

Hausdorff School: Economics and Tropical Geometry  
Bonn, May 9-13, 2016

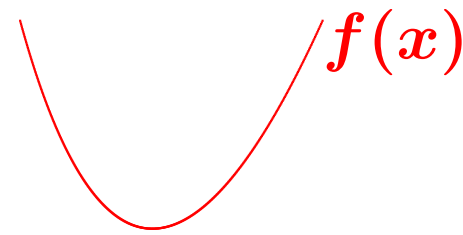
# **Discrete Convex Analysis I:**

## **Concepts of Discrete Convex Functions**

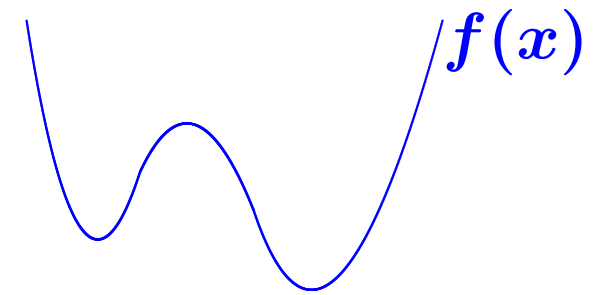
**Kazuo Murota**

**(Tokyo Metropolitan University)**

# Convex Function



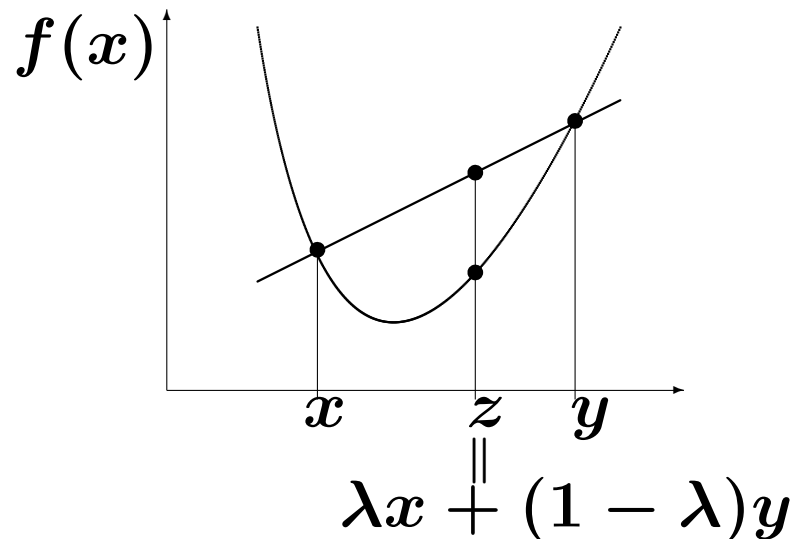
**convex**



**nonconvex**

$f : \mathbb{R}^n \rightarrow \bar{\mathbb{R}}$  is convex  $\iff$

$$\lambda f(x) + (1 - \lambda)f(y) \geq f(\lambda x + (1 - \lambda)y) \quad (0 < \forall \lambda < 1)$$



$$\bar{\mathbb{R}} = \mathbb{R} \cup \{+\infty\}$$

# Features of Convex Functions

- Occurrence in many models

motivations, applications

- Operations and transformations

- Sufficient structure for a theory

mathematically beautiful, practically useful

- Minimization algorithms

# Features of Convex Functions

**= Issues in discrete convex analysis**

- Occurrence in many models ?

motivations, applications

- Operations and transformations ?

- Sufficient structure for a theory ?

mathematically beautiful, practically useful

- Minimization algorithms ?

# Contents of Part I

## **C** Concepts of Discrete Convex Functions

**C1.** Univariate Discrete Convex Functions

**C2.** Classes of Discrete Convex Functions

**C3.** L-convex Functions

**C4.** M-convex Functions

**C5.** Remarks on Submodular Set Functions

**Part II: Properties,      Part III: Algorithms**

# C1.

## Univariate

## Discrete Convex Functions

**Ingredients of convex analysis**

# Definition of Convex Function

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

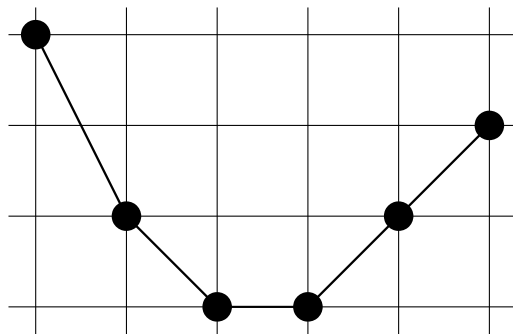
$$\overline{\mathbb{R}} = \mathbb{R} \cup \{+\infty\}$$

$$f(x-1) + f(x+1) \geq 2f(x)$$

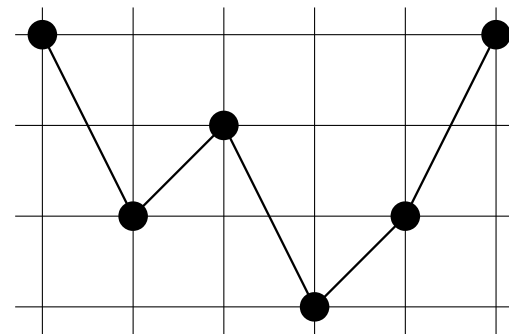
$$\iff f(x) + f(y) \geq f(x+1) + f(y-1) \quad (x < y)$$

$\iff f$  is **convex-extensible**, i.e.,

$\exists$  convex  $\bar{f} : \mathbb{R} \rightarrow \overline{\mathbb{R}}$  s.t.  $\bar{f}(x) = f(x) \quad (\forall x \in \mathbb{Z})$



convex



non-convex

# Local vs Global Optimality

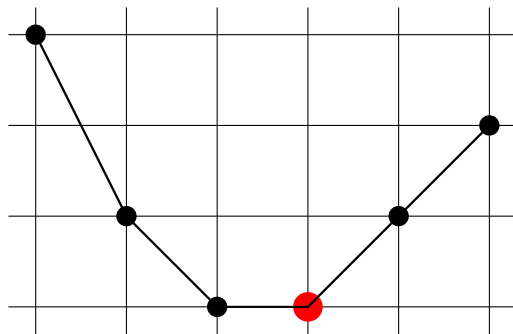
$$f : \mathbb{Z} \rightarrow \overline{\mathbb{R}}$$

**Theorem:**

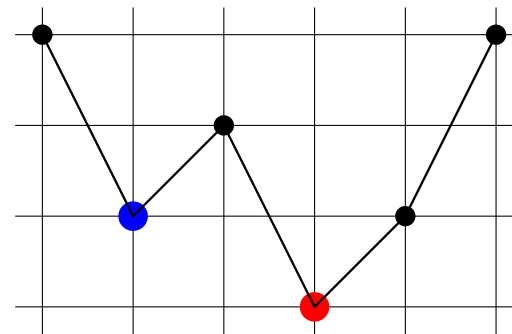
$x^*$ : global opt (min)

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

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



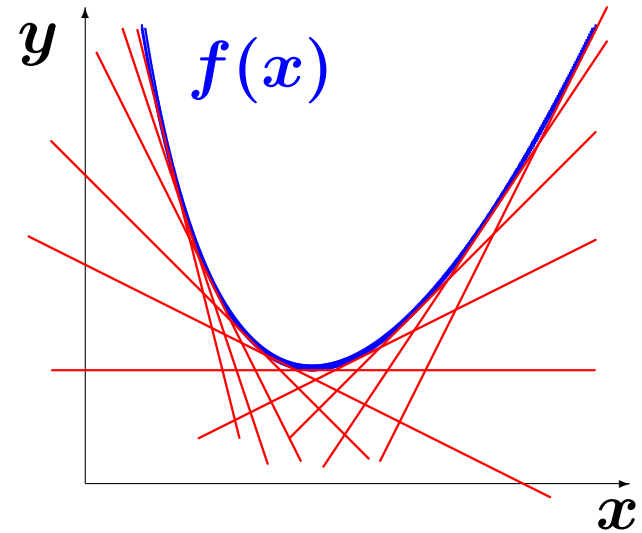
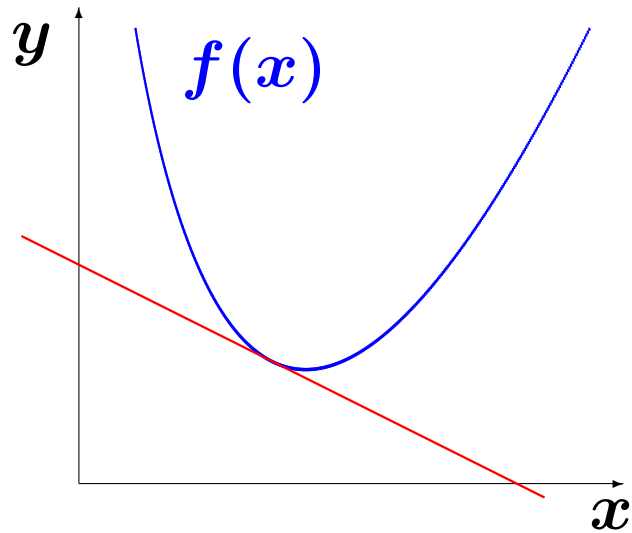
convex



