Polyhedral Analysis of a Polytope from a Service Center Location Problem with a Type of Decision-Dependent Customer Demand

Fengqiao Luo

*Northwestern University, Department of Industrial Engineering and Management Science, Evanston, IL 60202

Abstract

The polytope from a service center location problem with a special location-dependent demand has been investigated in this paper. The location-dependent demand is defined based on a maximum-attraction-principle assumption. Along this direction, the intersection polytope of the capacity constraints and demand constraints yields a novel polyhedral structure. The structure is investigated in this paper via a family of valid and facet-defining inequalities that can reflect the change of demand fulfillment in a neighborhood.

Keywords: valid and facet-defining inequalities, service center location, decision-dependent demand, maximum attraction principle.

1. Introduction

We investigate the polytope \( \text{conv}(\chi) \), where \( \chi \) is the set of points satisfying the following linear and mixed-binary constraints:

\[
\chi = \left\{ [x, y] \mid \sum_{i=1}^{m} x_{ij} \leq c_j y_j \quad \forall j \in [n], \right. \\
\left. \sum_{j=1}^{n} x_{ij} \leq \max_{j \in [n]} d_{ij} y_j \quad \forall i \in [m], \\
x_{ij} \geq 0, y_j \in \{0, 1\} \quad \forall i \in [m], \forall j \in [n] \right\},
\]

(1)
where \([m]\) represents the set of indices \(\{1, \ldots, m\}\). Note that \(\text{conv}(\chi)\) is indeed a polytope, which is shown in Proposition [4].

1.1. Motivation and background

The above linear system is motivated from a service center location problem with standard capacity constraints and additional decision-dependent constraints. More specifically, we consider a problem of allocating a set of service centers (facilities) in a region to meet the customer demand from the region. Similarly to the traditional facility location problem, the region consists of customer sites and candidate locations for the service centers. Each customer site has certain demand that needs to be fulfilled by a service center from its neighborhood. Especially, we consider the case that the demand of each customer site is a value which depends on the location pattern of service centers. Suppose there are \(m\) customer sites and \(n\) candidate locations for the service centers. Let \(c_j\) be the capacity of the service center at the location \(j \in [n]\). Let \(d_i(y)\) be the demand from customer site \(i\) for a given service-center-location vector \(y\). In particular, we consider a special type of dependency between the customer demand and the location vector, which is simple but sensible in depicting customers’ behavior in certain situations. We referred it as the maximum attraction principle. To explain this principle, we assume that when there is just one service center been allocated which is opened at \(j \in [n]\), it will attract \(d_{ij}\) amount of customer demand from the site \(i\). If there are multiple service centers opened at different locations, the maximum demand that can be attracted from \(i\) is equal to the maximum \(d_{ij}\) for \(j \in [n]\) that has an opened service center. This principle is formally described by the following equation:

\[
d_i(y) = \begin{cases} 
0 & \text{if } y_j = 0 \text{ for all } j \in [n], \\
d_{ij} & \text{if } y = e_j \text{ for some } j \in [n], \\
\max_{j \in [n]} d_{ij}y_j & \text{for other cases,}
\end{cases}
\]

where \(e_j\) is the \(n\)-dimensional vector with the \(j\)-th entry being 1 and other entries being 0.
The maximum attraction principle matches with our intuition from practice in certain situations. Note that the principle has incorporated the case that customers in site \( i \) are not willing to visit any service centers that are not within a certain range from \( i \). In this case, assume \( F_i \subseteq [n] \) is the subset of candidate locations that some customers from site \( i \) are willing to visit, and then we can set \( d_{ij} = 0 \) for every \( j \in [n] \setminus F_i \). In particular, \( F_i \) can be a subset of candidate locations in a neighborhood of site \( i \), as customers usually will not consider visiting a service center that is beyond a certain distance from their living place. For a given location vector \( y \), let \( F_i(y) = \{ j \in F_i \mid y_j = 1 \} \).

The maximum attraction principle further assumes that within the preferable opened service centers \( F_i(y) \), there exists one service center \( j^* \) (or multiple centers) that is (are) most attractive to the customers from site \( i \) in the sense that \( j^* \in \arg\max_{j \in F_i(y)} d_{ij} \), and the presence of multiple service centers in \( F_i(y) \) including \( j^* \) attracts the same amount of demand from site \( i \) as the presence of just a single service center at \( j^* \). This assumption is also consistent with our intuition in some situations. For example, if all service centers are identical in scale and service quality, the most attractive one to site \( i \) is likely the one that is closest to \( i \). Furthermore, customers who are willing to visit service centers with a longer distance are also willing to visit the one that is closest to their living place.

Notice that the polyhedral properties of the capacity constraints \( \sum_{i=1}^{m} x_{ij} \leq c_j y_j \) in (1) have been well studied in [1]. However, the demand attraction constraints \( \sum_{j=1}^{n} x_{ij} \leq \max_{j \in [n]} d_{ij} y_j \) concerned in this research are novel which brings new features to the polyhedral structure when interacting with the capacity constraints. We focus on investigating the polyhedral structure induced by the demand attraction constraints and the capacity constraints.

1.2. Literature review

The facility location problem [2] is one of the most fundamental problems investigated in operations research. In this problem, a decision maker needs to decide locations of a limited number of facilities (factories, retail centers,
power plants, service centers, etc.), and determine coverage of demand from different sites by the located facilities. The objective is to minimize the facility setup cost and the cost of production and delivery. This problem provides a basic framework to formulate related problems in resource allocation [3], supply chain management [4] and logistics [5, 6], etc. See [7] for a comprehensive review on this topic.

Facility location problems with non-deterministic demand have received plentiful investigation in the framework of stochastic optimization and robust optimization. In the stochastic optimization framework, customer demand are independent random parameters with known probability distributions and the problem can be formulated as a two-stage stochastic program in which the location vector is in the first-stage decision that needs to be made before realization of customer demand [8, 9]. In a special case when the customer demand rates and service rates of each facility are assumed to follow exponential probability distributions, the problem of minimizing long term average cost can be reformulated as a deterministic optimization problem with corresponding Poisson rates to characterize the demand and service. This setting has been applied to an allocation problem of ATMs [10]. The probability distribution of demand can also be used to define chance constraints to ensure a required service level under possible stockout and supply disruption [11, 12, 13, 14]. In the robust optimization framework for facility location problems, the information of demand is partially known to the decision maker, and the goal is to find a robust optimal location vector that optimizes the objective after a worst-case realization of demand information [15]. Following this direction of research, Baron and Milner [10] investigated a robust multi-period facility location problem with box and ellipsoid sets of uncertainty. Gourtani et. al. [17] investigated a distributionally-robust two-stage facility location problem with an ambiguity set defined corresponding to the mean and covariance matrix of a random parameter for expressing the demand. Facility location problems with demand uncertainty have been studied with a variety of novel application background, which includes but not limited to medication coverage and delivery under a large-scale bio-terror attack [11].
medical equipment (defibrillators) location problem to reduce cardiopulmonary resuscitation (CPR) risk [18], humanitarian relief logistics [19], and hazardous waste transportation [20], etc.

In a lot of real world problems, the parameters or the uncertainty of them can interplay with the decision to be made. This behavior is well observed especially in a sequential (multi-stage) decision-making process, in which information about system parameters are gradually revealed and the decisions made up until the current stage can reshape the uncertainty in future [21]. The customer demand could also depend on the location of facilities or service centers as revealed in [22, 23], which motivates the study of the facility location problem with location-dependent demand. As a robustification of it, Basciftci et al. [24] investigated a distributionally-robust facility location problem with decision-dependent ambiguity, where the ambiguity set is defined using disjoined lower and upper bounds on the mean and variance of the candidate probability distributions for customer demand, and these bounds are defined as linear functions of the location vector to admit a tractable MILP reformulation.

Identifying valid and facet-defining inequalities is an essential topic in understanding the polyhedral structure of a mixed-integer set of points [25, 26]. As a by-product, adding some of these inequalities may improve the computational performance of solving the corresponding mixed-integer linear program, as a result of strengthening the formulation [27, 28, 29, 30, 31, 32]. Some standard techniques are well developed to construct the face- and facet-defining inequalities such as lifting and simultaneous lifting [33, 34, 35, 36, 37, 38], lift-and-project [39, 40, 41, 42, 43, 44], and Fourier-Motzkin elimination [45, 46, 47], etc. By applying these techniques, different families of inequalities such as lifted cover inequalities [38, 36], lifted flow-cover inequalities [37], and disjunctive inequalities [48], etc. are constructed for corresponding basic polytopes that are often building blocks in the formulation of various mixed-integer linear programming problems in practice. In many mixed-integer linear programming problems from the application, construction of valid and facet-defining inequalities usually depends on ad hoc observations that identify sources leading to fractional solutions.
The key observation for constructing facet-defining inequalities in this paper is inspired by [49]. The work in [49] has constructed and generalized a family of facet-defining inequalities for the polytope of a standard capacitated facility location problem. Computational performance with the inequalities added to the linear relaxation problems have been investigated in [50, 51]. In the standard capacitated facility location problem considered in [49], the demand of each client site is given as a fixed constant, whereas the customer demand is considered to be decision-dependent in this work.

2. Polyhedral properties of $\text{conv}(\chi)$

We first show in the following proposition that the convex hull $\text{conv}(\chi)$ is indeed polytope.

