On the combinatorics of the 2-class classification problem

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Abstract

A set of points $\mathcal{X} = \mathcal{X}_B \cup \mathcal{X}_R \subseteq \mathbb{R}^d$ is linearly separable if the convex hulls of $\mathcal{X}_B$ and $\mathcal{X}_R$ are disjoint, hence there exists a hyperplane separating $\mathcal{X}_B$ from $\mathcal{X}_R$. Such a hyperplane provides a method for classifying new points, according to which side of the hyperplane the new points lie. When such a linear separation is not possible, it may still be possible to partition $\mathcal{X}_B$ and $\mathcal{X}_R$ into prespecified numbers of groups, in such a way that every group from $\mathcal{X}_B$ is linearly separable from every group from $\mathcal{X}_R$. We may also discard some points as outliers, and seek to minimize the number of outliers necessary to find such a partition. Based on these ideas, Bertsimas and Shioda proposed the classification and regression by integer optimization (CRIo) method in 2007. In this work we explore the integer programming aspects of the classification part of CRIo, in particular theoretical properties of the associated formulation. We are able to find facet-inducing inequalities coming from the stable set polytope, hence showing that this classification problem has exploitable combinatorial properties.

Keywords: classification, integer programming, polyhedral combinatorics
1 Introduction

The data classification problem is a widely-studied topic within the machine learning and the data mining communities. Briefly speaking, this problem aims at finding a partition of an euclidean space that represents the underlying pattern of a set of points. Many computational methods for tackling this problem exist, including decision trees [4], linear and quadratic programming [5, 7], and support vector machines [10], among others. Only recently has the use of discrete optimization, and in particular integer programming, been proposed within these settings [1, 3, 9].

In this work we are interested in the method called classification and regression via integer optimization (CRIO), proposed by Bertsimas and Shioda in [3]. This pioneering work presents a remarkable application of integer programming to this field, by proposing to classify points by hyperplanes separating pairs of groups of points. Given two sets of points in \( \mathbb{R}^d \), it is not always possible to find a hyperplane separating them, as Figure 1(a) shows. However, assuming that the underlying pattern can be expressed in terms of convex sets, it may be possible to subdivide each set into groups, and then find hyperplanes separating each pair of groups coming from different sets, as in Figure 1(b). In [3] a method is proposed to find such linearly separable groups, which includes the solution of a mixed integer program as a key step. This model also identifies a number of points that deviate from the underlying pattern and need to be disregarded in order to enable the desired classification. Further developments [1, 9] and computational experiments performed with instances from the literature show that this approach is promising as an effective alternative in practice.

There exist general algorithms for solving integer programs, and very strong implementations of these algorithms are available. However, since the general integer programming problem is NP-hard, the running times of these approaches may be intractably large, depending on the instance size and the model structure. Due to this fact, it is usual to study polyhedra associated with specific integer programming formulations, in order to find strong valid inequalities that may be helpful within these algorithms. Such polyhedral explorations may also reveal interesting structures and relations among different problems. In this work we are interested in such issues concerning a 0-1 integer program derived from CRIO in order to partition each set into groups, find the separating hyperplanes among them, and identify a minimum number of disregarded points. The 0-1 integer program studied results from a projection of a mixed integer one. The main goal that motivated this work is the question of whether there exist standard combinatorial
structures inherent to the CRIO method.

The remainder of this work is organized as follows. Section 2 introduces the problem in detail and sets out the notation used throughout the paper. Section 3 contains our initial polyhedral study, including several families of facet-inducing inequalities. A first affirmative answer to the existence of combinatorial structures within the associated polyhedra, in the form of facets coming from the stable set polytope, is the subject of Section 4. Finally, Section 5 closes the paper with conclusions and lines for future research.

2 The 2-class problem

Let $\mathcal{X} = \{x_1, \ldots, x_m\} \subset \mathbb{R}^d$ be a set of samples (also referred to as points) and consider a partition of the index set $[m] := \{1, \ldots, m\}$ into two classes $B$ and $R$, defining subsets $\mathcal{X}_B = \{x_i : i \in B\}$ and $\mathcal{X}_R = \{x_j : j \in R\}$. We start with some preliminary definitions before stating the mixed integer formulation.

2.1 Linear separability

The set $\mathcal{X}$ is linearly separable if and only if $\text{conv}(\mathcal{X}_B) \cap \text{conv}(\mathcal{X}_R) = \emptyset$, where $\text{conv}(X)$ denotes the convex hull of $X$. It is worth noting that being linearly separable is equivalent to be partitionable into two convex sets respecting $B$ and $R$. Hence, the linear separability of $\mathcal{X}$ is characterized by the fact that
the set of all $\lambda_i \geq 0$, $i \in [m]$, such that

$$\sum_{i \in B} \lambda_i x_i = \sum_{j \in R} \lambda_j x_j,$$

$$\sum_{i \in B} \lambda_i = \sum_{j \in R} \lambda_j = 1$$

must be empty. Applying Farkas’ Lemma, we get that this characterization is equivalent to state that there exist $p \in \mathbb{R}^d$ and $q, r \in \mathbb{R}$ such that

$$r - q < 0,$$

$$px_i + q \leq 0 \quad \text{for } i \in B,$$

$$px_j + r \geq 0 \quad \text{for } j \in R.$$ 

Adding $q$ in both sides of the last inequality and defining $\delta = q - r$, we get

$$px_i + q \leq 0 \quad \text{for } i \in B,$$

$$px_j + q \geq \delta \quad \text{for } j \in R.$$ 

If these conditions hold for some $p \in \mathbb{R}^d$ and $q, \delta \in \mathbb{R}$, then set $\delta' = \delta/2$, $q' = q - \delta'$ and divide $p$, $q'$, and $\delta'$ by $\delta'$ to conclude that $X$ is linearly separable if and only if the set $Q$ of hyperplanes $(p, q)$ such that

$$px_i + q \leq -1 \quad \text{for } i \in B,$$

$$px_j + q \geq 1 \quad \text{for } j \in R$$

is not empty. The example in Figure 1 clearly does not satisfy this property and, then, is not linearly separable.

### 2.2 Piecewise linear separability

In several situations of interest, the underlying pattern of the points in $X$ cannot be expressed by two convex sets only. In order to cope with such a scenario, let $L_B$ and $L_R$, $L_B \cap L_R = \emptyset$, be two sets of group indices specified for $B$ and $R$, respectively. An assignment of points in $X_B$ to indices in $L_B$ defines groups of points in $X_B$. Group $k \in L_B$ is the subset of $X_B$ assigned to index $k$. Groups of points in $X_R$ are defined similarly. A piecewise linear separation of $X$ is an assignment of points in $X_B$ to indices in $L_B$ and of points in $X_R$ to indices in $L_R$ such that groups $k$ and $\ell$ are linearly separable, for all $k \in L_B$ and $\ell \in L_R$. A hyperplane $(p_{k\ell}, q_{k\ell})$ separating groups $k$ and $\ell$ is such that

$$p_{k\ell} x_i + q_{k\ell} \leq -1 \quad \text{for } i \in B \text{ such that } x_i \text{ is assigned to } k,$$

$$p_{k\ell} x_j + q_{k\ell} \geq 1 \quad \text{for } j \in R \text{ such that } x_j \text{ is assigned to } \ell.$$
2.3 Mixed integer programming formulation

We now state a mixed integer programming formulation inspired by the one that constitutes the key step in the CRIO method [3]. The input of the problem is formed by the set of points $\mathcal{X}$, the partition of point indices $B$ and $R$, and the sets $L_B$ and $L_R$ of group indices. The objective is to find an assignment of points to groups that induces a piecewise linear separation of $\mathcal{X}$.

