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Convex Analysis Approach to Discrete Optimization, II

Properties of Discrete Convex Functions

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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)

P1.

Convex Extension

Convex Extension

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

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

Theorem:

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

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)

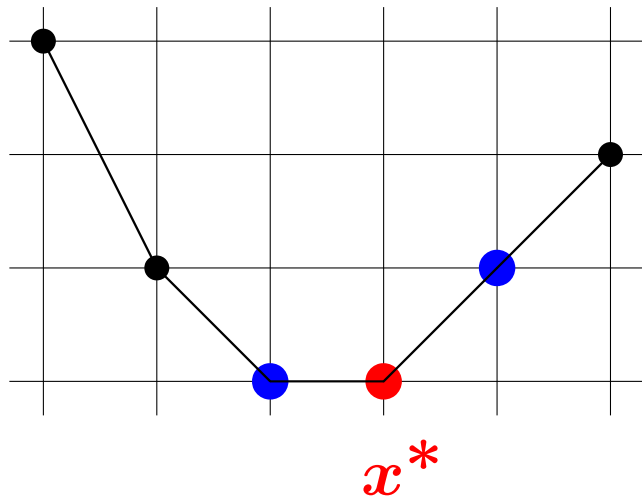
Local vs Global Optimality ($n = 1$)

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

x^* : global opt (min)

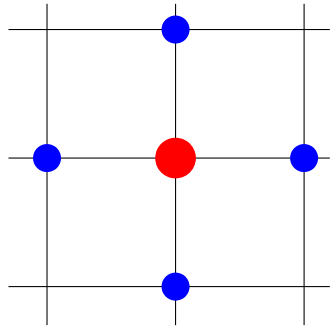
$\iff x^*$: local opt (min)

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



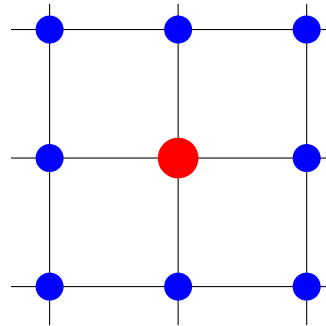
Neighborhood for Local Optimality

separable
convex



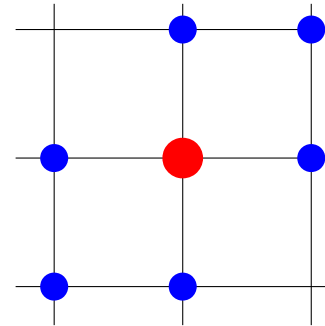
$$2n + 1$$

integrally
convex



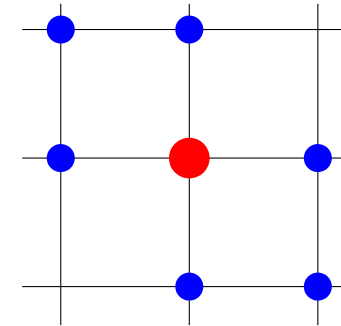
$$3^n$$

L^{\square} -
convex



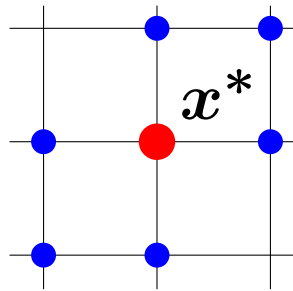
$$2^{n+1} - 1$$

M^{\square} -
convex



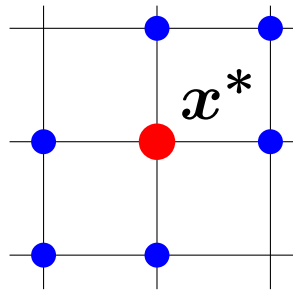
$$n(n + 1) + 1$$

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:** $\min_y f(x, y)$
- **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**

Sum and Convolution

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

Theorem:

(Murota 98)

