

Preliminaries

We need the following concepts introduced in [2].

Aggregation Structure

Definition 1. A sup-preserving aggregation structure (aggregation structure, for short) $\langle L_1, L_2, L_3, \square \rangle$, where $L_i = \langle L_i, \leq_i \rangle$, $i = 1, 2, 3$, are complete lattices and $\square : L_1 \times L_2 \rightarrow L_3$ is a function which commutes with suprema in both arguments. Define operations

$\circ : L_1 \times L_3 \rightarrow L_2$, $\circ : L_3 \times L_2 \rightarrow L_1$ by

$$a_1 \circ a_3 = \bigvee_2 \{a_2 \mid a_1 \square a_2 \leq_3 a_3\},$$

$$a_3 \circ a_2 = \bigvee_1 \{a_1 \mid a_1 \square a_2 \leq_3 a_3\}.$$

Moreover, define operation $\overset{\text{op}}{\square} : L_3 \times L_2 \rightarrow L_1$ (needed only for scalar-by-scalar equations) by

$$a_3 \overset{\text{op}}{\square} a_2 = \begin{cases} \bigwedge_1 \{a_1 \mid a_1 \square a_2 \geq_3 a_3\}, & \text{if exists } a_1 \in L_1 : \\ & a_1 \square a_2 \geq_3 a_3, \\ 0_2, & \text{otherwise.} \end{cases}$$

Consider two important examples of aggreg. struct. In both cases, $\langle L, \wedge, \vee, \otimes, \rightarrow, 0, 1 \rangle$ is a complete residuated lattice. $L_i = L$ and \leq_i is either \leq or the dual of \leq .

Example 1. $L_1 = \langle L, \leq \rangle$, $L_2 = \langle L, \leq \rangle$, $L_3 = \langle L, \leq \rangle$, $\square = \otimes$:

$$a_1 \circ a_3 = \bigvee \{a_2 \mid a_1 \otimes a_2 \leq a_3\} = a_1 \rightarrow a_3,$$

$$a_3 \circ a_2 = \bigvee \{a_1 \mid a_1 \otimes a_2 \leq a_3\} = a_3 \leftarrow a_2.$$

Example 2. $L_1 = \langle L, \leq \rangle$, $L_2 = \langle L, \leq^{-1} \rangle$, $L_3 = \langle L, \leq^{-1} \rangle$, $\square = \rightarrow$:

$$a_1 \circ a_3 = \bigwedge \{a_2 \mid a_1 \rightarrow a_2 \geq a_3\} = a_1 \otimes a_3,$$

$$a_3 \circ a_2 = \bigvee \{a_1 \mid a_1 \rightarrow a_2 \geq a_3\} = a_3 \rightarrow a_2.$$

General Product of Fuzzy Relations

Definition 2. For an aggregation structure $\langle L_1, L_2, L_3, \square \rangle$, and fuzzy relations $R \in L_1^{X \times Y}$, $S \in L_2^{Y \times Z}$, let a fuzzy relation $R \boxtimes S \in L_3^{X \times Z}$ be defined by

$$(R \boxtimes S)(x, z) = \bigvee_{y \in Y} (R(x, y) \square S(y, z)).$$

Product \boxtimes generalizes both sup-t-norm product (\circ) and inf-residuum product (\leftarrow):

- :: for the setting of Example 1: $R \boxtimes S = R \circ S$,
- :: for the setting of Example 2: $R \boxtimes S = R \leftarrow S$.

Definition 3. For $R \in L_1^{X \times Y}$ and $S \in L_2^{Y \times Z}$, let $R \triangleleft S \in L_3^{X \times Z}$ and $R \triangleleft S \in L_1^{X \times Z}$ be defined by

$$(R \triangleleft S)(x, z) = \bigwedge_{y \in Y} (R(x, y) \circ S(y, z)),$$

$$(R \triangleleft S)(x, z) = \bigwedge_{y \in Y} (R(x, y) \circ S(y, z)).$$

Fuzzy Relation Equations

Fuzzy relational equations play an important role in fuzzy set theory and its applications. Namely, it is often the case that the problem in a particular application of fuzzy logic may be transformed to the problem of identification an unknown fuzzy relation.