For every group $k \in L_B$ and every group $\ell \in L_R$, the formulation contains the variables $p_{k\ell} \in \mathbb{R}^d$ and $q_{k\ell} \in \mathbb{R}$, in such a way that $p_{k\ell}x = q_{k\ell}$ is the hyperplane separating the groups $k$ and $\ell$. For $i \in B$ and $k \in L_B$, the binary variable $z_{ik}$ represents whether $x_i$ is assigned to group $k$ or not. For $j \in R$ and $\ell \in L_R$, the binary variable $z_{j\ell}$ represents whether $x_j$ is assigned to group $\ell$ or not. In this setting, we can provide the formulation corresponding to the maximization of

$$\sum_{i \in B} \sum_{k \in L_B} z_{ik} + \sum_{j \in R} \sum_{\ell \in L_R} z_{j\ell} \quad (1)$$

subject to

$$p_{k\ell}x_i + q_{k\ell} \leq M - (M + 1)z_{ik} \quad \forall i \in B, \forall k \in L_B, \forall \ell \in L_R, \quad (2)$$

$$p_{k\ell}x_j + q_{k\ell} \geq -M + (M + 1)z_{j\ell} \quad \forall j \in R, \forall \ell \in L_R, \forall k \in L_B, \quad (3)$$

$$\sum_{k \in L_B} z_{ik} \leq 1 \quad \forall i \in B, \quad (4)$$

$$\sum_{\ell \in L_R} z_{j\ell} \leq 1 \quad \forall j \in R, \quad (5)$$

$$(p_{k\ell}, q_{k\ell}) \in \mathbb{R}^{d+1} \quad \forall k \in L_B, \forall \ell \in L_R, \quad (6)$$

$$z_{ik} \in \{0, 1\} \quad \forall (i, k) \in (B \times L_B) \cup (R \times L_R), \quad (7)$$

where $M$ is a big positive number. Note that the feasibility of the model does not depend on the actual value of $M$, namely if $M$ is small then some solutions are lost but the problem remains feasible.

**Definition 1.** Given an instance $\mathcal{I} = (\mathcal{X}, B, R, L_B, L_R)$ of the problem, we call $P_\mathcal{I}$ the convex hull of the points $(p, q, z)$ satisfying (2)-(7).

In addition, we can introduce a binary variable $o_i$ for each $i \in [m]$ to specify whether sample $x_i$ is an outlier or not, i.e., if it is not assigned to any group. With these settings, the objective function can be written as

$$\text{minimize } \sum_{i \in [m]} o_i$$
and constraints (4) and (5) should be replaced by the following constraints

\[ \sum_{k \in L_B} B z_{ik} = 1 - o_i \quad \forall i \in B, \tag{8} \]

\[ \sum_{l \in L_R} R z_{j\ell} = 1 - o_j \quad \forall j \in R. \tag{9} \]

2.4 Integer programming formulation

We now discuss an integer programming formulation resulting from a projection of (2)-(7) onto the space of the \( z \)-variables. For this purpose, let \( Q_{k\ell} \) denote the set of points \((p_{k\ell}, q_{k\ell}, z_{Bk}, z_{R\ell})\) satisfying (2)-(3) and (6)-(7) for \( \ell \in L_R \) and \( k \in L_B \), where \( z_{Bk} \) and \( z_{R\ell} \) are vectors constituted by the variables \( z_{ik} \), for all \( i \in B \), and \( z_{j\ell} \), for all \( j \in R \), respectively. Rewrite (2)-(3) for \( k \) and \( \ell \) as

\[
\begin{bmatrix}
X_B^T & \mathbf{1} \\
-X_R^T & -\mathbf{1}
\end{bmatrix} \begin{bmatrix} p_{k\ell} \\ q_{k\ell} \end{bmatrix} + (M + 1) \begin{bmatrix} z_{Bk} \\ z_{R\ell} \end{bmatrix} \leq M \cdot \mathbf{1},
\]

and denote by \( Q \) the combination of \( Q_{k\ell} \), for all \( \ell \in L_R \) and \( k \in L_B \). The projection of \( Q_{k\ell} \) onto \( \{0, 1\}^{B \cup R} \) is defined as

\[
Proj_z(Q_{k\ell}) = \{ (z_{Bk}, z_{R\ell}) \in \{0, 1\}^{B \cup R} : \exists (p_{k\ell}, q_{k\ell}, z_{Bk}, z_{R\ell}) \in Q_{k\ell} \}
\]

The combination of \( Proj_z(Q_{k\ell}) \), for all \( \ell \in L_R \) and \( k \in L_B \), gives \( Proj_z(Q) \), which in turn gives the set of all possibly intersecting group assignments. By Theorem 1.1 of [2], \( Proj_z(Q_{k\ell}) \) is given by the group assignments \( z \) satisfying

\[
(M + 1) \left( \sum_{i \in B} v_{ik} z_{ik} + \sum_{j \in R} v_{j\ell} z_{j\ell} \right) \leq M \left( \sum_{i \in B} v_{ik} + \sum_{j \in R} v_{j\ell} \right)
\]

for all \((v_{Bk}, v_{R\ell}) \geq 0\) such that

\[ \sum_{i \in B} v_{ik} x_i = \sum_{j \in R} v_{j\ell} x_j \quad (10) \]

\[ \sum_{i \in B} v_{ik} = \sum_{j \in R} v_{j\ell}. \quad (11) \]

It is worth noting that the convex hull of groups \( k \) and \( \ell \) intersect in a group assignment \( z \) if and only if there exists \((v_{Bk}, v_{R\ell})\) with \((z_{ik} = 0 \Rightarrow v_{ik} = 0)\),
(\(z_{j\ell} = 0 \Rightarrow v_{j\ell} = 0\)), and \(\sum_{i \in B} v_{ik} = 1\) such that (10)-(11) hold. Hence,

\[
(M + 1) \left( \sum_{i \in B} v_{ik}z_{ik} + \sum_{j \in R} v_{j\ell}z_{j\ell} \right) \leq 2M
\]

(12)

prevents such a group assignment to be chosen. The integer programming formulation consists in maximizing (1) over all binary points \(z\) of type (7) satisfying (4)-(5) and (12) for the extreme rays of the set defined by (10)-(11).

