

## A Note on Belief Structures and S-approximation Spaces

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**ABSTRACT.** We study relations between evidence theory and S-approximation spaces. Both theories have their roots in the analysis of Dempster's multivalued mappings and lower and upper probabilities, and have close relations to rough sets. We show that an S-approximation space, satisfying a monotonicity condition, can induce a natural belief structure which is a fundamental block in evidence theory. We also demonstrate that one can induce a natural belief structure on one set, given a belief structure on another set, if the two sets are related by a partial monotone S-approximation space.

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## 1. INTRODUCTION

**DEMPSTER-SHAFER THEORY OF EVIDENCE.** The Dempster-Shafer theory of evidence is a well-known method in dealing with uncertainty in problems. It originated in 1967 with the introduction of lower and upper probabilities by Dempster [2]. A *belief structure* is a fundamental concept in this theory which assigns two numeric values to each subset of a given set. These values are known as the *belief* and *plausibility* measures. See [8] for a detailed treatment.

**S-APPROXIMATION.** S-approximation spaces are a new way of handling uncertainty, which also originated from Dempster's concepts of lower and upper probabilities [4]. The motivation for this new approach is that it can be seen as a unifying view to rough sets and their extensions, such as [1, 6, 7, 13, 15, 16, 20], since they are all expressible in terms of S-approximation spaces [12, 4]. Hence, any results obtained over S-approximations can be naturally applied to rough sets and many of their extensions, too<sup>‡</sup>. However, S-approximations are capable of representing more than (extensions of) rough sets and model a very broad range of possible approximations (See [4, 12] for more examples).

**PREVIOUS WORKS ON S-APPROXIMATIONS.** The concept of S-approximation has been studied by several approaches and its relation to various theories have been examined. For example, S-approximations are studied in the context of Yao's three-way decisions theory [18, 10] and extended its results. Moreover, they have also been studied in the contexts of neighborhood systems [17, 9], intuitionistic fuzzy set theory [11] and with relations to topology [3].

**MOTIVATION.** Given the common background and overlap of goals, connections between evidence theory and other theories of approximation have been studied for a long time, e.g. its connections to the theory of rough sets are considered in [5, 14, 19]. The close links between S-approximation spaces and rough sets suggest that a study of relations between evidence theory and S-approximation spaces can yield to more general variants of these results. In this work, we obtain such results about the connections between evidence theory and S-approximation spaces and propose paths for future research.

**ORGANIZATION.** The paper is organized as follows: In Section 2, we first review some basic facts from evidence theory, S-approximation spaces, and their corresponding three-way decisions. Then, we study the connection between S-approximation spaces and evidence theory in Section 3. Finally, the paper concludes in Section 4 by suggesting interesting directions for future research.

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<sup>‡</sup>This includes all the results reported in the current paper.

## 2. PRELIMINARIES

**2.1. Dempster-Shafer Theory of Evidence.** In this section, we briefly discuss some background on the Dempster-Shafer theory of evidence. We follow the standard presentation in [8].

**BASIC PROBABILITY ASSIGNMENTS.** A *basic probability assignment*, or *bpa* for short, is a fundamental concept in evidence theory. Let  $W$  be a finite non-empty set. Then, a bpa over  $W$  is a mapping  $m : \mathcal{P}(W) \rightarrow [0, 1]$  satisfying the following conditions: (a)  $m(\emptyset) = 0$ , and (b)  $\sum_{X \subseteq W} m(X) = 1$ .

**BELIEF STRUCTURES.** A set  $X \subseteq W$  is called a *focal element* of  $m$  if  $m(X) \neq 0$ . Let  $\mathcal{M}$  be the collection of all focal elements of  $m$ , then the pair  $(\mathcal{M}, m)$  is called a *belief structure* on  $W$ .

