



On Two Operators of Composition in Dempster-Shafer Theory

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Composition in Probability Theory

Factorization Lemma.

For prob. distribution $\pi(X, Y, Z)$ there exist functions

$$\phi : \mathbb{X} \times \mathbb{Z} \longrightarrow \mathbb{R}^+,$$

$$\psi : \mathbb{Y} \times \mathbb{Z} \longrightarrow \mathbb{R}^+,$$

such that

$$\pi(\mathbf{a}) = \phi(\mathbf{a}^{\downarrow\{X,Z\}}) \cdot \psi(\mathbf{a}^{\downarrow\{Y,Z\}})$$

for all $\mathbf{a} \in \mathbb{X} \times \mathbb{Y} \times \mathbb{Z}$ iff $X \perp\!\!\!\perp Y | Z$, and therefore also

$$\pi(X, Y, Z) = \pi(X, Z) \triangleright \pi(Y, Z).$$

Definition of Composition.

Let $\kappa(K)$ and $\lambda(L)$ be probability distributions.

If $\lambda^{\downarrow K \cap L} \gg \kappa^{\downarrow K \cap L}$ then

$$\kappa(K) \triangleright \lambda(L) = \frac{\kappa(K) \triangleright \lambda(L)}{\lambda(K \cap L)},$$

otherwise the composition is *undefined*.

Properties of Composition.

- Operator \triangleright is neither commutative nor associative.
- $\kappa(K) \triangleright \lambda(L)$ is a prob. distribution of $K \cup L$.
- $\kappa(K) \triangleright \lambda(L)$ is an extension of κ .
- If κ and λ are consistent then $\kappa \triangleright \lambda$ is their joint extension.
- $\kappa(K) \triangleright \lambda(L)$ factorizes with respect to (K, L) ,
- and therefore also $K \setminus L \perp\!\!\!\perp L \setminus K | K \cap L$.
- If $X \in K$ and $X \notin L$ then $(\kappa \triangleright \lambda)^{-X} = \kappa^{-X} \triangleright \lambda$.

After defining the operator of composition in Shenoy's Valuation-Based Systems (VBS) [2] it appeared that there are (at least) two possibilities how to define an operator of composition in D-S theory:

Dempster operator of composition

Definition.

Consider two arbitrary bpas, μ_1 on \mathbb{X}_K and μ_2 on \mathbb{X}_L ($K \neq \emptyset \neq L$) and assume that $\mu_2^{\downarrow K \cap L} \gg \mu_1^{\downarrow K \cap L}$. Let θ_1 and θ_2 be their commonality functions, respectively. A composition $\theta_1 \triangleright \theta_2$ is defined for each nonempty $\mathbf{c} \subseteq \mathbb{X}_{K \cup L}$ by the following formula:

$$(\theta_1 \triangleright \theta_2)(\mathbf{c}) = \begin{cases} \alpha^{-1} \frac{\theta_1(\mathbf{c}^{\downarrow K}) \cdot \theta_2(\mathbf{c}^{\downarrow L})}{\theta_2^{\downarrow K \cap L}(\mathbf{c}^{\downarrow K \cap L})} & \text{if } \theta_2^{\downarrow K \cap L}(\mathbf{c}^{\downarrow K \cap L}) > 0, \\ 0 & \text{otherwise,} \end{cases}$$

where α is a normalization constant defined as

$$\alpha = \sum_{\mathbf{d} \in 2^{\mathbb{X}_{K \cup L}} : \theta_2^{\downarrow K \cap L}(\mathbf{d}^{\downarrow K \cap L}) > 0} (-1)^{|\mathbf{d}|+1} \frac{\theta_1(\mathbf{d}^{\downarrow K}) \cdot \theta_2(\mathbf{d}^{\downarrow L})}{\theta_2^{\downarrow K \cap L}(\mathbf{d}^{\downarrow K \cap L})}.$$

Factorizing operator of composition

Definition.

