

Computing the nc-rank via discrete convex optimization on CAT(0) spaces

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Edmonds Problem Edmonds 1967

Can we compute the rank of

$$A = A_1x_1 + A_2x_2 + \cdots + A_mx_m$$

in polynomial time ?

x_i : variables, A_i : $n \times n$ matrices over field \mathbb{K}

A : matrix over $\mathbb{K}[x_1, x_2, \dots, x_m] \subset \mathbb{K}(x_1, x_2, \dots, x_m)$

- RP, but P ? (for large field)
- Related to fundamental problems in diverse areas
~ combinatorial optimization, rigidity theory, TCS,...

Non-commutative Edmonds Problem

Ivanyos-Qiao-Subrahmanyam 2017

Can we compute the rank (*nc-rank*) of

$$A = A_1x_1 + A_2x_2 + \cdots + A_mx_m$$

in polynomial time ?

x_i : noncommutative variables, A_i : matrices over field \mathbb{K}

A : matrix over free ring $\mathbb{K}\langle x_1, x_2, \dots, x_m \rangle$

\cap

free skew field $\mathbb{K}(\langle x_1, x_2, \dots, x_m \rangle)$

- Broaden the literature:
Noncommutative algebra, Invariant theory, ...

Null cone membership for left-right action

$$GL_n \times GL_n \ni (S, T) \mapsto (SA_kT)_{k=1,2,\dots,m}$$

nc-rank in \mathbf{P}

- Garg-Gurvits-Oliveira-Wigderson 2019 : $\mathbb{K} = \mathbb{Q}, \mathbb{R}, \mathbb{C}$

Gurvits' operator scaling \approx alternating minimization for
a *geodesically convex* function
on symmetric space GL_n/O_n

$$\text{nc-nonsingularity} \Leftrightarrow \inf \left\{ \log \frac{\det \sum A_i X A_i^T}{\det X} : X \succ 0 \right\} > -\infty$$

→ *Noncommutative optimization* (Bürgisser et al.)

- Ivanyos-Qiao-Subrahmanyam 2018: \mathbb{K} arbitrary

Wong sequence --- vector-space analogue of augmenting path

Matrix completion by cyclic division algebra

Our contribution [Hamada-H 2020]

A different polytime algorithm to compute nc-rank

Features / techniques:

- Submodular optimization on modular lattice
- Geodesically convex optimization on CAT(0) space

non-manifold

For nc-rank over \mathbb{Q} (to avoid bit-complexity issue):

- reduction to nc-rank over $GF(p)$ by p -adic valuation

→ discrete convex optimization

on Euclidean building for $GL_n(\mathbb{Q})$

Submodularity in nc-rank

Thm (Fortin-Reutenauer 2004)

$$\text{nc-rank } \sum_k A_k x_k = 2n - \text{Max. } r + s$$

$$\text{s.t. } PA_k Q = \begin{matrix} & \begin{matrix} * & & * \\ & * & & \\ \hline \mathbf{0} & & & * \\ & & & \end{matrix} & \\ & \begin{matrix} r & & & \\ & s & & \end{matrix} & \end{matrix} \quad (\forall k)$$

$$P, Q \in GL_n(\mathbb{K})$$

$$\text{Max. } \dim X + \dim Y \quad \text{s.t. } A_k(X, Y) = \{0\} \quad (\forall k)$$

$$X, Y \subseteq \mathbb{K}^n \text{ vector subspaces}$$

$$\text{where } A_k(x, y) := x^T A_k y$$

$$\text{Max. } \dim X + \dim Y \quad \text{s.t. } A_k(X, Y) = 0 \quad (\forall k)$$

