

Lift & Project

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Have seen: strengthening LP relaxations by adding strong valid inequalities.

E.g.: Knapsack (\rightarrow Cornu & Shmoy) adding 10 new inequalities reduces gap to ≤ 2

in fact: the gap of the strengthened LP is ≈ 2 (see Carr et al. '03)

There is of course a PTAJ for max- and min Knapsacks. Having an LP with gap 2 not impressive.

Today: Lift-and Project hierarchies

- strengthening weak LPs systematically

- Idea: (i) find valid relaxation of given problem in higher-dim space "lift"

- (ii) projection to native space is tighter relaxation of problem "project"

There are many Lift & Project hierarchies, we focus on Lasserre that produces SDP relaxations.

Positive Def. Matrices

Symmetric matrix M is positive semidef (psd)

$$\text{i)} \quad X^T M X \geq 0 \quad \forall X$$

Equiv: (i) any principal submatrix has

$$\text{det} \geq 0$$

(ii) \exists vect. v_i s.t. $M_{ij} = \langle v_i, v_j \rangle$

Round- t Lasserre relax

linear relax of dicing problem

Given $K = \{x \in \mathbb{R}^n : Ax \geq b\}$ A mxn mat.

Introduce new variables $y_I \quad \forall I \subseteq [n], |I| \leq 2t+1$

Intuition think of y_I as representing event

$$\bigwedge_{i \in I} (x_i = 1)$$

want: $y_\emptyset = 1$ and $y_{\{i\}} = x_i$

$$\boxed{\text{Last}_t(K)} \quad \boxed{(y_{I \cup J})_{|I|, |J| \leq t}} \stackrel{(M_t(y))}{\geq 0}$$

$$\left(\sum_{i \in [n]} A_{e,i} \cdot y_{I \cup J \cup \{i\}} - b_e y_{I \cup J} \right)_{|I|, |J| \leq t} \stackrel{\forall e \in [m]}{\leq 0}$$

$$(M_t^e(y)) \quad y_\emptyset = 1$$

consisting

$\boxed{M_t(y)}$ moment matrix $\boxed{M_t^e(y)}$ moment matrix
of slack

enforcing constraints

$$\text{Last}_t^{\text{proj}}(K) = \{(y_{\{1\}}, \dots, y_{\{n\}}) : y \in \text{Last}_t(K)\}$$

projection of $\text{Last}_t(K)$ onto original space.

Note: set of PSD matrices of certain dim
norm polyhedral cone

$\Rightarrow \text{Las}_t(K)$ is polyhedron

Separation \Rightarrow computing neg. EV
of involved mat if there exists

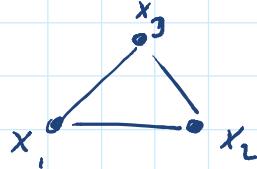
Can be done in polytime

\Rightarrow can (approx.) optimize over $\text{Las}_t(K)$
in $n^{O(t)} m^{O(1)}$ time.

Ex: Independent set $G = (V, E)$

$$K = \{x \in \mathbb{R}^n : x_u + x_v \leq 1 \quad \forall u, v \in E \\ 0 \leq x \leq 1\}$$

$$M_1(y) = \begin{pmatrix} 1 & y_1 & y_2 & y_3 \\ y_1 & 1 & y_1 & y_2 \\ y_2 & y_1 & 1 & y_3 \\ y_3 & y_2 & y_3 & 1 \end{pmatrix}$$



$$M_1^{12} = \begin{pmatrix} 1 - y_1 - y_2 & y_1 - y_1 - y_{12} & \dots \\ \vdots & y_1 - y_1 - y_{12} & \ddots \end{pmatrix}$$

Basic Properties

(P1) $\text{conv}(K \cap \{0,1\}^n) \subseteq \text{Las}_t^{\text{proj}}(K) \quad \forall t \geq 1$

give $x \in K \cap \{0,1\}^n$, w.l.

$$y_I = \begin{cases} 1 & : x_i = 1 \quad \forall i \in I \\ 0 & : \text{otherwise} \end{cases}$$

$$\text{note: } (M_t(\gamma))_{I,J} = \gamma_{I \cup J} = \gamma_I \cdot \gamma_J = (\gamma \gamma^T)_{I,J}$$

$$\text{and } \gamma \gamma^T \succeq 0$$

$$\text{Similarly } M_t^e(\gamma) \succ 0 \quad \forall e$$

So: γ feas. for $\text{Lass}_t(K)$ ✓

(P2) $\gamma \in \text{Lass}_t(K)$ Then

② $\underbrace{(\gamma_1, \dots, \gamma_n)}_{\bar{\gamma}} \in K$

$$M_t^e(\gamma)_{\emptyset, \emptyset} : \sum_{i=1}^n \alpha_i \gamma_i - b_e \gamma_{\emptyset} \stackrel{\text{psd}}{=} 0$$

so: $\bar{\gamma}$ sat. all constraints

also: consider principal subm. def.

