

Combining Problem Structure with Basis Reduction to solve a Class of Hard Integer Programs

Quentin Louveaux*
CORE, UCL, Belgium[†]

Laurence A. Wolsey
CORE, UCL, Belgium[†]

October 2000

Abstract

Recently Aardal et al. [2] have successfully solved some small difficult equality constrained integer programs by using basis reduction to reformulate the problems as inequality constrained integer programs in a different space. Here we adapt their method to solve integer programs that are larger, but have special structure. The practical problem motivating this work has variables x_{ij} and is a variant of the market share problem. More formally the problem can be viewed as finding a matrix $X \in \mathbb{Z}_+^{m \times n}$ satisfying $XA = C$, $BX = D$ where A, B, C, D are matrices of compatible dimensions, and the approach requires us to find a reduced basis of the lattice $\mathcal{L} = \{X \in \mathbb{Z}^{m \times n} : XA = 0, BX = 0\}$.

The main topic of this paper is a study of the lattice $\mathcal{L}$. It is shown that an integer basis of $\mathcal{L}$ can be obtained by taking the Kronecker product of vectors from integer bases of two much smaller lattices. Furthermore the resulting basis is a reduced basis if the integer bases of the two small lattices are reduced bases, and a suitable ordering is chosen.

Finally some limited computational results are presented showing the benefits of making use of the problem structure.

Keywords: Integer Programming, Basis reduction, Lattices, Market share problem.

*This author would like to thank the Belgian National Fund for Scientific Research (FNRS) for its financial support.

[†]Center of Operations Research and Econometrics, Université catholique de Louvain. 34, voie du Roman Pays. 1348 Louvain-la-Neuve, Belgium

1 Introduction

Consider a banker who must establish a certain number of portfolios for his clients. Client i 's portfolio must consist of d_i shares, and the banker holds a_j shares of type j whose estimated profit per share is p_j . The banker's problem is to divide up the $\sum a_j$ shares among the clients so that the expected profit per share of each client is as close as possible to the average value. Mathematically, he has the problem of finding a solution of

$$\begin{cases} \sum_j x_{ij} = d_i & 1 \leq i \leq m \\ \sum_i x_{ij} = a_j & 1 \leq j \leq n \\ \frac{1}{d_i} \left(\sum_j p_j x_{ij} \right) \simeq \bar{c} & 1 \leq i \leq m \\ x \in \mathbb{Z}_+^{mn}, \end{cases} \quad (1)$$

where $\bar{c}$ is the average expected profit $\frac{\sum_i \sum_j p_{ij} x_{ij}}{\sum x_{ij}}$ and $\simeq$ means that we want to be as close as possible to equality. One natural way to attempt to solve this problem is to introduce nonnegative slack variables s_i^+ and s_i^- , write $\frac{1}{d_i} \sum_j p_j x_{ij} + s_i^+ - s_i^- = \bar{c}$, take as objective function: $\min \sum_i s_i^+ + \sum_i s_i^-$, and then solve the resulting mixed integer program with a commercial system such as Cplex or Xpress. However this does not work. Even for small problems with $m = 6$ clients and $n = 15$ share types, an optimal solution cannot be found within hours.

The banker's problem is a variant with integer variables of the market share problem [12]. An alternative viewpoint is to see it as a closest vector problem. Specifically we have to find a nonnegative integer combination of the vectors $(\underline{e}_i, \underline{e}_j, p_j \underline{e}_i)^T$ that is as close as possible to the vector $(\underline{d}, \underline{a}, \bar{c} \underline{d})^T$.

We now develop another approach allowing us to tackle the problem. Consider the problem of finding feasible solutions of equality constrained integer programs, or more specifically of finding points in the set

$$Z = \{x \in \mathbb{Z}_+^N : Ax = b\},$$

where $A \in \mathbb{Z}^{M \times N}$ and $b \in \mathbb{Z}^M$. For this problem, Aardal et al. [2] have recently developed a successful two-step approach based on basis reduction [3, 6, 11].

In step 1 they use basis reduction on the associated lattice $\begin{pmatrix} I & 0 \\ 0 & 1 \\ A & -b \end{pmatrix}$ of dimension $N + M + 1$ to construct an alternative representation of the feasible set Z of the form

$$Z = \{x : x = q + P\lambda, \lambda \in \mathbb{Z}^{N-M}, x \geq 0\}, \quad (2)$$

where q, P are integer, $Aq = b$, P is an integral basis of the null space of A and due to the basis reduction algorithm, q and the columns of P are "short". In step 2, they apply an MIP system to the reformulated set Z with $\lambda \in \mathbb{Z}^{N-M}$ as

the variables. For small instances of the banker's problem, this approach works whereas the original MIP approach does not.

However difficulties arise when we try to solve larger instances by this approach. The dimension of the lattice to be reduced is $(M + N + 1)$, and the basis reduction algorithm is $\mathcal{O}((M + N)^4)$ [3, 4]. For $M + N$ greater than 300, the basis reduction algorithm becomes too time consuming. To overcome this difficulty, we propose to take advantage of the special structure of Z . Specifically we consider sets of the form

$$Z = \{X \in \mathbb{Z}_+^{m \times n} : XA = C, BX = D\}, \quad (3)$$

where $A \in \mathbb{Z}^{n \times K}$, $B \in \mathbb{Z}^{L \times m}$, $C \in \mathbb{Z}^{m \times K}$ and $D \in \mathbb{Z}^{L \times n}$. Note that with $A = \begin{pmatrix} 1 & \cdots & 1 \\ p_1 & \cdots & p_n \end{pmatrix}^T$ and $B = \begin{pmatrix} 1 & \cdots & 1 \end{pmatrix}$, we obtain the system (1) arising in the banker's problem presented above.

For the system (3), the direct approach of Aardal et al. involves reducing a basis of dimension $(mn + K + L + 1)$ which is impractical. Instead, we work with the separate lattices $\mathcal{L}_A = \{\underline{x} \in \mathbb{Z}^n : \underline{x}A = 0\}$ and $\mathcal{L}_B = \{\underline{x} \in \mathbb{Z}^m : B\underline{x} = 0\}$, and we use the same approach to compute bases of $\mathcal{L}_A$ and $\mathcal{L}_B$, and use the resulting basis vectors to construct a reduced basis for the large lattice

$$\mathcal{L} = \{X \in \mathbb{Z}^{m \times n} : XA = 0, BX = 0\}.$$

In section 2 we present the class of problems to be studied, and we show how its structure allows us to construct an integral basis of the large lattice $\mathcal{L}$ from integral bases of the two small lattices. In section 3 we show that, when the bases of the small lattices are reduced bases, the basis constructed in Section 2 is also a reduced basis. In section 4 we present computational results for the banker's problem with the original reduced basis approach and with our decomposition approach. Finally in section 5 we conclude and discuss some open questions.

Notation 1.1 *Throughout the paper, we use the following notation:*

- An underlined letter $\underline{a}$ represents a vector (but of any dimension).
- a_i represents the i^{th} component of the vector $\underline{a}$.
- $\underline{a}^p$ represents the p^{th} vector of a collection of vectors.
- a_i^p represents the i^{th} component of the p^{th} vector of the collection.
- A big letter A or a double underlined letter $\underline{\underline{a}}$ represents a matrix.

