A Consistency Maintenance Mechanism for Subjective Judgments and Its Application

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ABSTRACT

In this paper, we propose a mechanism for maintaining consistency of subjective judgments. In the mechanism, subjective judgments of decision makers are extracted and used to select alternatives effectively. The subjective judgments are quantified by using AHP based pairwise comparisons. In order to maintain consistency of the comparisons, we exploit the TMS based nonmonotonic inference in the Prolog-based production system KORE/IE. By using the inference, the consistency of the comparisons is effectively maintained.

1. Introduction

Knowledge based systems support decision makers by using knowledge bases. It is necessary to construct and modify the knowledge bases beforehand in the decision support. It is tedious and consumes a lot of time for assembling and codifying knowledge in the knowledge bases. As an alternative means of articulating knowledge beforehand, we implement a system which extracts and dynamically uses information of decision makers in a decision process. The system is called CDSS(Choice Design Support System). CDSS is implemented on production system KORE/IE(Shintani 1988). KORE/IE is a fast production system on Prolog. The inference mechanism is realized by performing a recognize-act cycle in a similar manner as OPS5(Forgy 1981) in which the type of reasoning is forward chaining.

CDSS supports decision makers by structuring their knowledge and extracting their subjective judgments. Concretely, CDSS helps decision makers make a reasonable choice from alternatives. In the choice process, two processes are used. One is a process which clarifies attributes used for estimating the alternatives. The other is a process which determines the priorities of the alternatives by estimating the attributes. In the estimation, a pairwise comparison method based on AHP (Analytic Hierarchy Process)(Saaty 1980) is used. AHP is one of the weighting methods, and has an advantage of being able to treat subjective judgments of decision makers. However, in the method, it is difficult to maintain consistency of the judgments.

In this paper, especially, we focus on the mechanism for maintaining teonsistency and show the method for constructing the mechanism in CDSS.

2. The outline of CDSS

Fig.1 illustrates processes for decision support in CDSS. CDSS supports decision makers to create goals and their subgoals for problem solving. In the process CDSS uses subjective judgments of decision makers, and offers functions to clarify a structure of the problems. Knowledge of the decision makers is structured hierarchically through the process. In the hierarchy, elements which should be considered about the problems are arranged. The knowledge structured like this will be used as basic information for knowledge acquisition tools.

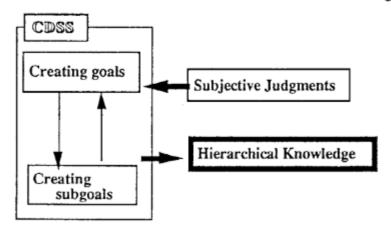


Fig.1. The Processes for decision support in CDSS

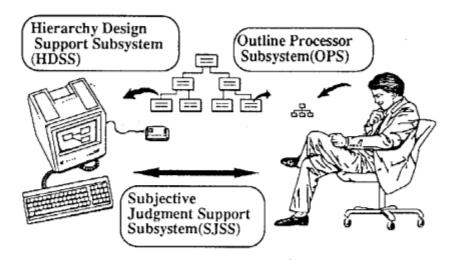


Fig.2. The subsystems in CDSS

CDSS is composed of (1)the Hierarchy Design Support Subsystem(HDSS), (2)the Subjective Judgments Support Subsystem(SJSS), and (3)the Outline Processor

Subsystem(OPS) as shown in Fig.2. The decision support processes in CDSS are achieved by combining and using these subsystems. HDSS helps to clarify elements which should be considered in the decision making process by arranging them hierarchically. The complexity in the problem is canceled through extracting relations among the elements and arranging the elements hierarchically. The hierarchy is treated visually as a graph by using a graph editor KORE/GET as shown in Fig.3. The editor provides functions for manipulating the graph. The functions of the editor include (1)editing nodes and arcs of the graph, and (2)drawing the graph in a visually well-organized form which enables us to easily grasp the structure of the graph. In the example illustrated in Fig.3, the graph is used for clarifying superiority or inferiority among alternatives at the third level by enumerating and estimating the elements (at the second level) which should be considered concerning a goal. The goal is shown at the first level of the graph.

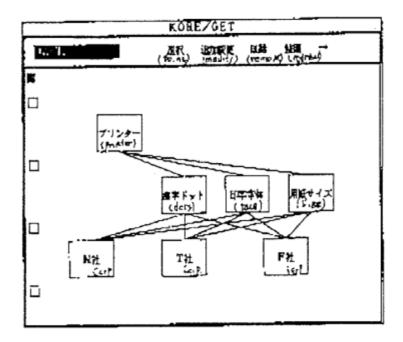


Fig.3. An example of the graph (in Japanese)

SJSS is used to quantify subjective judgments of decision makers by using AHP based pairwise comparisons. SJSS maintains consistency of the judgments. SJSS is discussed in more detail in Section 3.

