical intervention these adverse event reporting systems contain rather references to pathological processes that happened in a specific timeframe after a medical intervention, some of which might be indeed adverse events in the AEO sense.

The data stored in such an adverse event reporting system is typically used to generate hypotheses about whether what is reported as adverse events and medical interventions are causally linked. Such a hypothesis, represented by the term *adverse event hypothesis* in AEO, becomes critical when a dramatically large amount of cases are reported following the same medical intervention. Therefore, adverse event reporting is not an end. To find potential safety problems is an ultimate goal of reporting adverse events. This is one reason why AEO aims to represent not only the adverse event hypothesis, but also the final causal association.

Finally, note that when a clinician or a patient reports an event after some medical intervention for which it is only later proven to have caused the event, this event does not 'become' an AEO adverse event: it was an instance thereof from the very beginning, although unknown as such until the proof was delivered.

It is possible to reconcile AEO and AERO in a future time. While the events included in AERO for a specific medical intervention may be larger in number than the true adverse events caused by this intervention, AEO has more depth and targets for representation of a knowledgebase of adverse events truly caused by medical interventions. How to find out the cause-and-effect relation from the reported adverse events in adverse event reporting systems is often a challenge. Rehan *et al.*, for instance, provides physicians with a guide how to assess causal relations of adverse events induced by drug administration [5]. It will surely benefit the public health and has been a critical research topic ever since such an adverse event reporting system is invented.

4.2 Comparison with Other Adverse Event Representation Systems

Here we particularly compare our AEO approach with the representation model for adverse drug reactions (ADRs) provided by Bosquet *et al.* [3].

Bosquet *et al.* generated an ADR model that contains 19 semantic categories, and the categorical structure consists of 8 semantic categories within that model. Sixteen semantic links are described in their ontology. The set of minimal constraints are 4: an ADR should be classified as a disorder, an accident, an investigation, or a syndrome. A structural disorder is defined by at least one location and one morphology. A functional disorder is defined by at least one abnormal function. There are at least one semantic link *is_related_to* and one semantic category "Drug".

The work by Bosquet *et al.* largely differs from ours. First, their ontology is based on categorial design, while AEO is based on OBO foundry ontology design. Second, their approach does not model time dependency between a drug administration and an adverse event. Third, a causal relation between a drug administration and an adverse event is not clearly specified in their system, although it can be assumed to be the case under some interpretation of 'resulting from' in their definition.

4.3 Example: Vaccine-Induced Adverse Events

In the USA, more than 10 million vaccines per year are administrated to children less than 1 year old, usually between 2 and 6 months of age. At this age, infants are at greatest risk for many medical adverse events such as high fevers, seizures, and sudden infant death syndrome.



Figure 3. Modeling of vaccination-induced fever adverse event in AEO.

Fig. 3 provides an example of how AEO represents at instance level a specific vaccineinduced fever adverse event. In this example, it is represented that Bob was vaccinated with an Afluria flu vaccination at time t1, and then had a fever at time t2, t1 and t2 being instances of temporal instant. That Bob is an instance of human being at each of these time instants and that he participates at these time instants in the respective processes is represented as well (the little triangles in Fig. 3 indicate that the participation and instance relations involving continuants are threeplace relations). Since it is notified in the vaccine instruction that fever generation is an expected adverse event and Bob was in good health before the vaccination, Bob's fever generation is considered as an adverse event induced by the vaccination process. The term fever generation is imported from the Gene Ontology (GO).

The Brighton Collaboration is a global research network that set vaccine safety research standards and does not either assume a cause-and-effect relation. According to the Brighton Collaboration, fever is defined as an elevation of body temperature above the normal [14]. Similar to other Brighton

Collaboration definitions, the fever definition itself defines a clinical entity without inference of a causal relation to a given exposure. Therefore, the time interval from immunization until onset of the event cannot be part of the definition itself [14]. However, since AEO assumes such a cause-and-effect relation, this time interval is an important study topic in the AEO representation of an influenza vaccination and a fever vaccine adverse event. Therefore, we argue that AEO and those domain-specific adverse event ontologies aligned with AEO represent a knowledgebase of adverse events caused by medical interventions, where the data stored in regular adverse event reporting systems contain many random (coincident) and false positive events that are not induced by medical interventions.

5 Conclusion

Adverse events endanger patient safety and result in considerable extra healthcare costs. A community-based ontological representation of adverse events is crucial for improving adverse event research. The advent of AEO provides an opportunity for the adverse event research community to work together towards realism-based adverse event information representation and data analysis.

To monitor and study these adverse events, many vaccine and drug adverse event reporting systems have been established to collect information about adverse events that occur after the administration of licensed vaccines. The examples of national vaccine safety surveillance programs include the VAERS in the USA and the Adverse Events Following Immunization Reporting program by the Public Health Canada. These systems contain reported data about both coincidental events and those truly caused by vaccines. In our view, an ontological representation using AEO will provide a unified and machinereadable representation of various adverse events and support more advanced adverse event data analysis.

Many efforts are required to improve AEO. For example, for better adverse event data representation and knowledgebase establishment it is important to link AEO to adverse event terminologies such as MedDRA and WHO-ART, although caution is here required because of the lack of formal rigor in these systems [15]. It will be challenging and rewarding to predict and identify which events that are temporally associated with medical interventions exhibit causal relations with these interventions using informatics approaches (e.g., statistical algorithms, and literature mining). The drug adverse events are often affected by the genetic background (e.g., SNPs) of the patient. The intricate drugpatient and drug-drug interactions are crucial to determine the final adverse event outcomes. Some adverse events happen due to crossinteractions between drug and non-drugs (e.g., grapefruit). Sometimes, an adverse event emerges when a drug is removed. It would be ideal to model these interactions in AEO with a purpose to understand the fundamental adverse event mechanisms.

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