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Discrete Convex Analysis II: Properties of Discrete Convex Functions

Kazuo Murota
(Tokyo Metropolitan University)

Contents of Part II

Properties of Discrete Convex Functions

P1. Convex Extension

P2. Optimality Criterion (local = global)

P3. Operations

P4. Conjugacy (Legendre transform)

P5. Duality (separation, Fenchel)

Classes of Discrete Convex Functions

- 1. Submodular set fn (on $\{0,1\}^n$)
- 1. Separable-convex fn on \mathbb{Z}^n
- 1. Integrally-convex fn on \mathbb{Z}^n

- 2. L-convex (L^\natural -convex) fn on \mathbb{Z}^n
- 2. M-convex (M^\natural -convex) fn on \mathbb{Z}^n

- 3. M-convex fn on jump systems
- 3. L-convex fn on graphs

P1.

Convex Extension

Convex Extension

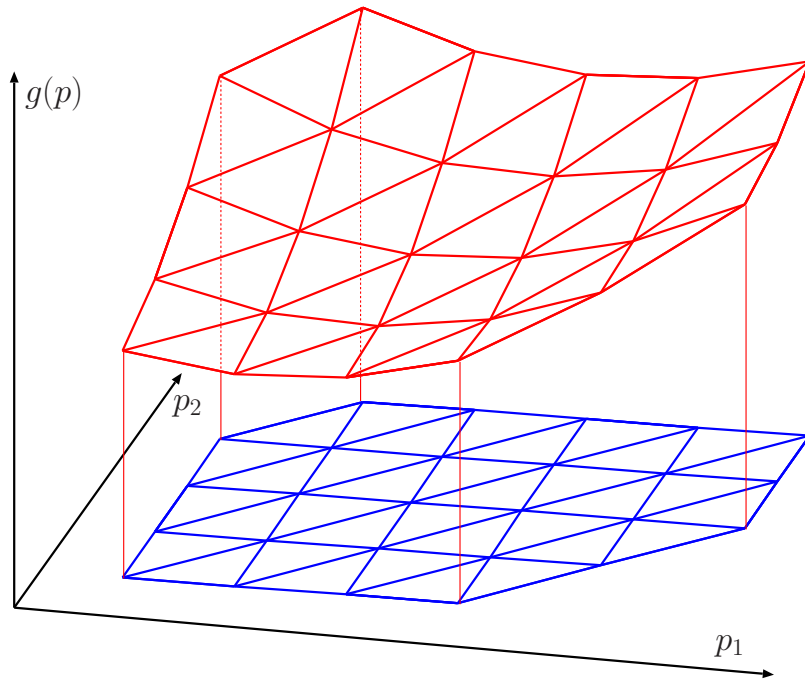
$f : \mathbb{Z}^n \rightarrow \bar{\mathbb{R}}$ is **convex-extendible**

$$\Leftrightarrow \exists \text{ convex } \bar{f} : \mathbb{R}^n \rightarrow \bar{\mathbb{R}}: \bar{f}(x) = f(x) \quad (\forall x \in \mathbb{Z}^n)$$

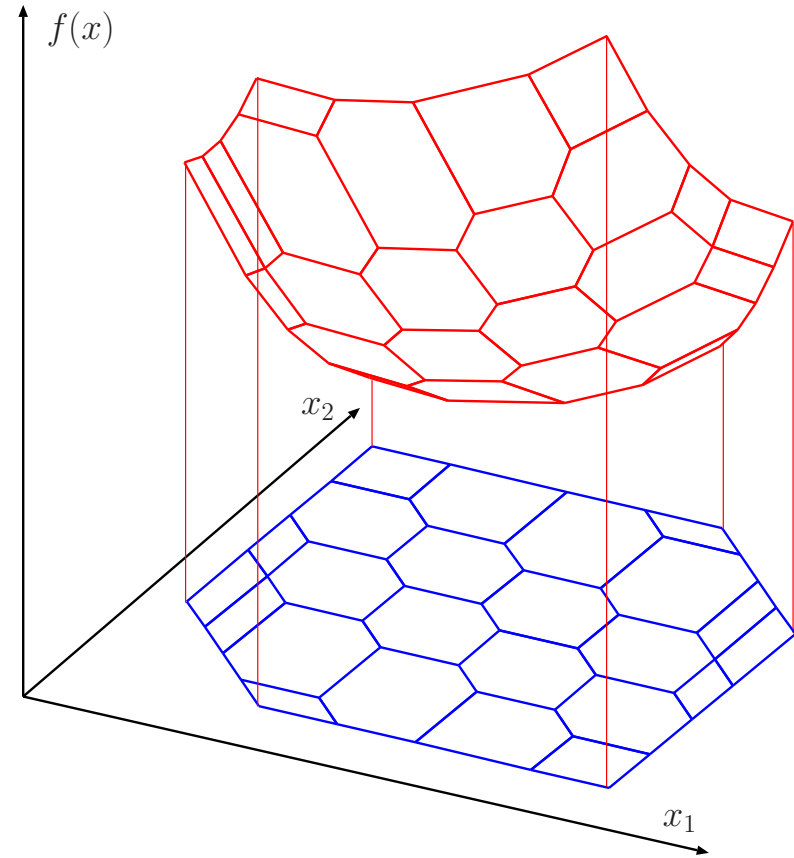
Theorem:

- (1) **Separable-convex** fns are convex-extendible
- (2) **Integrally-convex** fns are convex-extendible (by def)
- (3) **L^{\natural} -convex** fns are convex-extendible (Murota 98)
- (4) **M^{\natural} -convex** fns are convex-extendible (Murota 96)

