

Finding close T -indistinguishability Operators to a given Proximity

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Abstract

Two ways to approximate a proximity relation R (i.e. a reflexive and symmetric fuzzy relation) by a T -transitive one where T is a continuous archimedean t-norm are given.

The first one aggregates the transitive closure \bar{R} of R with a (maximal) T -transitive relation B contained in R .

The second one modifies the values of \bar{R} or B to better fit them with the ones of R .

Keywords: Proximity, Transitive Closure, Opening, T -indistinguishability Operator, Aggregation Operator, Quasi Arithmetic Mean, Representation Theorem.

1 Introduction

A proximity matrix or relation on a finite universe X is a reflexive and symmetric fuzzy relation R on X . In many applications transitivity of R with respect to a t-norm T is required. In these cases, R must be replaced by a new relation E also satisfying transitivity, such relations called T -indistinguishability operators. Of course, it is desirable that E is as close as possible to R . This paper presents a couple of ways to find close transitive relations to R in a reasonable way - i.e.: easy and rapid to generate- when the t-norm is continuous archimedean.

There are of course several ways to calculate the closeness of two fuzzy relations, many of them related to some metric. In this paper we propose a way related to the natural indistinguishability operator E_T associated to T , so that the degree of closeness or similarity between two fuzzy relations R and S is calculated aggregating the similarity of their respective entries using the quasi-arithmetic mean generated by an additive generator of T .

Also the euclidean metric will be used as an alternative method to compare fuzzy relations.

Trying to find the closest E to R is very expensive. Indeed, if n is the cardinality of the universe X , the transitivity of T -indistinguishability operators can be modeled by $3\binom{n}{3}$ inequalities and they lay in the region of the $\binom{n}{2}$ -dimensional space defined by them. The calculation of E becomes then a non-linear programming problem. Therefore, simpler methods to find a close E to R are desirable.

There are several algorithms to find the transitive closure \bar{R} of a proximity relation R and it is well known that $\bar{R} \geq R$. There are also algorithms to find maximal T -indistinguishability operators B among the set of T -indistinguishability operators smaller or equal than R and also the Representation Theorem gives a T -indistinguishability operator \underline{R} smaller or equal than R . It appears reasonable to aggregate \bar{R} and B or \underline{R} to obtain a new T -indistinguishability operator closer to R than \bar{R} , B or \underline{R} . This idea will be developed in Section 3.

If E is a T -indistinguishability operator, then the powers $E^{(p)}$ $p > 0$ are T -indistinguishability operators as well. This allows us to increase or decrease the values of E , since $E^{(p)} \leq E^{(q)}$ for $p \geq q$. So, we can decrease the values of the transitive closure or increase the ones of an operators smaller than R to find better approximations of it. Section 4 is devoted to this idea.

2 Preliminaries

This Section contains some results on t-norms and indistinguishability operators that will be needed later on in the paper. Besides well known definitions and theorems, the power T^n of a t-norm is generalized to irrational exponents in Definition 2.2. and given explicitly for continuous archimedean t-norms in Proposition 2.3.

Though many results remain valid for arbitrary t-

norms and especially for left continuous ones, for the sake of simplicity we will assume continuity for the t-norms throughout the paper.

Since a t-norm T is associative, we can extend it to an n -ary operation in the standard way:

$$T(x) = x$$

$$T(x_1, x_2, \dots, x_n) = T(x_1, T(x_2, \dots, x_n)).$$

In particular, $T(\overbrace{x, x, \dots, x}^{n \text{ times}})$ will be denoted by $x_T^{(n)}$ or simply by $x^{(n)}$ if the t-norm is clear.

If T is continuous, the n -th root $x_T^{(\frac{1}{n})}$ of x wrt T is defined by

$$x_T^{(\frac{1}{n})} = \sup\{z \in [0, 1] \mid z_T^{(n)} \leq x\}$$

and for $m, n \in N$, $x_T^{(\frac{m}{n})} = \left(x_T^{(\frac{1}{n})}\right)_T^{(m)}$.

Lemma 2.1. [8] If $k, m, n \in N$, $k, n \neq 0$ then $x_T^{(\frac{km}{kn})} = x_T^{(\frac{m}{n})}$.