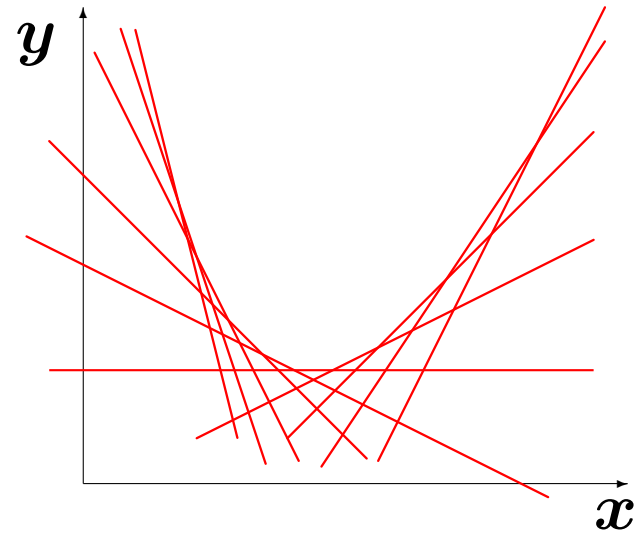
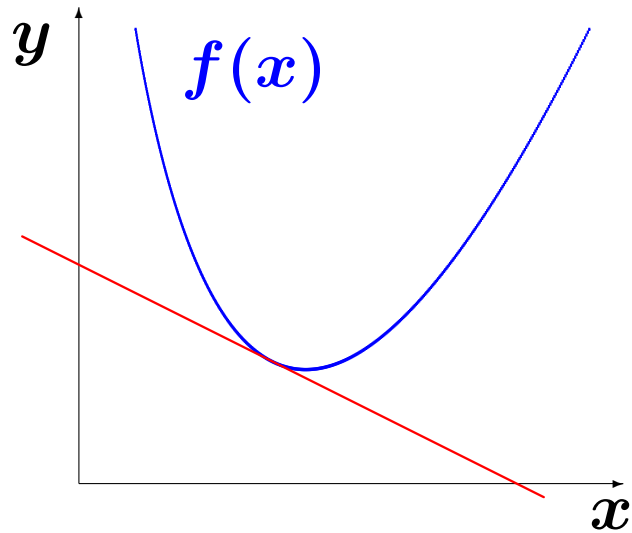
non-convex



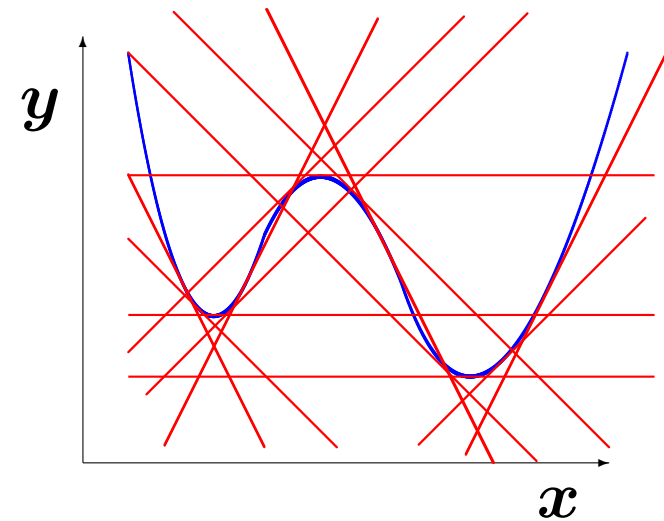
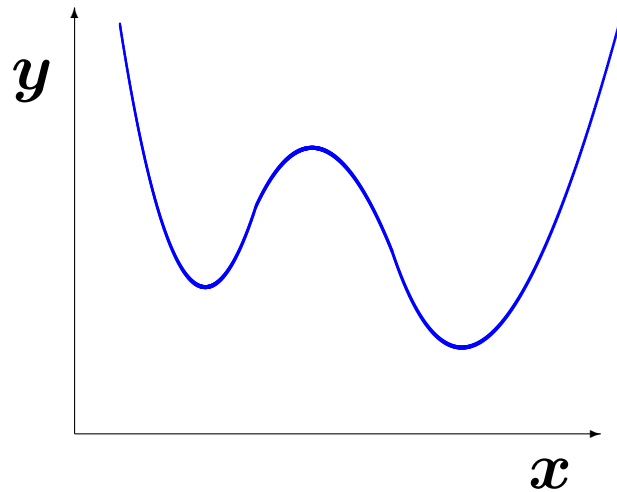
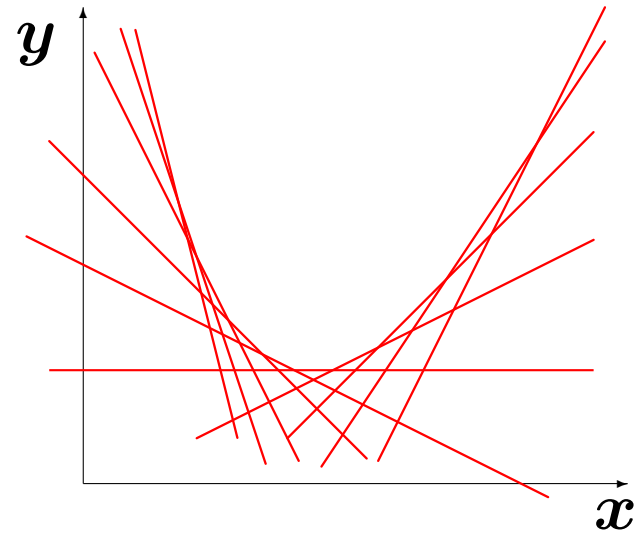
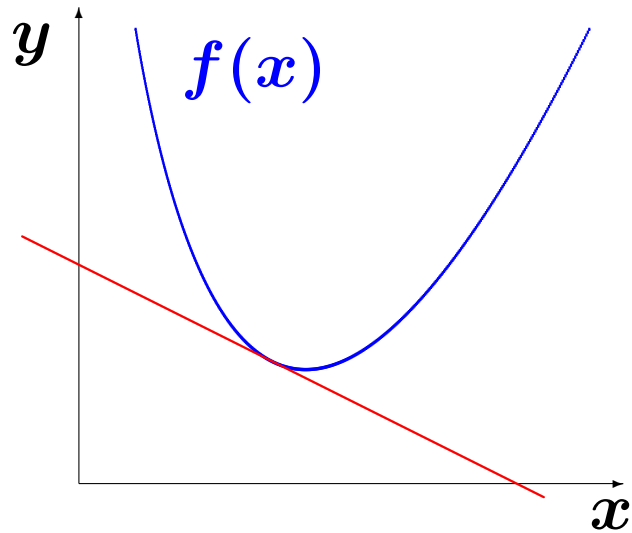
# Intuition of Legendre Transformation



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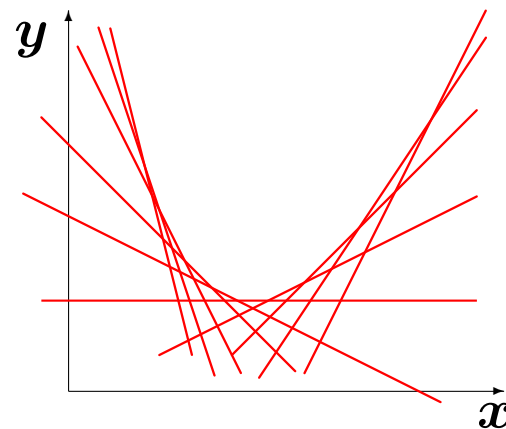
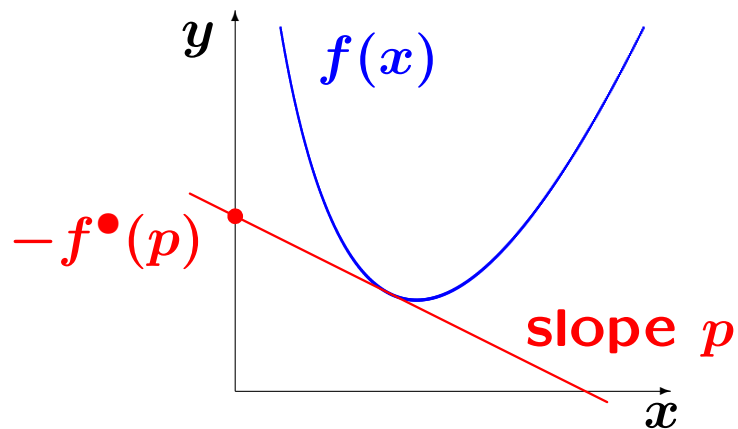


# Legendre Transformation

$f : \mathbb{Z} \rightarrow \overline{\mathbb{Z}}$  (integer-valued)

Define **discrete Legendre transform** of  $f$  by

$$f^\bullet(p) = \sup\{px - f(x) \mid x \in \mathbb{Z}\} \quad (p \in \mathbb{Z})$$



**Theorem:**

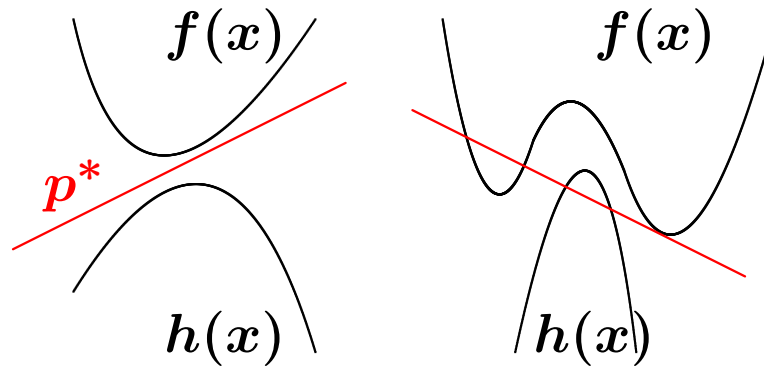
(1)  $f^\bullet$  is  $\mathbb{Z}$ -valued convex function,  $f^\bullet : \mathbb{Z} \rightarrow \overline{\mathbb{Z}}$

