

## Reasoning under inconsistency based on implicitly-specified partial qualitative probability relations: a unified framework

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### Abstract

Coherence-based approaches to inconsistency handling proceed by selecting preferred consistent subbases of the belief base according to a predefined method which takes advantage of explicitly stated priorities. We propose here a general framework where the preference relation between subsets of the belief base is induced by a system of constraints directly expressed by the user. Postulates taking their source in the qualitative modelling of uncertainty, either probabilistic or possibilistic, are used for completing the implicit specification of the preference relations. This enables us to define various types of preference relations, including as particular cases several well-known systems such as Brewka's preferred sub-theories or the lexicographical system. Since the number of preferred consistent subbases may be prohibitive, we propose to compile the inconsistent belief base into a new one from which it is easier to select one preferred consistent subbase.

and which consist in specifying an ordering relation on the formulas in  $\mathbf{K}$ , from which the preference relation on  $2^{\mathbf{K}}$  is induced. This is the case with Nebel (1991; 1994)'s syntax-based entailment, Brewka(1989)'s preferred sub-theories, Williams(1996)'s approach to belief revision, Geffner(1992)'s conditional entailment, the lexicographical system (Benferhat et al., 1993), (Lehmann, 1995), etc.

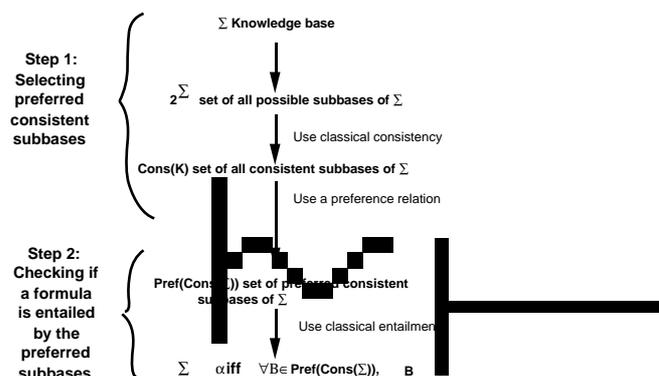


Figure 1

### 1. Introduction

Inconsistency may appear when a plausible consequence, obtained under incomplete information, has to be revised because further information is available. This issue has been extensively investigated in the nonmonotonic reasoning literature. In this paper we rather view inconsistency as being caused by the use (and the fusion) of multiple sources of information. Coherence-based approaches (Rescher; 1976) (Benferhat et al., 1995a) to inconsistency have two main steps, as shown in Fig. 1: i) build one or several preferred consistent subbases of the belief base  $\mathbf{K}$ , and ii) use classical entailment on these subbases.

A preference relation between subbases is a reflexive and transitive relation  $\geq$  on  $2^{\mathbf{K}}$ . Thus specifying it *explicitly* would need, in the extreme case,  $O(2^{2*|\mathbf{K}|})$  space which is not reasonable. It is why we have to *implicitly* specify a "small" set of constraints bearing on sets of formulas, which can be completed into a preference relation over  $2^{\mathbf{K}}$  using a set of postulates.

Two kinds of such constraints have been considered:

(a) *priority* constraints, which are qualitative in essence,

(b) *numerical* constraints, which consist in attaching to each formula a numerical weight which can be a probability degree (infinitesimal or not), a possibility degree (Dubois et al., 1994), a penalty value (Dupin et al., 1994), or an infinitesimal belief degree in belief functions theory (Benferhat et al., 1995b).

However, all these methods for specifying preferences implicitly lack flexibility on the way the preference relation is induced from the constraints, since the set of postulates is fixed once for all and is not a part of the representation language. In the first part of the article (Sections 2-4) we propose a general representation framework for coherence-based reasoning, which encompasses the well-known systems mentioned above. This framework offers a lot of flexibility, and for instance makes it possible to the user to define, a system which lays between Brewka's preferred sub-theories and the lexicographical system.

In the second part of the paper, we deal with the problem of representing the preferred consistent subbases compactly, in order to perform task 2 (see Fig. 1)

efficiently. For this purpose we propose a compilation technique which transforms  $\mathbf{K}$  into a new belief base

$(\mathbf{K})$  (consisting of additional formulas). This combination technique is inspired from recently introduced in the possibilistic logic setting (Benferhat et al., 1997). We show that this syntactic combination of belief bases defines a method for tracking inconsistency in the sense that plausible conclusion inferred from  $\mathbf{K}$  using some entailment inconsistency-tolerant consequence relation, is also consequence from a unique consistent subbase of  $(\mathbf{K})$ .