Assuming continuity for the t-norm T , the powers $x_T^{(\frac{m}{n})}$ can be extended to irrational exponents in a straightforward way.

Definition 2.2. If $r \in R^+$ is a positive real number, let $\{a_n\}_{n \in N}$ be a sequence of rational numbers with $\lim_{n \rightarrow \infty} a_n = r$. For any $x \in [0, 1]$, the power $x_T^{(r)}$ is

$$x_T^{(r)} = \lim_{n \rightarrow \infty} x_T^{(a_n)}.$$

Continuity assures the existence of last limit and independence of the sequence $\{a_n\}_{n \in N}$.

Proposition 2.3. Let T be an archimedean t-norm with additive generator t , $x \in [0, 1]$ and $r \in R^+$. Then

$$x_T^{(r)} = t^{[-1]}(rt(x)).$$

Proof. Due to continuity of t we need to prove it only for rational r .

If r is a natural number m , then trivially $x_T^{(m)} = t^{[-1]}(mt(x))$.

If $r = \frac{1}{n}$ with $n \in N$, then $x_T^{(\frac{1}{n})} = z$ with $z_T^{(n)} = x$ or $t^{[-1]}(nt(z)) = x$ and $x_T^{(\frac{1}{n})} = t^{[-1]} \left(\frac{t(x)}{n} \right)$.

For a rational number $\frac{m}{n}$,

$$x_T^{(\frac{m}{n})} = \left(x_T^{(\frac{1}{n})}\right)_T^{(m)} = t^{[-1]} \left(mt \left(x_T^{(\frac{1}{n})} \right) \right) =$$

$$t^{[-1]} \left(mt \left(t^{[-1]} \left(\frac{t(x)}{n} \right) \right) \right) = t^{[-1]} \left(\frac{m}{n} t(x) \right).$$

□

Definition 2.4. The residuation \vec{T} of a t-norm T is defined by

$$\vec{T}(x|y) = \sup\{\alpha \in [0, 1] \mid T(x, \alpha) \leq y\}.$$

Definition 2.5. The natural T -indistinguishability E_T associated to a given t-norm T is the fuzzy relation on $[0, 1]$ defined by

$$E_T(x, y) = T(\vec{T}(x|y), \vec{T}(y|x)).$$

E_T is indeed a special kind of T -indistinguishability operator (Definition 2.6) [2] and in a logical context where T plays the role of the conjunction, E_T is interpreted as the bi-implication associated to T [5].

Definition 2.6. Given a t-norm T , a T -indistinguishability operator E on a set X is a fuzzy relation $E : X \times X \rightarrow [0, 1]$ satisfying for all $x, y, z \in X$

1. $E(x, x) = 1$ (Reflexivity)
2. $E(x, y) = E(y, x)$ (Symmetry)
3. $T(E(x, y), E(y, z)) \leq E(x, z)$ (T -transitivity).

Example 2.7.

1. If T is the Lukasiewicz t-norm, then $E_T(x, y) = 1 - |x - y|$ for all $x, y \in [0, 1]$.
2. If T is the Product t-norm, then $E_T(x, y) = \text{Min}\left(\frac{x}{y}, \frac{y}{x}\right)$ for all $x, y \in [0, 1]$ where $\frac{z}{0} = 1$.
3. If T is the Minimum t-norm, then $E_T(x, y) = \begin{cases} \text{Min}(x, y) & \text{if } x \neq y \\ 1 & \text{otherwise.} \end{cases}$

Theorem 2.8. Representation Theorem [11]. Let R be a fuzzy relation on a set X and T a continuous t-norm. R is a T -indistinguishability operator if and only if there exists a family $(h_i)_{i \in I}$ of fuzzy subsets of X such that for all $x, y \in X$

$$R(x, y) = \inf_{i \in I} E_T(h_i(x), h_i(y)).$$

$(h_i)_{i \in I}$ is called a generating family of R .

