

Supplementary Material for:  
**Échantillonnage actif pour la découverte de règles  
classification via des comparaisons par paires**

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*Remark 1.* In this proof, we only consider rules with distinct scores. Since the rule space is finite, we order the rules by their score, i.e.,

$$\forall i > j \in \mathcal{D}, \quad g(r_i) > g(r_j),$$

where  $g(\cdot)$  is a rule scoring function.  $\mathcal{D}$  is the set of all possible rule indices.

**Definition 2.** Let  $\theta(\cdot, \cdot)$  be a rule out-ranking certainty function satisfying the following properties:

- $\forall i, j \in \mathcal{D}, \quad \theta(r_i, r_j) + \theta(r_j, r_i) = 1.$
- $\forall k \in \mathcal{D}, \forall i > j, \quad \theta(r_k, r_j) > \theta(r_k, r_i).$

**Lemma 3.**

$$\forall i \in \mathcal{D}, \quad \theta(r_i, r_i) = 0.5 \quad \text{and} \quad \forall i > j, \quad \theta(r_i, r_j) > \theta(r_j, r_i) \quad \text{and} \quad \forall k \in \mathcal{D}, i > j, \quad \theta(r_i, r_k) > \theta(r_j, r_k)$$

*Proof.* The first point is trivial from the first property of Def. 2:

$$\forall i \in \mathcal{D}, \quad \theta(r_i, r_i) + \theta(r_i, r_i) = 1 \Leftrightarrow \theta(r_i, r_i) = 0.5$$

The second point is proven as a direct consequence of the second property of Def. 2:

$$\forall k \in \mathcal{D}, \forall i, j \in \mathcal{D}, \quad \theta(r_k, r_j) > \theta(r_k, r_i)$$

We apply this property with  $k = i$  and  $j$  respectively:

$$\theta(r_i, r_j) > \theta(r_i, r_i) = 0.5 = \theta(\underbrace{r_j}_{r_k}, r_j) > \theta(r_j, r_i)$$

The third point is direct:

$$\theta(r_k, r_j) > \theta(r_k, r_i) \Leftrightarrow 1 - \theta(r_k, r_j) < 1 - \theta(r_k, r_i) \Leftrightarrow \theta(r_i, r_k) > \theta(r_j, r_k)$$

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**Definition 4** (Stochastic Matrix  $T$ ). Let  $T \in \mathcal{M}_n([0, 1])$  be the stochastic matrix associated with the Markov chain whose states are the decision rules. For all  $i, j$ ,  $T_{i,j}$  represents the probability of transitioning from state  $i$  to state  $j$ . The matrix  $T$  satisfies the following conditions:

1.  $\forall r_j \notin N(r_i), \quad T_{r_i, r_j} = 0.$
2.  $\forall r_j \in N(r_i), \quad T_{r_i, r_j} = \frac{1}{|N(r_i)|} \theta(r_j, r_i).$
3.  $T_{r_i, r_i} = \frac{1}{|N(r_i)|} \sum_{r_j \in N(r_i)} \theta(r_i, r_j).$

This ensures that the probabilities of exiting state  $i$  sum up to one.

**Theorem 5.** *For an irreducible chain, there is an equilibrium distribution.*

*Proof.* Let  $f_{i,i}(k)$  be the probability of the first return in  $i$  after exactly  $k$  steps. By definition, the periodicity of the Markov chain is  $\mathcal{P}_i = \gcd\{k : f_{i,i}(k) \neq 0\}$ . Since the Markov chain is irreducible, all states have the same periodicity. Let  $i \in \mathcal{D}$ , for all  $r_j \in N(r_i)$ ,  $\theta(r_i, r_j) > 0$  otherwise we wouldn't be able to access the state  $r_i$  contradicting irreducibility. Thus the period of the state  $r_i$  is one, showing the aperiodicity of the Markov chain. Since the chain is finite, we can conclude that it is ergodic and thus has an equilibrium distribution.  $\square$

**Theorem 6.** *Let  $\pi$  be the probability vector associated with the equilibrium distribution of the Markov chain of transition matrix  $T$ . In the general case, we have:*

$$\forall i, j \in \mathcal{D}, \quad i > j \implies \frac{\pi_i}{|N(r_i)|} > \frac{\pi_j}{|N(r_j)|}.$$

*Proof.* Suppose, for contradiction, that there exist  $i, j \in \mathcal{D}$  such that  $i > j$  but  $\pi_j \geq \pi_i$ . Since the chain is aperiodic and irreducible, the balance equation holds:

$$\forall i \in \mathcal{D}, \quad \sum_{k=1, k \neq i}^n [\pi_k \theta(r_i, r_k) - \pi_i \theta(r_k, r_i)] = 0.$$

Applying this to the supposed  $i$  and  $j$ , we obtain:

$$\begin{aligned} & [\pi_j \theta(r_i, r_j) - \pi_i \theta(r_j, r_i)] + \sum_{k=1, k \neq i, j}^n [\pi_k \theta(r_i, r_k) - \pi_i \theta(r_k, r_i)] \\ & - \left( [\pi_i \theta(r_j, r_i) - \pi_j \theta(r_i, r_j)] + \sum_{k=1, k \neq i, j}^n [\pi_k \theta(r_j, r_k) - \pi_j \theta(r_k, r_j)] \right) \\ & = 2 \underbrace{[\pi_j \theta(r_i, r_j) - \pi_i \theta(r_j, r_i)]}_{>0} + \sum_{k=1, k \neq i, j}^n \underbrace{[\pi_k (\theta(r_i, r_k) - \theta(r_j, r_k))]}_{>0} + \underbrace{[\pi_j \theta(r_k, r_j) - \pi_i \theta(r_k, r_i)]}_{>0} > 0. \end{aligned}$$

This contradiction establishes that  $\pi_i \geq \pi_j$ .  $\square$

*Remark 7.* Let  $N(\cdot)$  be a neighborhood such that  $|N(r_i)|$  increases with the rule's score  $g(r_i)$ . The equilibrium probability of a rule is increasing with its score:  $\forall i, j \in \mathcal{D}, \quad i > j \implies \pi_i > \pi_j$ . Here lies the inherent difficulty of an algorithm. The easiest way to satisfy this condition is to ensure that all of the rules have the same number of neighbors.