

# Mixed Integer Programming Models for Non-Separable Piecewise Linear Cost Functions

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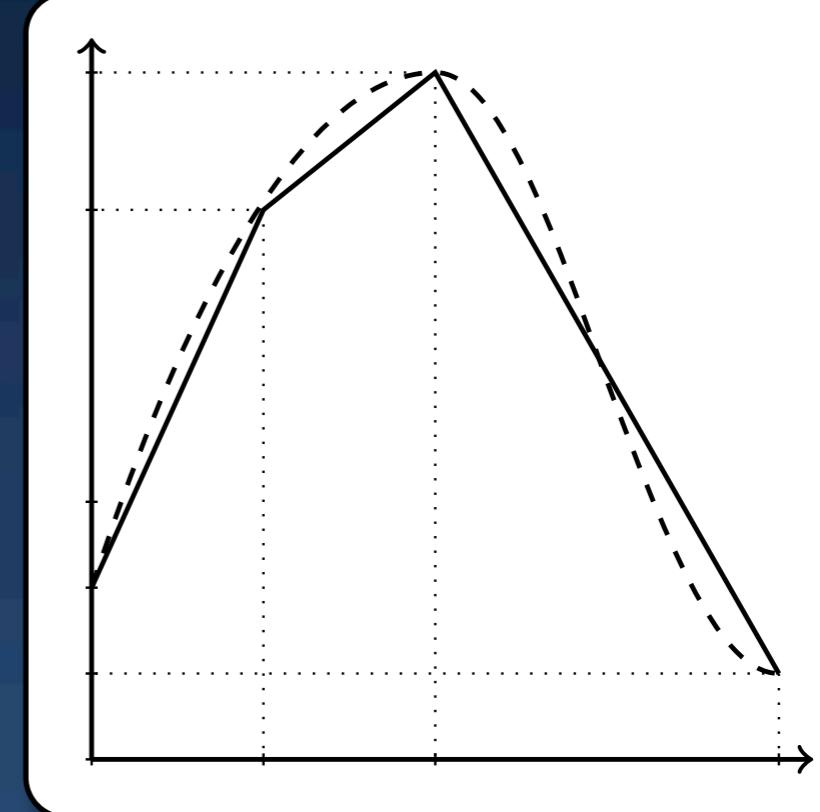
# Piecewise Linear Optimization

$$\min \quad f_0(x)$$

s.t.

$$f_i(x) \leq 0 \quad \forall i \in I$$

$$x \in X \subset \mathbb{R}^n$$



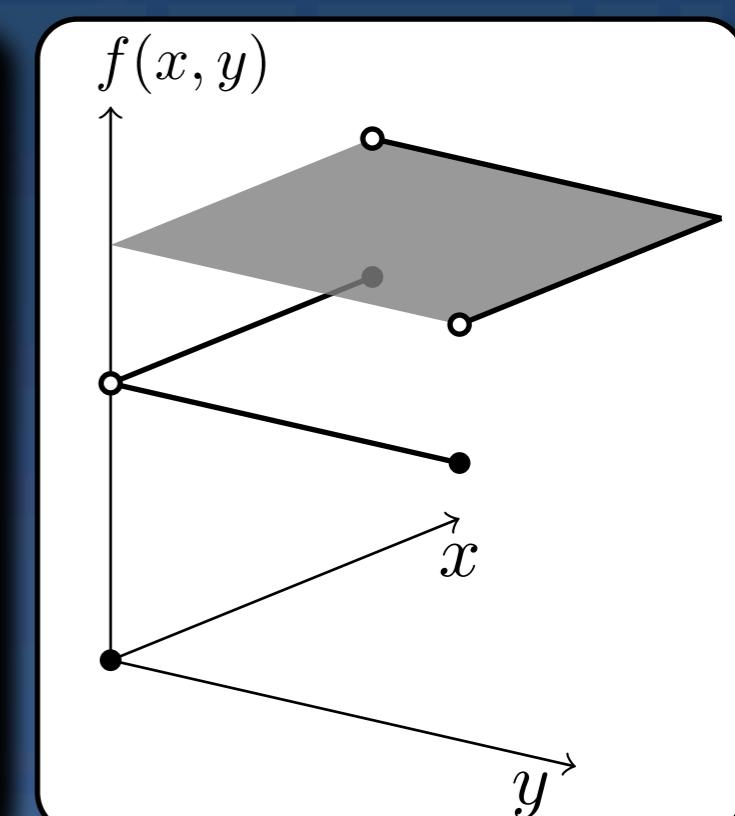
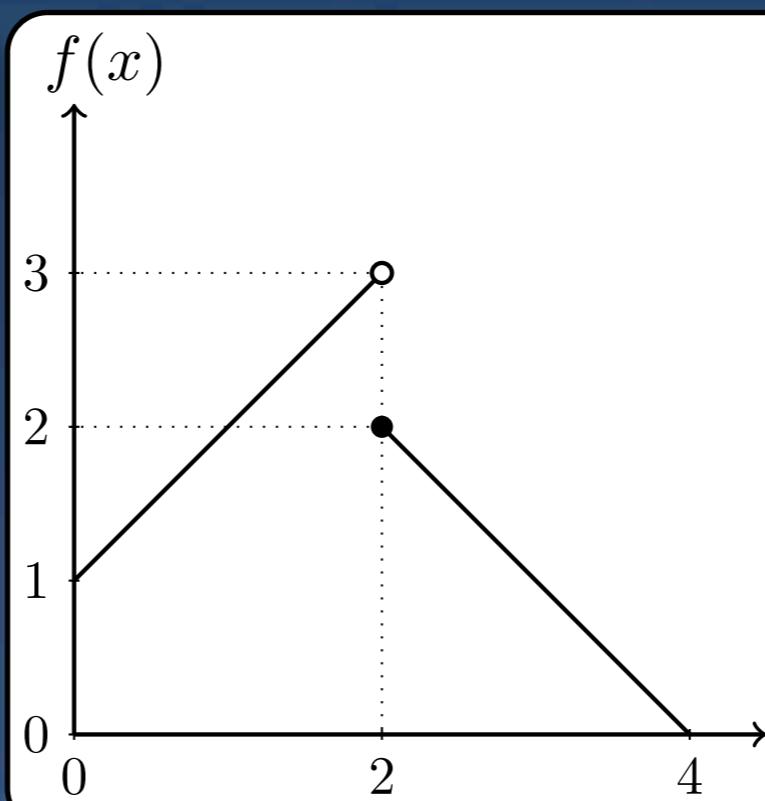
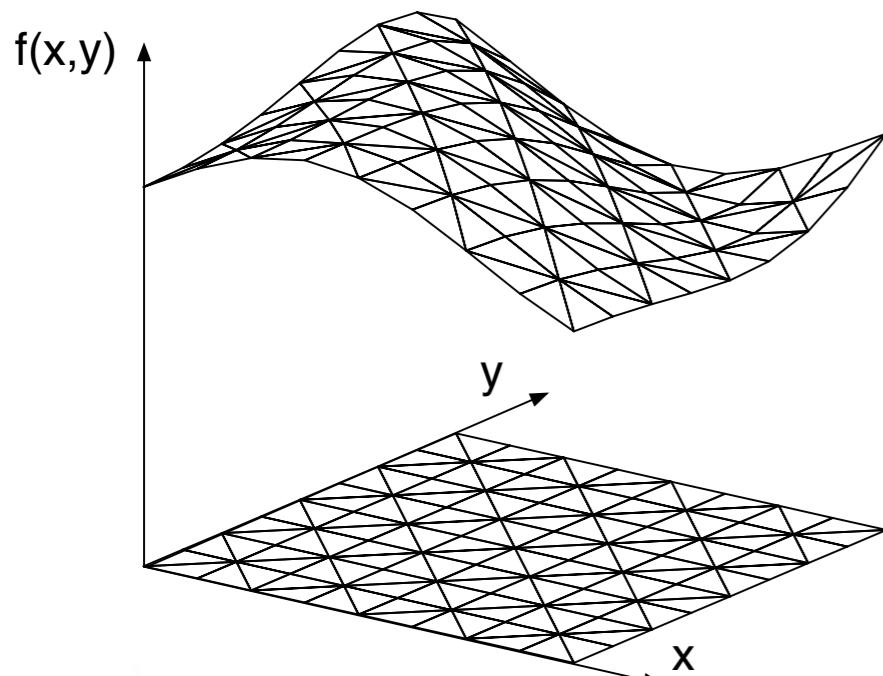
- $\forall i \in \{0\} \cup I \quad f_i(x) : D \rightarrow \mathbb{R}$  is a piecewise linear function (PLF) and  $X$  is any compact set.
- Convex = Linear Programming. Non-Convex = NP Hard.
- Specialized algorithms (Tomlin 1981, ..., de Farias et al. 2008 ) or Mixed Integer Programming Models (12+ papers).