The problem to determine an unknown fuzzy relation R between universe sets X and Y such that

$$R \boxtimes S = T,$$

where S and T are given (known) fuzzy relations between Y and Z , and X and Z , respectively, and \boxtimes is an operation of composition of fuzzy relations, is called the problem of fuzzy relational equations. Alternatively, given R and T , the problem is to determine S .

Denotation:

$$U \boxtimes S = T \quad \text{and} \quad R \boxtimes U = T,$$

where U is the unknown fuzzy relation.

Solvability Criteria

Theorem 4. Let $R \in L_1^{X \times Y}$, $S \in L_2^{Y \times Z}$, and $T \in L_3^{X \times Z}$ be fuzzy relations. Then

- :: $U \boxtimes S = T$ has a solution iff $T \triangleleft S^{-1}$ is its solution,
- :: $R \boxtimes U = T$ has a solution iff $R^{-1} \triangleleft T$ is its solution.

Theorem 5. If an equation $U \boxtimes S = T$ is solvable then the set of all solutions along with \triangleleft_1 forms a complete join-semilattice with the greatest element

$$T \triangleleft S^{-1}.$$

If an equation $R \boxtimes U = T$ is solvable then the set of all solutions along with \triangleleft_2 forms a complete join-semilattice with the greatest element

$$R^{-1} \triangleleft T.$$

Corollary 6. An equation $U \circ S = T$ is solvable iff $(S \triangleleft T^{-1})^{-1}$ is its solution. An equation $R \circ U = T$ is solvable iff $R^{-1} \triangleleft T$ is its solution.

All Solutions vs. Minimal Solutions

We assume two solutions R', R'' of $U \boxtimes S = T$, and fuzzy relation $R \in L_1^{X \times Y}$ such that $R' \triangleleft_1 R \triangleleft_1 R''$. Since \boxtimes is isotone in both arguments, we can write

$$T = R' \boxtimes S \triangleleft_1 R \boxtimes S \triangleleft_1 R'' \boxtimes S = T,$$

which implies that R is a solution of $U \boxtimes S = T$ as well.

In other words, a set of all solutions of $U \boxtimes S = T$ w.r.t. \triangleleft_1 is a convex set. Therefore, we just need to find all minimal solutions.

Minimal Solutions

“Scalar-by-Scalar Equation”

Let $u \in L_1$, $s \in L_2$, $t \in L_3$, define scalar-by-scalar equation:

$$u \square s = t. \quad (1)$$

Theorem 7. If equation (1) is solvable then for every its solution $r \in L_1$ it holds $r \in [t \overset{\text{op}}{\square} s, t \overset{\text{op}}{\square} s]$.

Lemma 8. Let L_1 be a finite chain. If there is $a_1 \in L_1$ such that $a_1 \square a_2 \geq_3 a_3$ then $a_1 \overset{\text{op}}{\square} a_3 = a_1 \circ a_3$.

Corollary 9. Let L_1 be a finite chain. Equation (1) is solvable iff $t \overset{\text{op}}{\square} s$ is its solution.

“Vector-by-Vector Equation”

Let $(u_j) \in L_1^Y$, $(s_j) \in L_2^Z$, $t \in L_3$, $j \in J = \{1, \dots, n\}$, define vector-by-vector equation:

$$(u_1 \dots u_n) \boxtimes \begin{pmatrix} s_1 \\ \vdots \\ s_n \end{pmatrix} = t. \quad (2)$$

Theorem 10. Let L_1 be a finite chain. If there is $j' \in J$ such that $u_{j'} \square s_{j'} = t$ is solvable then equation (2) has a minimal solution $R = (r_1 \dots r_n)$ such that

$$r_j = \begin{cases} t \overset{\text{op}}{\square} s_j, & \text{for } j = j', \\ 0_1, & \text{otherwise.} \end{cases}$$

“Vector-by-Matrix Equation”

Let $(u_j) \in L_1^Y$, $(s_{jk}) \in L_2^{Y \times Z}$, $(t_k) \in L_3^Z$, $j \in J = \{1, \dots, n\}$, $k \in K = \{1, \dots, p\}$, and L_1 be a finite chain, define vector-by-matrix equation:

$$(u_1 \dots u_n) \boxtimes \begin{pmatrix} s_{11} \dots s_{1p} \\ \vdots \\ s_{n1} \dots s_{np} \end{pmatrix} = (t_1 \dots t_p). \quad (3)$$