This paper emphasizes the syntactic aspect of the approach, while the companion paper (Benferhat et al., 1998) develops the semantic issues and concentrates on preference relations.

## 2. Specifying preferences explicitly

In this paper, we only consider a propositional language. The symbol  $\vdash$  represents the classical consequence relation, Greek letters  $\alpha, \beta, \dots$  present formulas. Let  $S = \{s_1, \dots, s_n\}$  be a set of sources (all  $s_i$  are different). A belief base  $\mathbf{K} = \{\phi(s_i) \mid s_i \in S\}$  is a multiset of the formulas  $\phi$  provided by sources. For the sake of simplicity, we simply write  $\phi$  instead of  $\phi(s_i)$ . The same belief can be present several times if it comes from different sources and this explains why we consider  $\mathbf{K}$  as a multiset. However,  $S$  is not a multiset but a set since all  $s_i$  are different. Thus, instead of working with  $2^{\mathbf{K}}$ , which would lead to ambiguity (since  $\mathbf{K}$  is a multiset), we prefer to work with  $2^S$ . Keep in mind that  $\phi$  is associated to  $s_i$ .

**Def. 1:** A partially quantitative probability (PQPP) relation  $\geq$  on  $2^S$  is a relation satisfying the following postulates (for  $X, Y, Z \subseteq 2^S$ ):  
 A1.  $\geq$  is transitive.  
 A2.  $X \geq Y$  implies  $X \supseteq Y$ ; (mean strict inclusion)  
 A3. If  $X, Y, Z$  are uniform subbases:  
 $X \cup Y \geq X \cup Z \Leftrightarrow Y \geq Z$ ,  
 where:  $X \supseteq Y$  means that  $X \geq Y$  and not  $Y \geq X$ .

The intuitive meaning of  $X \geq Y$  (resp.  $X \supseteq Y$ ) is that the set of sources in  $X$  is at least as preferred/prioritary/reliable as (resp. strictly more prioritary than) the set of sources in  $Y$ . Note that we do not require  $\geq$  to be connected ( $\geq$  is generally a partial order only), which entails that there may be incomparable subsets of  $2^S$ . The incomparability relation should not be confused with the equivalence (or indifference) relation defined by:  $X$  and  $Y$  are equivalent (denoted by  $X \approx Y$ ) iff  $X \geq Y$  and  $Y \geq X$ . The relation  $\geq$  is a kind of partially ordered qualitative probability (Lehmann, 1996), where all non-empty events have a "non-null" probability value due to A2. The cancellation property A3 is close to the one of comparative probabilities (Fishburn, 1986), although we remain in the qualitative framework.

Now, what has to be first specified is a set of constraints

being on the... included from the... repeatedly.

consists of... inequalities... there exist... (i.e.  $\Sigma$  ext... qualities of... The pr... closed...  $A = A$ ... properties... reaction... exists... sl...

**Proposition 1:** ... there exists a... denoted by  $\geq$ , such that for...  $\geq$  extend  $\Leftrightarrow$ ... be called... closure of... when... to... axioms A1-A3 then... the following way: let  $\mathbf{K} = \{(I, J) \in \mathcal{R} \mid I \geq J\}$  and  $\mathcal{R} \subseteq 2^S \times 2^S$ . Not complete...  $\mathcal{R} \subseteq 2^S \times 2^S$ . Finally... the transitive closure of  $\mathcal{R}$ .

**Example.** Let  $S = \{s_1, s_2, s_3, s_4\}$  and  $\mathcal{R} = \{(s_1, s_2), (s_2, s_3), (s_1, s_3)\}$ . The closure of  $\mathcal{R}$  is shown in Figure 2 (reflexivity and transitivity are not represented for sake of clarity). For instance  $\{s_1, s_4\} \supseteq \{s_2, s_3, s_4\}$  is derived from  $\{s_1\} \supseteq \{s_2, s_3\}$  and postulate A3.

