

A Preferential DL Approach to Model the *Non bis in idem* Principle for the Legal Domain

Cleyton Rodrigues^{a,b,*}, Fred Freitas^b, Ivan Varzinczak^c and Italo Oliveira^d

^a *University of Pernambuco, Garanhuns-PE, Brazil*

E-mail: cleyton.rodrigues@upe.br

^b *Center of Informatics, Federal University of Pernambuco, Recife-PE, Brazil*

E-mail: fred@cin.ufpe.br

^c *Centre de Recherche en Informatique de Lens, Université d'Artois, Lens, France*

E-mail: varzinczak@cril.fr

^d *Faculty of Law, Federal University of Pernambuco, Recife-PE, Brazil*

E-mail: italojsoliveira1@gmail.com

Abstract. Description Logics (DLs) are a family of formalisms that emerged to balance the trade-off between expressiveness and decidability for classical monotonic logic. Often, the research developed under the umbrella of *AI & Law* has relied on full synergy with DL to support argumentation reasoning, decision systems, legal compliance checking, and axiomatization of rationales and assumptions in the legal domain. Nevertheless, in many legal scenarios, regulations are defeasible. Inferences within the legal field are not purely deductive in nature, but retractable and ampliative, since generalizations mostly hold for normal or typical cases. This is absolutely true in the criminal domain, where general criminal types are usually described in the *caput* of the norms (e.g., a robbery), and other specific types unfold from these (e.g., robbery followed by death, which is known as “*Latrocínio*” in Brazilian Criminal law). Although the classical subsumption relation may seem a correct way to model the hierarchy of laws at first glance, if no contradiction arises between the more general and more specific, what it should be pointed out that the penalty for specific crimes cancel out the penalties foreseen by the more general laws. In other words, a hierarchy of norms must not rely on classical subsumption relation; instead, a non-monotonic approach suits better in this setting. Therefore, in this paper, we show that Preferential DL, a defeasible version of Description Logic, is better suited than classical DLs for a faithful representation of the content of legal regulatory knowledge; in particular, w.r.t. the representation of the principle of Double Jeopardy (a.k.a. *Non/Ne bis in idem*). In this paper, we make the case for the application of ontologies represented in defeasible, preferential DLs for modelling laws and penalties. Our solution focuses on overrule relations to organize a set of defeasible axioms in terms of specificity criteria.

Keywords: Description Logic, Non-monotonic Reasoning, Legal Domain, *Bis in idem*, Double Jeopardy

1. Introduction

As time passes, legislative chambers around the world have gradually engineered and upgraded their legal systems, reacting to social, cultural, economic, and political changes. Legal regulations need to consistently accommodate such changes, which, in gen-

eral, do not follow a systematized protocol, leading to potential inconsistencies in the system as a whole. Legal provisions (i.e., what is stated in a legal norm) use to be directed exclusively for trained professionals, i.e., judges and attorneys’ consumption. At present, as government data is becoming openly available, the legal domain should serve a wider audience. Concerns focus on tackling methodologies to represent legal rules, case interpretations and practitioner insights

*Corresponding author. E-mail: cleyton.rodrigues@upe.br.

1 into both human (professional or laymen people) and
2 machine-readable models.

3 In the early years of the last decade, we witnessed
4 a phenomenon – the Semantic Web [1] – aimed at ex-
5 tending the traditional web with semantically anno-
6 tated resources. The underlying logical layer of the
7 new web architecture [2] allows to axiomatize enti-
8 ties, types and roles of arbitrary domains through fine-
9 grained conceptual models. Logical inference capac-
10 ity became equally possible, opening new possibilities
11 not yet envisioned, particularly for the legal domain.
12 Since then, the related literature has experienced an ex-
13 plosion of logical models, of the most diverse realms,
14 functions, and ontological engagements. Additionally,
15 a synergetic relationship between the Semantic Web
16 and Ontologies has been observed; the latter served to
17 help standardize the open technologies of the former,
18 notably the Web Ontology Language (OWL¹) [3] and
19 its underlying logical formalism, Description Logic
20 (DL) [4].

21 Although several studies look at the legal domain
22 as an opportune landscape for Semantic-Web appli-
23 cations [5–7], the representation of legal knowledge
24 faces semantic and syntactic anomalies, such as nor-
25 mative conflicts. According to Lindahl (1992) [8], a
26 normative conflict occurs in a situation where it is
27 impossible to apply two rules together. Boer et al.
28 (2005) [9] list three principal normative conflicts: *Dis-*
29 *affirmation conflict* (two different deontic mode norms
30 in opposition), *Compliance conflict* (two norms with
31 the same deontic mode remain opposed), and conflicts
32 between defeasible rules. In this paper, we are partic-
33 ularly interested in the latter type, which prevents, for
34 example, situations in which one ought to do substan-
35 tially the same thing twice in different ways, introduc-
36 ing a violation of the globally accepted legal principle
37 known as Double Jeopardy.

38 The **Double Jeopardy** principle (a.k.a. the maxim
39 *non/ne bis in idem* [10]) accommodates the widely
40 adopted legal rule (it is part of the universal law of na-
41 tions [11]) that an individual should only be punished
42 once for one offense, avoiding a disproportionate pun-
43 ishment to the offender. That is, the principle poses sit-
44 uations susceptible to defeasible reasoning.

45 For instance, suppose a peculiar situation related
46 to the Brazilian Criminal Law. A person A deliber-
47 ately steals the wallet from person B; without fur-
48 ther information, the situation typically constitutes a

1 *Theft* (art. 155 of Brazilian penal code). Additional cir-
2 cumstances, such as stealing followed by aggression
3 (serious threat or violence) would draw a more seri-
4 ous crime, *Robbery* (art. 157 from the Brazilian pe-
5 nal code), that might lead to more severe penalties. In
6 other words, the former inference remains unchanged
7 until the insertion of specific circumstances triggers a
8 more exceptional case (e.g., robbery), whose penalty
9 should replace the one from the more general case
10 and removes the previous conclusion. From the *ca-*
11 *put* defining robbery, other more severe types are built
12 within this, such as robbery followed by death (“*la-*
13 *trocínio*”, art. 157. § 3.º). Even so, courts of law tend
14 to rule out robbery classification, particularly when the
15 crime of “*latrocínio*” occurs in the attempted mode (the
16 death of the victim did not materialize, even if the per-
17 petrador sought a fatal action). If it were not so, de-
18 fense lawyers would conveniently request the reclas-
19 sification of the crime of attempted “robbery followed
20 by death” as robbery only.

21 What is going on in the aforementioned cases is that
22 we might have two legal rules that apply to the same
23 facts, but only one rule should be applied. Moreover,
24 the rules have a “hierarchy”, where we should select
25 the superior one. As a result, it should be noted that
26 classical logic, which is the basis of DL, does not fully
27 fit into all juridical nuances [12]. Therefore, classical
28 DL may lead to problems or solutions that do not prop-
29 erly represent the usual human behavior tackled in the
30 aforementioned situation. On the other hand, Preferen-
31 tial DL [13] is a non-monotonic super set of DL, which
32 seems well-suited to represent related cases and laws
33 in the legal field. Preferential DL is more adequate than
34 its classical counterpart to model and reason with par-
35 tial or incomplete information, thereby avoiding the vi-
36 olation of ordinary legal principles, such as the *non bis*
37 *in idem*.

38 This work proposes a(n onto)logical approach with
39 defeasible axioms, instead of classical ones, to model
40 the criminal types of Brazilian law that abides by the
41 Double Jeopardy principle. Such an approach prevents
42 undesirable results with respect to the judicial practice.
43 Therefore, using the Preferential DL formalism and its
44 theoretical basis [13], whenever an illicit behavior is
45 classified into more than one criminal type, including
46 different punishments, a decision support system may
47 aptly infer the appropriate sanctions foreseen by the
48 legislation. This decision naturally takes into account
49 the criteria of specificity existing among those crimes.
50 Indeed, the solution presented here addresses a propo-
51 sal for preferential (and rational) overrule relations

51 ¹<https://www.w3.org/OWL>

to hierarchically organize a set of definitional axioms within an arbitrary legal knowledge base. With this solution at hand, penalties of the more specific axioms cancel out penalties of less specific ones.

The remainder of the paper is structured as follows. First, in Section 2 we provide the required background on classical description logic and its non-monotonic counterpart, preferential DLs. In Section 3, we pose the nuances of the legal universe and compare the classical and non-monotonic solutions to model legal domains, showing exactly where classical DL fails in such cases. In Section 4, we discuss how preferential DL offers more appropriate solutions to these settings. In addition, we present a general axiomatization for criminal norms in terms of defeasible axioms, along with a notion of overrule relations. In Section 5, related work is described. We conclude the paper along with a discussion about the results achieved so far and final remarks (Section 6).

2. Theoretical Background: Description Logic and Preferential DL

2.1. Description Logics

Description Logics (DLs) [4] are a family of knowledge representation formalisms, with support for reasoning tasks. DLs are a well-behaved fragment of L2, i.e., First-Order Logic (FOL) with 2 variables (decidable and NEXPTIME-complete), quite expressive and applicable to many areas.

Several (and heterogeneous) potential applications require different levels of expressiveness and complexity of reasoning, which justifies the use of different DL fragments. Furthermore, these formalisms stand as the formal foundation of the Semantic Web endeavour.

In what follows, we describe the syntax and semantics of a reasonably expressive DL family member, *SHIQ* [14], and the associated reasoning tasks. *SHIQ* is the foundation for the well-known OWL language (Ontology Web Language). OWL constitutes the Ontology Layer in the original Semantic Web vision [1].

2.1.1. Syntax of *SHIQ*

DL sublanguages are described by the constructors they provide; each one dictates a trade-off between expressiveness and reasoning computational complexity. In particular, *SHIQ* [14] is still one of the most expressive yet decidable DLs. The syntax of *SHIQ* is

arranged in terms of three pairwise disjoint finite sets of symbols: the set of concept names (N_C), the set of role names (N_R), and the set of nominals (N_O). The grammar below defines the structure of *SHIQ* concept expressions:

$$\begin{aligned} C, D &::= N_C \mid C \sqcap C \mid C \sqcup C \mid \neg C \mid \top \mid \perp \mid \exists r.C \mid \forall r.C \mid \\ &\quad \geq n r.C \mid \leq n r.C \mid \{N_O\} \\ R, S &::= N_R \mid N_R^- \end{aligned}$$

While the first definition describes the set of *SHIQ* concept expressions, the second defines role expressions. We denote \mathcal{L} as the set of *SHIQ* concepts. Concept constructors can be classified into Boolean constructs and relationship constraints. A DL knowledge base (\mathcal{KB}) is structured in three components, $\mathcal{KB} := \langle \mathcal{A}, \mathcal{T}, \mathcal{R} \rangle$ (actually, with nominals, \mathcal{A} becomes syntactic sugar, however, we keep it for reasons of clarity [14]), where:

- \mathcal{A} is the assertional axioms, in which, we find (suppose a, b are individuals of N_O):
 - * Concept assertion, $a : C$;
 - * Role assertion, $(a, b) : R$;
 - * Individual equality and inequality, $a \approx b$, $a \not\approx b$, respectively;
- \mathcal{T} is the terminological axioms, in which, we find:
 - * Concept inclusion, $C \sqsubseteq D$;
 - * Concept equivalence, $C \equiv D$.
- \mathcal{R} is the relational axioms, in which, we find:
 - * Role inclusion, $R \sqsubseteq S$;
 - * Role equivalence, $R \equiv S$;
 - * Complex role inclusion, $R_1 \circ R_2 \sqsubseteq R$;
 - * Inverse roles, $\text{Inv}(R) = R^-$;
 - * Transitive roles, $\text{Trans}(R) = {}^{(+)}R$.

2.1.2. Semantics of *SHIQ*

DL semantics assumes the Open-World Assumption (OWA) [15]. According to the OWA premise, given a state-of-affairs, the absence of information is treated as something unknown, unlike the Closed-World assumption [15, 16], where it is interpreted as negative information. OWA is the most reliable assumption for the representation of knowledge, whenever it is not possible to guarantee that all information has been provided, or have not been made available yet.

Baader et al.(2010) [4] outlines the semantics of Classical DL in terms of first-order interpretations.