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

$$f_1, f_2, \dots, f_k : \mathbb{L} (\mathbb{L}^\natural) \implies f_1 + f_2 + \dots + f_k : \mathbb{L} (\mathbb{L}^\natural)$$

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

Theorem:

(Murota 96)

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

$$f_1, f_2, \dots, f_k : \mathbb{M} (\mathbb{M}^\natural) \implies f_1 \square f_2 \square \dots \square f_k : \mathbb{M} (\mathbb{M}^\natural)$$

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

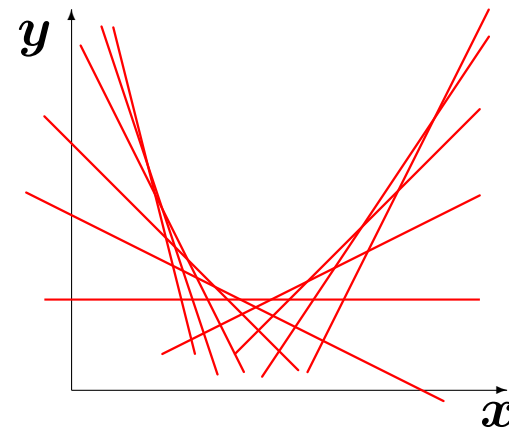
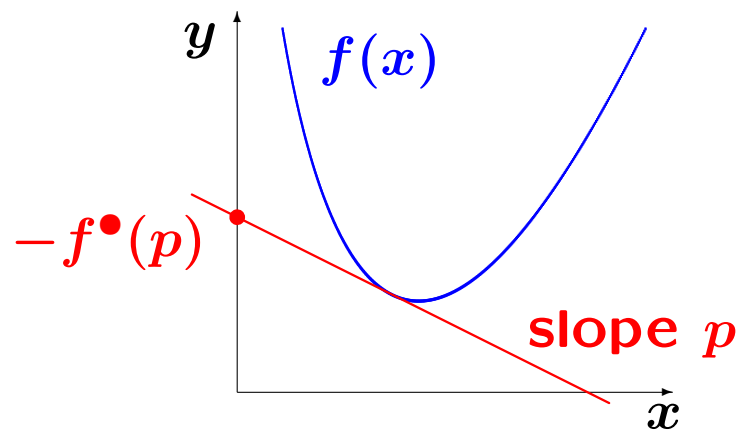
Sum/Convolution

	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

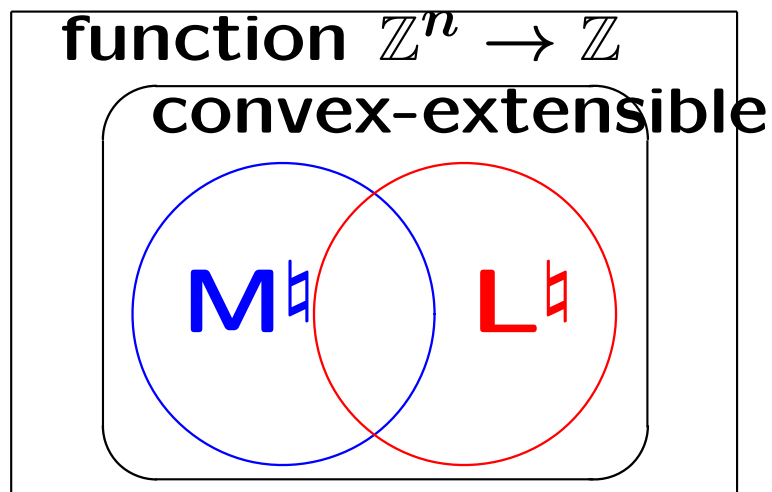
Integer-valued discrete fn $f : \mathbb{Z}^n \rightarrow \overline{\mathbb{Z}}$

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

Conjugacy and Biconjugacy

$f : \mathbb{Z}^n \rightarrow \bar{\mathbb{Z}}$	$f^\bullet : \mathbb{Z}^n \rightarrow \bar{\mathbb{Z}}$	$f^{\bullet\bullet} = f$
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	N	N
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