A final remark is in order with respect to this integer programming formulation. The projection of \(P_I\) onto \([0, 1]^{(B \times L_B) \cup (R \times L_R)}\), \(\text{Proj}_z(P_I)\), can be seen as the convex hull of the group assignments in \(\text{Proj}_z(Q)\) satisfying (4)-(5). In addition, the valid inequalities discussed in the next sections involve \(z\)-variables only. Consequently, Corollary 2.2 of [2] can then be applied to conclude that the facetness results of those sections are valid for \(\text{Proj}_z(P_I)\) as well.

3 Polyhedral study

In this section we are interested in facets of \(P_I\), with a particular interest in combinatorial structures originating facet-inducing inequalities. For \(k \in L_B\), \(\ell \in L_R\), and \(a \in [d]\), we denote by \(e^p_{k\ell a}\) the unit vector associated with the \(a\)-th coordinate of the variable vector \(p_{k\ell}\). Correspondingly, for \(k \in L_B\) and \(\ell \in L_R\), we denote by \(e^q_{k\ell}\) the unit vector associated with the variable \(q_{k\ell}\). For \(i \in B\) and \(k \in L_B\), let \(e^z_{ik}\) be the unit vector associated with the variable \(z_{ik}\). Finally, for \(j \in R\) and \(\ell \in L_R\), we denote by \(e^z_{j\ell}\) the unit vector associated with the variable \(z_{j\ell}\).

**Proposition 1.** \(P_I\) is full-dimensional.

**Proof.** In order to prove this proposition, we construct the following affinely independent feasible solutions.

1. Let \(b = (0, 0, 0)\) be the solution having all variables set to null values. Since \(z_{ik} = 0\) in \(b\) for every \(i \in [m]\), then no point is assigned to any group, and all constraints are satisfied, hence \(b\) is feasible.

2. For any \(k \in L_B\), \(\ell \in L_R\), and \(a \in [d]\), consider the solution \(b + \varepsilon e^p_{k\ell a}\), with \(0 < \varepsilon \leq M\). This solution is feasible since all points are outliers due to (2)-(3), and it is affinely independent w.r.t. the previous solutions, which have \(p_{k\ell a} = 0\).
3. Similarly, for any \( k \in L_B \) and \( \ell \in L_R \), the solution \( b + \varepsilon e_{k\ell}^q \) is feasible and affinely independent w.r.t. the previously-constructed solutions.

4. For any \( i \in B \) and \( k' \in L_B \), construct the solution \( b + e_{ik'} - \sum_{\ell \in L_R} e_{k'\ell}^q \). This solution trivially satisfies constraints (2) for \( k \neq k' \) and constraints (3). Constraint (2) for \( k = k' \) and \( \ell \in L_R \) takes the form \(-1 \leq 1\) (since \( q_{k'\ell} = -1 \) in this solution), hence it is satisfied. The remaining constraints are trivially satisfied. Furthermore, this solution is affinely independent w.r.t. the previous solutions, which have \( z_{ik'} = 0 \).

5. Similarly, for any \( j \in E \) and \( \ell \in L_R \), the solution \( b + e_{j\ell} + \sum_{k \in L_B} e_{k\ell}^q \) is feasible and affinely independent w.r.t. the previous ones.

The existence of these solutions shows that \( P_T \) is full-dimensional.

The solutions constructed within the proof of Proposition 1 allow to show the following facetness results in a quite straightforward way, hence the proof of the following proposition is omitted.

**Proposition 2.**

(i) The model constraints (4) and (5) are facet-inducing.

(ii) The bound \( z_{ik} \geq 0 \) is facet-inducing, for every \( i \in B \) and \( k \in L_B \).

(iii) The bound \( z_{j\ell} \geq 0 \) is facet-inducing, for every \( j \in R \) and \( \ell \in L_R \).

### 3.1 Convex-inclusion inequalities

We now explore families of valid inequalities for \( P_T \), and study their facetness properties. We first present a family of valid inequalities involving a point \( x_j \in X_R \) and a set of points in \( X_B \) whose convex hull contains \( x_j \). In this setting, we may consider the valid inequality given by the following proposition. We adopt the notation \( x_S := \{ x_i : i \in S \} \), for any \( S \subseteq [m] \).

**Proposition 3.** Let \( k \in L_B \), \( j \in R \) and \( S = \{ i_1, \ldots, i_s \} \subseteq B \) such that \( x_j \in \text{conv}(x_S) \). The convex-inclusion inequality

\[
\sum_{i \in S} z_{ik} + \sum_{\ell \in L_R} z_{j\ell} \leq s
\]

is valid for \( P_T \).
Proof. Let \((p, q, z) \in P_I\) be an integer feasible solution. If \(x_j\) is an outlier in this solution (i.e., \(z_{j\ell} = 0\) for every \(\ell \in L_R\)), then (13) is trivially satisfied, so assume \(z_{j\ell} = 1\) for some \(\ell \in L_R\). Since \(x_j\) is contained in the convex hull of the points \(x_S\), then there is no hyperplane separating \(x_j\) from \(x_S\). This implies that a solution having all the points in \(S\) assigned to the same group would not be feasible, hence \(\sum_{i \in S} z_{ik} \leq |S| - 1 = s - 1\), and (13) is satisfied.

Xavier and Campêlo [11] proposed a general facet-generating procedure that takes a valid (facet-inducing) inequality \(\pi x \leq \pi_0\) for a polytope \(P\) and a valid (facet-inducing) inequality for the face of \(P\) defined by \(\pi x = \pi_0\), and produces a new valid (facet-inducing) inequality for \(P\) by combining them. It is interesting to note that inequalities (13) are obtained with the proposed procedure by using (5) as \(\pi x \leq \pi_0\) along with \(\sum_{i \in S} z_{ik} \leq s - 1\), which is valid when \(\sum_{\ell \in L_R} z_{j\ell} = 1\).

Theorem 1. Assume \(|L_R| \geq 2\). The inequality (13) defines a facet of \(P_I\) if and only if \(S\) is minimal w.r.t. the property \(x_j \in \text{conv}(x_S)\) (i.e., \(x_j \notin \text{conv}(x_{S'})\) for every \(S' \subset S\)).

Proof. Assume first that \(S\) is minimal w.r.t. the property \(x_j \in \text{conv}(x_S)\). Let \(F\) be the face of \(P_I\) defined by (13), and let \((\lambda, \mu, \gamma)\) and \(\lambda_0\) be such that \(\lambda p + \mu q + \gamma z = \lambda_0\) for every \((p, q, z) \in F\). We shall verify that \((\lambda, \mu, \gamma)\) is a multiple of the coefficient vector of (13), thus showing that \(F\) is a facet of \(P_I\). To this end, let \(b = (0, q, z)\) be the solution obtained by setting \(z_{ik} = 1\) for every \(i \in S\), and all the remaining \(z\)-variables to 0 (i.e., all the points in \(X\setminus x_S\) are outliers). We set \(q_{k\ell} = -2\) for all \(k \in L_B\), \(\ell \in L_R\), and the remaining \(q\)-variables are set to 0. Note that constraints (2) for \(i \in S\) and corresponding to the groups \(k\) and \(\ell\) are strictly satisfied.