**Proposition 1.** The convex hull $\text{conv}(\chi)$ is a polytope.

**Proof.** Let $\chi(\hat{y})$ be the subset of $\chi$ with a given binary vector $y$ at value $\hat{y}$. Once the variable $y$ is fixed, the right side of the inequalities $\sum_{i=1}^{m} x_{ij} \leq c_j \hat{y}_j$ and $\sum_{j=1}^{n} x_{ij} \leq \max_{j \in [n]} d_{ij} \hat{y}_j$ becomes a constant, and hence $\chi(\hat{y})$ is a polytope. Notice that the convex hull of $\chi$ can be rewritten as

$$\text{conv}(\chi) = \text{conv} \left( \bigcup_{\hat{y} \in \{0,1\}^n} \chi(\hat{y}) \right).$$

(3)

The right side of (3) is a convex hull of a finite union of polytopes, and hence the result is also a polytope. \qed

Notice that the representation (1) of $\chi$ involves a nonlinear term $\max_{j \in [m]} d_{ij} \hat{y}_j$ of the decision vector $y$, which is inconvenient for the polyhedral analysis. The nonlinear term can be avoided by reformulating $\chi$ in a lifted space. The lifted
reformulation of $\chi$ is given as follows:

$$
\chi^L = \left\{ [x, y, q] \mid \begin{array}{l}
\sum_{i=1}^{m} x_{ij} \leq c_j y_j \quad \forall j \in [n], \\
\sum_{j=1}^{n} x_{ij} \leq \sum_{j \in [m]} d_{ij} q_{ij} \quad \forall i \in [m], \\
\sum_{j=1}^{n} q_{ij} \leq 1 \quad \forall i \in [n], \\
q_{ij} \leq y_j \quad \forall i \in [n], \forall j \in [m], \\
x_{ij} \geq 0, q_{ij} \geq 0, y_j \in \{0, 1\} \quad \forall i \in [m], \forall j \in [n]
\end{array} \right\}.
$$

(4)

**Proposition 2.** The polytope $\chi$ is the projection of $\chi^L$ on to the $[x, y]$ space, i.e., $\text{conv}(\chi) = \text{proj}_{[x,y]}\text{conv}(\chi^L)$.

**Proof.** Notice that both $\chi$ and $\chi^L$ can be represented as the convex hull of a finite number of sub-polytopes as follows:

$$
\text{conv}(\chi) = \text{conv} \left( \bigcup_{q \in \{0,1\}^n} \chi(q) \right),
$$

(5)

$$
\text{conv}(\chi^L) = \text{conv} \left( \bigcup_{q \in \{0,1\}^n} \chi^L(q) \right).
$$

So it suffices to show that $\chi(q) = \text{proj}_x \chi^L(q)$. First, consider any $[\hat{x}, \hat{y}, \hat{q}] \in \chi^L(q)$. To show $[\hat{x}, \hat{y}] \in \chi(q)$, it suffices to verify that $\sum_{j=1}^{n} \hat{x}_{ij} \leq \max_{j \in [n]} d_{ij} \hat{y}_j$ holds for every $i \in [n]$. Indeed, if we let $J_i = \{ j \in [m] \mid \hat{y}_j = 1 \}$, we have

$$
\sum_{j=1}^{n} \hat{x}_{ij} \leq \sum_{j \in [n]} d_{ij} \hat{q}_{ij} = \sum_{j \in J_i} d_{ij} \hat{q}_{ij} + \sum_{j \in [m] \setminus J_i} d_{ij} \hat{q}_{ij} \\
\leq \sum_{j \in J_i} d_{ij} \hat{q}_{ij} + \sum_{j \in [m] \setminus J_i} d_{ij} \hat{y}_j = \sum_{j \in J_i} d_{ij} \hat{q}_{ij} \\
\leq \left( \max_{j \in J_i} d_{ij} \right) \sum_{k \in J_i} \hat{q}_{ik} \leq \max_{j \in J_i} d_{ij} \hat{y}_j = \max_{j \in [n]} d_{ij} \hat{y}_j.
$$

(6)

Therefore, we have shown that $\text{proj}_x \chi^L(q) \subseteq \chi(q)$. Second, consider any $[\hat{x}, \hat{y}] \in \chi(q)$. We need to show that there exists a $\hat{q}$ such that $[\hat{x}, \hat{y}, \hat{q}] \in \chi^L(q)$. We can construct a $\hat{q}$ and the as follows: Let $k_i \in \text{argmax}_{j \in J_i} d_{ij}$. For any $i \in [m]$, let $\hat{q}_{ik_i} = 1$ and $\hat{q}_{ij} = 0$ for all $j \in [m] \setminus \{k_i\}$. Clearly the constraints $\sum_{j=1}^{n} \hat{q}_{ij} \leq 1$ and $\hat{q}_{ij} \leq \hat{y}_j$ are satisfied for each $i \in [m]$ by the way of constructing $\hat{q}$. 

7
Furthermore, we have

\[ \sum_{j \in [n]} d_{ij} \hat{q}_{ij} = d_{i,k_i} = \max_{j \in J_i} d_{ij} = \max_{j \in [n]} d_{ij} \hat{q}_{ij} \geq \sum_{j=1}^n \hat{x}_{ij}, \]  

which shows that the constraint \( \sum_{j=1}^n \hat{x}_{ij} \leq \sum_{j \in [n]} d_{ij} \hat{q}_{ij} \) is satisfied, and hence \( \chi(\hat{y}) \subseteq \text{proj}_x \chi^L(\hat{y}) \).

Proposition 2 has revealed a type of equivalence between \( \chi \) and \( \chi^L \). We will hence focus on deriving valid and facet-defining inequalities for \( \chi^L \). As we will see later, these inequalities do not involve the variables \( q \), and hence they are also valid for \( \chi \).

To give an insight on how the valid and facet-defining inequalities are constructed, we consider the quantity \( \sum_{i \in I} \sum_{j \in J} x_{ij} \) for any subsets \( I \subseteq [m] \) and \( J \subseteq [n] \). The capacity and demand attraction constraints imply that \( \sum_{i \in I} \sum_{j \in J} x_{ij} \leq \sum_{j \in J} c_j y_j \) and \( \sum_{i \in I} \sum_{j \in J} x_{ij} \leq \sum_{i \in I} \max_{j \in [n]} d_{ij} y_j \). The upper bound of the quantity \( \sum_{i \in I} \sum_{j \in J} x_{ij} \) depends on which constraint is more restrictive. Figure 1 illustrates a critical situation that if any one of the service centers in \( J \) is not opened, the total attracted demand is greater than the total capacity. While if all service centers in \( J \) are opened, the total attracted demand is less than the total capacity. Based on this critical situation, we can derive a facet-defining valid inequality \( \text{conv}(\chi^L) \) under some mild conditions. The theoretical results on the polyhedral properties of \( \text{conv}(\chi^L) \) are given by Theorem 2 and Theorem 3.

2.1. Technical lemmas for proving facetness

We provide a sequence of technical lemmas that will be used to prove the facet-extension results given in Theorem 1.

**Lemma 1.** Let \( n \) be a positive integer. Then there exist \( n \) linearly independent vectors in any \( n + 1 \) affinely independent vectors.

**Proof.** Suppose \( S = \{v^k\}_{k=0}^n \) are \( n+1 \) affinely independent vectors. If \( \text{span}(S) \geq n \), then there must exist \( n \) linearly independent vectors in \( S \). We now suppose that \( \text{span}(S) \leq n - 1 \). Specifically, suppose \( \text{span}(S) = m \), with \( m \leq \)
Figure 1: An illustration of the competition between the capacity constraints and the demand attraction constraints. In all subfigures, three customer sites are represented as circles which are denoted as S1, S2, S3 (from top to bottom) and two candidate service center locations are represented as squares which are denoted as F1, F2 (from top to bottom). The number in a customer site (circle) indicates the attracted demand, and the number in a service center (square) indicates the capacity. In case (a) only F1 is opened which attracts 2 units of demand from S1 and S2 respectively and 1 unit from S3. In this case the total attracted demand is greater than the total capacity. In case (b) only F2 is opened and the situation is similar to (a). In case (c) two service centers are opened, which attracts 2 units of demand from all customer sites. The total attracted demand is less than the total capacity.

$n - 1$. Then there exist $m$ linearly independent vectors in $S$. We assume that $\{v^k\}_{k=n-m+1}^n$ are linearly independent without loss of generality. It follows that $v^i \in \text{span}\{v^k\}_{k=n-m+1}^n$, $\forall i \in \{0, 1, \ldots, n - m\}$. Therefore, we have $\text{span}\{v^k - v^0\}_{k=1}^n \subseteq \text{span}\{v^k\}_{k=n-m+1}^n$, and hence $\dim\{v^k - v^0\}_{k=1}^n \leq m$, which contradicts to that $\{v^k - v^0\}_{k=1}^n$ are linearly independent.

Lemma 2. Let $\{v_i\}_{i=1}^k$ be affinely independent vectors in $\mathbb{R}^n$ and $\{u_i\}_{i=1}^m$ be linearly independent vectors in $\mathbb{R}^n$. If $\text{span}\{u_i\}_{i=1}^m \cap \text{span}\{v_i\}_{i=1}^k = \{0\}$, then $\{v_i\}_{i=1}^k \cup \{u_i\}_{i=1}^m$ are affinely independent vectors in $\mathbb{R}^n$.