2 Constructing an integral basis for $\mathcal{L}$

The problem we address in this section is to find an alternative representation of

$$\mathcal{L} = \{X \in \mathbb{Z}^{m \times n} : XA = \underline{0}, BX = \underline{0}\}, \quad (4)$$

with given matrices $A \in \mathbb{Z}^{n \times K}$, $K \leq n$ of rank K and $B \in \mathbb{Z}^{L \times m}$, $L \leq m$ of rank L .

Considering the second equation of (4), each column of X has to be a solution of the system $B\underline{x} = \underline{0}$. Let $\underline{\beta} = \begin{pmatrix} \underline{\beta}^1 & \dots & \underline{\beta}^{m-L} \end{pmatrix}$ denote an integral basis of the lattice $\mathcal{L}_B = \{x \in \mathbb{Z}^m : B\underline{x} = \underline{0}\}$. Thus each matrix $X \in \mathcal{L}$ can be written

$$X = \underline{\beta} \begin{pmatrix} \underline{\lambda}^1 \\ \vdots \\ \underline{\lambda}^{m-L} \end{pmatrix}, \quad (5)$$

with $\underline{\lambda}^1, \dots, \underline{\lambda}^{m-L} \in \mathbb{Z}^m$. Similarly we can use an integer basis of the lattice

$$\mathcal{L}_A = \{x \in \mathbb{Z}^n : \underline{x}A = \underline{0}\} \text{ to describe the vectors of } \mathcal{L}. \text{ Letting } \underline{\alpha} = \begin{pmatrix} \underline{\alpha}^1 \\ \vdots \\ \underline{\alpha}^{n-K} \end{pmatrix}$$

be such an integral basis, each solution X of (4) can be written

$$X = (\underline{\mu}^1 \dots \underline{\mu}^{n-K}) \underline{\alpha}, \quad (6)$$

where $\underline{\mu}^1, \dots, \underline{\mu}^{n-K} \in \mathbb{Z}^n$.

Now we can state the main result of this section.

Theorem 2.1 *The matrices $X \in \mathcal{L}$ are precisely the matrices of the form*

$$X = \underline{\beta} \Lambda \underline{\alpha}, \quad (7)$$

with $\Lambda \in \mathbb{Z}^{(m-L) \times (n-K)}$.

To prove this theorem, we need several intermediate results.

Observation 2.2 *For fixed $M \in \mathbb{Z}^{m \times m}$ and $N \in \mathbb{Z}^{n \times n}$ unimodular matrices, any $Y \in \mathbb{Z}^{m \times n}$ can be written as*

$$Y = M \Lambda N, \quad (8)$$

with $\Lambda \in \mathbb{Z}^{m \times n}$.

Proof: This is obvious by taking $\Lambda = M^{-1}Y N^{-1}$ which is integral. $\square$

For details about the Smith normal form of a matrix used in the next two lemmas, see the appendix and for example [10].

Lemma 2.3 Consider the lattice $\{\underline{x} \in \mathbb{Z}^n : C\underline{x} = \underline{0}\}$, where $C \in \mathbb{Z}^{m \times n}$, $m \leq n$ and $\text{rank}(C) = m$. Let $Y = (\underline{y}^1, \dots, \underline{y}^{n-m}) \in \mathbb{Z}^{n \times (n-m)}$ be a set of solutions of $C\underline{x} = \underline{0}$, then the following statements are equivalent

(i) The Smith normal form of Y is $\begin{pmatrix} I \\ 0 \end{pmatrix}$.

(ii) Y is an integer basis of the lattice.

Proof: This result appears to be known, but for completeness a proof is given in the appendix. $\square$

Lemma 2.4 A matrix $Y \in \mathbb{Z}^{n \times (n-m)}$ whose Smith normal form is $\begin{pmatrix} I \\ 0 \end{pmatrix}$ can be extended to a unimodular matrix.

Proof: There exist unimodular matrices M and N such that

$$Y = \begin{pmatrix} M_1 & M_2 \end{pmatrix} \begin{pmatrix} I \\ 0 \end{pmatrix} N,$$

with $M_1 \in \mathbb{Z}^{m \times n}$, $M_2 \in \mathbb{Z}^{m \times (m-n)}$. Thus, we have that

$$Y = M_1 N.$$

Now, consider the completion $\begin{pmatrix} Y & M_2 \end{pmatrix}$, we have that

$$\begin{pmatrix} Y & M_2 \end{pmatrix} = \begin{pmatrix} M_1 N & M_2 \end{pmatrix} = \begin{pmatrix} M_1 & M_2 \end{pmatrix} \begin{pmatrix} N & 0 \\ 0 & I \end{pmatrix}.$$

Thus,

$$|\det \begin{pmatrix} Y & M_2 \end{pmatrix}| = |\det M| |\det N| |\det I| = 1,$$

and M_2 completes Y as a unimodular matrix. $\square$

Proof of Theorem 2.1: By using the two lemmas, we know that we can complete $\underline{\alpha}$ and $\underline{\beta}$ into unimodular matrices. We denote these completions by $\underline{\underline{\alpha}}^c$ and $\underline{\underline{\beta}}^c$. Therefore, by Observation 2.2, $X \in \mathbb{Z}^{m \times n}$ can be expressed as

$$X = \begin{pmatrix} \underline{\beta} & \underline{\beta}^c \\ \underline{\underline{\alpha}} & \underline{\underline{\alpha}}^c \end{pmatrix} \begin{pmatrix} \Lambda_1 & \Lambda_2 \\ \Lambda_3 & \Lambda_4 \end{pmatrix} \begin{pmatrix} \underline{\underline{\alpha}} \\ \underline{\underline{\alpha}}^c \end{pmatrix},$$

or expanding

$$\begin{aligned} X &= \begin{pmatrix} \underline{\beta}\Lambda_1 + \underline{\underline{\beta}}^c\Lambda_3 & \underline{\beta}\Lambda_2 + \underline{\underline{\beta}}^c\Lambda_4 \end{pmatrix} \begin{pmatrix} \underline{\underline{\alpha}} \\ \underline{\underline{\alpha}}^c \end{pmatrix} \\ &= (\underline{\beta}\Lambda_1 + \underline{\underline{\beta}}^c\Lambda_3)\underline{\underline{\alpha}} + (\underline{\beta}\Lambda_2 + \underline{\underline{\beta}}^c\Lambda_4)\underline{\underline{\alpha}}^c. \end{aligned}$$

Recall that each solution of (4) can be written in both forms (5) and (6). From the necessary condition (6), we see that X must be a combination of the columns

of $\underline{\alpha}$. Since $\underline{\alpha}$ and $\underline{\alpha}^c$ are linearly independent, $\underline{\beta}\Lambda_2 + \underline{\beta}^c\Lambda_4 = \underline{0}$. But now, since the columns of $\underline{\beta}$ and $\underline{\beta}^c$ are linearly independent, we can conclude that, for each solution X of (4),

$$\Lambda_2 = \Lambda_4 = \underline{0}.$$

Identically, by taking the condition (5), we can conclude that

$$\Lambda_3 = \Lambda_4 = \underline{0}.$$

Hence,

$$X = \begin{pmatrix} \underline{\beta} & \underline{\beta}^c \end{pmatrix} \begin{pmatrix} \Lambda_1 & \underline{0} \\ \underline{0} & \underline{0} \end{pmatrix} \begin{pmatrix} \underline{\alpha} \\ \underline{\alpha}^c \end{pmatrix}.$$

and the set of solutions of (4) can be characterized as

$$X = \underline{\beta}\Lambda_1\underline{\alpha},$$

for a $\Lambda_1 \in \mathbb{Z}^{(m-L) \times (n-K)}$. Conversely, it is clear that every X of this form is a member of (4). $\square$

We have a general form of the vectors of (4). Now, we introduce some classical notions that allow us to simplify notation.