**BELIEF AND PLAUSIBILITY.** Given a belief structure  $(\mathcal{M}, m)$ , a *belief function*  $\text{Bel} : \mathcal{P}(W) \rightarrow [0, 1]$  and a *plausibility function*  $\text{Pl} : \mathcal{P}(W) \rightarrow [0, 1]$  can be derived, which are defined as follows for every  $X \subseteq W$ :

$$\text{Bel}(X) := \sum_{Y \subseteq X} m(Y), \quad (2.1)$$

and

$$\text{Pl}(X) := \sum_{Y \cap X \neq \emptyset} m(Y), \quad (2.2)$$

respectively. Note that the Bel and Pl functions are duals, i.e.  $\text{Bel}(X) = 1 - \text{Pl}(X^c)$ . Moreover,  $[\text{Bel}(X), \text{Pl}(X)]$  and  $\text{Pl}(X) - \text{Bel}(X)$  are called the *confidence interval* and the *ignorance level* of  $X$ , respectively.

**AXIOMATIC APPROACH.** A belief function can equivalently be defined in an axiomatic manner, i.e. it must satisfy the following axioms:

- $\text{Bel}(\emptyset) = 0$ ,
- $\text{Bel}(W) = 1$ ,
- $\text{Bel}(\cup_{i=1}^{\ell} X_i) \geq \sum_{\emptyset \neq I \subseteq \{1, \dots, \ell\}} (-1)^{|I|+1} \text{Bel}(\cap_{i \in I} X_i)$  for  $\{X_1, \dots, X_{\ell}\} \subseteq \mathcal{P}(W)$  and  $\ell > 0$ .

**2.2. S-approximation spaces.** In this section, some basic facts and definitions for S-approximation spaces are presented. We follow the notation of [10, 4].

**S-APPROXIMATION SPACES.** An *S-approximation space* is formally defined as a quadruple  $G = (U, W, T, S)$ , where  $U$  and  $W$  are finite non-empty sets,  $T$  is a multi-valued mapping  $T : U \rightarrow \mathcal{P}(W)$ , called a *knowledge component*, and  $S$  is a mapping  $S : \mathcal{P}(W) \times \mathcal{P}(W) \rightarrow \{0, 1\}$ , called a *decider*.

**LOWER AND UPPER APPROXIMATIONS.** Given an S-approximation space  $G = (U, W, T, S)$ , the lower and upper approximations of  $X \subseteq W$  are defined as

$$\underline{G}(X) = \{x \in U \mid S(T(x), X) = 1\}, \quad (2.3)$$

and

$$\overline{G}(X) = \{x \in U \mid S(T(x), X^c) = 0\}, \quad (2.4)$$

respectively, where  $X^c$  denotes the complement of  $X$  with respect to  $W$ .

GENERALITY AND SPECIAL CASES. Note that the mapping  $S$  can model a large class of measures, of which set inclusion, i.e.  $S_{\subseteq}(A, B) = \begin{cases} 1 & A \subseteq B \\ 0 & \text{otherwise} \end{cases}$ , is a special case. If we set  $S$  to  $S_{\subseteq}$  and consider the sets of form  $T(x)$  as blocks, we can model rough sets and some of their generalizations as special cases. For more information and other examples of decider functions consult [4, 10, 9, 12]. Moreover, other definitions and extensions have also been proposed for decider mappings, e.g. refer to [11] to see an instance suitable for intuitionistic fuzzy sets. However, in this paper we stick to the standard and general definition of S-approximation spaces as defined above.

TRICHOTOMY REGIONS. For any set  $X \subseteq W$ , the three pair-wise disjoint sets of positive, negative and boundary regions are defined as follows:

$$\begin{aligned} \text{POS}_G(X) &:= \{x \in U \mid S(T(x), X) = 1 \wedge S(T(x), X^c) = 0\} && \text{(Positive Region)} \\ &= \underline{G}(X) \cap \overline{G}(X), \\ \text{NEG}_G(X) &:= \{x \in U \mid S(T(x), X) = 0 \wedge S(T(x), X^c) = 1\} && \text{(Negative Region)} \\ &= U \setminus (\underline{G}(X) \cup \overline{G}(X)), \\ \text{BR}_G(X) &:= \{x \in U \mid S(T(x), X) = S(T(x), X^c)\} && \text{(Boundary Region)} \\ &= \underline{G}(X) \Delta \overline{G}(X), \end{aligned} \quad (2.5)$$

where  $A \Delta B = (A \setminus B) \cup (B \setminus A)$  for  $A, B \subseteq U$ .