Proof. Consider any coefficients $\{a_i\}_{i=2}^k$ and $\{b_i\}_{i=1}^m$ that satisfy the following equation

$$\sum_{i=2}^k a_i(v_i - v_1) + \sum_{j=1}^m b_j(u_j - v_1) = 0. \quad (8)$$

If $b_j = 0 \ \forall j \in [m]$, then (8) implies $\sum_{i=2}^k a_i(v_i - v_1) = 0$, and hence $a_i = 0 \ \forall i \in$
\{2, \ldots, k\} due to the affinely independence of \(\{v_i\}_{i=1}^k\). Now let \(J = \{j \in [m]: b_j \neq 0\}\) and assume \(|J| \geq 1\), then (8) becomes

\[
\sum_{j \in J} b_j u_j = \sum_{j \in J} b_j v_1 - \sum_{i=2}^k a_i (v_i - v_1).
\] (9)

By linearly independence, we have \(\sum_{j \in J} b_j u_j \neq 0\). Since \(\text{span}\{u_i\}_{i=1}^m \cap \text{span}\{v_i\}_{i=1}^k = \{0\}\), we must have \(\sum_{j \in J} b_j u_j = 0\), which leads to a contradiction. Therefore, we have proved that \(a_i = 0\) \(\forall i \in \{2, \ldots, k\}\) and \(b_j = 0\) \(\forall j \in [m]\), which shows that \(\{v_i\}_{i=1}^k \cup \{u_i\}_{i=1}^m\) are affinely independent vectors.

**Lemma 3.** Let \(l, m, n\) be three non-negative integers such that \(m \geq \max\{n, l\}\) and \(n > 0\). Let \(\{v_k\}_{k=1}^n\), \(\{v'_k\}_{k=1}^l\) and \(\{u_k\}_{k=1}^m\) be three sets of vectors in \(\mathbb{R}^n\). The following properties hold.

(a) If \(\{v_k\}_{k=1}^n\) and \(\{v'_k\}_{k=1}^l\) are \(n + l\) linearly independent vectors, then for any \(r > 0\) there exists an \(0 < \epsilon < r\) such that \(\{u_k + \epsilon v_k\}_{k=1}^n\) and \(\{v'_k\}_{k=1}^l\) are also \(n + l\) linearly independent vectors.

(b) If \(\{v_k\}_{k=1}^n\) are \(n\) affinely independent vectors, \(\{v'_k\}_{k=1}^l\) are \(l\) linearly independent vectors, \(\text{span}\{v'_k: \forall k \in [l]\} \subseteq \text{span}\{u_k: \forall k \in [n]\}\) and \(\text{span}\{v_k: \forall k \in [n]\} \cap \text{span}\{u_k: \forall k \in [n]\} = \{0\}\), then for any \(r > 0\) there exists an \(0 < \epsilon < r\) such that \(\{u_k + \epsilon v_k\}_{k=1}^n\) and \(\{v'_k\}_{k=1}^l\) are \(n + l\) affinely independent vectors.

**Proof.** (a) Note that the set \(\{v_k\}_{k=1}^n \cup \{v'_k\}_{k=1}^l\) contains \(n + l\) linearly independent vectors if and only if \(\det A^T A \neq 0\), where \(A = [v_1, \ldots, v_n, v'_1, \ldots, v'_l]\). It amounts to show that there exists an \(0 < \epsilon < r\) such that \(\det B^T(\epsilon)B(\epsilon) \neq 0\), where \(B(\epsilon) = [u_1 + \epsilon v_1, \ldots, u_n + \epsilon v_n, v'_1, \ldots, v'_l]\). We prove it by contradiction. Assume that \(\det B^T(\epsilon)B(\epsilon) = 0\) for all \(0 < \epsilon < r\). Note that since \(\det B^T(\epsilon)B(\epsilon)\) is a polynomial of \(\epsilon\) on \(\mathbb{R}\), this hypothesis implies that the polynomial \(\det B^T(\epsilon)B(\epsilon)\) has infinite number of real roots. However, only zero polynomial has infinitely many real roots. If we can show that \(\det B^T(\epsilon)B(\epsilon)\) is not a zero polynomial, we deduce a contradiction. The polynomial \(\det B^T(x)B(x)\) can be rewritten as
\[
\det B^\top(x)B(x) = \det \left( [u_1^\top + xv_1^\top, \ldots, u_n^\top + xv_n^\top, v_1'^\top, \ldots, v_l'^\top] \right)^	op
\]
\[
= \det A^\top A x^{2n} + f(x),
\]
where \( f(x) \) is a polynomial of \( x \) with degree at most \( 2n - 1 \). Therefore we have
\[
\lim_{x \to \infty} \frac{\det B^\top(x)B(x)}{x^{2n}} = \det A^\top A \neq 0,
\]
which shows that \( \det B^\top(x)B(x) \) is not a zero polynomial.

(b) We first make the following claim. **Claim:** For any \( r > 0 \) there exists an \( 0 < \epsilon < r \) such that the set \( \{ u_k + \epsilon v_k \}_{k=1}^n \) contains \( n \) affinely independent vectors. **Proof of Claim:** Note that the vectors \( \{ v_k \}_{k=1}^n \) are affinely independent if and only if \( \{ [1, v_k^\top] \}_{k=1}^n \) are linearly independent. Since by assumption we have \( \text{span}\{ v_k : \forall k \in [n] \} \cap \text{span}\{ u_k : \forall k \in [n] \} = \{0\} \), it amounts to show that there exists an \( 0 < \epsilon < r \) such that \( \{ [1, u_k^\top + \epsilon v_k^\top] \}_{k=1}^n \) are linearly independent, which is equivalent to that \( \{ [\epsilon, u_k^\top + \epsilon v_k^\top] \}_{k=1}^n \) are linearly independent. Since we have \( [\epsilon, u_k^\top + \epsilon v_k^\top] = [0, u_k^\top] + \epsilon[1, v_k^\top] \), we can apply the result of Part (a) to conclude the proof of the claim.

Suppose \( \{ u_k + \epsilon v_k \}_{k=1}^n \) are \( n \) affinely independent vectors. We now show that \( \text{span}\{ u_k + \epsilon v_k : \forall k \in [n] \} \cap \text{span}\{ v'_k : \forall k \in [l] \} = \{0\} \). Then by Lemma 2, we can deduce that \( \{ u_k + \epsilon v_k \}_{k=1}^n \) and \( \{ v'_k \}_{k=1}^l \) are \( n + l \) affinely independent vectors. Assume that \( \text{span}\{ u_k + \epsilon v_k : \forall k \in [n] \} \cap \text{span}\{ v'_k : \forall k \in [l] \} \neq \{0\} \). Then there exist coefficients \( \{ \alpha_k \}_{k=1}^n \) and \( \{ \beta_k \}_{k=1}^l \) which are not all zero, such that
\[
\sum_{k=1}^n \alpha_k (u_k + \epsilon v_k) = \sum_{k=1}^l \beta_k v'_k,
\]
which implies that
\[
\epsilon \sum_{k=1}^n \alpha_k v_k = \sum_{k=1}^l \beta_k v'_k - \sum_{k=1}^n \alpha_k u_k.
\]
Since \( \sum_{k=1}^n \alpha_k v_k \in \text{span}\{ v_k : \forall k \in [n] \} \) and \( \sum_{k=1}^l \beta_k v'_k = \sum_{k=1}^n \alpha_k u_k \in \text{span}\{ u_k : \forall k \in [n] \} \), but span\{ \epsilon v_k : \forall k \in [n] \} \cap \text{span}\{ v'_k : \forall k \in [l] \} = \{0\} \) holds by the given hypothesis. It follows that \( \sum_{k=1}^n \alpha_k v_k = 0 \) and hence \( \alpha_k = 0 \ \forall k \in [n] \) and \( \beta_k = 0 \ \forall k \in [l] \), which leads to a contradiction. \( \square \)

**Lemma 4.** (a) Let \( n \) be a positive integer. The following set of vectors are affinely independent: \( U = \{ e^n_i - e^n_{i+1} : \forall i \in [n] \} \) with the convention \( n + 1 \to 1 \),
where $e_i^n$ is a $n$-dimensional vector with the $i$-th entry being 1 and other entries being 0.

(b) Let $\{n_i\}_{i=1}^k$ be $k$ positive integers with $n_i \geq 2 \ \forall i \in [k]$. Consider the following sets of vectors in $\mathbb{R}^N$ with $N = \sum_{i=1}^k n_i$:

$$U_l = \{0^{n_1}, \ldots, 0^{n_l-1}, e_i^{n_l}, 0^{n_l+1}, \ldots, 0^{n_k} : \forall i \in [n_l - 1] \} \ \forall l \in [k],$$

$$V = \{0^{n_1}, \ldots, 0^{n_l-1}, e_i^{n_l}, -e_i^{n_l+1}, 0^{n_l+2}, \ldots, 0^{n_k} : \forall i \in [k] \},$$

with the convention $k+1 \rightarrow 1$. The $N$ vectors in $(\bigcup_{l=1}^k U_l) \cup V$ are affinely independent.

