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# **Reasoning with Inconsistencies in Propositional Peer-to-Peer Inference Systems** <sup>1</sup>

**Ph. Chatalic** <sup>2</sup> and **G.H. Nguyen**<sup>3</sup> and **M.Ch. Rousset** <sup>3</sup>

#### **Abstract.**

In a peer-to-peer inference system, there is no centralized control or hierarchical organization: each peer is equivalent in functionality and cooperates with other peers in order to solve a collective reasoning task. Since peer theories model possibly different viewpoints, even if each local theory is consistent, the global theory may be inconsistent. We exhibit a distributed algorithm detecting inconsistencies in a fully decentralized setting. We provide a fully distributed reasoning algorithm, which computes only *well-founded* consequences of a formula, i.e., with a consistent set of support.

#### **1 Introduction**

Recently peer-to-peer (P2P) systems have received considerable attention because their underlying infrastructure is appropriate to scalable and flexible distributed applications over Internet. In P2P systems, there is no centralized control or hierarchical organization: each peer is equivalent in functionality and cooperates with other peers in order to solve a collective task. P2P systems have evolved from simple keyword-based file sharing systems like Napster [2] and Gnutella [1] to semantic data management systems like EDUTELLA [22], PIAZZA [18] or SOMEWHERE [5]. Reasoning in P2P Inference Systems (P2PIS) has been considered very recently (e.g., [3, 12, 4]). The dynamicity of P2PIS imposes to revisit many reasoning problems in order to address them in a decentralized manner. In particular, it is neither feasible to bring all the information to a single server and use standard reasoning algorithms, nor to compute its best partitioning for using partition-based reasoning like in [15, 8].

The local theories of the P2PIS considered in [3, 4] are sets of clauses defined upon sets of propositional variables (the local vocabularies of the peers). A new peer joins an existing P2PIS by establishing *mappings* with other peers, called its *acquaintances*. Mappings are clauses involving variables of distinct peers that state semantic correspondences between different vocabularies. The decentralized algorithm DECA [6] computes the consequences (in some target language) of an input formula w.r.t. the global theory of the P2PIS (i.e. the union of all peer theories). The point is to compute those consequences, without having access to the global theory. DECA is anytime, sound, and complete (under some conditions). It has been implemented in the SOMEWHERE platform and experiments on synthetic data have shown its scalability [4, 6].

This paper focuses on the problem of reasoning with inconsistencies in a P2PIS. Since peer theories model possibly different view-

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points, the global theory may be inconsistent even if each local theory is consistent. Given the lack of centralized control, all peers should be treated equally. It would be unfair to refuse the join of a new peer just because the resulting P2PIS becomes inconsistent. Our choice is to accept the presence of inconsistency. The problem is first to *detect* inconsistencies, second to *reason* in spite of them in a satisfactory way. We thus compute only *well-founded* consequences of a formula, i.e., consequences of the formula w.r.t. to a consistent subset of the global theory. Such an approach is not novel in the centralized case but raises new algorithmic issues in the decentralized case because the computation and the storage of *nogoods* (accounting for inconsistencies) are distributed. One has to be able to check the consistency of distributed sets of formulas, w.r.t. distributed sets of nogoods.

We assume each local theory to be consistent. Therefore, the possible inconsistencies result from interactions between local theories and are caused by *mappings*. Before adding a mapping, a peer checks whether this mapping (possibly with other mappings) can be the cause of some inconsistency, i.e., if the empty clause can be produced as one of its consequences. In that case, the peer stores locally as a *nogood* the set of mappings involved in the corresponding proof. At reasoning time, the concerned distributed nogoods must be collected to check whether the proof under construction is well-founded.

In Section 2, we formally define the considered P2PIS. In Section 3, we present a decentralized algorithm for detecting inconsistencies and computing the corresponding nogoods. Section 4 provides a decentralized reasoning algorithm which is well-founded despite possible inconsistencies. In Section 5, we conclude by relating our approach w.r.t. existing work.