Bivariate L^{\natural} - and M^{\natural} -convex Functions



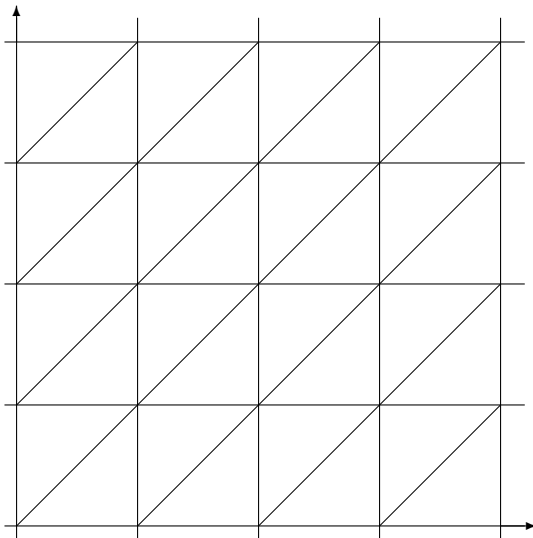
L^{\natural} -convex fn



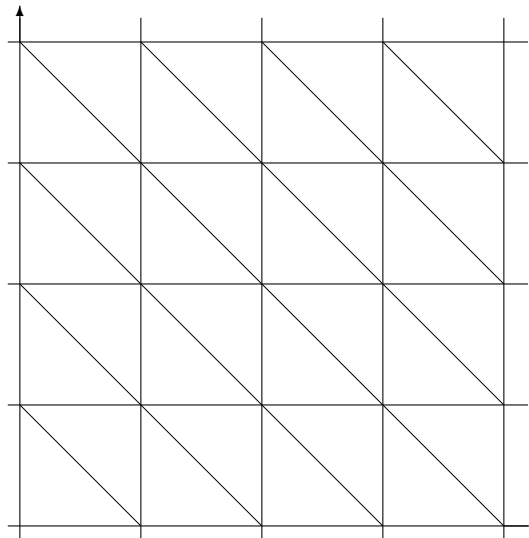
M^{\natural} -convex fn

Triangulation for Discrete Convex Functions

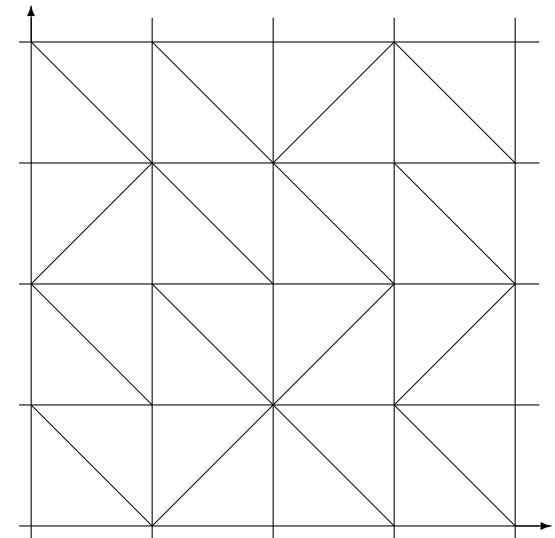
$(n = 2)$



L-convex

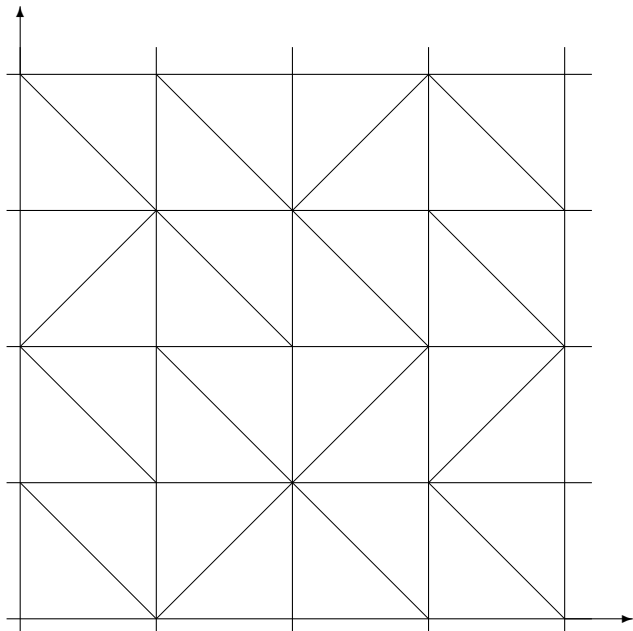


M-convex



Integrally convex

Bivariate Integrally-convex Functions



0	2	4	4	6 = $g(4, 4)$
0 ^M	1 ^M	2 ^L	3 ^M	4 = $g(4, 3)$
0 ^L	2 ^M	2 ^M	2 ^M	2
0 ^M	1 ^M	0 ^L	1 ^L	2
0 ^M	0 ^L	0 ^M	0 ^M	0
0				

$$g(x_1, x_2) = \#(\mathbf{M}) - \#(\mathbf{L}) \text{ in } [0, (x_1, x_2)]$$

$$f(x_1, x_2) = A(x_1^2 + x_2^2) + g(x_1, x_2)$$

Convex Extension — Computation

$f : \mathbb{Z}^n \rightarrow \bar{\mathbb{R}}$ is **convex-extendible**

$$\Leftrightarrow \exists \text{ convex } \bar{f} : \mathbb{R}^n \rightarrow \bar{\mathbb{R}} : \bar{f}(x) = f(x) \quad (\forall x \in \mathbb{Z}^n)$$

Theorem:

(1) **Separable-convex**

easy to compute (consecutive points)

(2) **Integrally-convex**

difficult (exp-time) to compute

(3) **L^{\natural} -convex** (Favati–Tardella 90, Murota 98)

easy to compute (Lovász ext.)

(4) **M^{\natural} -convex** (Shioura 09,15)

poly-time to compute (via conjugacy)

Classes of Discrete Convex Functions

$$f : \mathbb{Z}^n \rightarrow \overline{\mathbb{R}}$$

convex-extensible

integrally convex

M^{\natural} -convex

**separable
convex**

L^{\natural} -convex

$$M^{\natural} \cap L^{\natural} = \text{separable}$$

P2.

Optimality Criterion

(local opt = global opt)

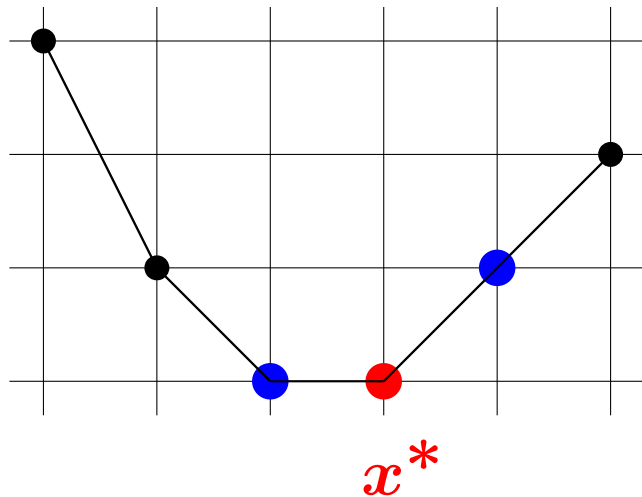
Local vs Global Optimality ($n = 1$)

$$f : \mathbb{Z} \rightarrow \overline{\mathbb{R}}$$

x^* : global opt (min)