(2)  $(f^\bullet)^\bullet = f$  (biconjugacy)

# Separation Theorem

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

$h : \mathbb{Z} \rightarrow \underline{\mathbb{R}}$   
concave



## Theorem (Discrete Separation Theorem)

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

$$f(x) \geq \alpha^* + p^* x \geq h(x) \quad (\forall x \in \mathbb{Z})$$

(2)  $f, h$ : integer-valued  $\Rightarrow \alpha^* \in \mathbb{Z}, p^* \in \mathbb{Z}$

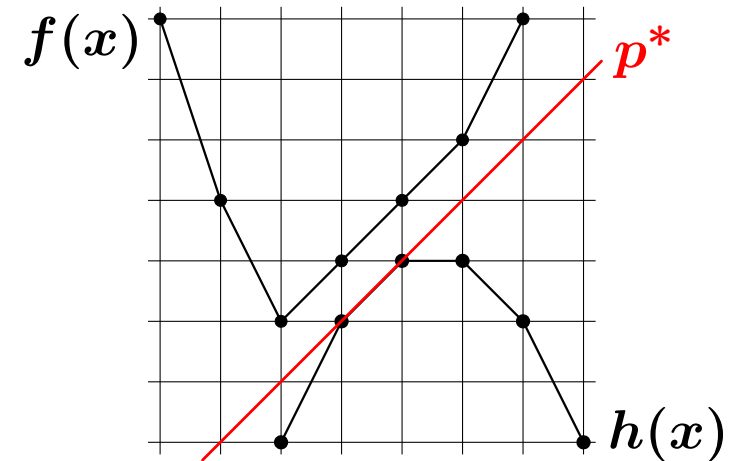
# Separation Theorem

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

**convex**

$$h : \mathbb{Z} \rightarrow \underline{\mathbb{R}}$$

**concave**



## **Theorem** (Discrete Separation Theorem)

$$(1) f(x) \geq h(x) \quad (\forall x \in \mathbb{Z}) \Rightarrow \exists \alpha^* \in \mathbb{R}, \exists p^* \in \mathbb{R}:$$

$$f(x) \geq \alpha^* + p^* x \geq h(x) \quad (\forall x \in \mathbb{Z})$$

$$(2) f, h: \text{integer-valued} \Rightarrow \alpha^* \in \mathbb{Z}, p^* \in \mathbb{Z}$$

# Fenchel Duality (Min-Max)

$f : \mathbb{Z} \rightarrow \overline{\mathbb{Z}}$ : convex,       $h : \mathbb{Z} \rightarrow \underline{\mathbb{Z}}$ : concave

Legendre transforms:

$$f^\bullet(p) = \sup\{px - f(x) \mid x \in \mathbb{Z}\}$$

$$h^\circ(p) = \inf\{px - h(x) \mid x \in \mathbb{Z}\}$$

**Theorem:**

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

# 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 **univariate discrete convex functions**



# C2.

## Classes of

## Discrete Convex Functions

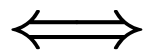
# Classes of Discrete Convex Functions

- 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

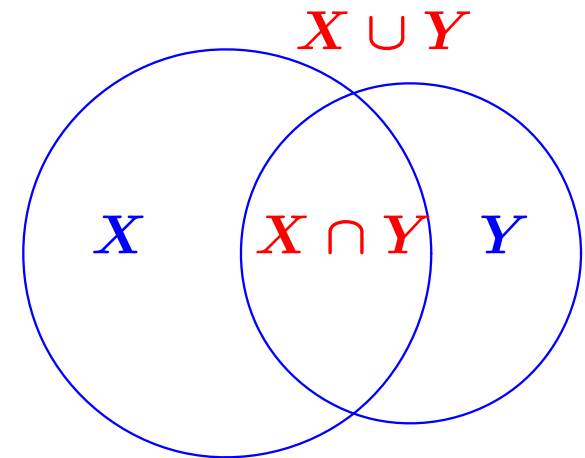
# Submodular Function

$$\overline{\mathbb{R}} = \mathbb{R} \cup \{+\infty\}$$

Set function  $\rho : 2^V \rightarrow \overline{\mathbb{R}}$  is **submodular**



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



cf.  $|X| + |Y| = |X \cup Y| + |X \cap Y|$

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

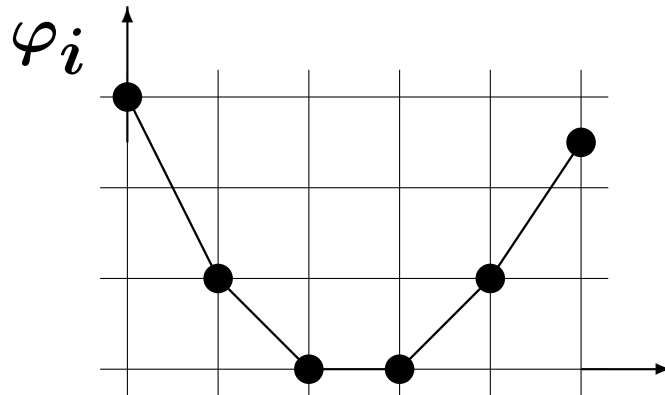
# Separable-convex Function

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

$\iff$

$$f(x) = \varphi_1(x_1) + \varphi_2(x_2) + \cdots + \varphi_n(x_n)$$

$\varphi_i$ : univariate convex



# 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 discrete convex functions**

# Some History

1935	Matroid	Whitney, Nakasawa
1965	Submodular function	Edmonds
1969	Convex network flow (electr.circuit)	Iri
<b>1982</b>	<b>Submodularity and convexity</b>	Frank, Fujishige, Lovász
1990	Valuated matroid	Dress–Wenzel
	Integrally convex fn	Favati–Tardella
<b>1996</b>	<b>Discrete convex analysis</b>	Murota
2000	Submodular minimization algorithm	Iwata–Fleischer–Fujishige, Schrijver
2006	M-convex fn on jump system	Murota
2012	L-convex fn on graph	Hirai, Kolmogorov

# Motivations/Applications/Connections

1. submodular	<b>MANY</b> problems graph cut, convex game
1. separable-conv	<b>MANY</b> problems min-cost flow, resource allocation
1. integrally-conv	[mathematical aesthetics]
2. L-conv ( $\mathbb{Z}^n$ )	network tension, image processing OR ( <b>inventory</b> , scheduling)
2. M-conv ( $\mathbb{Z}^n$ )	network flow, congestion game economics (game, <b>auction</b> ) <b>mixed polynomial matrix</b>
3. M-conv (jump)	deg sequence, <b>(2-)matching</b> polynomial (half-plane property)
3. L-conv (graph)	<b>multiflow</b> , multifacility location

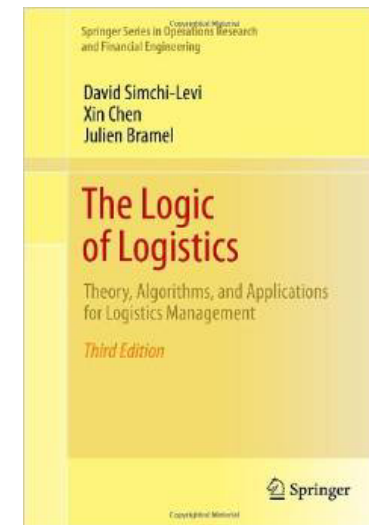
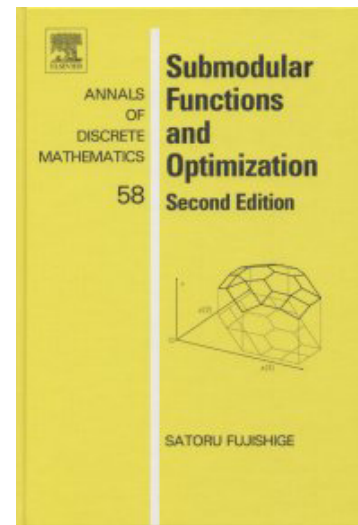
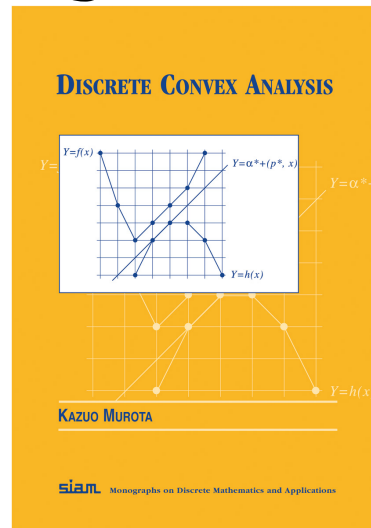
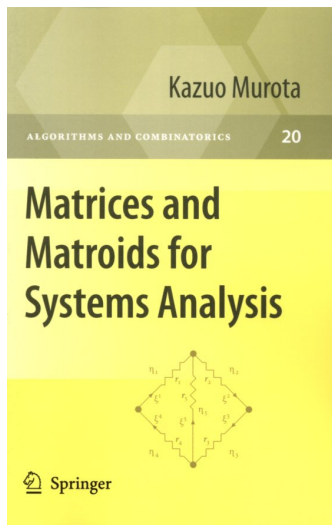
# Books (discrete convex analysis)