Claim 1: \(\lambda = 0\). For \(k' \in L_B\), \(\ell \in L_R\), and \(a \in [d]\), consider the solution \(b + \varepsilon e_{k'\ell a}^p\), which is feasible if \(\varepsilon\) is small enough. Indeed, constraints (2) for \(k = k'\) and \(i \in S\) are satisfied since they are strictly satisfied by \(b\). In this solution only one group is nonempty, and the solution satisfies (13) with equality. Together with \(b\), the existence of this solution implies \(\lambda_{k'\ell a} = 0\).

Claim 2: \(\mu = 0\). Similarly, for \(k' \in L_B\) and \(\ell \in L_R\), the solution \(b + \varepsilon e_{k'\ell}^q\) is feasible (for a small enough \(\varepsilon\)) and also satisfies (13) with equality. Again, the combination of this solution with \(b\) implies \(\mu_{k'\ell} = 0\).
Claim 3: $\gamma_{ik} = \gamma_{j\ell}$ for $i \in S$ and $\ell \in L_R$. Let $S' := S\setminus \{i\}$ and consider the solution $b_{i\ell}$ constructed by setting $z_{i'k} = 1$ for $i' \in S'$, $z_{j\ell} = 1$, and the remaining $z$-variables to 0 (including $z_{ik} = 0$). Since $x_j \not\in \text{conv}(x_{S'})$, then there exists a hyperplane $\pi y = \pi_0$ separating $x_j$ from $x_{S'}$. Let $p_{k\ell} = \pi$ and $q_{k\ell} = -\pi_0$, and set the remaining $p$- and $q$-variables to 0. The solution $b_{i\ell}$ thus constructed is feasible and satisfies (13) with equality, hence $(\lambda, \mu, \gamma) b = (\lambda, \mu, \gamma) b_{i\ell}$. Since $\lambda = 0$ and $\mu = 0$, we conclude that $\gamma_{ik} = \gamma_{j\ell}$. ◦

Claim 4: $\gamma_{j'\ell} = 0$ for $j' \neq j$ and $\ell \in L_R$. If $x_{j'}$ is not contained in $\text{conv}(x_S)$, then there exists a hyperplane separating $x_{j'}$ from $x_S$. Otherwise, if $x_{j'}$ is contained in $\text{conv}(x_S)$, since the points in $x_S$ are affinely independent (due to the minimality of $S$), then there exists $i \in S$ such that $x_{j'} \not\in \text{conv}(x_S\setminus \{i\})$. In both cases, there exists some $i \in S$ such that we can find a hyperplane separating $x_{j'}$ from $\text{conv}(x_S\setminus \{i\})$. Construct a solution $b_{j'\ell}$ by setting $z_{i'k} = 1$ for all $i' \in S\setminus \{i\}$, $z_{j\ell} = 1$, and $z_{j'\ell} = 1$ for any $\ell \in L_R \setminus \{\ell'\}$ (recall that $|L_R| \geq 2$), and the remaining $z$-variables to 0. Finally, set $(p_{k\ell}, q_{k\ell})$ equal to the hyperplane separating $x_S\setminus \{i\}$ from $x_j$, and $(p_{k\ell'}, q_{k\ell'})$ equal to the hyperplane separating $x_S\setminus \{i\}$ from $x_{j'}$. This solution is feasible and satisfies (13) with equality. By resorting to Claim 1 Claim 2 and Claim 3, the existence of $b$ and $b_{j'\ell}$ shows that $\gamma_{j'\ell} = 0$. ◦

Claim 5: $\gamma_{ik'} = 0$ for $i \in S$ and $k' \neq k$. Take any $\ell \in L_R$ and consider the solution $b_{k'}$ obtained from $b - e_{ik'}^* + e_{ik'}^* + e_{j\ell}^*$ by setting $(p_{k\ell}, q_{k\ell})$ equal to the hyperplane separating $x_{S}\setminus \{i\}$ from $x_j$. This solution is feasible and satisfies (13) with equality. By resorting to the previous claims, the existence of $b_{k'}$ and $b$ shows that $\gamma_{ik'} = 0$. ◦

Claim 6: $\gamma_{ik'} = 0$ for $i \not\in S$ and $k' \in L_B$. Similarly to the proof of Claim 5, the solution $b_{ik'} = b + e_{ik'}^*$ is feasible and satisfies (13) with equality, so $\gamma_{ik'} = 0$. ◦

By combining these claims, we conclude that $(\lambda, \mu, \gamma)$ is a multiple of the coefficient vector of (13), hence $F$ is a facet of $P_L$.

For the converse direction, suppose that $x_j \in \text{conv}(x_S\setminus \{i\})$ for some $i \in S$. This implies that any solution having $z_{ik} = 0$ cannot satisfy (13) with equality, since such a solution must have $z_{i'k} = 1$ for every $i' \in S\setminus \{i\}$ and $z_{j\ell} = 1$ for some $\ell \in L_R$, in order to attain equality. However, since $x_j \in \text{conv}(x_S\setminus \{i\})$ then no hyperplane separating $x_j$ from $\text{conv}(x_S\setminus \{i\})$
exists, hence such a solution cannot be feasible. This implies that every solution in the face of $P_I$ induced by (13) satisfies $z_{ik} = 1$ and, since $P_I$ is full-dimensional, such a face is not a facet of $P_T$.

The symmetrical inequalities considering a point $x_i \in X_B$ and a set of points in $X_R$ whose convex hull includes $x_i$ have the same properties as (13).

3.2 Obstacle inequalities

Given two distinct points $y_1, y_2 \in \mathbb{R}^d$, a set $Y \subseteq \mathbb{R}^d$ is an obstacle between $y_1$ and $y_2$ if $\text{conv}(Y) \cap \text{conv}\{y_1, y_2\} \neq \emptyset$ (see Figure 2). We say that $Y$ is a trivial obstacle if $y_1 \in \text{conv}(Y)$ or $y_2 \in \text{conv}(Y)$, and that $Y$ is a minimal obstacle if $\text{conv}(Y \setminus \{y_i\}) \cap \text{conv}\{y_1, y_2\} = \emptyset$, for every $y_i \in Y$.

The presence of an obstacle between two points of $X_R$ implies that $X_B$ and $X_R$ are not linearly separable. However, it is interesting to remark that the converse is not true. There may not exist an obstacle between two points of $X_R$ even when $X_B$ and $X_R$ are not linearly separable; although in such a case an obstacle will exist between some pair of points of $\text{conv}(X_R)$ (recall that $X_B$ and $X_R$ are not polytopes but finite sets of points). An example of this is given by the sets

$$
X_B = \{(1, 1, 0, 0), (-2, 1, 0, 0), (1, -2, 0, 0)\}
$$

$$
X_R = \{(0, 0, 1, 1), (0, 0, -2, 1), (0, 0, 1, -2)\}
$$

for which $\text{conv}(X_B) \cap \text{conv}(X_R) = \{(0, 0, 0, 0)\}$.

