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Discrete Convex Analysis II: Properties of Discrete Convex Functions

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Features of [Discrete] Convex Functions

- Occurrence in many models

motivations, applications

- Operations and transformations

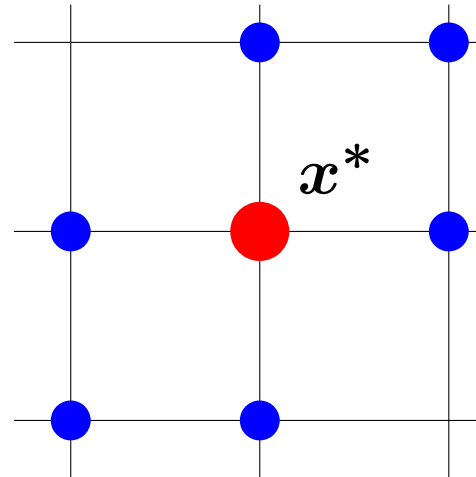
- Sufficient structure for a theory

mathematically beautiful, practically useful

- Minimization algorithms \implies **Part III**

Local min = Global min

	=?	neighbors	poly-time alg
submodular (set fn)	–		
separable-conv	Y		
integrally-conv	Y		
L-conv (\mathbb{Z}^n)	Y		
M-conv (\mathbb{Z}^n)	Y		



Contents of Part II

Properties of Discrete Convex Functions

P1. Convex Extension

P2. Operations

P3. Conjugacy

P4. Duality

P1.

Convex Extension

Convex Extension

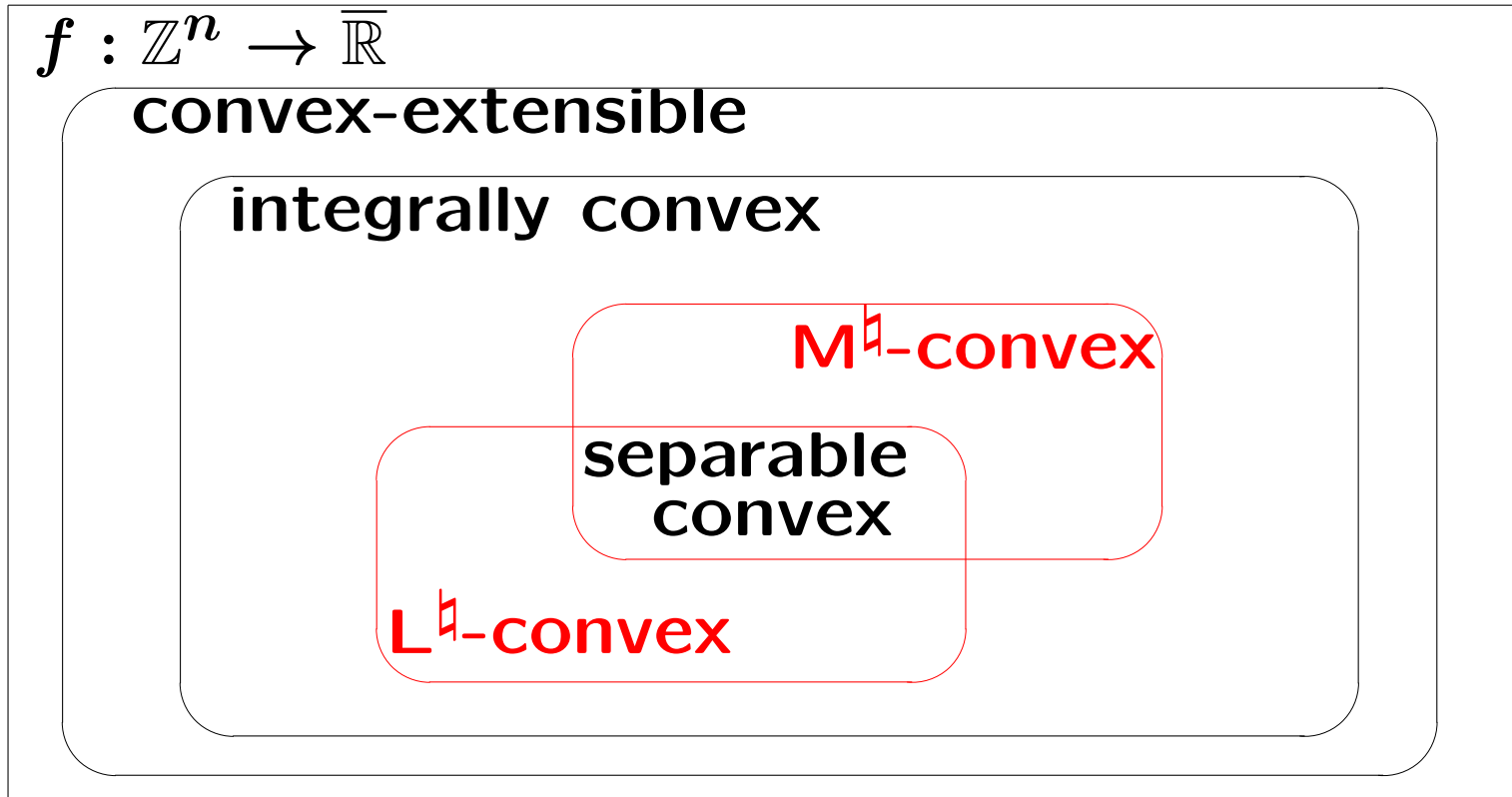
$f : \mathbb{Z}^n \rightarrow \bar{\mathbb{R}}$ is **convex-extensible**