It is noteworthy that the intuition behind Equation 2.5 is very similar to that of [18] (Equation 1). Refer to [9] for more discussion on this point. It is also the case that  $\text{POS}_G(X) = \text{NEG}_G(X^c)$  and  $\text{BR}_G(X) = \text{BR}_G(X^c)$  for any  $X \subseteq W$  [9]. We will routinely use these facts throughout the paper.

PARTIAL MONOTONICITY. A decider mapping  $S : \mathcal{P}(W) \times \mathcal{P}(W) \rightarrow \{0, 1\}$  is called *partial monotone* if  $X \subseteq Y \subseteq W$  and  $S(A, X) = 1$  imply that  $S(A, Y) = 1$  for any  $A \subseteq W$ . An S-approximation space  $G = (U, W, T, S)$  with a partial monotone decider mapping  $S$  is called a partial monotone S-approximation space. The lower and upper approximation operators and the three decision regions of such S-approximation spaces satisfy several important properties which are listed in the following proposition:

**Proposition 2.1** ([10, 12]). *Let  $G = (U, W, T, S)$  be a partial monotone S-approximation space. For all  $X, Y \subseteq W$ , we have:*

- (1)  $X \subseteq Y$  implies  $\overline{G}(X) \subseteq \overline{G}(Y)$ ,
- (2)  $X \subseteq Y$  implies  $\underline{G}(X) \subseteq \underline{G}(Y)$ ,
- (3)  $\overline{G}(X \cup Y) \supseteq \overline{G}(X) \cup \overline{G}(Y)$ ,
- (4)  $\overline{G}(X \cap Y) \subseteq \overline{G}(X) \cap \overline{G}(Y)$ ,

- (5)  $\underline{G}(X \cup Y) \supseteq \underline{G}(X) \cup \underline{G}(Y)$ ,
- (6)  $\underline{G}(X \cap Y) \subseteq \underline{G}(X) \cap \underline{G}(Y)$ ,
- (7)  $\overline{G}(X) = (\underline{G}(X^c))^c$ ,
- (8)  $\underline{G}(X) = (\overline{G}(X^c))^c$ ,
- (9)  $X \subseteq Y$  implies  $POS_G(X) \subseteq POS_G(Y)$ ,
- (10)  $X \subseteq Y$  implies  $NEG_G(Y) \subseteq NEG_G(X)$ ,
- (11)  $POS_G(X \cup Y) \supseteq POS_G(X) \cup POS_G(Y)$ ,
- (12)  $NEG_G(X \cup Y) \subseteq NEG_G(X) \cup NEG_G(Y)$ ,
- (13)  $POS_G(X \cap Y) \subseteq POS_G(X) \cap POS_G(Y)$ ,
- (14)  $NEG_G(X \cap Y) \supseteq NEG_G(X) \cap NEG_G(Y)$ ,
- (15)  $POS_G(X) \cap NEG_G(Y) \subseteq POS_G(X) \cap NEG_G(X \cap Y)$ .

**INFLECTION SETS.** Partial monotone S-approximation spaces can be represented by an equivalent form, which is called an *inflection set*. A pair  $(x, X) \in U \times \mathcal{P}(W)$  is called an *inflection point* with respect to  $G$  whenever  $S(T(x), X) = 1$  and for all  $Y \subsetneq X$ , we have  $S(T(x), Y) = 0$  [10]. The inflection set of a partial monotone  $G$ , which is denoted by  $\mathcal{IS}(G)$ , is defined as the set of all of its inflection points. Moreover, for  $x \in U$  we use  $\mathcal{IP}_G(x)$  to represent the collection of  $X \subseteq W$  where  $(x, X) \in \mathcal{IS}(G)$ , so that  $\mathcal{IS}(G) = \cup_{x \in U} \{(x, X) | X \in \mathcal{IP}_G(x)\}$ .