Definition 2.1 *Given a matrix Y , $\text{vec}(Y)$ denotes the column composed of the first column of Y followed by the second, etc.*

Definition 2.2 *The Kronecker product of two matrices $C \in \mathbb{R}^{m \times n}$ and $D \in \mathbb{R}^{p \times q}$ is*

$$C \otimes D = \begin{pmatrix} c_{11}D & \cdots & c_{1n}D \\ \vdots & \ddots & \vdots \\ c_{m1}D & \cdots & c_{mn}D \end{pmatrix},$$

belonging to $\mathbb{R}^{mp \times nq}$.

Proposition 2.5 *For matrices C , Y , D such that CYD exists,*

$$\text{vec}(CYD) = (C^T \otimes D)\text{vec}(Y).$$

Proof: See, for example, [5]. $\square$

Proposition 2.6 *$(\underline{\alpha}^T \otimes \underline{\beta})$ is a basis for the vectorized solutions of (4).*

Proof: Applying Proposition 2.5 to the matrix X given by Theorem 2.1, we obtain

$$\text{vec}(X) = (\underline{\alpha}^T \otimes \underline{\beta})\text{vec}(\Lambda_1).$$

$\square$

So in other words, a basis for $\mathcal{L}$ is obtained by taking the Kronecker product of the bases $\underline{\alpha}$ and $\underline{\beta}$ of $\mathcal{L}_A$ and $\mathcal{L}_B$ respectively.

Example Consider the following linear system with $m = 4$ and $n = 3$.

$$\left. \begin{array}{l} x_{11} + x_{21} + x_{31} = 0 \\ x_{12} + x_{22} + x_{32} = 0 \\ x_{13} + x_{23} + x_{33} = 0 \\ x_{14} + x_{24} + x_{34} = 0 \end{array} \right\} \text{Equations for the first dimension}$$

$$\left. \begin{array}{l} x_{11} + x_{12} + x_{13} + x_{14} = 0 \\ 16x_{11} + 57x_{12} + 23x_{13} + 66x_{14} = 0 \\ x_{21} + x_{22} + x_{23} + x_{24} = 0 \\ 16x_{21} + 57x_{22} + 23x_{23} + 66x_{24} = 0 \\ x_{31} + x_{32} + x_{33} + x_{34} = 0 \\ 16x_{31} + 57x_{32} + 23x_{33} + 66x_{34} = 0 \end{array} \right\} \text{Equations for the second dimension}$$

$$x_{ij} \in \mathbb{Z}, \quad \text{for } i = 1, \dots, 4 \text{ and } j = 1, \dots, 3.$$

The set of solutions form a lattice $\mathcal{L} = \{X \in \mathbb{Z}^{4 \times 3} : XA = 0, BX = 0\}$, with

$$A = \begin{pmatrix} 1 & 16 \\ 1 & 57 \\ 1 & 23 \\ 1 & 66 \end{pmatrix}, B = \begin{pmatrix} 1 & 1 & 1 \end{pmatrix} \text{ and } X = \begin{pmatrix} x_{11} & x_{12} & x_{13} & x_{14} \\ x_{21} & x_{22} & x_{23} & x_{24} \\ x_{31} & x_{32} & x_{33} & x_{34} \end{pmatrix}.$$

By Proposition 2.6 an integer basis can be found by computing the integer bases of the two lattices separately. First of all, we consider the lattice

$$\begin{pmatrix} 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = 0, \quad y_1, y_2, y_3 \in \mathbb{Z}.$$

It is readily verified that $\underline{\beta} = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \end{pmatrix}^T$ is an integer basis of the solutions (such a basis can be found by basis reduction). Similarly, for the system

$$\begin{pmatrix} y_1 & y_2 & y_3 & y_4 \end{pmatrix} \begin{pmatrix} 1 & 16 \\ 1 & 57 \\ 1 & 23 \\ 1 & 66 \end{pmatrix} = \begin{pmatrix} 0 & 0 \end{pmatrix}, \quad y_1, y_2, y_3, y_4 \in \mathbb{Z},$$

an integer basis $\underline{\alpha} = \begin{pmatrix} -1 & -4 & 2 & 3 \\ 10 & -3 & -11 & 4 \end{pmatrix}$ is obtained. Taking the Kronecker product of the two bases, we obtain that

$$(\underline{\alpha}^T \otimes \underline{\beta}) = \begin{pmatrix} -1 & 1 & 0 & -4 & 4 & 0 & 2 & -2 & 0 & 3 & -3 & 0 \\ 0 & -1 & 1 & 0 & -4 & 4 & 0 & 2 & -2 & 0 & 3 & -3 \\ 10 & -10 & 0 & -3 & 3 & 0 & -11 & 11 & 0 & 4 & -4 & 0 \\ 0 & 10 & -10 & 0 & -3 & 3 & 0 & -11 & 11 & 0 & 4 & -4 \end{pmatrix}^T$$

is a basis of the complete system.

3 A reduced basis of $\mathcal{L}$

In this section we show that, up to a reordering of the vectors, the basis constructed by computing the Kronecker product of reduced bases of the small lattices is itself reduced.

Observe that each row of the Kronecker product of the two matrices $\underline{\alpha}$ and $\underline{\beta}$ is the Kronecker product of a row (basis vector) of $\underline{\alpha}$ with a column of $\underline{\beta}$ (basis vector). We also need to take scalar products of such vectors.

Proposition 3.1 *Let $\underline{v}^1 := \underline{\gamma}^1 \otimes \underline{\delta}^1$ and $\underline{v}^2 := \underline{\gamma}^2 \otimes \underline{\delta}^2$ with $\underline{\gamma}^i \in \mathbb{R}^n$ and $\underline{\delta}^i \in \mathbb{R}^m$. Then*

$$\langle \underline{v}^1, \underline{v}^2 \rangle = \langle \underline{\gamma}^1, \underline{\gamma}^2 \rangle \langle \underline{\delta}^1, \underline{\delta}^2 \rangle,$$

where $\langle \underline{x}, \underline{y} \rangle$ denotes the scalar product.