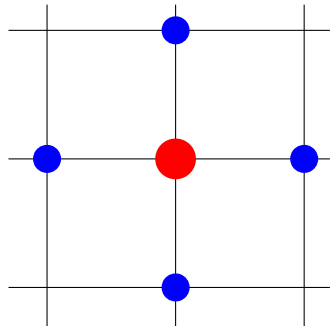
$\iff x^*$: local opt (min)

$$f(x^*) \leq \min\{f(x^* - 1), f(x^* + 1)\}$$



Neighborhood for Local Optimality

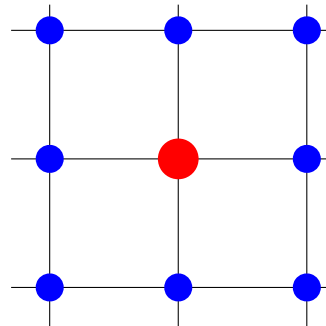
separable
convex



$$2n + 1$$

$$\{\pm e_1, \dots, \pm e_n\}$$

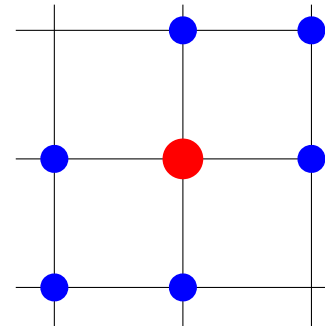
integrally
convex



$$3^n$$

$$\{\chi_X - \chi_Y\}$$

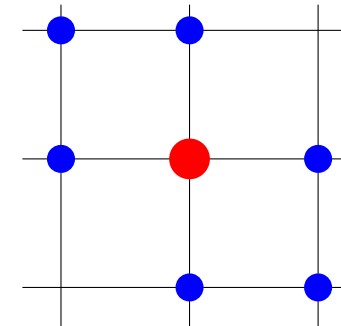
L_{\square} -
convex



$$2^{n+1} - 1$$

$$\{\pm \chi_X\}$$

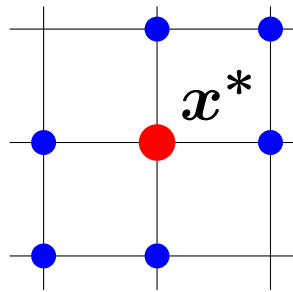
M_{\square} -
convex



$$n(n + 1) + 1$$

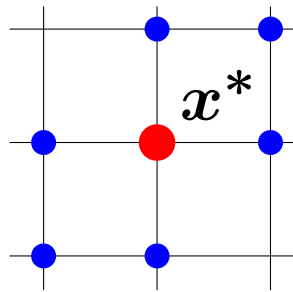
$$\{e_i - e_j\}$$

Local min = Global min



	= ?	#neigh -bors	poly-time local opt	algorithm global opt
submodular (set fn)	Y	2^n		
separable-conv	Y	$2n$		
integrally-conv	Y	3^n		
L^{\natural} -conv (\mathbb{Z}^n)	Y	2^n		
M^{\natural} -conv (\mathbb{Z}^n)	Y	n^2		

Local min = Global min



	= ?	#neigh -bors	poly-time local opt	algorithm global opt
submodular (set fn)	Y	2^n	Y	
separable-conv	Y	$2n$	Y	
integrally-conv	Y	3^n	N	
L^{\natural} -conv (\mathbb{Z}^n)	Y	2^n	Y	
M^{\natural} -conv (\mathbb{Z}^n)	Y	n^2	Y	

P3.

Operations

Operations

- **scaling:** $af(x) + b, \quad f(ax + b)$
- **linear addition:** $f(x) + \langle p, x \rangle$
- **section:** $f(x, 0)$
- **projection (partial minimization):** $\min_y f(x, y)$
- **sum:** $f_1(x) + f_2(x)$
- **convolution:** $(f_1 \square f_2)(x) = \min_y (f_1(y) + f_2(x - y))$
- **transformation by graphs/networks**

Scaling/Linear Addition

	$af(x)$ ($a > 0$)	$f(sx)$ ($s \in \mathbb{Z}_+$)	$f(-x)$	$f(x) + \langle p, x \rangle$
submodular (set fn)	Y	—	Y* $V \setminus X$	Y
separable-conv	Y	Y	Y $\pm x_i$	Y
integrally-conv	Y	N	Y $\pm x_i$	Y
L-conv (\mathbb{Z}^n)	Y	Y	Y	Y
M-conv (\mathbb{Z}^n)	Y	N	Y	Y

Section/Projection

	section $f(x, 0)$	projection $\min_y f(x, y)$
submodular (set fn)	Y restriction	Y contraction*
separable-conv	Y	Y
integrally-conv	Y	Y
L-conv (\mathbb{Z}^n)	N	Y
L ^h -conv	Y	Y
M-conv (\mathbb{Z}^n)	Y	N
M ^h -conv	Y	Y

contraction: $\rho_A(X) = \rho(X \cup A) - \rho(A)$

Sum and Convolution

- $(f_1 + f_2)(x) = f_1(x) + f_2(x)$

Theorem:

(Murota 98)

$$\begin{array}{ccc} f_1, f_2 : \mathbf{L} & \implies & f_1 + f_2 : \mathbf{L} \\ & & \mathbf{L}^\natural \implies \mathbf{L}^\natural \end{array}$$

$$f_1, f_2, \dots, f_k : \mathbf{L} / \mathbf{L}^\natural \implies f_1 + f_2 + \dots + f_k : \mathbf{L} / \mathbf{L}^\natural$$

- $(f_1 \square f_2)(x) = \min_y (f_1(y) + f_2(x - y))$

Theorem:

(Murota 96)

$$\begin{array}{ccc} f_1, f_2 : \mathbf{M} & \implies & f_1 \square f_2 : \mathbf{M} \\ & & \mathbf{M}^\natural \implies \mathbf{M}^\natural \end{array}$$

$$f_1, f_2, \dots, f_k : \mathbf{M} / \mathbf{M}^\natural \implies f_1 \square f_2 \square \dots \square f_k : \mathbf{M} / \mathbf{M}^\natural$$

Rem: $\mathbf{M} + \mathbf{M}$ is **not** \mathbf{M} , $\mathbf{L} \square \mathbf{L}$ is **not** \mathbf{L}

Significance of M-convolution Thm

Concave convolution:

$$(U_1 \square U_2)(x) = \max_y (U_1(y) + U_2(x - y))$$

U_1, U_2, \dots, U_k : gross-substitute (M^\sharp -concave)

\Rightarrow **aggregated utility** $U_1 \square U_2 \square \dots \square U_k$ **is**
gross-substitute (M^\sharp -concave)