Proof. Part (a) can be easily verified. We now focus on proving Part (b). Without loss of generality, we prove for the case that $n_l \geq 4$ for all $l \in [k]$. It is easy to verify that all vectors in $\bigcup_{l=1}^k U_l$ are linearly independent, and by Part (a), the vectors in $V$ are affinely independent. We claim that $\text{span}(V) \cap \text{span}(\bigcup_{l=1}^k U_l) = \{0^N\}$. If the claim holds then by Lemma 2, the $N$ vectors in $(\bigcup_{l=1}^k U_l) \cup V$ are affinely independent. To prove this claim, we label the vectors in $U_l$ as $u_i$ for $l \in [k]$ and $i \in [n_l - 1]$, and label the vectors in $V$ as $v_l$ for $l \in [k]$. We can verify that $\{v_l\}_{l=1}^{k-1}$ are linearly independent and $v_k = -\sum_{l=1}^{k-1} v_l$. Suppose $\text{span}(V) \cap \text{span}(\bigcup_{l=1}^k U_l) \neq \{0^N\}$ and let $\{\alpha_l | l \in [k], i \in [n_l - 1]\}$ and $\{\beta_l | l \in [k-1]\}$ be the coefficients such that

$$w = \sum_{l \in [k]} \sum_{i \in [n_l-1]} \alpha_l u_i = \sum_{l \in [k-1]} \beta_l v_l \neq 0^N. \quad (11)$$

Consider the $e_i^{n_s}$ entry of $w$ for $s \in [k]$ and $i \in \{2, \ldots, n_l - 1\}$. Evaluating this coefficient based on $\sum_{l \in [k-1]} \beta_l v_l$ implies that the coefficient should be 0. On the other side, evaluating this coefficient based on $\sum_{l \in [k]} \sum_{i \in [n_l-1]} \alpha_l u_i$ implies that it should be equal to $\alpha_{s,i} - \alpha_{s,i-1}$. Therefore, we must have $\alpha_{s,i} = \alpha_{s,i-1}$ for $s \in [k]$ and $i \in \{2, \ldots, n_l - 1\}$. For $l \in [k]$, let $a_l = \alpha_{l,i}$ for $i \in \{2, \ldots, n_l - 1\}$. By matching the coefficient corresponding to the $e_i^{n_s}$ entry of $w$, we get $a_1 = \beta_1$. By matching the coefficient corresponding to the $e_i^{n_l}$ entry of $w$, we get $a_l = 0$ for $l \in [k]$. This shows that $w = 0$, which leads to a contradiction. \qed
Lemma 5. Let $l, m, n$ be three positive integers. Let $u^k \in \mathbb{R}^l \otimes \mathbb{R}^n$ and $v^k \in \mathbb{R}^m \otimes \mathbb{R}^n$ for all $k \in [n]$ be arbitrary vectors with dimension $l \times n$ and $m \times n$, respectively. Let $u' \in \mathbb{R}^l \otimes \mathbb{R}^n$, and $v' \in \mathbb{R}^m \otimes \mathbb{R}^n$ be two arbitrary vectors. Let \( \{w^i\}_{i=1}^n \) be $n$ linearly independent vectors in $\mathbb{R}^n$ with $w^i_j \in \{0,1\}$ \( \forall i, j \in [n] \). Let $J_i = \{ j \in [n] : w^i_j = 1 \}$. Consider the following sets of vectors in the $\mathbb{R}^{l \times n} \oplus \mathbb{R}^{m \times n} \oplus \mathbb{R}^n$ space:

$$U^k = \{ [u^i, v^i, w^i] : \forall k \in [n], \forall i, i' \neq i', \forall j \in J_k \}$$

$$V^k = \{ [u^i, v^i, w^i] : \forall k \in [n], \forall j \in J_k \}$$

$$Y = \{ [u^i, v^i, 1^m] : \forall k \in [n] \}$$

where $1^m$ is the $n$-dimensional vector with all entries being 1, $[e_{ij}, 0, 0] := [e_{ij}^l \otimes e_j^m, 0^m \otimes 0^m]$ and $[0, e_{ij}, 0] := [0^l \otimes e_i^m, e_i^m \otimes 0^m]$. Then the set $S = \bigcup_{k=1}^n (U^k \cup V^k) \cup Y \cup W$ contains $(l + m + 1)n$ affinely independent vectors.

Proof. Since \( \{w^i\}_{i=1}^n \) are $n$ linearly independent binary vectors in $\mathbb{R}^n$, for each $j \in [n]$, there exists an $i \in [n]$ (depending on $j$) such that $j \in J_i$. Therefore, there exists a mapping $\pi : [n] \to [n]$ such that $\pi$ is surjective and $j \in J_{\pi(j)}$. We define the following sets of vectors:

$$U^j = \{ [u^i, v^i, w^i] : \forall j \in [n], \forall i \in [l-1] \}$$

$$V^j = \{ [u^i, v^i, w^i] : \forall j \in [n], \forall i \in [m] \}$$

Clearly, the sets \( \{U^j : \forall j \in [n]\} \) are disjoint and $\cup_{j \in [n]} U^j \subseteq \cup_{k \in [n]} U^k$. Similarly, the sets \( \{V^j : \forall j \in [n]\} \) are disjoint and $\cup_{j \in [n]} V^j \subseteq \cup_{k \in [n]} V^k$. We claim that all vectors in the following sets are affinely independent:

$$U'' = \{ [e_{ij}, 0, 0] - [e_{i+1,j}, 0, 0] : \forall i \in [l-1], \forall j \in [n] \},$$

$$V'' = \{ [0, e_{ij}, 0] : \forall i \in [m], \forall j \in [n] \},$$

$$Y'' = \{ [e_{ij}, 0, 0] - [e_{1,j+1}, 0, 0] : \forall j \in [n] \}.$$
The claim can be proved using Lemma 4(b) and Lemma 2 with following argument. By Lemma 4(b), all vectors in $U'' \cup Y''$ are affinely independent. One can easily verify that all vectors in $V''$ are independent and span $\{U'' \cup Y''\} \cap \text{span}\{V''\} = \{(0,0,0)\}$, and hence by Lemma 2 all vectors in $U'' \cup V'' \cup Y''$ are affinely independent. Then by Lemma 3(b), the following $(l + m + 1)n$ vectors are affinely independent:

$$
U^* = \{[a^{\pi(j)},v^{\pi(j)},w^{\pi(j)}] + [\epsilon e_{i,j},0,0] - [\epsilon e_{i+1,j},0,0] : \forall i \in [l-1], \forall j \in [n]\},
$$

$$
V^* = \{[a^{\pi(j)},v^{\pi(j)},w^{\pi(j)}] + [0,\epsilon e_{i,j},0] : \forall i \in [m], \forall j \in [n]\},
$$

$$
Y^* = \{[u',v',1^n] + [\epsilon e_{1,j},0,0] - [\epsilon e_{1,j+1},0,0] : \forall j \in [n]\},
$$

$$
W = \{[u^k,v^k,w^k] : \forall k \in [n]\}.
$$

Since we have $U^* \cup V^* \cup Y^* \cup W \subseteq S$, it implies that $S$ should contain $(l + m + 1)n$ affinely independent vectors.

2.2. Valid and facet-defining inequalities

Our goal is to construct valid and facet-defining inequalities for the convex hull $C(I,J)$ induced by the subsets $I, J$ of indices for the customer sites and candidate locations, and then show that some of them are also valid and facet-defining inequalities for $\text{conv}(\chi L)$ and $\text{conv}(\chi)$:

$$
C(I,J) = \{[x,y,q] \in \text{conv}(\chi L) \mid y_j = 0 \ \forall j \in [n] \setminus J, x_{ij} = 0, q_{ij} = 0 \ \forall (i,j) \notin I \times J\}. 
$$

(12)

The conversion from the facets of $C(I,J)$ to the facets of $\text{conv}(\chi L)$ is depicted by the following theorem:

**Theorem 1** (Facet extension). Consider the polytope $\text{conv}(\chi L)$ defined in (4) induced by the index subsets $I \subseteq [m]$ and $J \subseteq [n]$. Suppose $\text{conv}(\chi L)$ is in full dimension, i.e., $\text{dim} \text{conv}(\chi L) = (2m+1)n$, and $d_{ij} > 0$ for $i \in [m]$, $j \in [n]$. Suppose an inequality

$$
\sum_{j \in J} \alpha_{j}y_{j} + \sum_{i \in I} \sum_{j \in J} (x_{ij} + \beta_{ij}q_{ij}) \leq \gamma
$$

(13)
with coefficients $\alpha_j$, $\beta_{ij}$ and $\gamma$ define a facet of $C(I,J)$ and it is valid for $\text{conv}(\chi^L)$. If there exists a set $V$ of $(2|I|+1)|J|$ affinely independent points in $C(I,J)$ that are on the plane defined by this inequality and $V$ satisfies the following conditions:

1. For any point $[x,y,q] \in V$, it holds that $y_j \in \{0,1\}$ $\forall j \in J$;

2. $\forall j \in J$ there exists an point $[x^{(j)},y^{(j)},q^{(j)}] \in V$ such that $y_j^{(j)} = 1$ and $\sum_{i \in I} x_{ij}^{(j)} < c_j$;

3. For each $i \in I$, there exists a point $[\tilde{x}^{(i)},\tilde{y}^{(i)},\tilde{q}^{(i)}] \in V$ such that $\sum_{j \in J} q_{ij}^{(i)} < 1$.

Then the inequality (13) defines a facet of $\text{conv}(\chi^L)$.

Proof. We first show that $\text{conv}(\chi^L)$ is of full dimension. Since $\text{conv}(\chi^L)$ is defined by $(2m + 1)n$ variables, we will prove $\dim \text{conv}(\chi^L) = (2m + 1)n$ by constructing $(2m + 1)n + 1$ affinely independent points in $\text{conv}(\chi^L)$ as follows:

$$[0,0,0], \ [0,e_j,0] \ \forall j \in F, \ [0,e_j,e_{ij}] \ \forall i \in S \ \forall j \in F, \ [e_{ij},e_j,e_{ij}] \ \forall i \in S \ \forall j \in F,$$

where $\epsilon$ is a sufficient small positive constant. In particular, $\epsilon < \min\{d_{ij},c_j\}$. It is easy to verify the above $(2m + 1)n + 1$ points are affinely independent points in $\text{conv}(\chi^L)$, and hence $\dim \text{conv}(\chi^L) = (2m + 1)n$. Using a similar argument, we can show that $\dim C(I,J) = (2|I|+1)|J|$. We now focus on proving the theorem.