$X, Y \subseteq \mathbb{K}^n$ vector subspaces

$2n + 1$

III

$$\text{Min. } -\dim X - \dim Y + C \sum_k \text{rank } A_k|_{X \times Y}$$

$$\text{s.t. } (X, Y) \in \mathcal{L} \times \mathcal{L}$$

\mathcal{L} : the modular lattice of all vector subspaces of \mathbb{K}^n

partial order reversed

Lem: The objective function is submodular in $\mathcal{L} \times \mathcal{L}^*$

$$f(X, Y) + f(X', Y') \geq f(X + X', Y \cap Y') + f(X \cap X', Y + Y')$$

$$\wedge = (+, \cap), \vee = (\cap, +)$$

Our approach to solve SFM for nc-rank

$$\text{Min. } -\dim X - \dim Y + C \sum_i \text{rank } A_i |_{X \times Y}$$

$$\text{s. t. } (X, Y) \in \mathcal{M} = \mathcal{L} \times \mathcal{L}^*$$

- Continuous & convex relaxation to *CAT(0) space*

$$\mathcal{M} \xRightarrow{\substack{\text{orthoscheme} \\ \text{complex}}} K(\mathcal{M}), \text{ a CAT(0) space}$$

$$f \xRightarrow{\substack{\text{Lovász} \\ \text{extension}}} \bar{f}: K(\mathcal{M}) \rightarrow \mathbb{R}, \text{ geodesically convex}$$

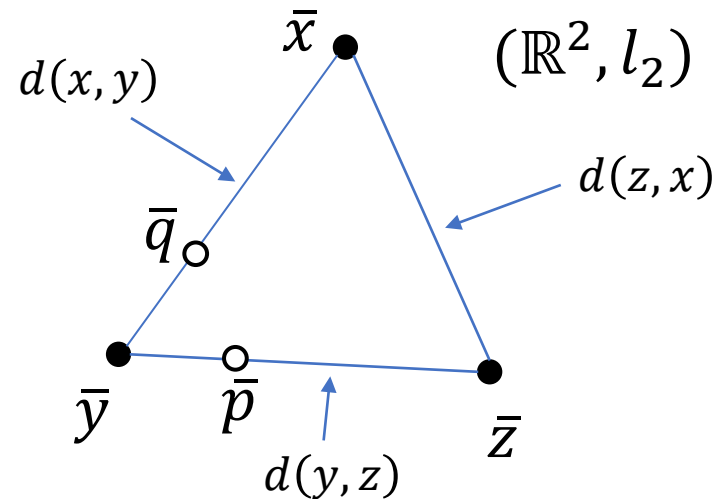
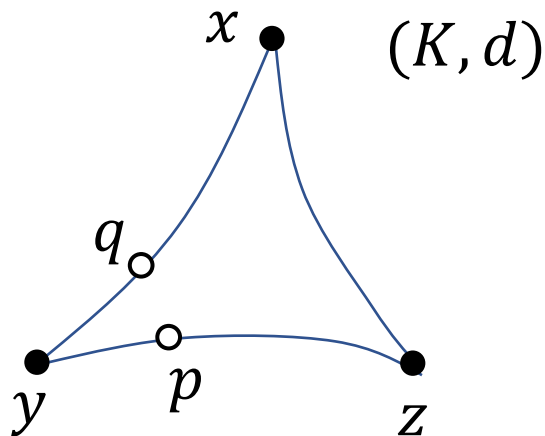
- Minimize \bar{f} by an optimization algorithm on CAT(0) space

Inspired by the classical SFM [Lovász-Grötschel-Schrijver] :

- $f: \{0,1\}^n \rightarrow \mathbb{R}$, submodular $\Rightarrow \bar{f}: [0,1]^n \rightarrow \mathbb{R}$, convex
- Minimize \bar{f} by ellipsoid method

CAT(0) space (Gromov 1987)

A geodesic metric space (K, d) s.t. every triangle is “slimmer”



$$\text{CAT}(0)\text{-inequality: } d(p, q) \leq \|\bar{p} - \bar{q}\|_2$$

Ex: \mathbb{R}^n , hyperbolic space, symmetric space of noncompact type
tree, Euclidean building, ...