In particular, given a proximity matrix or relation R on X (i.e. a reflexive and symmetric fuzzy relation), we can build the T -indistinguishability operator \underline{R} generated by the set of the columns of R (i.e. the fuzzy subsets $R(x, \cdot)$, $x \in X$).

Proposition 2.9. $\underline{R} \leq R$.

Definition 2.10. Let R be a proximity matrix or relation (i.e. a reflexive and symmetric fuzzy relation) on X and T a continuous t -norm. The T -transitive closure \overline{R} of R is the smallest T -indistinguishability operator on X satisfying $R \leq \overline{R}$.

Definition 2.11. Let R and S be two fuzzy relations on X and T a continuous t -norm. The Sup- T product of R and S is the fuzzy relation $R \circ S$ on X defined for all $x, y \in X$ by

$$(R \circ S)(x, y) = \sup_{z \in X} T(R(x, z), S(z, y)).$$

Since the Sup- T product is associative or continuous t -norms, we can define for $n \in \mathbb{N}$ the n^{th} power R_T^n of a fuzzy relation R :

$$R_T^n = \overbrace{R \circ \dots \circ R}^{n \text{ times}}.$$

Definition 2.12. Let R be a fuzzy relation on a set X and T a continuous t -norm. The transitive closure of R with respect to T is the fuzzy relation

$$R_T = \sup_{n \in \mathbb{N}} R_T^n.$$

Proposition 2.13. Let R be a proximity relation on a finite set X of cardinality n . Then

$$R^T = \sup_{s \in \{1, \dots, n-1\}} R_T^s.$$

3 Aggregating the transitive closure and a T -indistinguishability smaller than R

Given a proximity relation R on X , it is necessary in many cases to replace it by a T -indistinguishability operator E , since T -transitivity is required. In these cases, we want to find E close to R , where the closeness or similarity between fuzzy relations can be defined in many different ways.

Let X be a finite set of cardinality n . Ordering its elements linearly, we can view the fuzzy subsets of X as vectors: $X = \{x_1, \dots, x_n\}$ and a fuzzy set h is the vector $(h(x_1), \dots, h(x_n))$. A proximity relation R on X can be represented by a matrix (also called R) determined by the $\binom{n}{2}$ entries r_{ij} $1 \leq i < j \leq n$ of R above the diagonal.

Proposition 3.1. Let $E = (e_{ij})_{i,j=1,\dots,n}$ be a proximity matrix on a set X of cardinality n and T a continuous archimedean t -norm with additive generator t . E is a T -indistinguishability operator if and only if for all i, j, k $1 \leq i < j < k \leq n$

$$\begin{aligned} t(e_{ij}) + t(e_{jk}) &\geq t(e_{ik}) \\ t(e_{ij}) + t(e_{ik}) &\geq t(e_{jk}) \\ t(e_{ik}) + t(e_{jk}) &\geq t(e_{ij}) \end{aligned}$$

Example 3.2. If T is the Lukasiewicz t -norm, then we can take $t(x) = 1 - x$ and last inequalities become

$$\begin{aligned} e_{ij} + e_{jk} - e_{ik} &\leq 1 \\ e_{ij} + e_{ik} - e_{jk} &\leq 1 \\ e_{ik} + e_{jk} - e_{ij} &\leq 1 \end{aligned}$$

Example 3.3. If T is the Product t -norm, then we can take $t(x) = -\log(x)$ and last inequalities become

$$\begin{aligned} e_{ij} \cdot e_{jk} &\leq e_{ik} \\ e_{ij} \cdot e_{ik} &\leq e_{jk} \\ e_{ik} \cdot e_{jk} &\leq e_{ij} \end{aligned}$$

Given a proximity matrix R , we must then search for (one of) the closest matrices E satisfying the last $3\binom{n}{3}$ inequalities which is a non-linear programming problem.

Instead of this, we propose alternative methods to obtain not the best but reasonably good approximations of proximity relations by T -indistinguishability operators.

Definition 3.4. [1], [8] Given a continuous monotonic map $t : [0, 1] \rightarrow [-\infty, \infty]$ and p, q positive integers with $p + q = 1$, the weighted quasi-arithmetic mean $m_t^{p,q}$ generated by t and weights p and q is defined for all $x, y \in [0, 1]$ by

$$m_t^{p,q}(x, y) = t^{-1}(p \cdot t(x) + q \cdot t(y)).$$

m_t is continuous if and only if $\text{Ran } t \neq [-\infty, \infty]$.