**TRIVIAL ELEMENTS.** An element  $x \in U$  is called *trivial* if we have either  $\mathcal{IP}_G(x) = \emptyset$  or  $\mathcal{IP}_G(x) = \{\emptyset\}$ . In the former case, we have  $S(T(x), X) = 0$  for all  $X \subseteq W$ , so  $x$  appears in none of the lower approximations  $\underline{G}(X)$  and in every upper approximation  $\overline{G}(X^c)$ . So, the element  $x$  is not providing any useful information, i.e. it cannot be used to distinguish any pair of subsets of  $W$ . Similarly, in the latter case,  $S(T(x), \emptyset) = 1$ , which, due to partial monotonicity, implies  $S(T(x), X) = 1$  for all  $X \subseteq W$ . Hence, for all  $X \subseteq W$ , we have  $x \in \underline{G}(X)$  and  $x \notin \overline{G}(X^c)$ . So  $x$  does not provide any useful information in this case, either.

**REDUCIBILITY.** As argued above, if  $x$  is a trivial element, one can remove  $x$  and get a smaller system from which one can get just as much information as the initial system. A partial monotone S-approximation space is called *reducible* if it contains a trivial element, otherwise we call it *irreducible*.

### 3. S-APPROXIMATION SPACES AND BELIEF STRUCTURES

In this section, we study the relationship between S-approximation spaces and belief structures.

The qualities of lower and upper approximations with respect to an S-approximation space are defined as follows:

**Definition 3.1.** Let  $G = (U, W, T, S)$  be an S-approximation space. The qualities of lower and upper approximations of a set  $X \subseteq W$  with respect to  $G$

are defined as:

$$\underline{Q}_G(X) = \frac{|\text{POS}_G(X)|}{|U|}, \quad (3.1)$$

and

$$\overline{Q}_G(X) = \frac{|\text{POS}_G(X)| + |\text{BR}_G(X)|}{|U|}. \quad (3.2)$$

The qualities defined in Equations 3.1 and 3.2 are dual. This is stated more formally in the following proposition:

**Proposition 3.2.** *Let  $G = (U, W, T, S)$  be an  $S$ -approximation space. Then, for all  $X \subseteq W$  we have  $\underline{Q}_G(X) = 1 - \overline{Q}_G(X^c)$ .*

*Proof.* The proof is as follows and uses the fact that  $\text{POS}_G(X) = \text{NEG}_G(X^c)$ :

$$\begin{aligned} \underline{Q}_G(X) &= \frac{|\text{POS}_G(X)|}{|U|} = \frac{|\text{NEG}_G(X^c)|}{|U|} \\ &= \frac{|U \setminus (\text{POS}_G(X^c) \cup \text{BR}_G(X^c))|}{|U|} \\ &= 1 - \frac{|\text{POS}_G(X^c)| + |\text{BR}_G(X^c)|}{|U|} \\ &= 1 - \overline{Q}_G(X^c). \end{aligned} \quad (3.3)$$

□

Next, we consider the properties of these quality values for a partial monotone  $S$ -approximation space.

