

CONSONANT CREDIBILITY STRUCTURES WITHIN  
MULTI-EXPERT SYSTEMS BELIEF SPACES

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**Abstract**

A particular class of belief spaces is introduced, as an illustration of uncertain knowledge processing within multi-expert systems.

Under the basic assumption that the context generating a belief space is heterogeneous, containing both evidences with known probability and hypotheses, with known credibility and plausability, three main consequences are analyzed: (a) the splitting of the belief space into three categories of facts, according to their heterogeneous derivational support; (b) the construction of adequate probability and mass distribution functions; (c) the construction of specific credibility and plausability functions that "reveal" the consonant credibility structure, inner to the belief space.

**1. Introduction**

A special approach to credibility functions concerns their property of consonance, namely the possibility to determine nested levels of evolving belief within a universe of discourse, considered as a frame of discernment.

The consonance principles were introduced by Shafer /7/ and represent a specific aspect within the Dempster-Shafer belief theory /1,7/. They were largely debated in /2,3/.

A credibility structure is called consonant if its focal elements are nested, namely they can be indexed such that  $A_n \supset \dots \supset A_1$ , for  $m(A_1) \geq \dots \geq m(A_n)$  where  $m(A_i)$  is the mass attached to  $A_i$ . An important property /7/ states that a belief function derives from a consonant credibility structure iff

$$Pl(A \cup B) = \max (Pl(A), Pl(B)) \quad (1.1)$$

or, equivalently,

$$Cr(A \cap B) = \min (Cr(A), Cr(B)) \quad (1.2)$$

We recall that in the framework of belief spaces theory /5/ a basic concept is that of context, seen as a set  $\Theta = \{H_1, \dots, H_n\}$  of hypotheses.

As well, a belief space is actually a context, along with all the facts derivable from it.

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The elements used to derive a fact are called the origin set(OS) of that fact. A fact is called supported if  $OS \in \mathcal{P}(C)$ .

## 2. Consistent credibility within heterogeneous belief spaces

In the present approach we use the basic assumptions of /6/. The knowledge that an expert system deals with is assimilated to a belief space HB, where the context is heterogeneous. Namely  $C = E \cup H$ , where  $e \in E$  are evidences with known probability  $p(e)$ , while  $h \in H$  are hypotheses with a given credibility  $Gr(h)$  and plausability  $Pl(h)$ . The sets of rules are expressed by a multivalued mapping  $HR: HB \rightarrow \mathcal{P}(HB)$ . Thus, the same conclusion  $F \in HB$  may be supported by several origin sets(sets of premises)  $OS = \{A_1, \dots, A_n\}, A_i \subset HB$ .

A fact is called: (i) completely supported, when  $HR(F) \subset \mathcal{P}(C)$ ; (ii) partially hypothetical, when  $HR(F) \subset \mathcal{P}(HB)$ ; (iii) completely hypothetical, when  $HR(F) = \emptyset$ . The three classes will be denoted by CS, PH, and CH, respectively.

Two probability type distributions are defined on C:

$$- \text{ a lower one, cred: } C \rightarrow [0,1], \text{ with } cred(e) = \begin{cases} p(e), & e \in E \\ Gr(\{e\}), & e \in H \end{cases} \quad (2.1)$$

$$- \text{ a higher one, plaus: } C \rightarrow [0,1], \text{ with } plaus(e) = \begin{cases} p(e), & e \in E \\ Pl(\{e\}), & e \in H \end{cases} \quad (2.2)$$

Two mass type distributions are further associated to  $F \in CS$ :

$$- \text{ lm: } 2^{HB} \rightarrow [0,1], \text{ lm}(F) = \min_{OS} \sum_{e \in OS} cred(e) \quad (2.3)$$

$$- \text{ hm: } 2^{HB} \rightarrow [0,1], \text{ hm}(F) = \max_{OS} \sum_{e \in OS} plaus(e) \quad (2.4)$$

Credibility and plausability are defined, for facts of CS and PH, at the level of a single origin set OS, by:

$$GR-S(F) = \left[ \sum_{A_i \in OS} cr(A_i) \times dc-i \times N(F|A_i) \right] \times s$$

$$PL-S(F) = \left[ \sum_{A_i \in OS} pl(A_i) \times dc-i \times \overline{N}(F|A_i) \right] \times s$$

where  $dc-i$  is the degree of contribution of  $A_i$  in concluding  $F$ ,  $s$  is the certainty degree attached to  $F$ , due to OS, and:

$$cr(A_i) = \begin{cases} lm(A_i), & A_i \in CS \\ GR-C(A_i), & A_i \in PH \\ GR(A_i), & A_i \in CH \end{cases} \quad pl(A_i) = \begin{cases} hm(A_i), & A_i \in CS \\ PL-C(A_i), & A_i \in PH \\ PL(A_i), & A_i \in CH \end{cases}$$