by ϕ and Σ of $M_t(\gamma)$:

$$\begin{bmatrix} \gamma_\emptyset & \gamma_I \\ \gamma_I & \gamma_I \end{bmatrix} = \Gamma$$

$$\text{det}(\Gamma) = \gamma_I (1 - \gamma_I) \stackrel{\text{psd}}{\geq} 0$$

$$\Rightarrow 0 \leq \gamma_I \leq 1 \quad \forall I$$

③ $\gamma_J = \gamma_I$ for $I \subseteq J$, $|I| \leq |J| \leq t$

Consider submatrix for I, J

$$\begin{aligned} \left| \begin{bmatrix} \gamma_I & \gamma_J \\ \gamma_J & \gamma_J \end{bmatrix} \right| &= \gamma_I \gamma_J - \gamma_J^2 \\ &= \gamma_J (\gamma_I - \gamma_J) \geq 0 \end{aligned}$$

$$\Rightarrow \gamma_J \leq \gamma_I$$

Lemma 1 $y \in \text{Last}_t(\kappa) \quad i \in [n], 0 < y_i < 1$
 $\implies y \in \text{Conv} \{ z \in \text{Last}_{t-1}(\kappa) \mid z_i \in \{0, 1\}\}$

Pf: Define

$$z_I^{(1)} = \frac{y_{I \cup \{i\}}}{y_i} \quad z_I^{(0)} = \frac{y_I - y_{I \cup \{i\}}}{1 - y_i}$$

$$\implies y = y_i z_I^{(1)} + (1 - y_i) z_I^{(0)}$$

$$\text{and } z_i^{(1)} = \frac{y_i}{y_i} = 1, z_i^{(0)} = \frac{y_i - y_i}{1 - y_i} = 0$$

To Do show psd constraints

$$y \in \text{Last}_t(\kappa) \implies M_t(y) \succeq 0$$

$$\implies \exists v_I \text{ s.t. } M_t(y)_{IJ} = \langle v_I, v_J \rangle$$

$$\implies \sum_{\substack{I, J \subseteq [n] \setminus i \\ |I|, |J| < t}} \left\langle \frac{v_{I \cup i}}{\sqrt{y_i}}, \frac{v_{J \cup i}}{\sqrt{y_i}} \right\rangle = \frac{1}{y_i} \cdot y_{I \cup J \cup i} = z_{I \cup J}^{(1)}$$

$$\implies M_{t-1}(z^{(1)}) \succeq 0$$

$$\begin{aligned} \text{Also: } & \left\langle \frac{v_I - v_{I \cup i}}{\sqrt{1 - y_i}}, \frac{v_J - v_{J \cup i}}{\sqrt{1 - y_i}} \right\rangle \\ &= \frac{v_I v_J - v_I v_{J \cup i} - v_J v_{I \cup i} + v_{I \cup i} v_{J \cup i}}{1 - y_i} \end{aligned}$$

$$= \frac{y_{I \cup J} - y_{I \cup J \cup i}}{1 - y_i} = z_{I \cup J}^{(0)}$$

$$\implies M_{t-1}(z^{(0)}) \succeq 0$$

Similar to slack matrices. \square

Immediate consequence via induction:

Lemma 1 $y \in \text{Las}_t(k)$, $S \subseteq [n]$, $|S| \leq t$

$$\Rightarrow y \in \text{Conv} \{ z \in \text{Las}_{t-|S|}(k) \mid z_i \in \{0,1\} \forall i \in S \}$$

Actually can get an explicit convex combination.

Let

$$(1) \quad y_I = \sum_{H \subseteq I} (-1)^{|H|} y_{I \cup H, k}$$

Lemma 2 $y \in \text{Las}_t(k)$, $S \subseteq [n]$

$$\Rightarrow y = \sum_{\substack{J \supseteq S \\ J \neq \emptyset \\ z_j > 0}} y_J \left(\frac{y_J}{y_\emptyset} \right) \quad \text{(*)}$$

EX

check that

$$\sum y_J = 1$$

and $z^J \in \text{Las}_{t-|S|}(k)$, and

$$z_i^J = \begin{cases} 1 & : i \in J, \\ 0 & : i \notin J, \end{cases}$$

Proof: fairly straight forward via induction
and Lemma 1.

