

Distributed Reasoning with Conflicts in an Ambient Peer-to-Peer Setting



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Artificial Intelligence Methods for Ambient Intelligence

Aim of this Study

- An algorithm for reasoning with distributed rule theories in an ambient setting
- The algorithm
 - models the participating agents as nodes in a P2P system
 - takes into account some special characteristics of context knowledge and ambient agents
 - considers the potential conflicts that may arise during the integration of the distributed knowledge

Talk Outline

- Context Knowledge and Ambient Agents
 - Notion of Context
 - Special Characteristics of Ambient Agents
 - Challenges of Reasoning in an Ambient Setting
- Related Work
- Algorithm Description
 - General Approach
 - Problem Statement - Definitions
 - Steps of the *P2P_DR* Algorithm
 - The Algorithm in Action
 - Algorithm Properties
- Future Work

Notion of Context

- Context can be described as

".. any information that can be used to characterize the situation of an entity. An entity is a person, place or object that is considered relevant to the interaction between a user and an application, including the user and application themselves.." [Dey and Abowd, 1999]

- Context Characteristics
 - Multiple heterogeneous formats
 - Dynamic
 - Unknown
 - Ambiguous
 - Imprecise
 - Erroneous

Ambient Agents

- Diverse goals, experiences and perceptive capabilities
- Distinct vocabularies
- Different levels of sociality
- Dynamic behavior
 - Nodes join and leave the system randomly
 - An ambient agent is not able to know a priori all other entities that are present at a specific time instance
 - It cannot communicate directly with all other ambient agents

Challenges

- Reasoning with the highly dynamic and ambiguous context
- Managing the potentially huge piece of context data, in a real-time fashion, taking into account the restricted computational, storage and communication capabilities of some mobile devices
- Collective intelligence, by supporting information sharing, and distributed reasoning between the entities of the ambient environment.

Centralized reasoning is not a good solution, as it is too costly and cannot handle the system dynamics

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Reasoning in PDMS (1/2)

- First Order Logic Interpretations of P2Ps
 - proposed by [*Bernstein et al.,2002; Halevy et al.,2003*]
 - problems regarding **modularity**, **generality** and **decidability**
- Semantics based on Epistemic Logic
 - proposed by [*Calvanese et al.,2004*]
 - does not deal with **inconsistencies**
- Autoepistemic Semantics
 - proposed by [*Franconi et al.,2003*]
 - formalizes **local inconsistency**
 - guarantees that a locally inconsistent database base will not render the entire knowledge base inconsistent

Reasoning in PDMS (2/2)

- Non-monotonic Epistemic Logic Semantics
 - proposed by [Calvanese et al.,2005]
 - enables isolating local inconsistency
 - handles peers that provide mutually inconsistent data
- Propositional P2P Inference System
 - proposed by [Chatalic et al.,2006]
 - detects mutually inconsistent data and reasons without them
- Common Deficiencies
 - Conflicts are not actually resolved; they are rather isolated
 - Trust is not part of the model
 - In most cases, the participating peers share a common alphabet

Non-monotonic Reasoning in MCS

- Rule-based MCS with Default Negation
 - proposed by [Roelofsen & Serafini, 2005]
- Contextual Default Reasoning
 - proposed by [Brewka et al.,2007]
 - models bridging rules between different contexts as default rules
 - closer to implementation due to the well-studied relation between Default Logic and Logic Programming
 - does not provide reasoning algorithms, leaving some practical issues unanswered

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Our Approach

- A P2P rule-based model that captures
 - local knowledge
 - bridging rules
 - trust
- Why P2P?
 - Each different peer independently collects and processes in its own way the available context information.
 - Each peer may not have (immediate) access to all information sources.
 - The peers share their knowledge through messages with their neighboring nodes.
 - Each peer may not trust all the other peers at the same level.
 - Peers join and leave the system randomly.

Definitions

- We assume a peer-to-peer system P as a collection of peer local theories

$$P = \{P_i, i=1,2,\dots,n\}$$

- Each peer has a proper distinct vocabulary V_{P_i} and a unique identifier i .
- Each local theory is a set of rules that contain only local literals

$$r_i : a_i, b_i, \dots, k_i \rightarrow x_i$$

- Each peer also defines mappings that associate local literals with literals from the vocabulary of other peers (remote literals):

$$m_i : a_i, b_j, \dots, z_k \rightarrow x$$

Problem Statement

- *Given a peer-to-peer system P and a query about a literal x_i issued at peer P_i , find the truth value of x_i considering P_i 's local theory, its mappings and the theories of other system nodes.*

- We assume that the local theories are consistent,
- ..but this is not necessarily true for the unification of the system peer theories (local rules and mappings).
- The inconsistencies result from interactions between local theories and are caused by mappings.

P_1

$$r_{11} : a_1 \rightarrow x_1$$

$$m_{11} : a_2 \rightarrow a_1$$

$$m_{12} : a_3 \rightarrow \neg a_1$$

P_2

$$r_{21} : \rightarrow a_2$$

P_3

$$r_{31} : \rightarrow a_3$$

Steps of the *P2P_DR* Algorithm (1/3)

- **Step 1**
 - Use P_i 's local theory to prove x_i .
- **Step 2**
 - Collect P_i 's local and mapping rules that support x_i .
 - For each such rule, check the truth value of the literals in its body. For each local / remote literal, issue similar queries (*recursive calls of the algorithm*) to P_i (local literals) or to the appropriate P_i 's acquaintances (remote literals).
 - To avoid *circles*, before each new call, check if the same query has been issued before during the same call of the algorithm.
 - Build the *supportive set of x_i* ; this contains the 'strongest' set of mapping rules (defined either locally or remotely) that can be used to prove x_i in the absence of contradictions.

Steps of the *P2P_DR* Algorithm (2/3)

- **Step 3**
 - Collect P_i 's local and mapping rules that support $\neg x_i$ (contradict x_i).
 - In the same way with Step 2, build the *supportive set of $\neg x_i$* (*conflicting set of x_i*)
- **Step 4**
 - Compare the *supportive* with the *conflicting* set of x_i .
 - If the *supportive set* is stronger set than the *conflicting set*, return *Yes* and terminate. Otherwise, return *No* and terminate.

Steps of the *P2P_DR* Algorithm (3/3)

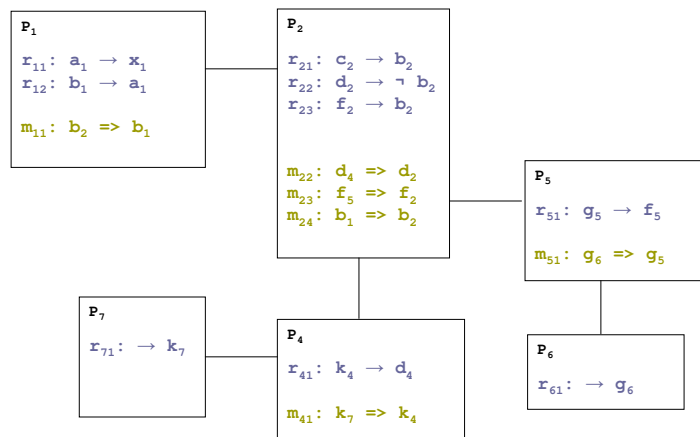
- How to compare two mapping sets
 - Each peer defines an *order* of the system peers, based on the *trust* it has on each one of them. According to this ordering, for two mapping rules, m_k and m_l *m_k is stronger than m_l from P_i 's viewpoint if P_k precedes P_l in P_i 's order*
 - The strength of a mapping set is determined by the strength of the weakest rule in this set.

