
Column Basis Reduction and Decomposable Knapsack Problems

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OUTLINE OF THE PRESENTATION

- ① Basis reduction
- ② Column Basis Reduction (CBR)
 - CBR in Range Space
 - CBR in Null Space
 - CBR with rhs reduction
- ③ Branching on a constraint
- ④ Decomposable knapsack problems

WHAT IS COLUMN BASIS REDUCTION?

Given integral matrix A , compute unimodular U s.t.

- columns of AU have small Euclidean norm
- and nearly orthogonal (angle between any column and the linear space spanned by other columns is ≥ 60 degrees)

Methods are:

- LLL-reduction by Lenstra, Lenstra and Lovasz
- Korkhine-Zolotarev (KZ) reduction

$$A = \begin{pmatrix} 289 & 18 \\ 466 & 29 \\ 273 & 17 \end{pmatrix}, U = \begin{pmatrix} 1 & -15 \\ -16 & 241 \end{pmatrix}, AU = \begin{pmatrix} 1 & 3 \\ 2 & -1 \\ 1 & 2 \end{pmatrix}$$

THE OUTLINE OF LLL-BR METHOD

Given matrix $B \in \mathbb{R}^{m \times n}$ with m independent columns,
 $\mathbb{L}(B) = \{Bv \mid v \in \mathbb{Z}^n\}$ Basically, set of all integral combinations
of the columns of B .

- Finding shortest vector in L is believed to be NP-complete.
- LLL lattice basis reduction is approximation algorithm in polynomial time

The Algorithm:

1. Find Gram-Schmidt basis of columns of B . Let them be b_1^*, \dots, b_m^* .

$$b_1^* = b_1, b_k^* = b_k - \sum_{j=1}^{k-1} \mu_{kj} b_j^*, \mu_{kj} = \frac{b_k \cdot b_j^*}{b_j^* \cdot b_j^*}$$

The Algorithm: Cont.

2. b_1, b_2, \dots, b_m is reduced if $|\mu_{kj}| \leq 1/2$ for $1 \leq j < k \leq m$ and

$$b_k^* \cdot b_k^* \geq (\alpha - \mu_{kk-1}^2) b_{k-1}^* \cdot b_{k-1}^* \quad (**)$$

for $1 < k \leq m$ and $1/4 < \alpha \leq 1$

We say that b_k is size-reduced if $|\mu_{kj}| \leq 1/2$ for $1 < j \leq k$

- 3 Let $b_k \leftarrow b_k - \lceil \mu_{kk-1} \rceil b_{k-1}$.

If (**) holds, size-reduce b_k completely, do

$b_k \leftarrow b_k - \lceil \mu_{kj} \rceil b_j$ for $j = k - 2, \dots, 1$, increment k

- 4 Otherwise swap b_k and b_{k-1} , decrement k .

COLUMN BASIS REDUCTION (CBR)

CBR in the Range Space:

- Change

$$\begin{aligned} (IP) b' &\leq Ax \leq b \\ x &\in \mathbb{Z}^n \end{aligned}$$

- to

$$\begin{aligned} (\widetilde{IP}) b' &\leq AUy \leq b \\ y &\in \mathbb{Z} \end{aligned}$$

- The relation between x and y is $Uy = x, y = U^{-1}x$

Example: The infeasible problem,

$$106 \leq 21x_1 + 19x_2 \leq 113$$

$$0 \leq x_1, x_2 \leq 6$$

$$x_1, x_2 \in \mathbb{Z}$$

- Branching on either variable will create at least 5 feasible nodes.

Apply CBR:

$$106 \leq -2y_1 + 7y_2 \leq 113$$

$$0 \leq -y_1 - 6y_2 \leq 6$$

$$0 \leq y_1 + 7y_2 \leq 6$$

$$y_1, y_2 \in \mathbb{Z}$$

- Branching on either variable y_1 would create 4 feasible branches, but branching on y_2 immediately proves infeasibility.

CBR in the Null Space: Let $A_1x = b_1$ be a system of equalities in $b' \leq Ax \leq b$.