OPS supports decision makers to arrange and structure their knowledge for problem solving. The organized knowledge is used and visually clarified as basic information for the decision support process in HDSS. CDSS provides a simulation function in the decision process by using HDSS and SJSS cooperatively. The function corresponds to a sensitivity

analysis which is used to see how changing a judgment has an effect on a conclusion. Through the simulation, decision makers get information to evaluate and organize knowledge for the decision process.

3. The consistency maintenance mechanism

3.1. Quantifying subjective judgments

A decision maker can make better choices between elements of a problem by using the ratio scales which are obtained from his subjective judgments by using AHP based pairwise comparisons. The comparisons are used to determine the relative importance of the elements in a level with respect to the elements in the level immediately above it. A matrix is set up by carrying out all the comparisons. The matrix is called the pairwise comparison matrix. The pairwise comparison matrix is used to generate the ratio scale for clarifying the priority of the element in the level. The pairwise comparison matrix is a square matrix whose row (or column) consists of elements used for the comparisons. The content of the matrix is weighting numbers for the comparisons where if the value V for element a_{ij} is determined then the reciprocal value 1/V is automatically entered in a_{ji} . The value V takes an integer which is less than 9. The larger the number, the more importantly the judgment is agreed upon.

In AHP, consistency can be checked by using the inconsistency ratio(I.R.) which can be represented as follows;

$$(\lambda_{max} - n)/(n - 1)$$

where λ_{max} , n are the maximum eigenvalue and the size of the pairwise comparison matrix, respectively. If the I.R. is less than 0.1, the consistency is considered acceptable; otherwise it is inconsistent. If inconsistency has occurred, we need to maintain the consistency by revising a few inadequate pairwise comparisons. However, it is difficult to find the comparisons. In the worst case it forces us to revise all the comparisons.

3.2. The Subjective Judgments Support Subsystem

The Subjective Judgments Support Subsystem(SJSS) provides a mechanism for automatically maintaining the consistency of the pairwise comparisons. The maintenance mechanism is realized by using rule based programming and a nonmonotonic inference function in KORE/IE(Shintani 1989). If inconsistency has occurred, inadequate pairwise comparisons are checked and revised first for maintaining the consistency.

3.2.1. The rule based process for pairwise comparisons

Fig.4 illustrates the process for pairwise comparisons in SJSS. In Fig.4, the symbol surrounded by a square represents a name of a rule. The rule "start" is used for initializing a state for pairwise comparisons.

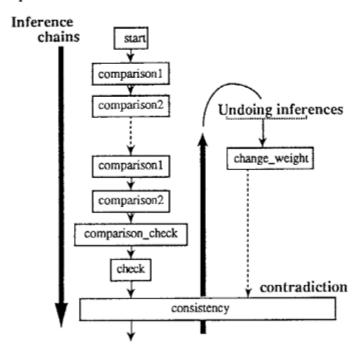


Fig.4. The Process for pairwise comparisons in SJSS

The rules "comparison1" and "comparison2" are used for obtaining a pairwise comparison matrix. The rule comparison2 determines the reciprocal values and enters them into the matrix. The rule comparison2 means that "If the value of the cell in the C-th column, the R-th row of the matrix X is the weight W, and the value of the cell in the R-th column, the C-th row of the matrix X is undefined, then put the weight 1/W into the cell in the R-th column, the C-th row of the matrix X". The rule comparison2 can be described as follows;

where a rule in KORE/IE consists of (1)the name of the rule (such as *comparison2*), (2)the symbol ":", (3)the symbol "if", (4)the LHS(Left Hand Side) which has some LHS patterns (such as "matrix_element(name=X, ...)"), (5)the symbol "then", (6)the RHS(Right Hand Side) which has some RHS actions (such as "add"), and (7) the symbol ".". The symbol "&"

is a delimiter. The patterns correspond to compound terms of Prolog. The arguments are a sequence of one or more slot-value pairs. The pairs correspond to terms which have functions of arity 2 (e.g. =, $\==$, <, >). The second LHS pattern with the symbol "-" represents a negative pattern. In a matching process, the negative pattern is satisfied if no working memory(WM) elements match the pattern. The action add is used to create a new WM element and put the weight W into the cell in the R-th column, the C-th row of the matrix X.