#### **2 Peer-to-peer inference systems**

The P2PIS that we consider are networks of peer clausal theories  $\mathcal{P} = \{P_i\}_{i=1..n}$ , such that each peer has a proper *vocabulary* (denoted by  $\mathcal{V}_{P_i}$ ). We suppose that each peer has a unique identifier (for example, its IP address) and that variable names use in some way this identifier. For simplicity, we just use the index  $i$  as the peer identifier and denote a variable A of the peer  $P_i$  by  $A_i$ . Each local theory of a peer  $P_i$  is a set of clauses without duplicated literals. We denote by  $\mathcal{L}_{P_i}$  the language of clauses involving only variables of  $\mathcal{V}_{P_i}$ .

#### **Definition 1 (Mappings, shared variables and acquaintances)**

*A* mapping *is a clause of a peer involving at least a variable of the vocabulary of another peer. A variable of some peer is said to be* shared *if it appears in a mapping of another peer. The* acquaintances *of a peer are the peers in the network with which it shares variables.*

The *global theory*  $T(\mathcal{P})$  of a P2PIS  $\mathcal{P} = \{P_i\}_{i=1..n}$  is:  $\bigcup_{i=1..n} P_i$ . Since mappings play a special role in our approach, we distinguish  $\mathcal{M} = \bigcup_{i=1...n} M_i$ , where  $M_i$  is the set of mappings of  $P_i$  from

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 $\mathcal{O} = \bigcup_{i=1..n} O_i$ , where each  $O_i$  is the complementary of  $M_i$ . In addition, we assume that each mapping of  $P_i$  has a unique identifier prefixed by  $P_i$ .

A shared literal is a shared variable or its negation. For a clause  $c$ , we denote by  $S(c)$  the disjunction of its shared literals and by  $L(c)$  the disjunction of its other literals. We suppose that each peer  $P$  knows its acquaintances, and given a shared literal l we denote by  $ACQ(l, P)$ the set of peers with which  $P$  shares the variable of  $l$ .

In contrast with other approaches [17, 12], we do not adopt an epistemic or modal semantics for interpreting a P2PIS but we interpret it with the standard semantics of propositional logic. In particular, a variable shared by two peer theories of a given P2PIS is interpreted by the same value in the two peers.

**Definition 2 (Semantics of a P2PIS)** *Let*  $P = \{P_i\}_{i=1..n}$  *be a P2PIS, an* interpretation I *of* P *is an assignment of the variables of*  $\mathcal{T}(\mathcal{P})$  *to* true *or* false. *I is a* model of a clause *c iff one of the literals of* c *is evaluated to* true *in* I*.* I *is a* model of a set of clauses *iff it is a model of all the clauses of the set.*

*-* P *is* consistent *iff* T (P) *has a model.*

*- The* consequence relation *is the standard one:*  $P \models c$  *iff every model of* P *is a model of* c*. We say that* c *is an implicate of* P*.*

#### **Example**

Let us consider the P2PIS corresponding to Figure 1.  $P_1$  can be asked by researchers for choosing where to submit their results (demos or papers). For instance, part of its knowledge can model that: PODS06 is open for submission; submitting to PODS06 entails submitting to PODS; only theoretical results are submitted to  $PODS$ ; a demo cannot be submitted to JAIR.  $P_2$  distinguishes proceedings from journals and knows that: submitting to PODS entails submitting to a conference with proceedings; submitting to JAIR entails submitting to a journal; a same result cannot be submitted in a conference and in a journal; patented results cannot be submitted to a journal. P<sup>3</sup> has some knowledge about research valorization policy: software should be patented or presented as demos; theoretical results should be submitted to journals. The knowledge expressed separately by  $P_1$ ,  $P_2$  and  $P_3$  using their respective vocabularies can be respectively modelled by the set of clauses  $O_1$ ,  $O_2$  and  $O_3$ .

The set  $M_2$  of mappings stored at  $P_2$  states the equivalence between  $PODS_1$  and  $PODS_2$  (reps.  $JAIR_1$  and  $JAIR_2$ ) through the mappings identified by  $P_2.1$ ,  $P_2.2$ ,  $P_2.3$  and  $P_2.4$ . The set  $M_3$ of mappings stored at  $P_3$  also establishes equivalences between variables of  $P_3$  and variables of the two other peers. Note however that mappings clauses do not necessarily result from equivalences.