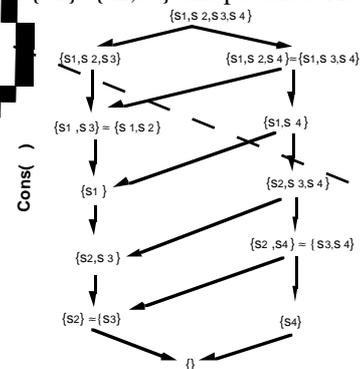


Figure 2

## 3. PQPP-preference-based entailment

Roughly speaking,  $>$  helps selecting preferred subbases of  $\mathbf{K}$  by removing all those which are not maximal.  $X > Y$  means that, in order to restore the consistency of  $\mathbf{K}$ , we prefer to maintain the set of beliefs supported by the sources in  $X$  rather than to maintain the set of beliefs supported by the sources in  $Y$ . In this way, the condition A2 becomes very natural since it corresponds to the idea of minimal change: we try to maintain as many pieces of information from  $\mathbf{K}$  as possible. A3 simply means that

only the form  $\alpha$  which does not contain  $Z$  should be taken into account. This is the case if  $Z$  is a bridge. Now, coherence based on  $\alpha$  is a special case of the initial specificity  $\alpha$ . Since  $\alpha$  is a bridge, the theories implied by  $\alpha$  are  $\alpha$ -maximal. If  $\alpha$  is not a bridge, then  $\alpha$  and a preference relation  $\geq$ , we can define a usual way  $\text{Max}(\text{Cons}(\alpha, \geq))$ . The minimal bridge can be defined in the usual way. For conjunction, we have:

**Def. 2:**  $X$  is a  $\geq$ -consistent element iff  $X$  is a consistent element and  $X \in \text{Max}(\text{Cons}(\alpha, \geq))$  for all consistent elements  $\alpha$  of  $\mathcal{K}$ .

**Def. 3:**  $K$  is a  $\geq$ -consistent theory iff  $K \in \text{Max}(\text{Cons}(\alpha, \geq))$  for all consistent theories  $\alpha$  of  $\mathcal{K}$ .

**Example 1:** Let  $\mathcal{K} = \langle \mathcal{L}, \text{AND} \rangle$  and  $\alpha = \{s_1, s_2, s_3, s_4, s_5\}$  with  $\phi_1 = s_1 \wedge s_2$ ,  $\phi_2 = s_2 \wedge s_3$ ,  $\phi_3 = s_3 \wedge s_4$ ,  $\phi_4 = \neg a \wedge c$ . Let  $\geq$  be a preference relation on  $\mathcal{K}$  such that  $\{s_2, s_3, s_4, s_5\} \geq \{s_1, s_2, s_3, s_4\}$ . Then  $\text{Max}(\text{Cons}(\alpha, \geq)) = \{s_2, s_3, s_4, s_5\}$  and therefore we have  $\text{Max}(\text{Cons}(\alpha, \geq)) = \{s_2, s_3, s_4, s_5\}$ .

Note that  $\{s_1, s_2, s_3, s_4\}$  is not a consistent theory w.r.t.  $\geq$ .

**Prop. 2:** If  $\alpha \geq \beta$  then  $\text{Max}(\text{Cons}(\alpha, \geq)) \subseteq \text{Max}(\text{Cons}(\beta, \geq))$ .

This is due to the fact that  $\text{Max}(\text{Cons}(\alpha, \geq))$  is a  $\geq$ -consistent theory. The non-monotonicity of  $\text{Max}(\text{Cons}(\alpha, \geq))$  is due to the fact that  $\alpha$  is not a bridge.

**Example 1 (continued):** Let  $\mathcal{K}' = \mathcal{K} \cup \{s_1, s_2, s_3, s_4, s_5\}$  and  $\phi_5 = \neg a \wedge \neg c$ . Let  $\geq$  be a preference relation on  $\mathcal{K}'$  such that  $\{s_2, s_3, s_4, s_5\} \geq \{s_1, s_2, s_3, s_4, s_5\}$ . Then  $\text{Max}(\text{Cons}(\alpha, \geq)) = \{s_2, s_3, s_4, s_5\}$  and  $\text{Max}(\text{Cons}(\alpha', \geq)) = \{s_2, s_3, s_4, s_5\}$ . We have:  $\text{Max}(\text{Cons}(\alpha, \geq)) = \text{Max}(\text{Cons}(\alpha', \geq))$  since  $\{s_2, s_3, s_4, s_5\} \geq \{s_1, s_2, s_3, s_4, s_5\}$ .

#### 4. Recovering coherence theories

This section shows that the element based P-QPP relation allows us to recover several coherence theories. Coherence based theories generally assume a stratification of sources  $\{s_1, \dots, s_p\}$ . We note  $s_i$  the source (according to an arbitrary numbering) in  $\mathcal{K}$ . This stratification expresses a total pre-order between the sources:  $\forall s_i, s_j \in \mathcal{K}, s_i > s_j$  iff  $i < j$ .