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

Theorem:

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

Classes of Discrete Convex Functions



P2.

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**

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	Y* ($n = 2$)	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

Sum and Convolution

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

Theorem:

(Murota 98)

$$\begin{array}{l} f_1, f_2 : \text{L-convex} \quad \implies \quad f_1 + f_2 : \text{L-convex} \\ \text{L}^\natural\text{-convex} \quad \implies \quad \text{L}^\natural\text{-convex} \end{array}$$

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

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

Theorem:

(Murota 96)

$$\begin{array}{l} f_1, f_2 : \text{M-convex} \quad \implies \quad f_1 \square f_2 : \text{M-convex} \\ \text{M}^\natural\text{-convex} \quad \implies \quad \text{M}^\natural\text{-convex} \end{array}$$

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

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

Significance of M-convolution Thm

Concave convolution:

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

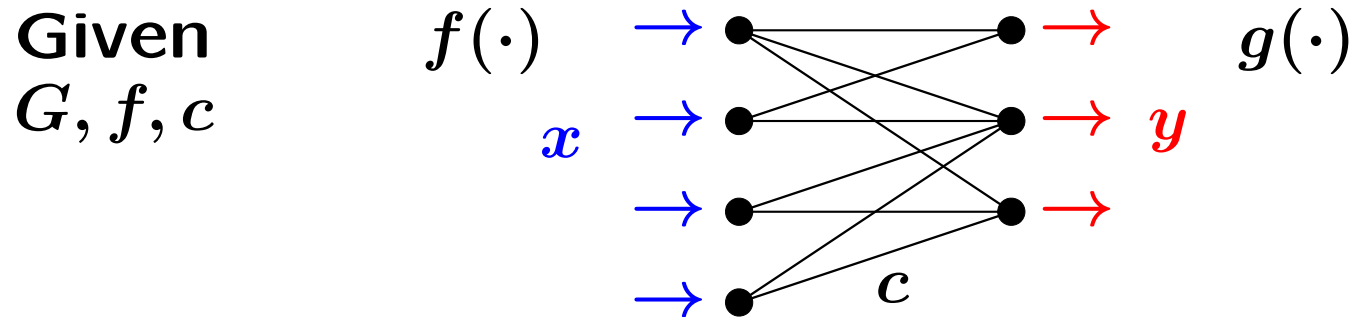
U_1, U_2, \dots, U_k : gross-substitute (M[♯]-concave)