**Proposition 4.** Let $S = \{i_1, \ldots, i_s\} \subseteq B$ be such that $x_S$ is an obstacle between two points $x_{j_1}, x_{j_2} \in X_R$, $j_1 \neq j_2$. For $k \in L_B$ and $\ell \in L_R$, the
obstacle inequality
\[ z_{j_1\ell} + z_{j_2\ell} + \sum_{i \in S} z_{ik} \leq s + 1 \] (14)
is valid for \( P_T \).

**Proof.** Let \((p, q, z) \in P_T\) be an integer feasible solution. Since the left-hand side of (14) contains \( s + 2 \) binary variables, we need only consider the case \( z_{j_1\ell} = z_{j_2\ell} = 1 \) and \( z_{ik} = 1 \) for every \( i \in S \). These variable values imply that \( x_{j_1} \) and \( x_{j_2} \) are assigned to the group \( \ell \), whereas all the points in \( x_S \) are assigned to the group \( k \). This is not possible in a feasible solution, since \( \text{conv}(x_S) \cap \text{conv}(\{x_{j_1}, x_{j_2}\}) \neq \emptyset \) implies that there is no hyperplane separating \( \{x_{j_1}, x_{j_2}\} \) from \( x_S \). Hence, \((p, q, z)\) satisfies (14). Since \((p, q, z)\) is an arbitrary solution, then (14) is valid for \( P_T \). \( \square \)

**Lemma 1.** Let \( S = \{i_1, \ldots , i_s\} \subseteq B \) be such that \( x_S \) is a nontrivial obstacle between \( x_{j_1}, x_{j_2} \in X_\ell \), \( j_1 \neq j_2 \). Let \( i \in B \setminus S \), and define \( S' = S \cup \{i\} \). Then, \( x_{j_1} \notin \text{conv}(x_S) \) or \( x_{j_2} \notin \text{conv}(x_{S'}) \).

**Proof.** Since \( x_{j_1} \notin \text{conv}(x_S) \), there exists an inequality \( \pi y \leq \pi_0 \) separating \( x_{j_1} \) from \( x_S \), namely \( \pi x_{i'} \leq \pi_0 \) for \( i' \in S \) and \( \pi x_{j_1} > \pi_0 \). We have \( \pi x_{j_2} \leq \pi_0 \), since otherwise \( \text{conv}(\{x_{j_1}, x_{j_2}\}) \cap \text{conv}(x_S) = \emptyset \). If \( x_{j_1} \in \text{conv}(x_{S'}) \) then \( \pi x_{i'} > \pi_0 \), since otherwise we would have \( \pi x_{i'} \leq \pi_0 \) for every \( i' \in S' \) but \( x_{j_1} \) does not satisfy this inequality. We have \( \pi x_{i'} \leq \pi_0 \) for every \( i' \in S \) and also \( \pi x_{j_2} \leq \pi_0 \). Since \( x_{j_2} \notin \text{conv}(x_S) \) (since \( x_S \) is nontrivial) and \( \pi x_{i'} > \pi_0 \), then \( x_{j_2} \notin \text{conv}(x_{S'}) \). \( \square \)

We say that a hyperplane \( \pi y \leq \pi_0 \) strictly separates the sets \( Y \subseteq \mathbb{R}^d \) and \( Y' \subseteq \mathbb{R}^d \) if \( \pi y < \pi_0 - 1 \) for every \( y \in Y \) and \( \pi y > \pi_0 + 1 \) for every \( y \in Y' \). If \( Y = \{y_1\} \) is a singleton, we also say that \( \pi y \leq \pi_0 \) strictly separates \( y_1 \) from \( Y' \) if \( \pi y \leq \pi_0 \) strictly separates \( Y \) and \( Y' \). Note that a strictly separating hyperplane between two sets of points exists whenever a separating plane exists between those sets.

**Theorem 2.** The inequality (14) defines a facet of \( P_T \) if and only if \( x_S \) is a non-trivial minimal obstacle between \( x_{j_1} \) and \( x_{j_2} \).

**Proof.** Assume first that \( x_S \) is non-trivial and minimal. Let \( F \) be the face of \( P_T \) defined by (14), and let \(( \lambda, \mu, \gamma) \) and \( \lambda_0 \) such that \( \lambda p + \mu q + \gamma z = \lambda_0 \) for every \((p, q, z) \in F\). Define \( b = (p, q, z) \) to be the solution obtained by setting \( z_{ik} = 1 \) for every \( i \in S \), \( z_{j_1\ell} = 1 \), and all the remaining z-variables to 0. Also, set \( p_{k\ell} \) and \( q_{k\ell} \) equal to a hyperplane strictly separating \( x_S \) from \( z_{j_1\ell} \) (which exists since \( x_{j_1} \notin \text{conv}(x_S) \)), \( q_{k\ell'} = -2 \) for \( \ell' \neq \ell \), and the rest of the \( p \)- and \( q \)-variables equal to 0.
Claim 1: $\lambda = 0$. For $k' \in L_B$, $\ell' \in L_R$, and $a \in [d]$, consider the solution $b_{k'\ell'a} := b + \varepsilon e_{k'\ell'a}^p$ for a small enough $\varepsilon$, which is feasible since only two groups are nonempty, and are strictly separated in $b$ (so a minor change in one coefficient keeps the separation). The existence of the solutions $b$ and $b_{k'\ell'a}$ implies $\lambda_{k'\ell'a} = 0$. \hfill \diamond

Claim 2: $\mu = 0$. Similarly, for $k' \in L_B$ and $\ell' \in L_R$, the solution $b_{k'\ell'} := b + \varepsilon e_{k'\ell'}^q$ is feasible and also satisfies (14) with equality. Again, the existence of $b$ and $b_{k'\ell'}$ implies $\mu_{k'\ell'} = 0$. \hfill \diamond

Claim 3: $\gamma_{ik} = \gamma_{j_1\ell} = \gamma_{j_2\ell}$ for $i \in S$. Let $S' := S \setminus \{i\}$ and consider the solution $b_i$ constructed by setting $z'_{ik} = 1$ for $i' \in S'$, $z_{j_1\ell} = z_{j_2\ell} = 1$, and the remaining $z$-variables to 0. Since $x_{S'}$ is not an obstacle between $x_{j_1}$ and $x_{j_2}$, then there exists a hyperplane $\pi y = \pi_0$ separating $\{x_{j_1}, x_{j_2}\}$ from $x_{S'}$. Let $p_{k\ell} = \pi$, $q_{k\ell} = -\pi_0$, $q_{k'\ell'} = -2$ for $\ell' \neq \ell$, $q_{k'\ell'} = 2$ for $k' \neq k$, and set the remaining $p$- and $q$-variables to 0. Claim 1 and Claim 2 ensure that $\lambda = 0$ and $\mu = 0$, hence the existence of $b_i$ and $b$ in $F$ implies $\gamma_{ik} = \gamma_{j_1\ell} = \gamma_{j_2\ell}$. A symmetric argument shows that $\gamma_{ik} = \gamma_{j_1\ell}$. \hfill \diamond