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Assume that we issue a query about x_1 to P_1

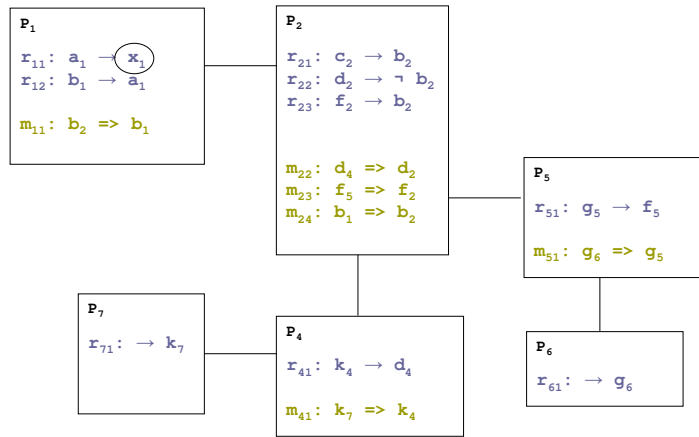


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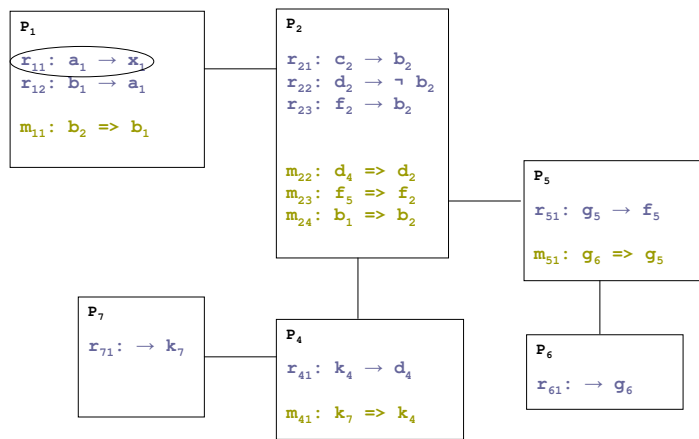
Neither x_1 nor $\neg x_1$ derive from P_1 's local theory



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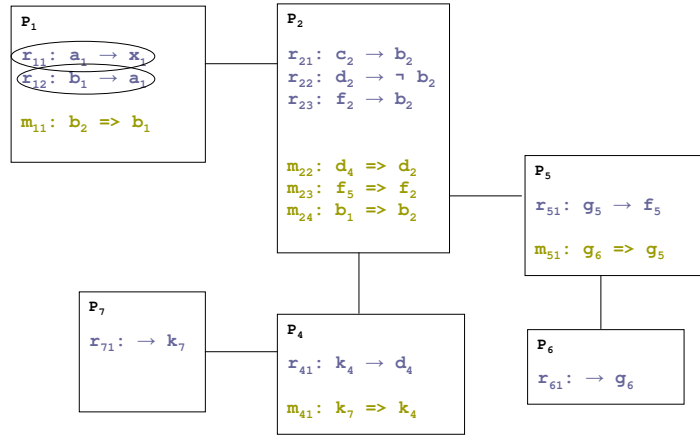
r_{11} is a supportive rule for x_1 , which has a_1 as its only premise.



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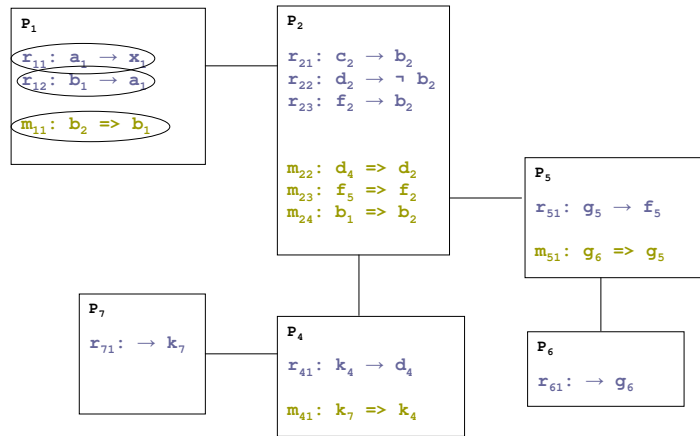
r_{12} is a supportive rule for a_1 , which has b_1 as its only premise.



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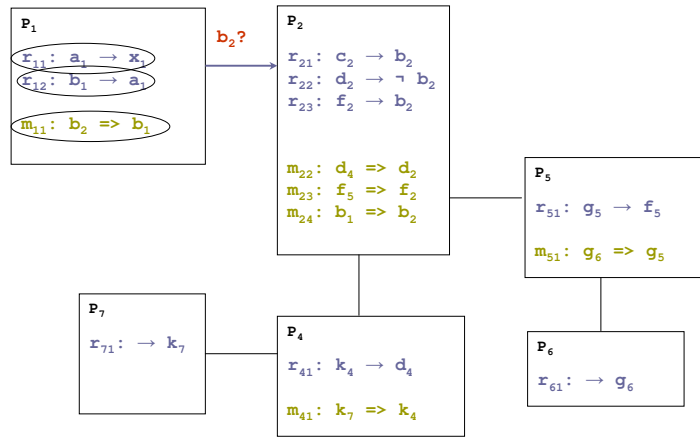
m_{11} is a supportive rule for b_1 , which has b_2 as its only premise



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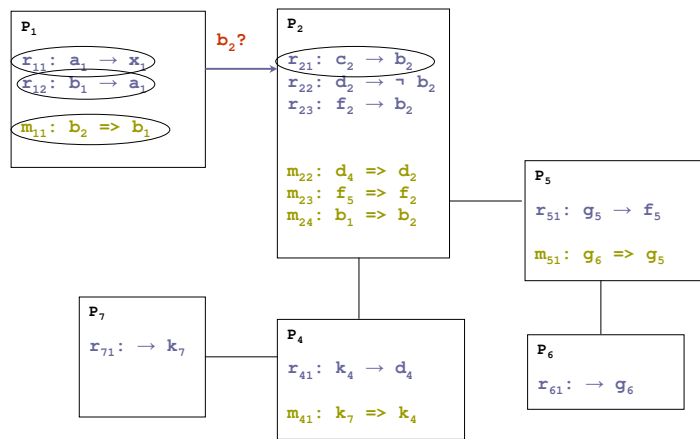
b_2 belongs to P_2 's published theory, so P_1 queries P_2 about b_2