#### **The reasoning problem**

For each peer  $P_i$ , we consider a set  $\mathcal{TV}_{P_i} \subseteq \mathcal{V}_{P_i}$  of *target variables*, supposed to represent the variables of interest for the application, (e.g., observable facts in a model-based diagnosis application, or classes storing data in an information integration application). For a set SP of peers of a P2PIS, we define its *target language*  $Target(SP)$  as the language of clauses (including the empty clause) involving only variables of  $\bigcup_{P \in SP} \mathcal{TV}_P$ .

#### **Definition 3 (P2PIS consequence finding problem)**

*Given a P2PIS* P, a peer P of P and a clause  $q \in \mathcal{L}_P$ , the P2PIS consequence finding problem *is to find the set of* proper prime implicates *of* q *w.r.t.*  $T(P)$  *that belong to*  $Target(P)$ *.* 



 $Journal_2, Patent_2$ 

**Figure 1.** Example P2PIS network (edges labelled by shared variables).

A *proper prime implicate* of  $q$  w.r.t. a (distributed) theory  $T$  is a prime implicate of  $T \cup \{q\}$ , which is not a prime implicate of T.

DECA [3, 6] is the first fully decentralized algorithm being able to solve this problem without having a global view of the system. Details and illustrations of DECA behavior may be found in [6]. Roughly summarized, when running at a peer  $P$  and asked to compute the proper prime implicates of a literal  $q$ , it first computes all proper prime implicates of this literal w.r.t. the local theory of  $P$ , selects those of interest w.r.t. the target language and then split each clause c obtained in this way in two subclauses :  $L(c)$  and  $S(c)$ .  $S(c)$  is in turn splitted and for each shared literal l of  $S(c)$  DECA asks its appropriate acquaintances (which are running the very same algorithm) to further propagate  $l$  in the P2PIS and to return the corresponding results. As soon as consequences are obtained for all literals of  $S(c)$  they are recombined together, as well as with  $L(c)$ , to produce consequences of the splitted clause c. Different branches of reasoning are thus developed throughout the network, following the shared variables, and thus the mappings. In order to ensure termination in presence of cycles as well as to handle the transmission of the results back to the appropriate peers, an *history* is associated to each propagated literal and updated each time that the corresponding reasoning branch goes out from one peer into another peer.

#### **3 P2P detecting inconsistencies and nogoods**

As outlined in the introduction, even if each  $P_i$  is locally consistent, this is not necessarily the case for  $T(P)$ . We assume that the causes of inconsistencies are only due to mappings and define a **nogood** ng as a set of mappings such that  $\mathcal{O} \cup nq$  is inconsistent. Note that in a nogood  $nq$  there necessarily exists some mapping  $m$  from which one can derive the empty clause (i.e., with a proof rooted in  $m$ ). We exploit this property in the decentralized algorithm P2P-NG to detect nogoods accounting for inconsistencies. P2P-NG runs at each peer and is used before adding a new mapping  $m$  to a peer  $P$ , to check whether the propagation of  $m$  into the existing P2PIS produces the empty clause. If that is the case,  $P$  stores locally as new nogoods the mappings (including  $m$ ) involved in the corresponding derivations. We call the set of mappings used in a derivation, its **mapping support**. The P2P-NG algorithm computes the (possibly empty) set of mapping supports of the derivations of the empty clause rooted in

 $m$ , starting at the peer  $P$ . Since mappings have a unique identifier, mapping supports can efficiently be encoded as sets of mapping identifiers (and similarly for nogoods computed from mapping supports).

P2P-NG is an adaptation of the DECA consequence finding algorithm. It follows the same split-recombine strategy but it has the following significant differences:

- The stopping conditions **must** be changed. The ones in the original DECA algorithm were designed for the computation of *proper prime* implicates only. Here we need to find all the possible ways of deriving the empty clause, we cannot stop the reasoning as soon as we produce the empty clause, or as soon as we find a unit clause in a local peer which is the same as the literal under processing.

- Because we are only looking for  $\Box$  as a consequence, locally produced consequences c such that  $L(c) \neq \square$  can be filtered out.

- While DECA would return  $\{\Box\}$  as its result if there exists a derivation of the empty clause, P2P-NG returns as many mapping supports as different ways of deriving  $\Box$ .