**Proposition 3.3.** *Let  $G = (U, W, T, S)$  be a partial monotone  $S$ -approximation space. Then,  $\underline{Q}_G(\emptyset) = 0$ .*

*Proof.* It suffices to show that  $\text{POS}_G(\emptyset) = \emptyset$ . The proof is by contradiction. Suppose there exists some  $x \in U$  such that  $x \in \text{POS}_G(\emptyset)$ . So, it is the case that  $S(T(x), \emptyset) = 1$  and  $S(T(x), W) = 0$ . This is a contradiction with partial monotonicity of  $G$ , since  $\emptyset \subseteq W$  and we need to have  $S(T(x), W) = 1$ . Therefore the desired result is obtained. □

**Proposition 3.4.** *Let  $G = (U, W, T, S)$  be an irreducible partial monotone  $S$ -approximation space. Then,  $\underline{Q}_G(W) = 1$ .*

*Proof.* Note that  $G$  is irreducible, hence for every  $x \in U$ , there exists  $X \subseteq W$ , such that  $S(T(x), X) = 1$ <sup>§</sup>. Therefore, by partial monotonicity, we have  $S(T(x), W) = 1$  for all  $x \in U$ . Moreover,  $S(T(x), \emptyset) = 0$  for all  $x \in U$ . Hence,  $x \in \text{POS}_G(W)$  for all  $x \in U$  and  $\text{POS}_G(W) = U$ . So,  $\underline{Q}_G(W) = \frac{|U|}{|U|} = 1$ . □

<sup>§</sup>Otherwise  $x$  is trivial and  $G$  is reducible, which is a contradiction.

**Proposition 3.5.** *Let  $G = (U, W, T, S)$  be a partial monotone S-approximation space. Then, for all  $\ell \in \mathbb{N}$  we have*

$$\underline{Q}_G(\cup_{i=1}^{\ell} X_i) \geq \sum_{\emptyset \neq I \subseteq \{1, \dots, \ell\}} (-1)^{|I|+1} \underline{Q}_G(\cap_{i \in I} X_i), \quad (3.4)$$

where  $X_i \subseteq W$ .

*Proof.* By the definition, we have

$$\underline{Q}_G(\cup_{i=1}^{\ell} X_i) = \frac{|\text{POS}_G(\cup_{i=1}^{\ell} X_i)|}{|U|}. \quad (3.5)$$

By partial monotonicity of  $G$ , we have

$$\frac{|\text{POS}_G(\cup_{i=1}^{\ell} X_i)|}{|U|} \geq \frac{|\cup_{i=1}^{\ell} \text{POS}_G(X_i)|}{|U|}, \quad (3.6)$$

since  $\text{POS}_G(\cup_{i=1}^{\ell} X_i) \supseteq \cup_{i=1}^{\ell} \text{POS}_G(X_i)$ . Now the desired result can be obtained by applying the inclusion-exclusion principle.  $\square$

Propositions 3.3 to 3.5 result in the following:

**Proposition 3.6.** *Let  $G = (U, W, T, S)$  be an irreducible partial monotone S-approximation space. The quality of lower approximation, as defined in Definition 3.1, is a belief function.*

Similarly, for an irreducible partial monotone S-approximation space, the quality of upper approximation is a plausibility function. This is treated more formally in the following proposition:

**Proposition 3.7.** *Let  $G = (U, W, T, S)$  be an irreducible partial monotone S-approximation space. Then the quality of upper approximation, as defined in Definition 3.1, is a plausibility function.*

*Proof.* By the duality of belief and plausibility functions, we have  $\text{Pl}_G(X) = 1 - \text{Bel}_G(X^c)$  and this is all we have to show. By the definition, we have

$$\begin{aligned} \underline{Q}_G(X^c) &= \frac{|\text{POS}_G(X^c)|}{|U|} \\ &= \frac{|\text{NEG}_G(X)|}{|U|} \\ &= \frac{|U \setminus (\text{POS}_G(X) \cup \text{BR}_G(X))|}{|U|} \\ &= 1 - \frac{|\text{POS}_G(X)| + |\text{BR}_G(X)|}{|U|} \\ &= 1 - \overline{Q}_G(X). \end{aligned} \quad (3.7)$$

By applying Proposition 3.6, the desired result is obtained.  $\square$

By Propositions 3.6 and 3.7, it can be said that every irreducible partial monotone S-approximation space induces a belief structure on  $W$ .