2000: Murota, **Matrices and Matroids for Systems Analysis**, Springer

2003: Murota, **Discrete Convex Analysis**, SIAM

2005: Fujishige, **Submodular Functions and Optimization**, 2nd ed., Elsevier

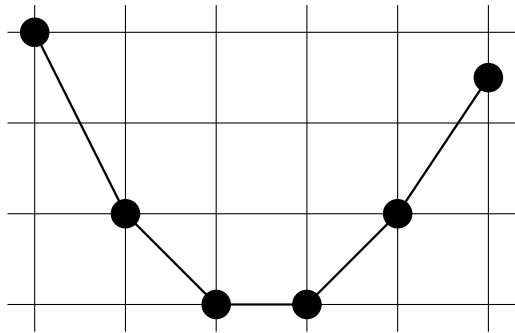
2014: Simchi-Levi, Chen, Bramel, **The Logic of Logistics**, 3rd ed., Springer



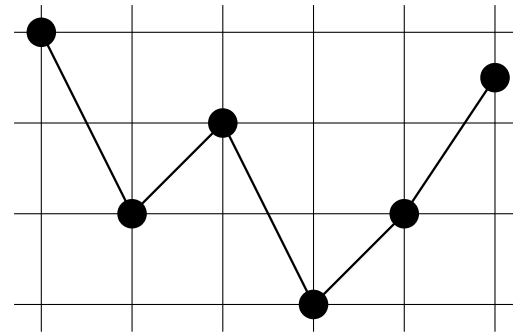


# Convex Extension

- $f : \mathbb{Z}^n \rightarrow \overline{\mathbb{R}}$  is **convex-extensible**  
 $\Leftrightarrow \exists$  convex  $\bar{f} : \mathbb{R}^n \rightarrow \overline{\mathbb{R}} : \bar{f}(x) = f(x) \ (\forall x \in \mathbb{Z}^n)$
- $\bar{f}$  is a **convex extension** of  $f$
- **convex closure**  
= (pointwise) max convex extension



convex-extensible

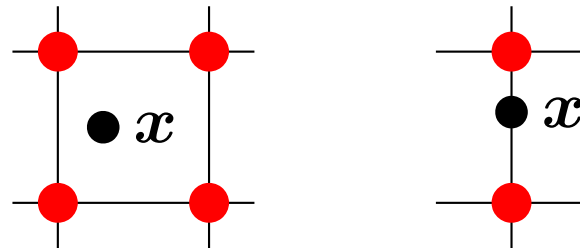


NOT convex-extensible

# Integrally Convex Function

(Favati-Tardella 1990)

$$N(x) = \{y \in \mathbb{Z}^n \mid \|x - y\|_\infty < 1\} \quad (x \in \mathbb{R}^n)$$

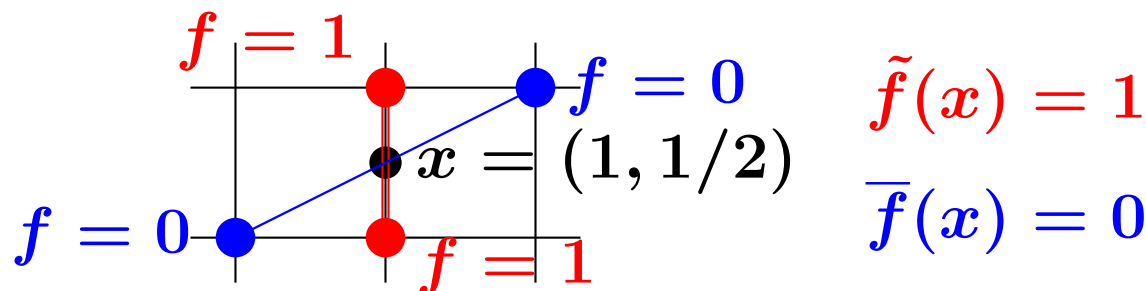


Local convex extension:

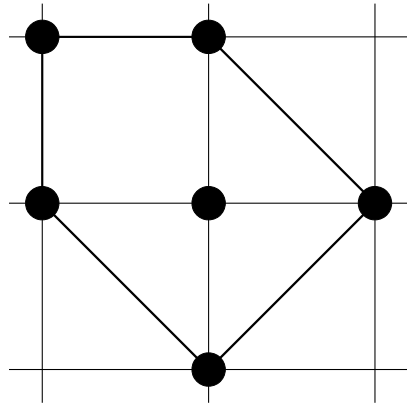
$$\tilde{f}(x) = \sup_{p, \alpha} \{ \langle p, x \rangle + \alpha \mid \langle p, y \rangle + \alpha \leq f(y) \ (\forall y \in N(x)) \}$$

Def:  $f$  is integrally convex  $\iff \tilde{f}$  is convex

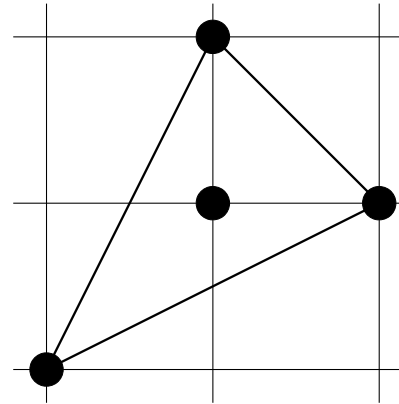
Ex:  $f(x_1, x_2) = |x_1 - 2x_2|$  is NOT integrally convex



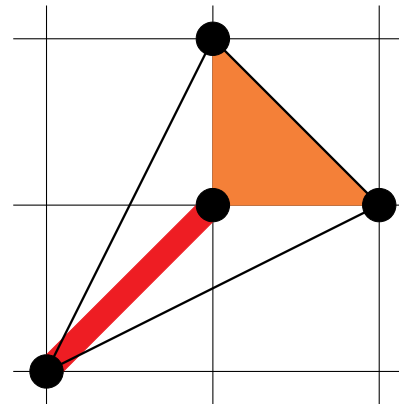
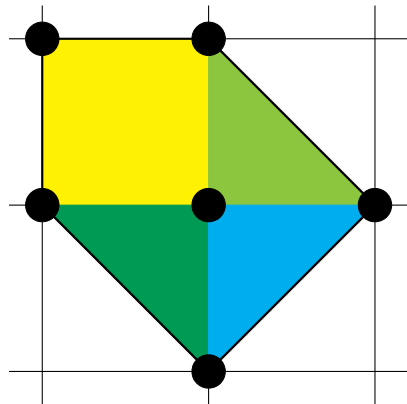
# Integrally Convex Set



**YES**



**NO**



# Five Properties of “Convex” Functions

1. convex extension
2. local opt = global opt

hold for **integrally convex functions**

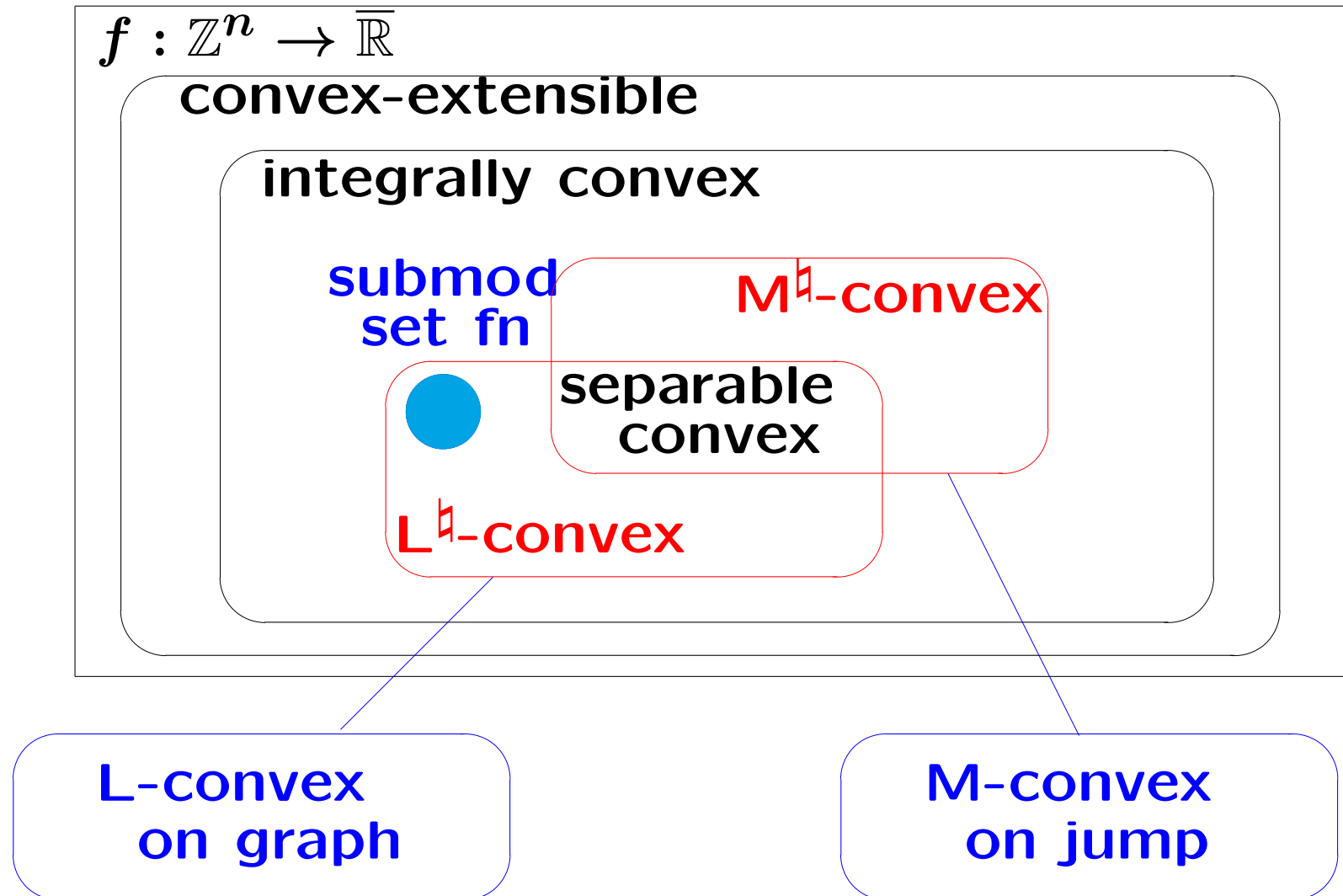
3. Legendre transform (biconjugacy)
4. separation theorem
5. Fenchel duality