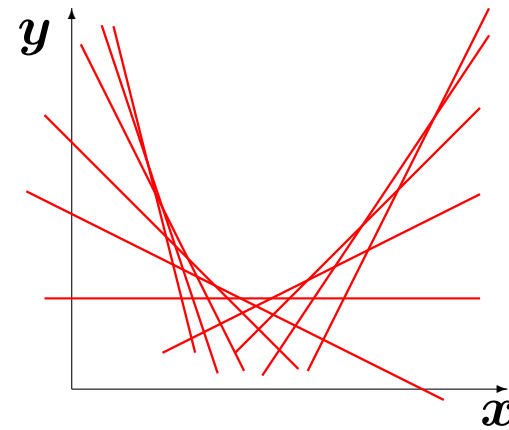
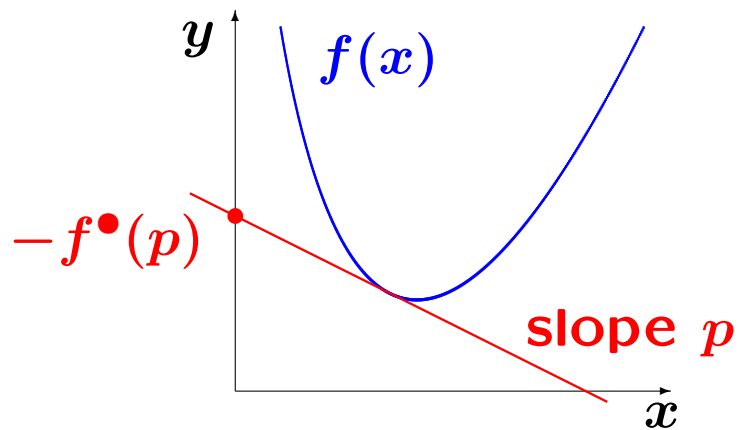
Sum/Convolution

	sum $f_1 + f_2$	convolution $f_1 \square f_2$
submodular (set fn)	Y	N matroid intersec $\min_{Y \subseteq X} (\rho_1(Y) + \rho_2(X \setminus Y))$
separable-conv	Y	Y
integrally-conv	N	N
L-conv (\mathbb{Z}^n)	Y	N \rightarrow L₂-convex
M-conv (\mathbb{Z}^n)	N \rightarrow M₂-conv matr.intersec	Y matroid union

P4. Conjugacy

(Legendre transform)

Conjugacy: Discrete Legendre Transform



$$f^\bullet(p) = \sup_{x \in \mathbb{Z}^n} \{ \langle p, x \rangle - f(x) \}$$

\Rightarrow If $f : \mathbb{Z}^n \rightarrow \overline{\mathbb{Z}}$, then $f^\bullet : \mathbb{Z}^n \rightarrow \overline{\mathbb{Z}}$
(integer-valued)

M-L Conjugacy Theorem

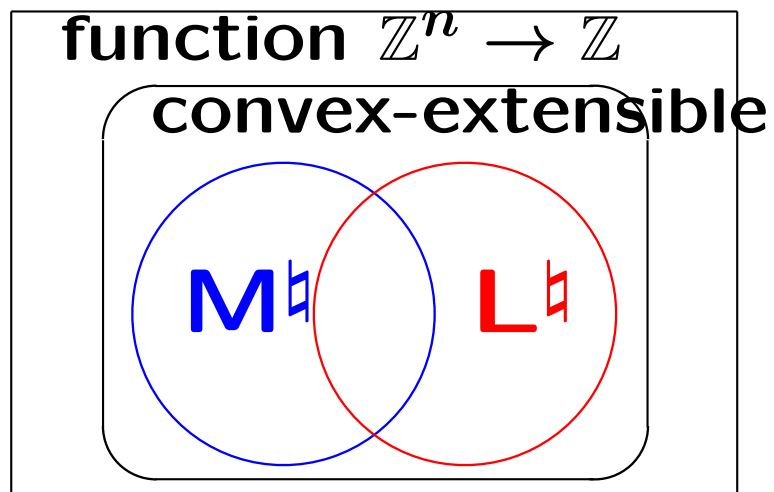
Integer-valued discrete fn $f : \mathbb{Z}^n \rightarrow \overline{\mathbb{Z}}$

Legendre transform: $f^\bullet(p) = \sup_{x \in \mathbb{Z}^n} [\langle p, x \rangle - f(x)]$

(1) **M and L are conjugate** (Murota 98)

(2) **M^\natural and L^\natural are conjugate**

$$f \mapsto f^\bullet = g \mapsto g^\bullet = f$$



(3) **biconjugacy**

$$f^{\bullet\bullet} = f$$

for $f \in M^\natural \cup L^\natural$

Significance of M-L Conjugacy

- Economics (game, auction)

x : commodity bundle, p : price vector

- Network flow (min-cost flow)

x : flow, p : tension (potential)

- Electrical network

(Iri's book 69)

x : current, p : voltage (potential)

- Discrete DC programming

(Maehara-Murota 15)

- Convexity view on matroids

Conjugacy in Linear Algebra

$$[a_1, \dots, a_5] = \begin{array}{|c|c|c|c|c|} \hline 1 & 0 & 0 & 1 & 0 \\ \hline 0 & 1 & 0 & 1 & 1 \\ \hline 0 & 0 & 1 & 0 & 1 \\ \hline \end{array}$$

Bases $\mathcal{B} = \{ \{1, 2, 3\}, \{1, 2, 5\}, \{1, 3, 4\}, \{1, 3, 5\},$
 $\{1, 4, 5\}, \{2, 3, 4\}, \{2, 4, 5\}, \{3, 4, 5\} \}$

Rank fn $\rho(X) = \text{rank} \{a_j \mid j \in X\}$

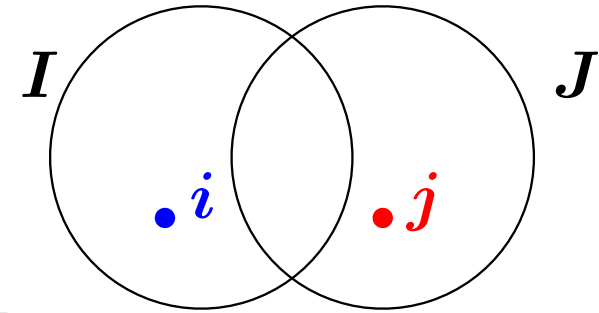
Equivalence $\mathcal{B} \iff \rho$

$$\rho(X) = \max\{|X \cap J| \mid J \in \mathcal{B}\} \quad (X \subseteq V)$$

$$\mathcal{B} = \{J \subseteq V \mid \rho(J) = |J| = \rho(V)\}$$

Axioms of Matroid

Basis axiom (set family \mathcal{B}):



$$\forall I, J \in \mathcal{B}, i \in I \setminus J, \exists j \in J \setminus I:$$

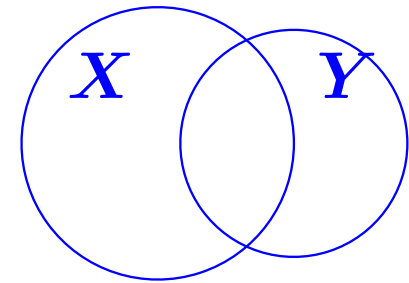
$$I - i + j \in \mathcal{B}, J + i - j \in \mathcal{B}$$

