

9. Similarity Relations

Example 9.1

Specification of a partial control mapping (“good control actions“)

		gradient						
		-40	-6	-3	0	3	6	40
deviation	-70	22,5	15,0	15,0	10,0	10,0	5,0	5,0
	-50	22,5	15,0	10,0	10,0	5,0	5,0	0,0
	-30	15,0	10,0	50,0	5,0	0,0	0,0	0,0
	0	5,0	5,0	0,0	0,0	0,0	-10,0	-15,0
	30	0,0	0,0	0,0	-5,0	-5,0	-10,0	-10,0
	50	0,0	-5,0	-5,0	-10,0	-15,0	-15,0	-22,5
	70	-5,0	-5,0	-15,0	-15,0	-15,0	-15,0	-15,0
			current					

Interpolation of this table

Additional knowledge was available: Some values are indistinguishable (from a measurement point of view) or should be treated in a similar way.

Problem 1: How to model such similarity information?

Problem 2: How to interpolate in the case of existing similarity information?

How to model similarity?

Proposal 1: equivalence relation

- (I) $x \approx x$ (reflexivity)
- (II) $x \approx y \leftrightarrow y \approx x$ (symmetry)
- (III) $x \approx y \wedge y \approx z \rightarrow x \approx z$ (transitivity)

- x and y similar ($x \approx y$), if and only if $|x-y| < \varepsilon$, (ε fixed)

\approx is not transitive, Poincaré paradox

$x \approx y, y \approx z, x \not\approx z$

- counterintuitive !

Proposal 2: similarity relation (multi-valued equivalence relation)

$[x \approx y]$ degree, to which $x \approx y$ holds

$$E_{\approx}: X \times X \rightarrow [0,1], (x,y) \rightarrow [x \approx y]$$

$$(1) \quad E_{\approx}(x,x) = 1$$

$$(2) \quad E_{\approx}(x,y) = E_{\approx}(y,x)$$

$$(3) \quad \Pi(E_{\approx}(x,y), E_{\approx}(y,z)) \leq E_{\approx}(x,z), \text{ where } \Pi \text{ is a } t\text{-norm}$$

E_{\approx} is called fuzzy equality relation, similarity relation, indistinguishability operator or tolerance relation.

- **Example**

δ pseudo metric on X

$\Pi(\alpha, \beta) = \max\{\alpha + \beta - 1, 0\}$ Lukasiewicz t -norm,

Then

$E_{\delta}(x, y) = 1 - \min\{\delta(x, y), 1\}$ fuzzy equality relation

$\delta_E(x, y) = 1 - E(x, y)$ induced pseudo metric

i.e. fuzzy equality and distance are dual notions

- **Formal Definition**

$E: X \times X \rightarrow [0, 1]$ similarity relation, iff

(1) $E(x, x) = 1$

(2) $E(x, y) = E(y, x)$

(3) $\max\{(E(x, y) + E(y, z) - 1), 0\} \leq E(x, z)$

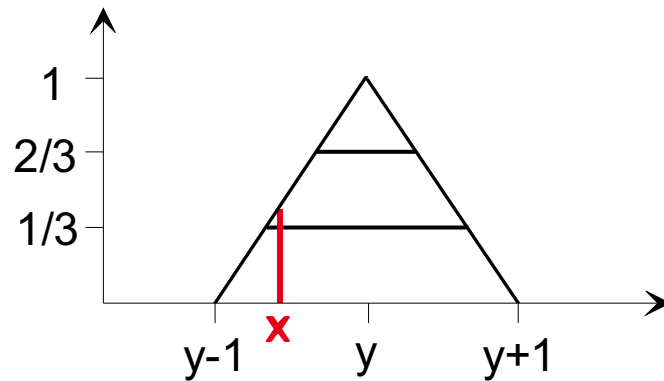
Fuzzy Sets as a Derived Concept

$$\delta(x,y) = |x-y|$$

metric

$$E_{\delta}(x,y) = 1 - \min \{|x-y|, 1\}$$

similarity relation



$$\mu_y: X \rightarrow [0,1]$$

$x \rightarrow E_{\delta}(x,y)$ fuzzy singleton

μ_y describes the “local“ similarities

Example 9.2

$$E : \mathcal{R}x\mathcal{R} \rightarrow [0,1], (x, y) \mapsto 1 - \min\{|x - y|, 1\}$$

is a similarity relation w.r.t the t-norm T_{Luka}

Def. 9.3

Let E be a similarity relation on X w.r.t. T .