We refer the inequality $\sum_{j \in J} \alpha_j y_j + \sum_{i \in I} \sum_{j \in J} (x_{ij} + \beta_{ij} q_{ij}) \leq \gamma$ as $\text{ineq}$ in the proof. Since $\text{ineq}$ defines a facet of $C(I,J)$, there exist $(2|I|+1)|J|$ affinely independent points in $C(I,J)$ that are on the plane defined by $\text{ineq}$. We denote this set of the $(2|I|+1)|J|$ affinely independent points as $V$. We will construct additional $(2m + 1)n - (2|I|+1)|J|$ linearly independent points of $\text{conv}(\chi^L)$ that are on the plane defined by $\text{ineq}$. Let $[x,y,q]$ be an arbitrary point in $V$, we
first construct the following set of points:

\[ V^1 = \{ [x, y, q] + [0, e_j, 0] \mid \forall j \in [n] \setminus J \} \]

\[ \cup \{ [x, y, q] + [0, e_j, 0] \mid \forall i \in [m] \setminus I, \forall j \in [n] \setminus J \} \]

\[ \cup \{ [x, y, q] + [\varepsilon e_{ij}, e_j, e_{ij}] \mid \forall i \in [m] \setminus I, \forall j \in [n] \setminus J \}, \]

with \( \varepsilon > 0 \) sufficient small. We can verify that all points in \( V^1 \) are linearly independent points of \( \text{conv}(\chi^L) \) that are on the plane defined by \( \text{ineq} \), and \( |V^1| = (2m - 2|I| + 1)(n - |J|) \). Let \( [x^{(j)}, y^{(j)}, q^{(j)}] \in V \) for every \( j \in J \) be the vector satisfying the condition (2). We construct the following set of points:

\[ V^2 = \{ [x^{(j)}, y^{(j)}, q^{(j)}] + [0, 0, e_{ij}] \mid \forall i \in [m] \setminus I, \forall j \in J \} \]

\[ \cup \{ [x^{(j)}, y^{(j)}, q^{(j)}] + [\varepsilon e_{ij}, 0, e_{ij}] \mid \forall i \in [m] \setminus I, \forall j \in J \}, \]

with sufficient small \( \varepsilon > 0 \). One can verify that all points in \( V^2 \) are linearly independent points of \( C(I, J) \) that are on the plane defined by \( \text{ineq} \), and \( V^2 \) contains \( 2(m - |I|)|J| \) linearly independent points. Finally, we construct the following set of points:

\[ V^3 = \{ [\tilde{x}^{(i)}, \tilde{y}^{(i)}, \tilde{q}^{(i)}] + [0, e_j, \varepsilon e_{ij}] \mid \forall i \in I, \forall j \in [n] \setminus J \} \]

\[ \cup \{ [\tilde{x}^{(i)}, \tilde{y}^{(i)}, \tilde{q}^{(i)}] + [\varepsilon' e_{ij}, e_j, \varepsilon e_{ij}] \mid \forall i \in I, \forall j \in [n] \setminus J \}, \]

where \( [\tilde{x}^{(i)}, \tilde{y}^{(i)}, \tilde{q}^{(i)}] \) is the point satisfying the condition (3) for each \( i \in I \), and \( \varepsilon, \varepsilon' > 0 \) are sufficient small numbers. Again, it is straightforward to verify that all points in \( V^3 \) are linearly independent points of \( \text{conv}(\chi^L) \) that are on the plane defined by \( \text{ineq} \), and \( |V^3| = 2|I|(|n - |J||) \). By looking at the perturbation terms involved in \( V^1 \), \( V^2 \) and \( V^3 \) we see that the vectors in \( V^1 \), \( V^2 \) and \( V^3 \) are linearly independent. Furthermore, since \( \text{span}\{V^1, V^2, V^3\} \cap \text{span}(V) = \{0, 0, 0\} \), then by Lemma 2 all vectors in \( V \cup V^1 \cup V^2 \cup V^3 \) are affinely independent. The number of points we have constructed is

\[ |V| + |V^1| + |V^2| + |V^3| = (2|I| + 1)|J| + (2m - 2|I| + 1)(n - |J|) \]

\[ + 2(m - |I|)|J| + 2|I|(|n - |J||) \]

\[ = 2mn + n, \]

which shows that the \( \text{ineq} \) defines a facet of \( \text{conv}(\chi^L) \). \( \square \)
Theorem 2. Suppose $d_{ij} > 0$ for all $i \in [m], j \in [n]$. The following hold:
(a) For any subsets of indices $I \subseteq [m], J \subset [n]$ and index $j' \in [n] \setminus J$ the following inequality is valid for $\text{conv}(\mathbf{x}^L)$
\[
\sum_{i \in I} \sum_{j \in J \cup \{j'\}} x_{ij} - \left( \sum_{i \in I} \max_{j \in [n]} d_{ij} - \sum_{j \in J} c_j \right) y_{j'} \leq \sum_{j \in J} c_j. \tag{14}
\]
(b) Let $I \subseteq [m]$ and $J \subset [n]$ be two subset of indices, if $\sum_{j \in J} c_j \leq \sum_{i \in I} \max_{j \in J} d_{ij}$, then the following inequality is valid for $\text{conv}(\mathbf{x}^L)$
\[
\sum_{i \in I} \sum_{j' \in [n] \setminus J} x_{ij} - \sum_{j' \in [n] \setminus J} \left( \sum_{i \in I} \max_{j \in [n]} d_{ij} - \sum_{j \in J} c_j \right) y_{j'} \leq \sum_{j \in J} c_j. \tag{15}
\]
Proof. Part (a): It suffices to show the validness for the two cases: $y_{j'} = 0$ and $y_{j'} = 1$. If $y_{j'} = 0$, then we have
\[
\sum_{i \in I} \sum_{j \in J \cup \{j'\}} x_{ij} - \left( \sum_{i \in I} \max_{j \in [n]} d_{ij} - \sum_{j \in J} c_j \right) y_{j'} = \sum_{i \in I} \sum_{j \in J} x_{ij} \leq \sum_{j \in J} \sum_{i \in I} x_{ij} \leq \sum_{j \in J} c_j, \tag{16}
\]
where we use $y_{j'} = 0, y_j \leq 1$ and the constraint $\sum_{i \in [m]} x_{ij} \leq c_j y_j$ to get the above inequalities. For the case of $y_{j'} = 1$, we have
\[
\sum_{i \in I} \sum_{j \in J \cup \{j'\}} x_{ij} - \left( \sum_{i \in I} \max_{j \in [n]} d_{ij} - \sum_{j \in J} c_j \right) y_{j'} = \sum_{i \in I} \sum_{j \in J \cup \{j'\}} x_{ij} - \sum_{i \in I} \max_{j \in [n]} d_{ij} + \sum_{j \in J} c_j \leq \sum_{i \in I} \sum_{j \in [n]} x_{ij} - \sum_{i \in I} \max_{j \in [n]} d_{ij} + \sum_{j \in J} c_j \leq \sum_{i \in I} \sum_{j \in [n]} d_{ij} q_{ij} - \sum_{i \in I} \max_{j \in [n]} d_{ij} + \sum_{j \in J} c_j \leq \sum_{j \in [n]} c_j, \tag{17}
\]
where we use $y_{j'} = 1$, and the constraint $\sum_{j \in [n]} x_{ij} \leq \sum_{j \in [n]} d_{ij} q_{ij}$ in (17).

Part (b): For a feasible point $[x, y, q]$, let $J' = \{ j \in [n] \setminus J \mid y_j = 1 \}$. For any $i \in I$, let $\pi(i) = \arg\max_{j \in J \cup J'} d_{ij}$ where ties are broken arbitrarily. Let $I_j = \{ i \in I \mid \pi(i) = j \}$ for every $j \in J \cup J'$, and by definition the subsets $\{I_j\}_{j \in J \cup J'}$ form a partition of $I$. If $|J'| \leq 1$, then (15) reduces to (14). Therefore without
loss of generality, we assume \(|J'| \geq 2\). The expression on the left can be relaxed as
\[
\sum_{i \in I} \sum_{j \in [n] \setminus J} x_{ij} - \sum_{j' \in [n] \setminus J} \left( \sum_{i \in I} \max_{j \in J \cup \{j'\}} d_{ij} - \sum_{j \in J} c_j \right) y_{j'} \\
\leq \sum_{i \in I} \sum_{j \in J \cup J'} d_{ij} q_{ij} - \sum_{j' \in J' \setminus J} \left( \sum_{i \in I} \max_{j \in J \cup \{j'\}} d_{ij} - \sum_{j \in J} c_j \right). \tag{18}
\]

Notice that we have the following inequalities hold
\[
\sum_{i \in I} \sum_{j \in J \cup J'} d_{ij} q_{ij} \leq \sum_{i \in I} d_{i \pi(i)} = \sum_{j \in J \cup J'} \sum_{i \in I} j \max_{j \in J \cup \{j'\}} d_{ij} - \sum_{j \in J} c_j. \tag{19}
\]