**Theorem 3.8.** *Let  $G = (U, W, T, S)$  be an irreducible partial monotone S-approximation space. Then,  $G$  induces a belief structure  $(\mathcal{M}, m)$  on  $W$  where*

$$m(X) = \sum_{Y \subseteq X} (-1)^{|X \setminus Y|} \underline{Q}_G(Y), \quad (3.8)$$

and

$$\mathcal{M} = \{X \subseteq W \mid m(X) \neq 0\}, \quad (3.9)$$

for  $X \subseteq W$ .

*Proof.* The bpa can be defined from a belief function, which is the quality of lower approximation (by Proposition 3.6), by the following relation

$$n(A) = \sum_{B \subseteq A} (-1)^{|A \setminus B|} \text{Bel}(B), \quad (3.10)$$

where  $n$  is a bpa [19]. This concludes the proof.  $\square$

Next, we show that belief structures can induce S-approximation spaces.

**Theorem 3.9.** *Suppose that  $(\mathcal{M}, m)$  is a given belief structure over a finite non-empty set  $W$  such that for all focal elements  $X \in \mathcal{M}$ , there exist  $a, b \in \mathbb{Z}^+$  such that  $m(X) = \frac{a}{b}$ . Then, there exists an S-approximation space  $G = (U, W, T, S)$  such that the quality of lower and upper approximations with respect to  $G$  are the corresponding belief and plausibility functions.*

*Proof.* The proof is by construction. Without loss of generality, we assume that there exists a constant  $d \in \mathbb{Z}^+$  such that for all focal elements  $X \in \mathcal{M}$ , we have  $m(X) = \frac{c}{d}$  for some  $c \in \mathbb{Z}^+$ . This is easy to obtain by computing the least common multiple.

Now define the set  $U$  as  $U = \{1, \dots, d\}$ . For each  $X \in \mathcal{M}$  with  $m(X) = \frac{l_X}{d}$ , we choose a subset  $A_X$  of size  $l_X$  of  $U$ . We assume that the  $A_X$ 's are pairwise disjoint. We can always find such disjoint  $A_X$ 's, since  $\sum_{X \in \mathcal{M}} m(X) = 1$  and hence  $\sum_{X \in \mathcal{M}} l_X = d$ . Now for each  $i \in A_X$ , we let  $T(i) = X$ . Finally, we let the decider mapping  $S$  be the ordinary set inclusion operator  $S_{\subseteq}$ .

Next, it is easy to see that  $G$  satisfies the conditions of Propositions 3.6 and 3.7. Therefore, the qualities of lower and upper approximations with respect to  $G$  are belief and plausibility functions, respectively.

Finally, we show that for all  $X \subseteq W$ , the belief and plausibility values of  $X$  with respect to  $(\mathcal{M}, m)$  are equal to the corresponding values with respect to  $G$ . Since the belief and plausibility functions are dual, it suffices to show the result for belief. This can be done as follows:



$$\begin{aligned}
\underline{Q}_G(X) &= \frac{|\text{POS}_G(X)|}{|U|} \\
&= \frac{|\{x \in U \mid T(x) \subseteq X\}|}{|U|} \\
&= \sum_{Y \subseteq X} m(Y) = \text{Bel}(X).
\end{aligned} \tag{3.11}$$

This concludes the proof.  $\square$

Now suppose that we are given a belief structure  $(\mathcal{M}, m)$  over  $U$  and an irreducible partial monotone S-approximation space  $G = (U, W, T, S)$ . Then we can induce a belief structure  $(\mathcal{M}', m')$  on  $W$  by declaring  $\mathcal{M}'$  as

$$\mathcal{M}' = \{Z \subseteq W \mid \exists x \in U, (x, Z) \in \mathcal{IS}(G)\}, \tag{3.12}$$

and the bpa  $m'$  as

$$m'(Y) = \begin{cases} \sum_{X \in \mathcal{M}} \frac{m(X)}{|X|} \times \left( \sum_{x \in X, Y \in \mathcal{IP}_G(x)} \frac{1}{|\mathcal{IP}_G(x)|} \right) & \text{if } Y \in \mathcal{M}', \\ 0 & \text{otherwise.} \end{cases} \tag{3.13}$$

The intuition behind Equation 3.13 is that the bpa value of every  $X \in \mathcal{M}$  is divided between each  $x \in X$  equally likely, which are called their shares. Then, the bpa  $m'$  of  $Y \in \mathcal{M}'$  receives the shares of those  $x \in X$  for which  $Y \in \mathcal{IP}_G(X)$ .