Definition 2.1. Classical DL Interpretation. An Interpretation is a tuple $\mathcal{I} = \langle \Delta^{\mathcal{I}}, \cdot^{\mathcal{I}} \rangle$, where $\Delta^{\mathcal{I}}$ represents the non-empty set known as the domain of \mathcal{I} , while $\cdot^{\mathcal{I}}$ is a function that maps concepts to subsets of $\Delta^{\mathcal{I}}$, relations to subsets of $\Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$ and instances to elements of $\Delta^{\mathcal{I}}$.

In addition, in this same context, it is necessary to describe the definition of Model.

Definition 2.2. Model. An interpretation \mathcal{I} is a model of a concept C if $C^{\mathcal{I}} \neq \emptyset$. In addition, an interpretation \mathcal{I} is a model of a $\mathcal{KB} := \langle \mathcal{A}, \mathcal{T} \rangle$ if \mathcal{I} is a model of \mathcal{T} and a model of \mathcal{A} .

Table 1 highlights the self-explanatory semantics for DL grammar constructors and TBox/ABox/RBox axioms. For a short and didactic introduction to DL and its applications, as well as semantics for remaining constructors, check additional reference [4, 14].

2.1.3. Reasoning over Concept Expressions and Knowledge Bases

From the definition of Interpretation in DL, we can state some basic tasks of reasoning over concept expressions such as:

- **Concept Satisfiability:** Given a concept C , C is satisfiable iff it admits a model.
- **Concept Subsumption:** An interpretation \mathcal{I} is a model of a general concept subsumption $C \sqsubseteq D$ if $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$.

In addition, DLs have associated a number of reasoning tasks that are important from the standpoint of knowledge representation and reasoning. Among them are:

- **Knowledge Base Satisfiability:** Given a knowledge base \mathcal{KB} , and two concepts C and D , \mathcal{KB} is satisfiable if it admits a model, that is, an Interpretation \mathcal{I} , which for every axiom $C \sqsubseteq D$ in \mathcal{KB} , $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$.
- **Concept Satisfiability w.r.t. Knowledge Base** ($\mathcal{KB} \models C \equiv \perp$): Given a knowledge base \mathcal{KB} , and a concept C , C is satisfiable w.r.t. \mathcal{KB} if there is an interpretation \mathcal{I} , which is a model for \mathcal{KB} , and further a model for C , that is, $C^{\mathcal{I}} \neq \emptyset$.
- **Logical Implication** ($\mathcal{KB} \models C \sqsubseteq D$): Given a knowledge base \mathcal{KB} , and two concepts C and D , D subsumes C , if for all models \mathcal{I} of \mathcal{KB} , $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$.

The reasoning tasks described hitherto, however, address only the terminological portion (\mathcal{T}) of a knowl-

Table 1
DL SHIQ Syntax \times Semantics

| | Syntax | Semantics |
|-------------------------|-------------------------------|---|
| Individuals | | |
| Individual Name | a | $a^{\mathcal{I}}$ |
| Roles | | |
| Atomic Role | R | $R^{\mathcal{I}}$ |
| Concepts | | |
| Atomic Concept | A | $A^{\mathcal{I}}$ |
| Intersection | $C \sqcap D$ | $C^{\mathcal{I}} \cap D^{\mathcal{I}}$ |
| Union | $C \sqcup D$ | $C^{\mathcal{I}} \cup D^{\mathcal{I}}$ |
| Complement | $\neg C$ | $\Delta^{\mathcal{I}} \setminus C^{\mathcal{I}}$ |
| Top Concept | \top | $\Delta^{\mathcal{I}}$ |
| Bottom Concept | \perp | \emptyset |
| Existential Restriction | $\exists R.C$ | $\{x \in \Delta^{\mathcal{I}} \mid \exists y, (x, y) \in R^{\mathcal{I}}, y \in C^{\mathcal{I}}\}$ |
| Universal Restriction | $\forall R.C$ | $\{x \in \Delta^{\mathcal{I}} \mid \forall y, (x, y) \in R^{\mathcal{I}} \Rightarrow y \in C^{\mathcal{I}}\}$ |
| TBox | | |
| Subsumption | $C \sqsubseteq D$ | $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$ |
| Equivalence | $C \equiv D$ | $C^{\mathcal{I}} = D^{\mathcal{I}}$ |
| RBox | | |
| Role Inclusion | $R \sqsubseteq S$ | $R^{\mathcal{I}} \subseteq S^{\mathcal{I}}$ |
| Role Equivalence | $R \equiv S$ | $R^{\mathcal{I}} = S^{\mathcal{I}}$ |
| Complex Role Inclusion | $R_1 \circ R_2 \sqsubseteq S$ | $R_1^{\mathcal{I}} \circ R_2^{\mathcal{I}} \subseteq S^{\mathcal{I}}$ |
| Transitive Role | $(+)R$ | $R^{\mathcal{I}} = (R^{\mathcal{I}})^+$ |
| Inverse Role | R^- | $\{(y, x) \mid (x, y) \in R^{\mathcal{I}}\}$ |
| ABox | | |
| Concept Assertion | $a : C$ | $a^{\mathcal{I}} \in C^{\mathcal{I}}$ |
| Role Assertion | $(a, b) : R$ | $\langle a^{\mathcal{I}}, b^{\mathcal{I}} \rangle \in R^{\mathcal{I}}$ |

edge base. Therefore, given individual names a, b in N_O^2 , for the assertional component, we have:

- **Concept Instantiation** ($\mathcal{KB} \models a : C$): Given a knowledge base \mathcal{KB} , and an individual a , a is an instance of concept C w.r.t. \mathcal{KB} if $a^{\mathcal{I}} \in C^{\mathcal{I}}$ holds for all models \mathcal{I} of \mathcal{KB} ;
- **Role Name Instantiation** ($\mathcal{KB} \models (a, b) : R$): Given a knowledge base \mathcal{KB} , and some individuals a, b , the pair of individuals (a, b) is an instance of role name R w.r.t. \mathcal{KB} if $\langle a^{\mathcal{I}}, b^{\mathcal{I}} \rangle \in R^{\mathcal{I}}$ holds for all models \mathcal{I} of \mathcal{KB} ;

Horrocks (2005)[17] argues that those decidable key inference problems – concept satisfiability and logical subsumption – were the key to DL becoming the ba-

²With nominals, we have: $a : C$ equivalent to $\{a\} \sqsubseteq C$, and $(a, b) : R$ equivalent to $\{a\} \sqsubseteq \exists R.\{b\}$

1 sic formalism of many applications. Among others, we
 2 highlight information retrieval to bio-medical [18] and
 3 legal [19] documents, practical application for internet
 4 of things (iot) [20], databases and schema design [21],
 5 and query answering [22]. In particular, DL stood out
 6 as the formalism of the ontological language for the
 7 semantic-web applications. Not only that, but the possi-
 8 bility of providing efficient decision-making systems
 9 also weighed on the choice of DL, provided by opti-
 10 mized implementations of tableaux procedures. Un-
 11 fortunately, exceptions and non-monotonic reasoning,
 12 such as the kind that occurs so often in legal reasoning,
 13 cannot be well represented in description logic. Actu-
 14 ally, legal regulations are usually defeasible [23].

15 2.2. Preferential DLs

16 In order to leverage the DL semantics bearing non-
 17 monotonic reasoning, Britz et al. (2011) [13] suggest
 18 a DL extension addressing a defeasible subsumption
 19 constructor \sqsubset to axiomatize exceptions for the typi-
 20 cal objects, known as Preferential Description Logic.
 21 Consider, for example, the universe of flying birds and
 22 penguins. Instead of stating, for example, that birds
 23 fly ($Birds \sqsubseteq Fly$), through a preferential DL syntax,
 24 we state that *typical* birds (the most base cases) fly
 25 ($Birds \sqsubset Fly$). In order to cope with exceptionality,
 26 \sqsubset should not be monotonic. Under such conditions, it
 27 is common to speak of a *defeasible knowledge base*,
 28 that is, one with a set of classical and a set of defea-
 29 sible axioms. In the following, we highlight the pref-
 30 erential and rational subsumption relation defined by
 31 Britz et al. (2011) [13], and derived from Kraus et al.
 32 (1990) [24] (also known as KLM theory).

33 **Definition 2.3. (Preferential Subsumption [13]).** A
 34 subsumption relation $\sqsubset \subseteq \mathcal{L} \times \mathcal{L}$ is a preferential sub-
 35 sumption relation iff it satisfies the following proper-
 36 ties, where Ref stands for Reflexivity, LLE for Left Log-
 37 ical Equivalence, RW stands for Right Weakening, and
 38 CM stands for Cautious Monotonicity:

$$\begin{array}{ll}
 (Ref) C \sqsubset C & (LLE) \frac{C \equiv D, C \sqsubset E}{D \sqsubset E} \\
 (And) \frac{C \sqsubset D, C \sqsubset E}{C \sqsubset D \sqcap E} & (Or) \frac{C \sqsubset E, D \sqsubset E}{C \sqcup D \sqsubset E} \\
 (RW) \frac{C \sqsubset D, D \sqsubseteq E}{C \sqsubset E} & (CM) \frac{C \sqsubset D, C \sqsubset E}{C \sqcap D \sqsubset E}
 \end{array}$$

1 **Definition 2.4. (Rational Subsumption [13]).** A sub-
 2 sumption relation $\sqsubset \subseteq \mathcal{L} \times \mathcal{L}$ is a rational subsumption
 3 relation iff in addition to being a preferential subsump-
 4 tion relation, it also satisfies the Rational Monotonicity
 5 property (RM).

$$(RM) \frac{C \sqsubset D, C \not\sqsubset \neg E}{C \sqcap E \sqsubset D}$$

6 Extending DL with preferential/rational subsump-
 7 tion relations favors non-monotonic reasoning capabil-
 8 ities. For defeasible subsumption inferences, an intu-
 9 itive formal semantics is further proposed by Britz et
 10 al. (2011) [13]. In effect, the semantics of preferential
 11 DLs is organized in terms of strictly partially-ordered
 12 structures, $\mathcal{P} := \langle \Delta^{\mathcal{P}}, \cdot^{\mathcal{P}}, \prec^{\mathcal{P}} \rangle$, where:

- 13 – $\langle \Delta^{\mathcal{P}}, \cdot^{\mathcal{P}} \rangle$ is an ordinary DL interpretation;
- 14 – $\prec^{\mathcal{P}}$ is an irreflexive, anti-symmetric and transitive
 15 partial order on $\Delta^{\mathcal{P}}$;
- 16 – and $\prec^{\mathcal{P}}$ is, additionally, smooth: for every $C \in \mathcal{L}$,
 17 if $C^{\mathcal{P}} \neq \emptyset$, then $\min_{\prec^{\mathcal{P}}}(C^{\mathcal{P}}) \neq \emptyset$, where $\min_{\prec^{\mathcal{P}}}$
 18 denotes the minimal elements in $C^{\mathcal{P}}$.

19 The partial order organizes the elements of an arbi-
 20 trary domain $\Delta^{\mathcal{P}}$ in a stratification corresponding to
 21 several “levels of typicality”, decreasing from bottom
 22 to top; that is, the lower-level objects must be the most
 23 normal. The set of allowed stratifications is informed
 24 by the knowledge specified in the knowledge base in
 25 the form of defeasible concept inclusions.³ Given a
 26 preferential DL interpretation \mathcal{P} and a defeasible sub-
 27 sumption statement $C \sqsubset D$, the semantics of the latter
 28 is given by:

$$\mathcal{P} \models C \sqsubset D \text{ iff } \min_{\prec^{\mathcal{P}}}(C^{\mathcal{P}}) \subseteq D^{\mathcal{P}}.$$

29 ³We can make an analogy with what happens when specifying
 30 propositional knowledge bases. Under an empty knowledge base, all
 31 possible propositional valuations are allowed; then if one adds $p \wedge q$
 32 to the base, only the valuations satisfying $p \wedge q$ remain; and so on
 33 as more sentences are added. Here, in the preferential case, at the
 34 starting point we have all possible preferential interpretations, with
 35 all possible orderings on the domain. After adding the statements
 36 $C \sqsubset D$ and $C \sqcap C' \sqsubset \neg D$ to the knowledge base, only those preferential
 37 interpretations which stratify the extension of C into the more typical
 38 ones (those that are not C' and that fall under D) and the less typical
 39 ones (those that are also C' and are not D) remain. The process then
 40 continues as more knowledge is added to the ontology and in the end
 41 only the orderings that are compatible with the knowledge base are
 42 the allowed ones.

From the rational and preferential subsumption definitions, we can proceed with the associated entailment definitions [13].

