

Algebraic Product is the Only t-Norm for Which Optimization Under Fuzzy Constraints is Scale-Invariant

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[Need for Optimization...](#)

[Bellman-Zadeh...](#)

[Problem: the Value...](#)

[What We Show in...](#)

[Definitions](#)

[Main Results](#)

[Proof of Proposition 1](#)

[Proof of Proposition 2](#)

[Acknowledgments](#)

[Home Page](#)

[Title Page](#)

[◀◀](#)

[▶▶](#)

[◀](#)

[▶](#)

[Page 1 of 11](#)

[Go Back](#)

[Full Screen](#)

[Close](#)

[Quit](#)

1. Need for Optimization under Fuzzy Constraints

- Example: we need to build a chemical plant for producing chemicals needed for space exploration.
- Among all designs x with small effect on environment, we need to select the most profitable one.
- Here, for each alternative x , we can compute the value $f(x)$ of the objective function.
- Constraints are formulated by using imprecise words from a natural language (like “small”).
- In fuzzy logic, to each alternative x , we assign degree $\mu_c(x)$ to which x is, e.g., small.
- E.g., if a user marks smallness of x by 7 on a scale 0 to 10, we take $\mu_c(x) = 7/10$.
- Problem: find x such that $f(x) \rightarrow \max$ under constraint $\mu_c(x)$.

2. Bellman-Zadeh Approach to Optimization under Fuzzy Constraints

- First, we find the smallest value m of the objective function $f(x)$ among all possible solutions x .
- Then, we find the largest possible value M of the objective function over all possible constraints.
- We form the degree to which x is maximal:

$$\mu_m(x) \stackrel{\text{def}}{=} \frac{f(x) - m}{M - m}.$$

- We want to find an alternative which satisfies the constraints *and* maximizes the objective function.
- In fuzzy techniques, the degree of truth in “and”-statement is described by an appropriate *t-norm* $f(a, b)$.
- So, we select x for which the degree $\mu_s(x) = f_{\&}(\mu_c(x), \mu_m(x))$ is the largest.

3. Problem: the Value M is Not Well Defined

- Usually, we have experience with similar problems, so we know previously selected alternative(s) x .
- The value $f(x)$ for such “status quo” alternatives can be used as the desired minimum m .
- Finding M is much more complicated, we do not know which alternatives to include and which not to include.
- If we replace the original value M with a new value $M' > M$, then the maximizing degree changes:

$$\mu_m(x) = \frac{f(x) - m}{M - m} \rightarrow \mu'_m(x) = \frac{f(x) - m}{M' - m}.$$

- Here, $\mu'_m(x) = \lambda \cdot \mu_m(x)$ for $\lambda \stackrel{\text{def}}{=} \frac{M - m}{M' - m} < 1$.
- In general, diff. alternatives $\max \mu_s(x) = f_{\&}(\mu_c(x), \mu_m(x))$ and $\mu'_s(x) = f_{\&}(\mu_c(x), \mu'_m(x)) = f_{\&}(\mu_c(x), \lambda \cdot \mu_m(x))$.

4. What We Show in This Talk

In this paper:

- we show that the dependence on M disappears if we use algebraic product t-norm $f_{\&}(a, b) = a \cdot b$.
- We also show that this is the only t-norm for which decisions do not depend on M .

Need for Optimization...

Bellman-Zadeh...

Problem: the Value...

What We Show in...

Definitions

Main Results

Proof of Proposition 1

Proof of Proposition 2

Acknowledgments

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 5 of 11

Go Back

Full Screen

Close

Quit

5. Definitions

- By a t-norm, we mean a f-n $f_{\&} : [0, 1] \times [0, 1] \rightarrow [0, 1]$ s.t. $f_{\&}(a, b) = f_{\&}(b, a)$ and $f_{\&}(1, a) = a$ for all a, b .
- We say that optimization under fuzzy constraints is scale-invariant for the t-norm $f_{\&}(a, b)$ if
 - for every set X , for every two functions $\mu_c : X \rightarrow [0, 1]$ and $\mu_m : X \rightarrow [0, 1]$, and
 - for every real number $\lambda \in (0, 1)$,

we have $S = S'$, where:

- S is the set of all $x \in X$ for which the function $\mu_s(x) = f_{\&}(\mu_c(x), \mu_m(x))$ attains its maximum;
- S' is the set of all $x \in X$ for which the function $\mu'_s(x) = f_{\&}(\mu_c(x), \lambda \cdot \mu_m(x))$ attains its maximum.