Rank axiom (set function ρ):

(R1) $0 \leq \rho(X) \leq |X|$

(R2) $X \subseteq Y \implies \rho(X) \leq \rho(Y)$

(R3) $\rho(X) + \rho(Y) \geq \rho(X \cup Y) + \rho(X \cap Y)$



Equivalence $\mathcal{B} \Leftrightarrow \rho$ (M \leftrightarrow L)

$$\rho(X) = \max\{|X \cap J| \mid J \in \mathcal{B}\} \quad (X \subseteq V)$$

$$\mathcal{B} = \{J \subseteq V \mid \rho(J) = |J| = \rho(V)\}$$

Conjugacy in Matroid

Bases $\mathcal{B} = \{ \{1, 2, 3\}, \{1, 2, 5\}, \{1, 3, 4\}, \{1, 3, 5\},$
 $\{1, 4, 5\}, \{2, 3, 4\}, \{2, 4, 5\}, \{3, 4, 5\} \}$

Rank fn $\rho(X) = \text{rank} \{a_j \mid j \in X\}$

Equivalence $\mathcal{B} \iff \rho$

$$\rho(X) = \max\{|X \cap J| \mid J \in \mathcal{B}\} \quad (X \subseteq V)$$

$$\mathcal{B} = \{J \subseteq V \mid \rho(J) = |J| = \rho(V)\}$$

Bases \leftarrow conjugate \Rightarrow **Rank fn**
(M[♯]-convex) (L[♯]-convex)

Dual Character of Matroid Rank

$$\rho(X) = \max\{|I| \mid I : \text{independent}, I \subseteq X\}$$

is **M[♯]-concave** and **L[♯]-convex**

Edmonds' matroid union formula:

$$\max_X \{\rho_1(X) + \rho_2(V \setminus X)\} = \min_Y \{\rho_1(Y) + \rho_2(Y) + |V \setminus Y|\}$$

submod maximization
(M[♯]-concave \square M[♯]-concave)

submod minimization
(L[♯]-convex + L[♯]-convex)

Self-Conjugacy: $\rho(X) = |X| - \rho^\bullet(\chi_X)$

Conjugacy in Polymatroids

Polyhedron S

$$S = \{x \mid x(A) \leq \rho(A) \quad \forall A\} \quad \leftarrow$$

Submodular fn ρ

$$\rightarrow \rho(A) = \max_{x \in S} x(A)$$

Conjugacy in Polymatroids

Polyhedron S

$$S = \{x \mid x(A) \leq \rho(A) \ \forall A\} \leftarrow$$

Submodular fn ρ

$$\rightarrow \rho(A) = \max_{x \in S} x(A)$$

Indicator fn of S

$$f(x) \in \{0, +\infty\}$$

$$\rightarrow: g(p) = \max_{x \in S} \langle p, x \rangle = \max_x [\langle p, x \rangle - f(x)] = f^\bullet(p)$$

$$\leftarrow: f(x) = \max_p [\langle p, x \rangle - g(p)] = g^\bullet(x)$$

Lovász ext. of ρ

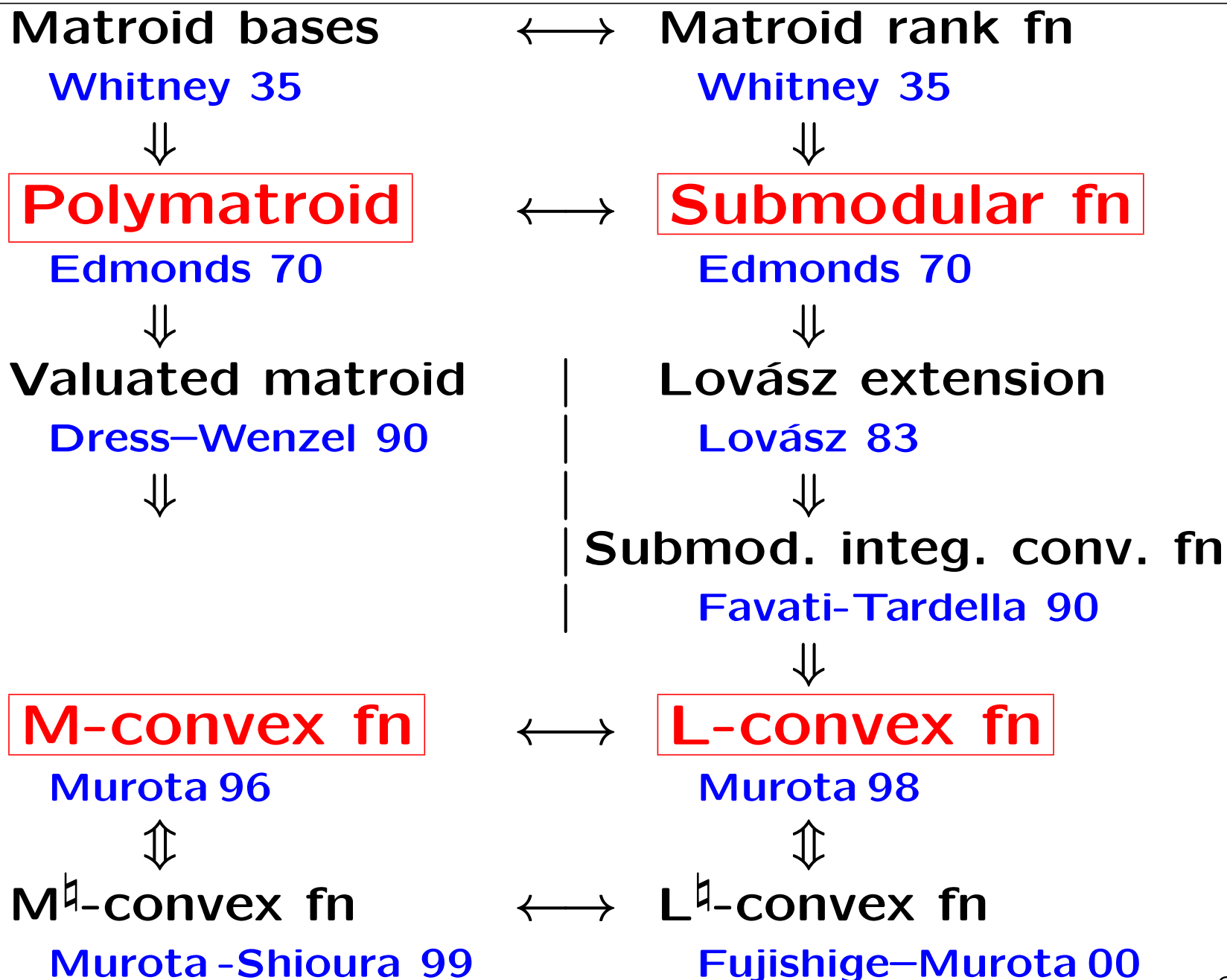
$$g(p)$$

Legendre transform

M[♯]-convex

L[♯]-convex

History of Discrete Conjugacy



Integral Subgradient & Biconjugacy

Subdifferential: $\partial f(x)$

$$= \{\mathbf{p} \in \mathbb{R}^n \mid f(y) - f(x) \geq \langle \mathbf{p}, y - x \rangle \ (\forall y)\}$$