Claim 4: $\gamma_{j_1\ell} = \gamma_{j_2\ell} = 0$ for $\ell' \in L_R$, $\ell' \neq \ell$. Construct a solution $b_{\ell'}$ by setting $z_{j_2\ell'} = 1$, and setting $(p_{k\ell'}, q_{k\ell'})$ as a hyperplane strictly separating conv($x_{S'}$) from $x_{j_2}$. This solution is feasible and, together with $b$, Claim 1 and Claim 2 shows that $\gamma_{j_2\ell'} = 0$. A symmetric argument shows that $\gamma_{j_2\ell} = 0$. \hfill \diamond

Claim 5: $\gamma_{i'k} = 0$ for $i' \notin S$. Let $S' := S \cup \{i'\}$. Since $x_S$ is a nontrivial obstacle between $x_{j_1}$ and $x_{j_2}$, then Lemma 1 implies that $x_{j_1} \notin \text{conv}(x_{S'})$ or $x_{j_2} \notin \text{conv}(x_{S'})$. If $x_{j_1} \notin \text{conv}(x_{S'})$, then the solution $b_{j'} := b + \varepsilon e_{i'k}^p$ is feasible since there is a hyperplane separating $x_{j_1}$ from $x_{S'}$, and the existence of $b$ and $b_{j'}$ implies the claim. If $x_{j_2} \notin \text{conv}(x_{S'})$ then a symmetrical argument also settles the claim. \hfill \diamond

Claim 6: $\gamma_{j'\ell'} = 0$ for $j' \neq j_1, j_2$ and $\ell' \in L_R$. If $x_{j'}$ is not contained in conv($x_S$), then a similar argument as in the proof of Claim 5 settles this case. Otherwise, if $x_{j'}$ is contained in conv($x_S$), since the points in $x_S$ are affinely independent (due to the minimality of $x_S$), then there exists $i \in S$ such that $x_{j'} \notin \text{conv}(x_{S'})$, where $S' := S \setminus \{i\}$. This implies that we can find a hyperplane separating $\{x_{j_1}, x_{j'}\}$ from conv($x_{S'}$). Construct a solution $b_{j'\ell'}$ by setting $z_{i'k} = 1$ for all $i' \in S'$, and $z_{j_1\ell} = z_{j'\ell'} = 1$, and the remaining $z$-variables to 0.
• If $\ell \neq \ell'$, then set $(p_{k\ell}, q_{k\ell})$ equal to the hyperplane separating $x_{S'}$ from $x_p$, and $(p_{k\ell'}, q_{k\ell'})$ equal to the hyperplane separating $x_{S'}$ from $x_{j'}$.
• If $\ell = \ell'$, then set $(p_{k\ell}, q_{k\ell})$ equal to the hyperplane separating $x_{S'}$ from $\{x_{j_1}, x_{j_2}\}$.

This solution is feasible and satisfies (14) with equality. By resorting to Claim 1 and Claim 2, the existence of $b$ and $b_{j'\ell}$ shows that $\gamma_{j'\ell} = 0$. ♦

Claim 7: $\gamma_{ik'} = 0$ for $i \in S$ and $k' \neq k$. Define $S' = S \setminus \{i\}$. By the minimality of $x_S$, the solution $b_{ik}$ obtained by setting $z_{ik} = 1$ for $i' \in S'$, $z_{j_1} = z_{j_2} = 1$, and $z_{ik'} = 1$ is feasible, since the sets $S'$ and $\{x_{j_1}, x_{j_2}\}$ are linearly separable, and so are the sets $\{x_i\}$ and $\{x_{j_1}, x_{j_2}\}$. Hence we can set the variables $(p_{k\ell}, q_{k\ell})$ and $(p_{k'\ell}, q_{k'\ell})$ to the corresponding separating hyperplanes. Since $\lambda = 0$ and $\mu = 0$, the existence of $b$ and $b_{ik'}$ implies that $\gamma_{ik'} = 0$. ♦

Claim 8: $\gamma_{ik'} = 0$ for $i \notin S$ and $k' \neq k$. Similarly to the proof of Claim 5, the solution $b_{ik'} = b + e_{ik'}$ is feasible and satisfies (14) with equality, so $\gamma_{ik'} = 0$. ♦

By combining these claims, we conclude that $(\lambda, \mu, \gamma)$ is a multiple of the coefficient vector of (14), hence $F$ is a facet of $P_T$.

For the converse direction, if $x_S$ is trivial with $x_{i_1} \in \text{conv}(x_S)$, then (14) is the sum of $z_{j_2\ell} \leq 1$ and the inequality (13) associated with $x_{i_1}$ and $S$ (analogously if $x_{j_2} \in \text{conv}(x_S)$). On the other hand, if $S$ is not minimal, say $S' := S \setminus \{i'\}$ for some $i' \in S$ is still an obstacle between $x_{j_1}$ and $x_{j_2}$, then any solution having $z_{i'k} = 0$ cannot satisfy (14) with equality, since such a solution must have $z_{ik} = 1$ for every $i \in S'$ and $z_{j_1\ell} = z_{j_2\ell} = 1$ in order to attain equality. However, in this setting no hyperplane separating $\{x_{j_1}, x_{j_2}\}$ from $\text{conv}(x_{S'})$ exists, hence such a solution is not feasible. We thus conclude that every solution in the face of $P_T$ induced by (14) satisfies $z_{i'k} = 1$, hence this face is not a facet of $P_T$. ☐

The symmetrical inequalities considering two points $x_{i_1}, x_{i_2} \in X_B$ and a set of points in $X_R$ whose convex hull includes $\text{conv}(\{x_{i_1}, x_{i_2}\})$ have the same properties as (14).

Inequalities (13) and (14) are based on similar ideas, and we can consider a natural generalization of these inequalities in the following way. If $S \subseteq B$
and \( T \subseteq R \) are two sets of points with \( \text{conv}(x_S) \cap \text{conv}(x_T) \neq \emptyset \), then the inequality

\[
\sum_{i \in S} z_{ik} + \sum_{j \in T} z_{j\ell} \leq |S| + |T| - 1
\]

is valid for \( P_T \) for every \( k \in L_B \) and \( \ell \in L_R \). However, this inequality may not be not facet-inducing when \( \text{conv}(x_S) \cap \text{conv}(x_T) \) has nonzero dimension.