fail for **integrally convex functions**

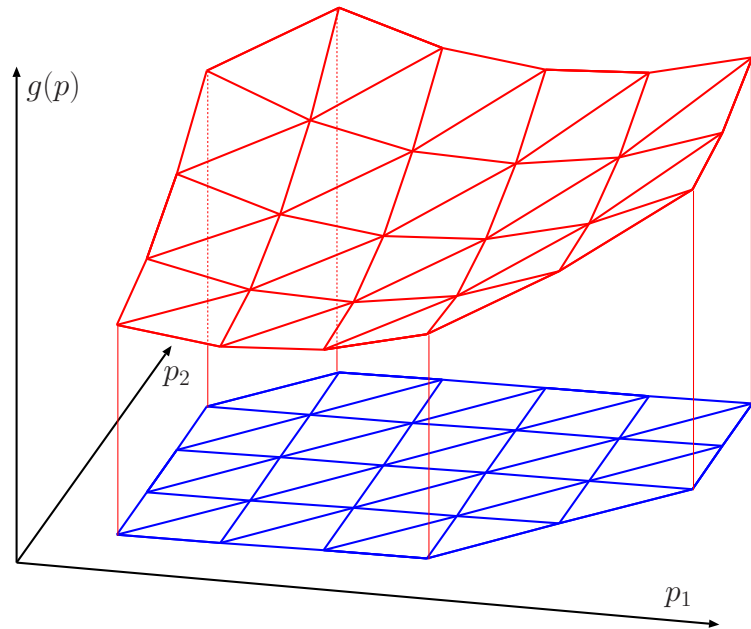
# Definitions

<p>1. submodular (set fn)</p>	$\rho(X) + \rho(Y) \geq \rho(X \cup Y) + \rho(X \cap Y)$
<p>1. separable -conv</p>	$f(x) = \varphi_1(x_1) + \varphi_2(x_2) + \cdots + \varphi_n(x_n)$ $\varphi_i(t-1) + \varphi_i(t+1) \geq 2\varphi_i(t) \quad (\forall t \in \mathbb{Z})$
<p>1. integrally -conv</p>	<p>Local convex ext <math>\tilde{f}(x)</math> is convex</p>
<p>2. L-conv(<math>\mathbb{Z}^n</math>)</p>	
<p>2. M-conv(<math>\mathbb{Z}^n</math>)</p>	
<p>3. M-conv(jump)</p>	
<p>3. L-conv(graph)</p>	

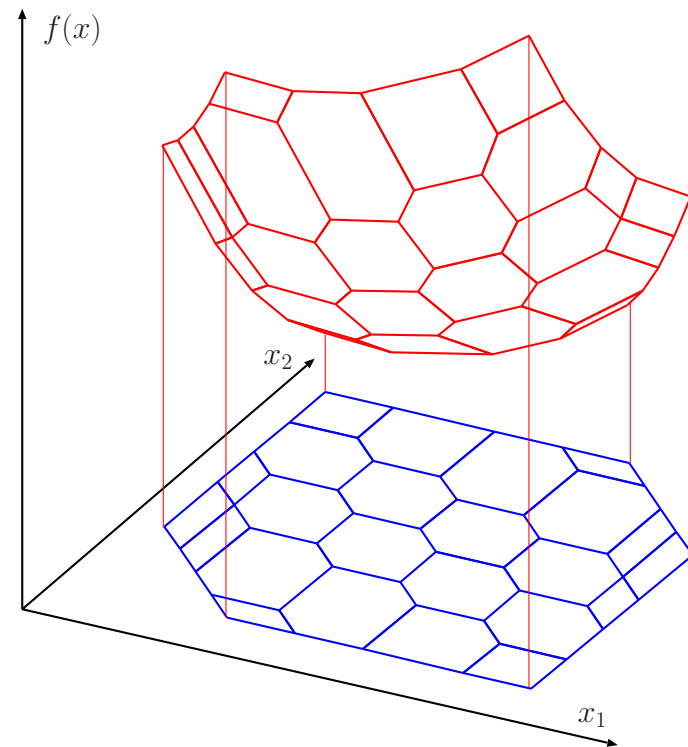
# Classes of Discrete Convex Functions



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



**$L^{\natural}$ -convex fn**



**$M^{\natural}$ -convex fn**

# C3.

## L-convex Functions



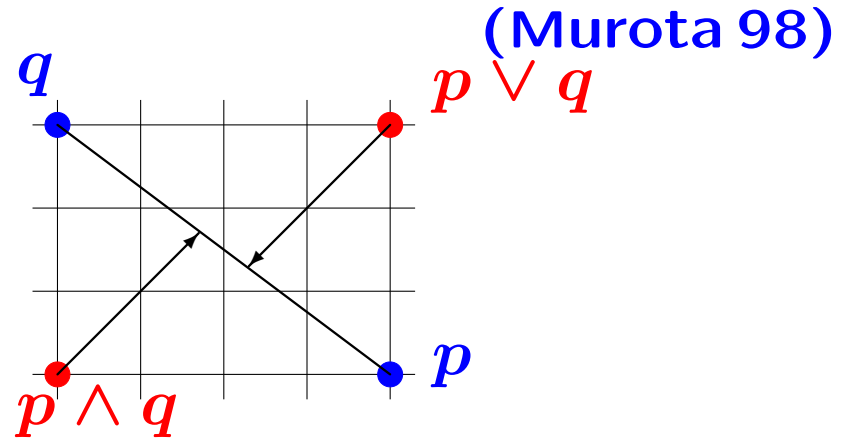
# L-convex Function

(L = Lattice)

$$g : \mathbb{Z}^n \rightarrow \mathbb{R} \cup \{+\infty\}$$

$p \vee q$     **compnt-max**

$p \wedge q$     **compnt-min**

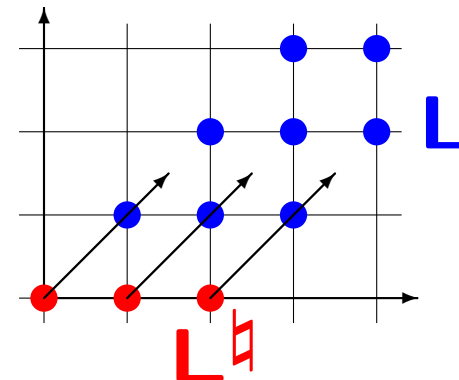


**Def:**  $g$  is L-convex  $\iff$

• Submodular:  $g(p) + g(q) \geq g(p \vee q) + g(p \wedge q)$

• Translation:  $\exists r, \forall p: g(p + 1) = g(p) + r$

$$1 = (1, 1, \dots, 1)$$



# $L^{\natural}$ -convexity from Submodularity

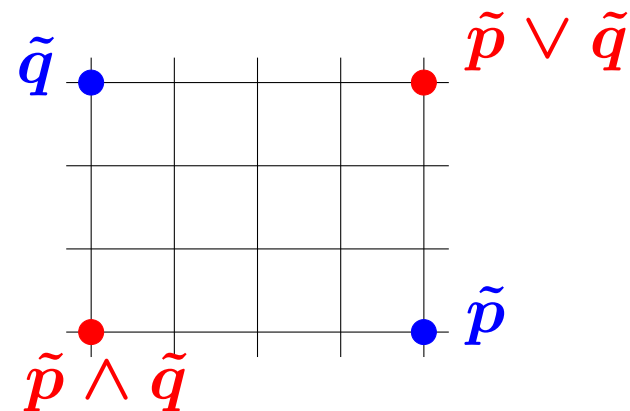
(Murota 98, Fujishige–Murota 2000)

$$g : \mathbb{Z}^n \rightarrow \mathbb{R} \quad \mathbf{L^{\natural}\text{-convex}} \iff$$

$$\tilde{g}(p_0, p) = g(p - p_0 \mathbf{1}) \text{ is submodular in } (p_0, p)$$

$$\tilde{g} : \mathbb{Z}^{n+1} \rightarrow \mathbb{R}, \quad \mathbf{1} = (1, 1, \dots, 1)$$