# Mixed Integer Models for PLFs

- Existing studies are for separable functions:

$$f(x) = \sum_{j=1}^n f_j(x_j) \text{ for } f_j(x_j) : \mathbb{R} \rightarrow \mathbb{R}$$

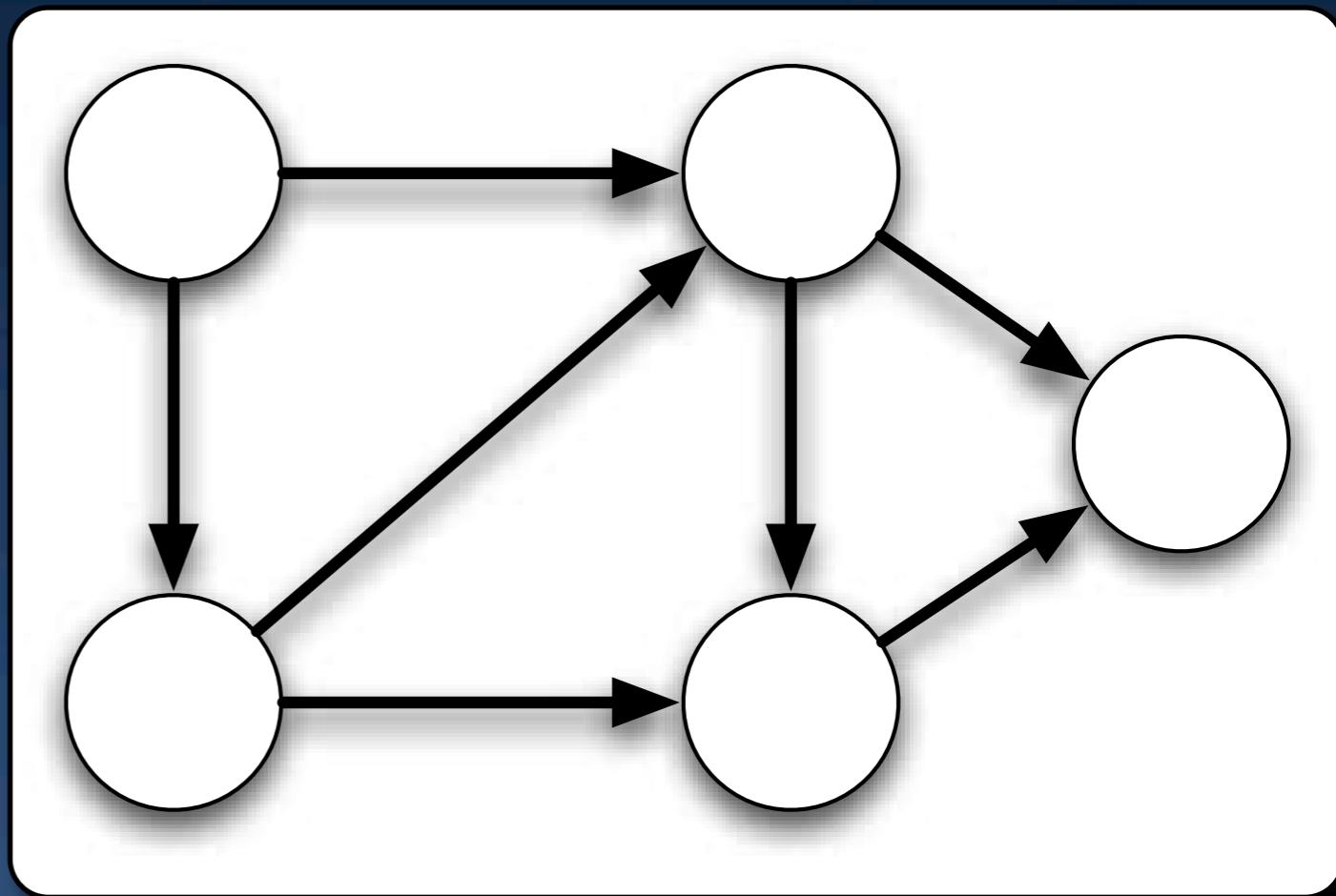
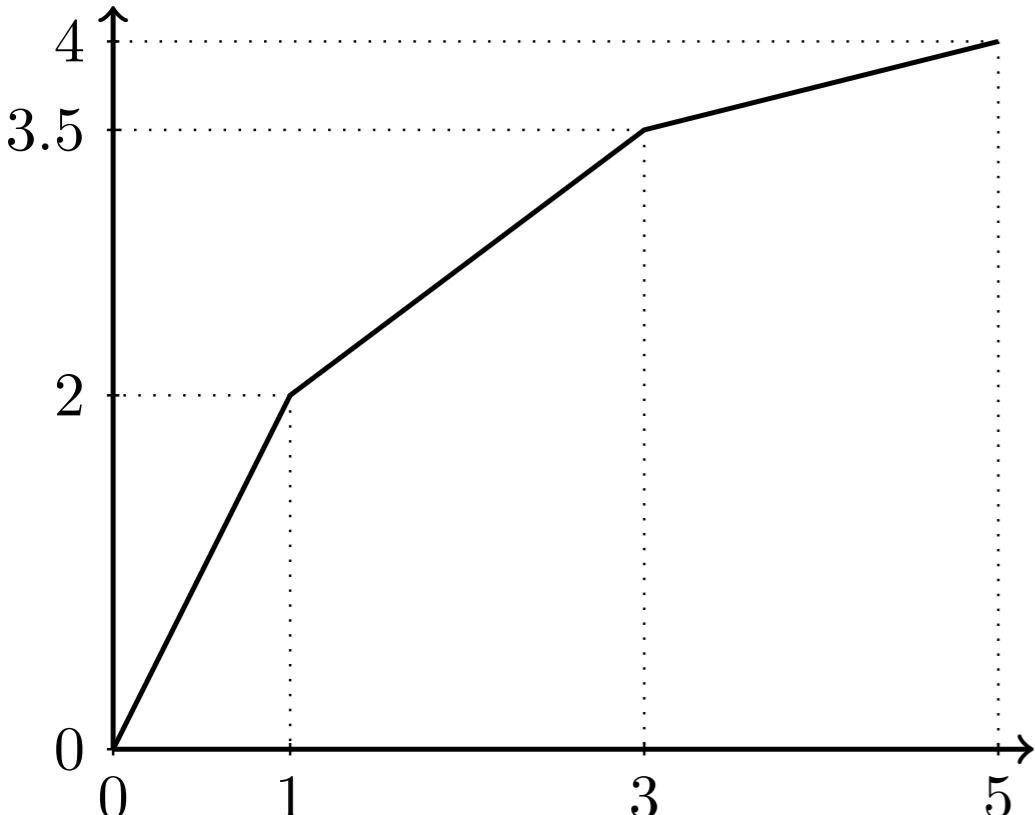
- We emphasize Non-Separable and/or discontinuous:



# Outline

- Applications of Piecewise Linear Functions.
- Modeling Piecewise Linear Functions.
- Computational Results.
- Extension to Discontinuous Functions.
- Final Remarks.

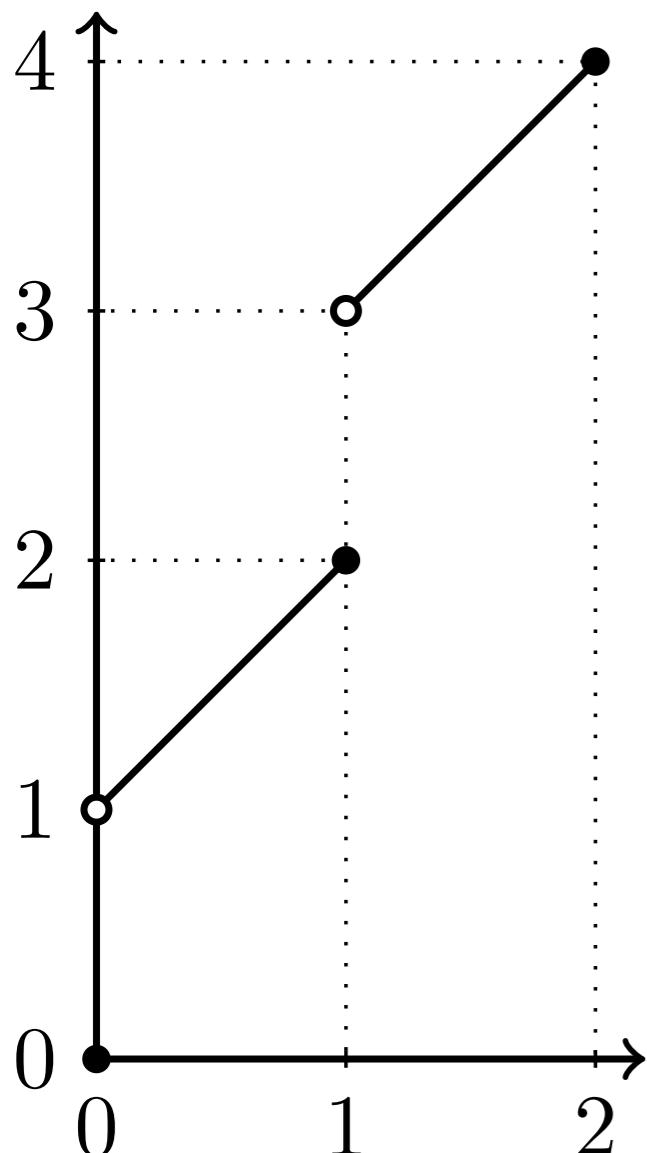
# Economies of Scale: Concave



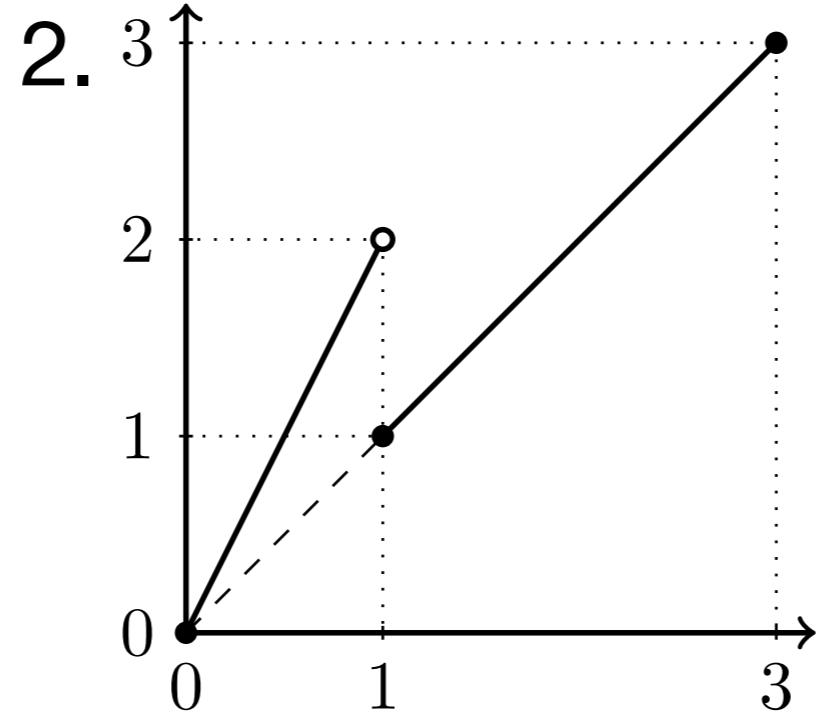
- Single and multi-commodity network flow.
- Applications in telecommunications, transportation, and logistics.
- (Balakrishnan and Graves 1989, ..., Croxton, et al. 2007).