Definition 2.5. (Preferential Entailment). A defeasible subsumption statement $C \sqsubseteq D$ is preferentially entailed by a given defeasible knowledge base \mathcal{KB} iff $C \sqsubseteq D$ is a statement of the preferential closure of \mathcal{KB} , i.e., it is a derivation from \mathcal{KB} using the rules of Preferential Subsumption.

In such context, nevertheless, given a preferential interpretation \mathcal{P} , we can not conclude (in general) $\mathcal{P} \Vdash C \sqcap E \sqsubseteq D$ from $\mathcal{P} \Vdash C \sqsubseteq D$. Britz et al. (2011) [13] mention that “preferential entailment is thus too weak”. Inferences drawn in the legal context should be retractable and ampliative [23]. In other words, plausible (though provisional) conclusions need to be inferred in the absence of conflicting information. According to it, such conclusions should be removed in the light of new conflicting information. Therefore, we leave aside the monotonic preferential entailment, and adopt here the non-monotonic rational entailment [25].

Rational entailment solves the unwanted property of monotonicity induced by preferential interpretations through a preference ordering on models [25]. The motivation for this is to realize that some models are more important than other.

Definition 2.6. (Rational Entailment). A defeasible subsumption statement $C \sqsubseteq D$ is rationally entailed by a given defeasible knowledge base \mathcal{KB} iff $C \sqsubseteq D$ is a statement of the rational closure of \mathcal{KB} , i.e., it is a derivation from the minimal model (the model in which objects are as typical as possible) of \mathcal{KB} .

In the following example, we address these particularities

Example 2.1. Suppose a knowledge base \mathcal{KB}_{bf} conceptualizing the domain of birds. We have three concepts: Bird, Penguin, and Fly (representing entities with flying capabilities, like birds, mammals, airplanes). For \mathcal{KB}_{bf} , terminological axioms define classical and defeasible subsumption statements, while the Abox asserts four famous cartoon instances, namely happyFeet, skipper, woody and bartok.

$$\mathcal{P} \Vdash \mathcal{KB}_{bf} : \left\{ \begin{array}{l} \mathcal{T} = \left\{ \begin{array}{l} \text{Bird} \sqsubseteq \text{Fly}, \\ \text{Penguin} \sqsubseteq \text{Bird}, \\ \text{Penguin} \sqsubseteq \neg \text{Fly} \end{array} \right\} \\ \mathcal{A} = \left\{ \begin{array}{l} \text{Penguin}(\text{happyFeet}), \\ \text{Bird}(\text{skipper}), \\ \text{Bird}(\text{woody}), \\ \text{Fly}(\text{bartok}) \end{array} \right\} \end{array} \right\}$$

$$\mathcal{KB}_{bf} = \mathcal{T} \cup \mathcal{A} \models \left\{ \begin{array}{l} \text{Fly}(\text{skipper}), \\ \text{Fly}(\text{woody}) \end{array} \right\}$$

By axiomatizing the knowledge of flying objects, happyFeet was asserted as a Penguin. Nothing very specific was said about skipper and woody except that they are birds. Under the principle of *presumption of typicality* [26], it is plausible to infer that these individuals do fly.⁴ It is assumed, therefore, that they are typical instances of birds. Nevertheless, taking into account that we are dealing with partially observable environments, new acquired information may require the withdrawal of some previously inferred consequences. This is what happens upon discovering that skipper is a penguin. In these terms, Fly(skipper) needs to be retracted from the set of inferences produced.⁵

Figure 1 illustrates a possible interpretation (\mathcal{P}) satisfying the new knowledge base \mathcal{KB}'_{bf} . \mathcal{P} is defined in terms of $\langle \Delta^{\mathcal{P}}, \cdot^{\mathcal{P}}, \prec^{\mathcal{P}} \rangle$, where:

- $\Delta^{\mathcal{P}} = \{\text{happyFeet}, \text{skipper}, \text{woody}, \text{bartok}\};$
- * $\text{Bird}^{\mathcal{P}} = \{\text{happyFeet}, \text{skipper}, \text{woody}\};$
- * $\text{Penguin}^{\mathcal{P}} = \{\text{happyFeet}, \text{skipper}\};$
- * $\text{Fly}^{\mathcal{P}} = \{\text{woody}, \text{bartok}\};$
- $\prec^{\mathcal{P}} = \{(\text{woody}, \text{happyFeet}), (\text{woody}, \text{skipper}), (\text{bartok}, \text{happyFeet}), (\text{bartok}, \text{skipper})\}.$

$$\mathcal{KB}'_{bf} = \left\{ \begin{array}{l} \mathcal{T} = \left\{ \begin{array}{l} \text{Bird} \sqsubseteq \text{Fly}, \\ \text{Penguin} \sqsubseteq \text{Bird}, \\ \text{Penguin} \sqsubseteq \neg \text{Fly} \end{array} \right\} \\ \mathcal{A} = \left\{ \begin{array}{l} \text{Penguin}(\text{happyFeet}), \text{Penguin}(\text{skipper}), \\ \text{Bird}(\text{skipper}), \text{Bird}(\text{woody}), \text{Fly}(\text{bartok}) \end{array} \right\} \end{array} \right\}$$

In the class of birds, woody is the minimal object with respect to penguins happyFeet and skipper. Even outside the set of birds, so does bartok in relation to the penguins. It is assumed that the most typical elements of Bird and Fly indeed fly. Remaining objects (arranged at the highest level) are exceptions to these, i.e., cases

⁴The principle of presumption of typicality is one of the fundamental principles of rationality in non-monotonic reasoning. It is at the heart of a form of ampliative reasoning and states that we shall always assume that we are dealing with the most typical possible situation compatible with the information at our disposal.

⁵It is worth pointing out here that rational entailment check (i.e., the computation of rational closure) has to be recalculated. Our use of retraction here refers to what happens at the level of all logical consequences of a base: from the user’s perspective, it is as if a piece of knowledge had been retracted.

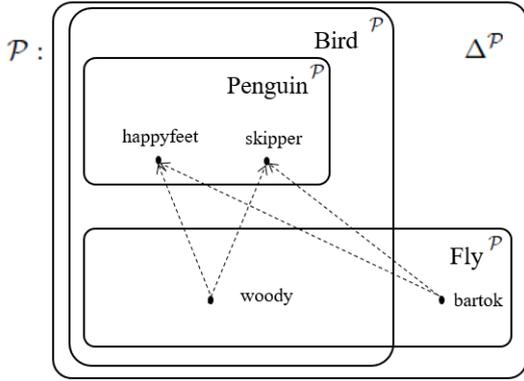


Fig. 1. A preferential DL interpretation for the bird example

where generalizations do not apply. If, instead, we claim that Fly subsumes Bird in the traditional sense, that is, $\text{Bird} \sqsubseteq \text{Fly}$, the base would be inconsistent: $\mathcal{KB}'_{bf} \models \top \sqsubseteq \perp$, since $\mathcal{KB}'_{bf} \models \text{Penguin} \sqsubseteq \perp$ (from $\mathcal{KB}'_{bf} \models \text{Penguin} \sqsubseteq \text{Fly}$ and $\mathcal{KB}'_{bf} \models \text{Penguin} \sqsubseteq \neg\text{Fly}$) and $\text{Penguin}(\text{happyfeet}) \in \mathcal{KB}'_{bf}$.

We end the present section with a discussion on the advantages of preferential description logics over their classical counterparts and on some philosophical and technical reasons for employing them rather than their classical cousins.

One may object that exceptions and overruling of properties can be formalized in classical DLs by fully specifying in the left-hand side of subsumption statements all the conditions for a rule to hold. For example, one could state $\text{Bird} \sqcap \neg\text{Penguin} \sqsubseteq \text{Fly}$ and $\text{Penguin} \sqsubseteq \neg\text{Fly}$. As it turns out, from a knowledge representation and reasoning perspective, this is not suitable. The reasons are as follows:

- The first reason relates to readability of the statements in the knowledge base. The more conditions one explicitly adds to the left-hand side of subsumption statements, the longer and harder to understand such statements become.
- The second reason is that every discovery of a new exception will require remodelling of the whole knowledge base, which is a computationally expensive task, even when performed offline.
- The third reason relates to the way we humans actually reason under incomplete information. When performing reasoning, we humans do not explicitly think of all special cases that would prevent a conclusion from being drawn. Instead, we

base our reasoning only on the information at our disposal and provisionally jump to the conclusion. It is only upon facing new information that we accommodate it with the previous knowledge we had and, usually, we do it in a non-disruptive way.

- The fourth reason is that classical DL entailment is non-ampliative and non-defeasible, features that are sought for when reasoning with incomplete information.

We contend the preferential approach to default reasoning addresses the four points above in a satisfactory manner:

- First, our defeasible concept inclusions are shorter than the classical ones explicitly specifying the exceptions and therefore are user-friendly.
- Second, since the statements in the knowledge base already cope with exceptions, the discovery of a new one is automatically dealt with. Remodelling is only required if really desired, e.g., if one wants to refine the levels of typicality, and even in this case changes to the knowledge base are localized.
- Third, the reading of a defeasible statement of the form $C \sqsubset D$, i.e., “usually, Cs are Ds” is in line with human actual reasoning.
- Finally, rational closure is a well-accepted form of ampliative and defeasible entailment relation.

3. Nuances of the Legal Domain

Throughout this section, we argue on the limitations of classical DL to tackle a kind of normative conflict inherent to the legal domain, in particular a type often observed within the normative *corpus* of criminal law.

3.1. Normative Conflicts and the Double Jeopardy Principle

According to Lindahl (1992) [8], a normative conflict occurs in a situation where it is impossible to apply two rules together. Boer et al. (2005) [9] list the three main types of normative conflicts: Disaffirmation conflict, Compliance conflict, and conflicts between rules that are defeasible. Disaffirmation refers to situations where an event is qualified by two different deontic mode norms (permissive and prohibitive, for example). The conflict known as compliance refers to situations where norms with the same deontic qualifica-

tion still remain opposed. Finally, the last conflict deals with normative rules that are defeasible, to prevent, for example, situations in which one ought to do substantially the same thing twice in different ways. The latter is the type of conflict addressed in this manuscript.

Usually, legal systems explicitly define a preference among norms, which are hierarchically ordered. From more general and comprehensive norms, such as the sovereign constitution, other laws, decrees and regulations are derived. In legal theory [27], usually when two laws govern the same facts, a law governing a specific subject matter overrides a law governing only general matters. As a result, legal reasoning is defeasible by nature. Inferred conclusions can potentially be withdrawn, with the inclusion of new information that triggers more specific situations. Celano (2012) [28] argues that, although seen as conditional rules of type IF <CONDITIONS> THEN <CONSEQUENT>, the antecedent part of a norm must be satisfied only under normal conditions. This is because the application of legal rules are open to be canceled by specific circumstances (which are either hidden within the lines of normative texts, or are clearly explained by them). In general, however, those situations are difficult to specify prior to the application (*ex ante*) of the law for particular cases [23]. Worse yet, it is humanly impossible to list all circumstances and sufficient conditions for practical applications [29]. Figures 2 and 3 depict a situation represented in classical and preferential DL, along with inferences of (in)consistency.

Within the Penal Code, ordinary crimes are built from more typical situations (written down in the *caput* of articles), and other criminal types (more serious or soft) unfold from this. In the criminal realm, for example, one cannot condemn the behavior of an agent such as theft (art. 155) and robbery (art. 157) at the same time. In Figure 2, using classical DL, events of stealing match a theft; these events accompanied by aggression, characterize a robbery. Besides, each crime defines its unique sanction.⁶ On these conditions, it is inferred that Robbery is a Theft. However, from this new knowledge, with those already laid down at the knowledge base, two penalties are imposed on the crime of robbery, leading to an inconsistency, since only one penalty shall be prescribed to a crime. Although not explicitly illustrated, we assume that the inconsistency is reached by TBox together with ABox; this latter is represented by *e1* and *e2* individuals. In this case, *e2*

should be classified as Theft and Robbery, and therefore, relating the instance to both penalties.

Nonetheless, in Figure 3, there is no longer any inconsistency in the imposition of penalties, due to the use of Preferential DL. According to the Preferential DL semantics, individuals can be ordered by typicality inside a class: therefore, only the most typical stealing events are classified as theft. Robbery events would be at a level above. Note that this is due to the use of the preferential subsumption relation, contrary to the classical, monotonic DL semantics. Such an example shows why preferential approaches seem to fit more naturally to law subsumption in general, and helps building law hierarchies where, despite specific laws are viewed as special cases of more general laws, the penalty conflict is dealt with the usual way law practitioners do. The axioms will be explained in the following sections.

