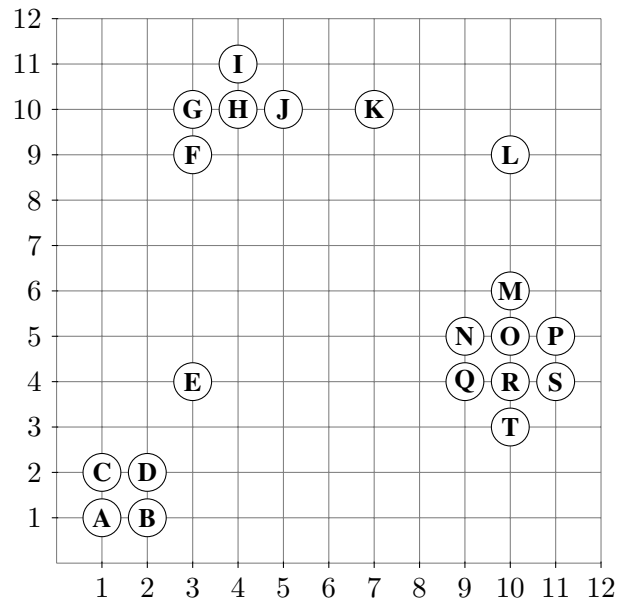


Knowledge Discovery and Data Mining I
 WS 2019/20

Exercise 8: Agglomerative Clustering, OPTICS, Clustering Evaluation

Exercise 8-1 Hierarchical Clustering

Given the following data set:



As distance function, use Manhattan Distance:

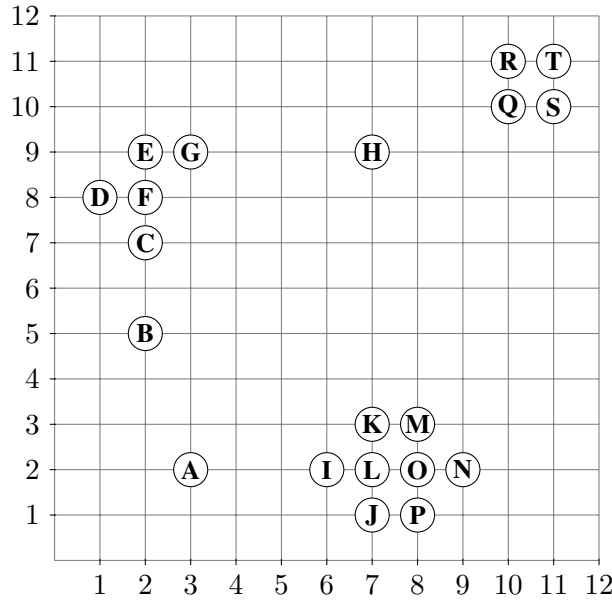
$$L_1(x, y) = |x_1 - y_1| + |x_2 - y_2|$$

Compute two dendrograms for this data set. To compute the distance of sets of objects, use

- the single-link method
- the complete-link method

Hint: With discrete distance values, nodes may have more than two children.

Exercise 8-2 OPTICS



As distance function, use Manhattan distance $L_1(a, b) := |a_1 - b_1| + |a_2 - b_2|$.

Construct an OPTICS reachability plot for each of the following parameter settings. In case of a tie always proceed with the first candidate in alphabetical order.

- (a) $\epsilon = 5$ and $minPts = 2$
- (b) $\epsilon = 5$ and $minPts = 4$
- (c) $\epsilon = 2$ and $minPts = 4$
- (d) $\epsilon = \infty$ and $minPts = 4$