⇒ **aggregated utility** $U_1 \square U_2 \square \dots \square U_k$ **is**
gross-substitute (M[♯]-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

Transformation by Graph/Network



$$\min\{f(x) + c(\xi) \mid x \longleftrightarrow y\} =: g(y)$$



$$\exists \text{ flow } \xi : \partial\xi = x \oplus (-y)$$

$$\exists \text{ matching } M : \partial M = X \cup Y$$

Theorem:

(Murota 96)

$$f : \text{M-convex} \implies g : \text{M-convex}$$

$$\text{M}^{\natural}\text{-convex} \implies \text{M}^{\natural}\text{-convex}$$

Independent Assignment Valuation

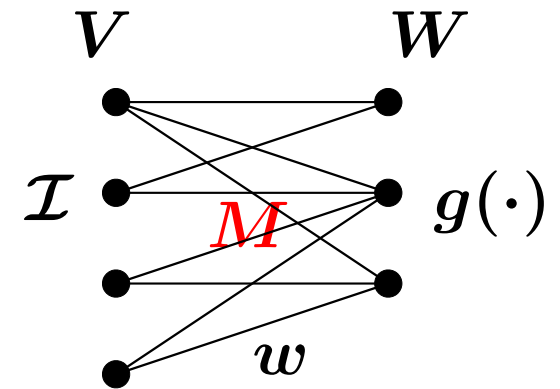
$G = (V, W; E)$: bipartite graph, w : edge weight

(V, \mathcal{I}) : matroid (\mathcal{I} : indep. sets)

Corollary:

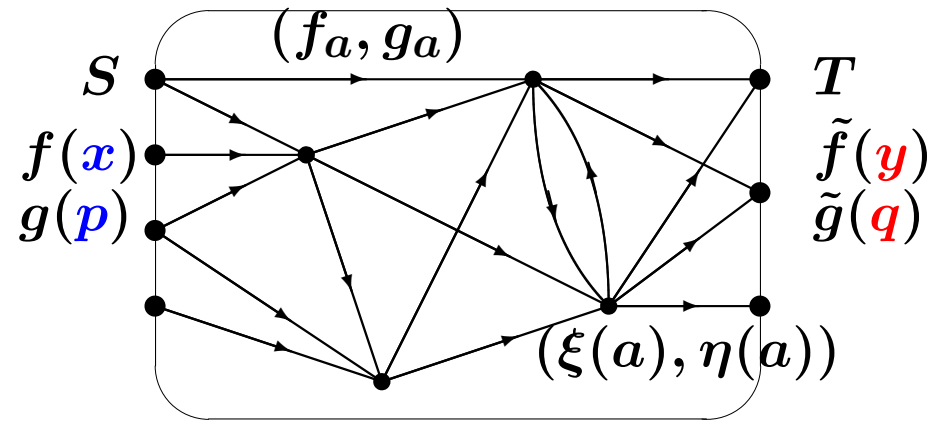
$$g(Y) = \max\{w(M) \mid M: \text{matching}, \\ V \cap \partial M \in \mathcal{I}, W \cap \partial M = Y\}$$

is M^\sharp -concave



Transformation by Network

x, y : flow
 p, q : potential



M-convex (\mathbb{Z}^n): $(y \in \mathbb{Z}^T)$

$$\tilde{f}(y) = \inf_{\xi, x} \left\{ f(x) + \sum_{a \in A} f_a(\xi(a)) \mid \partial \xi = (x, -y, 0), \right. \\ \left. \xi \in \mathbb{Z}^A, (x, -y, 0) \in \mathbb{Z}^S \times \mathbb{Z}^T \times \mathbb{Z}^{V \setminus (S \cup T)} \right\}$$