### 4 The obstacle graph

If \( V \subseteq R \) is the index set of a subset of points in \( X_R \), we say that \( S \subseteq 2^B \) is a family of obstacles within \( V \) if each \( S \in S \) induces an obstacle \( x_S \) between two points in \( V \). For any set \( V \subseteq R \) and any family \( S \) of obstacles within \( V \), we define \( G(V,S) = (V,E) \) to be the obstacle graph associated with \( V \) and \( S \), where \( E = \{ij : \text{there exists an obstacle between } x_i \text{ and } x_j \text{ induced by a member of } S\} \). In other words, \( G(V,S) \) has an edge for each pair of points in \( V \) having \( x_S \) as an obstacle for some \( S \in S \).

A stable set in a graph \( G = (V,E) \) is a set \( I \subseteq V \) of vertices such that \( ij \notin E \) for every \( i,j \in I \). The maximum cardinality of a stable set in \( G \) is the stability number of \( G \), and is denoted by \( \alpha(G) \).

**Proposition 5.** Let \( V \subseteq R \) and let \( S \) be a family of obstacles within \( V \). Let \( k : S \rightarrow L_B \) be a function associating a group \( k_S \in L_B \) to each obstacle-inducing member of \( S \). For \( \ell \in L_R \), the obstacle rank inequality

\[
\sum_{j \in V} z_{j\ell} + \sum_{S \in S} \sum_{i \in S} z_{ik_S} \leq \alpha(G(V,S)) + \sum_{S \in S} (|S| - \sum_{i \in S} z_{ik_S})
\]

is valid for \( P_T \).

**Proof.** Let \( (p,q,z) \in P_T \) be an integer solution and \( G' = (V,E') \subseteq G(V,S) \) be the subgraph of the obstacle graph defined by \( E' = \{ij : \text{there exists some } S \in S \text{ such that } x_S \text{ is an obstacle between } x_i \text{ and } x_j \text{ and } \sum_{r \in S} z_{rk_S} = |S|\} \). Note that \( ij \in E' \) implies that \( x_i \) and \( x_j \) cannot belong to the same group in \( L_R \) (although the converse implication may not hold, e.g., if the obstacle between them is not minimal), hence the variables \( \{z_{j\ell}\}_{j \in V} \) induce a stable set in \( G' \). Since the deletion of an edge in a graph increases its stability number by at most 1, then

\[
\alpha(G') \leq \alpha(G(V,S)) + |S \setminus E'| \leq \alpha(G(V,S)) + \sum_{S \in S} \left[ |S| - \sum_{i \in S} z_{ik_S} \right].
\]
By combining this assertion with $\sum_{i \in S} z_{ik} \leq \alpha(G')$, we conclude that $(p, q, z)$ satisfies (15). Since $(p, q, z)$ is an arbitrary solution, then (15) is valid for $P_I$. \hfill \Box

The incidence vector of a stable set $I$ is $x^I \in \{0, 1\}^{|V|}$ defined by $x^I_i = 1$ if and only if $i \in I$. The stable set polytope of a graph $G$ is $STAB(G) = \text{conv}\{x^I : I \text{ is a stable set of } G\}$.

**Theorem 3.** Let $V \subseteq R$ and let $S$ be a nonempty family of obstacles within $V$. If

(A) $\sum_{i \in V} x_i \leq \alpha(G(V, S))$ induces a facet of $STAB(G(V, S))$,

(B) $x_I$ and $S$ are linearly separable, for every maximum stable set $I$ of $G(V, S)$ and every $S \in S$,

(C) there exists a bijection between $E$ and $S$, such that the obstacle associated with $ij \in E$ is an obstacle between $x_i$ and $x_j$,

(D) $k_S \neq k_{S'}$ for every $S, S' \in S$, $S \neq S'$,

(E) every obstacle induced by a member of $S$ is minimal,

(F) for every $j \notin V$, there exists a maximum stable set $I$ of $G(V, S)$ such that $I \cup \{j\}$ is strictly separable from $x_S$ for each $S \in S$, and

(G) for every $i \in B \cup \bigcup_{S \in S} S$ and every $S \in S$, there exists a maximum stable set $I$ of $G(V, S)$ such that $I$ is strictly separable from $x_{S'}$ where $S' := S \cup \{i\}$,

then (15) induces a facet of $P_I$.

**Proof.** Let $F$ be the face of $P_I$ defined by (15), and suppose that $\lambda p + \mu q + \gamma z = \lambda_0$ for every solution $(p, q, z) \in F$. We shall prove that $(\lambda, \mu, \gamma)$ is a multiple of the coefficient vector of (15), thus showing that $F$ is indeed a facet of $P_I$.

If $I \subseteq V$ is a stable set, we define $b_I^j$ to be the solution given by $z_{j\ell} = 1$ for every $j \in I$, $z_{ikS} = 1$ for every $S \in S$ and every $i \in S$, and by setting the remaining $z$-variables to 0. For every $S' \in S$, the hypothesis (B) ensures that there exists a hyperplane strictly separating $\{x_j\}_{j \in I}$ from $\{x_i : i \in S, for every S \in S with k_S = k_{S'}\}$, and we set $(p_{kS\ell}, q_{kS\ell})$ to be such a hyperplane. The solution thus constructed is feasible, and the hypothesis (C) ensures that it satisfies (15) with equality.
Claim 1: $\lambda = 0$ and $\mu = 0$. Let $I \subseteq V$ be a maximum-size stable set of $G(V,S)$. For $k \in L_B$, $\ell \in L_R$, and $a \in [d]$, consider the solution $b_{k\ell a}^I := b^I + \varepsilon e^p_{k\ell a}$, which is feasible since either $k$ and $\ell$ are strictly separated, or else one of these groups is empty. The existence of these solutions implies $\lambda_{k\ell a} = 0$. A similar argument shows $\mu_{k\ell} = 0$. $\Diamond$

Claim 2: $\gamma_{j\ell} = \gamma_{j'\ell}$ for every $j, j' \in V$. The hypothesis $\Box (A)$ ensures that there exist stable sets $I_1, \ldots, I_{|V|}$ of size $\alpha(G(V,S))$ such that $x^{I_1}, \ldots, x^{I_{|V|}}$ are affinely independent. The solutions $\{b_{i}^{I_i}\}_{i=1}^{V}$ are, consequently, affinely independent and only differ in the variables $\{z_{j}\}_{j \in V}$, hence settling the claim. $\Diamond$

Claim 3: $\gamma_{ikS} = \gamma_{j\ell}$ for every $i \in S$, $S \in \mathcal{S}$ and every $j \in V$. Let $I$ be a maximum stable set of $G(V,S)$ and consider the solution $b_{ji}^I$, defined from $b^I$ by setting $z_{ikS} = 0$. Let $j_1$ and $j_2$ be the members of $V$ whose corresponding points are separated by $x_S$. Since $x_S$ is minimal, the points in the group $kS$ do not constitute an obstacle between $x_{j_1}$ and $x_{j_2}$, hence we can set $z_{j_1\ell} = z_{j_2\ell} = 1$ and the solution $b_{ji}^I$ is thus feasible. Since $b_{ji}^I$ satisfies (15) with equality, then the fact that $b^I$ also belongs to $\mathcal{F}$ implies $\gamma_{ikS} = \gamma_{j\ell}$. $\Diamond$

