

# CREPO: An Open Repository to Benchmark Credal Network Algorithms (Supplementary Material)

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## Appendix A. Supplementary Material

Consider a closed and convex set  $C$  in  $\mathbb{R}^d$ . Let  $d$  denote also the dimension of  $C$ . The vertices of  $C$  are assumed to be finite and denoted as  $e(C)$ . A hyperplane  $H$  in  $\mathbb{R}^d$  can be parametrized by a pair  $(\mathbf{v}, w)$ , with  $\mathbf{v} \in \mathbb{R}^d$  and  $w \in \mathbb{R}$  as follows:

$$H_{\mathbf{v},w} := \{\mathbf{x} \in \mathbb{R}^d : \mathbf{v} \cdot \mathbf{x} = w\}. \quad (1)$$

The segment  $S$  connecting points  $\mathbf{a}, \mathbf{b} \in \mathbb{R}^d$  is instead:

$$S_{\mathbf{a},\mathbf{b}} := \{\mathbf{x} \in \mathbb{R}^d := \lambda \mathbf{a} + (1 - \lambda) \mathbf{b}, 0 \leq \lambda \leq 1\}. \quad (2)$$

**Definition 1**  $H_{\mathbf{v},w}$  is a supporting hyperplane for  $C$  passing through  $\mathbf{x}^* \in C$  if and only if  $\mathbf{x}^* \in P_{\mathbf{v},w}$  and  $\mathbf{v} \cdot \mathbf{x} \leq \mathbf{v} \cdot \mathbf{x}^*$  for each  $\mathbf{x} \in C$ .

**Definition 2** A point  $\mathbf{x}^*$  belongs to the boundary  $b(C)$  of the convex set  $C$  if and only if there is at least a supporting hyperplane for  $C$  passing through  $\mathbf{x}^*$ .

We use notation  $\text{CH}$  for the convex hull of a set of points, e.g.,  $C := \text{CH}[e(C)]$ . The following result holds.

**Lemma 3** Let  $C$  be a convex set in  $\mathbb{R}^d$  such that  $d$  is also its dimension (i.e., Given  $\mathbf{a}, \mathbf{b} \in e(C)$ , let  $\mathbf{x}^* := \frac{1}{2}(\mathbf{a} + \mathbf{b})$ . Let also  $C' := \text{CH}[e(C) \setminus \{\mathbf{a}, \mathbf{b}\}]$ , while  $H_{\mathbf{v},w}$  denotes a supporting hyperplane for  $C$  through  $\mathbf{x}^*$ . It holds that, if  $S_{\mathbf{a},\mathbf{b}} \subset H_{\mathbf{v},w}$ , then  $\mathbf{x}^* \notin C'$ .

**Proof** By construction  $\mathbf{x} \in S_{\mathbf{a},\mathbf{b}}$ . Assume, ad absurdum,  $\mathbf{x}^* \in C'$ . Thus,  $\mathbf{x}^*$  should be a convex combination of the vertices of  $C'$ , i.e.

$$\mathbf{x}^* = \sum_{\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}} \lambda_{\mathbf{z}} \mathbf{z}, \quad (3)$$

where  $\lambda_{\mathbf{z}} \geq 0$  for each  $\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}$  and  $\sum_{\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}} \lambda_{\mathbf{z}} = 1$ . Take the scalar product by  $\mathbf{v}$ :

$$\mathbf{v} \cdot \mathbf{x}^* = \mathbf{v} \cdot \left[ \sum_{\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}} \lambda_{\mathbf{z}} \mathbf{z} \right] = \sum_{\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}} \lambda_{\mathbf{z}} \mathbf{v} \cdot \mathbf{z}. \quad (4)$$

By supporting hyperplane definition and simple algebra:

$$\mathbf{v} \cdot \mathbf{x}^* = \sum_{\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}} \lambda_{\mathbf{z}} \mathbf{v} \cdot \mathbf{z} \leq \sum_{\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}} \lambda_{\mathbf{z}} \mathbf{v} \cdot \mathbf{x}^* = \mathbf{v} \cdot \mathbf{x}^*. \quad (5)$$

This implies  $\mathbf{z} \in H_{\mathbf{v},w}$  for each  $\mathbf{z} \in e(C) \setminus \{\mathbf{a}, \mathbf{b}\}$ . As also  $\mathbf{a}$  and  $\mathbf{b}$  belong to  $H_{\mathbf{v},w}$ , we have  $e(C) \subset H_{\mathbf{v},w}$ . In other words  $C$  is included in a hyperplane and it coincides with its boundary, but this is against the original assumption about the dimension of  $C$ . ■

As a consequence of this lemma we have that, in Definition 1, the midpoint of the two vertices added to  $C$  is a vertex of the new set. This simply following that the two points at minimum (Euclidean) distance belong to a same edge of a convex polytope and the credal set can be always parametrized in order to have full dimension (see discussion in Section 3. When coping with non-Euclidean distances, to have the same result, the two points at minimum distance in Definition 1 should be detected with the additional condition of belonging to a same edge.