L-convex (\mathbb{Z}^n): $(q \in \mathbb{Z}^T)$

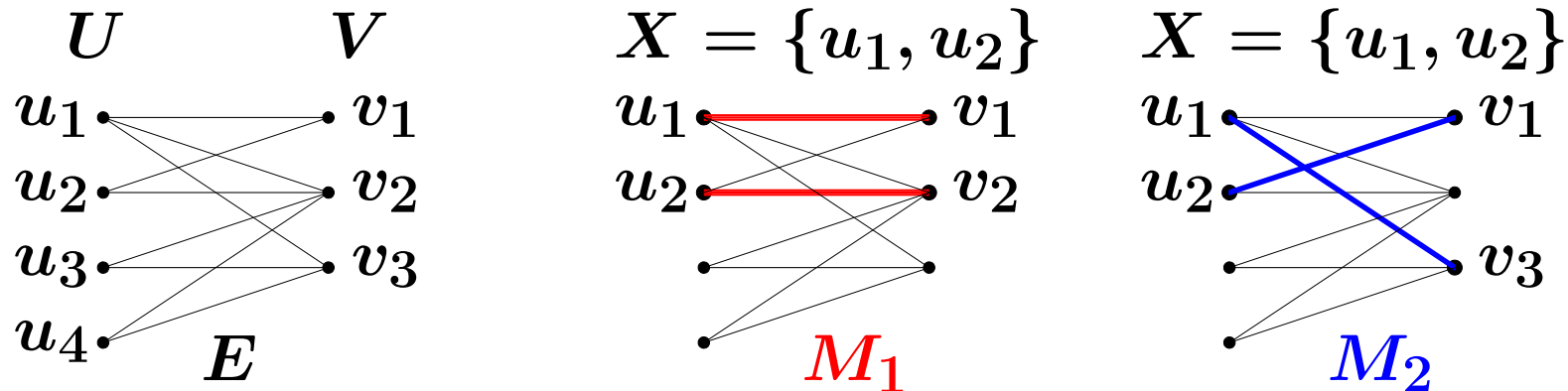
$$\tilde{g}(q) = \inf_{\eta, p, r} \left\{ g(p) + \sum_{a \in A} g_a(\eta(a)) \mid \eta = -\delta(p, q, r), \right. \\ \left. \eta \in \mathbb{Z}^A, (p, q, r) \in \mathbb{Z}^S \times \mathbb{Z}^T \times \mathbb{Z}^{V \setminus (S \cup T)} \right\}$$

Transformation by Graphs/Networks

submodular (set fn)	Y (conjugate formula) matroid \mapsto matroid
separable-conv	N
integrally-conv	N
L-conv (\mathbb{Z}^n)	Y (conjugate formula)
M-conv (\mathbb{Z}^n)	Y

Representation Problem

Matching / Assignment



Max weight for $X \subseteq U$ (w : given weight)

$$f(X) = \max \left\{ \sum_{e \in M} w(e) \mid M: \text{matching}, U \cap \partial M = X \right\}$$

Max-weight func f is M^{\sharp} -concave (cf. Murota 1996)

- Proof by augmenting path

Assignment valuation is GS (cf. Hatfield-Milgrom 2005)

Endowed Assignment Valuation

Assignment valuation (AV):

$$f_w(X) = \max w\text{-weight of a matching covering } X$$

Endowed assignment valuation (EAV):

$$f_{w,T}(X) = f_w(X \cup T) - f_w(T)$$

$$\mathbf{AV} \subseteq \mathbf{EAV} \subseteq \mathbf{GS}$$

(Hatfield-Milgrom 05)

$$\mathbf{Thm:} \quad \mathbf{EAV} \neq \mathbf{GS}$$

(Ostrovsky-Paes Leme 15)

Proof ideas

- transversal matroids are base orderable (strong exch)
- cycle matroid $\mathcal{M}(K_4)$ is not base orderable
- rank of $\mathcal{M}(K_4) \in \mathbf{GS} \setminus \mathbf{EAV}$

Matroid-based Valuation (MBV)