Equation (3) can be rewritten using a table \mathfrak{T} of dimension $(n+1) \times p$ (meaning, the last row is \vee_3 of the rows above):

$$\begin{array}{cccc} u_1 \square s_{11} & \dots & u_1 \square s_{1p} & \\ \vdots & & \vdots & \\ u_n \square s_{n1} & \dots & u_n \square s_{np} & \\ \hline t_1 & \dots & t_p & \end{array}$$

Now, see what happen to the table \mathfrak{T} when we take $R = (r_1 \dots r_n) = T \triangleleft S^{-1}$ (the greatest solution) as a solution of (3).

For every $j \in J$ there must be $k' \in K$ such that

$$r_j = \bigwedge_{k' \in K} (t_{k'} \circ s_{jk'}) = t_{k'} \circ s_{jk'}.$$

By K_j we denote the set of all indices $k' \in K$ such that $u_j \square s_{jk'} = t_{k'}$ is solvable and $r_j = t_{k'} \circ s_{jk'}$.

Minimal Solutions

Important assertions:

- :: for each $k' \in K_j$, $r_j = t_{k'} \circ s_{jk'}$ is the only solution of $u_j \square s_{jk'} = t_{k'}$
- :: for each $k'' \in K \setminus K_j$ we have $r_j \square s_{jk''} <_3 t_{k''}$.

Define a binary table \mathfrak{B} of the dimension $(n+1) \times p$:

$$\mathfrak{B}_{jk} = \begin{cases} 1, & \text{if } \mathfrak{T}_{jk} = t_k, \\ 0, & \text{if } \mathfrak{T}_{jk} <_3 t_k. \end{cases}$$

Obviously, the last row of the table \mathfrak{B} is filled by ones.

Definition 11. $J_{\text{cov}} \subseteq J$ is a covering of the last row of \mathfrak{B} if $\max_{j \in J_{\text{cov}}} \mathfrak{B}_{jk} = 1$ for all $k \in K$. Covering $J_{\text{cov}} \subseteq J$ is a minimal one if there is no covering J'_{cov} such that $J'_{\text{cov}} \subseteq J_{\text{cov}}$.

Example 3. Assume an equation (3) with table \mathfrak{T} :

$$\begin{array}{cccccccc} r_1 \square s_{11} = t_1 & r_1 \square s_{12} = t_2 & r_1 \square s_{13} <_3 t_3 & r_1 \square s_{14} <_3 t_4 & & & & \\ r_2 \square s_{21} <_3 t_1 & r_2 \square s_{22} <_3 t_2 & r_2 \square s_{23} = t_3 & r_2 \square s_{24} = t_4 & & & & \\ r_3 \square s_{31} = t_1 & r_3 \square s_{32} <_3 t_2 & r_3 \square s_{33} <_3 t_3 & r_3 \square s_{34} = t_4 & & & & \\ r_4 \square s_{41} <_3 t_1 & r_4 \square s_{42} <_3 t_2 & r_4 \square s_{43} = t_3 & r_4 \square s_{44} <_3 t_4 & & & & \\ \hline & t_1 & t_2 & t_3 & t_4 & & & \end{array}$$

Table \mathfrak{B} can be easily derived from \mathfrak{T} :

$$\begin{array}{cccc} 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ \hline 1 & 1 & 1 & 1 \end{array}$$

There exist several coverings, but just two of them are the minimal ones: $\{1, 2\}$, $\{1, 3, 4\}$.

Theorem 12. Let (3) be an equation with $R = (r_1 \dots r_n)$ being the greatest solution. Every minimal solution $M = (m_1 \dots m_n)$ of (3) is in the form:

$$m_j = \begin{cases} r_j, & \text{for } j \in J_{\text{cov}}, \\ 0_1, & \text{otherwise,} \end{cases}$$

where J_{cov} is a minimal covering of the last row of the corresponding table \mathfrak{B} .

Minimal solutions of $U \circ S = T$ (sup-t-norm equation) and $U \triangleleft S = T$ (inf-residuum equation) can be described using direct consequences of the previous results.

Future Research

- :: developing algorithms, efficient computation of all solutions (removing duplicities), complexity issues

References

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