\uparrow **subgradient**

Integral subdifferential: $\partial_{\mathbb{Z}} f(x) = \partial f(x) \cap \mathbb{Z}^n$

$$= \{\mathbf{p} \in \mathbb{Z}^n \mid f(y) - f(x) \geq \langle \mathbf{p}, y - x \rangle \ (\forall y)\}$$

\uparrow **integral subgradient**

Prop: $f : \mathbb{Z}^n \rightarrow \mathbb{Z} \cup \{+\infty\}$ (Murota 98)

If $\partial_{\mathbb{Z}} f(x) \neq \emptyset$ for all $x \in \text{dom } f$, then $f^{\bullet\bullet} = f$

$$f^{\bullet}(p) = \sup_{x \in \mathbb{Z}^n} [\langle p, x \rangle - f(x)]$$

Integ. Subgradient & Biconjugacy: Example

(Discrete life is not easy)

$$f : \mathbb{Z}^n \rightarrow \overline{\mathbb{Z}}$$

$$\partial_{\mathbb{Z}} f(x) \neq \emptyset ?$$

$$f^{\bullet\bullet} = f ?$$

Example: $D = \{(0, 0, 0), \pm(1, 1, 0), \pm(0, 1, 1), \pm(1, 0, 1)\}$

$$f(x_1, x_2, x_3) = \begin{cases} (x_1 + x_2 + x_3)/2, & x \in D, \\ +\infty, & \text{o.w.} \end{cases}$$

D is “convex”: $\text{conv}(D) \cap \mathbb{Z}^n = D$

$$\partial f_{\mathbb{R}}(0) = \{(1/2, 1/2, 1/2)\}$$

$$\partial_{\mathbb{Z}} f(0) = \emptyset$$

$$f^{\bullet\bullet}(0) = - \inf_{p \in \mathbb{Z}^3} \max\{0, |p_1 + p_2 - 1|, |p_2 + p_3 - 1|, |p_3 + p_1 - 1|\}$$

$$f^{\bullet\bullet}(0) = -1 \neq 0 = f(0)$$

Integral Subgradient & Biconjugacy

Thm:

(Murota 98)

f : integer-valued $L^{\mathbb{Z}}_- / M^{\mathbb{Z}}_- / L^{\mathbb{Z}}_{\leq 2} / M^{\mathbb{Z}}_{\leq 2}$ -convex

- Subdifferential $\partial f(x)$ is an integral polyhedron
- Hence integral subgradient p exists
- Hence $f^{\bullet\bullet} = f$

Thm:

(Murota-Tamura 18)

f : integer-valued **integrally convex**

- Subdifferential $\partial f(x)$ is NOT an integral polyhedron
- But integral subgradient p exists
- Hence $f^{\bullet\bullet} = f$

Conjugacy and Biconjugacy

Legendre trans: $f^\bullet(p) = \sup_{x \in \mathbb{Z}^n} [\langle p, x \rangle - f(x)]$
--

$f : \mathbb{Z}^n \rightarrow \bar{\mathbb{Z}}$	$f^\bullet : \mathbb{Z}^n \rightarrow \bar{\mathbb{Z}}$	$f^{\bullet\bullet} = f$
submodular (set fn)	submodular polyhedron $\{x \in \mathbb{Z}^n \mid x(A) \leq \rho(A)\}$	Y
separable-convex $f(x) = \sum \varphi_i(x_i)$	separable-convex $\varphi_1^\bullet(p_1) + \dots + \varphi_n^\bullet(p_n)$	Y
integrally-convex	Not integrally-convex (characterization: open)	Y
L-convex (\mathbb{Z}^n)	M-convex	Y
L[‡]-convex	M[‡]-convex	Y
M-convex (\mathbb{Z}^n)	L-convex	Y
M[‡]-convex	L[‡]-convex	Y

P5. Duality

(separation theorem)
(Fenchel duality)

Conjugacy/Duality in Matroids

Conjugacy

Exchange axiom \Leftrightarrow Submodularity of rank function

Duality

Matroid intersection theorem (Edmonds)

Discrete separation (Frank)

Fenchel-type duality (Fujishige)

Matroid Intersection Problem

Given two matroids

- Find a common indep. set X with $\max |X|$
- Find a common base B (if any)

Given two matroids and weight w

- Find a common indep. set X with $\max w(X)$
- Find a common base B with $\max w(B)$

LP formulation with integrality, combinatorial duality,
efficient algorithms, many applications

Edmonds' Intersection Theorem

Submodular polyhedron $(\rho(\emptyset) = 0, \rho(V) < +\infty)$

$$P(\rho) = \{x \in \mathbb{R}^n \mid x(X) \leq \rho(X) \ (\forall X \subseteq V)\} \quad (|V| = n)$$

Theorem:

(Edmonds 70)

(1) For $\rho_1, \rho_2 : 2^V \rightarrow \bar{\mathbb{R}}$: submodular,

$$\max_x \{x(V) \mid x \in P(\rho_1) \cap P(\rho_2)\} = \min_X \{\rho_1(X) + \rho_2(V \setminus X)\}$$

(2) If ρ_1 and ρ_2 are integer-valued, then

$$P(\rho_1) \cap P(\rho_2) = \overline{P(\rho_1) \cap P(\rho_2) \cap \mathbb{Z}^n}$$

and there exists $x^* \in \mathbb{Z}^n$ that attains the maximum

Frank's Discrete Separation

(Frank 82)