Modeling & Algorithm & Optimization on CAT(0) spaces

Fact: CAT(0) space is uniquely geodesic

→ Convexity is naturally defined

- The space of phylogenetic trees (BHV-tree space) [Billera et al 2001]
- Configuration spaces of robots [Abram & Ghrist 2004]
- Polynomial-time geodesic computation
 - for BHV-tree space [Owen 2011]
 - for CAT(0) cubical complex [Hayashi 2021]
- Proximal point algorithm for convex optimization [Bačák 2014]
- Combinatorially-defined CAT(0) spaces:
 - median graph → CAT(0) cubical complex [Chepoi 2000]
 - poset / lattice → orthoscheme complex [Brady-McCammond 2012]

Orthoscheme Complex Brady-McCammond 2012

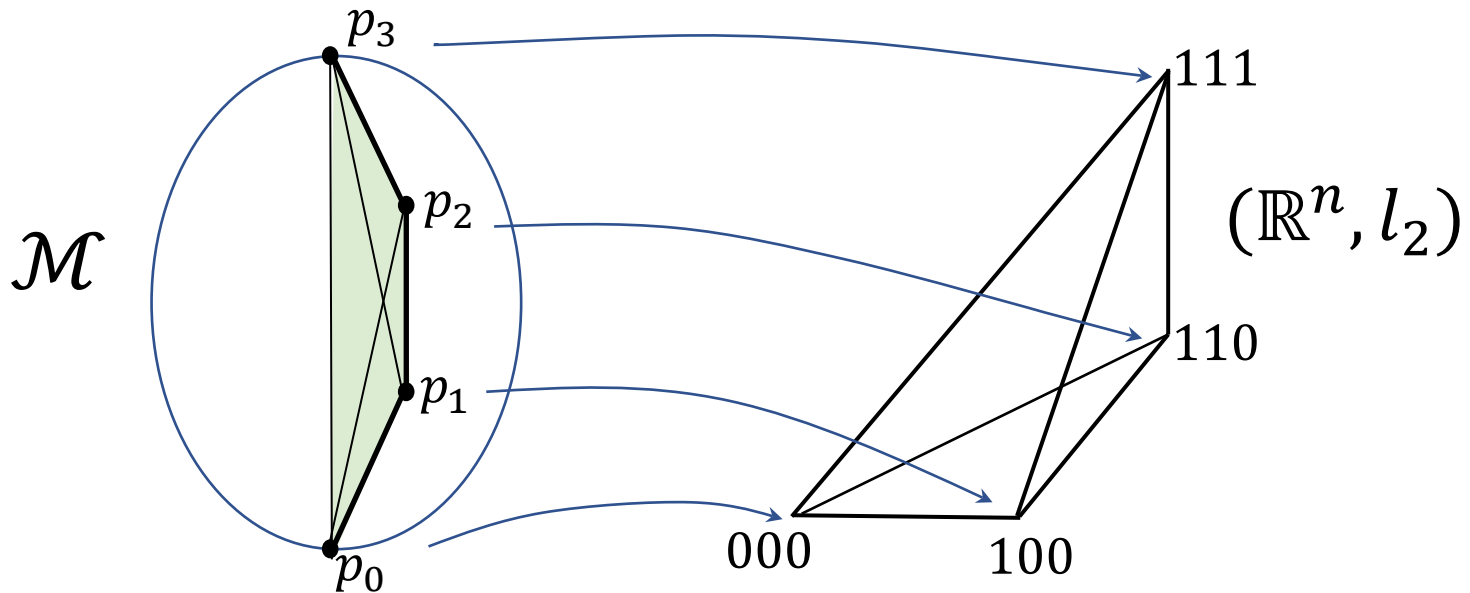
\mathcal{M} : graded poset

formal convex combination $\sum_i \mu_i = 1, \mu_i \geq 0$

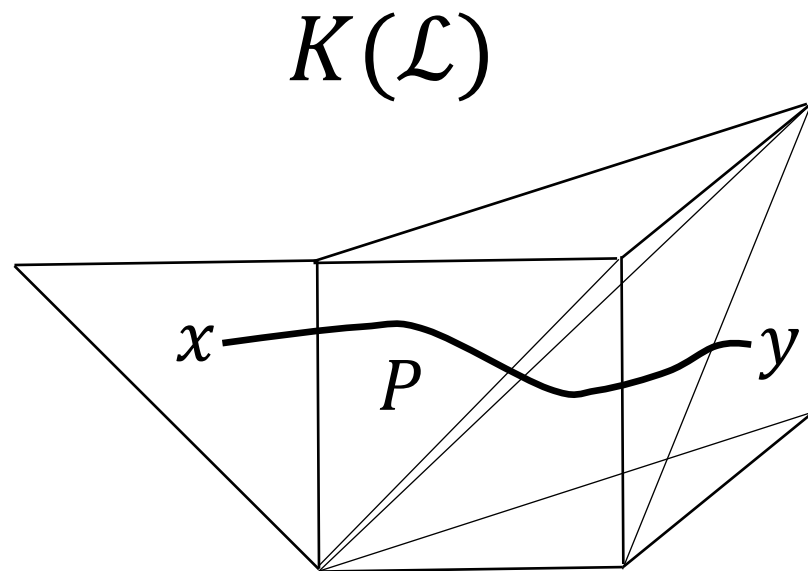
