

Algorithm 1: CCC algorithm**1 Function** get_partitions(\mathbf{v} , k_{\max}):**Output:** Ω_r : clustering with r clusters over n objects**2 if** $\mathbf{v} \in \mathbb{R}^n$ **then****3 for** $r \leftarrow 2$ **to** $\min\{k_{\max}, |\mathbf{v}| - 1\}$ **do****4** $\boldsymbol{\rho} \leftarrow (\rho_\ell \mid \Pr(v_i < \rho_\ell) \leq (\ell - 1)/r), \forall \ell \in [1, r + 1]$ **5** $\Omega_{r\ell} \leftarrow \{i \mid \rho_\ell < v_i \leq \rho_{\ell+1}\}, \forall \ell \in [1, r]$ **6 else****7** $\mathcal{C} \leftarrow \cup_j \{v_i\}$ **8** $r \leftarrow |\mathcal{C}|$ **9** $\Omega_{rc} \leftarrow \{i \mid v_i = \mathcal{C}_c\}, \forall c \in [1, r]$ **// Remove singleton partitions****10** $\Omega \leftarrow \{\Omega_r \mid |\Omega_r| > 1\}, \forall r$ **11 return** Ω **13 Function** ccc(\mathbf{x} , \mathbf{y} , k_{\max}):**Input:** \mathbf{x} : feature values on n objects \mathbf{y} : feature values on n objects k_{\max} : maximum number of internal clusters**Output:** c : similarity value for \mathbf{x} and \mathbf{y} ($c \in [0, 1]$)**14** $\Omega^{\mathbf{x}} = \text{get_partitions}(\mathbf{x}, k_{\max})$ **15** $\Omega^{\mathbf{y}} = \text{get_partitions}(\mathbf{y}, k_{\max})$ **16** $c \leftarrow \max\{\mathcal{A}(\Omega_p^{\mathbf{x}}, \Omega_q^{\mathbf{y}})\}, \forall p, q$ **17 return** $\max(c, 0)$