By the rule *comparison1*, weighting numbers are inserted and grouped into the matrix. The rule *comparison1* can be described as follows.

where the execution of RHS actions correspond to that of Prolog goals. The fourth RHS action is executed based on the OR control structure in Prolog. In the execution of the action qa_weight, pairwise comparisons are obtained, and a decision maker declares his belief for the pairwise comparisons. The fifth argument Default of the action qa_weight returns the status of the belief. The action qa_weight keeps the information of the comparisons in a database. The information is used to avoid the comparisons which had been already performed previously when the rule will be triggered by the nonmonotonic inference mechanism in KORE/IE. If a cell in the C-th column, the R-th row of the matrix X is already filled by the weight number W. the action qa_weight returns the number W with the belief Default. There are two type of the belief which are (1)assumption and (2)fact. Concretely, the decision maker can group the weighting numbers into two classes such as (1)temporary(or assumption) numbers and (2)reliable(or fact) numbers which can be generated by using the actions add assumption and add, respectively. The actions are defined as predicates of Prolog. add_assumption is used to create a new assumption in the WM, and keep the temporary number W and the information of the cell in the database. If inconsistency has occurred, the temporary numbers are checked and revised first for maintaining the consistency.

The rule "comparison_check" checks whether the matrix is filled or not. The rule "check" is used for getting the I.R. of the matrix.

3.2.2. The TMS based mechanism for maintaining the consistency

If the I.R. comes to be greater than 0.1, SJSS finds the causes for the inconsistency, and revises the causes by using the TMS(Doyle 1979) based inference mechanism in KORE/IE. The causes correspond to inadequate pairwise comparisons. The TMS based mechanism provides some advantages as follows: (1)In order to realize flexible pairwise comparisons, a decision maker can represent his belief for the pairwise comparisons. (2)The system can dynamically manage the order of the comparisons for realizing efficient comparisons. For example, SJSS can omit some redundant pairwise comparisons and check the consistency of each comparison dynamically by exploiting the Harker's method(Harker 1987).

The rule "consistency" is used for achieving the action "contradiction" if the I.R. is greater than 0.1. The action is used to find an inadequate assumption and resolve an inconsistency by revising the assumption. The action provides a TMS based function for maintaining consistency among pairwise comparisons. In SJSS, by using the action, the inconsistency is resolved by decreasing a temporary number of I.R. in which the number will come to be less than 0.1.

The outline of the algorithm for the action contradiction can be shown as in Fig.5.

- Step 1: Get a list L of assumption on which an inconsistency depends. Go to Step 1-1.
- Step 1-1: Sort the list L by using a specified strategy. Go to Step 2.
- Step 2: If the list L is empty then stop the action and the action becomes a failure; otherwise go to Step 3.
- Step 3: Get an assumption from the list L, which is the first element(that is, Lhead) of the list L (that is, L=[Lhead|Ltail]), and backtrack to the inference step which includes the assumption. Then, assert a negation of the assumption. Go to Step 4.
- Step 4: If a new inconsistency is detected, then undo the result of Step 3, let L=Ltail, and go to Step 2; otherwise remove the data which depend on the assumption, and go to 5.
- Step 5: Stop the action. The action succeeds.
 - Fig.5. Sketching the algorithm for the action contradiction.

The list L corresponds to the nogood-set of TMS. In TMS, assumptions in the nogood-set are checked in an arbitrary order. The feature of the algorithm shown in Fig.5 is to provide a

function to check the assumption in a specified order. By specifying the order the algorithm can resolve the inconsistency effectively. The rule consistency can be described as follows:

```
consistency:
    if matrix(name=X,inconsistency_ratio > 0.1)
    then
        (contradiction(Assumption,decrease_IR)
        ;
        otherwise(Assumption,decrease_IR,new_weight).
```

where the action contradiction is used in the RHS. If the action contradiction fails, the first argument of the action contradiction returns a list of assumptions which corresponds to the nogood-set. The second argument of the action is used for indicating a strategy which sorts the list of the assumptions. The strategy is defined as a Prolog predicate which has an argument. The strategy is used at Step1-1 in Fig.5. In SJSS, by using the strategy decrease_IR, the assumptions in the list are arranged in decreasing order of the amount of number which is increased and decreased for revising the temporary numbers. If the second argument is not indicated, the list obtained at Step 1 is directly used in which the assumptions are arranged in order of generating them.

In the RHS, if the action contradiction fails, the action otherwise is activated. The action otherwise resolves inconsistency in the same manner as the action contradiction. However, there are two main differences between the actions. One is that the action otherwise is used for revising some assumptions at the same time. Second is that the action otherwise directly revises the temporary numbers in the list Assumption by using a specified procedure which is indicated at the third arguments. The reason for the direct revision is that in a forward chaining production system the simultaneous revision of the numbers corresponds to achieving the revision during each recognize-act cycle. Namely, the action otherwise includes a function for the rule change_weight mentioned in the next section. The simultaneous revision needs to be achieved only when the rule consistency is firing.