We use the following notations:

- For a clause c,  $Resolvent+SMS(c, P)$  computes the couples of pairs  $(r \, SMS(r, P))$  such that r is a local consequence of c w.r.t. P and  $SMS(r, P)$  is its corresponding set of local mapping supports. - For a literal  $q$ ,  $\bar{q}$  denotes its complementary literal.

- A history hist is a sequence of tuples  $(l, P, c)$  where l is a literal,  $P$  a peer, and  $c$  a clause which is a consequence of  $l$  on the peer *P*. A history  $[(l_n, P_n, c_n), \ldots, (l_1, P_1, c_1), (l_0, P_0, c_0)]$  represents a branch of reasoning initiated by the propagation of the literal  $l_0$ within the peer  $P_0$ , which either has produced locally the clause  $c_0$ in  $P_0$  (in that case,  $c_0$  may have been splitted into its different literals among which  $l_1$  is propagated in  $P_1$ ), or not (in that case  $l_0$  is simply propagated from  $P_0$  to  $P_1$  and  $l_0 = c_0 = l_1$ ). For every  $i \in [0..n-1]$ ,  $c_i$  is a consequence of  $l_i$  and  $P_i$ , and  $l_{i+1}$  is a literal of  $c_i$ , which is propagated in  $P_{i+1}$ .

- ⊗ is the distribution union operator on sets of sets:

 $SS_1 \otimes \cdots \otimes SS_n = \{S_1 \cup \cdots \cup S_n | S_1 \in SS_1, \ldots, S_n \in SS_n\}.$  If  $L = \{l_1, \ldots, l_p\}, \otimes_{l \in L} SS_l$  denotes  $SS_{l_1} \otimes \cdots \otimes SS_{l_p}$ .

Each literal resulting from the splitting of a clause (Line (4) of the P2P-NG algorithm) is processed independently by the P2P- $NG_l$  algorithm. P2P-N $G_l(q, SP, hist)$  checks whether there exists a derivation of  $\Box$  rooted in the literal q, starting with the computation of local consequences of  $q$ , and then recursively following the acquaintances of the visited peers. To ensure termination, it is necessary to keep track of the literals already processed by peers. This is done thanks to  $hist$ , where  $hist$  is the history of the reasoning branch ending up to the propagation of the literal  $q$  in  $SP$ , which is the set of acquaintances of the last peer added to the history.

Theorem 1 states that the result of  $P2P-NG(m, P)$  can be used to characterize nogoods involving the mapping  $m$  (by assimilating mappings to their corresponding index).

**Theorem 1** *Let* m *be a mapping and* P *be a peer such that* P2P- $NG(m, P) \neq \emptyset$ .  $\forall ms \in P2P\text{-}NG(m, P), ms \cup \{m\}$  *is a nogood.* 

Before adding a new mapping  $m$  to its local theory, each peer  $P$  first computes P2P-NG $(m, P)$  and stores locally all the nogoods  $\{m\}$ ms, such that  $ms \in P2P-NG(m, P)$  is minimal (for inclusion).

Theorem 2 states that the P2P-NG algorithm is complete, i.e., it enables to find the mapping supports of all the irredundant derivations of the empty clause from a given mapping added to a P2PIS. The proof of Theorem 2 relies on Lemma 1.

**Definition 4 (Irredundant derivation)** *A derivation is irredundant if it does not involve two identical applications of the resolution rule.* **Algorithm 1:** Detection of the nogoods caused by adding a mapping  $P2P-NG(m, P)$ 

- (1)  $\text{LOCAL}(P) \leftarrow \text{Resolvent} + \text{SMS}(m, P)$
- (2) RESULT  $\leftarrow \emptyset$
- (3) **foreach**  $(c \text{ sms}) \in \text{LOCAL}(P) \text{ s.t. } S(c) \neq \square \text{ and } L(c) = \square$
- (4) **foreach** literal  $q \in S(c)$
- (5)  $NOGOODS(q) \leftarrow P2P-NG_l(q, ACQ(q, P), \emptyset)$
- (6) **if** for every  $q \in S(c)$  NOGOODS $(q) \neq \emptyset$
- (7) UNIONCOMB ←  $sms \otimes (\otimes_{a \in S(c)} \text{NOGOOD}(q))$
- (8) RESULT ← RESULT  $\cup$  UNIONCOMB (9) **return** RESULT