Proof:

$$\begin{aligned} \langle \underline{v}^1, \underline{v}^2 \rangle &= \langle (\gamma_1^1 \underline{\delta}^1, \dots, \gamma_n^1 \underline{\delta}^1), (\gamma_1^2 \underline{\delta}^2, \dots, \gamma_n^2 \underline{\delta}^2) \rangle \\ &= \gamma_1^1 \gamma_1^2 \langle \underline{\delta}^1, \underline{\delta}^2 \rangle + \dots + \gamma_n^1 \gamma_n^2 \langle \underline{\delta}^1, \underline{\delta}^2 \rangle \\ &= \langle \underline{\gamma}^1, \underline{\gamma}^2 \rangle \langle \underline{\delta}^1, \underline{\delta}^2 \rangle. \end{aligned} \quad \square$$

We also need to work with reduced bases.

Definition 3.1 *Given r linearly independent vectors $(\underline{b}^j) \in \mathbb{R}^n$, $(\underline{b}^j)_{j=1, \dots, r}$ is a reduced basis if*

$$(i) \quad |\mu_{ij}| \leq \frac{1}{2} \quad \text{for all } i < j$$

$$(ii) \quad \|\hat{\underline{b}}^{j+1} + \mu_{j,j+1} \hat{\underline{b}}^j\|^2 \geq \frac{3}{4} \|\hat{\underline{b}}^j\|^2 \quad \text{for } 1 \leq j \leq r-1$$

where $\mu_{ij} = \frac{(\hat{\underline{b}}^j, \hat{\underline{b}}^i)}{(\hat{\underline{b}}^i, \hat{\underline{b}}^i)}$, and $(\hat{\underline{b}}^j)_{j=1, \dots, r}$ is the Gram-Schmidt orthogonalization of $(\underline{b}^j)_{j=1, \dots, r}$.

For the rest of this section, we study the integral basis $V = (\underline{v}^{pq}) = (\underline{\alpha}^p \otimes \underline{\beta}^q)$ constructed in Section 2, where $(\underline{\alpha}^p)_{p=1}^P$ and $(\underline{\beta}^q)_{q=1}^Q$ are now *reduced* bases with $P = n - K$ and $Q = m - L$. Let $(\hat{\underline{\alpha}}^p)_p$, $(\hat{\underline{\beta}}^q)_q$ denote the Gram-Schmidt orthogonalization of the bases $(\underline{\alpha}^p)_p$ and $(\underline{\beta}^q)_q$ respectively, with Gram-Schmidt coefficients denoted μ^α and μ^β respectively.

We now consider possible orderings of the set V of basis vectors. Throughout this section $(p, q) < (p', q')$ means that $p \leq p'$ and $q \leq q'$ and $(p, q) \neq (p', q')$.

Definition 3.2 *A total ordering of the basis V is called a monotone ordering if, for any distinct pair of vectors $\underline{v}^{pq}$ and $\underline{v}^{p'q'}$ with $(p, q) < (p', q')$, $\underline{v}^{pq}$ precedes $\underline{v}^{p'q'}$ in the ordering.*

Proposition 3.2 *For any monotone ordering of the integral basis V , $(\hat{\underline{v}}^{pq}) = (\hat{\underline{\alpha}}^p \otimes \hat{\underline{\beta}}^q)$ is the Gram-Schmidt orthogonalization of V .*

Proof: Clearly, $\underline{v}^{\hat{1}1} = \underline{v}^{11} = \underline{\alpha}^1 \otimes \underline{\beta}^1$. We then proceed by induction. We have

$$\begin{aligned}
\underline{\hat{\alpha}}^p \otimes \underline{\hat{\beta}}^q &= (\underline{\alpha}^p - \sum_{j=1}^{p-1} \mu_j^\alpha \underline{\hat{\alpha}}^j) \otimes (\underline{\beta}^q - \sum_{i=1}^{q-1} \mu_i^\beta \underline{\hat{\beta}}^i) \\
&= \underline{v}^{pq} - \underline{\alpha}^p \otimes (\sum_{i=1}^{q-1} \mu_i^\beta \underline{\hat{\beta}}^i) - (\sum_{j=1}^{p-1} \mu_j^\alpha \underline{\hat{\alpha}}^j) \otimes \underline{\beta}^q + \sum_{i=1}^{q-1} \sum_{j=1}^{p-1} \mu_j^\alpha \mu_i^\beta (\underline{\hat{\alpha}}^j \otimes \underline{\hat{\beta}}^i) \\
&= \underline{v}^{pq} - (\sum_{k=1}^p \lambda_k \underline{\hat{\alpha}}^k) \otimes (\sum_{i=1}^{q-1} \mu_i^\beta \underline{\hat{\beta}}^i) - (\sum_{j=1}^{p-1} \mu_j^\alpha \underline{\hat{\alpha}}^j) \otimes (\sum_{l=1}^q \gamma_l \underline{\hat{\beta}}^l) \\
&\quad + \sum_{i=1}^{q-1} \sum_{j=1}^{p-1} \mu_j^\alpha \mu_i^\beta (\underline{\hat{\alpha}}^j \otimes \underline{\hat{\beta}}^i) \\
&= \underline{v}^{pq} - \sum_{(i,j) < (p,q)} \tilde{\mu}_{ij} \underline{\hat{v}}_{ij}.
\end{aligned}$$

We only need to show now that this vector is orthogonal to all the others already considered. Indeed, we have

$$\begin{aligned}
\langle (\hat{\alpha}^p \otimes \hat{\beta}^q), \hat{v}^{ij} \rangle &= \langle \hat{\alpha}^p, \hat{\alpha}^i \rangle \langle \hat{\beta}^q, \hat{\beta}^j \rangle \\
&= 0 \quad \text{for all } (i, j) < (p, q).
\end{aligned}$$

This gives us a complete characterization of $\hat{\alpha}^p \otimes \hat{\beta}^q$ as the Gram-Schmidt orthogonalization of $\alpha^p \otimes \beta^q$. $\square$

Proposition 3.3 *For any monotone ordering of the basis V , condition (i) of Definition 3.1 of a reduced basis is satisfied.*

Proof: For any pair $(p_1, q_1), (p_2, q_2)$, with $(p_1, q_1) < (p_2, q_2)$,

$$\begin{aligned}
\mu_{(p_1, q_1)(p_2, q_2)} &= \frac{\langle \underline{v}^{p_2 q_2}, \underline{\hat{v}}^{p_1 q_1} \rangle}{\langle \underline{\hat{v}}^{p_1 q_1}, \underline{\hat{v}}^{p_1 q_1} \rangle} \\
&= \frac{\langle \underline{\alpha}^{p_2}, \underline{\hat{\alpha}}^{p_1} \rangle \langle \underline{\beta}^{q_2}, \underline{\hat{\beta}}^{q_1} \rangle}{\langle \underline{\hat{\alpha}}^{p_1}, \underline{\hat{\alpha}}^{p_1} \rangle \langle \underline{\hat{\beta}}^{q_1}, \underline{\hat{\beta}}^{q_1} \rangle} \quad \text{by Proposition 3.1} \\
&= \mu_{p_1 p_2}^\alpha \mu_{q_1 q_2}^\beta.
\end{aligned}$$

Now as $(\underline{\alpha}^j)$ and $(\underline{\beta}^i)$ are reduced bases, $|\mu_{p_1 p_2}^\alpha|, |\mu_{q_1 q_2}^\beta| \leq \frac{1}{2}$, and thus

$$\begin{aligned}
|\mu_{(p_1, q_1)(p_2, q_2)}| &= |\mu_{p_1 p_2}^\alpha| |\mu_{q_1 q_2}^\beta| \\
&\leq \frac{1}{4}.
\end{aligned}$$

$\square$

Now we need to refine the ordering of the basis V in order to satisfy condition (ii) of Definition 3.1.