Consider two normal bpas, μ_1 on \mathbb{X}_K and μ_2 on \mathbb{X}_L ($K \neq \emptyset \neq L$). A composition $\mu_1 \triangleright \mu_2$ is defined for each nonempty $\mathbf{c} \subseteq \mathbb{X}_{K \cup L}$ by one of the following expressions:

(i) if $\mu_2^{\downarrow K \cap L}(\mathbf{c}^{\downarrow K \cap L}) > 0$ and $\mathbf{c} = \mathbf{c}^{\downarrow K} \bowtie \mathbf{c}^{\downarrow L}$ then

$$(\mu_1 \triangleright \mu_2)(\mathbf{c}) = \frac{\mu_1(\mathbf{c}^{\downarrow K}) \cdot \mu_2(\mathbf{c}^{\downarrow L})}{\mu_2^{\downarrow K \cap L}(\mathbf{c}^{\downarrow K \cap L})};$$

(ii) if $\mu_2^{\downarrow K \cap L}(\mathbf{c}^{\downarrow K \cap L}) = 0$ and $\mathbf{c} = \mathbf{c}^{\downarrow K} \times \mathbb{X}_{L \setminus K}$ then

$$(\mu_1 \triangleright \mu_2)(\mathbf{c}) = m_1(\mathbf{c}^{\downarrow K});$$

(iii) in all other cases, $(\mu_1 \triangleright \mu_2)(\mathbf{c}) = 0$.

Properties of the Dempster operator

- Meets all the properties of composition (can be used for multidimensional model representation, makes local computations in decomposable models possible).
- Computationally **extremely complex**.
- Can be used in the process of Iterative Proportional Fitting but **Csiszár's convergence theorem [1] does not hold**.
- **Defines conditionals**: for bpa μ on \mathbb{X}_M and $X_j, X_k \in M$,

$$\mu(X_k | X_j = \mathbf{a}) = (\nu_{X_j=\mathbf{a}} \triangleright \mu)^{\downarrow X_k},$$

where $\nu_{X_j=\mathbf{a}}$ is a one-dimensional bpa on \mathbb{X}_j having just one focal element $\{\mathbf{a}\} \subset \mathbb{X}_j$, for which $\nu_{X_j=\mathbf{a}}(\{\mathbf{a}\}) = 1$.

Conditional Independence Lemma.

For commonality function $\theta(X, Y, Z)$ the conditional independence relation $X \perp\!\!\!\perp Y | Z$ holds true iff

$$\theta(X, Y, Z) = \theta(X, Z) \triangleright \theta(Y, Z).$$

Properties of the factorizing operator

- Meets all the properties of composition (can be used for multidimensional model representation, makes local computations in decomposable models possible).
- Computationally **much simpler** than Dempster operator.
- Can be used in the process of Iterative Proportional Fitting; **Csiszár's convergence theorem [1] holds true**.
- **Cannot be used to express conditionals**.

Factorization Lemma.

For bpa $\mu(X, Y, Z)$ there exist functions

$$\phi : 2^{\mathbb{X} \times \mathbb{Z}} \longrightarrow \mathbb{R}^+, \quad \psi : 2^{\mathbb{Y} \times \mathbb{Z}} \longrightarrow \mathbb{R}^+,$$

such that

$$\mu(\mathbf{a}) = \begin{cases} \phi(\mathbf{a}^{\downarrow\{X,Z\}}) \cdot \psi(\mathbf{a}^{\downarrow\{Y,Z\}}) & \text{if } \mathbf{a} = \mathbf{a}^{\downarrow\{X,Z\}} \bowtie \mathbf{a}^{\downarrow\{Y,Z\}} \\ 0 & \text{otherwise} \end{cases}$$

iff $\mu(X, Y, Z) = \mu(X, Z) \triangleright \mu(Y, Z)$.

Main Open Problem - Conjecture

Suppose μ_1 , μ_2 and μ_3 are bpas on \mathbb{X}_K , \mathbb{X}_L , and \mathbb{X}_M , respectively. If $L \supset (K \cap M)$ then, $(\mu_1 \triangleright \mu_2) \triangleright \mu_3 = \mu_1 \triangleright (\mu_2 \triangleright \mu_3)$.

References

- [1] I. Csiszár. I-divergence geometry of probability distributions and minimization problems. *Ann. Probab.* 3, pp 146–158, 1975.
[2] R. Jiroušek and P. P. Shenoy. Compositional models in valuation-based systems. *Int. J. Approx. Reasoning*, 55, 1, pp. 277–293, 2014.