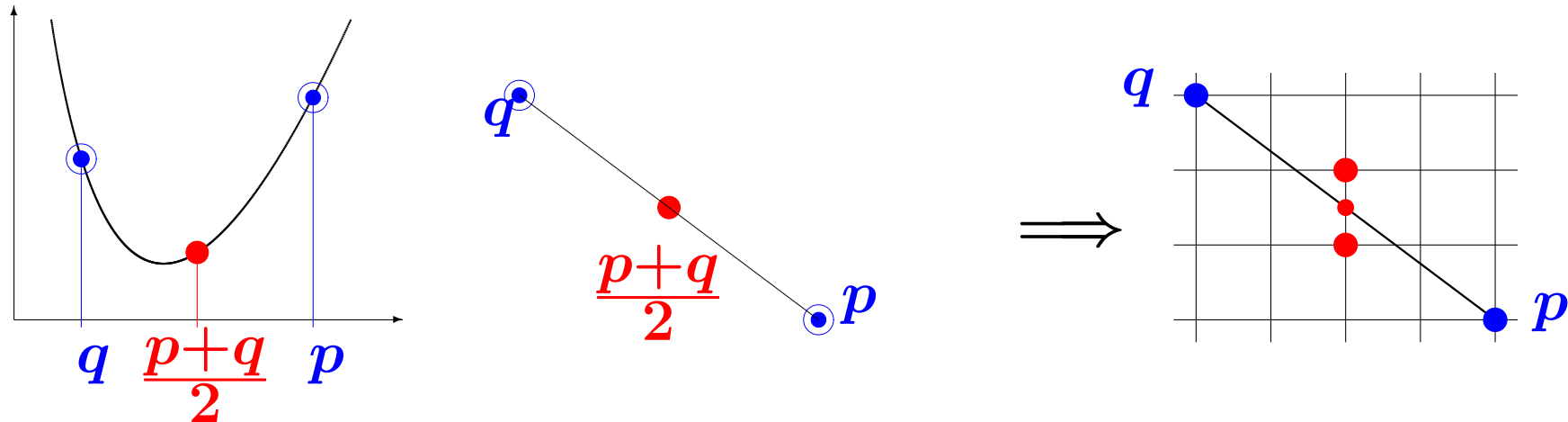
$$\tilde{g}(\tilde{p}) + \tilde{g}(\tilde{q}) \geq \tilde{g}(\tilde{p} \vee \tilde{q}) + \tilde{g}(\tilde{p} \wedge \tilde{q})$$



$$\mathbf{L}_{n+1} \simeq \mathbf{L}_n^{\natural} \supsetneq \mathbf{L}_n$$

# $L^{\natural}$ -convexity from Mid-pt-convexity

(Favati-Tardella 1990, Fujishige–Murota 2000)



Mid-point convex ( $g : \mathbb{R}^n \rightarrow \mathbb{R}$ ):

$$g(p) + g(q) \geq 2g\left(\frac{p+q}{2}\right)$$

$\Rightarrow$  **Discrete mid-point convex ( $g : \mathbb{Z}^n \rightarrow \mathbb{R}$ )**

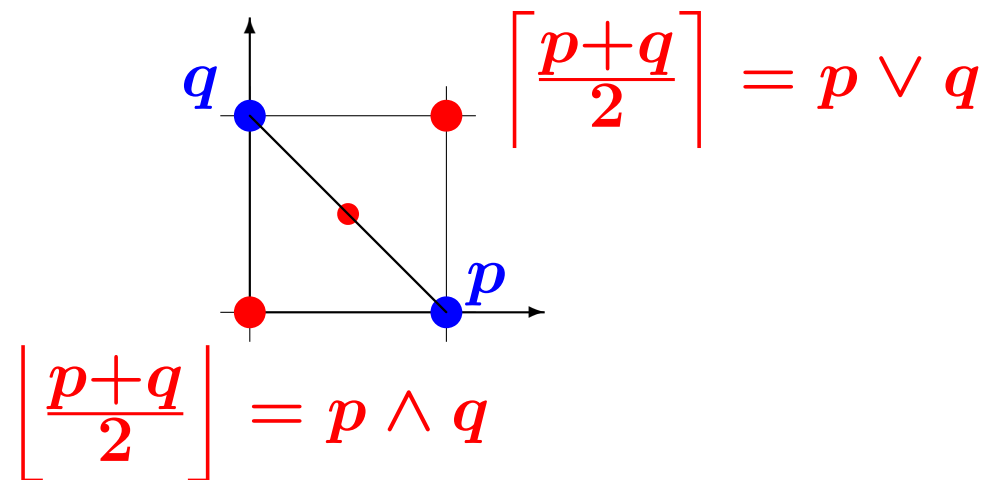
$$g(p) + g(q) \geq g\left(\left\lceil \frac{p+q}{2} \right\rceil\right) + g\left(\left\lfloor \frac{p+q}{2} \right\rfloor\right)$$

**$L^{\natural}$ -convex function**

( $L = \text{Lattice}$ )

# Mid-pt Convexity for 01-Vectors

For  $p, q \in \{0, 1\}^n$



**Discrete mid-pt convexity:**

$$g(p) + g(q) \geq g\left(\left\lceil \frac{p+q}{2} \right\rceil\right) + g\left(\left\lfloor \frac{p+q}{2} \right\rfloor\right)$$

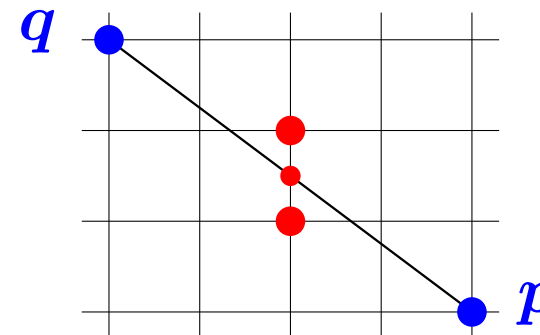
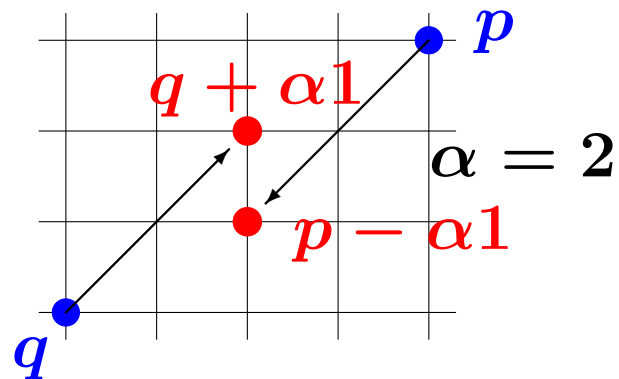
$\iff$  **Submodularity:**

$$g(p) + g(q) \geq g(p \vee q) + g(p \wedge q)$$

# Translation Submodularity ( $L^{\natural}$ )

$$g(p) + g(q) \geq g((p - \alpha 1) \vee q) + g(p \wedge (q + \alpha 1))$$

$$(\alpha \geq 0)$$

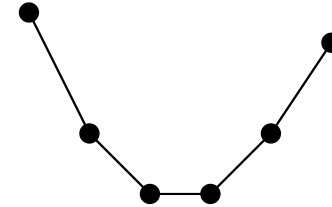
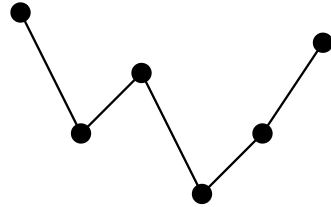


discrete mid-pt convex

- $\tilde{g}(p_0, p) = g(p - p_0 1)$  is submodular in  $(p_0, p)$   
 (Fujishige-Murota 00)
- $\Leftrightarrow$  translation submodular  
 (Fujishige-Murota 00)
- $\Leftrightarrow$  discrete mid-pt convex  
 (Favati-Tardella 90)
- $\Leftrightarrow$  submod. integ. convex

# Rem: $L^{\natural}$ -convex vs Submodular

$n = 1$



**Fact 1:** Every  $g : \mathbb{Z} \rightarrow \mathbb{R}$  is **submodular**

**Fact 2:** Function  $g : \mathbb{Z} \rightarrow \mathbb{R}$  is  **$L^{\natural}$ -convex**

$$\iff g(p-1) + g(p+1) \geq 2g(p) \text{ for all } p \in \mathbb{Z}$$

# L<sup>‡</sup>-convex Function: Examples

**Quadratic:**  $g(p) = \sum_i \sum_j a_{ij} p_i p_j$  is L<sup>‡</sup>-convex

$$\Leftrightarrow a_{ij} \leq 0 \quad (i \neq j), \quad \sum_j a_{ij} \geq 0 \quad (\forall i)$$

**Energy function:** For univariate convex  $\psi_i$  and  $\psi_{ij}$

$$g(p) = \sum_i \psi_i(p_i) + \sum_{i \neq j} \psi_{ij}(p_i - p_j)$$

**Range:**  $g(p) = \max\{p_1, p_2, \dots, p_n\} - \min\{p_1, p_2, \dots, p_n\}$

**Submodular set function:**  $\rho : 2^V \rightarrow \overline{\mathbb{R}}$

$$\Leftrightarrow \rho(X) = g(\chi_X) \quad \text{for some L}^{\ddagger}\text{-convex } g$$

**Multimodular:**  $h : \mathbb{Z}^n \rightarrow \overline{\mathbb{R}}$  is multimodular  $\Leftrightarrow$

$h(p) = g(p_1, p_1 + p_2, \dots, p_1 + \dots + p_n)$  for L<sup>‡</sup>-convex  $g$