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

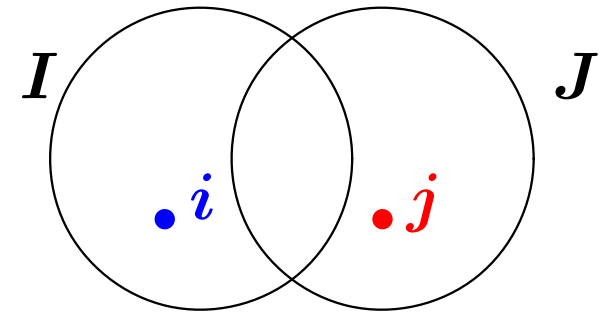
Equivalence $\mathcal{B} \iff \rho$

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

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

Axioms of Matroid

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



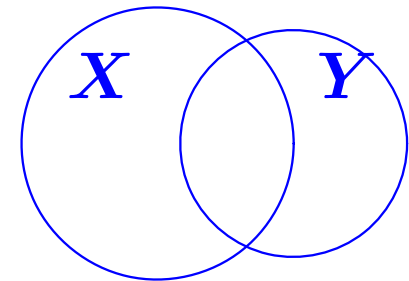
$$\forall I, J \in \mathcal{B}, i \in I \setminus J, \exists j \in J \setminus I:$$
$$I - i + j \in \mathcal{B}, J + i - j \in \mathcal{B}$$

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 by Legendre tr.)

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

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

Dual Character of Matroid Rank

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

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

Edmonds' matroid union formula:

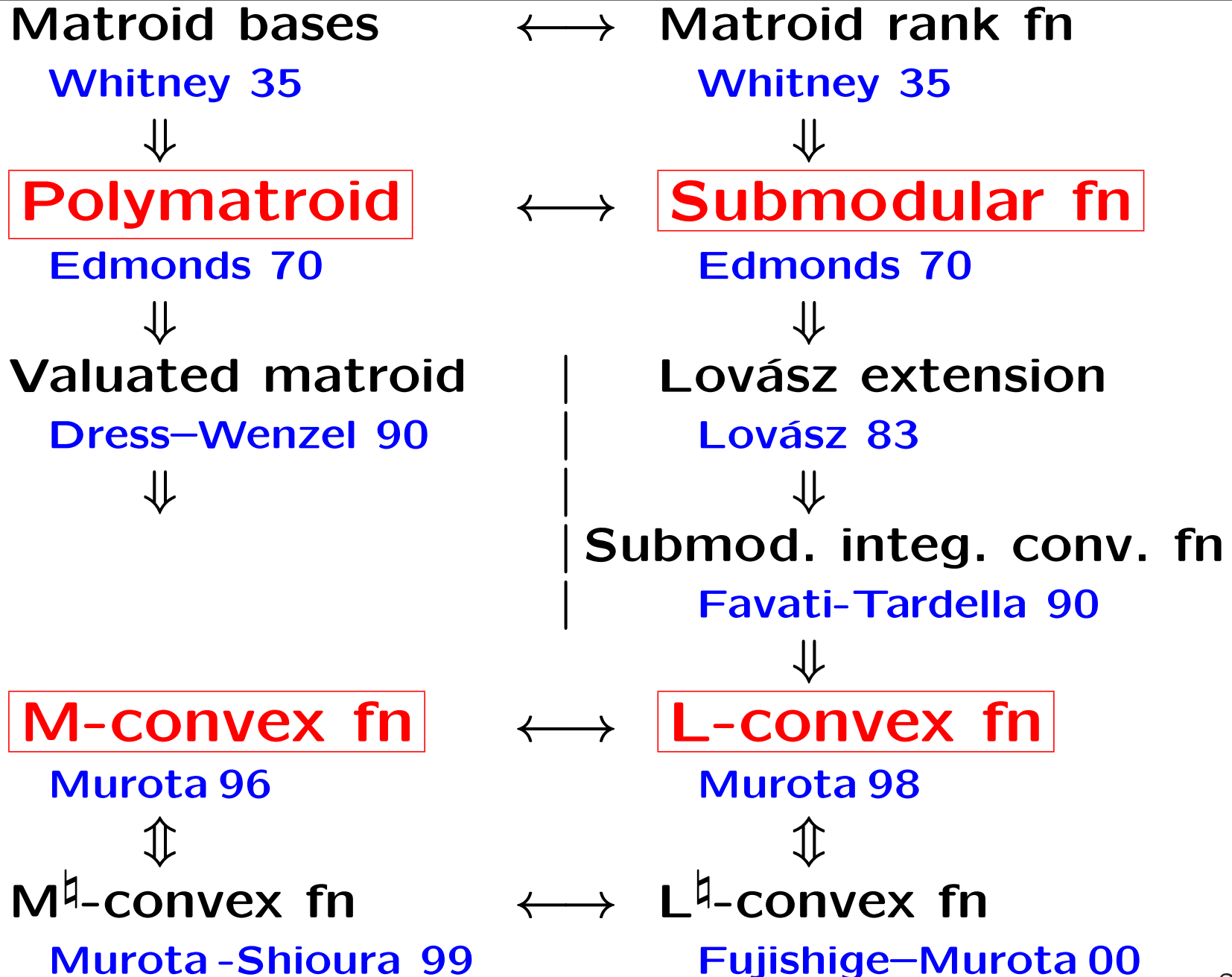
$$\max_X \{\rho_1(X) + \rho_2(V \setminus X)\} = \min_Y \{\rho_1(Y) + \rho_2(Y) + |V \setminus Y|\}$$