Let  $\alpha$  be a consistent theory of  $\mathcal{K}$ . We define  $\alpha$  as a  $\geq$ -consistent theory iff  $\alpha \in \text{Max}(\text{Cons}(\alpha, \geq))$ . We define  $\alpha$  as a  $\geq$ -consistent theory iff  $\alpha \in \text{Max}(\text{Cons}(\alpha, \geq))$ . We define  $\alpha$  as a  $\geq$ -consistent theory iff  $\alpha \in \text{Max}(\text{Cons}(\alpha, \geq))$ .

The  $\geq$ -maximal consistent elements are called preferred consistent elements (Brewka, 1989) and  $\geq$ -lex maximal consistent elements are called lexicographically preferred consistent elements (Benferhat et al., 1993; Schumann, 1995).

There are also some other coherence theories which associate positive integer numbers  $c_i$  (resp. symbolic numbers  $\alpha_i$ ) to the sources  $s_i$ . In this case, we use a symbolic penalty logic proposed (Dupin et al., 1994), which uses such assignments. The rational orders  $\text{Cons}(\alpha, \geq)$  in the following are:

(symbolically based ordering defined by  $X >_{\text{pen}} Y$  iff  $\sum\{c_i \mid s_i \in X\} < \sum\{c_j \mid s_j \in Y\}$  (resp.  $X >_{\text{spe}} Y$  iff  $\sum\{\alpha_i \mid s_i \in X\} < \sum\{\alpha_j \mid s_j \in Y\}$ ).

Coherence based on  $\geq$  implies  $X >_{\text{pen}} Y$ . Symbolic penalty logic is equivalent to an infinitesimal version of belief functions (Benferhat et al., 1995b).

The following conditions for recovering the previous coherence theories are summarized in Fig. 3, see (Benferhat et al., 1995b):

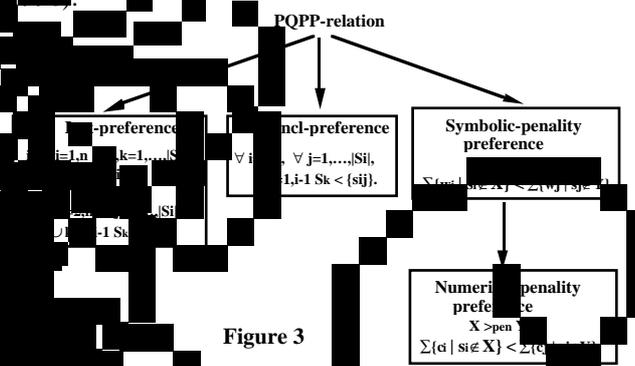


Figure 3

A natural question suggested by this picture is: can a penalty logic be represented by a numerical representation? PQPP-relation. Namely, we may wonder if for any PQPP-relation there exists a set of weights  $c_i$  associated with the sources  $s_i$  such that  $X >_{\text{pen}} Y$  iff  $\sum\{c_i \mid s_i \in X\} < \sum\{c_j \mid s_j \in Y\}$ . The answer is no, and we can use the same counter-example given by Kraft, Pratt and Seidenberg (1959) where they show that qualitative probability relations cannot be represented by a probability distribution. Let  $\mathcal{K} = \{s_1, s_2, s_3, s_4, s_5\}$  and  $\varepsilon = 1$ . Let  $c_1 = 400 - \varepsilon$ ,  $c_2 = 100 - \varepsilon$ ,  $c_3 = 300 - \varepsilon$ ,  $c_4 = 200$  and  $c_5 = 600$ . Then

define  $\geq$  in the following way:

- for any couple  $(X, Y) \neq (\{s_1, s_2, s_3\}, \{s_4, s_5\})$ :  
 $X \geq Y$  iff  $\sum\{c_i \mid s_i \in X\} \leq \sum\{c_j \mid s_j \in Y\}$
- $\{s_1, s_2, s_3\} > \{s_4, s_5\}$  (although  $c_4 + c_5 > c_1 + c_2 + c_3$ )

It can be shown that  $\geq$  cannot be represented by any numerical penalty logic. This is due to the following facts:

- Note that the qualitative preference relation  $\geq$  satisfies the following inequalities:  
 $\{s_1, s_4, s_5\} > \{s_2, s_3, s_4, s_5\}$ ,  $\{s_3, s_4, s_5\} > \{s_1, s_2, s_5\}$ ,  
 $\{s_2, s_4, s_5\} > \{s_1, s_3, s_4\}$ ,  $\{s_1, s_2, s_3\} > \{s_4, s_5\}$ ,  
 and that there is no  $(c_1, c_2, c_3, c_4, c_5)$  integer costs which satisfy the above inequalities. Indeed  $c_2 + c_3 < c_4$ ,  $c_1 + c_2 < c_3 + c_4$ ,  $c_1 + c_3 < c_4 + c_5$  implies  $c_1 + c_2 + c_3 < c_4 + c_5$ .
- The preference relation is a P-ordinality ordering defined on the numbers satisfies the axioms A1, A2, and A3 in the direction of the link  $\{s_1, s_2, s_3\} \rightarrow \{s_4, s_5\}$  has no effect on the satisfaction of formulas since  $X \neq \{s_1, s_2, s_3\} \rightarrow K \rightarrow \{s_4, s_5\} \rightarrow X \geq X \geq \{s_1, s_2, s_3\}$ .

Apart from the well known results, we can generalize the general framework, we may define a preference relation by specifying a set of constraints on the preference relation.

**Example 2: multi-criteria preference + Pareto**  
 Agent A:  $\{s_1, s_2, s_3\}$  and  $\{s_4, s_5\}$   
 of constraints  $\{s_1, s_2, s_3\} \geq A$ .  
 Similarly for Agent B we have  $\{s_4, s_5\} \geq B$ .  
 Letting  $K = K_A \cup K_B$  and adding the constraints  $\forall X \subseteq K, \forall Y \subseteq K$ ,  
 $\forall X \subseteq K_A, \forall Y \subseteq K_B$  we get a global preference relation, where the preferences are not commensurable.

Some systems, in particular systems based on the selection of a unique preferred consistent subbase, cannot be recovered using OPP-relations. An example of such systems is the possibilistic logic approach (Dubois et al., 1994), which uses satisfiability and where the ordering on  $\text{Cons}(K)$  is defined as follows:  
 - "best out" ordering: defined by  $X >_{Bo} Y$  if  $\text{Max}\{i \mid \exists s \in X_i \text{ and } s \notin Y_i\} > \text{Max}\{j \mid \exists s' \in Y_j \text{ and } s' \notin X_j\}$ .  
 convention  $\text{Min}\emptyset = 0$ .

Possibilistic logic can be recovered using the following postulates: (A1), (A'2)  $Y \subseteq X \Rightarrow X \geq Y$  and (A'3)  $Y \geq Z \Rightarrow X \cup Y \geq X \cup Z$ , instead of A1-A3. A'3 is the main axiom of qualitative possibility theory (Dubois et al., 1994). See (Dubois et al., 1992) for an explicit possibilistic handling of the sources.

### 5. Compiling inconsistent belief bases

Coherence-based approaches take blindly into account all the beliefs in  $K$  and compute the set of preferred consistent subsets of  $K$  although some beliefs are not

used in inferences. This is not the case of belief bases  $K = \{\alpha, \neg\alpha, \beta\}$  where all the beliefs are assumed to be equally reliable. This belief base is consistent and admits two preferred consistent subsets  $A = \{\alpha, \beta\}$  and  $B = \{\neg\alpha, \beta\}$ . We can see that for any formula  $\psi$ , we have both  $A \models \psi$  and  $B \models \psi$ . This means that we can ignore the beliefs  $\{\alpha, \neg\alpha\}$  without changing the set of plausible consequences of  $K$  in the sense of preferred consistent subsets inference.

Generalizing this idea we propose an alternative for representing preference relations. This is done in the following way:

- The first step is to define a preference relation  $\geq$  on the belief bases  $K$  by viewing the couple  $(K, \geq)$  as a preference relation on the set of formulas  $\mathcal{F}$ . Let  $\text{OR}(X) = \bigvee\{\phi_i \mid \phi_i \in X\}$  the disjunction of the original formulas  $X$ . We provide the source in  $\text{OR}(X)$  formally as follows in the following way:

**Def. 4:** Let  $(K, \geq)$  be a preference relation on the set of formulas  $\mathcal{F}$ . A belief base  $X$  is said to be subsumed by  $Y$  (written  $X \geq Y$ ) if  $\text{OR}(X) \geq \text{OR}(Y)$ .

Let  $\subseteq_2$ , and  $B = \{s_1, s_2, s_3\}$ . For any  $\text{OR}(Y)$  is said to be subsumed by  $\text{OR}(X)$  if  $\exists C \subseteq B$  such that  $X \geq Y$  and  $\text{OR}(X) \geq \text{OR}(Y)$ .

$\text{OR}(X)$  is unique in general, however  $\text{OR}(K)$  is equivalent in the sense that in the second step all the selected consistent subsets are classically equivalent.