Proposition 3.5. Fixed the weights p and q , the map assigning to every continuous Archimedean t -norm T with generator t the weighted mean $m_t^{p,q}$ generated by t is a bijection between the set of continuous Archimedean t -norms and the set of continuous quasi-arithmetic means with these weights.

Proposition 3.6. Let T be a continuous archimedean t -norm with additive generator t , $p \in [0, 1]$ and E, F two T -indistinguishability operator on X . The weighted quasi-arithmetic mean $m_t^{p,1-p}$ with weights p and $1 - p$ of E and F is a T -indistinguishability operator.

Thanks to this last proposition, given a proximity matrix R we can calculate its transitive closure \overline{R} and a smaller T -indistinguishability operator than R , for example \underline{R} and find the weights $p, 1 - p$ to obtain the closest average of \overline{R} and \underline{R} to R .

The similarity between two fuzzy relations on X will be calculated in the following way.

Definition 3.7. Let T be a continuous archimedean t -norm with additive generator t and R, S two fuzzy

relations on a finite set X of cardinality n . The degree $DS(R, S)$ of similarity or closeness between R and S is defined by

$$DS(R, S) = t^{-1} \left(\frac{\sum_{1 \leq i, j \leq n} |t(r_{ij}) - t(s_{ij})|}{n} \right).$$

Proposition 3.8. DS is a T -indistinguishability operator on the set of fuzzy relations on X .

Corollary 3.9. Let $R = (r_{ij})$ be a proximity matrix on a finite set X of cardinality n , T a continuous archimedean t -norm with additive generator t , $\bar{R} = (\bar{r}_{i,j})$ its transitive closure, $\underline{R} = (\underline{r}_{i,j})$ the T -indistinguishability operator obtained from R with the Representation Theorem, $p \in [0, 1]$ and $m_t^{p, 1-p}(\bar{R}, \underline{R})$ the T -indistinguishability operator quasi-arithmetic mean of \bar{R} and \underline{R} with weights p and $1 - p$. Then

$$DS(R, m_t^{p, 1-p}(E, F)) = t^{-1} \left(\frac{\sum_{1 \leq i, j \leq n} |p \cdot t(\bar{r}_{ij}) + (1-p) \cdot t(\underline{r}_{ij}) - t(r_{ij})|}{n} \right).$$

We are looking for the value (or values) of p that maximize the last equality. Since t^{-1} is a decreasing map, this is equivalent to minimize

$$\sum_{1 \leq i, j \leq n} |p \cdot t(\bar{r}_{ij}) + (1-p) \cdot t(\underline{r}_{ij}) - t(r_{ij})|$$

and, since R is reflexive and symmetric, is equivalent to minimize

$$f(p) = \sum_{1 \leq i < j \leq n} |p \cdot t(\bar{r}_{ij}) + (1-p) \cdot t(\underline{r}_{ij}) - t(r_{ij})|$$

Proposition 3.10. Let $f_1, \dots, f_n : [0, 1] \rightarrow R$ be n concave functions. Then $\sum_1^n f_i$ is a concave function.

Proof. By definition, given two points x_1, x_2 of $[0, 1]$, the segments joining their images by f_i $i = 1, \dots, n$ are above f_i . $\sum_1^n f_i$ will then be below the sum of all the segments. \square

Corollary 3.11. $f(p)$ is a concave function.

Proof. Each summand $|p \cdot t(\bar{r}_{ij}) + (1-p) \cdot t(\underline{r}_{ij}) - t(r_{ij})|$ of f is a concave function. \square

Proposition 3.12. The set of minima of $f(p)$ consists of a single point or of a closed interval.

Proof. f is a concave function and its graphic is a polygonal line. \square

Proposition 3.13. The computation of $m_t(\bar{R}, \underline{R})$ with maximum $DS(R, m_t(\bar{R}, \underline{R}))$ can be done taking $O(n^3)$ time complexity.