Furthermore, the second term in (15) is evaluated as
\[
\sum_{j \in [n] \setminus J} \left( \sum_{i \in I} \max_{j \in J \cup \{j'\}} d_{ij} - \sum_{j \in J} c_j \right) y_{j'} = \sum_{j \in J'} \sum_{i \in I} \max_{j \in J \cup \{j'\}} d_{ij} - \sum_{j \in J} c_j \tag{20}
\]

The first term in (20) can be decomposed as
\[
\sum_{j \in [n] \setminus J} \sum_{i \in I} \max_{j \in J \cup \{j'\}} d_{ij} \tag{21}
\]

where \(T_1\) and \(T_2\) represent the first and second term on the right side of (19), respectively. The term \(T_1\) is evaluated as
\[
T_1 = \sum_{j \in J'} \sum_{i \in I_j} \max_{j \in J \cup \{j'\}} d_{ij} = \sum_{j \in J} \sum_{i \in I_j} \max_{j \in J \cup \{j'\}} d_{ij} \tag{22}
\]

where we use the fact that for any \(j^* \in J, i \in I_j, \) and \(j' \in [n] \setminus J,\) we have
\[ d_{ij^*} = \max_{j \in J \cup \{j^*\}} d_{ij} = \max_{j \in J} d_{ij}. \] The term \( T_2 \) can be lower bounded as:

\[
T_2 = \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} \\
= \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} \\
= \sum_{j^* \in J'} \sum_{i \in I_{j^*}} d_{ij^*} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} \\
\geq \sum_{j^* \in J'} \sum_{i \in I_{j^*}} d_{ij^*} + (|J'| - 1) \sum_{i \in I} \sum_{j \in J} \max d_{ij}. \tag{23}
\]

Substituting (22) and (23) into (21) yields the following inequalities:

\[
\sum_{j^* \in J'} \sum_{j \in J \cup \{j^*\}} \sum_{i \in I_{j^*}} \max d_{ij} \geq |J'| \sum_{j^* \in J} \sum_{i \in I_{j^*}} \max d_{ij} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} \\
+ (|J'| - 1) \sum_{i \in I} \sum_{j \in J} \max d_{ij} \\
= \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \max d_{ij} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} \\
= \sum_{j^* \in J'} \sum_{i \in I_{j^*}} d_{ij^*} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \sum_{j \in J \cup \{j^*\}} \max d_{ij} \\
+ (|J'| - 1) \sum_{i \in I} \sum_{j \in J} \max d_{ij}, \tag{24}
\]

where we use the fact that \( \{I_{j^*} \mid j^* \in J \cup \{j^*\} \} \) form a partition of \( I \). Substituting (24) into (20) yields

\[
\sum_{j^* \in [n] \setminus J} \left( \sum_{i \in I} \max_{j \in J \cup \{j^*\}} d_{ij} - \sum_{j \in J} c_j \right) y_{j^*} \geq \sum_{j^* \in J} \sum_{i \in I_{j^*}} \max d_{ij} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \max d_{ij^*} \\
+ (|J'| - 1) \sum_{i \in I} \max d_{ij} - |J'| \sum_{j \in J} c_j \\
\geq \sum_{j^* \in J} \sum_{i \in I_{j^*}} \max d_{ij} + \sum_{j^* \in J'} \sum_{i \in I_{j^*}} \max d_{ij^*} - \sum_{j \in J} c_j \tag{25}
\]

where we use the assumption \( \sum_{i \in I} \max_{j \in J} d_{ij} - \sum_{j \in J} c_j \geq 0 \). Substituting (19) and (25) into (18) yields

\[
\sum_{i \in I} \sum_{j \in [n] \setminus J} x_{ij} - \sum_{j^* \in [n] \setminus J} \left( \sum_{i \in I} \max_{j \in J \cup \{j^*\}} d_{ij} - \sum_{j \in J} c_j \right) y_{j^*} \leq \sum_{j \in J} c_j. \tag{26}
\]

This shows that (15) is valid for \( \text{conv}(\chi^L) \). \( \square \)
Theorem 3. Let $I \subseteq [m]$ and $J \subseteq [n]$. Suppose the following conditions are satisfied

$$
\sum_{j \in J} c_j \geq \sum_{i \in I} \max_{j \in J} d_{ij}, \quad \max_{j \in J} d_{ij} = \max_{j \in [n]} \max_{j \in J} d_{ij} \quad \forall i \in I,
$$

(27)

$$
\sum_{j \in J \setminus \{j_0\}} c_j < \sum_{i \in I} \max_{j \in J \setminus \{j_0\}} d_{ij} \quad \forall j_0 \in J.
$$

Then the following inequality defines a facet of $\text{conv}(\chi^L)$

$$
\sum_{i \in I} \sum_{j \in J} x_{ij} + \sum_{j \in J} \left( \sum_{j' \in J \setminus \{j\}} c_{j'} - \sum_{i \in I} \max_{j' \in J \setminus \{j\}} d_{ij'} \right) y_j \leq (|J|-1) \left( \sum_{j \in J} c_j - \sum_{i \in I} \max_{j \in J \setminus \{j_0\}} d_{ij} \right).
$$

(28)

**Proof.** We will first provide an intuition of how the coefficients of the inequality (28) are determined given the condition in (27). Since we restrict the decision variables induced by the subsets $I$ and $J$, a facet-defining valid inequality is likely in the form of

$$
\sum_{i \in I} \sum_{j \in J} x_{ij} + \sum_{j \in J} a_j y_j \leq b.
$$

(29)

We need to determine the coefficients $a_j$ and $b$. Notice that there are $|J| + 1$ coefficients to be determined, and we will set up $|J| + 1$ linear equations corresponding to the following $|J| + 1$ situations: (1) opening service centers at all locations in the subset $J$, (2) opening service centers at all locations in $J \setminus \{j\}$ for all $j \in J$. Notice that (2) contains $|J|$ different situations referred as situation $(2,j)$. In the first situation, all locations in $J$ have service centers. So the maximum value of $\sum_{i \in I} \sum_{j \in J} x_{ij}$ is $\sum_{i \in I} \max_{j \in J} d_{ij}$, i.e., the aggregated demand constraint is binding. Similarly, in the situation $(2,j_0)$, locations in $J \setminus \{j_0\}$ have service centers, and the maximum value of $\sum_{i \in I} \sum_{j \in J} x_{ij}$ is $\sum_{j \in J \setminus \{j_0\}} c_j$, i.e., the aggregated capacity constraint is binding. These observations help us set up the following linear equations to determine $a_j$ and $b$:

$$
\sum_{i \in I} \max_{j \in J} d_{ij} + \sum_{j \in J} a_j = b, \quad (30a)
$$

$$
\sum_{j' \in J \setminus \{j\}} c_{j'} + \sum_{j' \in J \setminus \{j\}} a_{j'} = b \quad \forall j \in J. \quad (30b)
$$
By solving the above linear equation system, we can determine the values of all coefficients:

\[ b = (|J| - 1) \left( \sum_{j \in J} c_j - \sum_{i \in I} \max_{j' \in J} d_{ij'} \right), \]

\[ a_j = \sum_{j' \in J \setminus \{j\}} c_{j'} - \sum_{i \in I} \max_{j' \in J} d_{ij'} \quad \forall j \in J, \]

which leads to the inequality (28).

Now we give a formal proof of that (28) is a facet-defining valid inequality of \( \operatorname{conv}(\chi^L) \). To see the validness, we divide the analysis into four cases: (a) \( y_j = 1 \) for all \( j \in J \); (b) \( y_j = 1 \) and \( y_{j'} = 0 \) for all \( j' \in J \setminus \{j\} \); (c) there exists a \( J_0 \subset J \) satisfying \( 2 \leq |J_0| \leq |J| - 1 \) and \( y_j = 1 \) for all \( j \in J \setminus J_0 \) and \( y_{j'} = 0 \) for all \( j \in J_0 \); (d) \( y_j = 0 \) for all \( j \in J \). The validness of (28) clearly holds in the cases (a) and (b) by the way of determining the coefficients. For the case (d), (28) becomes

\[ \sum_{i \in I} \sum_{j \in J} x_{ij} \leq (|J| - 1) \left( \sum_{j \in J} c_j - \sum_{i \in I} \max_{j' \in J} d_{ij'} \right). \]

It is valid because in this case we have \( \sum_{i \in I} \sum_{j \in J} x_{ij} = 0 \) and \( c_j - \sum_{i \in I} \max_{j' \in J} d_{ij'} > 0 \) by assumption. It remains to verify the validness in the case (c). Substitute \( y_j = 1 \) for \( j \in J \setminus J_0 \) into (28), and notice that we have

\[ (|J| - 1) \left( \sum_{j \in J} c_j - \sum_{i \in I} \max_{j' \in J} d_{ij'} \right) - \sum_{j \in J \setminus J_0} \left( \sum_{j' \in J \setminus \{j\}} c_{j'} - \sum_{i \in I} \max_{j' \in J} d_{ij'} \right) \]

\[ = \sum_{j_0 \in J_0} \sum_{j \in J \setminus \{j_0\}} c_j - (|J_0| - 1) \sum_{i \in I} \max_{j' \in J} d_{ij'} \]

\[ \geq \sum_{j_0 \in J_0} \sum_{j \in J \setminus \{j_0\}} c_j - (|J_0| - 1) \sum_{j \in J} c_j \]

\[ = (|J_0| - 1) \sum_{j \in J_0} c_j + |J_0| \sum_{j \in J \setminus J_0} c_j - (|J_0| - 1) \sum_{j \in J} c_j \]

\[ = \sum_{j \in J \setminus J_0} c_j \sum_{i \in I} \sum_{j \in J} x_{ij}, \]

where the assumed condition is used to get the first inequality above. It proves the validness in this case.