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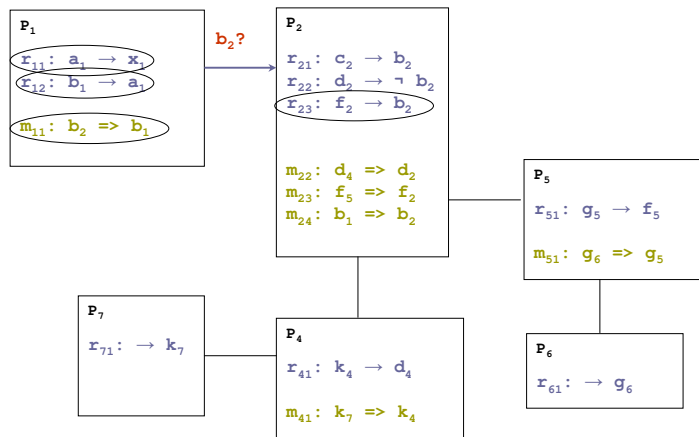
r_{21} is a supportive rule for b_2 , which has c_2 as its only premise. There is no supportive rule for c_2 , so r_{21} cannot be used to prove b_2



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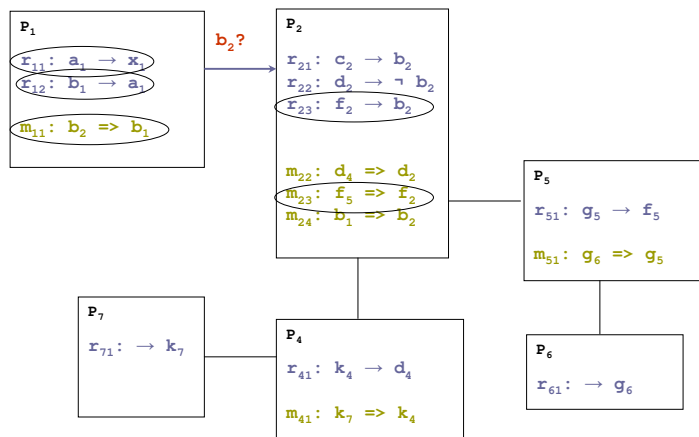
r_{23} is another supportive rule for b_2 , which has f_2 as its only premise.



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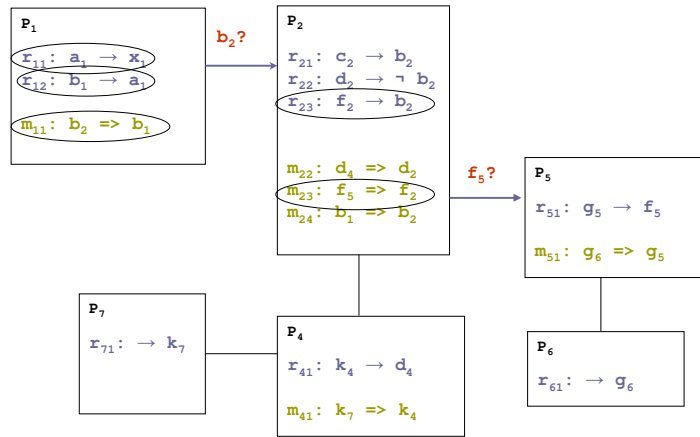
m_{23} is a supportive rule for f_2 , which has f_5 as its only premise.



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f_5 belongs to P_5 's published theory, so P_2 queries P_5 about f_5

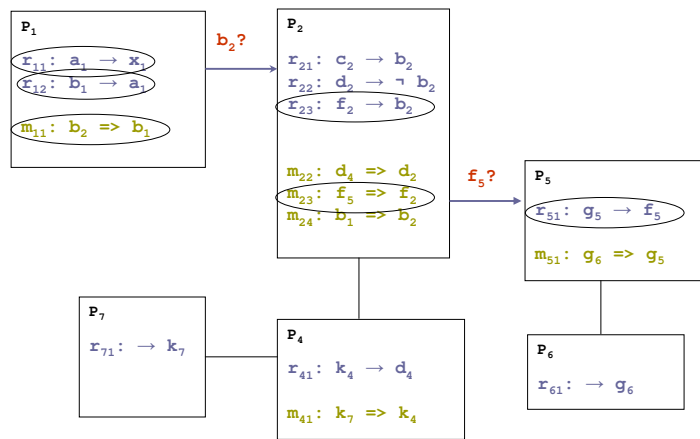


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r_{51} is a supportive rule for f_5 , which has g_5 as its only premise.

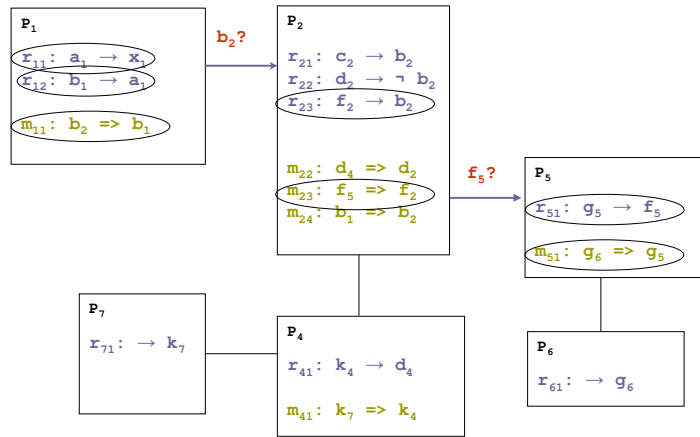


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m_{51} is a supportive rule for g_5 , which has g_6 as its only premise.

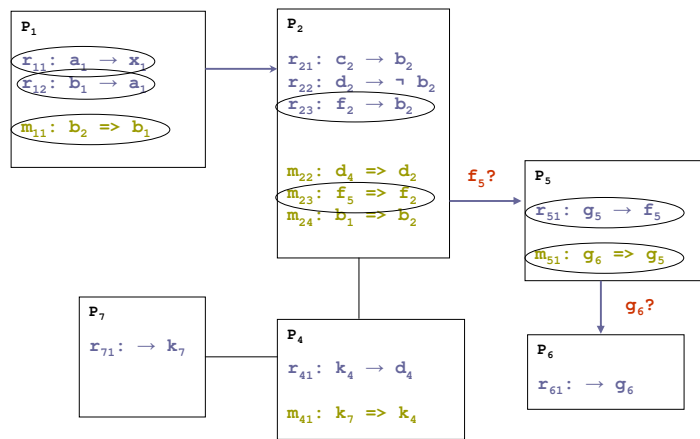


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g_6 belongs to P_6 's published theory, so P_5 queries P_6 about g_6

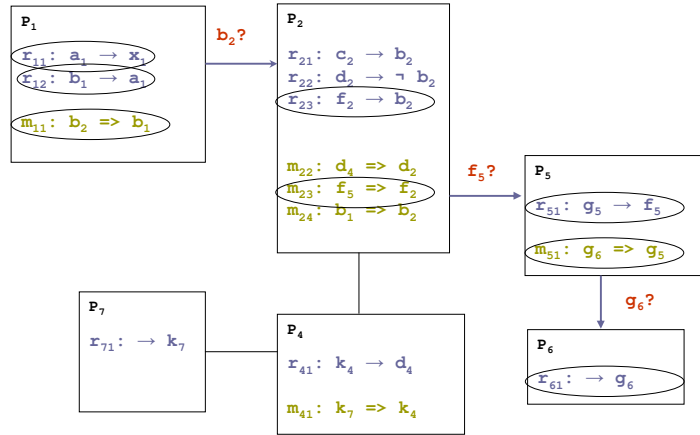


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g_6 derives from P_6 's local theory