# Fixed Charges and Discounts

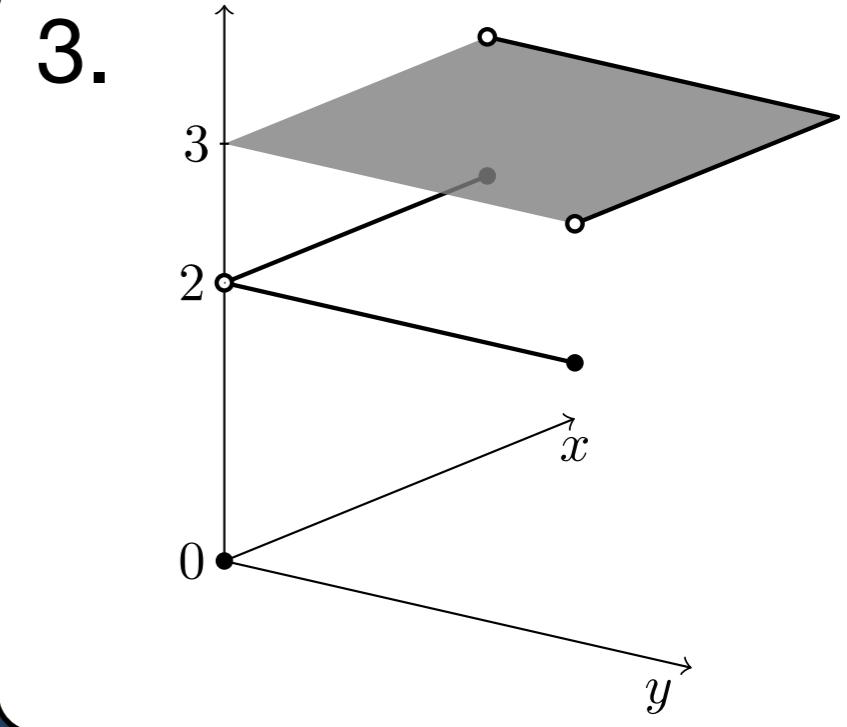
1.



2.

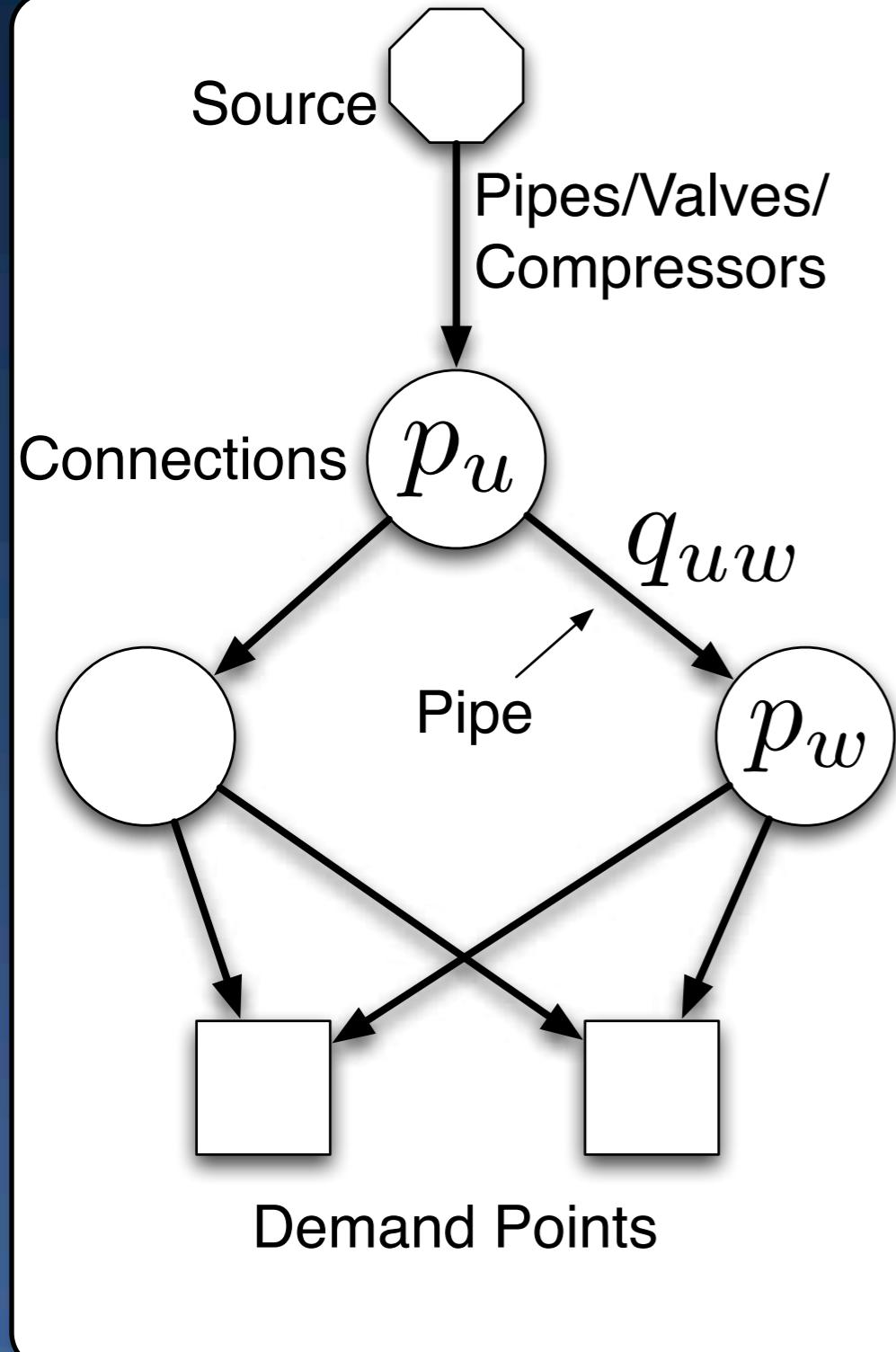


3.



1. Fixed Costs in Logistics.
2. Discounts (e.g. Auctions: Sandholm, et al. 2006, CombineNet).
3. Discounts in fixed charges (Lowe 1984).