**Example 1 (continued):** Let  $K = \{\phi_1, \phi_2, \phi_3, \phi_4\}$  with  $\phi_1 = a, \phi_2 = \neg a \vee b, \phi_3 = \neg a \vee \neg b, \phi_4 = \neg a \vee a \wedge c$ . After removing tautologies  $\{s_1, s_2, s_3\} \geq X \mid X \subseteq K$  we get  $\{s_1, s_2, s_3\} \geq X$ .

$\{s_1, s_2, s_3\} \geq X$  where  $X = \{\phi_1, \phi_2, \phi_3, \phi_4\}$ . Note that  $\phi_1$  is subsumed by  $\phi_2$  since  $\phi_1 \geq \phi_2$  and  $\{s_1, s_4\} > \{s_2\}$  and that  $\phi_3$  is also subsumed by  $\phi_2$  since  $\phi_3 \geq \phi_2$  and  $\{s_3, s_4\} > \{s_2\}$ . Therefore

$\text{OR}(K) = \text{OR}(\{s_1, s_2, s_3\}) = \phi_1 \vee \phi_2 \vee \phi_3 \vee \phi_4$ .  
 $\text{OR}(K) = \text{OR}(\{s_1, s_2, s_3\}) = \phi_1 \vee \phi_2 \vee \phi_3 \vee \phi_4$ .

In (Benferhat et al., 1998) an incremental algorithm has been proposed for computing  $\text{OR}(K)$ . This is possible by applying an associative and commutative binary combination operator, denoted by  $\diamond$ . We briefly recall the definition of this operator. Let  $K_1 = \{\text{OR}(X_i) \mid X_i \in \text{Cons}(K), i=1, k\}$  and  $K_2 = \{\text{OR}(Y_j) \mid Y_j \in \text{Cons}(K), j=1, m\}$ . The operator  $\diamond$  is defined by:

$K_1 \diamond K_2 = K_1 \cup K_2 \cup \{\text{OR}(X_i \cup Y_j) \mid X_i \in K_1, Y_j \in K_2, \text{OR}(X_i \cup Y_j) \text{ is not a tautology}\}$ .

As it can be seen, the operator  $\diamond$  introduces new disjunctions of formulas which are not necessarily substituted (in the sense of Def. 5). Let  $\mathbf{K} = \{K_i \mid i=1, n\}$  be a belief base. Let  $K_i = \{\phi_i\}$  be one formula belief base. We can show that  $\mathbf{K} \models \phi$  iff  $\bigwedge_{i=1}^n K_i \diamond \phi$ .

## 6. Selecting consistent subbases in $\mathbf{K}$

In the previous section, if the belief base  $\mathbf{K}$  has been compiled into a PQP, we know that all formulas in  $\mathbf{K}$  are consistent with  $\text{Min}(\mathbf{K})$ . In this section, we select a maximal consistent subset of  $\mathbf{K}$  (denoted by  $\text{BC}(\mathbf{K})$ ), consisting of beliefs and roughly speaking, these pieces of information in  $\mathbf{K}$  are not blocked by any minimally inconsistent subbase of  $\mathbf{K}$ . The following definitions are not in our PQP, but they formally define the notion of accepted beliefs.

**Def. 6:** A subbase  $\mathbf{K}'$  of  $\mathbf{K}$  is called a minimally inconsistent subbase if  $\mathbf{K}' \models \text{OR}(X) \in \text{Incons}(\mathbf{K})$ .

Let  $\text{Nogood}(\mathbf{K})$  be the set of all minimally inconsistent subbases of  $\mathbf{K}$ .

**Def. 7:** A formula  $\text{OR}(X)$  of  $\mathbf{K}$  is called an escape from a minimally inconsistent subbase  $\mathbf{K}'$  if  $\text{OR}(Y) \in \text{Nogood}(\mathbf{K})$  such that  $X > Y$ .

Finally, accepted beliefs are defined by:

**Def. 8:** A formula  $\text{OR}(X)$  of  $\mathbf{K}$  can be accepted in  $\mathbf{K}$  iff  $\text{OR}(X) \in \text{BC}(\mathbf{K})$ ,  $\text{OR}(X)$  escapes from  $\mathbf{K}'$  for all  $\mathbf{K}' \in \text{Nogood}(\mathbf{K})$ .

The following proposition shows that the notion of accepted beliefs can be recovered using accepted beliefs of the previous procedure (Benferhat et al., 1998).