- compute an integral matrix B_1 and an integral vector x_0 such that $\{x \in \mathbb{Z}^n | A_1x = b_1\} = \{B_1\lambda + x_0 | \lambda \in \mathbb{Z}^{n-m_1}\}$
- $A_1x_0 = b_1$ and $A_1B_1 = 0$.
- B_1 and x_0 computed with Hermite Normal Form computation.
- Substitute $B_1\lambda + x_0$ for x in original problem and apply CBR in range space.

CBR in the Null Space:

- Change

$$Ax = b$$

$$l \leq x \leq u$$

$$x \in \mathbb{Z}^n$$

- to

$$l \leq B_1\lambda + x_0 \leq u$$

$$\lambda \in \mathbb{Z}^{n-m_1}$$

- And apply CBR in Range Space.

CBR with Right Hand Side Reduction:

- On several instances RHS reduction gives better reformulations. Write IP as

$$\begin{aligned} Dx &\leq f \\ x &\in \mathbb{Z}^n \end{aligned}$$

- Reformulate as

$$\begin{aligned} (DU)y &\leq f - Dx_r \\ y &\in \mathbb{Z}^n \end{aligned}$$

- x_r is calculated (with Babai's algorithm) s. t. Dx_r is an approximation to a **closest** vector to f in the $\mathbb{L}(D)$.

BRANCHING ON A CONSTRAINT

Given P and integral vector c , the width of P in the direction of c is

- $width(c, P) = \max\{cx \mid x \in P\} - \min\{cx \mid x \in P\}$
- Branching on cx means creating
 $cx = \lceil min \rceil, cx = \lceil min \rceil + 1, \dots, cx = \lfloor max \rfloor$ branches
- If $[min, max]$ does not contain any integer then P is infeasible
- If c is unit vector then it is regular x branching.

Example

$$106 \leq 21x_1 + 19x_2 \leq 113$$

$$0 \leq x_1, x_2 \leq 6$$

$$x_1, x_2 \in \mathbb{Z}$$

- Branching on $x_1 + x_2$ will immediately prove the problem is infeasible, since $\min = 5.04$ and $\max = 5.94$.

T+1-LEVEL DECOMPOSIBLE KNAPSACK PROBLEM

Assume

1. Given matrix $P \in \mathbb{Z}^{t \times n}$, row vectors $a, r \in \mathbb{Z}^n$, a column vector $u \in \mathbb{Z}_{++}^n$. u might have components equal to $+\infty$. and p_i represent a row of P .
2. Given a row vector $M \in \mathbb{Z}_{++}^t$ with $M_1 > M_2 > \dots > M_t$
3. $a = MP + r$

Definition: The feasibility problem

$$\beta' \leq ax \leq \beta$$

$$0 \leq x \leq u$$

$$x \in \mathbb{Z}^n$$

is called **t+1-level decomposable knapsack problem**.

2-LEVEL DECOMPOSIBLE KNAPSACK PROBLEM

$$\begin{aligned}\beta' &\leq ax \leq \beta \\ 0 &\leq x \leq u \\ x &\in \mathbb{Z}^n\end{aligned}$$

where

- $a = pM + r$ with $p \in \mathbb{Z}_+^n, r \in \mathbb{Z}^n$; M large
- β, β' are chosen, so the instance is LP-feasible.
- IP-infeasibility can be proven by branching px
- The previous example is 2-level decomposable knapsack problem with $p = (1, 1), r = (1, -1), u = (6, 6), M = 20,$
 $a = pM + r = (21, 19)$
- Remember, branching on $px = x_1 + x_2$ proves infeasibility at root node

REFORMULATION WITH CBR IN RANGE SPACE

Calculate U such that

$$A = \begin{pmatrix} a \\ I \end{pmatrix} = \begin{pmatrix} pM + r \\ I \end{pmatrix} U \text{ is reduced.}$$

Theorem 1: If M is sufficiently large then

$$pU = (0, 0, \dots, \alpha) \text{ for some } \alpha \in \mathbb{Z} \setminus \{0\}$$

Corollary: $Uy = x \Rightarrow pUy = px \Rightarrow \alpha y_n = px$

\Rightarrow branching on y_n proves infeasibility.