# Non-Linear and PDE Constraints



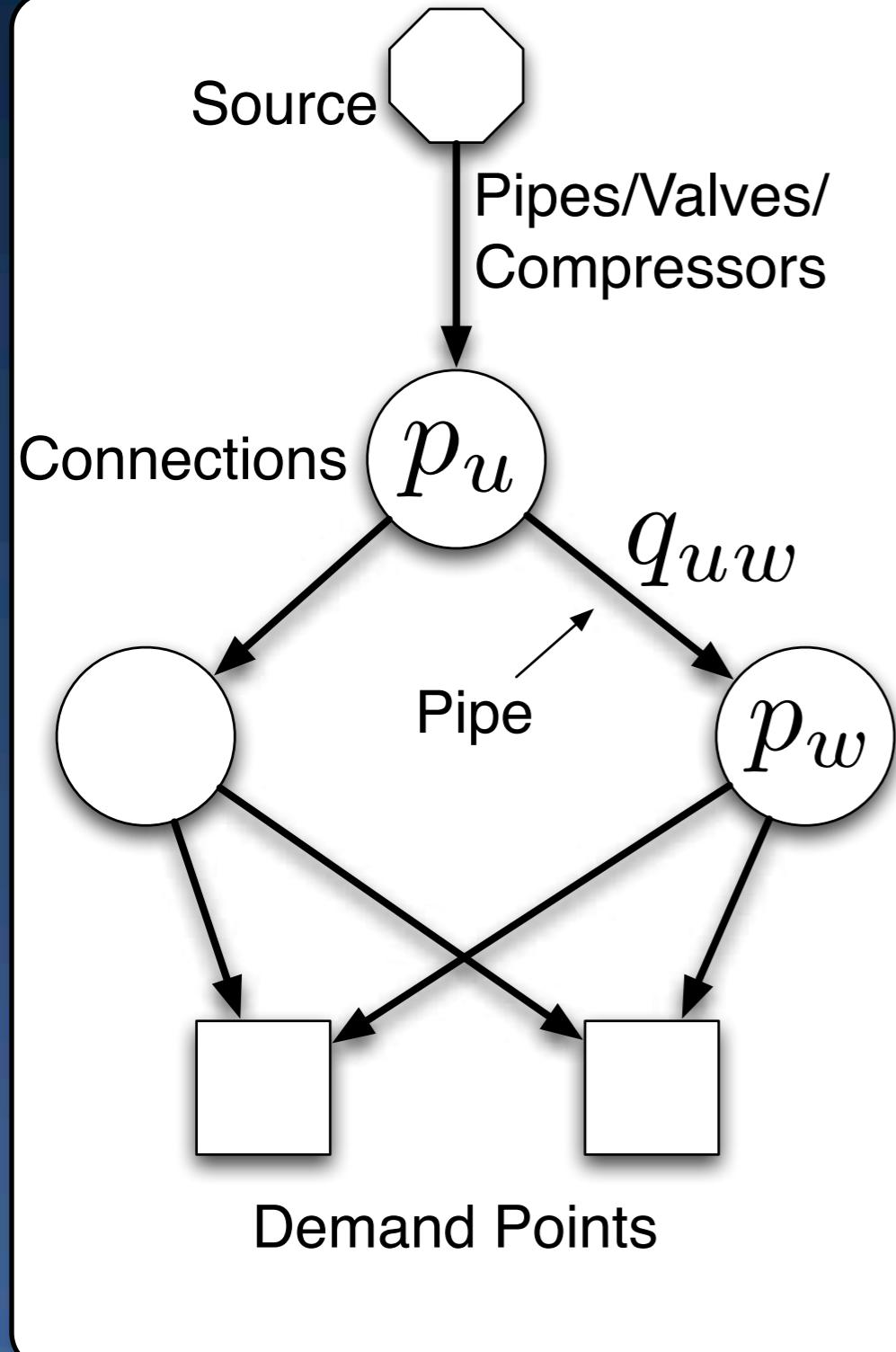
$p(x, t)$  = gas pressure  
 $q(x, t)$  = gas volume flow

$$A \frac{\partial \rho}{\partial t} + \rho_0 \frac{\partial q}{\partial x} = 0,$$

$$\frac{\partial p}{\partial x} = -\lambda \frac{|v|v}{2D} \rho.$$

- Gas Network Optimization  
(Martin et al. 2006).

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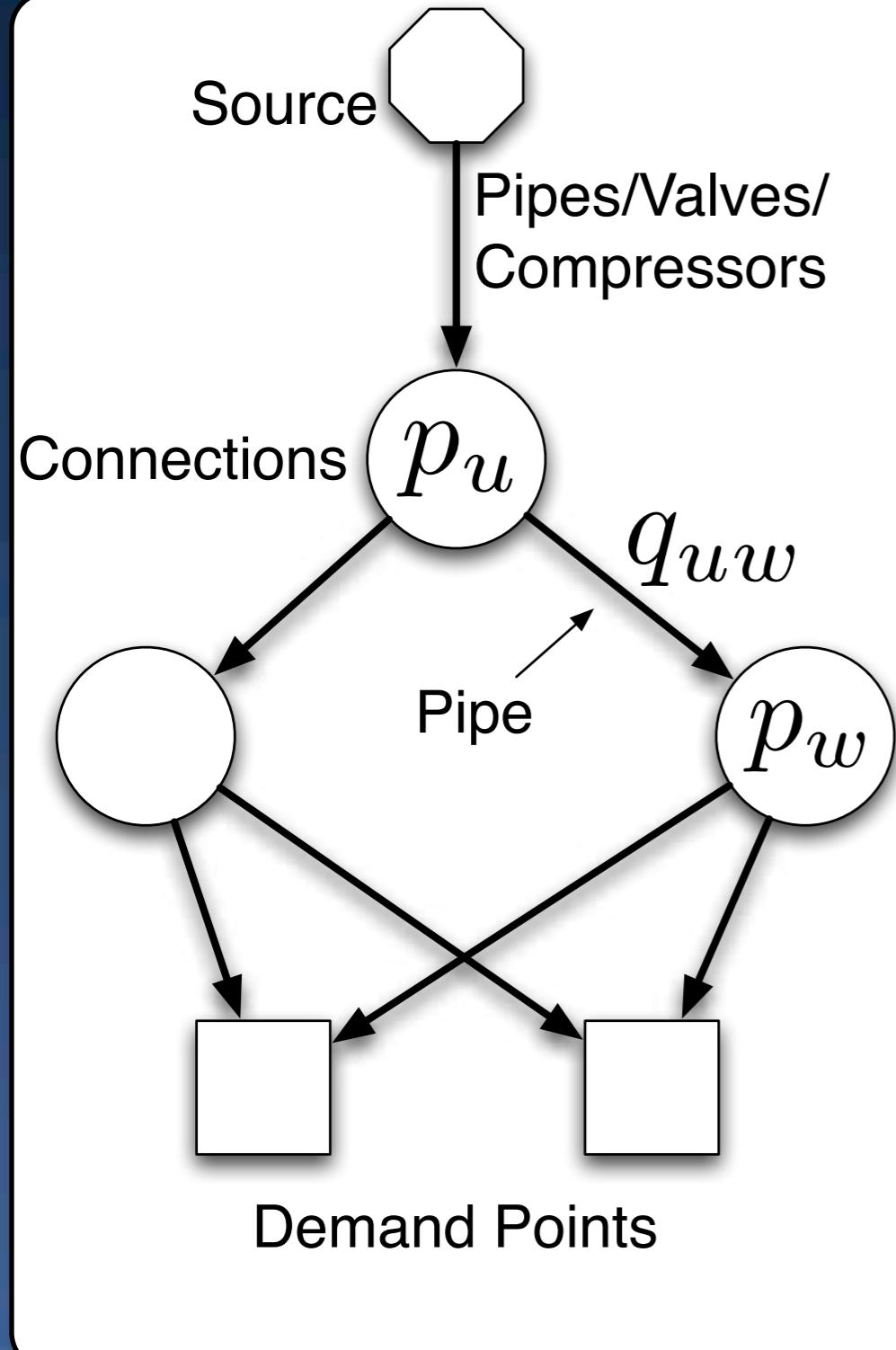
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Discretize non-linear stationary solution  $p_v = g(p_u, q_{uv})$

- Gas Network Optimization (Martin et al. 2006).