MBV = (Ostrovsky-Paes Leme 15)

(0) Start with a weighted matroid valuation:

$$f(X) = \max\{w(Y) \mid Y: \text{indep} \subseteq X\}$$

(*) Repeat

(i) convolution: $\tilde{f}(X) = \max_{Y \subseteq X} (f_1(Y) + f_2(X \setminus Y))$

(ii) contraction: $\tilde{f}(X) = f(X \cup T) - f(T)$

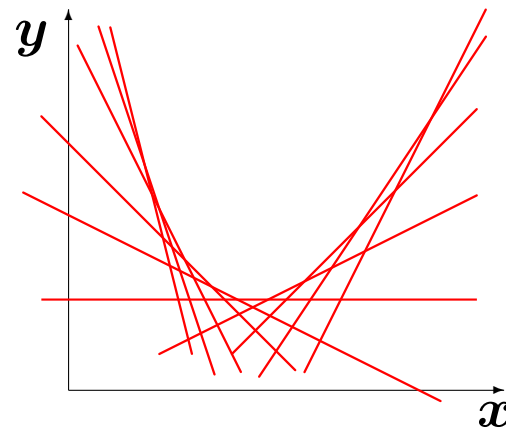
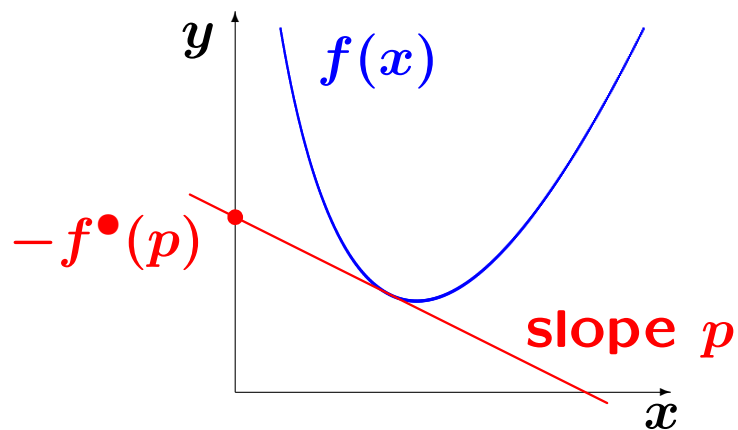
$$\text{EAV} \subsetneq \text{MBV} \subseteq \text{GS}$$

Conj.: MBV = GS (Ostrovsky-Paes Leme 15)

P3.

Conjugacy

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

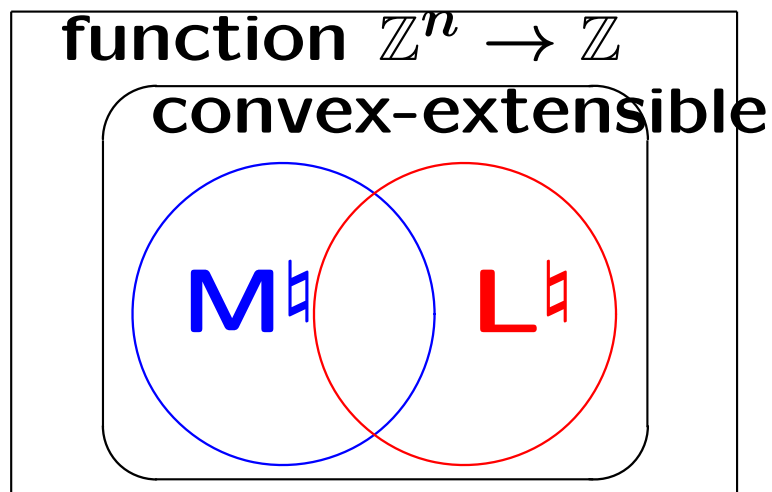
Integer-valued discrete fn $f : \mathbb{Z}^n \rightarrow \overline{\mathbb{Z}}$