To show that (28) defines a facet of \( \operatorname{conv}(\chi^L) \), we will first show that it defines a facet of \( \mathcal{C}(I, J) \) (see (12)), and apply Theorem 1 to conclude that it also defines
a facet of $\text{conv}(\chi^L)$ via verifying the three conditions in Theorem 1 are satisfied. Let $V = \{[x, y, q] \in C(I, J) \mid [x, y, q] \text{ satisfies (28) as equality}\}$. Clearly, there exists a facet of $C(I, J)$ containing all points in $V$, and suppose this facet is represented in the following general form:

$$
\sum_{i \in I} \sum_{j \in J} \alpha_{ij} x_{ij} + \sum_{j \in J} \beta_j y_j + \sum_{i \in I} \sum_{j \in J} \gamma_{ij} q_{ij} \leq \lambda. \tag{31}
$$

It suffices to show that the above coefficients are different from the coefficients in (28) by a positive constant. Note that all points in $V$ should satisfy (31) as an equality. We are going to propose some specific points from $V$ and plug them into (31) to determine the coefficients. First, we notice that there must exist a vector $[\sum_{j \in J} x^0_{ij}, e_j, q^0] \in V$ such that $x^0_{ij} > 0$, $0 < q^0_{ij} < 1$ for all $(i, j) \in I \times J$, and $\sum_{i \in I} x^0_{ij} < e_j$ for all $j \in J$. It can be deduced that the points $[x^0 + \epsilon e_{ij_1} - \epsilon e_{ij_2}, \sum_{j \in J} e_j, q^0]$ for all $i \in I$, $j_1, j_2 \in J$, $j_1 \neq j_2$, and $[\sum_{j \in J} e_j, x^0 + \epsilon e_{ij_1} - \epsilon e_{ij_2}, q^0]$ for all $i_1, i_2 \in I$, $j \in J$ are also in $V$ for some sufficient small $\epsilon$. Substituting these points into (31) as an equality, we conclude that $\alpha_{ij_1} = \alpha_{ij_2}$ and $\alpha_{ij_1} = \alpha_{ij_2}$, which implies that $\alpha_{ij} = \alpha \forall (i, j) \in I \times J$ for some $\alpha > 0$. Without loss of generality, we can set $\alpha = 1$ since the inequality is invariant under multiplication of a positive constant. Similarly, the exists a point $[x^1, \sum_{J \setminus \{j_0\}} e_j, q^1] \in V$ such that $0 < q^1_{ij} < 1$ for all $(i, j) \in I \times (J \setminus \{j_0\})$. It can be deduced that the points $[x^1, \sum_{J \setminus \{j_0\}} e_j, q^1 + \epsilon e_{ij_1}] \forall (i_1, j_1) \in I \times (J \setminus \{j_0\})$ are all in $V$. Substituting this point into (31) as an equality, we conclude that $\gamma_{ij} = 0$ for all $(i, j) \in I \times (J \setminus \{j_0\})$. Since $j_0$ is arbitrary, we conclude that $\gamma_{ij} = 0$ for all $(i, j) \in I \times J$. Therefore, (31) becomes

$$
\sum_{i \in I} \sum_{j \in J} x_{ij} + \sum_{j \in J} \beta_j y_j \leq \lambda. \tag{32}
$$

Clearly, since the previously considered point $[x^0, \sum_{j \in J} e_j, q^0] \in V$ should satisfy $\sum_{i \in I} \sum_{j \in J} x^0_{ij} = \sum_{i \in I} \max_{j \in J} d_{ij}$ (this can be obtained by substituting the point into (28) as an equality). Then substituting this point into (32) as an equality gives an equation in the same form of (30a). Similarly, substituting the point $[x^1, \sum_{J \setminus \{j_0\}} e_j, q^1] \in V$ into (32) as an equality gives an equation in
the same form of (30b). Therefore the to be determined coefficients \([\beta, \gamma]\) are the solution of (30), which shows that the facet defining valid inequality (32) of \(\mathcal{C}(I, J)\) is exactly (28).

Now we show that (28) is also a facet defining inequality of \(\text{conv}(\chi^L)\) using Theorem 1. To achieve this, we first need to show that (28) is valid for \(\text{conv}(\chi^L)\). The validness of (28) for \(\text{conv}(\chi^L)\) is based on the following two observations: if a subset \(J' \subseteq [n] \setminus J\) of service centers are opened, the maximum attracted demand from a customer site \(i\) is at most \(\max_{j \in J \cup J'} d_{ij} \leq \max_{j \in [n]} d_{ij}\). But \(\max_{j \in [n]} d_{ij} = \max_{j \in J} d_{ij}\) by assumption, which implies that once all places in \(J\) have service centers, the aggregated demand constraint \(\sum_{i \in I} \sum_{j \in J} x_{ij} \leq \sum_{i \in I} \max_{j \in [n]} d_{ij}\) is valid and it is independent of whether or not other places having service centers. Similarly, in the case that places at \(J \setminus \{j_0\}\) have service centers, the aggregated capacity constraint dominates the aggregated demand constraint which is also independent of service center locations at other places due to the fact \(\max_{j \in J \setminus \{j_0\}} d_{ij} \leq \max_{j \in J \cup J' \setminus \{j_0\}} d_{ij}\) for any \(J' \subseteq [n] \setminus J\). It remains to show that there exist \((2|I| + 1)|J|\) affinely independent vectors in \(\mathcal{C}(I, J)\) that satisfy the three conditions in Theorem 1. Since we have already proved that (28) is a facet defining inequality of \(\mathcal{C}(I, J)\), there must exist a set \(V\) of \((2|I| + 1)|J|\) affinely independent vectors in \(\mathcal{C}(I, J)\) that satisfy (28) as an equality. Furthermore, we can make sure that \(V\) includes the following two vectors: \([x^0, \sum_{j \in J} e_j, q^0]\) and \([x^1, \sum_{j \in J \setminus \{j_0\}} e_j, q^1]\) where the two vectors are introduced in the previous analysis of this proof. Notice that \([x^0, \sum_{j \in J} e_j, q^0]\) satisfies the condition (2) of Theorem 1 for all \(j \in J\), and \([x^1, \sum_{j \in J \setminus \{j_0\}} e_j, q^1]\) satisfies the condition (3) of Theorem 1 for each \(i \in I\). This concludes the proof.

Theorem 2 and 3 provide valid and facet-defining inequalities for \(\text{conv}(\chi^L)\). Notice that these inequalities do not involve variables \(q\) introduced to lift \(\chi\). This observation raises a question: are these inequalities also valid and facet-defining for \(\text{conv}(\chi)\)? The following proposition gives an affirmative answer to this question.
Proposition 3. Let \([x,y] \in \mathbb{R}^{m+n}\), \(P\) be a full dimensional polytope in \(\mathbb{R}^m\) represented by a linear-inequality system of variables \(x\), and \(Q\) be a full dimensional polytope in \(\mathbb{R}^{m+n}\) represented by a linear-inequality system of variables \(x,y\). Suppose \(P = \text{proj}_x Q\). If \(\alpha^\top x + \beta \leq 0\) is a valid (facet-defining) inequality of \(Q\), it is also a valid (facet-defining) inequality of \(P\).

Proof. (1) Suppose \(\alpha^\top x + \beta \leq 0\) is a valid inequality of \(Q\). Due to the projection relationship, for any \(x' \in P\), there exists a \([x',y'] \in Q\). Since \(\alpha^\top x + \beta \leq 0\) is valid for \(Q\), we have \(\alpha^\top x' + \beta \leq 0\), which implies that \(\alpha^\top x + \beta \leq 0\) is valid for \(P\). (2) Suppose \(\alpha^\top x + \beta \leq 0\) is a facet-defining inequality of \(Q\). We focus on the case that the facet does not contain the point \([0,0]\). By definition, there exists \(m+n\) linearly independent points \([x^k,y^k] : k \in [m+n]\) satisfying \(\alpha^\top x^k + \beta = 0\). Since every \(y^k\) is a \(n\)-dimensional vector, the set \([x^k] : k \in [m+n]\) contains \(m\) linearly independent vectors. In other words, there exist \(m\) linearly independent vectors on the hyperplane \(\alpha^\top x + \beta = 0\), which shows that \(\alpha^\top x + \beta \leq 0\) defines a facet of \(P\).

Corollary 1. The inequality (14) is valid for \(\text{conv}(\chi)\), the inequality (15) is valid for \(\text{conv}(\chi)\) given that the condition in Theorem 2 is satisfied, and the inequality (28) defines a facet of \(\text{conv}(\chi)\) given that the conditions in Theorem 3 is satisfied.

3. Conclusion

The facet-defining inequalities given in Theorem 4 hold when the parameters of demand and capacity satisfy a certain condition for a subset of client locations and candidate facility locations. The same spirit can be used to construct various conditions that lead to different families of facet-defining inequalities. Identifying the subsets that satisfy the conditions is also a NP-hard problem, which prevents a systematic application of these inequalities in computation at low cost of computing power. This indicates that the facet-defining inequalities identified in this work is of more theoretical interest for understanding the
facility-location polyhedral structure under competing constraints of capacity and demand.

Acknowledgement

This research was partially supported by the ONR grant N00014-18-1-2097.

References

[1] K. Aardal, Y. Pochet, L. A. Wolsey, Capacitated facility location: valid inequalities and facets, Mathematics of Operations Research 20 (3) (1995) 562–582.