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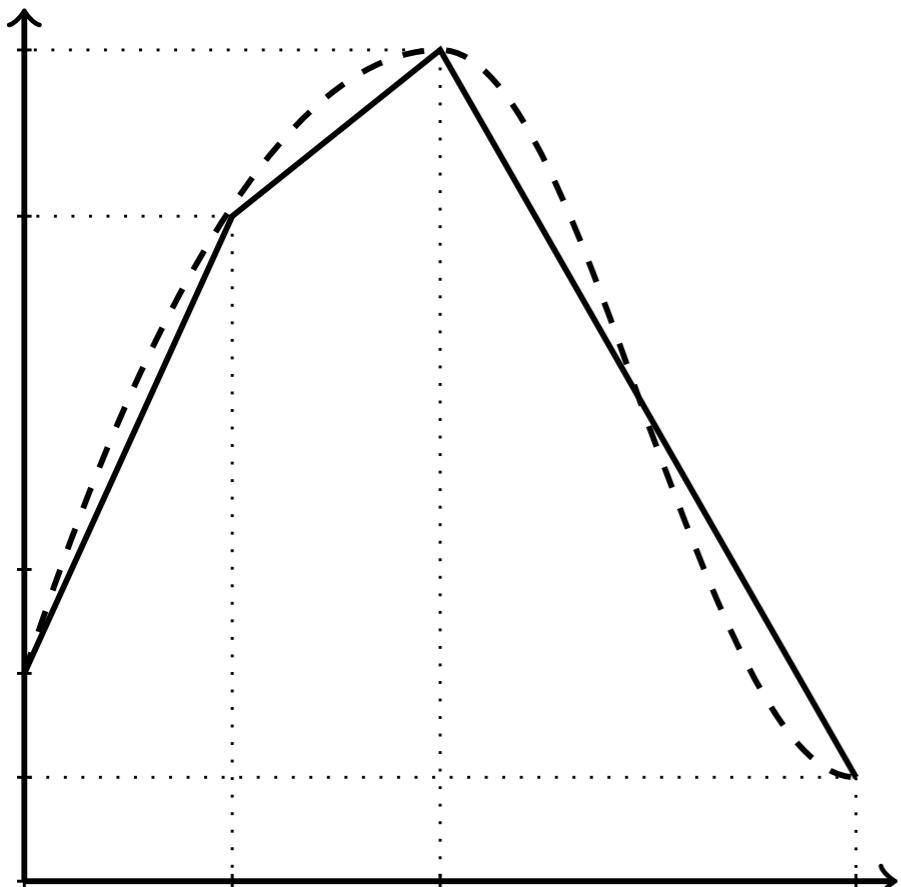
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Discretize PDE  
(Fügenschuh, et al. 2008)

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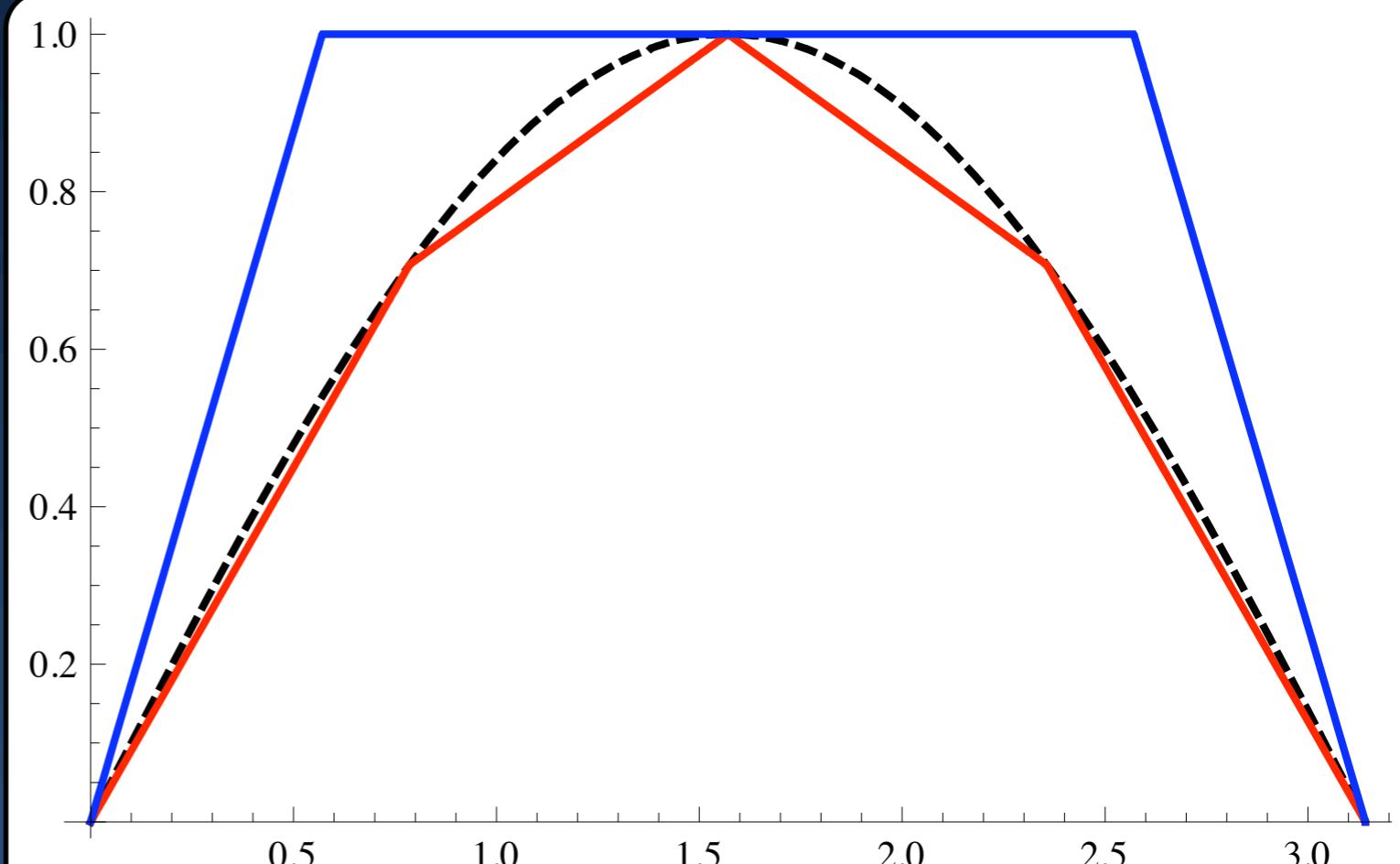
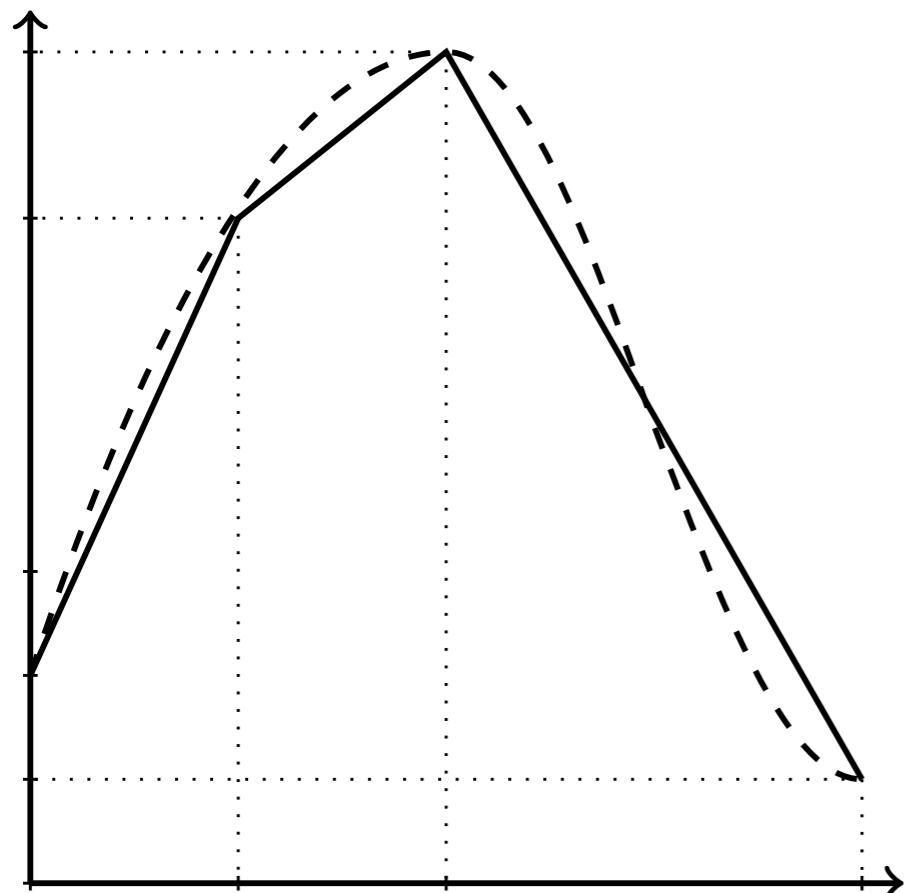
# Numerically Exact Global Optimization