(\rightarrow sandwhich)

Note: Suppose $y \in \text{Las}_n(k)$ then there is a random variable $X \in \mathbb{K}^n \{0,1\}^n$ with

$$(**) \quad \Pr \left[\bigwedge_{i \in I} X_i = 1 \right] = y_I$$

Simply take decomposition of Lemma 2
and draw $\mathbb{P}_{\partial, H}$ with prob

$$\mathbb{P}_{\partial, H}$$

$\Rightarrow \text{** holds.}$

Inclusion-Exclusion:

$$\Pr[\bigvee_{i \in J_0} (X_i = 1)] = \sum_{\emptyset \subset H \subseteq J_0} (-1)^{|H|+1} \Pr[\bigwedge_{i \in H} (X_i = 1)]$$

$$\begin{aligned} \Rightarrow \Pr[\bigwedge_{i \in J_0} (X_i = 0)] &= 1 - \Pr[\bigvee_{i \in J_0} (X_i = 1)] \\ &= \sum_{H \subseteq J_0} (-1)^{|H|} \Pr[\bigwedge_{i \in H} (X_i = 1)] \end{aligned}$$

Intersecting all events with $\bigwedge_{i \in J_1} (X_i = 1)$ gives

generalized inclusion-exclusion formula:

$$\begin{aligned} \Pr[\bigwedge_{i \in J_1} (X_i = 1) \wedge \bigwedge_{i \in J_0} (X_i = 0)] \\ = \sum_{H \subseteq J_0} (-1)^{|H|} \Pr[\bigwedge_{i \in J_1 \cup H} X_i = 1] \end{aligned}$$

$$\text{**} = \sum_{H \subseteq J_0} (-1)^{|H|} \mathbb{P}_{\partial, H}$$

$$\stackrel{\text{(DEF)}}{=} \mathbb{P}_{\emptyset}$$

Hence:

.
as.

$$\begin{aligned}
 \frac{\mathbb{P}_{Z \sim \Phi}(\mathbf{z})}{\mathbb{P}_{Z \sim \Phi}(\mathbf{z}^*)} &= \frac{\Pr[\bigwedge_{i \in \mathcal{I} \cup \mathcal{I}'} (X_i = 1) \wedge \bigwedge_{i \in \mathcal{J}_0} (X_i = 0)]}{\Pr[\bigwedge_{i \in \mathcal{J}_0} (X_i = 0) \wedge \bigwedge_{i \in \mathcal{J}_1} (X_i = 1)]} \\
 &= \Pr\left[\bigwedge_{i \in \mathcal{I}} (X_i = 1) \mid \bigwedge_{i \in \mathcal{J}_0} (X_i = 0) \wedge \bigwedge_{i \in \mathcal{J}_1} (X_i = 1)\right]
 \end{aligned}$$

Thus: $Z^{\mathcal{J}_0, \mathcal{J}_1}$ \equiv Y conditioned on $\mathcal{J}_0, \mathcal{J}_1$

This global probabilistic view holds even locally:

Lemma 3 $y \in \text{Last}_t(\kappa)$, $S \subseteq [n]$, $|S| \leq t$

$\Rightarrow \exists$ dist $D(S)$ over $\{0, 1\}^{|S|}$ w.h.

$$\Pr_{Z \sim D(S)} \left[\bigwedge_{i \in S} (Z_i = 1) \right] = b_S \quad \forall S \subseteq S$$

Pf: again use convex cond. in Lemma 2.

note: nice consistency property:

$S \neq S' \subseteq [n]$, $|S|, |S'| \leq t$ then

$$\begin{aligned}
 \Pr_{Z \sim \Phi(S')} \left(\bigwedge_{i \in S} (Z_i = 1) \right) &= \Pr_{Z \sim D(S)} \left(\bigwedge_{i \in S} (Z_i = 1) \right) \\
 &= b_S \quad \forall S \subseteq S \cap S'
 \end{aligned}$$

skip?

Ex: 3-colouring

Let $G = (V, E)$ be a 3-colourable graph; i.e.,
 \exists colouring $(x_v)_{v \in V}$, $x_v \in \{g, r, b\}$

$\Leftrightarrow \forall u, v \in E : x_u \neq x_v$.

$$K = \{x \in [0,1]^{3n} \mid x_{i,c} + x_{j,c} \leq 1 \quad \forall i, j \in E, c \in \{g, r, b\}\}$$

Lemma 4: $y \in \text{Las}_{\geq t}(K)$

$\Rightarrow \exists$ family of dist. $(D(S))_{\substack{S \subseteq V \\ |S| \leq t}}$ s.t.

(a) each event $x \sim D(S)$ is valid
3-colouring of $G[S]$

(b) $\Pr_{x \sim D(S)} [x(i_1) = c_1, \dots, x(i_k) = c_k] =$
 $y_{(i_1, c_1), \dots, (i_k, c_k)}$

$\forall i_1, \dots, i_k \in S, c_1, \dots, c_k \in \{g, r, b\}$



Decomposition Property

Let $K \subseteq \mathbb{R}^n$ be relax of 0,1-polyhedron. We know that $\text{Las}_n(K) = \text{conv}(K \cap \{0,1\}^n)$. But Las more is often stronger.