It is still worth mentioning that our objective is to help in classifying the different types of acts, with the consequent imposition of penalties. Certain idiosyncrasies resulting from classical reasoning are thus avoided.

3.2. A Case Study about Crimes Against Property

The Brazilian Criminal Law is referred to as the set of legal rules that define criminal offenses, establishing punishment and security measures. These norms establish definitions about the crimes, their types, and criminal penalties. Besides, they are structured by the Brazilian Penal Code⁷ (Decree-Law 2,848 dated 1940), having been extensively amended by Law 7209/1984. Notably, crimes against property correspond to the protected legal interest in the crimes set out in Articles 155-180. Beyond the economic value, for criminal purposes, the value of assets covers also the moral value of goods, as a letter, a stone, or any material object that has affection value to the owner, although it may not have an exchange value. Crimes against property include, but are not limited to (the original norms in native Portuguese are available in Appendix A):

- **Theft:** *To take away a chattel⁸, for himself or others.* (Art. 155);
- **Robbery:** *To take a chattel, for himself or others, by serious threat or violence [...] (Art. 157);*

⁶Crime \sqsubseteq 1hasPenalty.Penalty

⁷http://www.planalto.gov.br/ccivil_03/decreto-lei/lei2848.htm

⁸An item of personal property that is movable.

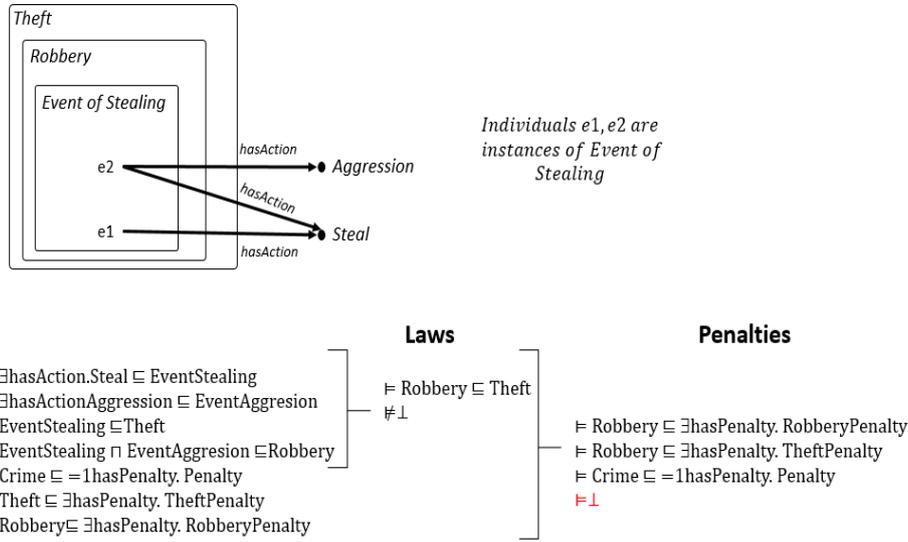


Fig. 2. Undesired Inferences in Description Logic

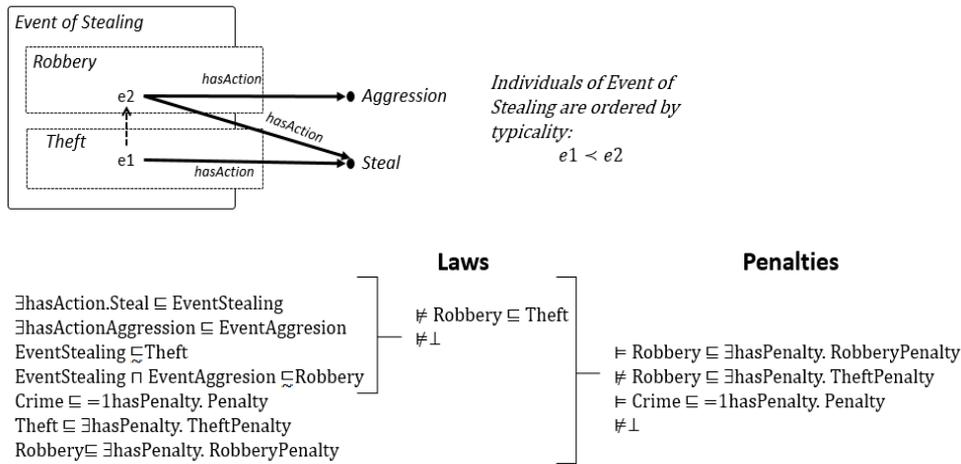


Fig. 3. Desired Inferences in Preferential Description Logic

– **Robbery followed by Death:** *If the violence results in [...] death, the jail time is twenty to thirty years.* (Art. 157, § 3º);

It is important to note that some concepts were slightly modified with respect to the English language to suit the specificities and types defined in the Brazilian criminal law. Drafting of the penal code has some pitfalls, which, if not properly addressed, lead us to wrong conclusions. As stated by the law itself, any layperson believes that Robbery is a sub type of Theft. Nevertheless, although the crime of Robbery is more

specific, since it requires an aggressive action beyond the very action of stealing, it is legally wrong to state that this criminal type is a sub type of Theft. In other words, both rules cannot apply simultaneously in the daily judicial practice. Besides, each crime defines its own imprisonment time. The Double Jeopardy principle prevents double penalty. Therefore, rather than talking about sub types of crimes, it is more consistent with judicial practice to speak of typical and specialization cases. Example 3.1 presents a situation in which an instance of a material good was violated.

Example 3.1. Bill is a graduate student and attends classes at night shift. One day, returning home, Bill was approached by a biker who ordered Bill to give him his wallet. The biker was armed and said that if Bill told anyone on the wallet, he would kill him.

For legibility, criminal domain concepts are structured in terms of Unified Modelling Language (UML) classes [30]. Figure 4 pictures a class diagram for a partial view of the General Theory of Crime, as stated in the Brazilian Penal Code.⁹ Herein, we have specialized the top ontology known as Unified Foundational Ontology (UFO), which deals with the duality of established philosophical theory between Endurant and Perdurant entities [32, 33]. Endurants are perennial particulars that exist as time goes by (like agents and objects). Situations are special sorts of endurants as well, a portion of reality recognized as a whole, a state of affairs. Conversely, Perdurants happen in time, that is, they are framed by intervals of time. Substantial instantiates a Kind category, i.e., a rigid type which provides the essential property (that is, any substantial instantiates imperiously in every situation). UFO provides further non-rigid categories, such as Role, which is instantiated only in certain contexts depending on a given relation, whereas Phase is a type instantiated in a context dependent on an intrinsic property.

Unlike (non-agentive) objects, herein we consider Agent as any entity that can initiate an Action. CrimeAgent is a top category for any Agent within the criminal realm. It specializes further in two disjoint roles, agents who perform a criminal act (ActiveAgent) and agents who have had some property protected by the State (honor, life, valuable object, and so on) violated (PassiveAgent). Situations are changed through a CriminalAct, that is, an Action, carried out by the ActiveAgent. CriminalAct violates a CrimeObject, which is in turn, associated to a PassiveAgent.

A ForbiddenSituation refers to a specific state-of-affairs prohibited by an IncriminatingRule. These situations are composed by violated crime objects, and injured passive agents. The kernel of a Crime is the criminal action taken. As secondary elements, a legal act may have additional (aggravating/mitigating) circumstances, modifying the crime type. In order not to overload the figure, we avoid illustrating details of norms and sanctions. Figure 4 pictures the axiomatization for the theory of crime.

⁹In [31], we have provided a detailed axiomatization with respect to the Theory of Crime.

Figure 5 illustrates the instantiation for Example 3.1. behaviorBob is a complex event involving two actions, carried out by the same agent. Each violates a legal asset associated with the passive agent. The resulting situation is forbidden by the article that typifies the robbery, but also by the theft article (when considering only part of the final situation).

The TBox \mathcal{T} models both types of crime. Conversely, the ABox \mathcal{A} models Bob's behavior.

| | | | |
|-----------------|------------------------|--|---|
| $\mathcal{T} =$ | Crime | \equiv Event \sqcap \exists realizedThrough.CriminalAct, | } |
| | CriminalAct | \sqsubseteq Action, | |
| | CriminalAct | \sqsubseteq \exists performanceOf.ActiveAgent \sqcap \exists violation.CrimeObject, | |
| | CrimeObject | \sqsubseteq \exists isAssociatedTo.PassiveAgent, | |
| | ForbiddenSituation | \sqsubseteq \exists hasCriminalAct.CriminalAct, | |
| | IncriminatingRule | \sqsubseteq \exists disallows.ForbiddenSituation \sqcap \forall disallows.ForbiddenSituation, | |
| | hasCriminalAct | \equiv hasDisallowedSituation $\bar{_}$, | |
| | hasDisallowedSituation | \sqsubseteq hasPosSituation, | |
| | Aggression | \sqsubseteq CriminalAct, | |
| | Steal | \sqsubseteq CriminalAct, | |
| | Event \sqcap | \exists realizedThrough.Steal \sqsubseteq Theft, | |
| | Event \sqcap | \exists realizedThrough.Steal \sqcap \exists realizedThrough.Aggression \sqsubseteq Robbery, | |
| | Situation | \sqsubseteq \exists hasEndurant.Endurant, | |
| | Circumstance | \equiv Aggravating \sqcup Mitigating, | |
| | Aggravating | \sqsubseteq \neg Mitigating, | |

| | | |
|---|---|---|
| $\mathcal{A} =$ | Event(behaviorBob), Aggression(threatenLife), | } |
| | Steal(takeTheWallet), Situation(billThreatenedAndStolen), | |
| | CrimeObject(billPsychological), | |
| | CrimeObject(billWallet), Agent(bob), PassiveAgent(bill), | |
| | IncriminatingRule(art157Law2848), | |
| | IncriminatingRule(art155Law2848), | |
| | performanceOf(takeTheWallet, bob), | |
| | performanceOf(threatenLife, bob), | |
| | realizedThrough(behaviorBob, takeTheWallet), | |
| | realizedThrough(behaviorBob, threatenLife), | |
| | hasCriminalAct(billThreatenedAndStolen, takeTheWallet), | |
| | hasCriminalAct(billThreatenedAndStolen, threatenLife), | |
| | violation(threatenLife, billPsychological), | |
| | violation(takeTheWallet, billWallet), | |
| | isAssociatedTo(billPsychological, bill), | |
| | isAssociatedTo(billWallet, bill), | |
| hasDisallowedSituation(behaviorBob, billThreatenedAndStolen), | | |
| disallows(art157Law2848, billThreatenedAndStolen), | | |
| disallows(art155Law2848, billThreatenedAndStolen) | | |

For simplicity, in \mathcal{T} we do not list the direct subsumption relationships, modelled as inheritance be-