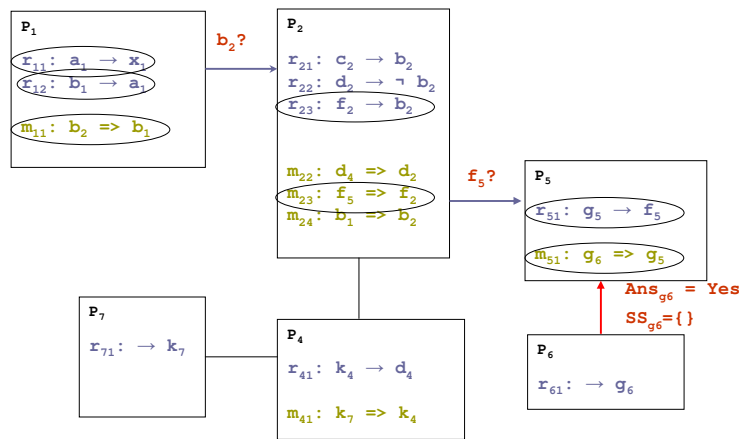


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P_6 returns the answer **Yes** to P_5 about g_6 with an empty set of supportive mappings (it was proved locally)

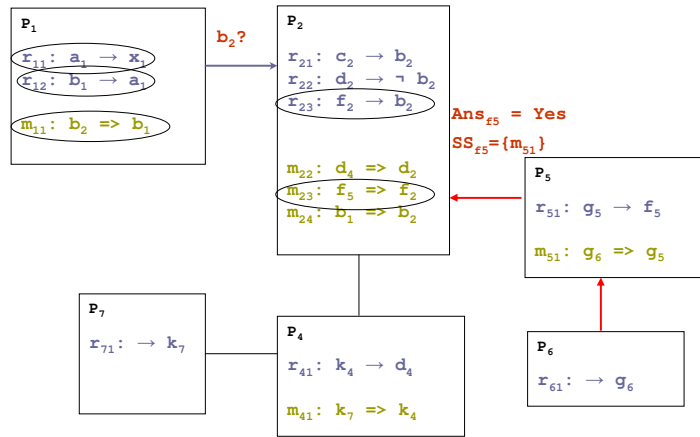


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P_5 returns the answer *Yes* to P_2 about f_5 with a supportive set $SS_{f_5}=\{m_{51}\}$

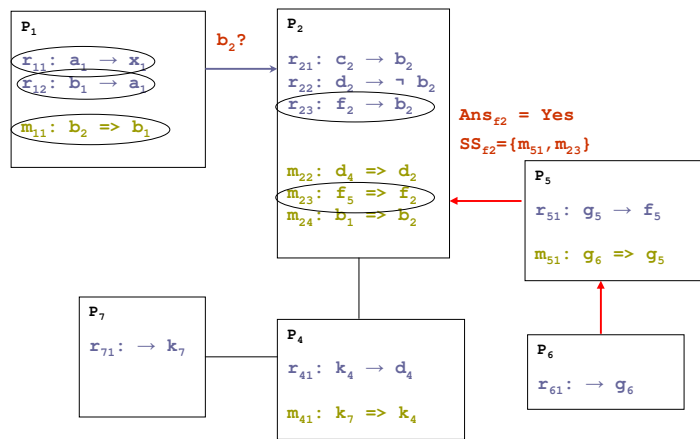


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33

P_2 returns the answer *Yes* for f_2 with a supportive set $SS_{f_2}=\{m_{51}, m_{23}\}$

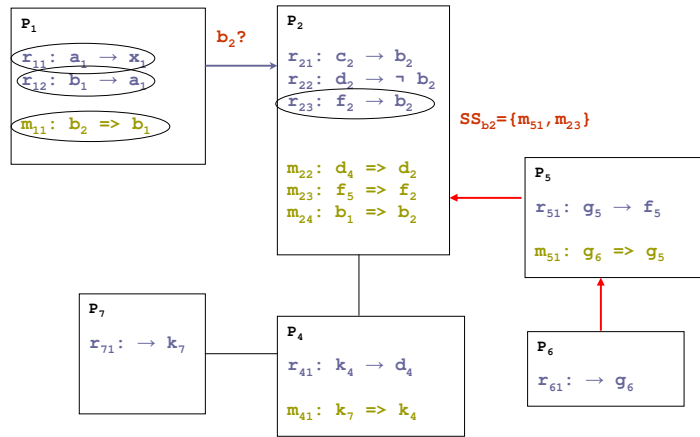


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P_2 builds an initial supportive set for b_2 , $SS_{b_2} = \{m_{51}, m_{23}\}$

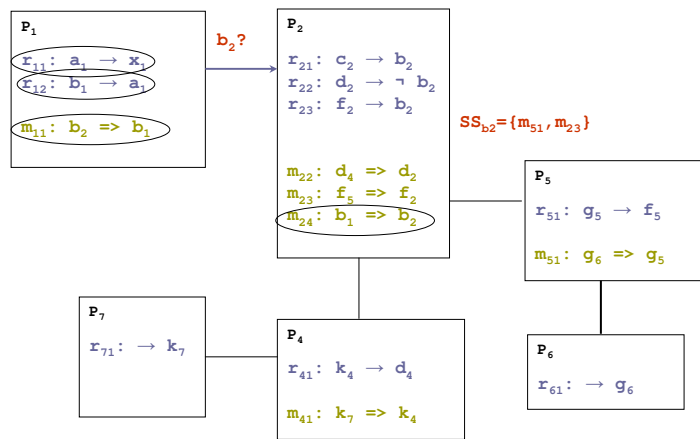


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m_{24} is another supportive rule for b_2 , which has b_1 as its only premise.

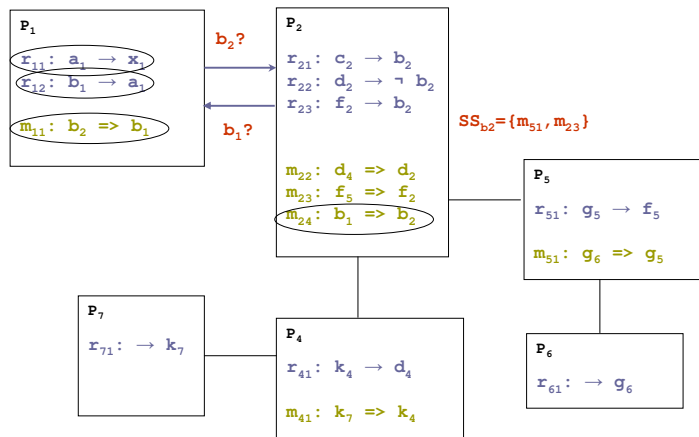


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b_1 belongs to P_1 's published theory, so P_2 queries P_1 about b_1

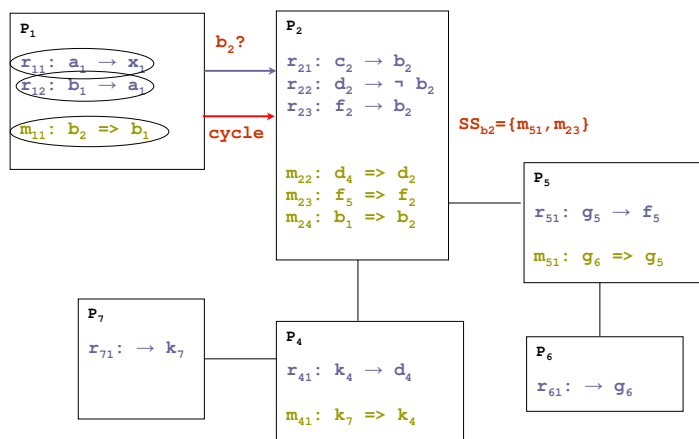


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A cycle is detected, so P_2 abandons m_{24}