# Five Properties of “Convex” Functions

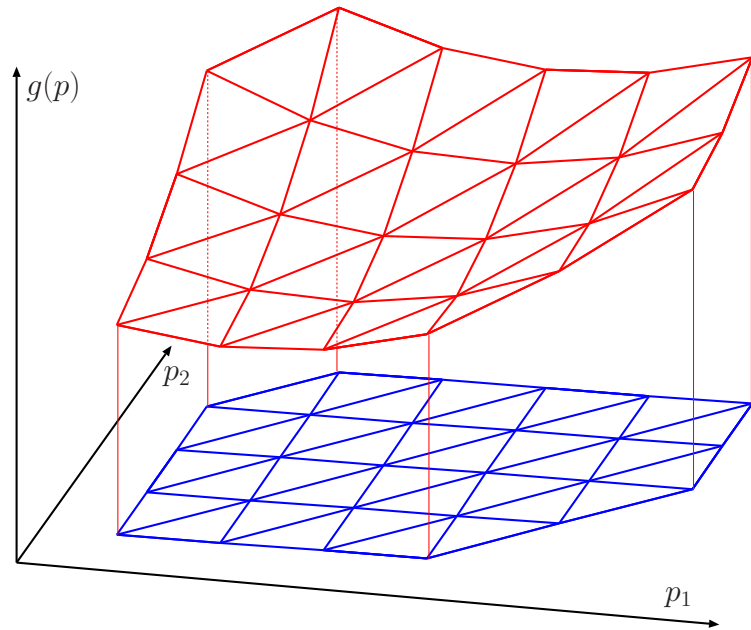
1. convex extension
2. local opt = global opt
3. Legendre transform (biconjugacy)
4. separation theorem
5. Fenchel duality

hold for **L-convex functions**

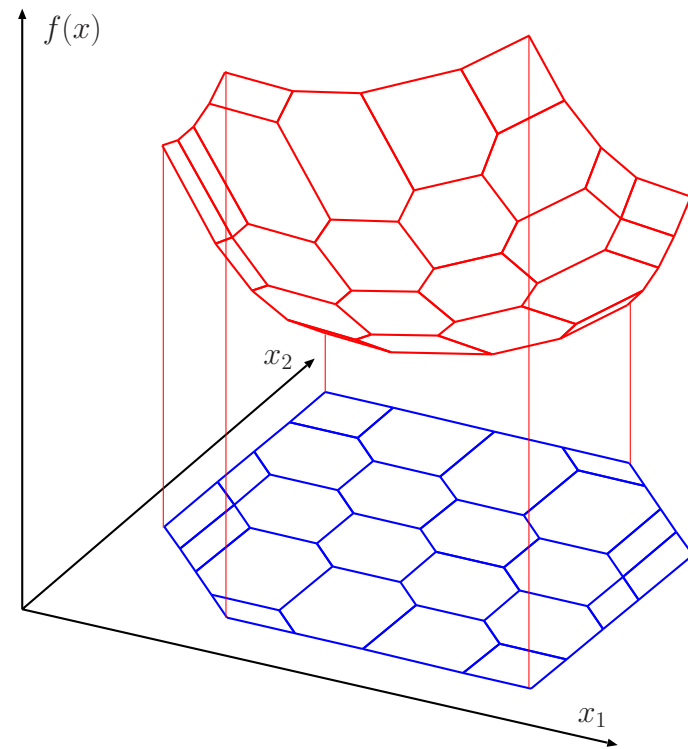
⇒ **Part II**



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



**$L^{\natural}$ -convex fn**



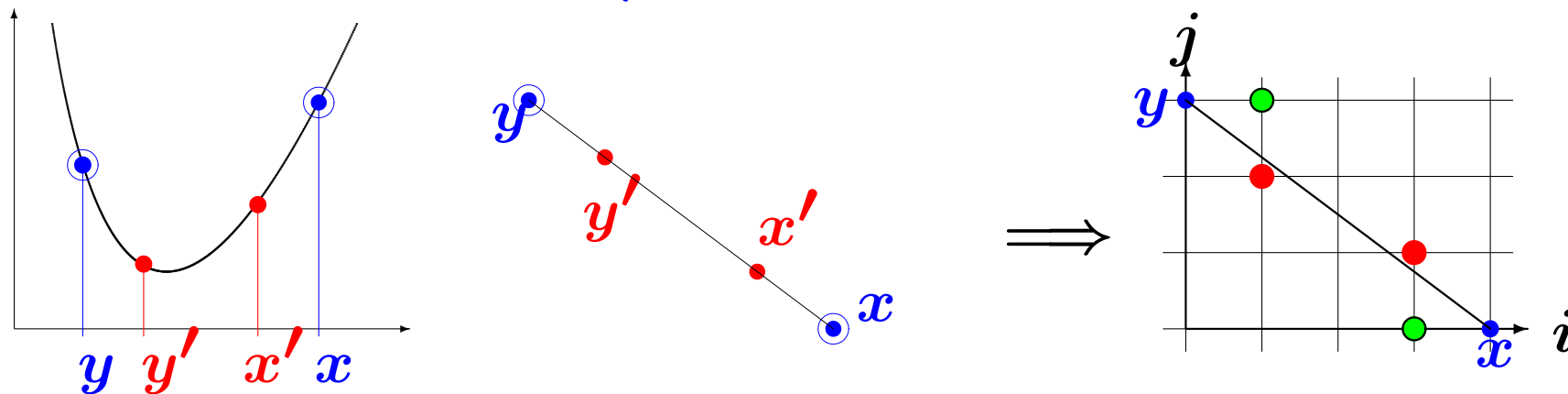
**$M^{\natural}$ -convex fn**

# C4.

## M-convex Functions

# M<sup>‡</sup>-convexity from Equi-dist-convexity

(Murota 1996, Murota–Shioura 1999)



Equi-distance convex ( $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ):

$$f(x) + f(y) \geq f(x - \alpha(x - y)) + f(y + \alpha(x - y))$$

$\implies$  Exchange ( $f : \mathbb{Z}^n \rightarrow \mathbb{R}$ )  $\forall x, y, \forall i : x_i > y_i$

$$f(x) + f(y) \geq \min [f(x - e_i) + f(y + e_i),$$

$$\min_{x_j < y_j} \{f(x - e_i + e_j) + f(y + e_i - e_j)\}]$$

M<sup>‡</sup>-convex function

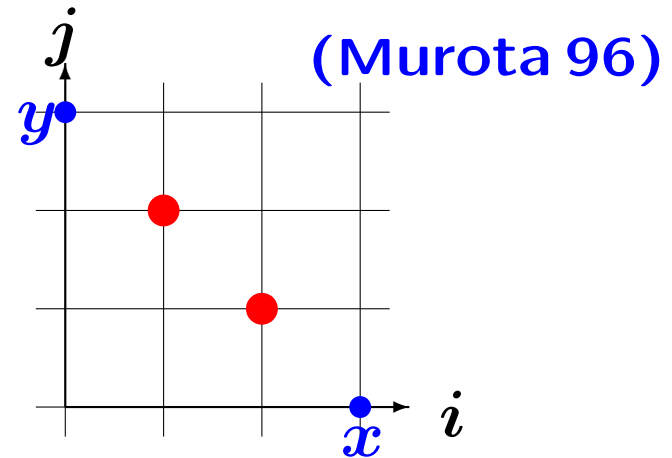
(M = Matroid)

# M-convex Function

(M = Matroid)

$$f : \mathbb{Z}^n \rightarrow \mathbb{R} \cup \{+\infty\}$$

$e_i$ :  $i$ -th unit vector



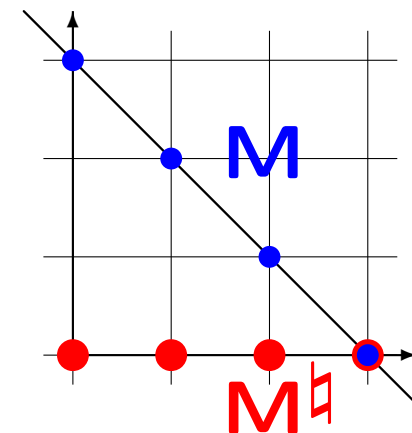
**Def:**  $f$  is M-convex

$$\iff \forall x, y, \quad \forall i : x_i > y_i, \quad \exists j : x_j < y_j :$$

$$f(x) + f(y) \geq f(x - e_i + e_j) + f(y + e_i - e_j)$$

$\text{dom } f \subseteq \text{const-sum hyperplane}$

$$\mathbf{M}_{n+1} \simeq \mathbf{M}_n^{\natural} \supsetneq \mathbf{M}_n$$



# Gross Substitutes (for set function)

$f : 2^V \rightarrow \mathbb{R}$  utility (reservation value) function

$p$  price vector

$D(p) = \arg \max(f - p) = \{X \mid f(X) - p(X) \text{ is maximum}\}$   
demand correspondence

**Gross substitutes property:** (Kelso–Crawford 82)

$X \in D(p), p \leq q$

$\Rightarrow \exists Y \in D(q) : \{i \in X \mid p_i = q_i\} \subseteq Y$

Equiv. cond. for  $D(p)$  (Gul–Stacchetti 99)

Equiv. cond. for  $f$  (Reijnierse–van Gallekom–Potters 02)

& equivalence to  $M^{\natural}$ -concavity (Fujishige–Yang 03)

# Gross Substitutes for $f$ (not for $D(p)$ )

$f : 2^V \rightarrow \mathbb{R}$  (set function)

$f$ : **gross substitutes**  $\iff$

(i)  $f(S \cup \{i, j\}) + f(S) \leq f(S \cup \{i\}) + f(S \cup \{j\})$

(submodular)