 $P2P-NG<sub>l</sub>(q, SP, hist)$ 

- (1) **if** for every  $P \in SP$ ,  $(q, P, \_) \in hist$
- (2) **return** ∅
- (3) **else**
- (4)  $SMS(q) \leftarrow \{\emptyset\}$
- (5) RESULT  $\leftarrow \emptyset$
- (6) **if**  $(\bar{q}, \ldots) \in hist$
- (7) RESULT ← RESULT  $\cup$  { $\emptyset$ }
- (8) **foreach**  $P \in SP$
- (9)  $\text{LOCAL}(P) \leftarrow \{(q \{\emptyset\})\} \cup Resolvent + SMS(q, P)$
- (10) RESULT ← RESULT  $\cup \bigcup_{P \in SP} SMS(\square, P)$
- (11) **foreach**  $P \in SP$  and  $(c \overline{sms}) \in \text{LOCAL}(P)$  s.t.  $S(c) \neq \Box$
- and  $L(c) = \Box$ (12) **foreach** literal  $l \in S(c)$
- (13)  $\text{SMS}(l) \leftarrow \text{P2P-NG}_l(l, \text{ACO}(l, P), [(q, P, c)|hist])$

$$
SMS(t) \leftarrow P2P-NQ_l(t, ACQ(t, F), [(q, F, c)]ntsl)
$$

- (14) **if** for every  $l \in S(c)$ ,  $\text{SMS}(l) \neq \emptyset$
- (15) UNIONCOMB ←  $sms \otimes (\otimes_{l \in S(c)} \text{SMS}(l))$
- (16) RESULT ← RESULT  $\cup$  UNIONCOMB

(17) **return** RESULT

**Theorem 2** *Let* P *be a P2PIS and* m *a mapping of a given peer* P *of* P*. Let* ms *be a mapping support of an irredundant derivation of*  $\Box$  *rooted in m. It will be returned by* P2P-NG $(m, P)$ *.* 

**Lemma 1** *Let* ms *be the mapping support of an irredundant derivation of*  $\Box$  *rooted in a clause c:*  $c_1 \lor ... \lor c_n$  *where every*  $c_i$  *is a clause (which can be a unit clause or not) such that there is no literal common to*  $c_i$  *and*  $c_j$  *for*  $i \neq j$ *. There exists*  $ms_1, \ldots, ms_n$  *where, for every i*,  $ms_i$  *is a mapping support of an irredundant derivation of*  $\Box$ *rooted in*  $c_i$ *, such that:*  $ms = ms_1 \cup ... \cup ms_n$ *.* 

Corollary 1 results directly from Theorem 2. It guarantees that all the minimal nogoods are computed ans stored in the P2PIS. It is the key for proving that the reasoning algorithm presented in the next section is well-founded.

**Corollary 1** *Let*  $T(P) = O \cup M$  *be the global theory of a P2PIS and let* S *be a set of mappings of* M*. If* S *is a minimal nogood then it is stored at some peer* P *of* P*.*

**Example (cont.):** In the example of Section 2, let us suppose that the different peers join in the following order:  $P_1$ , then  $P_2$ , then  $P_3$  and that their respective mappings are added according to their numbering. At the join of  $P_2$ , the successive adding of the 4 mappings causes no inconsistency. When  $P_3$  joins, the 4 first mappings cause no inconsistency. Let us focus on the 5th one. P2P- $NG(\neg Journal_3 \lor Journal_2, P_3)$  is triggered where  $P_3$  is the theory containing the clauses of  $P_3$  that are not mappings, and the 4 first mappings (which have been added since they have been checked as not deriving inconsistencies).  $\neg Theory_1 \lor Journal_2$  is produced