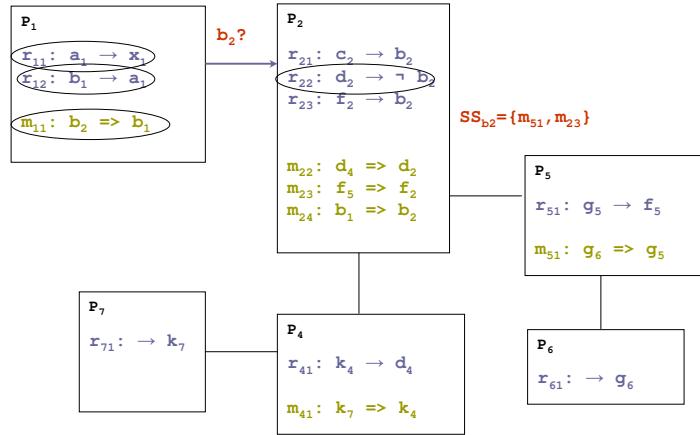


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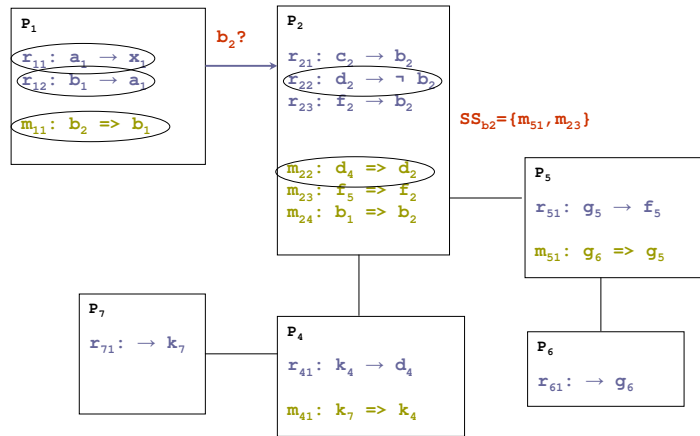
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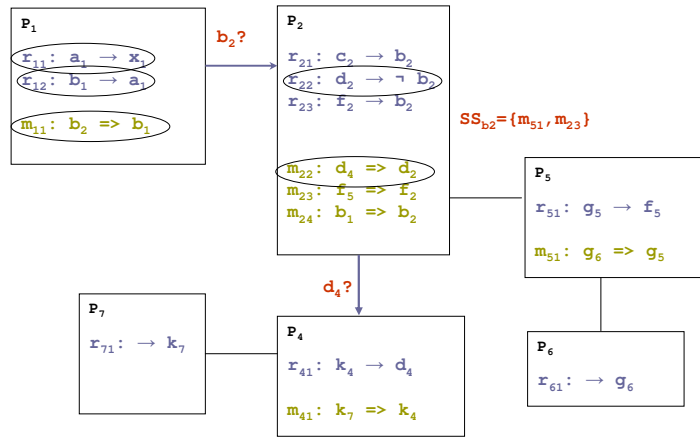
r_{22} is a conflicting rule for b_2 , which has d_2 as its only premise



m_{22} is a supportive rule for d_2 , which has d_4 as its only premise.



d_4 belongs to P_4 's published theory, so P_5 queries P_4 about d_4

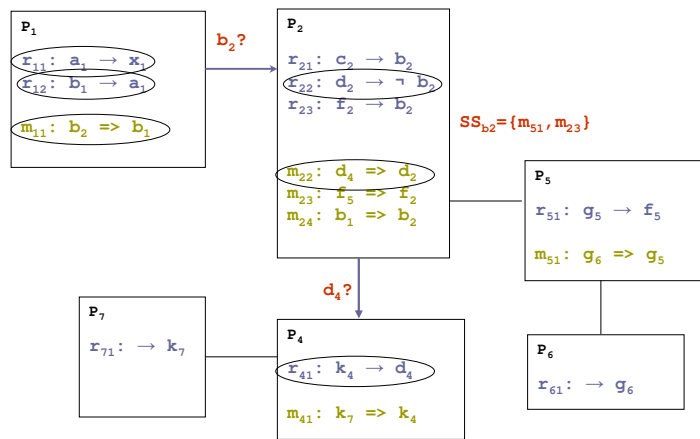


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r_{41} is a supportive rule for d_4 , which has k_4 as its only premise

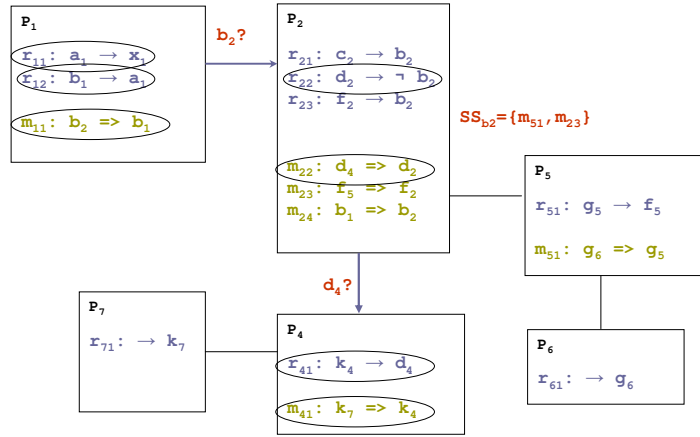


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m_{41} is a supportive rule for k_4 , which has k_7 as its only premise

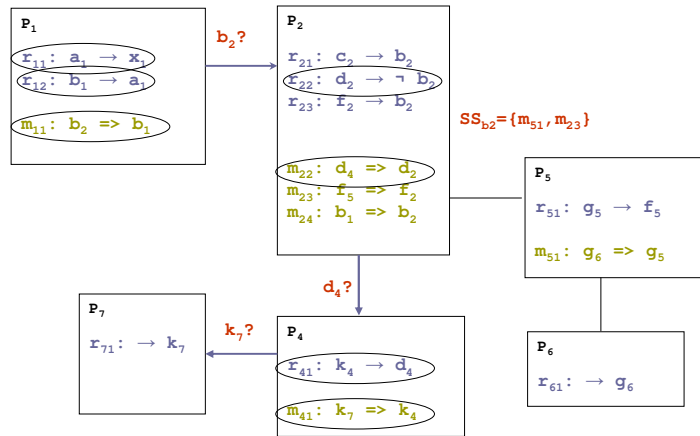


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k_7 belongs to P_7 's published theory, so P_4 queries P_7 about k_7