Definition 3.3 A monotone ordering of the basis V is called regular if whenever $\underline{v}^{p_1 q_1}$ directly precedes $\underline{v}^{p_2 q_2}$ in the ordering and $(p_1 - p_2)(q_1 - q_2) < 0$, $\|\underline{v}^{p_1 q_1}\| \leq \|\underline{v}^{p_2 q_2}\|$.

Proposition 3.4 If the basis V has a regular monotone ordering, it is a reduced basis.

Proof: It suffices to show that the Condition (ii) of Definition 3.1 of a reduced basis is satisfied. Consider two pairs $(p_1, q_1), (p_2, q_2)$ where (p_1, q_1) directly precedes (p_2, q_2) in the ordering. There are two cases.

Case 1: $(p_2 - p_1)(q_2 - q_1) \geq 0$. Because the ordering is monotone, this implies that either $p_2 = p_1 + 1$ and $q_1 = q_2$ or $p_1 = p_2$ and $q_2 = q_1 + 1$. Suppose without loss of generality that $p_2 = p_1 + 1$ and $q_1 = q_2$.

$$\begin{aligned}
& \|\underline{v}^{p_1+1, q_1} + \mu_{(p_1, q_1)(p_1+1, q_1)} \underline{v}^{p_1 q_1}\|^2 \\
&= \|\hat{\underline{\alpha}}^{p_1+1} \otimes \hat{\underline{\beta}}^{q_1} + \frac{\langle \underline{\alpha}^{p_1+1}, \hat{\underline{\alpha}}^{p_1} \rangle \langle \hat{\underline{\beta}}^{q_1}, \hat{\underline{\beta}}^{q_1} \rangle}{\langle \hat{\underline{\alpha}}^{p_1}, \hat{\underline{\alpha}}^{p_1} \rangle \langle \hat{\underline{\beta}}^{q_1}, \hat{\underline{\beta}}^{q_1} \rangle} (\hat{\underline{\alpha}}^{p_1} \otimes \hat{\underline{\beta}}^{q_1})\|^2 \\
&= \|\hat{\underline{\alpha}}^{p_1+1} \otimes \hat{\underline{\beta}}^{q_1} + \mu_{p_1, p_1+1}^\alpha (\hat{\underline{\alpha}}^{p_1} \otimes \hat{\underline{\beta}}^{q_1})\|^2 \quad \text{since } \langle \hat{\underline{\beta}}^{q_1}, \hat{\underline{\beta}}^{q_1} \rangle = \langle \hat{\underline{\beta}}^{q_1}, \hat{\underline{\beta}}^{q_1} \rangle \\
&= \|(\hat{\underline{\alpha}}^{p_1+1} + \mu_{p_1, p_1+1}^\alpha \hat{\underline{\alpha}}^{p_1}) \otimes \hat{\underline{\beta}}^{q_1}\|^2 \\
&= \|\hat{\underline{\alpha}}^{p_1+1} + \mu_{p_1, p_1+1}^\alpha \hat{\underline{\alpha}}^{p_1}\|^2 \|\hat{\underline{\beta}}^{q_1}\|^2 \\
&\geq \frac{3}{4} \|\hat{\underline{\alpha}}^{p_1}\|^2 \|\hat{\underline{\beta}}^{q_1}\|^2 \quad \text{since } (\underline{\alpha}^j) \text{ is reduced} \\
&= \frac{3}{4} \|\hat{\underline{\alpha}}^{p_1} \otimes \hat{\underline{\beta}}^{q_1}\|^2 = \frac{3}{4} \|\underline{v}^{p_1 q_1}\|^2.
\end{aligned}$$

Case 2: $(p_2 - p_1)(q_2 - q_1) < 0$.

Suppose without loss of generality that $p_1 > p_2$ and $q_1 < q_2$. We show that $\mu_{(p_1, q_1)(p_2, q_2)} = 0$. Indeed,

$$\mu_{(p_1, q_1)(p_2, q_2)} = \frac{\langle \underline{\alpha}^{p_2}, \hat{\underline{\alpha}}^{p_1} \rangle \langle \hat{\underline{\beta}}^{q_2}, \hat{\underline{\beta}}^{q_1} \rangle}{\langle \hat{\underline{\alpha}}^{p_1}, \hat{\underline{\alpha}}^{p_1} \rangle \langle \hat{\underline{\beta}}^{q_1}, \hat{\underline{\beta}}^{q_1} \rangle}.$$

Since $\underline{\alpha}^{p_2} = \sum_{k=1}^{p_2} \lambda^k \hat{\underline{\alpha}}^k$, we have

$$\langle \underline{\alpha}^{p_2}, \hat{\underline{\alpha}}^{p_1} \rangle = \sum_{k=1}^{p_2} \lambda^k \langle \hat{\underline{\alpha}}^k, \hat{\underline{\alpha}}^{p_1} \rangle = 0 \quad \text{as } p_2 < p_1.$$

So

$$\begin{aligned}
\|\underline{v}^{p_2 q_2} + \mu_{(p_1, q_1)(p_2, q_2)} \underline{v}^{p_1 q_1}\|^2 &= \|\underline{v}^{p_2 q_2}\|^2 \\
&\geq \|\underline{v}^{p_1 q_1}\|^2 \quad \text{as the ordering is regular} \\
&> \frac{3}{4} \|\underline{v}^{p_1 q_1}\|^2.
\end{aligned}$$

□

We now present an algorithm to construct a regular monotone ordering of the basis V .

Definition 3.4 *Given a vector $\underline{v}^{pq} \in V$, the direct successors of $\underline{v}^{pq}$ are the vectors $\underline{v}^{p+1,q}$ and $\underline{v}^{p,q+1}$ (if they exist) and the predecessors are the vectors $\underline{v}^{p',q'}$ with $(p', q') < (p, q)$.*

Ordering Algorithm: RB is an ordered set of vectors from V .
 Q is a set of vectors from V .

Initialization: $RB := \{\underline{v}^{11}\}$
 $Q := \{\underline{v}^{12}, \underline{v}^{21}\}$.

Loop: While $Q \neq \emptyset$ do
 Choose $\underline{v} \in Q$ such that $\underline{v} = \arg \min\{\|\underline{x}\| : x \in Q\}$
 $RB := RB \cup \{\underline{v}\}$
 $Q := Q \setminus \{\underline{v}\}$
 Add to Q any direct successor of $\underline{v}$
 that has all its predecessors in RB .
 end

Return RB in order.