locally at  $P_3$  as a local consequence of  $\neg Journal_3 \lor Journal_2$ , with a set of local mapping support equal to  $\{\{P_3, P_4\} \}$ . Then, ¬Theory1∨Journal<sup>2</sup> is splitted (Line (4) of P2P-NG):¬Theory<sup>1</sup> is processed by  $P_1$ , while  $Journal_2$  is processed by  $P_2$ . The propagation of  $\neg Theory_1$  produces  $\Box$  as a local consequence in  $P_1$  with a local set of mapping support equal to  $\{\emptyset\}$ . Thus  $\{\emptyset\}$  is returned to  $P_3$  as the set of mapping supports of the derivation of  $\Box$  from Theory<sub>1</sub>. The propagation of  $Journal_2$  produces  $\neg PODS_1$  as a local consequence in  $P_2$ , with a local set of mapping supports equal to  $\{\{P_2,1\}\}\$ , as well as  $\neg Patent_2$  with a local set of mapping support equal to  $\{\emptyset\}$ .  $\neg PODS_1$  is in turn propagated in  $P_1$ , where it produces  $\Box$  as a local consequence with a local set of mapping supports equal to  $\{\emptyset\}$ . It is transmitted back to  $P_2$  which, after combination of  $\{\emptyset\}$  and  $\{\{P_2,1\}\}\$  (Line 15) of P2P-NG<sub>l</sub>( $Journal_2, \{P_2\}, \emptyset$ )), transmits back to  $P_3$  the set of mapping supports  $\{\{P_2,1\}\}\)$  for the derivation  $\Box$  from  $Journal_2$ . By combination (Line 7) P2P- $NG(\neg Journal_3 \lor Journal_2, P_3)$  returns  $\{\{P_3.2, P_2.1\}\}\.$  The nogood  $\{P_3, 5, P_3, 2, P_2, 1\}$  is thus obtained and stored at  $P_3$ . No other nogood is obtained from the last mappings of  $P_3$ .

#### **4 Peer-to-peer well-founded reasoning**

First, we have to define the notion of *well-founded* consequences that can be derived from a given input clause and a possibly inconsistent P2PIS. Many semantics have been proposed and studied for reasoning with (centralized) inconsistent theories ([14] for a survey). We adopt the following one (which is one of the simplest ones), because it makes sense in a decentralized and dynamic setting.

**Definition 5 (P2P well-founded implicate)** *Let* P *be an inconsistent* P2PIS*:* r *is a* well-founded *implicate of* c *w.r.t.* P *if* r *is an implicate of c w.r.t. a* consistent subset *of*  $T(\mathcal{P})$ *.* 

The WF-DECA $(q, P)$  algorithm computes well-founded consequences of the (unit) clause  $q$ , starting at the peer  $P$ . This algorithm extends the original DECA consequence finding algorithm [3, 6] by computing the set of mapping supports of the derivations for each consequence, and by collecting the nogoods encountered during the reasoning. Because of the split/recombination technique used by the algorithm, mapping supports of derivations are only known after the recombination step, and the set of possibly relevant nogoods must be available at this step: if some mapping support includes a nogood, it is discarded; consequences that get an empty set of mapping supports after nogoods filtering are discarded as well.

We use the following notations :

-  $LocalConsSSNG(q, P)$  is a local procedure that computes the set of triples  $(c \, sms \, sng)$  such that c is a local consequence of q w.r.t. P, sms is its corresponding set of local mapping supports, and sng is the set of nogoods stored at the peer  $P$  that contain a mapping  $m$  of some mapping support  $ms$  of  $c$ .

-⊎ denotes the *merged union* of sets of consequences, i.e. the union of sets of triples of the form  $(c \, sms \, sng)$ , where triples corresponding to a same consequence  $c$  are merged together, by computing the union of their respective sms and sng.

 $\oslash$  is the distribution union operator on sets of triples of the form  $(c \text{ } sms \text{ } sng): S_1 \otimes \cdots \otimes S_n = \{(c_1 \vee \cdots \vee$  $c_n$  sms<sub>1</sub> ⊗ · · · ⊗ sms<sub>n</sub> sng<sub>1</sub> ∪ · · · ∪ sng<sub>n</sub>)/(c<sub>1</sub> sms<sub>1</sub> sng<sub>1</sub>) ∈  $S_1, \ldots, (c_n \, sms_n \, sing_n) \in S_n$ .