- Process engineering (Bergamini et al. 2005, 2008, Computers and Chemical Eng.)
- Wetland restoration (Stralberg et al. 2009).



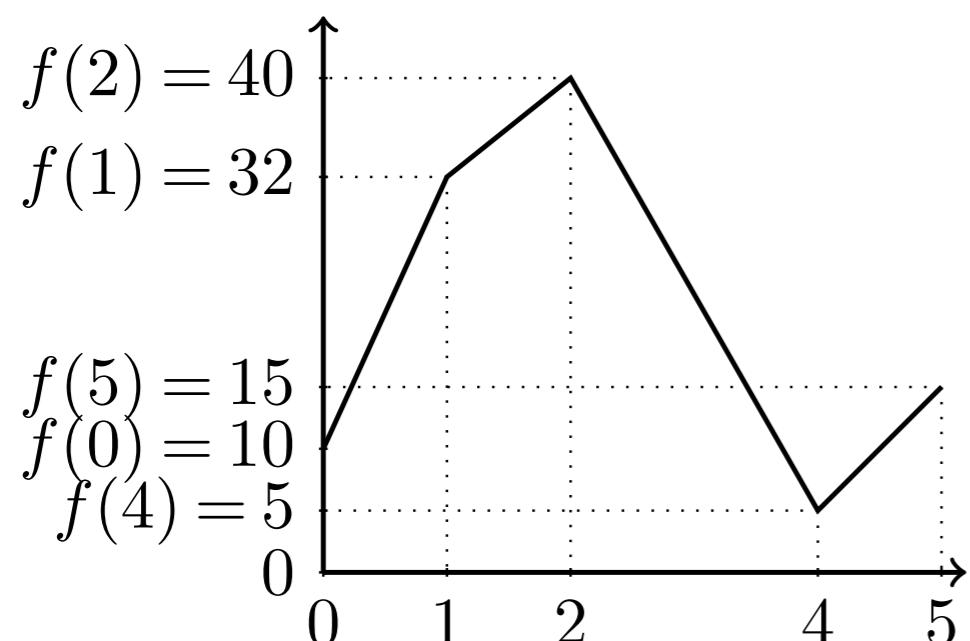
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# Piecewise Linear Functions: Definition



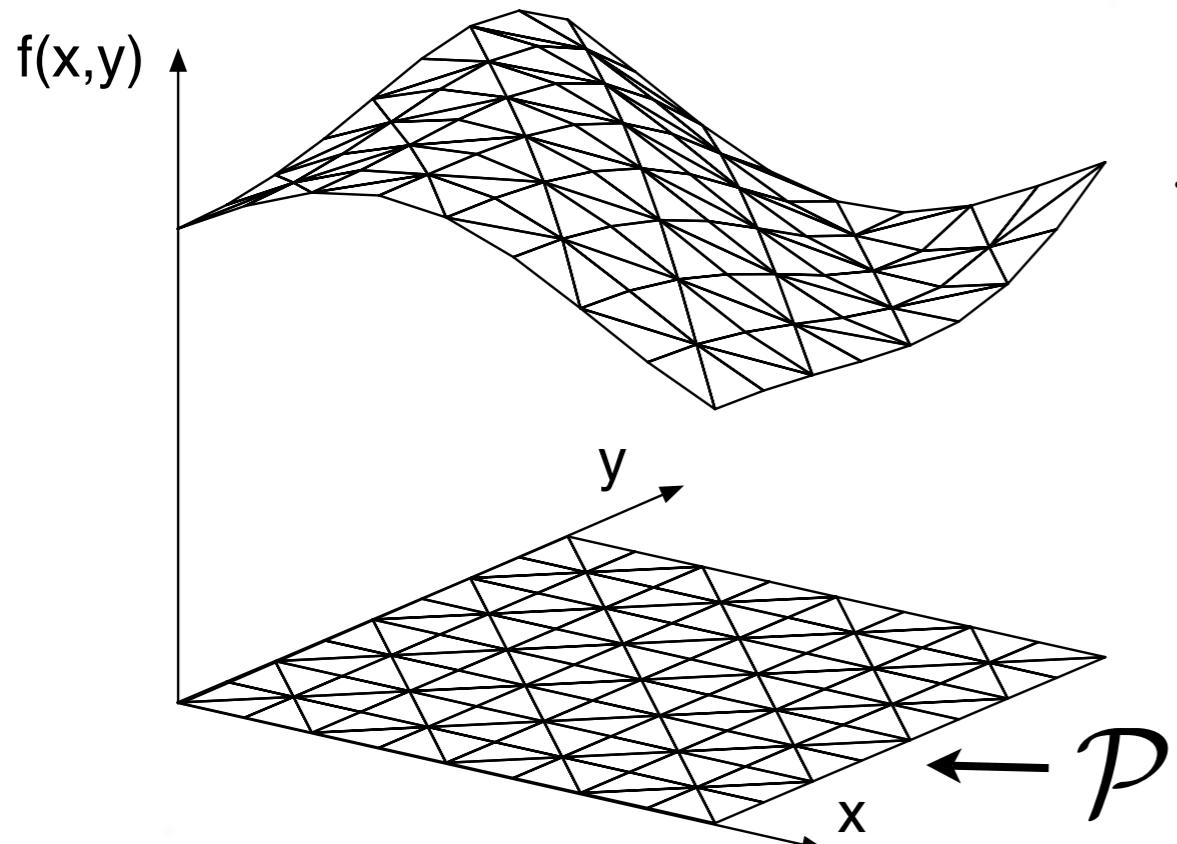
$$f(x) := \begin{cases} 22x + 10 & x \in [0, 1] \\ 8x + 24 & x \in [1, 2] \\ -17.5x + 75 & x \in [2, 4] \\ 10x - 35 & x \in [4, 5] \end{cases}$$