**Proposition 3:** Let  $\mathbf{K}'$  be a subbase of  $\mathbf{K}$  such that any formula  $\text{OR}(X)$  of  $\mathbf{K}$  is accepted in  $\mathbf{K}'$  iff  $\text{BC}(\mathbf{K}) \subseteq \mathbf{K}'$ .

Fortunately, we do not need to compute all minimal inconsistent subsets of  $\mathbf{K}$  for computing  $\text{BC}(\mathbf{K})$ . It will distinguish two cases:

1. If the PQP preference relation is complete (i.e.  $\forall X, Y \subseteq \mathbf{K}$ , we have  $X \geq Y$  or  $Y \geq X$ ), it is enough to compute a subset of  $\mathbf{K}$ , denoted  $\text{Min}(\mathbf{K})$ . Then,  $\text{OR}(X) \in \text{BC}(\mathbf{K})$  iff  $X > \text{Incons}(\mathbf{K})$ . This calculation is largely developed in (Benferhat et al., 1998). An efficient procedure for computing  $\text{Incons}(\mathbf{K})$  is also proposed

in (Benferhat et al., 1998). If the PQP preference relation is not complete, we need to consider a subset of the set of minimally inconsistent subbases of  $\mathbf{K}$ , denoted by  $\text{Nogood}(\mathbf{K})$ .

This section focuses on the case where the PQP preference relation is not complete. Computing  $\text{BC}(\mathbf{K})$  is then more tricky, even if  $\text{Min}(\mathbf{K})$  is computed. All the following definitions are for computing  $\text{BC}(\mathbf{K})$ .

The following definitions are for computing  $\text{BC}(\mathbf{K})$ . The first one facilitates the computation of  $\text{BC}(\mathbf{K})$ . The second improvement is to reduce the number of minimally inconsistent subbases, namely:

**Proposition 4:**  $\text{Nogood}(\mathbf{K}) = \{ \mathbf{K}' \mid \mathbf{K}' \models \text{OR}(X) \in \text{Incons}(\mathbf{K}) \}$ .

To make further improvements, we need to refine the definition of  $\text{Nogood}(\mathbf{K})$  with the following definition:

**Def. 9:**  $\text{Min}(\mathbf{K}) \subseteq \mathbf{K}' \subseteq \mathbf{K}$  iff  $\mathbf{K}' \models \text{OR}(X) \in \text{Incons}(\mathbf{K})$ .

This extension is defined as  $\text{Nogood}^*(\mathbf{K}) = \{ \mathbf{K}' \mid \text{Min}(\mathbf{K}) \subseteq \mathbf{K}' \subseteq \mathbf{K} \text{ and } \mathbf{K}' \models \text{OR}(X) \in \text{Incons}(\mathbf{K}) \}$ . We will denote by  $\text{Nogood}^*(\mathbf{K})$  the set of all non-dominated minimally inconsistent subbases of  $\mathbf{K}$ .

Such a  $\text{Nogood}^*(\mathbf{K})$  is not empty. In fact,  $\text{Min}(\mathbf{K}) \in \text{Nogood}^*(\mathbf{K})$ .

Such a  $\text{Nogood}^*(\mathbf{K})$  is not empty. In fact,  $\text{Min}(\mathbf{K}) \in \text{Nogood}^*(\mathbf{K})$ . The following proposition shows that in order to compute  $\text{BC}(\mathbf{K})$ , it is enough to consider  $\text{Nogood}^*(\mathbf{K})$  instead of  $\text{Nogood}(\mathbf{K})$ .

**Proposition 5:**  $\text{BC}(\mathbf{K}) = \{ \text{OR}(X) \in \mathbf{K} \mid \text{OR}(X) > \text{Min}(\mathbf{K}) \}$ .

The following improvement is important for computing  $\text{Nogood}^*(\mathbf{K})$  more easily.

**Proposition 6:** Let  $A$  be a minimally inconsistent subbase, and let  $\text{OR}(X)$  be a belief of  $\mathbf{K}$  such that  $X \geq A$ ,  $Y \geq X$ . Then any minimally inconsistent subbase  $\mathbf{K}'$  in  $\text{Nogood}(\mathbf{K})$  such that  $\text{Min}(\mathbf{K}) \subseteq \mathbf{K}'$  is not in  $\text{Nogood}^*(\mathbf{K})$ .

Based on the previous propositions, the following algorithm presents a way to compute  $\text{Nogood}^*(\mathbf{K})$ . It is more efficient as it starts with a minimally inconsistent subbase made of the most important beliefs in  $\mathbf{K}$ .