Ex: Knapsack $K = \{x \in \mathbb{R}_{\geq 0}^n : \sum_{i=1}^n x_i \leq 1.9\}$

v_i : value of object i

Consider $y \in \text{Las}_2(K)$ and let $0 < y_i < 1$.

Claim $y_{\{i,j\}} = 0 \quad \forall i \neq j$

Pf: Suppose not and $\exists i \neq j$ s.t.

$$\gamma_{\{i,j\}} > 0$$

Then pick convex comb. from before:

$$\gamma = \gamma_i \gamma^{(1)} + (1-\gamma_i) \gamma^{(2)}$$

$$\text{know } P2d: \quad \gamma_{\{i,j\}}^{(o)} \leq \gamma_i^{(o)} = 0$$

$$\Rightarrow \gamma_{\{i,j\}}^{(1)} > 0 \Rightarrow \gamma_j^{(1)} > 0$$

But this cannot be, since by Lemma 1:

$$\exists \gamma^{(10)}, \gamma^{(11)} \in \text{Lass}_0(K) = K \text{ s.t.}$$

$$\gamma^{(1)} = \gamma_j^{(1)} \gamma^{(11)} + (1-\gamma_j^{(1)}) \gamma^{(10)}$$

and $\gamma^{(11)}, \gamma^{(10)} \in K$ but $\gamma^{(1)}$ viol. cap constraint.

□

Let $\{v_I\}_{\substack{I \subseteq \mathbb{F}_n \\ |I| \leq 2}}$ 3-dim vect. s.t.

$$\begin{aligned} \langle v_I, v_J \rangle &= \gamma_{I \cup J} \\ \Rightarrow \langle v_i, v_j \rangle &= \gamma_{\{i,j\}} = 0 \quad \forall i \neq j \end{aligned}$$

$$\begin{aligned} \Rightarrow \sum_{i=1}^n \gamma_i &= \sum_{i=1}^n \|v_i\|_2^2 = \underbrace{\sum_{i=1}^n \langle v_\emptyset, v_i \rangle}_{\langle v_i, v_i \rangle = \langle v_i, v_\emptyset \rangle} = \sum_{i=1}^n \left\langle v_\emptyset, \frac{v_i}{\|v_i\|_2} \right\rangle^2 \\ &\leq (\cos \varphi)^2 \|v_\emptyset\|_2^2 \\ &\leq \|v_\emptyset\|_2^2 = 1 \end{aligned}$$

$K + \mathbf{1}\mathbf{1}^T \subseteq 1$ defines $\text{Conv}(K \cap \{0,1\}^n)$.

$$\Rightarrow \text{Lass}_2(K) = \text{Conv}(K \cap \{0,1\}^n)$$

General result:

Lemma 5 $K = \{x \in \mathbb{R}^n : \forall x \geq \delta\}$

Suppose that any $x \in K$ has $\leq t$ ones.

Then

$$\text{Lar}_{t+1}(K) = \text{Conv}(K \cap \{0,1\}^n)$$

Pf: Know: $M_{t+1}(y) \geq 0$. Similar argument as
before shows that

$$y_I = 0$$

$$\forall I \subseteq [n] : |I| < |I| \leq 2(t+1)+1$$

index set dim in $\text{Lar}_{t+1}(K)$

Define: $y_I = 0 \quad \forall I, |I| > 2t+3$.

$$M_n(y) = \begin{bmatrix} M_{t+1}(y) & 0 \\ 0 & 0 \end{bmatrix} \quad \begin{array}{l} |I| \leq t+1 \\ |I| > t+1 \end{array}$$

$|I| \leq t+1 \quad |I| > t+1$

Clearly $M_n(y) \geq 0$ as $M_{t+1}(y) \geq 0$

One can also show that $M_n^e(y) \geq 0 \quad \forall 1 \leq e \leq m$.

$$\Rightarrow \text{Lar}_{t+1}(K) \subseteq \text{Lar}_n(K) \quad \square$$

Decomposition thm (Kardia, Mathieu, Nguyen '11)

Suppose that all $x \in K$ have $\leq k$ ones in $S \subseteq [n]$
and $k \leq t$. Then

$$y \in \text{Lar}_n(K) \Rightarrow y \in \text{Conv}\{z : z \in \text{Lar}_{S, \dots}(K)\}$$

v - t

$z_i \in \{0,1\}^{\frac{n}{k}}$ $\forall i \in S\}$