Proof:

The computation of \bar{R} and \underline{R} can be done in $O(n^3)$ complexity time [9].

The addition (aggregation of distances) takes $O(n^2)$ time complexity.

The minimization of $f(p)$ takes at most $O(n^2)$ time complexity.

So the most complex part of this process is the computation of \bar{R} and \underline{R} , which still takes $O(n^3)$ complexity time.

Example 3.14. Let X be a set of cardinality 7 and R the proximity relation given by

$$R = \begin{pmatrix} 1 & 1 & 0.3 & 0.3 & 0.1 & 0.3 & 0.4 \\ 1 & 1 & 0.6 & 0.4 & 0.5 & 0.4 & 0.2 \\ 0.3 & 0.6 & 1 & 0.1 & 0.3 & 0.2 & 0.5 \\ 0.3 & 0.4 & 0.1 & 1 & 1 & 1 & 1 \\ 0.1 & 0.5 & 0.3 & 1 & 1 & 1 & 1 \\ 0.3 & 0.4 & 0.2 & 1 & 1 & 1 & 1 \\ 0.4 & 0.2 & 0.5 & 1 & 1 & 1 & 1 \end{pmatrix}.$$

Then, for T the Lukasiewicz t -norm,

$$\bar{R} = \begin{pmatrix} 1 & 1 & 0.6 & 0.4 & 0.5 & 0.4 & 0.4 \\ 1 & 1 & 0.6 & 0.5 & 0.5 & 0.5 & 0.5 \\ 0.6 & 0.6 & 1 & 0.5 & 0.5 & 0.5 & 0.5 \\ 0.4 & 0.5 & 0.5 & 1 & 1 & 1 & 1 \\ 0.5 & 0.5 & 0.5 & 1 & 1 & 1 & 1 \\ 0.4 & 0.5 & 0.5 & 1 & 1 & 1 & 1 \\ 0.4 & 0.5 & 0.5 & 1 & 1 & 1 & 1 \end{pmatrix}$$

and

$$\underline{R} = \begin{pmatrix} 1 & 0.6 & 0.3 & 0.1 & 0.1 & 0.1 & 0.1 \\ 0.6 & 1 & 0.3 & 0.2 & 0.1 & 0.2 & 0.2 \\ 0.3 & 0.3 & 1 & 0.1 & 0.1 & 0.1 & 0.1 \\ 0.1 & 0.2 & 0.1 & 1 & 0.8 & 0.9 & 0.6 \\ 0.1 & 0.1 & 0.1 & 0.8 & 1 & 0.8 & 0.7 \\ 0.1 & 0.2 & 0.1 & 0.9 & 0.8 & 1 & 0.7 \\ 0.1 & 0.2 & 0.1 & 0.6 & 0.7 & 0.7 & 1 \end{pmatrix}.$$

$$f(p) = |0.4p| + |0.3p - 0.3| + |0.3p - 0.1| + |0.4p - 0.4| + |0.3p - 0.1| + |0.3p| + |0.3p| + |0.3p - 0.1| + |0.4p| + |0.3p - 0.1| + |0.3p - 0.3| + |0.4p - 0.4| + |0.4p - 0.2| + |0.4p - 0.3| + |0.4p| + |0.2p| + |0.1p| + |0.4p| + |0.2p| + |0.3p| + |0.3p|$$

which attains its minimum for $p = \frac{1}{3}$.

A good T -transitive approximation of R (for T the Lukasiewicz t -norm) is then

$$\begin{pmatrix} 1 & 0.733 & 0.4 & 0.2 & 0.233 & 0.2 & 0.2 \\ 0.733 & 1 & 0.4 & 0.3 & 0.233 & 0.3 & 0.3 \\ 0.4 & 0.4 & 1 & 0.233 & 0.233 & 0.233 & 0.233 \\ 0.2 & 0.3 & 0.233 & 1 & 0.867 & 0.933 & 0.733 \\ 0.233 & 0.233 & 0.233 & 0.867 & 1 & 0.867 & 0.8 \\ 0.2 & 0.3 & 0.233 & 0.933 & 0.867 & 1 & 0.8 \\ 0.2 & 0.3 & 0.233 & 0.733 & 0.8 & 0.8 & 1 \end{pmatrix}.$$

