Outlier re-insertion via Mixed-Integer Linear Programming

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Outlier re-insertion 1

We assume that the distance metric d_{ij} is fixed for the remainder of this section. Let \mathcal{N} be the set of all points, C_i the set of co-class neighbors of point i, \bar{C}_i the set of non co-class neighbors of point i, and O the current set of outliers. Let $\mathcal{IK} = \{(i,k) \in \mathcal{N} \times \mathcal{N} : \nexists j \in \mathcal{C}_i \setminus \mathcal{O} : d_{ij} + \epsilon \leq d_{ik}, k \in \bar{\mathcal{C}}_i \}$. Set $M_{ik} = 1 - d_{ik}$. (We are making use of our normalization $d_{ij} \in [0,1]$ here.) For any outlier penalty parameter $\rho > 0$ (I suggest $\rho = 1$), the following mixed-integer linear program attempts to minimize the number of outliers given a fixed distance metric.

$$\min_{\mathbf{y}, \mathbf{z}} \quad \rho \sum_{i \in \mathcal{O}} z_i \tag{1a}$$

s.t.
$$\sum_{j \in \mathcal{C}_i} d_{ij} y_{ij} + \epsilon \le d_{ik} + M_{ik} z_k \qquad \forall (i, k) \in \mathcal{IK}$$

$$\sum_{j \in \mathcal{C}_i} y_{ij} = 1 \qquad \forall i \in \mathcal{N} \backslash \mathcal{O}, j \in \mathcal{C}_i$$
(1b)

$$\sum_{j \in \mathcal{C}_i} y_{ij} = 1 \qquad \forall i \in \mathcal{N} \backslash \mathcal{O}, j \in \mathcal{C}_i$$
 (1c)

$$\sum_{j \in \mathcal{C}_i} y_{ij} = 1 - z_i \qquad \forall i \in \mathcal{O}, j \in \mathcal{C}_i$$
 (1d)

$$y_{ij} \le 1 - z_j$$
 $\forall i \in \mathcal{N}, j \in \mathcal{C}_i \cap \mathcal{O}$ (1e)

$$y_{ij} \in \{0, 1\}$$
 $\forall i \in \mathcal{N}, j \in \mathcal{C}_i$ (1f)

$$z_i \in \{0, 1\} \qquad \forall i \in \mathcal{O} \tag{1g}$$

$$z_i = 0 \qquad \forall i \in \mathcal{N} \setminus \mathcal{O}$$
 (1h)

Several observations:

1. Bolun wrote the RHS of constraint (??) as $M_i(z_i + z_k)$. This is correct, but the formulation above is tighter, meaning the linear programming relaxation provides a better lower bound, and thus optimizers believe it will be better. Why can we replace $M_i(z_i + z_k)$ with $M_{ik}z_k$? Now that we have constraints (??), which effectively set the LHS of (??) to ϵ if point i is deemed an outlier $(z_i = 1)$, there is no reason to include $M_{ik}z_i$ on the RHS of (??) as well.

2. What is the set \mathcal{IK} doing? What is its purpose? Given that the distance metric is fixed, for any pair of points $(i, k) : i \in \mathcal{N}, k \in \bar{\mathcal{C}}_i$, if

$$\min\{d_{ij}: j \in \mathcal{C}_i \setminus \mathcal{O}\} + \epsilon \leq d_{ik}$$

which holds if and only if

$$\exists j \in \mathcal{C}_i \backslash \mathcal{O} : d_{ij} + \epsilon \le d_{ik} , \qquad (2)$$

then the corresponding constraint (??) is already satisfied (and will continue to be satisfied once outliers are considered) and is therefore redundant.

3. Is it really worth the trouble to create the set \mathcal{IK} ? Won't the solver take care of this? While solvers are good at eliminating redundant constraints, it is best to help them eliminate constraints a priori especially when we can immediately identify them. This saves the solver time when building the optimization model and in preprocessing. Example: Suppose there are 10 classes, 100 points per class, for a total of 1000 points. For simplicity, suppose there are 3 outliers per class (that is, our algorithm has thus far identified 3 outliers per class so that $|\mathcal{O}| = 30$). There are 900,000 constraints that must be included if constraints (??) are written " $\forall i \in \mathcal{N}, k \in \bar{\mathcal{L}}_i$ " since, for every point i (of which there are 1,000), there are 900 non-neighbors. However, if we were to replace " $\forall i \in \mathcal{N}, k \in \bar{\mathcal{C}}_i$ " with $\forall (i,k) \in \mathcal{IK}$, then at least 846,810 (94%) of the constraints of type (??) can be eliminated before doing a single optimization. Why? In this example, there are 970 non-outliers, each of which has 9×97 non-neighbors that are not outliers. These 846,810 pairs of points have already been shown to satisfy constraints (??) to arrive at the current distance metric. We write "at least" because there are likely many other pairs of points involving outliers (i.e., $k \in \mathcal{O}$) for which the condition (??) also holds.

1.1 AIMMS implementation

In AIMMS, there are several items to be aware of. First, define the set \mathcal{IK} as follows:

$$\mathcal{IK} = \{(i,k)|k \in \bar{\mathcal{C}}_i \text{ and not exists}(j \in \mathcal{C}_i \setminus \mathcal{O}|d_{ij} + \epsilon \leq d_{ik})\}$$

Second, z_i should only be defined for $i \in \mathcal{O}$. You should not define it for all $i \in \mathcal{N}$ and then set $z_i = 0$ for all $i \in \mathcal{N} \setminus \mathcal{O}$ as in (??). Note that AIMMS understands that if z_i appears in a constraint written for all $i \in \mathcal{N}$, it will only include it when z_i exists.

Third, in summary, the following formulation should be used in AIMMS:

$$\min_{\mathbf{y}, \mathbf{z}} \quad \rho \sum_{i \in \mathcal{O}} z_i \tag{3a}$$

$$\begin{array}{ll}
\mathbf{y}, \mathbf{z} & \neq \sum_{i \in \mathcal{O}} z_i \\
\text{s.t.} & \sum_{j \in \mathcal{C}_i} d_{ij} y_{ij} + \epsilon \leq d_{ik} + M_{ik} z_k & \forall (i, k) \in \mathcal{IK} \\
& \sum_{j \in \mathcal{C}_i} y_{ij} = 1 - z_i & \forall i \in \mathcal{N}, j \in \mathcal{C}_i
\end{array} \tag{3b}$$

$$\sum_{j \in \mathcal{C}_i} y_{ij} = 1 - z_i \qquad \forall i \in \mathcal{N}, j \in \mathcal{C}_i$$
 (3c)

$$y_{ij} \le 1 - z_j$$
 $\forall i \in \mathcal{N}, j \in \mathcal{C}_i \cap \mathcal{O}$ (3d)

$$y_{ij} \in \{0, 1\}$$
 $\forall i \in \mathcal{N}, j \in \mathcal{C}_i$ (3e)

$$z_i \in \{0, 1\} \qquad \forall i \in \mathcal{O} \tag{3f}$$