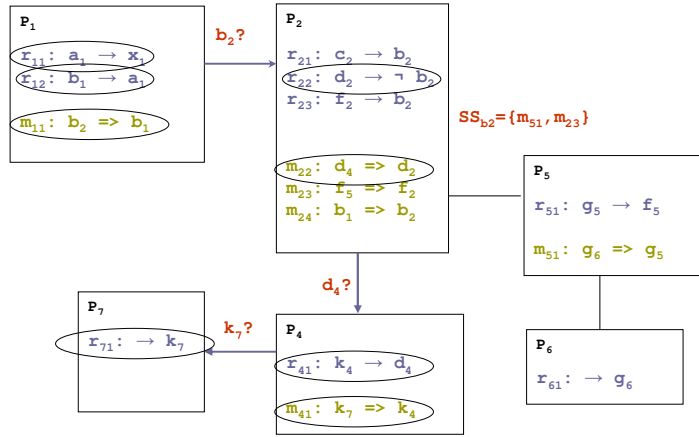


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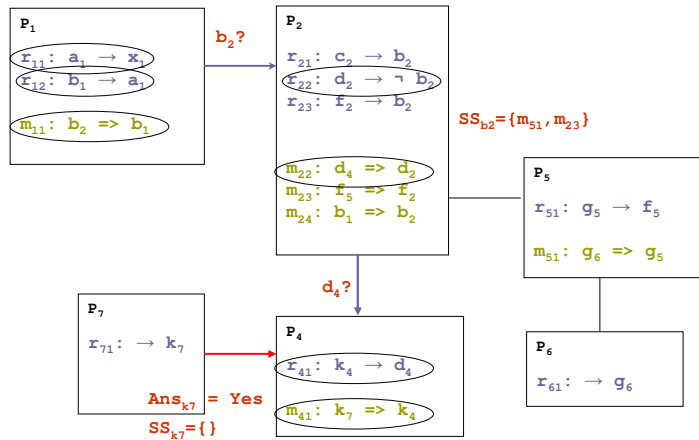
k_7 derives from P_7 's local theory



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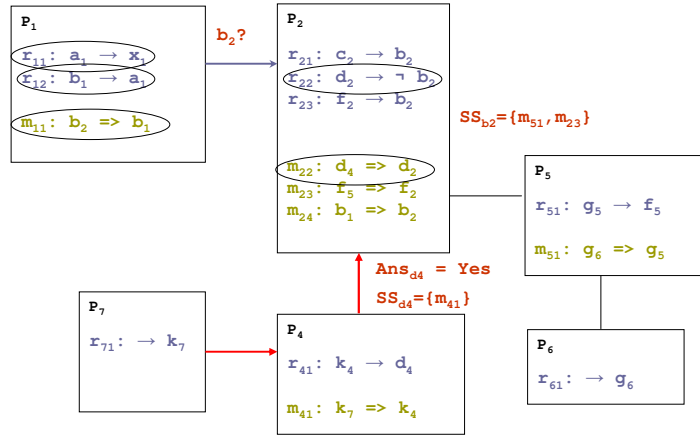
P_7 returns the answer Yes to P_4 about k_7 with an empty set of supportive mappings (it was proved locally)



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P_4 returns the answer Yes to P_2 about d_4 with a supportive set

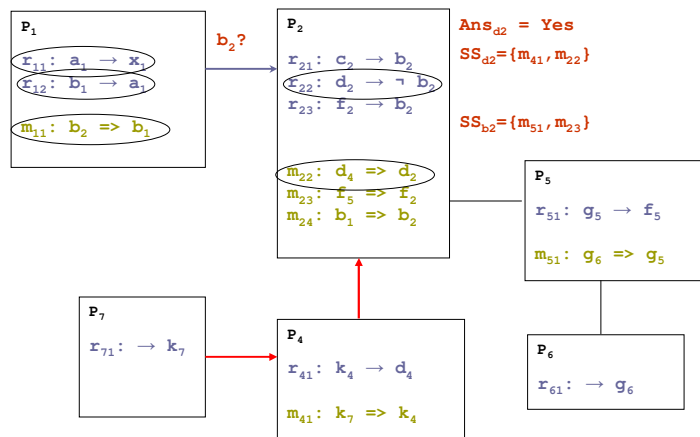


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P_2 returns the answer Yes for d_2 with a supportive set $SS_{d_2} = \{m_{41}, m_{22}\}$

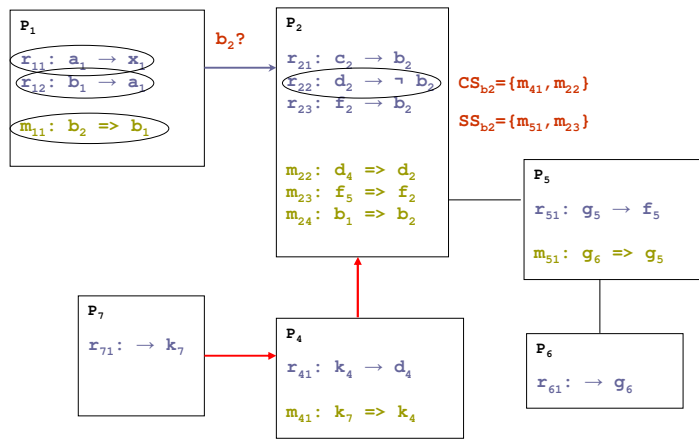


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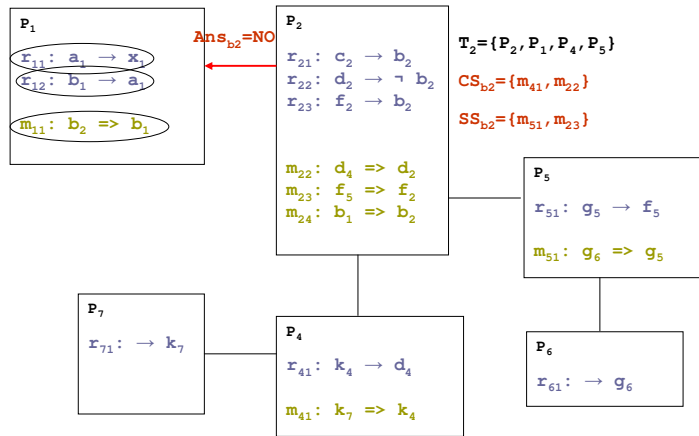
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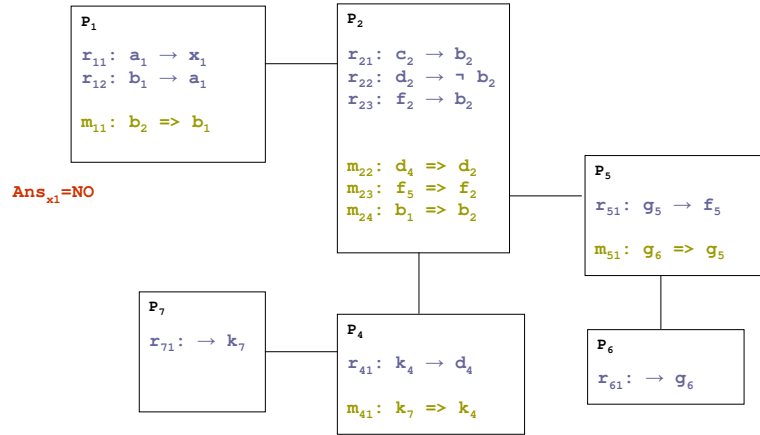
P_2 builds the conflicting set of b_2 , $CS_{b_2} = \{m_{41}, m_{22}\}$



Assuming that P_2 trusts P_4 more than P_5 , SS_{b_2} is not stronger than CS_{b_2} so P_2 cannot prove b_2 and returns **NO** to the query issued by P_1