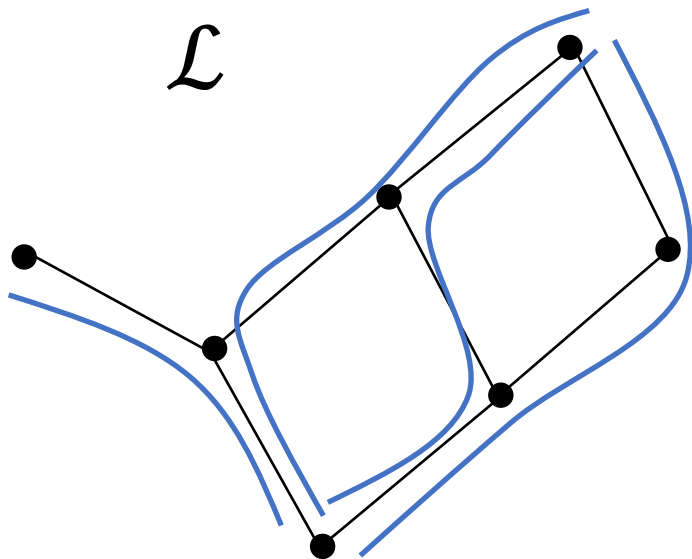
$K(\mathcal{M}) := \{ \sum_i \mu_i p_i \mid p_0 < p_1 < \dots < p_n, p_i \in \mathcal{M} \}$ = the order complex of \mathcal{M}

metrized by filling *orthoscheme* to each simplex:

$$\sum_i \mu_i p_i \mapsto \sum_i \mu_i (\overbrace{1, \dots, 1}^i, 0, \dots, 0)$$



Orthoscheme = $\text{conv} \{ (0,0, \dots, 0), (1,0, \dots, 0), (1,1,0, \dots, 0), \dots, (1,1,1, \dots, 1) \}$ 11



The length $d(P)$ of a path $P: [0,1] \rightarrow K(\mathcal{L})$ is well-defined

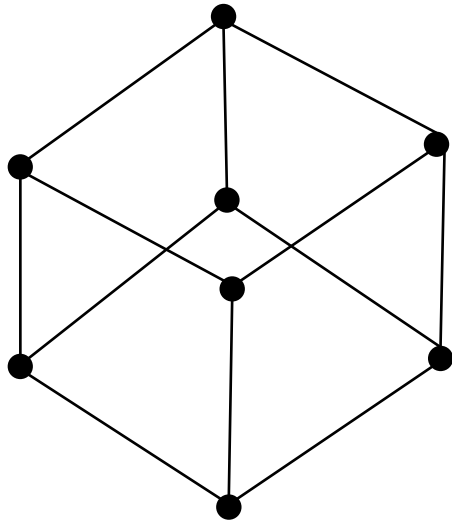
$$d(x, y) := \inf \{d(P) \mid P: (x, y)\text{-path}\}$$

$\rightarrow (K(\mathcal{L}), d)$ is a geodesic metric space

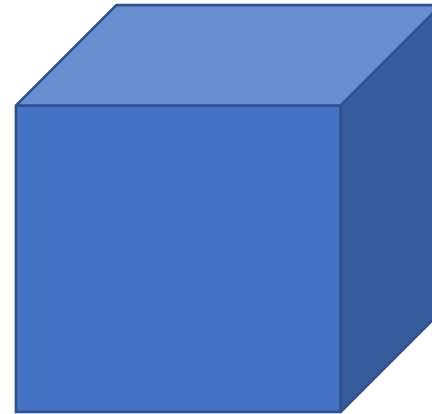
Question [Brady-McCammond 2012]:

Which is the class of posets \mathcal{L} having $CAT(0)$ $K(\mathcal{L})$?