Example 3.15. Let X be a set of cardinality 7 and R the proximity relation given by

$$R = \begin{pmatrix} 1 & 0.5 & 0.7 & 0.7 & 0.5 & 0.7 & 0.8 \\ 0.5 & 1 & 1 & 0.8 & 0.9 & 0.8 & 0.6 \\ 0.7 & 1 & 1 & 0.5 & 0.7 & 0.6 & 0.9 \\ 0.7 & 0.8 & 0.5 & 1 & 0.5 & 0.5 & 0.5 \\ 0.5 & 0.9 & 0.7 & 0.5 & 1 & 0.5 & 0.5 \\ 0.7 & 0.8 & 0.6 & 0.5 & 0.5 & 1 & 0.5 \\ 0.8 & 0.6 & 0.9 & 0.5 & 0.5 & 0.5 & 1 \end{pmatrix}.$$

Then, for T the Product t -norm,

$$\bar{R} = \begin{pmatrix} 1 & 0.7 & 0.72 & 0.7 & 0.5 & 0.7 & 0.8 \\ 0.7 & 1 & 1 & 0.8 & 0.9 & 0.8 & 0.9 \\ 0.72 & 1 & 1 & 0.8 & 0.9 & 0.8 & 0.9 \\ 0.7 & 0.8 & 0.8 & 1 & 0.72 & 0.64 & 0.56 \\ 0.5 & 0.9 & 0.9 & 0.72 & 1 & 0.72 & 0.63 \\ 0.7 & 0.8 & 0.8 & 0.64 & 0.72 & 1 & 0.56 \\ 0.8 & 0.9 & 0.9 & 0.56 & 0.63 & 0.56 & 1 \end{pmatrix}$$

and

$$\underline{R} = \begin{pmatrix} 1 & 0.5 & 0.5 & 0.625 & 0.5 & 0.625 & 0.714 \\ 0.5 & 1 & 0.625 & 0.5 & 0.625 & 0.555 & 0.555 \\ 0.5 & 0.625 & 1 & 0.5 & 0.555 & 0.555 & 0.6 \\ 0.625 & 0.5 & 0.5 & 1 & 0.5 & 0.5 & 0.5 \\ 0.5 & 0.625 & 0.555 & 0.5 & 1 & 0.5 & 0.5 \\ 0.625 & 0.555 & 0.555 & 0.5 & 0.5 & 1 & 0.5 \\ 0.714 & 0.555 & 0.6 & 0.5 & 0.5 & 0.5 & 1 \end{pmatrix}.$$

$f(p)$ attains its minimum for $p = 0.521$.

A good T -transitive approximation of R (for T the Product t -norm) is then

$$\begin{pmatrix} 1 & 0.587 & 0.595 & 0.660 & 0.5 & 0.660 & 0.754 \\ 0.587 & 1 & 0.783 & 0.626 & 0.744 & 0.662 & 0.700 \\ 0.595 & 0.783 & 1 & 0.626 & 0.700 & 0.662 & 0.729 \\ 0.660 & 0.626 & 0.626 & 1 & 0.595 & 0.563 & 0.528 \\ 0.5 & 0.744 & 0.700 & 0.595 & 1 & 0.595 & 0.559 \\ 0.660 & 0.662 & 0.662 & 0.563 & 0.595 & 1 & 0.528 \\ 0.754 & 0.700 & 0.729 & 0.528 & 0.559 & 0.528 & 1 \end{pmatrix}.$$

The degree of closeness between two fuzzy relations can also be calculated using the euclidean distance.

