

Algorithm 1: Clustermatch algorithm

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1 Function get_partitions( $\mathbf{v}$ ,  $k_{\max}$ ):  
   Output:  
    $\Omega_r$ : 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      $\rho \leftarrow (\rho_\ell \mid \Pr(v_i < \rho_\ell) \leq (\ell - 1)/r), \forall \ell \in [1, r + 1]$   
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  $\Omega$   
12  
13 Function clustermatch( $\mathbf{x}$ ,  $\mathbf{y}$ ,  $k_{\max}$ ):  
   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  $c$ 
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