Theorem 3 states that the WF-DECA $(q, P)$  algorithm terminates and returns only well-founded consequences. Its proof relies on showing that each triple  $(c \, sms \, sng)$  returned by WF- **Algorithm 2:** Well Founded Distributed Consequence Finding Algorithm

 $WF-DECA(q, P)$ 

(1) WF-DECAH $(q, {P}, \emptyset)$ 

 $WF-DECAH(q,SP, hist)$ 

- (1) **if** for every  $P \in SP$ ,  $(q, P, \_) \in hist$
- (2) **return** ∅
- (3) **else if**  $(\bar{q}, \, \square, \, \square) \in hist$
- (4) **return**  $\{(\Box \{\emptyset\}\emptyset)\}$
- (5) **else**
- (6) RESULT  $\leftarrow \emptyset$
- (7) **foreach**  $P \in SP$
- (8)  $\text{LOCAL}(P) \leftarrow \{(q \{\emptyset\} \emptyset)\} \oplus LocalConsSSNG(q, P)$
- (9) **foreach**  $P \in SP$  and  $(c \, sms \, sng) \in \text{LOCAL}(P)$  such that  $L(c) \in Target(P)$
- (10) **if**  $S(c) \in Target(P)$
- (11) RESULT ← RESULT  $\forall$  {(c sms snq)}
- $(12)$  **if**  $S(c) \neq \Box$
- (13) **foreach** literal  $l \in S(c)$
- $(14)$  ANSWER $(l)$  ←
- (15) WF-DECAH $(l, ACQ(q, P), [(q, P, c)]hist]$ 
	-
- (16) **if** for every  $l \in S(c)$  ANSWER(*l*)  $\neq \emptyset$ (17) UNIONCOMB  $\leftarrow$
- (18)  $\{ (L(c) \text{ sms} \text{ sng}) \} \otimes (\otimes_{l \in S(c)} \text{ANSWER}(l))$ <br>(19) **foreach**  $(c \text{ sms} \text{ sna}) \in \text{UNIONCOMP}$ foreach  $(c \, sms \, sng) \in UNIONCOMB$ (20)  $nsms \leftarrow \{ms \in sms/\forall ng \in sing, ng \not\subseteq ms\}$ (21) **if**  $nsms \neq \emptyset$
- (22) RESULT ← RESULT ⊎  $\{(c \text{ nsms snq})\}$

(23) **return** RESULT

 $DECA(q, SP, hist)$  is such that either c is a local consequence of q w.r.t. some peer  $P \in SP$ , or for every  $ms \in sms$ ,  $\mathcal{O} \cup ms$  is consistent (using Corollary 1) and  $c$  is the result of a derivation rooted in q that only uses clauses from  $\mathcal{O} \cup \{l_0, l_1, \ldots, l_n, q\} \cup ms$ , where  $l_0, l_1, \ldots, l_n$  are the literals in hist.

**Theorem 3** *Let* P *be a peer of a P2PIS* P *and* q *a literal belonging to the vocabulary of* P*.* WF-DECA(q,P) *terminates and for* all triples  $(c \, sms \, sng)$  returned by  $WF\text{-}\text{DECA}(q, P)$ ,  $c \, is \, a \, well$ *founded consequence of* q *w.r.t.* P*.*

**Example (cont.):** Let us illustrate the behaviour of WF- $DECA(Soft<sub>3</sub>, P<sub>3</sub>)$ , assuming that the only target variables are  $PODS<sub>1</sub>$  and  $JAIR<sub>1</sub>$ .  $Patent<sub>2</sub> \vee Demo<sub>1</sub>$  is the only clause produced locally on  $P_3$  with a local part of which (i.e.  $\Box$ ) in  $Target(P_3)$ . Its local sms is  $\{\{P_3.7, P_3.3\}\}\.$  The only nogood stored at  $P_3$  contains neither  $P_3.7$  nor  $P_3.3$ . The corresponding sng returned by  $LocalConsSSNG(Soft, P_3)$  is thus empty.  $Patent_2 \vee Demo_1$  is then splitted. When  $Patent_2$  is transmitted to  $P_2$ ,  $\neg JAIR_1$  is the only clause produced locally with a local part (i.e.  $\Box$ ) in  $Target(P_2)$ . Its local sms is  $\{\{P_2.3\}\}\$ and its sng is empty. The further propagation of  $\neg JAIR_1$  returns an empty result. So the triple  $(\neg JAIR_1\{\{P_2.3\}\}\emptyset)$  is sent back to  $P_3$  as a consequence of  $Patent_2$ . When  $Demo_1$  is transmitted to  $P_1$  the only clause produced locally is  $\neg JAIR_1$ , which is in  $Target(P_1)$ . Its local sms is empty, as well as its *sng*. So the triple  $(\neg JAIR_1{\{\emptyset\}\emptyset})$  is sent back to  $P_3$  as a consequence of  $Demo_1$ .  $P_3$  then combines these two triples obtained from  $P_2$  and  $P_1$  giving  $(\neg JAIR_1 \{ \{P2.3\} \} \emptyset)$ , which is the only final consequence of  $Soft_3$  being in the target language. Since the corresponding sng is empty, it is trivially a wellfounded consequence.