Definition 3.16. Let $R = (r_{ij})$ and $S = (s_{ij})$ be two fuzzy relations on a finite set X of cardinality n . The euclidean distance D between R and S is

$$D(R, S) = \left(\sum_{1 \leq i, j \leq n} (r_{ij} - s_{ij})^2 \right)^{\frac{1}{2}}$$

Corollary 3.17. Let $R = (r_{ij})$ be a proximity matrix on a finite set X of cardinality n , T a continuous archimedean t -norm with additive generator t , $\bar{R} = (\bar{r}_{i,j})$ its transitive closure, $\underline{R} = (\underline{r}_{ij})$ the T -indistinguishability operator obtained from \bar{R} with the Representation Theorem, $p \in [0, 1]$ and $m_t(\bar{R}, \underline{R})$ the T -indistinguishability operator quasi-arithmetic mean of \bar{R} and \underline{R} with weights p and $1 - p$. Then

$$D(R, m_t(E, F)) =$$

$$\left(\sum_{1 \leq i, j \leq n} (t^{-1}(p \cdot t(\bar{r}_{ij}) + (1-p) \cdot t(\underline{r}_{ij})) - t(r_{ij}))^2 \right)^{\frac{1}{2}}.$$

Proposition 3.18. Let T be the Lukasiewicz t -norm and R a proximity on a set X of cardinality n . The closest $m_t(\bar{R}, \underline{R})$ to R is attained for

$$p = \frac{\sum_{1 \leq i < j \leq n} (\bar{r}_{ij} - \underline{r}_{ij}) (r_{ij} - \underline{r}_{ij})}{\sum_{1 \leq i < j \leq n} (\bar{r}_{ij} - \underline{r}_{ij})^2}$$

Proof. Due to symmetry and reflexivity, it is enough to minimize

$$f(p) = \sum_{1 \leq i < j \leq n} (p(\bar{r}_{ij} - \underline{r}_{ij}) + \underline{r}_{ij} - r_{ij})^2.$$

$$f'(p) = 2 \sum_{1 \leq i < j \leq n} (p(\bar{r}_{ij} - \underline{r}_{ij}) + \underline{r}_{ij} - r_{ij}) (\underline{r}_{ij} - r_{ij}) = 0$$

and

$$p = \frac{\sum_{1 \leq i < j \leq n} (\bar{r}_{ij} - \underline{r}_{ij}) (r_{ij} - \underline{r}_{ij})}{\sum_{1 \leq i < j \leq n} (\bar{r}_{ij} - \underline{r}_{ij})^2}.$$

□

Example 3.19. Let X be a set of cardinality 4 and R the proximity relation on X given by

$$R = \begin{pmatrix} 1 & 0.8 & 0.2 & 0.4 \\ 0.8 & 1 & 0.7 & 0.1 \\ 0.2 & 0.7 & 1 & 0.6 \\ 0.4 & 0.1 & 0.6 & 1 \end{pmatrix}.$$

If T is the Lukasiewicz t -norm, the closest T -indistinguishability operator of the type $m_t(\bar{R}, \underline{R})$ (with respect to the euclidean distance) is attained for $p = 0.638889$.

A good T -approximation of R is then

$$\begin{pmatrix} 1 & 0.6917 & 0.3917 & 0.3639 \\ 0.6917 & 1 & 0.5917 & 0.2278 \\ 0.3917 & 0.5917 & 1 & 0.5278 \\ 0.3639 & 0.2278 & 0.5278 & 1 \end{pmatrix}$$

4 Applying a homotecy to a T -indistinguishability operator

In this Section, the fact that the power of a T -indistinguishability operator is again a T -indistinguishability operator will be exploited to modify the entries of \bar{R} or \underline{R} to find a better approximation of R .

Proposition 4.1. *Let T be a continuous t -norm, E a T -indistinguishability operator on X and $p > 0$. Then $E^{(p)}$ is a T -indistinguishability operator.*

Example 4.2.

- If T is a continuous archimedean t -norm with additive generator t and E a T -indistinguishability operator, then $t^{[-1]}(p \cdot t(E))$ is a T -indistinguishability operator.
- If T is the Lukasiewicz t -norm and E a T -indistinguishability operator, then $\text{Max}(0, 1 - p + p \cdot E)$ is a T -indistinguishability operator.
- If T is the Product t -norm and E a T -indistinguishability operator, then E^p is a T -indistinguishability operator.