$\mu \in F(X)$ is extensional iff $T(\mu(x), E(x, y)) \leq \mu(y)$ for all x, y .

Def. 9.4

Let E be a similarity relation on X w.r.t. T And $M \subseteq X$.

$$\mu_M : X \rightarrow [0,1], x \mapsto \sup\{E(x, x') \mid x' \in M\}$$

Is called extensional hull of M .

Example 9.5

A singleton is the extensional hull of $\{x_0\}$.

Def. 9.6

Let E, F be similarity relations on X and Y .

$\zeta: X \rightarrow Y$ is extensional with respect to E, F iff $E(x, x') \leq F(\zeta(x), \zeta(x'))$ holds.

Theorem 9.7

Let E_1, \dots, E_n similarity relations w.r.t. \prod on X_1, \dots, X_n .

Define $E: (X_1 \times \dots \times X_n)^2 \rightarrow [0, 1]$,

$((x_1, \dots, x_n), (y_1, \dots, y_n)) \mapsto \min\{E_1(x_1, y_1), \dots, E_n(x_n, y_n)\}$

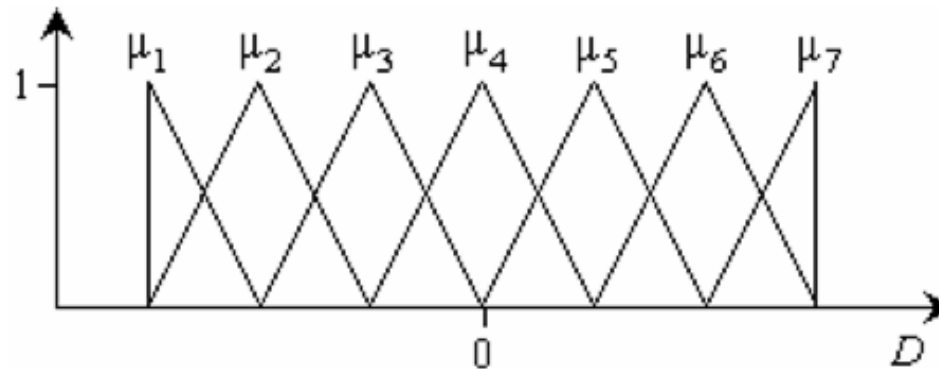
- E is a similarity relation w.r.t. \prod on $X_1 \times \dots \times X_n$.
- For all $i \in \{1, \dots, n\}$ the projection $\pi_i: X_1 \times \dots \times X_n \rightarrow X_i$ is extensional w.r.t. E and E_i .
- If E' is a similarity relation w.r.t. \prod on $X_1 \times \dots \times X_n$, i.e. all projections are extensional, then $E' \leq E$ holds.

Remark 9.8

E is the biggest similarity relation for which all projections are extensional.

Specification of similarity relations

given a family of fuzzy sets that describes the “local” similarities



then

- there is a similarity relation on D with induced fuzzy sets μ_i , iff

$$\sup_{x \in X} \{\mu_i(x) + \mu_j(x) - 1\} \leq \inf_{y \in X} \{1 - |\mu_i(y) - \mu_j(y)|\} \text{ for all } i, j$$

- if $\mu_i(x) + \mu_j(x) \leq 1$ for $i \neq j$ then there is a similarity relation E on X , where

$$E(x, y) = \inf_i \{1 - |\mu_i(x) - \mu_i(y)|\}$$

Necessity of scaling

Are there other similarity relations on the real numbers than

$$E(x,y) = 1 - \min\{|x-y|, 1\}?$$

Integration of scaling

A similarity relation is dependent on the measurement unit:

Celsius: $E(20^\circ\text{C}, 20.5^\circ\text{C}) = 0.5$ Fahrenheit: $E(68\text{F}, 68.9\text{F}) = 0.1$

scaling factor for Celsius/Fahrenheit: 1.8, since $F = 1.8C + 32$

$$E(x,y) = 1 - \min\{|c \cdot x - c \cdot y|, 1\}$$

$$X = [a, b]$$

Scaling

$$c: X \rightarrow [0, \infty),$$

Transformation

$$f: X \rightarrow [0, \infty), \quad x \rightarrow \int_a^x c(t) dt$$

Similarity relation

$$E: X \times X \rightarrow [0, 1], \quad (x, x') \rightarrow 1 - \min \{|f(x) - f(x')|, 1\}$$