#### **5 Conclusion and Related Work**

We have presented a fully decentralized approach for reasoning with inconsistencies in propositional P2PIS. Nogoods are discovered each time a new mapping is added. Though these are stored in a completely distributed way, the WF-DECA algorithm guarantees that all consequences it returns are well-founded. For lack of place, many optimization details have been omitted. In particular, calculating *all* possible mapping supports of each consequence is not necessary. Computing only the minimal ones is sufficient. The implementation of this approach is currently under process. An extensive experimental study in the spirit of that of [4, 6] is planned for the near future.

For efficiency reasons our approach for P2P reasoning with more powerful languages is to approximate the peer theories and the queries into propositional logic and to exploit the current infrastructure to compute the corresponding propositional answers. The actual answers are then computed by focusing on the relevant peers.

Reasoning under inconsistency has been widely studied in artificial intelligence, but mainly in the centralized case. There are two kinds of strategies for dealing with inconsistent knowledge bases. The first one is to restore consistency as in belief revision [7, 16]. In the context of peer-to-peer interconnected databases, although not strictly restoring consistency, the work of [9] characterizes consistent answers as those that can be drawn from all minimal repairs of all peer databases and that satisfy as well a set of data exchange constraints (similar to mappings) relating the different peer schemes.

The alternative is just to tolerate inconsistency [10]. In a fully decentralized setting, all peers playing the same role, this seems far more preferable. A wide range of paraconsistent logics have been proposed [20] to avoid the trivialization problem of classical logic. Our approach is closer to coherence systems and argumentative approaches [14], since we consider consequences that can be produced from consistent subsets of  $\mathcal{T}(\mathcal{P})$ . Since  $\mathcal O$  is known to be consistent, mapping supports can be considered as justifications supporting the consequences. Our approach is a credulous one, since any subset of  $M$  consistent with  $O$  is considered. Well-founded consequences can also be viewed as either local consequences or as the formulas of some extension of the supernormal default theory [11]  $\Delta = (0, M)$ .

The work of [19] on reasoning from inconsistent ontologies is another kind of coherence-based approach. For a given query, it aims at finding a specific consistent subset of the global theory, in which the query (or its negation) classically holds. If it is not possible, the answer is undetermined. The set is constructed by successive consistent expansions, according to some selection function measuring some relevance criterion with the query. They use a syntactical criteria, that selects only formulas sharing some variable with those already selected during the previous iterations. Limiting the choice for the consistent subset clearly reduces the set of accepted consequences.

An epistemic semantics has been proposed in [13] to deal with possible inconsistencies in a P2P Data Inference Systems formalized in a first order multi modal language. It considers the case of local inconsistency (a point not addressed by our approach) as well as global inconsistency. Mapping are formalised in such a way that they cannot be used to propagate information from some locally inconsistent theory. Moreover mappings can only be used to propagate information to a peer, as far as they do not contradict either local information or other non-local information that may be deduced on that peer.

Distributed CSP techniques [24, 23] aim at finding consistent assignments of a set of variables, satisfying a set of distributed constraints. They also propagate nogoods corresponding to invalid partial affectations, that are further stored on the peer that receives them. This tends to replicate among all agents some global knowledge. This is also the case in distributed ATMS [21]. In comparison, we exploit nogoods only at reasoning time. Only those that may interfere with some mapping support of a related *sms* are transmitted in the *sng*.

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