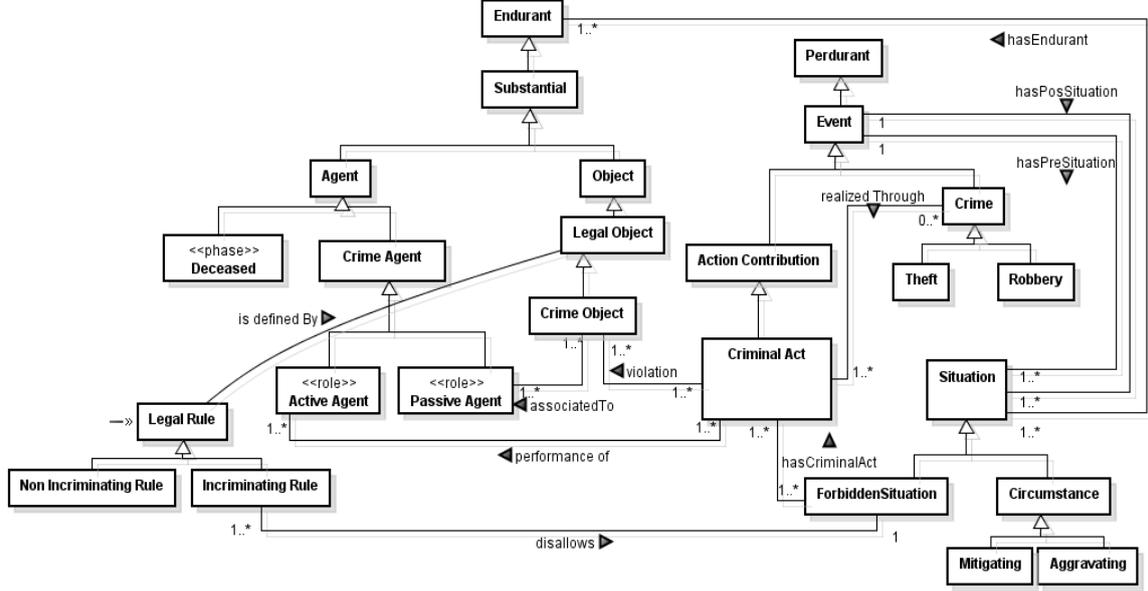


Fig. 4. A Partial Model for Crime and Agents

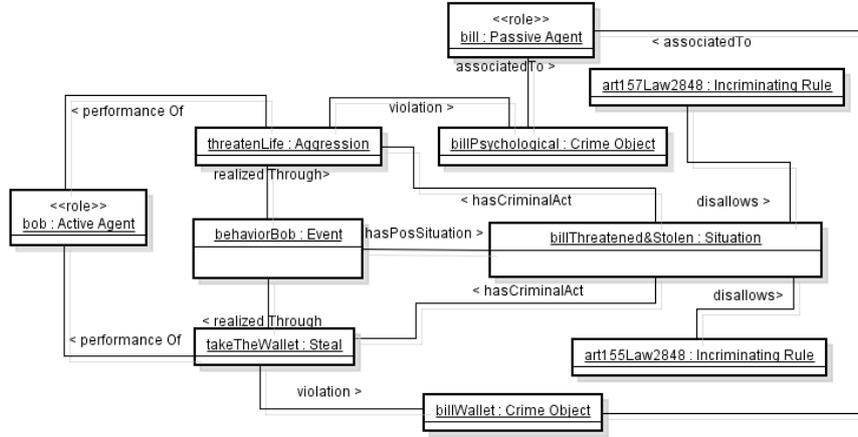


Fig. 5. Instantiation of Bob's behavior (Example 3.1)

tween classes in the UML class diagram. We also avoid overloading ABox \mathcal{A} with too many axioms. Among other results of logical subsumption, classical inference reasoning in DL shows further that (given $\mathcal{KB} = \mathcal{T} \cup \mathcal{A}$):

$$\mathcal{KB} \models \left\{ \begin{array}{l} \text{Theft}(\text{behaviorBob}), \\ \text{Robbery}(\text{behaviorBob}) \end{array} \right\}$$

From a logical point of view, the axiomatization in Example 3.1 is consistent. According to what has been stated, Bob is classified as an ActiveAgent instance. More importantly, his behavior is classified in two ways. The consequence entailed by \mathcal{KB} classifies behaviorBob as both a theft and a robbery. Although we have not shown in Figure 4, a criminal norm defines the sanctions associated. Instance *art155Law2848* defines as punishment a prison time between 1 and 4 years, while instance *art157Law2848* defines a prison time between 4 and 10 years.

It turns out that, if crimes are related by classical subsumption relations, they will potentially lead to different penalties, which would violate the double jeopardy principle. Suppose now that we provide an expert system for decision-making in the criminal realm. When reasoning about a given situation, and its possible relations with crimes and their respective penalties, we are actually seeking the most specific responses. This suggests that the knowledge base has to be built with a distinct subsumption relation; one whose semantics is able to accommodate a hierarchy of laws in which only the penalty of the most specific law is summoned. Such inference requirement is not fulfilled by classical DLs in a natural way.

4. Suitability of Preferential DL to the Legal Domain

Legal systems, governed by laws and principles, comprise many conditions that preclude the strict use of monotonic logics. Generalization (ruled by *modus ponens* inference rule) only addresses the most typical cases. As stated before, preferential DL overcomes this situation, by means of a non-monotonic notion of concept subsumption. Next, we present the defeasible TBox \mathcal{T}_d for crimes against property.

| | | |
|-------------------|---|--|
| $\mathcal{T}_d =$ | Crime | \equiv Event \sqcap \exists realizedThrough.CriminalAct, |
| | CriminalAct | \sqsubseteq Action, |
| | CriminalAct | \sqsubseteq \exists performanceOf.ActiveAgent \sqcap \exists violation.CrimeObject, |
| | CrimeObject | \sqsubseteq \exists isAssociatedTo.PassiveAgent, |
| | ForbiddenSituation | \sqsubseteq \exists hasCriminalAct.CriminalAct, |
| | IncriminatingRule | \sqsubseteq \exists disallows.ForbiddenSituation \sqcap \forall disallows.ForbiddenSituation, |
| | hasCriminalAct | \equiv hasDisallowedSituation $^-$, |
| | hasDisallowedSituation | \sqsubseteq hasPosSituation, |
| | Aggression | \sqsubseteq CriminalAct, |
| | Steal | \sqsubseteq CriminalAct, |
| | Event \sqcap \exists realizedThrough.Steal | \sqsubset Theft ⁽¹⁾ , |
| | Event \sqcap \exists realizedThrough.Steal \sqcap \exists realizedThrough.Aggression | \sqsubset Robbery ⁽²⁾ , |
| | Robbery | \sqsubseteq \neg Theft, |
| | Situation | \sqsubseteq \exists hasEndurant.Endurant, |
| | Circumstance | \equiv Aggravating \sqcup Mitigating, |
| | Aggravating | \sqsubseteq \neg Mitigating, |

Unlike the previous TBox \mathcal{T} , \mathcal{T}_d allows for both classical and defeasible subsumption statements. In \mathcal{T}_d

there are some slight modifications in the axioms (1) and (2). The latter axioms state that, typically, stealing something is a theft (1). New conditions may defeat this conclusion: stealing accompanied by violence or verbal threat is a robbery (2). Nevertheless, under the semantics of rational consequence (particularly, regarding the rational monotony property), given the premises:

$$\begin{aligned} \text{Event} \sqcap \exists \text{realizedThrough.Steal} &\sqsubset \text{Theft}, \\ \text{Event} \sqcap \exists \text{realizedThrough.Steal} &\sqsubset \neg \exists \text{realizedThrough.Aggression}, \end{aligned}$$

we still can infer:

$$\text{Event} \sqcap \exists \text{realizedThrough.Steal} \sqcap \exists \text{realizedThrough.Aggression} \sqsubset \text{Theft}$$

Based on this conclusion, even taking into account the new knowledge base $\mathcal{KB}_d = \mathcal{T}_d \cup \mathcal{A}$, as we know that Robbery $\sqsubseteq \neg$ Theft, then:

$$\mathcal{KB}_d \models \perp$$

The knowledge base $\langle \mathcal{A}, \mathcal{T}_d \rangle$ still does not match the expected practical result in the legal field, since dual classification violates the principle already mentioned and makes the knowledge base inconsistent. The RM property still forces the inference of theft. It is necessary to maximize the typicality of objects; in the context, to inform that, normally, events carried out through stealing are not accompanied by aggression:

$$\begin{aligned} \mathcal{T}'_d = \mathcal{T}_d \cup \{ \text{Event} \sqcap \\ \exists \text{realizedThrough.Steal} &\sqsubset \neg \exists \text{realizedThrough.Aggression} \} \end{aligned}$$

Thus, from $\mathcal{KB}'_d = \mathcal{T}'_d \cup \mathcal{A}$, we reach the desired conclusion:

$$\begin{aligned} \mathcal{KB}'_d &\models \text{Robbery}(\text{behaviorBob}), \\ \mathcal{KB}'_d &\not\models \text{Theft}(\text{behaviorBob}) \end{aligned}$$

Based on the notions of typicality and preference aforementioned, in the next subsection we propose a preferential and rational overrule relations to hierarchically organize a set of axioms within an arbitrary legal knowledge base. Some mathematical properties of these relations are given as well.

4.1. Preferential DL for Double Jeopardy Principle

The non-monotonic subsumption relation used in this study establishes an implicit priority among defeasible axioms. Given that, the contribution of the present paper is the provision of an approach for modelling legal domains in the presence of the Double Jeopardy principle. In other words, we propose a non-monotonic specification of crimes of the criminal code, based on an implicit ordering of precedence on norms. For this, we define overrule relations and denote when they are (implicitly) applied under preferential and rational semantics. In effect, we have addressed the Priority of Specificity (according to B eltran and Ratti (2013) [23]), which presupposes that more specific information cancels out more general ones. Consider that for each defeasible subsumption axiom γ , it has an antecedent (A) and a consequent (C) part, expressed as follows:

$$\gamma : A(\gamma) \sqsubseteq C(\gamma)$$

From this, we propose the following overrule relations.

Definition 4.1. Preferential Overrule Relation ($>_p$). Let us take \sqsubseteq as a preferential subsumption relation (Definition 2.3). Thus, given a consistent defeasible Tbox \mathcal{T} and two defeasible axioms (α and β), in which $\mathcal{T} \models \alpha$ and $\mathcal{T} \models \beta$, and neither $\mathcal{T} \models C(\alpha) \sqsubseteq \perp$ nor $\mathcal{T} \models C(\beta) \sqsubseteq \perp$. We say β overrules α (under the preferential semantics) if and only if $\mathcal{T} \models A(\beta) \sqsubseteq A(\alpha)$, and $\mathcal{T} \models C(\beta) \sqcap C(\alpha) \sqsubseteq \perp$. Thus, in general, $\beta >_p \alpha$ if:

$$\mathcal{T} \models \left\{ \begin{array}{l} \alpha : \prod_{i=1}^n D_i \sqsubseteq C(\alpha) \\ \beta : \prod_{i=1}^n D_i \prod_{j=1}^m E_j \sqsubseteq C(\beta) \\ \eta : C(\alpha) \sqcap C(\beta) \sqsubseteq \perp \end{array} \right\}$$

By handling the exception between α and β , it must be stated that their consequent portions are disjoint (η axiom). Otherwise, an individual could be classified as both concept expressions ($C(\alpha)$ and $C(\beta)$); which would not lead to an exception.

Lemma 1. The preferential overrule relation $>_p$ is ir-reflexive.

Proof. Assume that $>_p$ is reflexive. Thus, given a defeasible axiom α and a TBox \mathcal{T} , where $\mathcal{T} \models \alpha$, then $\alpha >_p \alpha$. This leads to a contradiction, with Definition 4.1, since $\mathcal{T} \not\models C(\alpha) \sqcap C(\alpha) \sqsubseteq \perp$, taking into account the hypothesis that $\mathcal{T} \not\models C(\alpha) \sqsubseteq \perp$. \square

Lemma 2. The preferential overrule relation $>_p$ is asymmetric.

Proof. Assume that $>_p$ is not asymmetric. Hence, there are at least two defeasible axioms α and β , and a TBox \mathcal{T} , where $\mathcal{T} \models \alpha$ and $\mathcal{T} \models \beta$, and $\beta >_p \alpha$ and $\alpha >_p \beta$. From Definition 4.1, we have $A(\beta) \sqsubseteq A(\alpha)$ and $A(\alpha) \sqsubseteq A(\beta)$. In other words, $A(\alpha) \equiv A(\beta)$. However, still according to Definition 4.1, the consequent part of the axioms are disjoint, and hence: $\mathcal{T} \models C(\alpha) \sqcap C(\beta) \sqsubseteq \perp$. Similarly to the last proof, assuming the hypotheses $\mathcal{T} \not\models C(\alpha) \sqsubseteq \perp$ and $\mathcal{T} \not\models C(\beta) \sqsubseteq \perp$, and as we already know that $A(\alpha) \equiv A(\beta)$, thus $A(\alpha) \sqsubseteq C(\alpha)$ and $A(\alpha) \sqsubseteq C(\beta)$. Clearly we would have $\mathcal{T} \models \top \sqsubseteq \perp$, since a typical instance of $A(\alpha)$, by the preferential/rational subsumption, could not be simultaneously classified as disjoint concepts. In other words, from the previous axioms, we would have $A(\alpha) \sqsubseteq C(\alpha) \sqcap C(\beta)$, thus $A(\alpha) \sqsubseteq \perp$. Hence, we confirm the asymmetry of the overrule relation. \square

Lemma 3. The preferential overrule relation $>_p$ is intransitive.