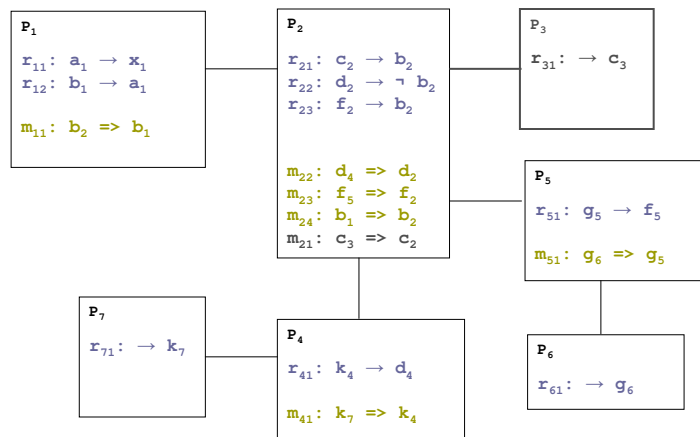
P_1 returns successively **NO** for b_j , a_j , and finally for x_j



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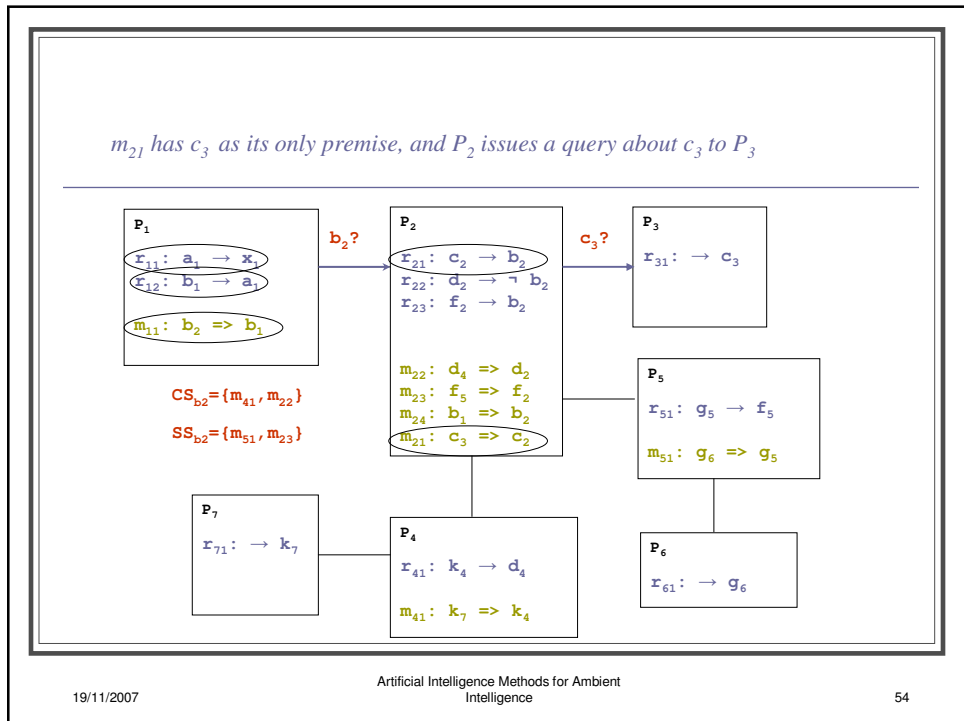
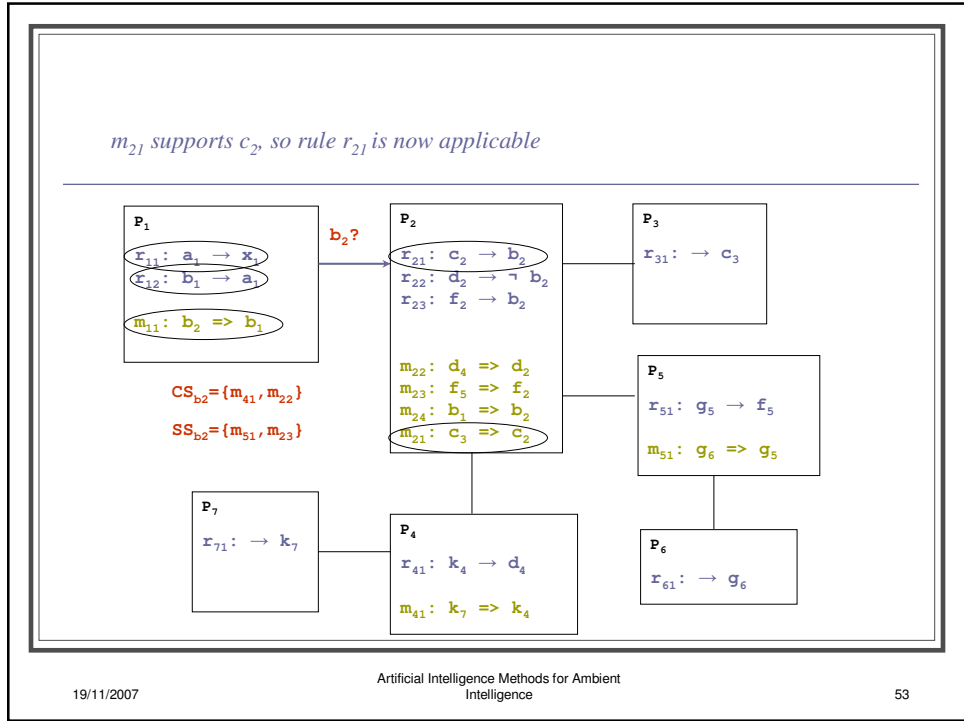
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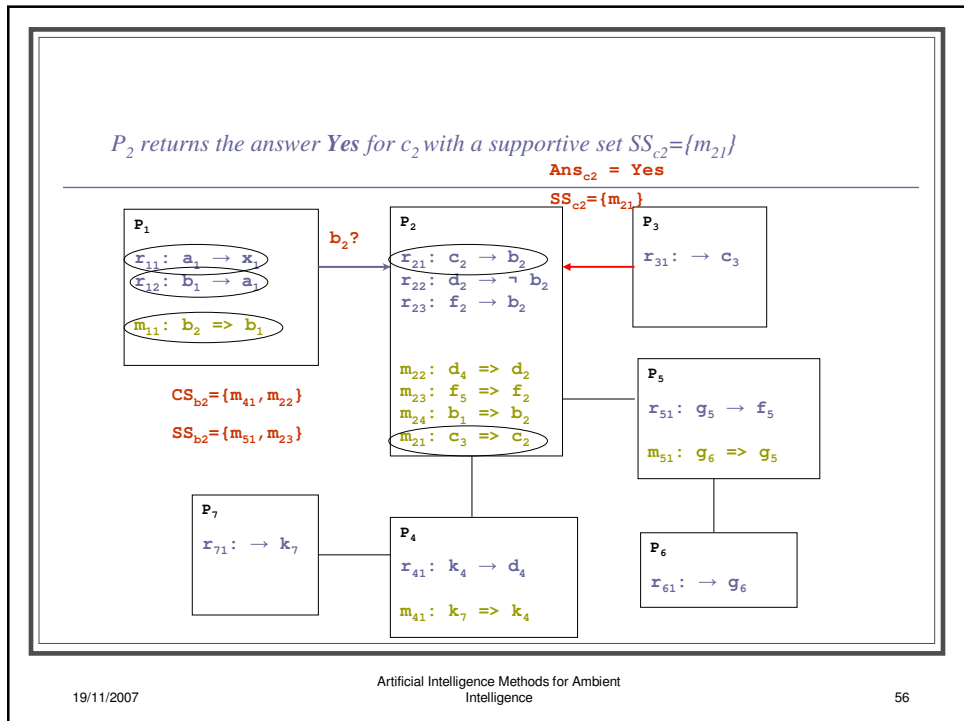
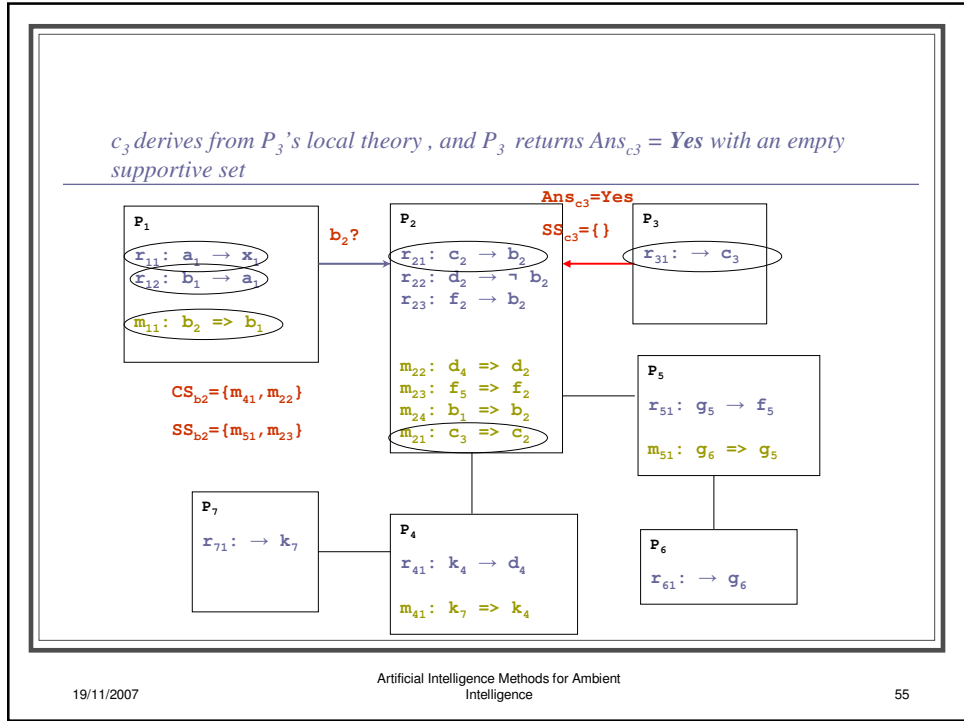
Imagine the case that a new peer (P_3) joins the system, and P_2 establishes a connection with the new peer through the mapping rule m_{21}