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

Conjugacy in Polymatroids

Polyhedron S

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

Submodular fn ρ

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

Dual Character of Matroid Rank

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

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

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

$$\begin{aligned} \text{Prf: } \rho^\bullet(\chi_X) &= \max_Y \{|X \cap Y| - \rho(Y)\} \\ &= \max_{Y \supseteq X} \{|X \cap Y| - \rho(Y)\} = |X| - \rho(X) \end{aligned}$$

$$\rho:\text{subm} \Rightarrow \rho:\text{L}^\sharp\text{-conv} \Rightarrow \rho^\bullet:\text{M}^\sharp\text{-conv} \Rightarrow \rho:\text{M}^\sharp\text{-concave}$$

Significance of Conjugacy

- **Economics (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)
Toland–Singer duality

Implication in Economics

$U(x) : \mathbb{Z}^n \rightarrow \mathbb{Z} \cup \{-\infty\}$: utility

$V(p) = \max_x \{U(x) - \langle p, x \rangle\}$: indirect utility

Theorem: (Murota-Shioura-Yang 16)

Utility $U(x)$ is gross-substitute (M^{\natural} -concave)

\iff Indirect utility $V(p)$ is L^{\natural} -convex

Application to Auction:

Ascending auction

\Rightarrow Lyapunov fn min. algorithm (Ausubel 06)

\Rightarrow L^{\natural} -convex fn min. algorithm \implies **Part III**

\Rightarrow Precise analysis (M.-Shioura-Yang 16)

Conjugacy/Duality in Matroids

Conjugacy

Exchange axiom \Leftrightarrow Submodularity of rank function

Duality

Matroid intersection theorem (Edmonds)

Discrete separation (Frank)

Fenchel-type duality (Fujishige)

P4. Duality

Convex & Concave = Linear

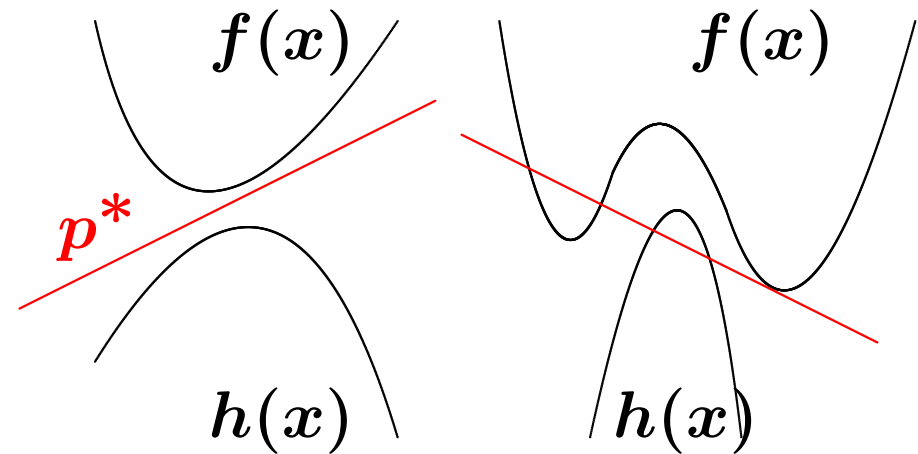
$$f, -f: \text{convex} \implies f(x) = \langle p, x \rangle + \alpha$$

submodular (set) (\mathbb{Z}^n)	Y N
separable-conv	Y
integrally-conv	N (every set fn is int-conv)
L-conv (\mathbb{Z}^n)	Y
M-conv (\mathbb{Z}^n)	Y