Proof. Consider the following \mathcal{KB} , with arbitrary concepts:

$$\mathcal{KB} \models \left\{ \begin{array}{l} \alpha : X \sqsubseteq Y \\ \beta : X \sqcap Z \sqsubseteq H \\ \gamma : X \sqcap Z \sqcap O \sqsubseteq P \\ Y \sqcap H \sqsubseteq \perp \\ H \sqcap P \sqsubseteq \perp \end{array} \right\}$$

From Definition 4.1, we have $\beta >_p \alpha$ and $\gamma >_p \beta$. Nevertheless, as $\mathcal{KB} \not\models Y \sqcap P \sqsubseteq \perp$, we can not infer $\gamma >_p \alpha$. On the other hand, considering three or more defeasible axioms, if all its consequent parts are pairwise-disjoint (as long as the criterion concerning its antecedent parts is also met), the transitivity property will be preserved. For the latter example, just note that adding $Y \sqcap P \sqsubseteq \perp$ in \mathcal{KB} will make the axioms α, γ match all the criteria of the overrule relation. \square

Lemma 4. The preferential overrule relation $>_p$ is acyclic.

From Lemma 2, it is trivially proven that there are no cycles between overrule relations. Once again, the veracity of the Lemma is assured only if the hypothesis that the consequent parts of the axioms are satisfiable. Therefore, if the cyclic property were true, for two axioms α and β , if $\beta >_p \alpha$ then $\alpha >_p \beta$. But this would

make the overrule relation also symmetrical, which has already been proven to be false in Lemma 2.

We provide the overrule relation for the rational subsumption relation as well. In order to facilitate understanding, we define an operator of difference between DL formulae.

Definition 4.2. Specificity Subtraction. Given two DL concepts (\mathcal{F}_1 and \mathcal{F}_2 , where $\mathcal{F}_1 \sqsubseteq \mathcal{F}_2$) of the form $\prod_{i=1}^j E_i$, \setminus_{DL} is an operator of specificity subtraction, where $\mathcal{F}_1 \setminus_{DL} \mathcal{F}_2$ results in a DL formula \mathcal{F}_3 , where $\mathcal{F}_2 \sqcap \mathcal{F}_3 \equiv \mathcal{F}_1$.

Definition 4.3. Rational Overrule Relation ($>_r$) Let us take \sqsubseteq as a rational subsumption relation (Definition 2.4). Thus, given a defeasible TBox \mathcal{T} and two defeasible axioms α and β , such that $\mathcal{T} \models \alpha$ and $\mathcal{T} \models \beta$, and neither $\mathcal{T} \models C(\alpha) \sqsubseteq \perp$ nor $\mathcal{T} \models C(\beta) \sqsubseteq \perp$. We say β overrules α if and only if $\mathcal{T} \models A(\beta) \sqsubseteq A(\alpha)$, $\mathcal{T} \models C(\beta) \sqcap C(\alpha) \sqsubseteq \perp$, and $\mathcal{T} \models A(\alpha) \sqsubset \neg \mathcal{F}$, where $\mathcal{F} = A(\beta) \setminus_{DL} A(\alpha)$. Thus, in general, $\beta >_r \alpha$, if:

$$\mathcal{T} \models \left\{ \begin{array}{l} \alpha : \prod_{i=1}^n D_i \sqsubset C(\alpha) \\ \beta : \prod_{i=1}^n D_i \prod_{j=1}^m E_j \sqsubset C(\beta) \\ \eta : C(\alpha) \sqcap C(\beta) \sqsubseteq \perp \\ \rho : \prod_{i=1}^n D_i \sqsubset \neg(\prod_{j=1}^m E_j) \end{array} \right\}$$

$$\text{and } \prod_{j=1}^m E_j \equiv A(\beta) \setminus_{DL} A(\alpha).$$

Lemmas 1-4 apply directly to rational overrule relation.

The next section details a case study within the Brazilian criminal code governing crimes against property.

4.2. Expanding Specificity Levels in Crimes against Property

In this section, we continue with the case study of crimes against property, leveraging the previous example with new additional circumstances. A new type of crime is introduced: Robbery Followed By Death. Henceforth, we use Example 4.1 to illustrate the levels of specificity between norms.

Example 4.1. Bill is a graduate student and attends classes at night shift. One day, returning home, Bill was approached by a biker who ordered Bill to give him his wallet. Frightened, Bill tried to run, but Bob shot him to death. Bob picked up Bill's wallet lying on the floor, and walked away.

We present the knowledge base for the subset of crimes against property (\mathcal{KB}_{cp}). Only for didactic reasons, we separate the Tbox into two distinct sets, splitting the classical and defeasible axioms (respectively, \mathcal{T}_c and \mathcal{T}_d). Additionally, the Example 4.1 is described shortly after by the corresponding assertional axioms. Figure 6 models the corresponding individuals.

This object diagram was also instantiated based on the classes and associations highlighted in Figure 4. We emphasize the concept of Deceased, that is, that agent who had his life (a crime object) violated by some criminal action.

| | | | |
|-------------------|--|--|---|
| $\mathcal{T}_c =$ | Crime | \equiv Event \sqcap \exists realizedThrough.CriminalAct, | } |
| | CriminalAct | \sqsubseteq Action, | |
| | CriminalAct | \sqsubseteq \exists performanceOf.ActiveAgent \sqcap \exists violation.CrimeObject, | |
| | CrimeObject | \sqsubseteq \exists isAssociatedTo.PassiveAgent, | |
| | ForbiddenSituation | \sqsubseteq \exists hasCriminalAct.CriminalAct, | |
| | IncriminatingRule | \sqsubseteq \exists disallows.ForbiddenSituation \sqcap \forall disallows.ForbiddenSituation, | |
| | hasCriminalAct | \equiv hasDisallowedSituation ⁻ , | |
| | hasDisallowedSituation | \sqsubseteq hasPosSituation, | |
| | Deceased | \equiv Agent \sqcap \exists hasViolatedObject.Life, | |
| | Life | \sqsubseteq CrimeObject, | |
| | hasViolatedObject | \equiv isAssociatedTo ⁻ , | |
| | Aggression | \sqsubseteq CriminalAct, | |
| | Situation | \sqsubseteq \exists hasEndurant.Endurant, | |
| | Steal | \sqsubseteq CriminalAct, | |
| | Event \sqcap \exists realizedThrough.Steal | \sqsubseteq EventOfSteal, | |
| | Event \sqcap \exists realizedThrough.Aggression | \sqsubseteq EventOfAggression, | |
| | Theft | \sqsubseteq \neg Robbery, | |
| | Robbery | \sqsubseteq \neg RobberyAndMurder, | |
| | RobberyAndMurder | \sqsubseteq \neg Theft | |

| | | | |
|-------------------|--|---|---|
| $\mathcal{T}_d =$ | EventOfSteal | \sqsubset Theft, | } |
| | EventOfSteal | \sqsubset \neg EventOfAggression, | |
| | EventOfSteal \sqcap EventOfAggression | \sqsubset Robbery, | |
| | EventOfSteal \sqcap EventOfAggression | \sqsubset \neg \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)), | |
| | EventOfSteal \sqcap EventOfAggression \sqcap \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)) | \sqsubset RobberyAndMurder | |
| | EventOfSteal \sqcap EventOfAggression \sqcap \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)) | \sqsubset RobberyAndMurder | |
| | EventOfSteal \sqcap EventOfAggression \sqcap \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)) | \sqsubset RobberyAndMurder | |
| | EventOfSteal \sqcap EventOfAggression \sqcap \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)) | \sqsubset RobberyAndMurder | |
| | EventOfSteal \sqcap EventOfAggression \sqcap \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)) | \sqsubset RobberyAndMurder | |
| | EventOfSteal \sqcap EventOfAggression \sqcap \exists hasPosSituation. (\exists hasEndurant. (PassiveAgent \sqcap Deceased)) | \sqsubset RobberyAndMurder | |

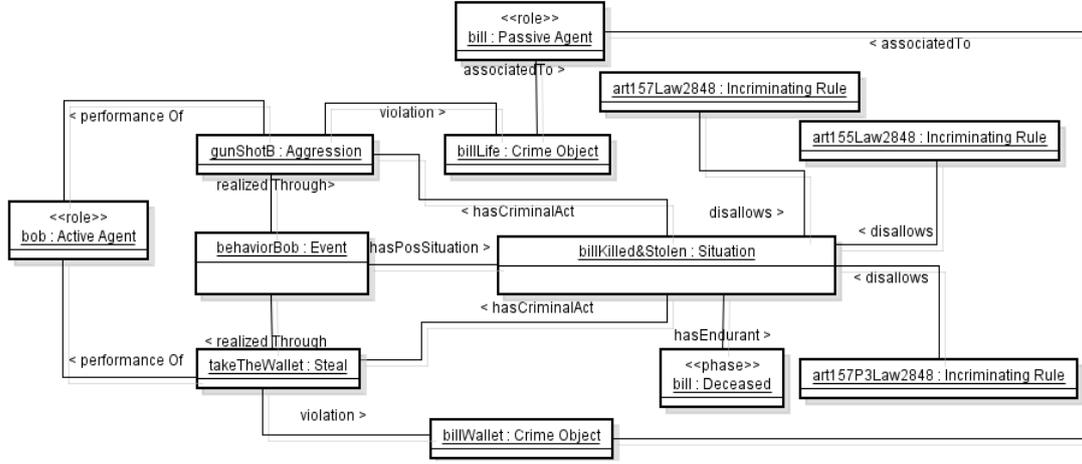


Fig. 6. Instantiation of Bob's behavior (Example 4.1)

$\mathcal{A} = \left\{ \begin{array}{l} \text{Event}(\text{behaviorBob}), \text{Aggression}(\text{gunShotB}), \\ \text{Steal}(\text{takeTheWallet}), \text{Situation}(\text{billKilledAndStolen}), \\ \text{Life}(\text{billLife}), \text{CrimeObject}(\text{billWallet}), \\ \text{Agent}(\text{bob}), \text{PassiveAgent}(\text{bill}), \\ \text{IncriminatingRule}(\text{art157Law2848}), \\ \text{IncriminatingRule}(\text{art155Law2848}), \\ \text{IncriminatingRule}(\text{art157P3Law2848}), \\ \text{performanceOf}(\text{takeTheWallet}, \text{bob}), \\ \text{performanceOf}(\text{gunShotB}, \text{bob}), \\ \text{realizedThrough}(\text{behaviorBob}, \text{takeTheWallet}), \\ \text{realizedThrough}(\text{behaviorBob}, \text{gunShotB}), \\ \text{hasCriminalAct}(\text{billKilledAndStolen}, \text{takeTheWallet}), \\ \text{hasCriminalAct}(\text{billKilledAndStolen}, \text{gunShotB}), \\ \text{violation}(\text{gunShotB}, \text{billLife}), \text{violation}(\text{takeTheWallet}, \text{billWallet}), \\ \text{isAssociatedTo}(\text{billLife}, \text{bill}), \text{isAssociatedTo}(\text{billWallet}, \text{bill}), \\ \text{hasDisallowedSituation}(\text{behaviorBob}, \text{billKilledAndStolen}), \\ \text{hasEndurant}(\text{billKilledAndStolen}, \text{bill}), \\ \text{disallows}(\text{art157Law2848}, \text{billKilledAndStolen}), \\ \text{disallows}(\text{art155Law2848}, \text{billKilledAndStolen}), \\ \text{disallows}(\text{art157P3Law2848}, \text{billKilledAndStolen}) \end{array} \right\}.$

The aforementioned TBox \mathcal{T}_d formalizes the defeasible axioms, starting with the most typical case. Normally, stealing something from someone is a Theft. Notwithstanding, whether the act occurs with the use of brutal force or verbal threat, the situation must be categorized just as a Robbery. Another level of specificity is established on Robbery: if the aggression results in the death of the passive agent, there is still a single criminal type, RobberyAndMurder. Even with this nomenclature, within Brazilian legal system, it does not make sense to claim the existence of two crimes, Robbery and Murder.

Thus, we can easily verify the following rational overrule relations (definition 4.3) between the defeasible axioms:

$$\begin{array}{l} \text{EventOfSteal} \sqcap \text{EventOfAggression} \sqcap \exists \text{hasPosState}. (\exists \text{hasEndurant}. (\text{PassiveAgent} \sqcap \text{Deceased})) \sqsubset \text{RobberyAndMurder} \\ \text{EventOfSteal} \sqcap \text{EventOfAggression} \sqsubset \text{Robbery} \\ \text{EventOfSteal} \sqsubset \text{Theft} \end{array}$$

Additionally, considering the Example 4.1, we have: $\mathcal{KB}_{cp} \models \text{RobberyAndMurder}(\text{behaviorBob})$ and $\mathcal{KB}_{cp} \not\models \text{Theft}(\text{behaviorBob})$ and $\mathcal{KB}_{cp} \not\models \text{Robbery}(\text{behaviorBob})$.