Proposition 3.5 *The ordering algorithm terminates with a regular monotone ordering of V .*

Proof: The monotonicity is obvious because of the criterion of selection of vectors entering Q . Indeed, if $(p, q) < (p', q')$, a vector $\underline{v}^{p'q'}$ cannot come before $\underline{v}^{pq}$ because it cannot enter Q until $\underline{v}^{pq}$ is in RB .

Now we have to prove the regularity. On each loop, we choose a vector $\underline{v} \in Q$. None of the remaining vectors in Q are direct successors of $\underline{v}$, and all the vectors that are added to Q during the loop are direct successors of $\underline{v}$. Therefore on the following loop, either we choose a vector $\underline{w}$ that is not a direct successor of $\underline{v}$ and thus $\|\underline{w}\| \geq \|\underline{v}\|$ because $\underline{v}$ was chosen ahead of $\underline{w}$ on the previous loop, or we choose a vector $\underline{w}$ that is a direct successor of $\underline{v}$. Therefore the condition of regularity is satisfied.

Finally we have to check that the algorithm terminates with $|RB| = |V|$. On each loop, exactly one vector is added to RB . So the algorithm has to stop. Suppose now that the algorithm terminates with $RB \subset V$. Select $p := \min\{p : \exists j \text{ with } \underline{v}^{pj} \notin RB\}$. Now select $q := \min\{q : \underline{v}^{pq} \notin RB\}$. Clearly, $\underline{v}^{pq}$ can be added to Q since $\underline{v}^{p-1,q} \in RB$ (or $p = 1$) and $\underline{v}^{p,q-1} \in RB$ (or $q = 1$) and thus all its direct predecessors are in RB . Therefore $Q \neq \emptyset$, and the algorithm cannot have ended, a contradiction. So $|RB| = |V|$.

Theorem 3.6 *RB is a reduced basis of the lattice $\mathcal{L}$.*

To end this section, we observe that stronger properties hold for the reduced basis $\mathcal{L}$ than for a general reduced basis. The The next two propositions give the results for a general lattice and for $\mathcal{L}$ respectively.

Proposition 3.7 *Let $(\underline{\gamma}^r)_r$ be a reduced basis of a lattice $\mathcal{L}_\gamma$ of $\mathbb{R}^n$. Then,*

- (i) $\|\underline{\gamma}^1\| \leq 2^{(r-1)/4} (\det \mathcal{L}_\gamma)^{\frac{1}{r}}$
- (ii) $\|\underline{\gamma}^1\| \leq 2^{(r-1)/2} \min\{\|\underline{w}\| : \underline{w} \in \mathcal{L}_\gamma, \underline{w} \neq \underline{0}\}$
- (iii) $\prod_{k=1}^r \|\underline{\gamma}^k\| \leq 2^{r(r-1)/4} \det(\mathcal{L}_\gamma).$

Proposition 3.8 *For the reduced basis $(\underline{v}^{pq})_{p=1, \dots, P, q=1, \dots, Q}$ constructed above,*

- (i) $\|\underline{v}^{11}\| \leq 2^{(P+Q-1)/4} (\det \mathcal{L})^{\frac{1}{PQ}}$
- (ii) $\|\underline{v}^{11}\| \leq 2^{(P+Q)/2} \min\{\|\underline{w}\| : \underline{w} \in \mathcal{L}, \underline{w} \neq \underline{0}\}$
- (iii) $\prod_{p=1}^P \prod_{q=1}^Q \|\underline{v}^{pq}\| \leq 2^{PQ(P+Q-2)/4} \det \mathcal{L}$

Proof: We only prove (iii). The other proofs are similar.
From Proposition 3.7,

$$\prod_{p=1}^P \|\underline{\alpha}^p\| \leq 2^{P(P-1)/4} \det(\mathcal{L}_A), \text{ and} \quad (9)$$

$$\prod_{q=1}^Q \|\underline{\beta}^q\| \leq 2^{Q(Q-1)/4} \det(\mathcal{L}_B). \quad (10)$$

For the lattice $\mathcal{L}$,

$$\begin{aligned} \prod_{p=1}^P \prod_{q=1}^Q \|\underline{v}^{pq}\| &= \prod_{p=1}^P \prod_{q=1}^Q \|\underline{\alpha}^p\| \|\underline{\beta}^q\| \\ &= \left(\prod_{p=1}^P \|\underline{\alpha}^p\| \right)^Q \left(\prod_{q=1}^Q \|\underline{\beta}^q\| \right)^P \\ &\leq 2^{QP(P-1)/4} (\det \mathcal{L}_A)^Q 2^{PQ(Q-1)/4} (\det \mathcal{L}_B)^P \quad (11) \\ &\leq 2^{PQ(P+Q-2)/4} \det \mathcal{L}, \quad (12) \end{aligned}$$

where the inequality (11) comes from (9) and (10), and (12) from the fact that $\det \mathcal{L} = (\det \mathcal{L}_A)^Q (\det \mathcal{L}_B)^P$. $\square$

4 The Banker's problem

4.1 Theoretical point of view

Here we show how the results of Section 2 can be used to tackle the banker's problem. Specifically we want to solve the problem

Problem 4.1

$$\begin{aligned}
\min \quad & \sum_{i=1}^m \frac{|s_i|}{d_i} \\
s.t. \quad & \sum_{i=1}^m x_{ij} = c_j & \text{for } j = 1, \dots, n \\
& \sum_{j=1}^n x_{ij} = d_i & \text{for } i = 1, \dots, m \\
& \sum_{j=1}^n p_j x_{ij} + s_i = d_i \frac{\sum_{j=1}^n c_j p_j}{\sum_{j=1}^n d_i} & \text{for } i = 1, \dots, m \\
& x \in \mathbb{Z}_+^{mn}, \quad s \in \mathbb{Z}^m.
\end{aligned} \tag{13}$$

We rescale the last set of equations so that all the coefficients and variables are integer. This yields

$$(\sum_{j=1}^n d_i) \sum_{j=1}^n p_j x_{ij} + \tilde{s}_i = d_i (\sum_{j=1}^n c_j p_j).$$

We can see that this system is of the form studied in Section 2 except for the additional variables s_i . However, the system still has the same structure. Indeed,

$$\text{taking } X = \begin{pmatrix} x_{11} & \cdots & x_{1n} & s_1 \\ \vdots & \ddots & \vdots & \vdots \\ x_{m1} & \cdots & x_{mn} & s_m \end{pmatrix}, A = \begin{pmatrix} 1 & \cdots & 1 & 0 \\ (\sum d_i)p_1 & \cdots & (\sum d_i)p_n & 1 \end{pmatrix}^T$$

and $B = \begin{pmatrix} 1 & \cdots & 1 \end{pmatrix}$, we get the system (13), with one more equation, namely $\sum_{i=1}^m s_i = 0$. However every solution of (13) automatically satisfy this equation.

4.2 Computational results

Our set of test instances are randomly generated feasible instances of the banker's problem with expected profits uniformly distributed between 5 and 105, based on a real instance. The supplies are uniformly distributed between 1 and 50, and the demands are randomly generated while maintaining $\sum_i d_i = \sum_j a_j$. In all the computations, LiDIA [7] was used to calculate the reduced bases and Cplex, version 6.6 to solve the MIPs, both running on a Sun Sparc Ultra 60. .