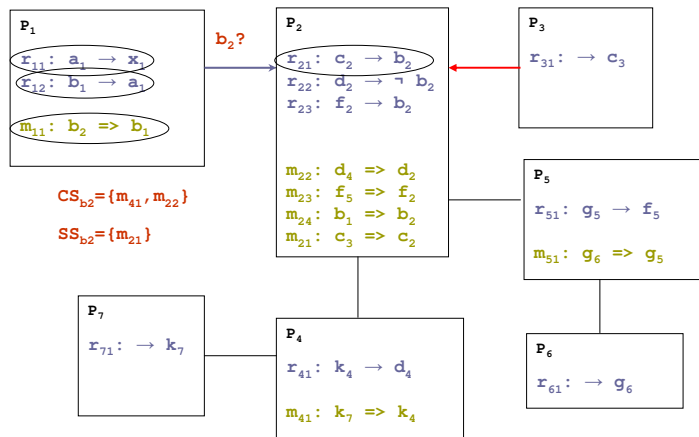
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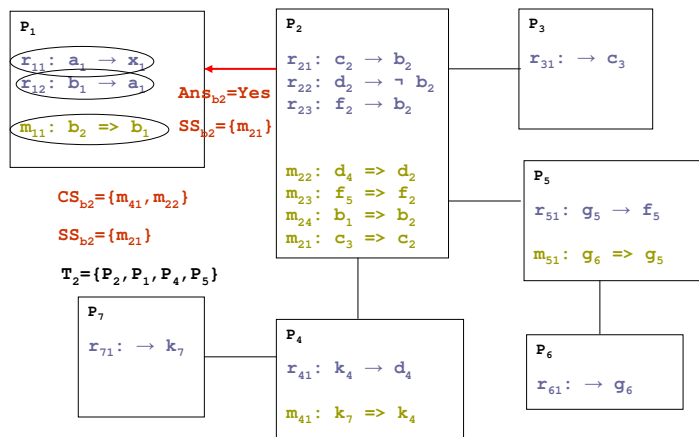




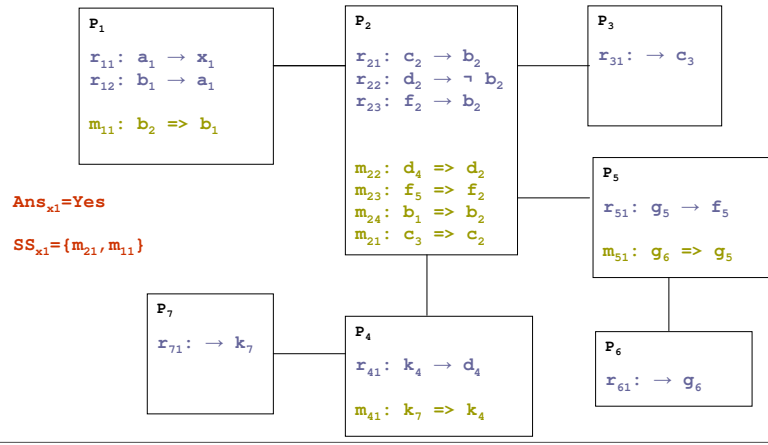
m_{21} is stronger than $\{m_{41}, m_{22}\}$, so $SS_{b_2} = \{m_{21}\}$



Now SS_{b_2} is stronger than CS_{b_2} as m_{21} is stronger than $\{m_{41}, m_{22}\}$, and P_2 returns $Ans_{b_2} = \text{Yes}$, with $SS_{b_2} = \{m_{21}\}$



P_1 returns successively *Yes* for b_1 , a_1 , and finally for x_1 , adding SS_{b_2} and m_{11} to SS_{x_1}



Properties of the Algorithm

- The algorithm is guaranteed to *terminate*
- The total *number of messages* that need to be exchanged for the evaluation of a single query is in the worst case $O(n^2)$ (n is the total number of system nodes)
- There is a *feasible theory* that derives from the unification of the distributed theories and derives the same conclusions.

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Future Work

- Extend the Algorithm to Support
 - Overlapping vocabularies
 - Defeasible Logic Local Theories
 - Non-Boolean queries
- Study Applications in the Aml Domain
 - Rules may represent ontological knowledge, policies, or regulations

Distributed Reasoning with Conflicts in an Ambient Peer-to-Peer Setting



Thank You!
Questions?

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