4.3. Reasoning with the DIP Protégé Plug-in

Given other numerous DL extensions with defeasibility (detailed in Section 5), as far as we know, few robust and scalable implementations have been made available. With respect to the Preferential DL, an existing Protégé plugin can be used, at least at the TBox level, for defeasible reasoning under the rational semantics, known as DIP (acronym for Defeasible Inference Platform) [34].

DIP is composed of three view windows: one for regular axioms (Figure 7 (1)), one for defeasible axioms (Figure 7 (2)), and the DIP view itself for reasoning (Figure 7 (3)). In order to tag an axiom as defeasible, a toggle button (d) is available. Another point to consider is that DIP makes call to a classical reasoner,

1 FACT++¹⁰. Figure 7 illustrates these windows within
2 DIP tab.

3 For the illustrated scenario in Figure 7, the query de-
4 scribes a situation where there is a dead agent (the vic-
5 tim). Subsequently, two actions were taken to arrive at
6 this state-of-affairs, a theft of an item, and an aggres-
7 sion. In addition to the rigid superclasses, DIP ranked
8 the situation as only a crime of RobberyAndMurder.
9 More importantly, no other crime has been mentioned,
10 although the query involves the inherent circumstances
11 of the robbery and theft crimes. Figure 8 illustrates the
12 situation for typical cases of robbery.

13 As a result of described actions, the dead person is
14 a crucial item to meet the exceptionality criteria. The
15 absence of this particularity classifies the situation as a
16 crime of Robbery, as depicted in Figure 8. Even so, the
17 presence of the steal event does not inflict the behavior
18 as a simple crime of Theft.

19 The fact that Preferential DL can be reasoned au-
20 tomatically within an ontology environment certainly
21 consists in one more advantage for Preferential DL, a
22 feature that most of the other non-monotonic alterna-
23 tives cannot offer.

24 5. Related Work

25 In this section, we highlight similar studies to deal
26 with non-monotonic nuances in the legal domain, and
27 alternative proposals to accommodate typical and non-
28 typical cases in DL formalism. In the end, we present a
29 comparative table between these last studies, singling
30 out why we chose the preferential approach.

31 5.1. Handling Conflicts in the Legal Domain

32 Since the 80s, debates about the need of non-
33 monotonicity as a necessary requirement for legal rea-
34 soning have been conveniently carried out, even per-
35 sisting to this day [35]. Situations such as the judicial
36 reasoning over incomplete (usually, unknown infor-
37 mation), the open-textured concepts [36], the hetero-
38 geneity of legal sources, and the very interplay among
39 laws with legal principles (which operate on a higher
40 level of reasoning w.r.t. laws, closing some gaps left
41 by them) [37] are potentially sources of conflicts.

42 In addition to the preferential approach, non-monotonic
43 logics have been widely used to deal with conflicts.
44 Some of these approaches are: Reiter’s default reason-

45 ing [38], McCarthy’s circumscription approach [39]
46 and the logical theory for defeasible reasoning of [40].
47 Other studies focused on approaches that set order
of preference between elements in conflicts, either in
a rule-based [41], [42], or on legal arguments per-
spective [43], [44], [45]. It is also important to note
that some papers [46], [47] have tried to define lev-
els of reasoning, where monotonic and non-monotonic
logic could coexist harmoniously. For such solutions,
at lower levels, from the current circumstances, facts
(arguably conflicting) are inferred. At a higher level,
if necessary, the problems would be solved. That is,
goal-setting rules at a second level would try to set pri-
orities (often presupposed by lawyers and judges), and
validate rules at a lower level.

48 5.2. Proposals for Handling Exceptions in DL

49 There are already studies about the problem of deal-
50 ing with exceptions¹¹ in ontologies, even inside the
51 Description Logic community. Let us consider a sketch
of five different approaches, in order to compare them
to our strategy:

- Neo-topological approach to reasoning on ontolo-
gies with exceptions proposed by [48];
- Defeasible description logics based on (defeasi-
ble) rules proposed by [49];
- A Non-monotonic description logic $\mathcal{ALC}+\mathbf{T}_{\min}$,
which defines an operator of typicality for de-
scription logic under a preferential semantics,
proposed by [50],
- A circumscriptive approach proposed by [51],
and finally
- A new semantics designed to address knowledge
engineering needs (called \mathcal{DL}^N), proposed by
[52].

“Neo-Topology” regards entities as points in space,
classes as clusters or groups of entities in space, and
classical topological operators (interior, border, and
closure). Additionally, neo-topology also uses the con-
cept of Typicality Degree to assign an element “L” (in
a class) which does not match all the properties of
a class, and creates the notion of “thickness” within
the border. This defines a topological area where it is
possible to assign those elements of the class which
are neither typical nor atypical, but rather falling in

¹¹In this subsection, as we are not specifically addressing the legal
domain, rather than conflicts, we speak of exceptions in a general
way.

¹⁰<http://owl.man.ac.uk/factplusplus/>

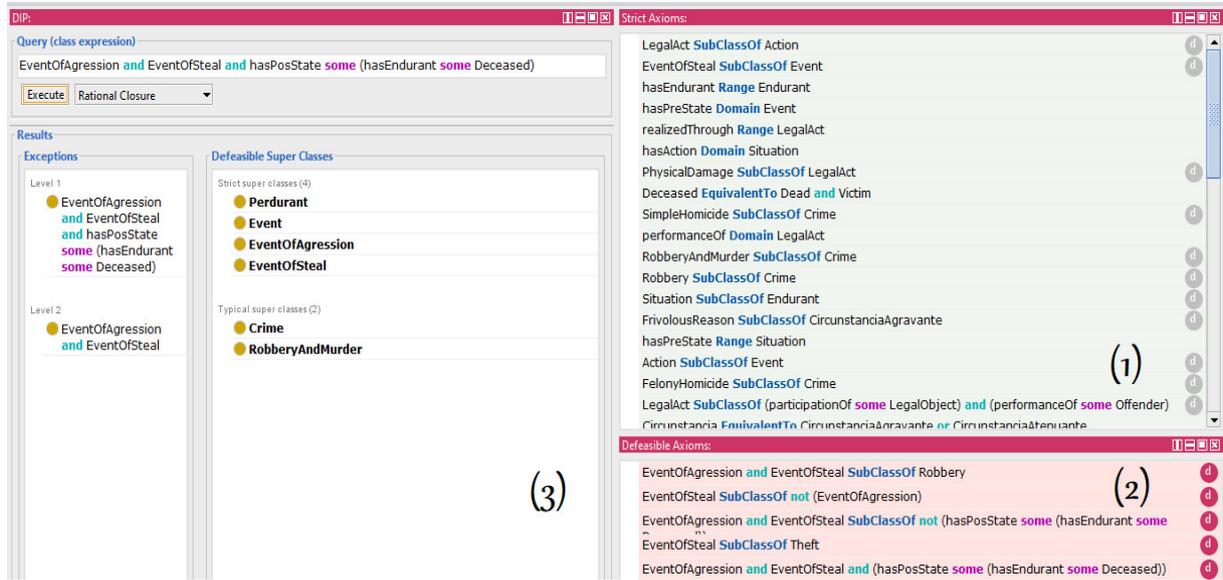


Fig. 7. DIP Reasoner

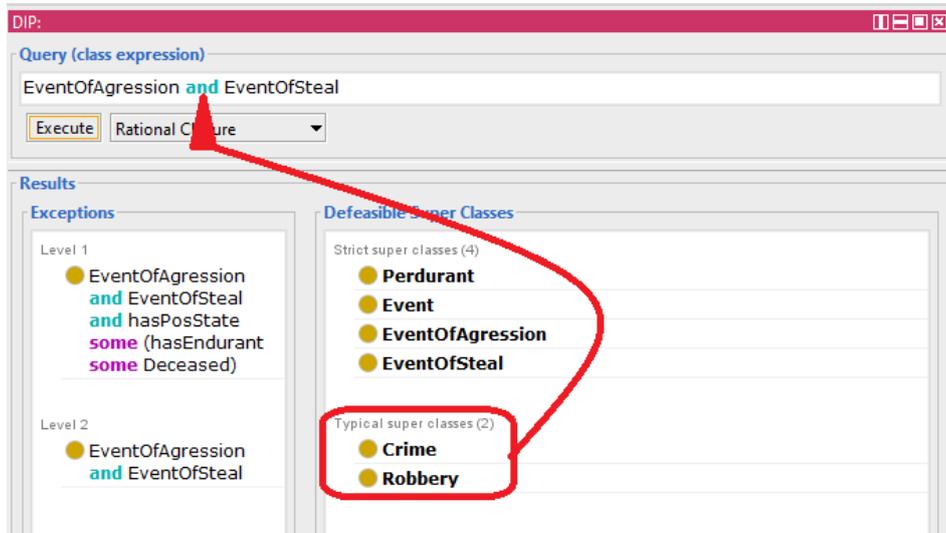


Fig. 8. Reasoning of Robbery crime

between. Doing so, neo-topology might express exceptions and even exceptions of exceptions. Although neo-topology has a user-friendly graphical representation, relationships between elements are not defined, different from description logic formalisms. Besides that, as far as we know, there are no results about complexity of neo-topology reasoning at all. In addition, the solution fails to have a clear underlying formalism,

and the rules of inference are limited to the operations of conjunctions.

A hybrid system to extend DL with defeasible rules is the proposal by [49]. Actually, the intention is to anchor DL in a Defeasible Logic, particularly the DefL (Defeasible Logic), proposed by [53]. Defeasible Logics are rooted in an underlying defeasible theory [40], [54], which predicts defeaters rules, and a superiority relation between (defeasible) rules. A defeasible the-

ory \mathcal{D} is a structure $\mathcal{D} = \langle F, R, > \rangle$ where F is a finite set of grounded literals (the facts), R is a finite set of FOL rules, and $>$ is an acyclic binary relation of superiority on R . Thus $\alpha > \beta$ states that α overrules β if both rules are applicable, and these rules have complementary head literals. DEF- \mathcal{ALC} is the result of a translation (τ) applied to the ABox/ TBox of a DL base in a language of a defeasible theory. In particular, the attempt to integrate the systems is only achieved through a new non-monotonic operator: \sqsubseteq , in order to provide an operator isomorphic to the DefL defeasible rules. Given $Cp1, Cp2$ as concepts, \rightarrow and \Rightarrow as the DefL classical strictly implication and defeasible implication (respectively), we have:

$$\begin{aligned} \tau(Cp1 \sqsubseteq Cp2) &= (\tau(Cp1) \rightarrow \tau(Cp2)) \\ \tau(Cp1 \sqsubset Cp2) &= (\tau(Cp1) \Rightarrow \tau(Cp2)) \end{aligned}$$

Still, as expected, the proposal provides a binary superiority relation ($>$) between the defeasible rules. The solution presented has several drawbacks, however. The translation does not address the blocking rules of the form $Cp1 \rightsquigarrow Cp2$ [54]. Furthermore, no data were presented about the proposal complexity, besides the overwork in translation between formalisms. Another point is the user's need to make explicit new rules, whenever new exceptions to superiority relations have emerged. Something that is often not a trivial task.

The DL extension proposed by [50], presents a semantics based on the KLM preferential semantics for non-monotonic reasoning. $\mathcal{ALC}+\mathbf{T}_{\min}$ contains a new operator (\mathbf{T}) of which the intuition is to single out the typical instances, rather than to extend some regular DL operator. As expected, operator \mathbf{T} preserves the postulates stated by KLM. In order to leverage the power of non-monotonic reasoning at its maximum, $\mathcal{ALC}+\mathbf{T}_{\min}$ strengthens the semantics of monotonic $\mathcal{ALC}+\mathbf{T}$ considering only the minimal model semantics. Therefore, given an arbitrary knowledge base \mathcal{KB} , and $\llbracket \mathcal{KB} \rrbracket$ as the set of models under $\mathcal{ALC}+\mathbf{T}$, $\mathcal{ALC}+\mathbf{T}_{\min}$ maximizes typicality selecting the models from $\llbracket \mathcal{KB} \rrbracket$ with the lowest number of atypical instances. By axiomatizing, for example, that, typically, stealing something from someone is a theft, is represented as follows:

$$\mathbf{T}(\text{EventOfSteal}) \sqsubseteq \text{Theft.}$$

Despite the well-founded semantics and the available tableau calculus decision procedure for checking

minimal entailment, as far as we know, there is no robust implementation for $\mathcal{ALC}+\mathbf{T}_{\min}$; conversely, tooling is crucial to engineer domain and assert reasoning capabilities.