3.2.3. Revising the pairwise comparisons

A rule is needed to concretely revise an assumption because the action *contradiction* only negates the assumption by asserting the negation of the assumption as shown at Step 3 in Fig.5.

The rule "change_weight" is used for concretely revising the assumption(that is, an inadequate pairwise comparison), which can be described as follows;

```
change_weight:
if \matrix_element(name=X,column=C,row=R,weight=W)
then
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new_weight(W,W2) & add(matrix_element(name=X,column=C, row=R,weight=W2).

where the LHS pattern corresponds to the negation of the inadequate comparison which needs to be revised. The negation is generated by the action contradiction if the action finds the comparison as an assumption. By firing the rule change_weight, the inadequate comparison is revised. In order to revise the comparison, the RHS action "new_weight" is used. The action changes the old weighting value "W" to the new weighting value "W2" which contributes to decrease the I.R. of the matrix. The range of the change takes ±2 because we assume that the old value is not so inadequate for the consistency. If the I.R. comes to be less than 0.1, the action contradiction stops and the consistency is automatically maintained.

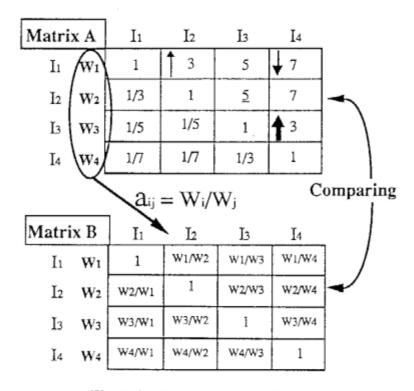


Fig.6. Setting up Matrix B from Matrix A

It is well known that it is difficult to determine the range of the change in which we need to decrease the maximum eigenvalue of a pairwise comparison matrix mentioned in Section 3.1. In order to concretely determine the range, we exploit the feature of the matrix based on AHP. Fig. 6 shows how the range is determined. It shows two pairwise comparison matrices which are Matrix A and Matrix B. An element aij in Matrix A is obtained by performing a pairwise comparison between elements Ii and Ij. In Matrix A, the direction of the arrows are used to represent the range of the direction in which the upper and down arrows indicate increasing and

decreasing, respectively. The amount of increasing and decreasing is represented by using three kinds of arrows(that is, thin, normal, and thick) in which the thicker the arrow, the more the change is taken.

In order to determine the arrows, we set up Matrix B from Matrix A as shown Fig.6. Namely, Matrix B can be set up by using the weight of each element, which is computed from Matrix A. By exploiting the theory of AHP, the element aij in Matrix B are defined as Wi/Wj in which Wi and Wj are weights of the elements Ii and Ij in Matrix A, respectively. The I.R. of Matrix B comes to be 0 by using the method based on AHP. Matrix B corresponds to the ideal matrix in which pairwise comparisons of any problem are consistently performed. So, by gradually bringing Matrix A close to Matrix B, the I.R. of Matrix A will decrease effectively. Namely, the range of the change mentioned above can be determined by comparing Matrix A and B.

4. Conclusions

Subjective judgments of decision makers are quantified easily by using the pairwise comparison method. In the method, it is difficult to maintain the consistency among the comparisons. In order to maintain consistency, tedious revisions for the comparisons are required. In CDSS consistency is automatically maintained by using the TMS based inference mechanism in KORE/IE. Generally speaking, the validity of the maintenance is to be checked by decision makers. However, decision makers can get a reasonable result after the automatic maintenance. The reasons of the results can be summarized as follows: (1) By using AHP method, decision makers can perform almost valid comparisons. The decision makers can assume that the initial comparisons is not so inadequate for the consistency. (2) By using the TMS based mechanism, decision makers can represent their belief for the comparisons. If the inconsistency is arisen, the inconsistency can be resolved effectively by focusing on the assumption based comparisons. Generally, in case of using TMS, a convergence becomes a matter of concern. In our approach, the mechanism converges effectively to the reasonable result by using domain knowledge (that is, Matrix B mentioned in Section 3.2.3). (3) By restricting the range of the change (that is, ±2) for the comparisons, decision makers get reasonable result after the automatic revisions.

CDSS is similar to the frameworks of MORE(Kahn 1985) and MOLE(Eshelman 1986) which are knowledge acquisition tools. MORE and MOLE is used to obtain diagnosis knowledge from decision makers and structure them by using a domain depend interview method. The main feature of CDSS is that CDSS supports decision makers without the domain knowledge and the specified interview method. The future subject is that we enhance CDSS in

order to realize a knowledge acquisition tool by adding functions for constructing and refining a model of a problem domain.

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