In Table 1, we compare the basis reduction approach with the direct MIP approach on five relatively small instances. The basis reduction approach permits us to solve all of the instances within a few seconds. On the other hand,

none of the instances were solved by the direct MIP approach because, for each instance, memory problems were encountered after more than an hour. In fact, the linear programming relaxations stay at zero throughout the enumeration tree, but good feasible solutions are found.

Sizes		Basis reduction			Direct MIP approach		
n	m	Time B&B	LB	UB	Time B&B	LB	UB
16	5	1	$1.5 \cdot 10^{-2}$		***	0	$1.5 \cdot 10^{-2}$
16	5	1	$2.5 \cdot 10^{-2}$		***	0	$2.5 \cdot 10^{-2}$
16	5	0	$8.9 \cdot 10^{-3}$		***	0	$8.9 \cdot 10^{-3}$
16	5	1	$1.4 \cdot 10^{-2}$		***	0	$1.4 \cdot 10^{-2}$
16	5	1	$1.7 \cdot 10^{-2}$		***	0	$1.7 \cdot 10^{-2}$
*** : Memory limit reached. Average time: 2 hours							

Table 1: Comparison of direct MIP or basis reduction

In Table 2, we compare the computation times of the basis composition approach using the product of two small reduced bases, and the direct basis reduction approach. For both approaches, Time part 1 is the time in seconds to construct the alternative MIP formulation by computing the integral basis of the homogeneous system, and Time B&B is the time spent to prove optimality with Cplex. For both approaches, the variable selection rule used in Cplex is “pseudo-reduced costs” as it leads to better and more stable results. For the direct basis reduction approach, we observe that the time required for basis reduction becomes a limiting factor as the problem gets bigger. On the other hand, using the composite basis approach, the time required in the MIP step increases, but only slightly. This suggests that the product vectors are perhaps not as short as those from direct basis reduction. But if we consider the total computation time, it is always more interesting to use the composite basis approach.

In Table 3 we run some larger instances with just the composite basis approach as the direct approach is now too time consuming. It appears that for the banker’s problem, the difficulty depends on the number m of clients. Instances with $m = 15$ are more difficult than instances with $m = 8$, even for large n . As m approaches 20, proving optimality becomes difficult.

5 Conclusions

The special structure of the linear integer system, or the corresponding lattice $\mathcal{L}$ is crucial in the results obtained here. It is natural to ask whether there are other structured linear integer programs for which a similar decomposition approach can be developed.

Theoretically, another interesting question arises. If $\underline{\alpha}^1$ and $\underline{\beta}^1$ are the shortest vectors in the lattices $\mathcal{L}_A$ and $\mathcal{L}_B$ respectively, $\underline{\alpha}^1 \otimes \underline{\beta}^1$ is the shortest vector

Sizes		Composite basis				Direct basis reduction			
n	m	Time part 1	Time B&B	LB	UB	Time part 1	Time B&B	LB	UB
22	9	1	1		$3.6 \cdot 10^{-2}$	552	2		$3.6 \cdot 10^{-2}$
22	9	1	2		$5 \cdot 10^{-2}$	527	1		$5 \cdot 10^{-2}$
22	9	1	3		$6 \cdot 10^{-2}$	632	1		$6 \cdot 10^{-2}$
22	9	1	1		$3.4 \cdot 10^{-2}$	554	1		$3.4 \cdot 10^{-2}$
22	9	1	1		$3.4 \cdot 10^{-2}$	519	1		$3.4 \cdot 10^{-2}$
30	12	1	12		$8.3 \cdot 10^{-2}$	4121	669		$8.3 \cdot 10^{-2}$
30	12	1	***	$5.5 \cdot 10^{-2}$	$6.7 \cdot 10^{-2}$	3955	7		$5.5 \cdot 10^{-2}$
30	12	1	36		$6.4 \cdot 10^{-2}$	4315	8		$6.4 \cdot 10^{-2}$
30	12	1	10		$6.6 \cdot 10^{-2}$	4003	16		$6.6 \cdot 10^{-2}$
30	12	1	11		$7.5 \cdot 10^{-2}$	4438	5		$7.5 \cdot 10^{-2}$
15	15	1	24		$2.1 \cdot 10^{-1}$	1110	10		$2.1 \cdot 10^{-1}$
15	15	1	438		$1.9 \cdot 10^{-1}$	1241	6		$1.9 \cdot 10^{-1}$
15	15	1	12		$2.1 \cdot 10^{-1}$	1415	23		$2.1 \cdot 10^{-1}$
15	15	1	128		$2.2 \cdot 10^{-1}$	1416	9		$2.2 \cdot 10^{-1}$
15	15	1	123		$3.1 \cdot 10^{-1}$	1083	6		$3.1 \cdot 10^{-1}$
60	8	2	11		$1.1 \cdot 10^{-2}$	10252	2		$1.1 \cdot 10^{-2}$
60	8	2	7		$9.5 \cdot 10^{-3}$	10482	6		$9.5 \cdot 10^{-3}$
60	8	2	17		$1.3 \cdot 10^{-2}$	11223	7		$1.3 \cdot 10^{-2}$
60	8	2	7		$1.4 \cdot 10^{-2}$	10965	6		$1.4 \cdot 10^{-2}$
*** = more than one hour									

Table 2: Comparison of constructed and direct basis reduction

in $\mathcal{L}$ among all the vectors of the form $\underline{\alpha} \otimes \underline{\beta}$ with $\underline{\alpha} \in \mathcal{L}_A$ and $\underline{\beta} \in \mathcal{L}_B$. How close is this vector to being the shortest vector in $\mathcal{L}$?

More generally a satisfying explanation of when the approach of Aardal et al. [2] is likely to be effective is still lacking.

Another question is more computational. Step 1 of the Aardal et al. approach leads to a reduced basis and thus an ordered set of columns. However the ordering plays no obvious role in the solution of the mixed integer program in Step 2. Some limited experiments [1, 8] have been carried out using priorities in an attempt to decide whether to branch on the first or last vector in the basis, but more seems to be needed. It would also be interesting to test whether the composite basis for $\mathcal{L}$ is a shorter basis than that produced by the direct reduced basis approach.

Acknowledgements

We would like to thank Paul Vandooren for suggesting appropriate notation, and the structure of the proof of Theorem 2.1.