Function: Computing  $\text{Nogood}^*(\phi)$

Input:  $\geq$ ,  $(\mathbf{K})$ ,

Begin

```

•  $\text{Nogood}^*(\phi) = \emptyset$ ;
• Let  $A$  a minimal inconsistent subbase of  $(\mathbf{K})$ 
While ( $A \neq \emptyset$ ) do Begin
  • Minimize  $A$  by removing  $\phi$  from  $(\mathbf{K})$ 
  •  $\exists \text{OR}(Y) \in A$ 
  • Refine  $A$ 
  s.t.  $(\text{OR}(Y) \in A, \geq)$ 
  •  $\text{Nogood}^*(\phi) = \text{Nogood}^*(\phi) \cup Y$ 
  • Minimize  $\text{Nogood}^*(\phi)$  by removing  $\phi$ 
  • Let  $B$  a minimal inconsistent subbase of  $(\mathbf{K})$ 
   $\text{Min}(B) = \text{Min}(A)$ 
end
Return ( $\text{Nogood}^*(\phi)$ )
end {Function}

```

Example 1 (continued) We assume that  $\phi_1 \vee \phi_4, \phi_2 \vee \phi_4, \phi_3 \vee \phi_4$  are consistent.

$\phi_1 \vee \phi_4 \equiv a \vee c$ ;  $\phi_2 \vee \phi_4 \equiv a \vee b$ ;  $\phi_3 \vee \phi_4 \equiv a \vee \neg b$ .  
 compute the consistent subbase  $\text{BC}(\mathbf{K})$  by first computing  $\text{Nogood}^*(\phi)$  using the previous algorithm. Let  $\phi_1 \vee \phi_4$  be a minimal inconsistent subbase.  $\phi_3 \vee \phi_4$  and  $\phi_2 \vee \phi_4$  are removed from  $(\mathbf{K})$  since  $\{s_1\} \geq \{s_4\}$  and  $\{s_1\} \geq \{s_3, s_4\}$ . At the end  $\text{Nogood}^*(\phi) = \{\phi_1 \vee \phi_4\}$  and  $(\mathbf{K}) = \{\phi_1 \vee \phi_4, \phi_2 \vee \phi_4, \phi_3 \vee \phi_4\}$ . Let us run again the previous algorithm. There is no minimal inconsistent subbase which is  $A = \{\phi_1 \vee \phi_4\}$ . However,  $A = \{\phi_1 \vee \phi_4, \phi_2 \vee \phi_4\}$  is considered since  $\{s_1, s_2\} \geq \{s_4\}$ .

$\text{Min}(\{\phi_1 \vee \phi_4, \phi_2 \vee \phi_4, \phi_3 \vee \phi_4\}) = \{\phi_1 \vee \phi_4\}$ .

Therefore:  $\text{Nogood}^*(\phi) = \{\phi_1 \vee \phi_4, \phi_2 \vee \phi_4\}$ .

Lastly the set of accepted beliefs is:

$$\text{BC}(\mathbf{K}) = \{\phi_1 \vee \phi_4, \phi_2 \vee \phi_4, \phi_3 \vee \phi_4\} \setminus \{a \vee c\}$$

and we can easily check that  $\forall \alpha, \mathbf{K}, \alpha \in \text{BC}(\mathbf{K})$ .

The proposed approach is incremental contrary to coherence based approaches, in the sense that when a new belief  $\phi$  provided by the source  $s_\phi$  is added to the original belief base  $\mathbf{K}$ , with a new set of constraints  $\phi$  is added to  $(\mathbf{K})$ , then  $(\mathbf{K} \cup \{\phi\})$  is derived from  $(\mathbf{K})$ .

is done by first computing  $\mathbf{K}^* = \text{BC}(\mathbf{K}) \diamond \{\phi\}$ . Then if  $\mathbf{K}^*$  is consistent then  $\text{BC}(\mathbf{K} \cup \{\phi\}) = \mathbf{K}^*$ , otherwise  $\text{BC}(\mathbf{K} \cup \{\phi\})$  contains accepted beliefs in  $\mathbf{K}^*$ .

### 7. Concluding remarks

Several well-known prioritized inconsistency handling methods have been unified into a powerful framework. A

flexible treatment of the priorities is provided by an explicit specification of the priority ordering through user-originated constraints and general postulates. Then a knowledge "compilation" technique enables us to explicit the formulas useful for the inference process, which amounts to select the consistent subpart of the new belief base obtained by compilation. This subpart is composed of the accepted formulas (i.e., those which escape from the minimally inconsistent subset of formulas of the belief base).

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