Sufficiently large means

- If LLL (Lenstra, Lenstra, Lovasz) reduction is used,
 $M > 2^{n+1} \|p\| \|r\|^2.$
- If KZ (Korkhine-Zolotarev) reduction is used,
 $M > \sqrt{n} \|p\| \|r\|^2.$

Strength of the BR algorithm is represented by c_n .

$$c_n(LLL) = 2^{n+1} \text{ and } c_n(KZ) = \sqrt{n}.$$

If c_n is smaller, the columns are more reduced.

CBR IN T+1-LEVEL KNAPSACK PROBLEMS

Let $A = \begin{pmatrix} a \\ I \end{pmatrix} = \begin{pmatrix} MP + r \\ I \end{pmatrix}$, $\tilde{P} = PU$ and \tilde{p}_i for the rows of \tilde{P} .

Theorem2: *There exists functions f_1, f_2, \dots, f_t with:*

(1) *Given $s \in \mathbb{Z}^t$ with $1 \leq s_t \leq \dots \leq s_1 \leq n - t$*

If

$$M_i > f_i(M_{i+1}, \dots, M_t, s_i, P, r, c_n) \quad (i = 1, \dots, t) \quad (*)$$

then

$$\tilde{p}_{i,1:s_i} = 0 \quad (i = 1, \dots, t)$$

(2) *There is M with*

$$\text{size}(M) = \text{poly}(\text{size}(P), \text{size}(r), \text{size}(c_n), n)$$

that satisfies ().*

What Theorem 2 says:

- If M_1 is sufficiently large compared to M_2, \dots, M_t , then $p_1 M_1$ contributes the most to the length of a .
- If M_2 is sufficiently large compared to M_3, \dots, M_t , then $p_2 M_2$ contributes the second most to the length of a and so on.
- To reduce the length of the columns of A , zero out many components of p_1 , fewer components of p_2 and so on.
- Let $n = 10, t = 4, s_1 = 6, s_2 = s_3 = 5, s_4 = 4$, the matrix \tilde{P} :

$$\begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & * & * & * & * & * \\ 0 & 0 & 0 & 0 & * & * & * & * & * & * \end{pmatrix}$$

BRANCHING IN CBR AND B & B

Since $\tilde{p}_i y = p_i x$ ($i = 1, \dots, t$)

- Branching on y_n, \dots, y_{s_1+1} in CBR in Range space has the same effect as branching on $p_1 x$ in original problem.
- Branching on $y_{s_1+1}, \dots, y_{s_2}$ in CBR in Range space has the same effect as branching on $p_2 x$ in original problem.

Thus, CBR

- takes the unknown **dominant** branching combinations
- transforms them into individual variables s.t. y_n has more significance than y_{n-1}

Example:

- An instance of 3-level knapsack problem with
 $n = 11, t = 2, u = e, M_1 = 220, M_2 = 11, \beta = 5661, \beta' = 5660$
- $p_1 = (2, 3, 5, 7, 8, 8, 9, 10, 10, 11, 11),$
 $p_2 = (7, 6, 5, 3, 3, 6, 4, 2, 6, 4, 7),$
 $r = (3, -1, 1, 1, 1, 3, -1, -1, 1, 1, -1)$
- $5660 \leq 520x_1 + 725x_2 + 1156x_3 + 1574x_4 + 1794x_5 + 1829x_6 +$
 $2023x_7 + 2221x_8 + 2267x_9 + 2465x_{10} + 2496x_{11} \leq 5661$
 $x_i \in \{0, 1\}$

Example: Cont.

- It is reasonably hard with pure B&B on x_i variables. Few hundred nodes to prove infeasibility
- Branching on p_1x is very useful. $24.30 \leq p_1x \leq 25.34$
- Then, branching on p_2x proves infeasibility, since $14.02 \leq p_2x \leq 14.93$
- After CBR, branch on $y_{11} = 25$ that results $24.30 \leq y_2 \leq 25.34$, then branching on y_{10} proves infeasibility since $14.02 \leq y_{10} \leq 14.93$

COMPUTATIONAL RESULTS

If maximization (or minimization) problem:

$$\max cx$$

$$b' \leq Ax \leq b$$

$$x \in \mathbb{Z}^n$$

Apply CBR to $\begin{pmatrix} c \\ A \end{pmatrix}$

The CBR is successful in the following problems:

- Subset sum problems
- Strongly correlated knapsack problems
- The market share problems