(ii)  $f(S \cup \{i, j\}) + f(S \cup \{k\}) \leq$

$\max[f(S \cup \{i, k\}) + f(S \cup \{j\}), f(S \cup \{j, k\}) + f(S \cup \{i\})]$

(Reijnierse–van Gallekom–Potters 02)

cf. Local exchange axiom of  $M^{\natural}$ -concave functions

# M<sup>h</sup>-concavity = Gross Substitutes

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

**M<sup>h</sup>-concave  $\iff$  Gross substitutes (+ \* \*)**

(Reijnierse–van Gallekom–Potters 02, Fujishige–Yang 03  
Danilov–Koshevoy–Lang 03, Murota–Tamura 03)

$\Rightarrow$  **Shioura–Tamura’s survey** (J. OR. Soc. Japan, 2015)  
[https://www.jstage.jst.go.jp/article/jorsj/58/1/58\\_61/\\_pdf](https://www.jstage.jst.go.jp/article/jorsj/58/1/58_61/_pdf)

**Gross substitutes:** ( $f$ : utility,  $p$ : price)

$$x \in \arg \max(f - p), \quad p \leq q,$$

$$\Rightarrow \exists y \in \arg \max(f - q) : y_i \geq x_i \quad \text{if } p_i = q_i$$

# M<sup>♯</sup>-convex Function: Examples

**Quadratic:**  $f(x) = \sum a_{ij}x_ix_j$  is M<sup>♯</sup>-convex

$$\Leftrightarrow a_{ij} \geq 0, \quad a_{ij} \geq \min(a_{ik}, a_{jk}) \quad (\forall k \notin \{i, j\})$$

**Min value:**  $f(X) = \min\{a_i \mid i \in X\}$  [unit preference]

**Cardinality convex:**  $f(X) = \varphi(|X|)$  ( $\varphi$ : convex)

**Separable convex:**  $f(x) = \sum \varphi_i(x_i)$  ( $\varphi_i$ : convex)

**Laminar convex:**  $f(x) = \sum_A \varphi_A(x(A))$  ( $\varphi_A$ : convex)

$\{A, B, \dots\}$ : laminar  $\Leftrightarrow A \cap B = \emptyset$  or  $A \subseteq B$  or  $A \supseteq B$



# M<sup>♯</sup>-concave Functions from Matroids

**Matroid rank:**  $f(X) = r(X)$  (rank of  $X$ )

**Matroid rank sum:**  $f(X) = \sum \alpha_i r_i(X)$

$r_i \leftarrow r_{i+1}$  (strong quotient),  $\alpha_i \geq 0$  (Shioura 12)

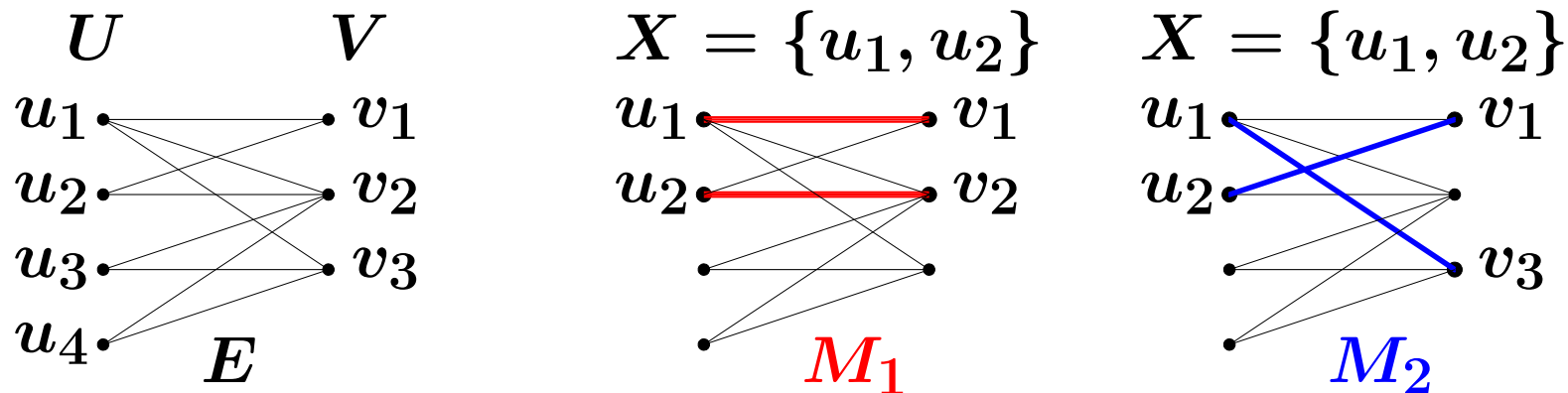
**Weighted matroid:**  $w$ : weight vector

$f(X) = \max\{w(Y) \mid Y: \text{indep} \subseteq X\}$  (Shioura 12)

**Valuated matroid:**  $\omega : 2^V \rightarrow \underline{\mathbb{R}}$

$\Leftrightarrow \omega(X) = f(\chi_X)$  for some M-concave  $f$

# 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)

# Polynomial Matrix

(Dress-Wenzel 90)  
Valuated Matroid

$$A = \begin{array}{|c|c|c|c|} \hline s+1 & s & 1 & 0 \\ \hline 1 & 1 & 1 & 1 \\ \hline \end{array} \quad \omega(J) = \deg \det A[J]$$

$\mathcal{B} = \{J \mid J \text{ is a base of column vectors}\}$

**Grassmann-Plücker  $\Rightarrow$  Exchange (M-concave)**

For any  $J, J' \in \mathcal{B}$ ,  $i \in J \setminus J'$ , there exists  $j \in J' \setminus J$   
s.t.  $J - i + j \in \mathcal{B}$ ,  $J' + i - j \in \mathcal{B}$ ,

$$\omega(J) + \omega(J') \leq \omega(J - i + j) + \omega(J' + i - j)$$

Ex.  $J = \{1, 2\}$ ,  $J' = \{3, 4\}$ ,  $i = 1$

$$\det A[\{1, 2\}] = \det A[\{3, 4\}] = 1, \quad \omega(J) = \omega(J') = 0$$

Can take  $j = 3$ :  $J - i + j = \{3, 2\}$ ,  $J' + i - j = \{1, 4\}$

$$\omega(J - i + j) = 1, \quad \omega(J' + i - j) = 1$$

# 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 **M-convex functions**

⇒ **Part II**

**C5.**

**Remarks on**

**Submodular Set Functions**

# Submodularity & Convexity in 1980's

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

- **min/max algorithms**

(Grötschel–Lovász–Schrijver/ Jensen–Korte, Lovász)

**min  $\Rightarrow$  polynomial,    max  $\Rightarrow$  NP-hard**

- **Convex extension**

(Lovász)

**set fn is submod  $\Leftrightarrow$  Lovász ext is convex**

- **Duality theorems**

(Edmonds, Frank, Fujishige)

**discrete separation,    Fenchel min-max**

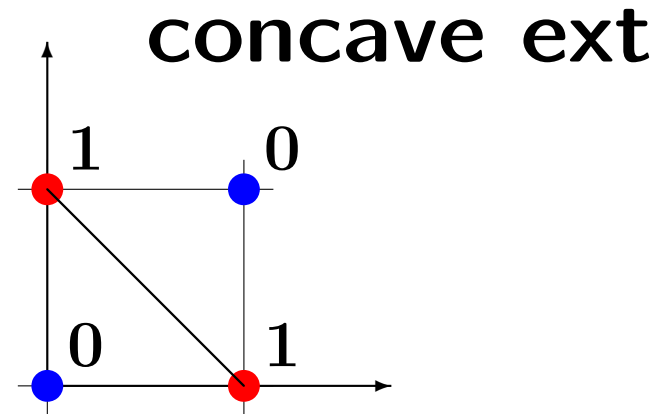
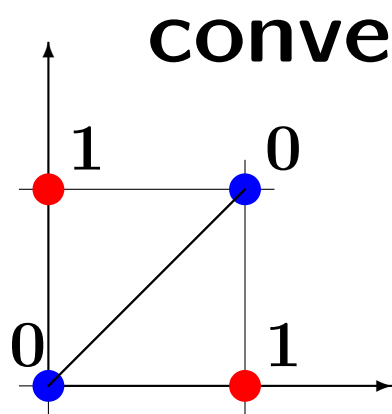
**Submodular set functions  
= Convexity + Discreteness**

# Set Function and Extensions

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

$$\rho(X) = \hat{\rho}(\chi_X)$$

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



cf. Lovász extension

# 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 **submodular set functions** (1980's)



# Submodular Set Function in DCA

- **Submodular** set fn = **L<sup>♯</sup>-convex** on  $\{0, 1\}^n$
- **M<sup>♯</sup>-concave** fns form a nice subclass
- **M<sup>♯</sup> + M<sup>♯</sup>** are polynomially tractable
- Sums of M<sup>♯</sup>-concave fns are still OK

$$f : \{0, 1\}^n \rightarrow \mathbb{R}$$

**submod=L<sup>♯</sup>-convex**

**M<sup>♯</sup> + M<sup>♯</sup>**

**M<sup>♯</sup>-  
concave**

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

**submodular**

**M<sup>♯</sup>-  
concave**

**L<sup>♯</sup>-  
convex**

# Dual Character of Matroid Rank Fn

Given a matroid:

$\rho(X) = \max\{|I| \mid I : \text{independent}, I \subseteq X\}$   
is **L<sup>♯</sup>-convex** and **M<sup>♯</sup>-concave**

Polymatroid rank function is **NOT** M<sup>♯</sup>-concave

**E N D**