**Theorem 3.10.** *Given a belief structure  $(\mathcal{M}, m)$  on a finite non-empty set  $U$  and an irreducible partial monotone S-approximation space  $G = (U, W, T, S)$ ,  $(\mathcal{M}', m')$  as defined in Equations 3.12 and 3.13 is a valid belief structure on  $U$ .*

*Proof.* The bpa  $m'$  needs to satisfy two conditions, i.e. (1)  $m'(\emptyset) = 0$  and (2)  $\sum_{Y \subseteq W} m'(Y) = 1$ . By the hypothesis that  $\emptyset \notin \mathcal{IP}_G(x)$  for all  $x \in U$ , we have  $\emptyset \notin \mathcal{M}'$  and therefore, its bpa value  $m'(\emptyset)$  is zero. The second property can be proven as follows (note that for all  $x \in X \in \mathcal{M}$ , we have  $\mathcal{IP}_G(x) \subseteq \mathcal{M}'$ ):

$$\begin{aligned}
\sum_{Y \subseteq W} m'(Y) &= \sum_{Y \in \mathcal{M}'} m'(Y) \\
&= \sum_{Y \in \mathcal{M}'} \sum_{X \in \mathcal{M}} \frac{m(X)}{|X|} \times \left( \sum_{x \in X \mid Y \in \mathcal{IP}_G(x)} \frac{1}{|\mathcal{IP}_G(x)|} \right) \\
&= \sum_{x \in X \in \mathcal{M}, Y \in \mathcal{IP}_G(x) \subseteq \mathcal{M}'} \frac{m(X)}{|X|} \times \frac{1}{|\mathcal{IP}_G(x)|} \\
&= \sum_{X \in \mathcal{M}} \frac{m(X)}{|X|} \times \left( \sum_{x \in X, Y \in \mathcal{IP}_G(x)} \frac{1}{|\mathcal{IP}_G(x)|} \right) \\
&= \sum_{X \in \mathcal{M}} \frac{m(X)}{|X|} \times \left( \sum_{x \in X} 1 \right) \\
&= \sum_{X \in \mathcal{M}} m(X) = 1.
\end{aligned} \tag{3.14}$$

□

#### 4. CONCLUSION AND FUTURE RESEARCH DIRECTIONS

In this paper, we studied some connections between the Dempster-Shafer's theory of evidence and the concept of S-approximation spaces. First, we defined two numeric measures called the qualities of lower and upper approximations for S-approximation spaces. Then, we showed that they can be used to derive a belief structure from an irreducible partial monotone S-approximation space in a natural way. Finally, we showed that given a belief structure on a set  $U$  and an irreducible partial monotone S-approximation space  $G = (U, W, T, S)$ , a valid natural belief structure can be induced on  $W$ .

The results obtained in this paper are the first ones settling a relation between the two theories and are extensible by trying to answer the following proposed problems:

- (1) Can belief structures be generalized to two universal sets with respect to an arbitrary S-approximation space in a natural or meaningful way?
- (2) Can the results of this paper be extended to neighborhood systems, especially the ones in [9]? For example, by fusing knowledge mappings of multiple S-approximation spaces with a similar approach to [5].
- (3) Can the qualities of lower and upper approximations be used to reduce the knowledge mappings in the context of [9]? For example, can one find a minimal set of knowledge mappings of multiple S-approximation spaces for which the amount of information one can obtain from that set does not change compared to the case when she uses all of them?

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