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

submod minimization
(L[♠]-convex $+$ L[♠]-convex)

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

History of Discrete Conjugacy



P5. Duality

(separation theorem)
(Fenchel duality)

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

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

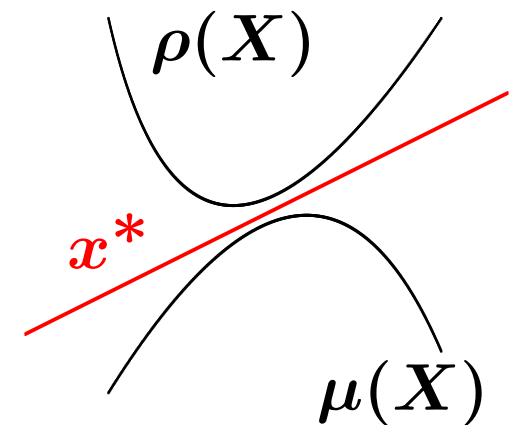
$\mu : 2^V \rightarrow \mathbb{R}$: supermodular

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

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

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

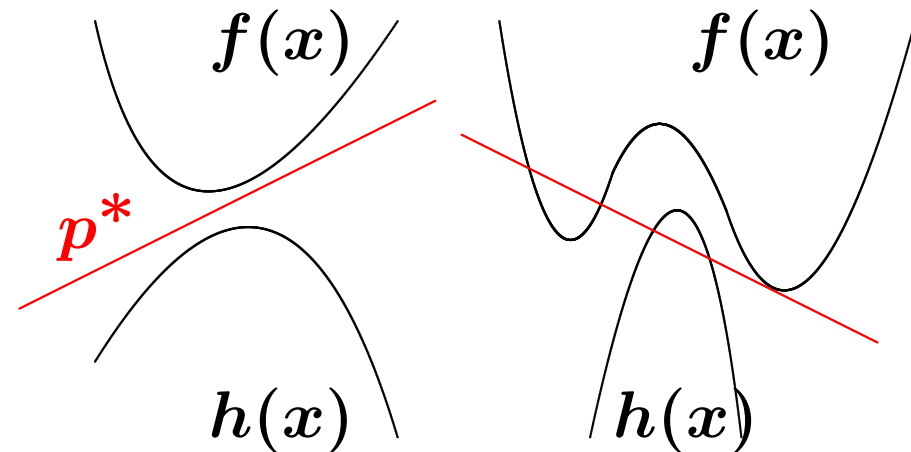
• ρ, μ : **integer-valued** $\Rightarrow x^* \in \mathbb{Z}^V$