Boolean Lattice



$$\mathcal{L} = \{0,1\}^n$$



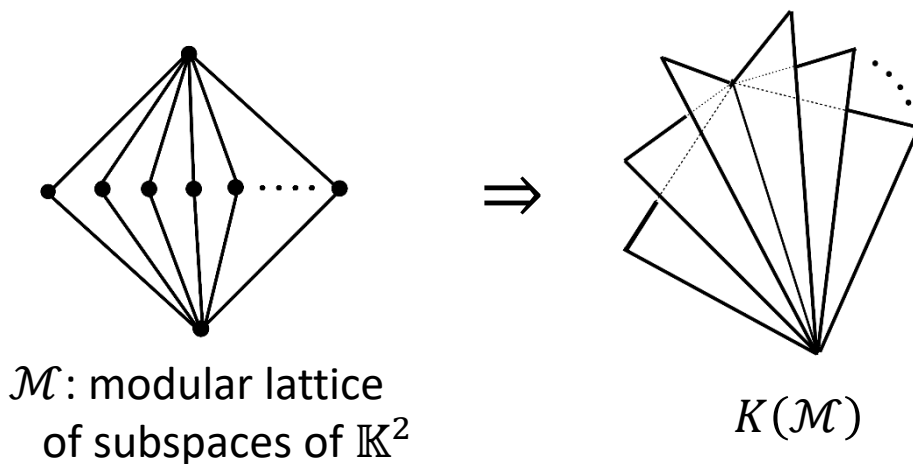
$$K(\mathcal{L}) \simeq [0,1]^n$$

$$\rightarrow \text{CAT}(0)$$

\mathcal{L} : distributive lattice \rightarrow $K(\mathcal{L})$: order polytope, CAT(0)

Thm [Haettel-Kielak-Schwer 2015, Chalopin-Chepoi-H-Osajda 2020]

\mathcal{M} : modular lattice $\Rightarrow K(\mathcal{M})$ is CAT(0)



Thm [H.2018] \mathcal{M} : modular lattice

$f: \mathcal{M} \rightarrow \mathbb{R}$ is submodular $\Leftrightarrow \bar{f}: K(\mathcal{M}) \rightarrow \mathbb{R}$ is convex

$$\bar{f}(x) := \sum_i \mu_i f(p_i) \quad \left(x = \sum_i \mu_i p_i ; p_0 < p_1 < \dots < p_n, \right)$$

Classical Lovász extension: $f: \{0,1\}^n \rightarrow \mathbb{R}$ is submodular $\Leftrightarrow \bar{f}: [0,1]^n \rightarrow \mathbb{R}$ is convex

$$\text{Min. } -\dim X - \dim Y + C \sum_i \text{rank } A_i|_{X \times Y}$$

$$\text{s. t. } (X, Y) \in \mathcal{L} \times \mathcal{L}^*$$

⇓ Lovász extension

$$\text{Min. } -\overline{\dim}(x) - \overline{\dim}(y) + C \sum_i \overline{\text{rank}} A_i(x, y)$$

$$\text{s. t. } (x, y) \in K(\mathcal{L} \times \mathcal{L}^*) = K(\mathcal{L}) \times K(\mathcal{L}^*)$$

→ convex optimization on CAT(0) space $K(\mathcal{L} \times \mathcal{L}^*)$
but we have no powerful method such as ellipsoid method