Exercise 8-3 Efficient Evaluation of Clusterings

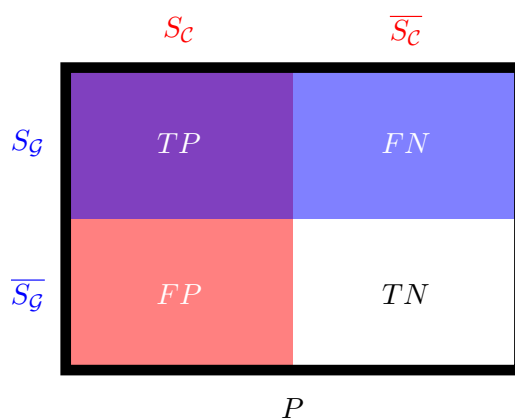
Let D be a database of size $n := |D|$, and let \mathcal{C}, \mathcal{G} be two partitionings of D . Furthermore, let $k := |\mathcal{C}|$ and $l := |\mathcal{G}|$ be the number of partitions, and assume that the contingency table is provided as a $(k \times l)$ matrix, where $N_{ij} = |C_i \cap G_j|$ denotes one cell in this table.

As in the lecture slides, let $P := \{(o, p) \in D^2 \mid o \neq p\}$ denote the set of all pairs, and $S_{\mathcal{C}} = \{(o, p) \in P \mid \exists C_i \in \mathcal{C} : \{o, p\} \subseteq C_i\}$ be the set of pairs that are contained in a common cluster C_i in \mathcal{C} . In addition, $\overline{S_{\mathcal{C}}}$ denotes the complement of $S_{\mathcal{C}}$ in P , i.e. $\overline{S_{\mathcal{C}}} = P \setminus S_{\mathcal{C}}$. $S_{\mathcal{G}}$ and $\overline{S_{\mathcal{G}}}$ are used analogously.

Using these four sets, we can now define the

- **True Positives (TP):** Same labelling in \mathcal{C} and same labelling in \mathcal{G} , i.e. $TP = |S_{\mathcal{C}} \cap S_{\mathcal{G}}|$
- **False Positives (FP):** Same labelling in \mathcal{C} , but different labelling in \mathcal{G} , i.e. $FP = |S_{\mathcal{C}} \cap \overline{S_{\mathcal{G}}}|$
- **False Negatives (FN):** Different labelling in \mathcal{C} , but same labelling in \mathcal{G} , i.e. $FN = |\overline{S_{\mathcal{C}}} \cap S_{\mathcal{G}}|$
- **True Negatives (TN):** Different labelling in \mathcal{C} , and different labelling in \mathcal{G} , i.e. $TN = |\overline{S_{\mathcal{C}}} \cap \overline{S_{\mathcal{G}}}|$

The relation of these four sets and $S_{\mathcal{C}}$ as well as $S_{\mathcal{G}}$ is also visualised in the following Venn diagram:



For each of these cardinalities, provide a method to obtain the numbers solely from the contingency table, i.e. without explicitly enumerating set of all pairs (which requires $\mathcal{O}(n^2)$ time).

- (a) $TP = |S_C \cap S_G|$,
- (b) $FP = |S_C \cap \overline{S_G}|$,
- (c) $FN = |\overline{S_C} \cap S_G|$,
- (d) $TN = |\overline{S_C} \cap \overline{S_G}|$.

Exercise 8-4 Mutual Information

Given are two clusterings of $D = \{A, \dots, Z\}$:

- $\mathcal{C} = \{\{A, D, E, I, K, L, M, N, T\}, \{B, F, O, Q, R, S, X, Y\}, \{C, G, H, P, V\}, \{J, U, W, Z\}\}$.
- $\mathcal{G} = \{\{A, F, M, Y\}, \{B, E, H, N, O, Q, R, Z\}, \{C, G, U, W\}, \{D, I, K, L, P, T, V\}, \{J, S, X\}\}$.

- (a) Setup the contingency table, i.e. compute the sizes $|C_i \cap G_j|$ for $i = 1, \dots, 4$, and $j = 1, \dots, 5$.
- (b) Using the contingency table from (a), compute the entropy of \mathcal{C} and \mathcal{G} , i.e. $H(\mathcal{C})$ and $H(\mathcal{G})$.
- (c) Using the contingency table from (a), compute the mutual entropy $H(\mathcal{C} | \mathcal{G})$.
- (d) Combine the results from (b) and (c) to obtain the normalised mutual information. What does this value tell about the two clusterings?