$\rho : 2^V \rightarrow \mathbb{R}$: submodular

$$(\rho(\emptyset) = 0)$$

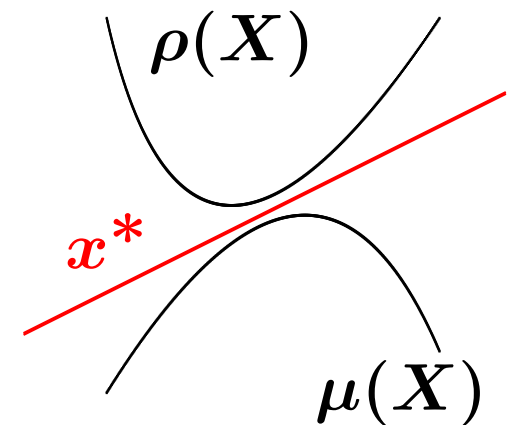
$\mu : 2^V \rightarrow \mathbb{R}$: supermodular

$$(\mu(\emptyset) = 0)$$

• $\rho(X) \geq \mu(X) \quad (\forall X \subseteq V) \Rightarrow \exists x^* \in \mathbb{R}^V$:

$$\rho(X) \geq x^*(X) \geq \mu(X) \quad (\forall X \subseteq V)$$

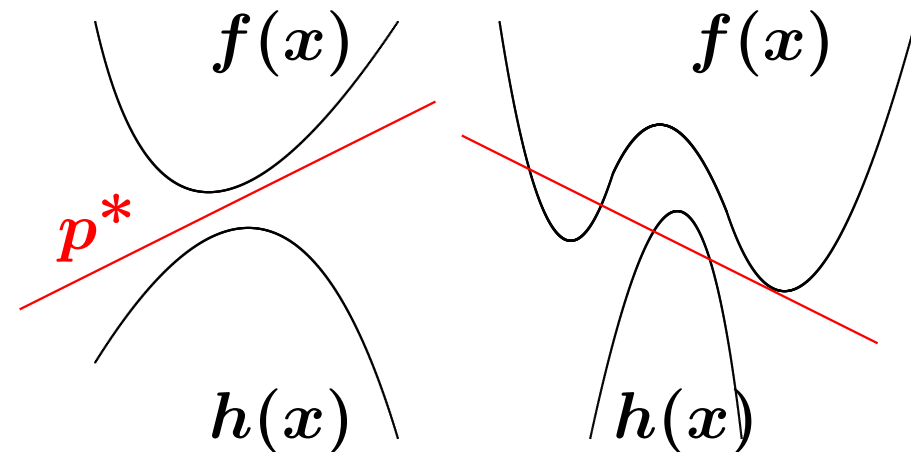
• ρ, μ : **integer-valued** $\Rightarrow x^* \in \mathbb{Z}^V$



Discrete Separation Theorem

$f : \mathbb{Z}^n \rightarrow \mathbb{R}$ “convex”

$h : \mathbb{Z}^n \rightarrow \mathbb{R}$ “concave”



• $f(x) \geq h(x) \quad (\forall x \in \mathbb{Z}^n) \Rightarrow \exists \alpha^* \in \mathbb{R}, \exists p^* \in \mathbb{R}^n:$

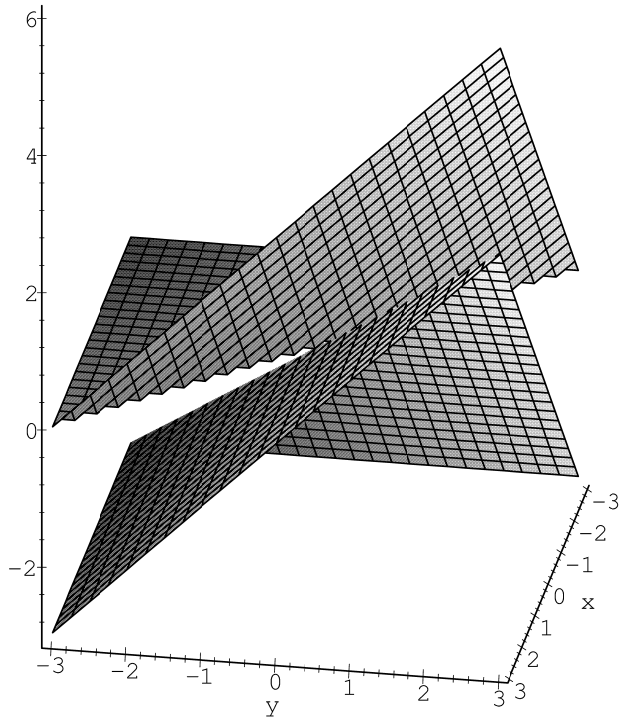
$$f(x) \geq \alpha^* + \langle p^*, x \rangle \geq h(x) \quad (x \in \mathbb{Z}^n)$$

• f, h : **integer-valued** $\Rightarrow \alpha^* \in \mathbb{Z}, p^* \in \mathbb{Z}^n$

Difficulty of Discrete Separation (1)

$$f(x, y) = \max(0, x + y) \quad \text{convex}$$

$$h(x, y) = \min(x, y) \quad \text{concave}$$



**nonintegral
separation**

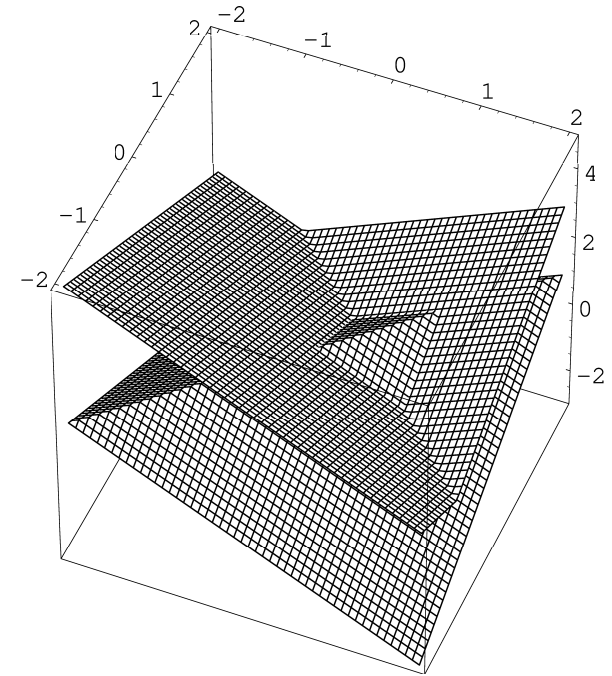
$p^* = (1/2, 1/2), \alpha^* = 0$ unique separating plane

Difficulty of Discrete Separation (2)

Even real-separation is nontrivial

$$f(x, y) = |x + y - 1| \quad \text{convex}$$

$$h(x, y) = 1 - |x - y| \quad \text{concave}$$

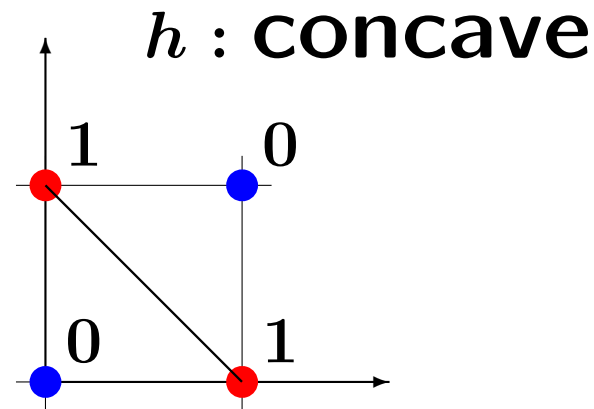
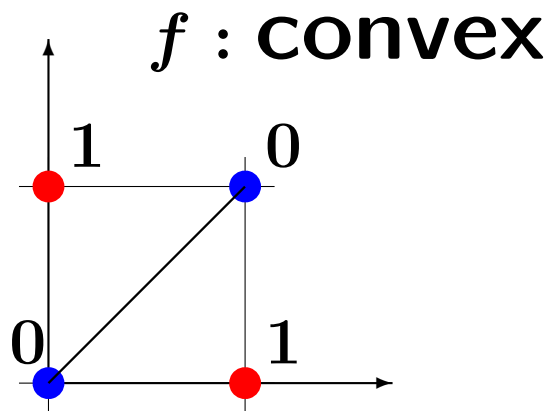