Feature: the objective is a sum of convex functions

→ splitting proximal point algorithm

Splitting Proximal Point Algorithm Bačák 2014

(K, d) : complete CAT(0) space, f_1, f_2, \dots, f_m : convex func on K

Goal: Minimize $f = \sum_{i=1}^m f_i$

SPPA: Iterate

$$z^{k+1} \leftarrow \operatorname{argmin}_{z \in K} f_{k \bmod m}(z) + \frac{1}{\lambda_k} d(z, z^k)^2$$

Thm (Ohta-Pálfia 2015)

$$f_i: L\text{-Lipschitz}, \quad f: \varepsilon\text{-strongly convex}, \quad \lambda_k := \frac{1}{2\varepsilon(k+1)}$$

$$\Rightarrow f(z^k) - \min f \leq \frac{\operatorname{poly}(L, m, \varepsilon^{-1}, \operatorname{diam} K)}{\sqrt{k}}$$

Min. $-\overline{\dim}(x) - \overline{\dim}(y) + C \sum_i \overline{\text{rank } A_i}(x, y)$ + Perturbation term
for ε -strong convexity

s. t. $(x, y) \in K(\mathcal{L} \times \mathcal{L}^*) = K(\mathcal{L}) \times K(\mathcal{L}^*)$

- Apply SPPA with:

$$f_0(x, y) := -\overline{\dim}(x) - \overline{\dim}(y)$$

$$f_i(x, y) := C \overline{\text{rank } A_i}(x, y) \quad (i = 1, 2, \dots, m)$$

- After $k = \text{poly}(n, m)$ iterations, we have $f(x^k, y^k) - \text{opt} < 1/2$

- By integrality of f , an optimum (X^*, Y^*) exists

$$\text{in the support of } (x^k, y^k) = \sum_i \mu_i(X_i, Y_i)$$

Main technical contribution

Thm: SPAA is implementable in polytime, i.e.,

$$\text{Min. } f_i(x, y) + \frac{1}{\lambda} d((x, y), (x_0, y_0))^2$$

$$\text{s. t. } (x, y) \in K(\mathcal{L} \times \mathcal{L}^*)$$

where $f_i(x, y) = -\overline{\dim}(x) - \overline{\dim}(y)$ or

$$f_i(x, y) = C \overline{\text{rank}} A_i(x, y)$$

is solvable in polytime.

Proof: Lattice theoretic / building theoretic argument

- An optimum exists in an *apartment* = the subcomplex of a Boolean sublattice

including the supports of $x_0, y_0, x_0^{\perp A_i}, y_0^{\perp A_i}$

- In the apartment $\approx [0, 1]^{2n}$, the problem is an easy convex quadratic program

Good / bad points

- Conceptually simple, described uniformly in an arbitrary field
- Applications to other problems ?
- Polytime, but very slow:
 - ← Less use of characteristics of the objective function
 - ← Undevelopment of optimization theory in CAT(0) spaces
- Bit complexity problem for $\mathbb{K} = \mathbb{Q}$:
 - ← Polynomial number of vector-space operations $X \cap X', X + X'$
can cause exponential bit explosion of bases

Reduction of nc-singularity over \mathbb{Q} to that over \mathbb{F}_p
 by p -adic valuation v_p & discrete convex optimization
 on Euclidean building

$$\begin{aligned} \text{Min.} \quad & v_p \det P + v_p \det Q \\ \text{s. t.} \quad & v_p (PA_k Q)_{ij} \geq 0 \quad (\forall ij, k), \\ & P, Q \in GL_n(\mathbb{Q}) \end{aligned}$$

- Nc-singular \Leftrightarrow optimal value = $-\infty$ $z = \underbrace{1011 \overbrace{000}^{v_2(z)}}_{\text{bitsize}}$
- $\text{opt} > -\infty \Rightarrow \text{opt} > -\text{poly}(\text{input}) \leftarrow v_p(|z|) \leq \text{bit-size of } z \text{ for } 0 \neq z \in \mathbb{Z}$
- *L-convex function* minimization on Euclidean building for $GL_n(\mathbb{Q})$
- Descent algorithm [H.2019], where each step is nc-rank computation over \mathbb{F}_p