Let $R = (r_{ij})$ be a proximity matrix on a set X of cardinality n , $p > 0$ and $E = (e_{ij})$ a T -indistinguishability operator on X with T a continuous archimedean t -norm with additive generator t . Then

$$DS(R, E^{(p)}) = t^{-1} \left(\frac{\sum_{1 \leq i, j \leq n} |t(r_{ij}) - p \cdot t(e_{ij})|}{n} \right).$$

To maximize the previous expression is equivalent to minimize

$$\sum_{1 \leq i, j \leq n} |t(r_{ij}) - p \cdot t(e_{ij})|.$$

Since R is reflexive and symmetric, this is equivalent to minimize

$$g(p) = \sum_{1 \leq i < j \leq n} |t(r_{ij}) - p \cdot t(e_{ij})|.$$

Again g is a sum of concave functions in $[0, 1]$ and therefore has a minimum or a close interval of minima.

Example 4.3. *Let us consider the same matrix*

$$R = \begin{pmatrix} 1 & 0.8 & 0.2 & 0.4 \\ 0.8 & 1 & 0.7 & 0.1 \\ 0.2 & 0.7 & 1 & 0.6 \\ 0.4 & 0.1 & 0.6 & 1 \end{pmatrix}.$$

Then, for T the Lukasiewicz t -norm,

$$\underline{R} = \begin{pmatrix} 1 & 0.5 & 0.2 & 0.3 \\ 0.5 & 1 & 0.4 & 0.1 \\ 0.2 & 0.4 & 1 & 0.4 \\ 0.3 & 0.1 & 0.4 & 1 \end{pmatrix}.$$

$$g(p) = |0.5 \cdot p - 0.2| + |0.8 \cdot p - 0.8| + |0.7 \cdot p - 0.6| + |0.6 \cdot p - 0.3| + |0.9 \cdot p - 0.9| + |0.6 \cdot p - 0.4|$$

which attains its minimum for $p = 0.857$.

A good approximation of R is then

$$\underline{R}^{(0.857)} = \begin{pmatrix} 1 & 0.5715 & 0.3144 & 0.4001 \\ 0.5715 & 1 & 0.4858 & 0.2287 \\ 0.3144 & 0.4858 & 1 & 0.4858 \\ 0.4001 & 0.2287 & 0.4858 & 1 \end{pmatrix}.$$

If we consider the euclidean distance between R and the power $E^{(p)}$ of a T -indistinguishability operator $E = (e_{ij})$, then

Proposition 4.4.

$$D(R, E^{(p)}) = \left(\sum_{1 \leq i, j \leq n} (t^{-1}(p \cdot t(e_{ij})) - r_{ij})^2 \right)^{\frac{1}{2}}.$$

Example 4.5. *Continuing the last example, $D(R, \bar{R}^{(p)})$ is maximum for $p = 1.208633$ and $D(R, \underline{R}^{(p)})$ is maximum for $p = 0.821306$.*

Good approximations of R are therefore

$$\bar{R}^{(1.208633)} = \begin{pmatrix} 1 & 0.7583 & 0.3957 & 0.2748 \\ 0.7583 & 1 & 0.6374 & 0.1540 \\ 0.3957 & 0.6374 & 1 & 0.5165 \\ 0.2748 & 0.1540 & 0.5165 & 1 \end{pmatrix}.$$

and

$$\underline{R}^{(0.821306)} = \begin{pmatrix} 1 & 0.8357 & 0.5893 & 0.5072 \\ 0.8357 & 1 & 0.7536 & 0.4251 \\ 0.5893 & 0.7536 & 1 & 0.6715 \\ 0.5072 & 0.4251 & 0.6715 & 1 \end{pmatrix}.$$

5 Concluding Remarks

In this paper we have presented two ways to find good approximations of a proximity relation by T -transitive ones (T archimedean) in a reasonable computational way.

The obtained approximation R' is in general not comparable with R in the sense that neither $R' \geq R$ nor $R \geq R'$ must hold.

The simple examples show that in general these approximations are better than the transitive closure or openings of the proximity R .

The methods of the paper cannot be applied to the Minimum t-norm. Other ways to obtain similar results for this t-norm are therefore needed and the authors will work on it in forthcoming papers.

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