Discrete Separation Theorem

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

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



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

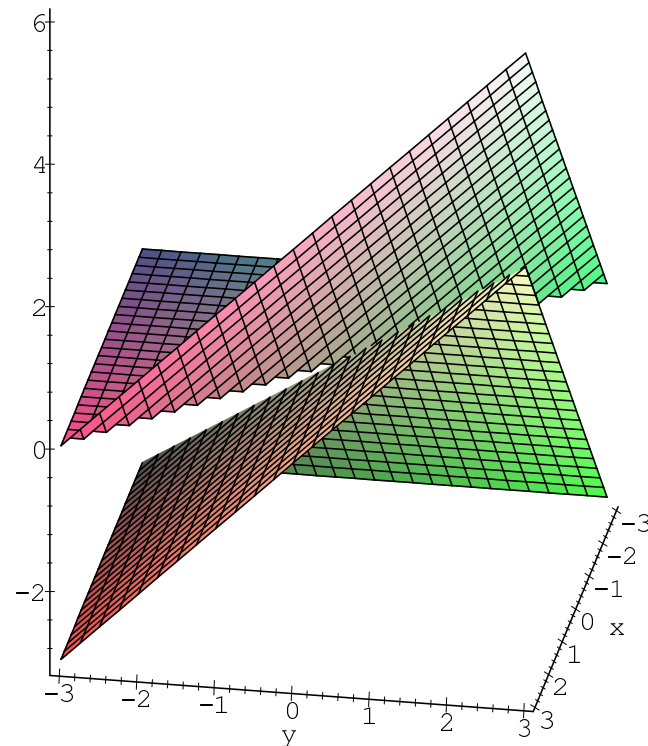
$$f(x) \geq \alpha^* + \langle p^*, x \rangle \geq h(x) \quad (x \in \mathbb{Z}^n)$$

• 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**

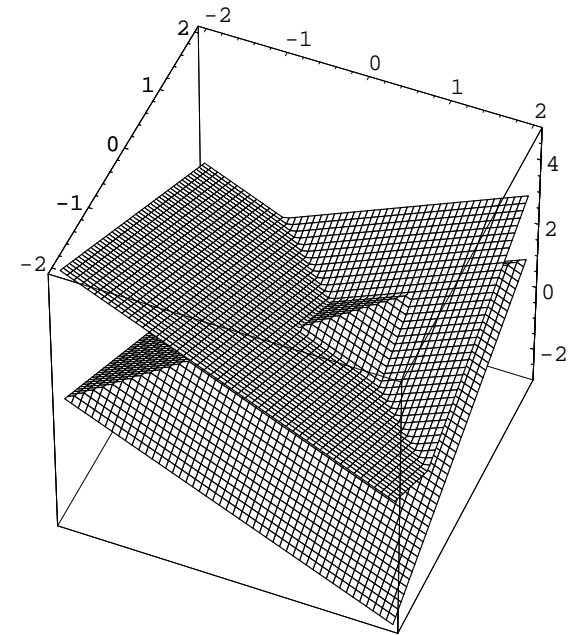
$p^* = (1/2, 1/2), \alpha^* = 0$ unique separating plane

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

Frank's Discrete Separation

(Frank 82)

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

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

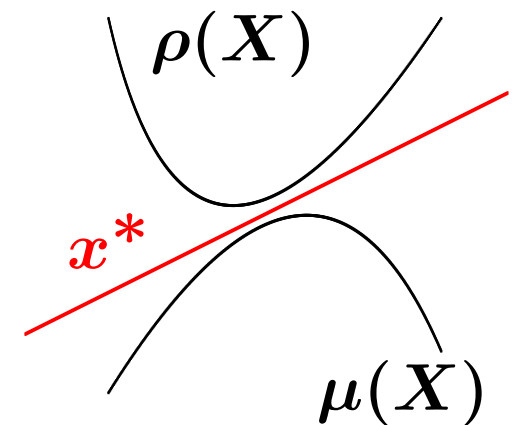
$\mu : 2^V \rightarrow \mathbb{R}$: supermodular

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

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

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

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



Discrete Separation 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)

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

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

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.

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)

DCA View on Matroid Union Formula

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

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

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

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 + M[♠]-concave)

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

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
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

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
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

Five Properties of “Convex” Functions

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

hold for

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

E N D