6. Main Results

- **Proposition 1.** *For the algebraic product t -norm $f_{\&}(a, b) = a \cdot b$, optimization under fuzzy constraints is scale-invariant.*
- **Proposition 2.** *$a \cdot b$ is the only t -norm for which optimization under fuzzy constraints is scale-invariant.*
- It is usually required that the t -norm is associative.
- Our result does not require associativity, so we can apply to non-associative and-operations.
- Such operations sometimes more adequately represent human reasoning (Zimmermann, Zysno).

Need for Optimization...

Bellman-Zadeh...

Problem: the Value...

What We Show in...

Definitions

Main Results

Proof of Proposition 1

Proof of Proposition 2

Acknowledgments

Home Page

Title Page



Page 7 of 11

Go Back

Full Screen

Close

Quit

7. Proof of Proposition 1

- For the algebraic product t-norm $f_{\&}(a, b) = a \cdot b$:
 - S is the set of all $x \in X$ for which the function $\mu_s(x) = \mu_c(x) \cdot \mu_m(x)$ attains its maximum, and
 - S' is the set of all $x \in X$ for which the function $\mu'_s(x) = \mu_c(x) \cdot \lambda \cdot \mu_m(x)$ attains its maximum.
- Here, $\mu'_s(x) = \lambda \cdot \mu_s(x)$ for a positive number λ .
- Clearly, $\mu_s(x) \geq \mu_s(y)$ if and only if $\lambda \cdot \mu_s(x) \geq \lambda \cdot \mu_s(y)$.
- So the optimizing sets S and S' for $\mu_s(x)$ and $\mu'_s(x) = \lambda \cdot \mu_s(x)$ indeed coincide.

8. Proof of Proposition 2

- Let $f_{\&}(a, b)$ be a t-norm for which optimization under fuzzy constraints is scale-invariant.
- Let $a, b \in [0, 1]$; let us prove that $f_{\&}(a, b) = a \cdot b$.
- Let us consider $X = \{x_1, x_2\}$ with $\mu_c(x_1) = \mu_m(x_2) = a$ and $\mu_c(x_2) = \mu_m(x_1) = 1$.
- Here, $\mu_s(x_1) = f_{\&}(\mu_c(x_1), \mu_m(x_1)) = f_{\&}(a, 1) = a$.
- Similarly, $\mu_s(x_2) = f_{\&}(\mu_c(x_2), \mu_m(x_2)) = f_{\&}(1, a) = a$.
- Since $\mu_s(x_1) = \mu_s(x_2)$, the optimizing set S consists of both elements x_1 and x_2 .
- Due to scale-invariance, for $\lambda = b$, $S' = S = \{x_1, x_2\}$ is optimizing for $\mu'_s(x) = f_{\&}(\mu_c(x), \lambda \cdot \mu_m(x))$.
- Thus, $\mu'_s(x_1) = \mu'_s(x_2)$, i.e., $f_{\&}(a, b \cdot 1) = f_{\&}(1, b \cdot a)$.
- So, $f_{\&}(a, b) = f_{\&}(1, b \cdot a) = a \cdot b$. Q.E.D.

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Need for Optimization...

Bellman-Zadeh...

Problem: the Value...

What We Show in...

Definitions

Main Results

Proof of Proposition 1

Proof of Proposition 2

Acknowledgments

Home Page

Title Page

◀◀

▶▶

◀

▶

Page 10 of 11

Go Back

Full Screen

Close

Quit

10. Bibliography

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[Need for Optimization...](#)[Bellman-Zadeh...](#)[Problem: the Value...](#)[What We Show in...](#)[Definitions](#)[Main Results](#)[Proof of Proposition 1](#)[Proof of Proposition 2](#)[Acknowledgments](#)[Home Page](#)[Title Page](#)[Page 11 of 11](#)[Go Back](#)[Full Screen](#)[Close](#)[Quit](#)