Sizes		Composite basis approach			
n	m	Time B&B	LB	UB	GAP
50	15	156	$6.4 \cdot 10^{-2}$		
50	15	140	$4.4 \cdot 10^{-2}$		
50	15	57	$5.2 \cdot 10^{-2}$		
50	15	***	$5.37 \cdot 10^{-2}$	$5.48 \cdot 10^{-2}$	2%
50	15	118	$4.5 \cdot 10^{-2}$		
30	20	***	$2 \cdot 10^{-1}$	$6.6 \cdot 10^{-1}$	69%
30	20	***	$8.54 \cdot 10^{-2}$	$6.8 \cdot 10^{-1}$	88%
30	20	1177	$2.2 \cdot 10^{-1}$		
30	20	409	$1.1 \cdot 10^{-1}$		
30	20	***	$2 \cdot 10^{-1}$	$2.4 \cdot 10^{-1}$	17%
*** = more than one hour					

Table 3: Critical sizes of the problems solved by this method

References

- [1] K. Aardal, R. E. Bixby, C. A. J. Hurkens, A. K. Lenstra, and J. W. Smeltink. Market split and basis reduction: towards a solution of the Cornuéjols-Dawande instances. *INFORMS Journal on Computing*, 12(3):192–202, 2000.
- [2] K. Aardal, C. Hurkens, and A. K. Lenstra. Solving a system of diophantine equations with lower and upper bounds on the variables. Technical Report UU-CS-1998-36, Departement of Computer Science, Utrecht University, 1998.
- [3] H. Cohen. *A Course in Computational Algebraic Number Theory*. Springer-Verlag, 1996.
- [4] A. Joux. *La réduction de réseaux en cryptographie*. PhD thesis, Ecole Polytechnique, Paris, 1998.
- [5] P. Lancaster and M. Tismenetsky. *The theory of matrices*. Computer Science and Applied Mathematics. Academic Press Inc., Orlando, Fla., second edition, 1985.
- [6] A. K. Lenstra, H.W. Lenstra, and L. Lovász. Factoring polynomials with rational coefficients. *Mathematische Annalen*, 261:515–534, 1982.
- [7] LiDIA. *A library for computational number theory*. TH Darmstadt/ Universität des Saarlandes, Fachbereich Informatik, Institut für Theoretische Informatik, <http://www.informatik.th-darmstadt.de/pub/TI/LiDIA>.

- [8] Q. Louveaux. Résolution de problèmes d'optimisation en nombres entiers par un algorithme de réduction de base dans les treillis. Mémoire de la Faculté des Sciences Appliquées, Université catholique de Louvain, Juin 1999.
- [9] G. L. Nemhauser and L. A. Wolsey. *Integer and Combinatorial Optimization*. Wiley, 1988.
- [10] M. Newman. *Integral Matrices*, volume 45 of *Pure and Applied Mathematics, a Series of Monographs and Textbooks*. Academic Press, 1972.
- [11] A. Schrijver. *Theory of linear and integer programming*. Wiley, Chichester, 1986.
- [12] H. P. Williams. *Model Building in Mathematical Programming*. Wiley & Sons, Chichester, 1993.

A Appendix

In Proposition A.1, we present the basic results about Smith Normal form, and we prove Theorem 2.1 characterizing when a set of integer vectors forms an integral basis of the null space of a matrix.

Proposition A.1 (Smith Normal Form) *Let A be an integer matrix from $\mathbb{Z}^{m \times n}$ with $m \geq n$ and $\text{rank}(A) = r$. There exist unimodular matrices $U \in \mathbb{Z}^{m \times m}$ and $V \in \mathbb{Z}^{n \times n}$ such that*

$$UAV = \begin{pmatrix} s_1 & & & & & \\ & \ddots & & & & \\ & & s_r & & & \\ & & & 0 & & \\ & & & & \ddots & \\ & & & & & 0 \\ & & & \underline{0} & & \end{pmatrix},$$

with $s_1 | s_2 | \dots | s_r$. For $1 \leq k \leq r$, $\prod_{i=1}^k s_i$ is the gcd of all $k \times k$ submatrices of A .

Proposition A.2 *Consider the system $C\underline{x} = \underline{0}$, where $C \in \mathbb{Z}^{m \times n}$ with $m \leq n$ and $\text{rank}(C) = m$. Let $Y \in \mathbb{Z}^{n \times (n-m)}$ be a set of solutions of $C\underline{x} = \underline{0}$, then the following statements are equivalent*

- (i) *The Smith normal form of Y is $\begin{pmatrix} I \\ 0 \end{pmatrix}$.*
- (ii) *Y is an integer basis of all the integer solutions of $C\underline{x} = \underline{0}$.*

Proof: (ii) $\Rightarrow$ (i) As $\text{rank}(Y) = n - m$, there exist two unimodular matrices U and V such that

$$Y = U \begin{pmatrix} s_1 & & & \\ & \ddots & & \\ & & s_{n-m} & \\ & & & \underline{0} \end{pmatrix} V,$$

with $U \in \mathbb{Z}^{n \times n}$ and $V \in \mathbb{Z}^{(n-m) \times (n-m)}$ and $s_{n-m} \geq 1$. As V is unimodular, its inverse exists, and we can write

$$YV^{-1} = U \begin{pmatrix} s_1 & & & \\ & \ddots & & \\ & & s_{n-m} & \\ & & & \underline{0} \end{pmatrix}.$$

Then

$$U \begin{pmatrix} s_1 & & & \\ & \ddots & & \\ & & s_{n-m} & \\ & & & 0 \end{pmatrix} = (s_1 \underline{u}^1 \quad \cdots \quad s_{n-m} \underline{u}^{n-m}) \quad (14)$$

is also an integer basis of the integer solutions of $C\underline{x} = \underline{0}$ since it is obtained by multiplying a basis by the unimodular matrix V . So $s_{n-m} \underline{u}^{n-m}$ is a solution of $C\underline{x} = \underline{0}$ and $\underline{u}^{n-m}$ is an integral solution as well. But it only lies in the basis (14) if $s_{n-m} = 1$. Hence, $s_{n-m} = 1$ and $s_i = 1$ for $i = 1, \dots, n - m$ by the divisibility property of the Smith normal form.

(i) $\Rightarrow$ (ii): There exists a basis of all the integer solutions of $C\underline{x} = \underline{0}$. Let us call it W . By hypothesis, we can find two unimodular matrices $U \in \mathbb{Z}^{n \times n}$ and $V \in \mathbb{Z}^{(n-m) \times (n-m)}$ such that

$$UYV = \begin{pmatrix} I_{n-m} \\ \underline{0}_{m \times (n-m)} \end{pmatrix}. \quad (15)$$

As each column of Y is a solution of $C\underline{x} = \underline{0}$, each of them is an integer combination of the columns of W . Thus, there exists a nonsingular integer matrix $\Lambda \in \mathbb{Z}^{(n-m) \times (n-m)}$ such that

$$Y = W\Lambda. \quad (16)$$

To prove that Y is an integer basis of the solutions of $C\underline{x} = \underline{0}$, we just have to prove that this matrix Λ is unimodular. We now show that ΛV is unimodular. Let S be its Smith normal form, so

$$P\Lambda VQ = S,$$

with P and Q unimodular. Rewriting (15), we now have

$$UW\Lambda V = UW P^{-1} S Q^{-1} = \begin{pmatrix} I \\ \underline{0} \end{pmatrix}.$$

As U and Q^{-1} are unimodular, the Smith normal form of $WP^{-1}S$ is $\begin{pmatrix} I \\ \underline{\underline{0}} \end{pmatrix}$. In $WP^{-1}S$, all the elements of the last column are multiples of the last element of the diagonal of S , and thus $s_{n-m} = 1$. Therefore the Smith normal form of ΛV is $\begin{pmatrix} I \\ \underline{\underline{0}} \end{pmatrix}$, and ΛV is unimodular. Therefore Λ is unimodular and Y is a basis. $\square$