At the level of the whole belief space we define:

$$GR-C(F) = \min_{OS_i \in HR(F)} GR-S_i(F); \quad PL-C(F) = \max_{OS_i \in HR(F)} PL-S_i(F)$$

GR and PL for facts of CH are taken from those facts of HB that are most similar to them, and also credible and plausible to a highest degree. The procedure used to detect these "reliable" similarities is a specific adaptation of some fuzzy pattern-matching and fuzzy decision-making methods described in /4/ and /8/. The concept of multi-expert system is supported by the acceptance of several  $HB_i$ , with  $\{OS_j\} j$ , located in  $\bigcup_1 \mathcal{P}(HB_i)$ . Under the above circumstances, the following propositions hold.

**Proposition 2.1.** The function  $lm$  in (2.3) is a mass distribution.

**Proof.** We have to show that:

$$lm(\emptyset) = 0 \quad (2.1.1.) \quad \text{and} \quad \sum_{A \subset HB} lm(A) = 1 \quad (2.1.2.)$$

$$\text{Indeed, } lm(\emptyset) = \begin{cases} p(\emptyset) = 0, E = \emptyset \\ cred(\emptyset) = Cr(\emptyset) = 0, H = \emptyset \end{cases} \quad \text{Thus, (2.1.1.)}$$

is satisfied.

For (2.1.2.), we have:

$$\sum_{A \subset HB} lm(A) = \begin{cases} \sum_{A \subset HB} \min_{OS \in HB(A)} \sum_{ei \in OS} p(ei), OS \subseteq E & (2a) \\ \sum_{A \subset HB} \min_{OS \in HB(A)} \sum_{ei \in OS} Cr(\{ei\}), OS \subseteq E & (2b) \end{cases}$$

In the case (2a), we have:

$$\sum_{A \subset HB} \min_{OS \in HB(A)} \sum_{ei \in OS} p(ei) = \min_{OS \in \mathcal{P}(HB)} \sum_{ei \in OS} p(ei)$$

Since  $p$  is a probability, in the case  $ci \in OS = E$  we have  $\sum_{ei \in E} p(ei) = 1$ .

$$\text{Therefore, } \min_{OS \in \mathcal{P}(HB), OS \subseteq E} \sum_{ei \in OS} p(ei) = 1.$$

In the case (2b), we have:

$$\sum_{A \subset HB} \min_{OS \in HB(A)} \sum_{ei \in OS} Cr(\{ei\}) = \min_{OS \in \mathcal{P}(HB)} \sum_{ei \in OS} Cr(\{ei\})$$

Since  $Cr$  is a credibility function,  $Cr(\{ei\}) = m(\{ei\})$ . Hence,

$$\text{for } ci \in OS = HB, \text{ we have } \sum_{\{ei\} \subset HB} Cr(\{ei\}) = \sum_{\{ei\} \subset HB} m(\{ei\}) = 1.$$

Thus,  $\min_{\{ei\} \subset HB} Cr(\{ei\}) = 1$  and (2.1.2.) is satisfied.

Similarly, it can be shown that  $hm$  in (2.4) is a mass distribution.

In this case, for  $OS = HB$ , we have:

$$\sum_{\{ei\} \subset HB} Pl(ei) = \sum_{\{ei\} \subset HB} \sum_{j | ei \in Aj} m(Aj) = \sum_j m(Aj) = 1.$$

**Proposition 2.2.** The plausability  $Pl-C$  derives from the consensus credibility structure of  $HB$ .

**Proof.** We have to show that for any  $F, G \subset HB$ ,

$$Pl-C(F \cup G) = \max(Pl-C(F), Pl-C(G))$$

Indeed, using the adequate definition-based substitution, we obtain:

$$\begin{aligned} & \max_{OSi \in HB(F \cup G)} Pl-Si(F \cup G) = \\ & = \max \left[ \max_{OSi \in HB(F)} \sum_{Aj \in OSi} pl(Aj) \times dc-j \times \prod (F|Aj) \times c, \right. \\ & \quad \left. \max_{OSi \in HB(G)} \sum_{Aj \in OSi} pl(Aj) \times dc-j \times \prod (G|Aj) \times c \right]. \end{aligned}$$

Hence, according to (1.1),  $HB$  underlies a consensus credibility structure, with respect to  $hm$ . The credibility consensus within  $HB$ , with respect to  $lm$ , results in a similar manner from (1.2), satisfied by  $CR-C$ .

### 3. Conclusions

An appropriate set of credibility functions was constructed, for a heterogeneous space of uncertain facts, in order to illustrate the convergence of belief levels towards a consensus credibility structure.

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