Claim 4: $\gamma_{j\ell'} = 0$ for every $j \in V$ and $\ell' \neq \ell$. Let $I$ be a maximum stable set of $G(V,S)$ and consider the solution $b_{ji}^I$, as defined in Claim 3. The solution $b_{ji}^{I'}$ obtained from $b_{ji}^I$ by setting $z_{j\ell'} = 1$ is feasible since the additional group $\ell'$ is a singleton and a subset of a member of $S$. Hence, the hypothesis $\Box (B)$ implies that $x_j$ can be strictly separated from the groups in $L_B$ so we set the variables $\{(p_{k\ell'k}, q_{k\ell'})\}_{k \in L_B}$ accordingly. Since the $z$-variables of $b_{ji}^I$ and $b_{ji}^{I'}$ only differ in the value of $z_{j\ell'}$, we conclude that $\gamma_{j\ell'} = 0$. $\Diamond$

Claim 5: $\gamma_{rk} = 0$ for every $S \in \mathcal{S}$, $i \in S$, and $k \neq k_S$. Again, let $I$ be a maximum stable set of $G(V,S)$ and consider the solution $b_{ji}^I$ as defined in Claim 3. Similarly to Claim 4, the solution $b_{ji}^{I_{ik}}$ obtained from $b_{ji}^I$ by setting $z_{ik} = 1$ is feasible, implying $\gamma_{ik} = 0$. $\Diamond$

Claim 6: $\gamma_{j\ell} = 0$ for every $j \notin V$. The hypothesis $\Box (F)$ ensures that there exists a maximum stable set $I$ of $G(V,S)$ such that $I \cup \{j\}$ is strictly separable from $x_S$, for every $S \in \mathcal{S}$. Consider the solution $b_{ji}^I$ defined from $b^I$ by setting $z_{j\ell} = 1$, and adjusting the variables
\[(p_{k\ell}, q_{k\ell})\}_{k \in L_B}\] accordingly. This construction, together with \(b^I\), establishes the claim. \hfill \diamondsuit

**Claim 7:** \(\gamma_{j\ell'} = 0\) for every \(j \notin V\) and \(\ell' \neq \ell\). Let \(I\) be a maximum stable set of \(G(V, S)\), and construct the solution \(b^I_{j\ell'}\) from \(b^I\) by setting \(z_{j\ell'} = 1\) and adjusting the variables \((p_{k\ell'}, q_{k\ell'})\)_{k \in L_B} accordingly. The existence of the solutions \(b^I_{j\ell'}\) and \(b^I\) implies, together with Claim 1 and Claim 2, that \(\gamma_{j\ell'} = 0\). \hfill \diamondsuit

**Claim 8:** \(\gamma_{ik} = 0\) for every \(i \notin \bigcup_{S \in \mathcal{S}} S\) and \(k \in L_B\). Let \(I\) be a maximum stable set of \(G(V, S)\) and consider the solution \(b^I_{ik}\) defined by setting \(z_{ik} = 1\) and adjusting the variables \((p_{k\ell'}, q_{k\ell'})\)_{\ell' \in L_B} accordingly. If \(k = k_S\) for some \(S \in \mathcal{S}\), we take \(I\) to be a stable set satisfying the hypothesis \((G)\). The existence of this solution, together with \(b^I\), implies \(\gamma_{ik} = 0\). \hfill \diamondsuit

By combining these claims, we conclude that \((\lambda, \mu, \gamma)\) is a multiple of the coefficient vector of (15), hence this inequality induces a facet of \(P_T\). \hfill \Box

Theorem 3 is a step towards an answer of our initial interest in the polytope \(P_T\), namely the existence of combinatorial structures within the polytope associated with the classification problem addressed in this work. Indeed, every facet of the stable set polytope \(STAB(G)\) gives rise to facet-inducing inequalities of \(P_T\), provided the technical conditions stated in Theorem 3 are satisfied. It is interesting to note that the weighted rank inequalities discussed in [6] can also be adapted in a similar way.

### 5 Concluding remarks

In this work we have started a polyhedral study of a particular 2-class classification problem asking for a classification of points in \(\mathbb{R}^d\) into linearly separable groups, and allowing for outliers not to be classified into any group. Our main interest is exploring the polyhedral structure of an integer programming formulation for this problem, and in this direction we were able to identify facet-inducing inequalities involving the binary variables that specify the points assigned to each group. The valid inequalities presented only involve these binary variables, and hint at combinatorial structures present in the classification problem studied, particularly in \(\text{Proj}_z(P_T)\).

A main result states that every facet-inducing inequality for the stable set polytope can be translated into facets of \(\text{Proj}_z(P_T)\). It would be interesting to explore whether similar ideas, namely considering the groups
as colors, can be applied to adapt valid inequalities from vertex coloring, $k$-partite subgraph, and $k$-partition polytopes to the polytope considered in this work. In this vein, a direction for further investigation is the polyhedral consequences of combining the structure of $\text{Proj}_{z}(P_T)$ with orbitope structure in order to eliminate symmetries [8].

An important aspect of our approach is that the integer programming formulation results from a projection of a mixed integer one including the $p$- and $q$-continuous variables. The particular structure of feasible solutions allows us to search for procedures that derive new inequalities from inequalities that are valid for subsets of points. More specifically, let $B' \subseteq B$ and $R' \subseteq R$, and define $Q_{B'R'} := \{ p \in \mathbb{R}^d, q \in \mathbb{R} : px_i + q \leq -1 \text{ for every } x_i \in B' \text{ and } px_j + q \geq 1 \text{ for every } x_j \in R' \}$. By definition, $Q_{B'R'}$ is a polytope, and let $\alpha^T(p,q) \leq \lambda_0$ be a valid inequality for $Q_{B'R'}$. Hence, for any $k \in L_B$ and $\ell \in L_R$, the inequality

$$\alpha^T(p_{k\ell},q_{k\ell}) \leq \lambda_0$$

is satisfied by the solutions having $z_{ik} = 1$ for every $i \in B'$ and $z_{j\ell} = 1$ for every $j \in R'$. We can now lift these inequalities in order to obtain an inequality valid for $P_T$. A straightforward (although not optimal) lifting is given by

$$\alpha^T(p_{k\ell},q_{k\ell}) \leq \lambda_0 + M' \left( |B'| + |R'| - \sum_{i \in B'} z_{ik} - \sum_{j \in R'} z_{j\ell} \right).$$

It would be interesting to explore whether there exist valid inequalities involving the $p$- and $q$-variables not coming from liftings of inequalities of the form (16) and their projections onto the $z$-variables space.

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