[51] introduce a circumscriptive extension of \mathcal{ALC} . The idea behind the proposal is to augment the knowledge base with "abnormality predicates" (Ab_P), of which the extension is meant to be minimized in the reasoning process. In order to represent the aforementioned theft example, we define one of the following equivalent axioms:

$$\begin{aligned} \text{EventOfSteal} \sqcap \neg \text{Ab}_{\text{EventOfSteal}} &\sqsubseteq \text{Theft} \\ \text{EventOfSteal} &\sqsubseteq \text{Theft} \sqcup \text{Ab}_{\text{EventOfSteal}} \end{aligned}$$

Similar to [50]'s approach, the goal is to restrict inferences to models in which the extension of abnormality predicates is as minimal as possible. Circumscribed knowledge bases are defined in terms of a tuple $(\prec, M, \text{Fix}, V)$, where M are the abnormality predicates to be minimized, Fix comprises the unchanged predicates, V corresponds to the predicates that vary, and \prec is a strict partial order over M . Suppose, for example, the following knowledge base:

$$\mathcal{KB} : \left\{ \begin{array}{l} \text{EventOfSteal} \sqcap \neg \text{Ab}_{\text{EventOfSteal}} \sqsubseteq \text{Theft,} \\ \text{EventOfSteal} \sqcap \text{EventOfAggression} \sqcap \neg \text{Ab}_{\text{EventOfSteal} \sqcap \text{EventOfAggression}} \sqsubseteq \text{Robbery,} \\ \text{EventOfSteal}(\text{behaviorX}), \\ \text{EventOfAggression}(\text{behaviorX}) \end{array} \right\}$$

For correct inferences about behaviorX, it is imperative to make explicit the priorities among the abnormality predicates. Therefore, the following axiom is added to the base, to infer that behaviorX is a Robbery: $\text{Ab}_{\text{EventOfSteal} \sqcap \text{EventOfAggression}} \prec \text{Ab}_{\text{EventOfSteal}}$. Deciding priority relationships adds an extra workload in ontological engineering. Besides, mistaken decisions can lead to counter-intuitive inferences. Still, the worst problem of this solution is the computational complexity with respect to the underlying entailment relation: NEXP^{NP} -complete [55].

A new non-monotonic description logic named \mathcal{DL}^{N} was proposed by [52]. The motivations for the new formalism revolved around two purposes: to address issues related to the computational complexity of non-monotonic reasoning, and to solve shortcomings of prior non-monotonicity strategies, such as *inheritance blocking*, that is, exceptional concepts do not directly inherit the default properties of their superclasses. \mathcal{DL}^{N} handles conflicts within a knowledge

base through a new approach: unresolved conflicts are evidence of missing knowledge, thus knowledge engineers should use them as support for knowledge base scaling and validation. Another point to be highlighted is that \mathcal{DL}^N axioms can be converted to classical DL axioms in polynomial time.

\mathcal{DL}^N address non-monotonicity through two constructs: Normality Concepts and Defeasible Inclusions (DIs). The former refers to the standard instances of a concept C . That is, for each \mathcal{DL} concept C , there is a new concept name NC . DIs are expression of the form $C \sqsubseteq_N D$ which means “by default, standard instances satisfy $C \sqsubseteq D$, unless stated otherwise”. In the latter case, $C \sqsubseteq_N D$ is overridden by a higher priority DIs. A \mathcal{DL}^N knowledge base is a disjoint union, $\mathcal{KB} = \mathcal{S} \cup \mathcal{D}$, where \mathcal{S} comprises a finite set of classical axioms (the strong portion), and \mathcal{D} is a finite set of DIs. The proposed logic resolves conflicts between non-monotonic axioms by a strict partial order (\prec), which is usually based on specificity notion.

Although the notion of “typical instances” is not addressed, \mathcal{DL}^N was constructed with postulates analogous to those found in KLM theory. However, the first version of the \mathcal{DL}^N did not completely satisfy its KLM version of the LLE postulate. A correction was proposed by [56]. Moreover, given the encouraging results of scalability tests [52] for bases with tens of thousands of concepts, the approach, although very recent, emerges as a future line of research to deal with exceptions in modelling legal-normative knowledge. It is worth pointing out though that the approach by Bonatti et al. assumes an underlying DL language that is much less expressive than \mathcal{SHQ} , which is the main reason for the good scalability of their results. Here we assume a more expressive description language, which caters for a wider class of applications. Moreover, an advantage of preferential DLs as proposed by Britz et al. and as used here over the approach by Bonatti et al. is the fact that preferential DLs satisfy all the basic properties that are acknowledge as important by the non-monotonic reasoning community. Bonatti et al.’s framework turns out to fail some of them, one of the reasons being the fact their semantic construction is not preferential. In moving down to a less expressive language but keeping our preferential semantics, we believe we can get performance results comparable to those by Bonatti and colleagues while preserving the preferential properties.

Table 2 summarizes a brief comparison between the approaches described in this section. Preferential DL brings together a set of conditions that guided us to de-

fine it as the approach used in this research. Besides being based on a solid theory for non-monotonicity [24], practical implementations that formalize the defeasible reasoning were conceived [57]. In especial DIP [34], which already implements the Rational semantics. It is important to point out that the computational complexity of the implementation is the same as that of the monotonic entailment relation; i.e., it is an **EXPTIME**-complete problem.

Table 2
Comparing the Related Work

| References | Computational Complexity of Entailment | Practical Impl. | Non-Monot. Underlying Theory |
|--|--|-----------------|------------------------------|
| Neo-Topological [48] | Unknown | X | Defeasible Logic |
| Defeasible DL w/Defeasible rules [49] | Unknown | X | Defeasible Logic (DefL) |
| $\mathcal{ALC}+\mathbf{T}_{\min}$ [50] | coNEXP^{NP} | ✓ | KLM [24] |
| Circumscription [55] | NEXP^{NP} -complete | ✓ | Circumscription Pattern [39] |
| Preferential DL [57] | EXPTIME -complete | ✓ | KLM [24] |
| \mathcal{DL}^N [52] | pN²EXPTIME | ✓ | KLM-like [56] |

6. Contributions and Final Remarks

The maturity of the Semantic Web technologies has fostered the creation of more intelligent and autonomous web services, capable of interacting with one another and making decisions. Although not relatively new, the legal area provides a challenging field for knowledge conceptualization and reasoning, with many possible semantic pitfalls. DLs have been widely employed in the engineering of legal semantic models, with norms and regulations broken down into manageable-sized ontologies. Nevertheless, little explored in the axiomatization of the legal domain are the possibilities to overcome the limitations of classical DL, in particular, to handle conflicts that remain in normative knowledge. Exceptions accommodate and regulate the particularities of men living in society. Using them is not synonymous to error in the drafting of

1 laws, but rather, solutions so that juridical doctrine can
2 evolve along with social and political changes.

3 In this paper, we have made a case for a defeasible
4 extension of DL as a more suitable logical formalism
5 for the nuances that are intrinsic to the legal domain.
6 What was written in the Brazilian penal code as classic
7 types of subsumption, in fact, represent special situations
8 that, when not well conceptualized, lead to conclusions
9 that contradict the doctrines of the juridical order. We
10 have defined a theoretical foundation for the principle of
11 specificity, where more specific laws subjugate other
12 generic regulations, whenever their sanctions are likely
13 to lead to inconsistency. In consequence, it preserves
14 further the principle of *Non bis in idem*.
15

16 We have presented conceptualizations of some
17 crimes from the Brazilian penal code through a slight
18 modification in the implication operator of DL: \sqsupset .
19 Such modification allows us to separate instances of a
20 concept into two groups, the most typical, which obey
21 the rules of *modus-ponens*, and those that represent
22 special cases with respect to the base case. Therefore,
23 inferences can be gradually dropped out, inasmuch as
24 new information is being gathered. As evidences and
25 other legal artifacts can be set while trials are proceeding
26 in court, it is of crucial importance to provide reasoning
27 systems that draw conclusions which are no longer
28 supported.

29 The applicability of this study emerges through
30 Scientific-Technological and Legal contributions. The
31 former addresses the advance of the use of DLs for the
32 unambiguous and shared representation of knowledge,
33 favoring transparency and effectiveness of law enforcement
34 and the elimination of the gap between the use of
35 technologies and legal systems. In the legal context,
36 as we tighten the links between these areas, we foster
37 a field for building systems that are skilled at carrying
38 out activities such as: legal action simulation, legal
39 compliance checking, conflicts' resolution, among others.
40

41
42 In this work, we have assumed a single objective
43 ordering on the objects of the domain, along the lines of
44 the tradition in non-monotonic reasoning and related
45 areas such as conditional logics and belief revision. Of
46 course, one of the drawbacks of having a single ordering
47 is the fact that, within a preferential interpretation,
48 objects are compared in absolute terms, not relative to
49 a context.

50 Recently, Britz and Varzinczak have proposed a
51 preferential extension of DLs addressing precisely this

1 issue [58]. There, a notion of context based on roles
2 is introduced, which gives rise to multiple preference
3 relations on the set of objects of the domain. With that,
4 it becomes possible to specify that an object x is more
5 normal than y w.r.t. a *context* r_1 , whereas y is more
6 typical than x w.r.t. a *different context* r_2 .

7 In the present work, we have decided to stick to the
8 single-preference version of preferential DLs because
9 it is the only one for which a working implementation
10 is available. An extension of DIP to handle multiple
11 preferences forms part of the upcoming tasks in the
12 development of non-monotonic DLs.
13

14 Looking forward, this research has shown encouraging
15 results, allowing the theoretical framework to be
16 expanded soon to deal with other principles. This is the
17 case of the “Trifle principle”, which removes any criminal
18 liability if the stolen property is of irrelevant value. This
19 is entirely related to the more general principle of
20 *De Minimis Non Curat Lex* [59], in which a behavior
21 with extremely low transgression of the law is not
22 classified as illegal. We also plan to provide a separate
23 ontology, capable of performing specific tasks such as
24 the applicability of laws. Thus, we would separate the
25 domain itself from the probable legal actions that can
26 be carried out.

27 An ongoing work is the development of a prototype,
28 in which an ordinary user can simulate lawsuits in
29 real or fictitious cases through a friendly experience,
30 without bothering with the low formal level of DL.
31 LEGIS (LEGAl analysIS) is a web-based front-end system,
32 through which one can make functional and affordable
33 checks carried out by the mapped criminal ontologies.
34 We expect the results obtained so far might improve the
35 juridical understanding of layperson and aid the labor-
36 intensive task of lawsuits performed by professional
37 jurists.

38 Given that knowledge-based systems aim to support
39 human decision-making, the need for legible descriptions
40 of how the results were inferred becomes indispensable.
41 Therefore, another proposal of future work is the
42 integration of these tasks of reasoning with the research
43 proposed by [60], in order to allow the user to understand
44 the explanations of the inferences made. [60] proposes
45 a conversion method which translates *ALC* connection
46 proofs into *ALC* sequent proofs. The connection method
47 [61] is an efficient proof system for first-order logic
48 and it already has a variant for reasoning over ontologies
49 written in the DL *ALC*. However, its proofs are not very
50 readable, which makes interaction with users in general
51 difficult. With the conver-

sion to sequents, a more readable and intelligible representation is obtained, since sequent calculus [62] is essentially a formal logic argument style. The conversion method might be used in practical applications, in areas that employ DL reasoning and generate descriptions on natural language inferences for lay users. The proof conversion can help users understand why a particular situation is characterized as a crime, making its use viable in practice.

Finally, it should be noted that this project comes from a joint effort of cooperation between computer scientists and legal experts. The assumptions made here were the result of a joint analysis between these teams. From this cooperation, once LEGIS is fully operational, we will apply the prototype as a subsidiary system for decisions in local courts.

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Appendix A. Criminal Types In Portuguese

In Portuguese, the Crimes against property read as follows:

- **Furto:** *Subtrair, para si ou para outrem, coisa alheia móvel. (Art. 155);*
- **Roubo:** *Subtrair coisa móvel alheia, para si ou para outrem, mediante grave ameaça ou violência [...] (Art. 157);*
- **Latrocínio:** *Se da violência resulta [...] morte, a reclusão é de vinte a trinta anos (Art. 157, § 3º);*