[2] M. S. Daskin, Network and Discrete Location: Models, Algorithms, and Applications, 2nd Edition, Wiley, 2013.

[3] S. Melkote, M. S. Daskin, An integrated model of facility location and transportation network design, Transportation Research Part A: Policy and Practice 35 (6) (2001) 515–538.

[4] M. T. Melo, S. Nickel, F. S. da Gama, Facility location and supply chain management - a review, European Journal of Operational Research 196 (2) (2009) 401–412.

[5] C. Boonmee, M. Arimura, T. Asada, Facility location optimization model for emergency humanitarian logistics, International Journal of Disaster Risk Reduction 24 (2017) 485–498.

[6] J. Allen, M. Browne, T. Cherrett, Investigating relationships between road freight transport, facility location, logistics management and urban form, Journal of Transport Geography 24 (2012) 45–57.

[7] S. H. Owen, M. S. Daskin, Strategic facility location: a review, European Journal of Operational Research 111 (3) (1998) 423–447.
[8] F. V. Louveaux, D. Peeters, A dual-based procedure for stochastic facility location, Operations Research 40 (3) (1992) 564–573.

[9] M. Albareda-Sambola, E. Fernández, F. S. da Gama, The facility location problem with Bernoulli demands, Omega 39 (2011) 335–345.

[10] Q. Wang, R. Batt, C. M. Rump, Algorithms for a facility location problem with stochastic customer demand and immobile servers, Annals of Operations Research 111 (2002) 17–34.

[11] P. Murali, F. Ordóñez, M. M. Dessouky, Facility location under demand uncertainty: response to a large-scale bio-terror attack, Socio-Economic Planning Sciences 46 (2012) 78–87.

[12] N. Gülpinar, D. Pachamanova, E. Çanakoğlu, Robust strategies for facility location under uncertainty, European Journal of Operational Research 225 (2013) 21–35.

[13] M. K. Lim, A. Bassamboo, S. Chopra, M. S. Daskin, Facility location decisions with random disruptions and imperfect estimation, Manufacturing & Service Operations Management 15 (2) (2013) 239–249.

[14] Y. Li, J. Shu, M. Song, J. Zhang, H. Zheng, Multisourcing supply network design: two-stage chance-constrained model, tractable approximations, and computational results, INFORMS Journal on Computing 29 (2) (2017) 287–300.

[15] L. V. Snyder, Facility location under uncertainty: a review, IIE Transactions 38 (2006) 537–554.

[16] O. Baron, J. Milner, Facility location: a robust optimization approach, Production and Operations Management 20 (5) (2010) 772–785.

[17] A. Gourtani, T. D. Nguyen, H. Xu, A distributionally robust optimization approach for two-stage facility location problems, EURO Journal on Computational Optimization 8 (2020) 141–172.
[18] T. Chan, Z. Shen, A. Siddiq, Robust defibrillator deployment under cardiac arrest location uncertainty via row-and-column generation, Operations Research 66 (2017) 358–379.

[19] A. Döyen, N. Aras, G. Barbarosoğlu, A two-echelon stochastic facility location model for humanitarian relief logistics, Optim. Lett. 6 (2012) 1123–1145.

[20] P. G. Berglund, C. Kwon, Robust facility location problem for hazardous waste transportation, Netw. Spat. Econ. 14 (2014) 91–116.

[21] V. Goel, I. E. Grossmann, A class of stochastic programs with decision dependent uncertainty, Math. Program. 108 (2006) 355–394.

[22] M. J. Eppli, J. D. Shilling, How critical is a good location to a regional shopping center?, The Journal of Real Estate Research 12 (3) (1996) 459–468.

[23] Rajagopal, Growing shopping malls and behavior of urban shoppers, Journal of Retail & Leisure Property 8 (2009) 99–118.

[24] B. Basciftci, S. Ahmed, S. Shen, Distributionally robust facility location problem under decision-dependent stochastic demand, https://doi.org/10.1016/j.ejor.2020.11.002 (2020).

[25] M. Conforti, G. Cornuéjols, G. Zambelli, Integer Programming, 2014th Edition, Springer, 2014.

[26] G. L. Nemhauser, L. A. Wolsey, Integer and Combinatorial Optimization, 1st Edition, Wiley-Interscience, 1999.

[27] Y. Pochet, L. A. Wolsey, Lot-sizing with constant batches: Formulation and valid inequalities, Mathematics of Operations Research 18 (4) (1993) 767–785.

[28] K. Aardal, Y. Pochet, L. A. Wolsey, Capacitated facility location: Valid inequalities and facets, Mathematics of Operations Research 20 (3) (1995) 562–582.
[29] G. Perboli, R. Tadei, R. Tadei, New families of valid inequalities for the two-
  echelon vehicle routing problem, Electronic Notes in Discrete Mathematics
  36 (1) (2010) 639–646.

[30] L. C. Coelho, G. Laporte, Improved solutions for inventory-routing prob-
  lems through valid inequalities and input ordering, International Journal
  of Production Economics 155 (1) (2014) 391–397.

[31] R. Baldacci, M. Battarra, D. Vigo, Valid inequalities for the fleet size and
  mix vehicle routing problem with fixed costs, Networks 54 (2009) 178–189.

[32] D. Huygens, M. Labbé, A. R. Mahjoub, P. Pesneau, The two-edge con-
  nected hop-constrained network design problem: Valid inequalities and
  branch-and-cut, Networks 49 (1) (2006) 116–133.

[33] Q. Louveaux, L. A. Wolsey, Lifting, superadditivity, mixed integer rounding
  and single node flow sets revisited, Annals of Operations Research 153 (1)
  (2007) 47–77. doi:10.1007/s10479-007-0171-7.

[34] J. Richard, I. de Farias Jr, G. Nemhauser, Lifted inequalities for 0-1 mixed
  integer programming: Basic theory and algorithm, Mathematical Program-
  ming 98 (2003) 89–113. doi:10.1007/s10107-003-0398-2.

[35] A. Atamtürk, On the facets of the mixed-integer knapsack polyhe-
  dron, Mathematical Programming 98 (2003) 145–175. doi:10.1007/s10107-
  003-0400-z.

[36] Z. Gu, G. L. Nemhauser, M. Savelsbergh, Lifted cover inequalities for mixed
  0-1 integer programs: Complexity, INFORMS Journal on Computing 11 (1)
  (1999) 117–123. doi:10.1287/ijoc.11.1.117.

[37] Z. Gu, G. L. Nemhauser, M. W. Savelsbergh, Lifted flow cover inequalities
  for mixed 0-1 integer programs, Mathematical Programming 85 (3) (1999)
  439–467. doi:10.1007/s101070050067.
[38] Z. Gu, G. L. Nemhauser, M. Savelsbergh, Lifted cover inequalities for mixed 0-1 integer programs: Computation, INFORMS Journal on Computing 10 (4) (1998) 427–437. doi:10.1287/ijoc.10.4.427

[39] Y. H. Au, L. Tunçel, A comprehensive analysis of polyhedral lift-and-project methods, SIAM Journal on Discrete Mathematics 30 (1) (2016) 411–451. doi:10.1137/130950173

[40] B. Kocuk, H. Jeon, S. S. Dey, J. Linderoth, J. Luedtke, X. A. Sun, A cycle-based formulation and valid inequalities for DC power transmission problems with switching, Operations Research 64 (4) (2016) 922–938. doi:10.1287/opre.2015.1471

[41] M. R. Kılınç, J. Linderoth, J. Luedtke, Lift-and-project cuts for convex mixed integer nonlinear programs, Mathematical Programming Computation 9 (4) (2017) 499–526. doi:10.1007/s12532-017-0118-1

[42] S. Burer, D. Vandenbussche, Solving lift-and-project relaxations of binary integer programs, SIAM Journal of Optimization 16 (3) (2006) 726–750. doi:10.1137/040609574

[43] N. E. Aguilera, S. M. Bianchi, G. L. Nasini, Lift and project relaxations for the matching and related polytopes, Discrete Applied Mathematics 134 (1-3) (2004) 193–212. doi:10.1016/s0166-218x(03)00337-8

[44] E. Balas, A modified lift-and-project procedure, Mathematical Programming 79 (1) (1997) 19–31. doi:10.1007/bf02614309

[45] G. B. Dantzig, B. C. Eaves, Fourier-Motzkin elimination and its dual, Journal of Combinatorial Theory, Series A 14 (3) (1973) 288–297. doi:10.1016/0097-3165(73)90004-6

[46] H. P. Williams, Fourier-Motzkin elimination extension to integer programming problems, Journal of Combinatorial Theory, Series A 21 (1) (1976) 118–123. doi:10.1016/0097-3165(76)90055-8
[47] M. Schechter, Integration over a polyhedron: An application of the Fourier-Motzkin elimination method, The American Mathematical Monthly 105 (3) (1998) 246–251. doi:10.1080/00029890.1998.12004874

[48] E. Balas, S. Ceria, G. Cornuéjols, A lifted-and-project cutting plane algorithm for mixed 0-1 programs, Mathematical Programming 58 (1993) 295–324.

[49] K. Aardal, Y. Pochet, L. A. Wolsey, Capacitated facility location: valid inequalities and facets, Mathematics of Operations Research 20 (3) (1995) 562–582.

[50] K. Aardal, Capacitated facility location: separation algorithms and computational experience, Mathematical Programming 81 (1998) 149–175.

[51] K. Aardal, A branch and cut algorithm for hub location problems with single assignment, Mathematical Programming 102 (2004) 371–405.