Application to related problems

(p, q) -scalability of $\{A_k\}$ by triangular matrices [Franks 2018]

$$\begin{aligned} \text{Max.} \quad & \sum_i (p_i - p_{i+1}) \dim E_i \cap X + \sum_i (q_i - q_{i+1}) \dim F_i \cap Y \\ & E_i, F_i: \text{standard flags} \\ \text{s. t.} \quad & A_k(X, Y) = \{0\} \quad (\forall k) \\ & X, Y: \text{vector subspaces} \end{aligned}$$

Brascamp-Lieb separation for BL-data (B, p) [Garg et al. 2018]

$$\begin{aligned} \text{Max.} \quad & \dim X + \sum_i p_i \dim Y_i \\ \text{s. t.} \quad & B_i(X, Y_i) = \{0\} \quad (\forall i) \\ & X, Y_i: \text{vector subspaces} \end{aligned}$$

Our approach is applicable, but it only gives pseudo-polynomial complexity
I don't know whether p-adic reduction can work

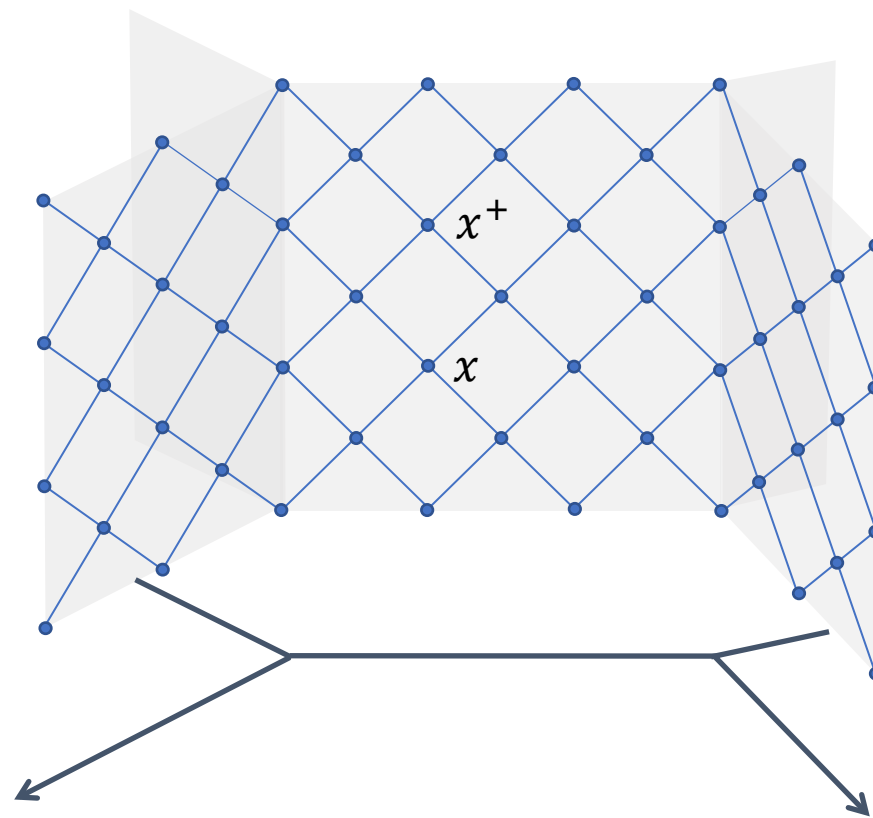
Reference

M. Hamada and H. Hirai: Computing the nc-rank via discrete convex optimization on CAT(0) spaces, *SIAM Journal on Applied Geometry and Algebra (SIAGA)*, to appear.

Papers and slides are available at

<http://www.misojiro.t.u-tokyo.ac.jp/~hirai/>

Thank you for your attention !



L-convex function on Euclidean building (of type A) [H. 2019]

$$f(x^+) = f(x) + \text{const}$$

$$f(x) + f(y) \geq f(x \wedge y) + f(x \vee y)$$

extends the one in *Discrete Convex Analysis* on \mathbb{Z}^n [Murota 1998]

→ *Discrete Convex Analysis beyond \mathbb{Z}^n* [H. 2013~]