Discrete Separation Theorem

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

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



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

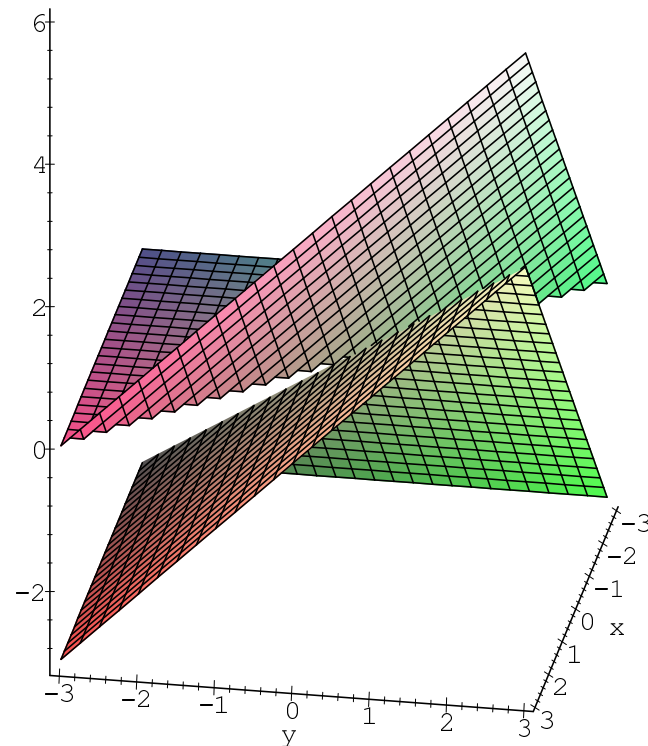
$$f(x) \geq \alpha^* + \langle p^*, x \rangle \geq h(x) \quad (x \in \mathbb{Z}^n)$$

• 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**

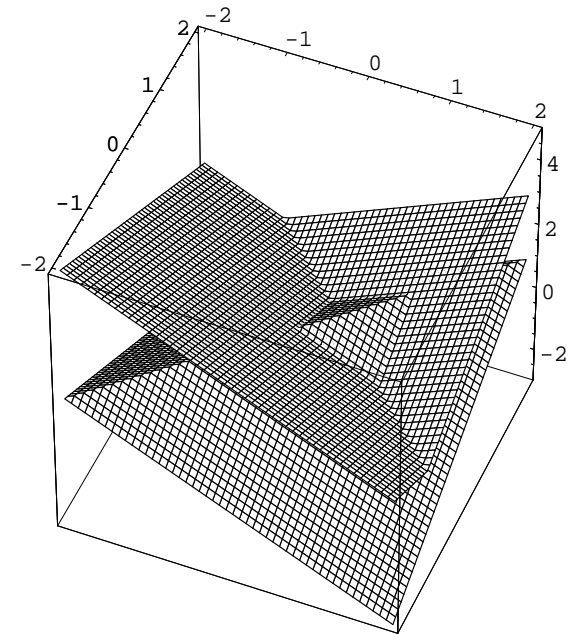
$p^* = (1/2, 1/2), \alpha^* = 0$ unique separating plane

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** $\alpha^* \in \mathbb{R}, p^* \in \mathbb{R}^2: \quad f(x) \geq \alpha^* + \langle p^*, x \rangle \geq h(x)$
 $\because f = 0 < h = 1 \quad \text{at} \quad (x, y) = (1/2, 1/2)$

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° : 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	Y*	N	N	N	Y	Y*	Y	N	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	Y*	N	N	N	Y	Y*	Y	N	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^q -convex functions
- M^q -convex functions

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