- $f(x, y) \geq h(x, y) \quad (\forall (x, y) \in \mathbb{Z}^2) \quad \text{true}$
- **No** $\alpha^* \in \mathbb{R}, p^* \in \mathbb{R}^2: \quad f(x) \geq \alpha^* + \langle p^*, x \rangle \geq h(x)$
 $\because f = 0 < h = 1 \quad \text{at} \quad (x, y) = (1/2, 1/2)$

Difficulty of Discrete Separation (3)

Set function \iff Function on $\{0, 1\}^n$

Every set function $\{0, 1\}^n \rightarrow \mathbb{R}$ can be extended to convex/concave function



- $f(x, y) = h(x, y) \quad (\forall (x, y) \in \{0, 1\}^2)$
- **No separation possible**

Discrete Separation Theorems

(Murota 96/98)

M-separation Thm (for M^{\natural} -convex)

⇒ Weight splitting for weighted matroid intersection
(Iri-Tomizawa 76, Frank 81)
(linear fn, indicator fn = M^{\natural} -convex fn)

L-separation Thm (for L^{\natural} -convex)

⇒ Discrete separ. for submod. set fn (Frank 82)
(submod. set fn = L^{\natural} -convex fn on 0-1 vectors)

Min-Max Duality

f : M^{\natural} -convex, h : M^{\natural} -concave $(\mathbb{Z}^n \rightarrow \bar{\mathbb{Z}})$

Legendre–Fenchel transform

$$f^{\bullet}(p) = \sup\{\langle p, x \rangle - f(x) \mid x \in \mathbb{Z}^n\}$$

$$h^{\circ}(p) = \inf\{\langle p, x \rangle - h(x) \mid x \in \mathbb{Z}^n\}$$

Fenchel-type duality thm (Murota 96, 98)

$$\inf_{x \in \mathbb{Z}^n} \{f(x) - h(x)\} = \sup_{p \in \mathbb{Z}^n} \{h^{\circ}(p) - f^{\bullet}(p)\}$$

self-conjugate $(f^{\bullet}: L^{\natural}$ -convex, $h^{\circ}: L^{\natural}$ -concave)

\implies Edmonds' intersection thm
Fujishige's Fenchel duality thm

Relation among Duality Thms

Discrete Convex

Combinatorial Opt.

M-separation

$$f(x) \geq \boxed{\text{Lin}} \geq h(x)$$



Fenchel duality

$$\inf\{f - h\} \\ = \sup\{h^\circ - f^\bullet\}$$



L-separation

$$f^\bullet(p) \geq \boxed{\text{Lin}} \geq h^\circ(p)$$

Fenchel duality (Fujishige 84)
matroid intersect. (Edmonds 70)



\Rightarrow **discrete separ. for submod**
(Frank 82)

\Rightarrow **valuated matroid intersect.**
(M. 96)



weighted matroid intersect.

(Edmonds 79, Iri-Tomizawa 76,
Frank 81)

Separation and Min-Max Theorems

	separation	min-max
submodular (set fn)	Y (Frank)	Y (Edmonds, Fujishige)
separable-conv	Y	Y
integrally-conv	N	N
L-conv (\mathbb{Z}^n)	Y	Y
M-conv (\mathbb{Z}^n)	Y	Y

Summary

	Operations				Minimize		Conjugacy/Duality			
	sca lng	sum	cnvl tion	graf tran	loc glob	prox imity	cnv ext	bi- cnj	sep thm	min max
submod (set fn)	—	Y	N	Y*	Y	—	Y	Y	Y	Y
separ -conv	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y
integ -conv	N	N	N	N	Y	Y*	Y	Y	N	N
L-conv (\mathbb{Z}^n)	Y	Y	N	Y*	Y	Y	Y	Y	Y	Y
M-conv (\mathbb{Z}^n)	N	N	Y	Y	Y	Y	Y	Y	Y	Y
M-conv (jump)	N	N	Y	Y	Y	?	N	N	N	N
L-conv (tree ⁿ)	?	Y	—	—	Y	Y*	Y*	?	?	?

Summary

	Operations				Minimize		Conjugacy/Duality			
	sca lng	sum	cnvl tion	graf tran	loc glob	prox imity	cnv ext	bi- cnj	sep thm	min max
submod (set fn)	—	Y	N	Y*	Y	—	Y	Y	Y	Y
separ -conv	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y
integ -conv	N	N	N	N	Y	Y*	Y	Y	N	N
L-conv (\mathbb{Z}^n)	Y	Y	N	Y*	Y	Y	Y	Y	Y	Y
M-conv (\mathbb{Z}^n)	N	N	Y	Y	Y	Y	Y	Y	Y	Y
M-conv (jump)	N	N	Y	Y	Y	?	N	N	N	N
L-conv (tree ⁿ)	?	Y	—	—	Y	Y*	Y*	?	?	?

Summary

	Operations				Minimize		Conjugacy/Duality			
	sca lng	sum	cnvl tion	graf tran	loc glob	prox imity	cnv ext	bi- cnj	sep thm	min max
submod (set fn)	—	Y	N	Y*	Y	—	Y	Y	Y	Y
separ -conv	Y	Y	Y	Y	Y	Y	Y	Y	Y	Y
integ -conv	N	N	N	N	Y	Y*	Y	Y	N	N
L-conv (\mathbb{Z}^n)	Y	Y	N	Y*	Y	Y	Y	Y	Y	Y
M-conv (\mathbb{Z}^n)	N	N	Y	Y	Y	Y	Y	Y	Y	Y
M-conv (jump)	N	N	Y	Y	Y	?	N	N	N	N
L-conv (tree ⁿ)	?	Y	—	—	Y	Y*	Y*	?	?	?

Five Properties of “Convex” Functions

1. convex extension
2. local opt = global opt
3. Conjugacy (Legendre transform)
4. separation theorem
5. Fenchel duality

hold for

- separable-convex functions
- L^{\natural} -convex functions
- M^{\natural} -convex functions

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