DEFINITION 1. Piecewise Linear  $f : D \subset \mathbb{R}^n \rightarrow \mathbb{R}$ :

$$f(x) := \begin{cases} m_P x + c_P & x \in P \quad \forall P \in \mathcal{P}. \end{cases}$$

for finite family of polytopes  $\mathcal{P}$  such that  $D = \bigcup_{P \in \mathcal{P}} P$

# Piecewise Linear Functions: Definition



$$f(x, y) := \begin{cases} 0.48x + 0.03y + 6 & (x, y) \in P_1 \\ \vdots \\ -0.4x - 0.04y + 8.45 & (x, y) \in P_{128} \end{cases}$$

$$P_1 := \{(x, y) \in \mathbb{R} : y \geq 0, x \leq 1, y - x \leq 0\}$$

$$\vdots$$

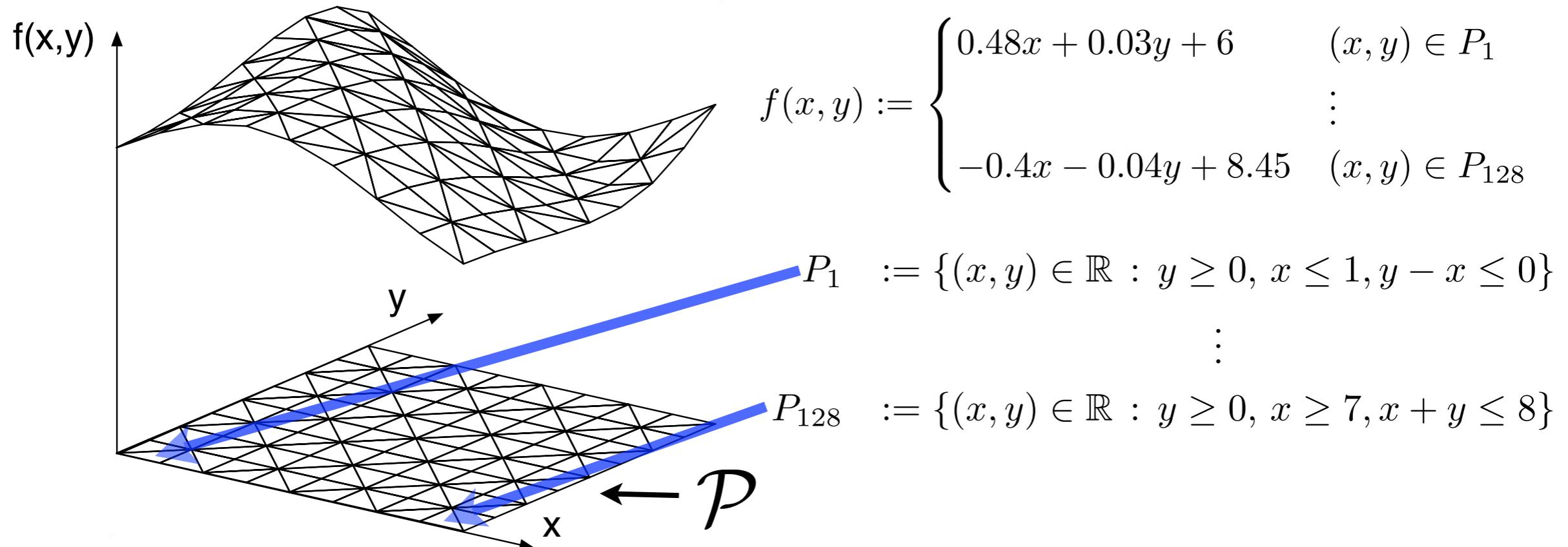
$$P_{128} := \{(x, y) \in \mathbb{R} : y \geq 0, x \geq 7, x + y \leq 8\}$$

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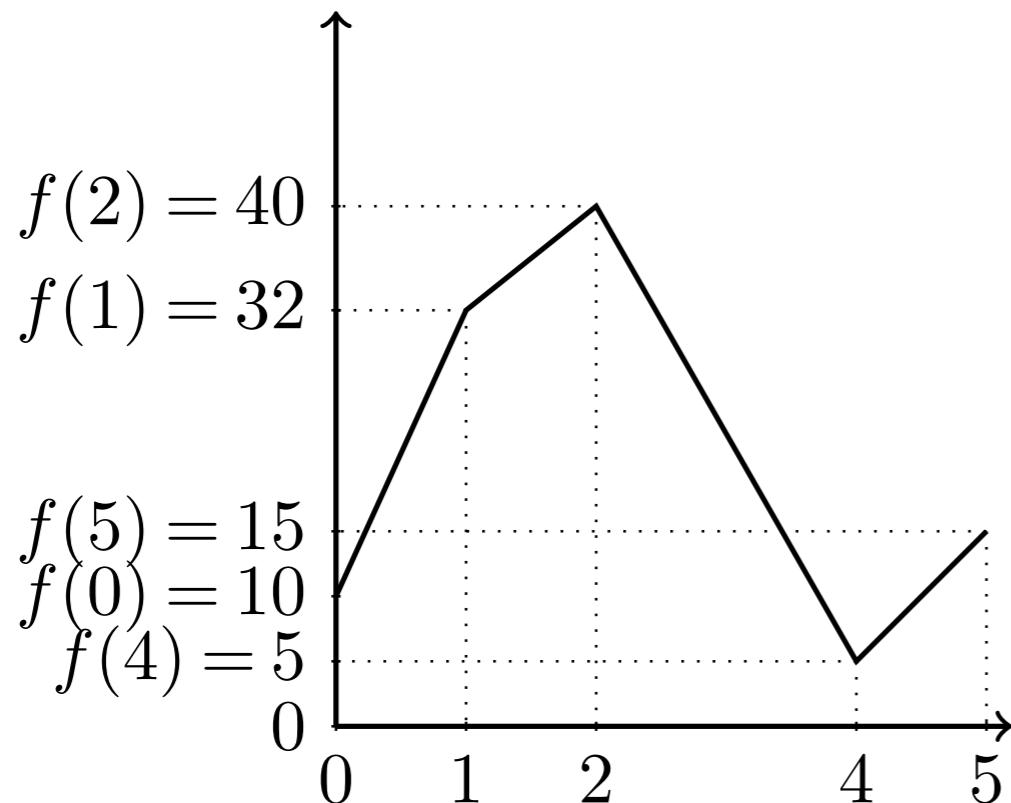
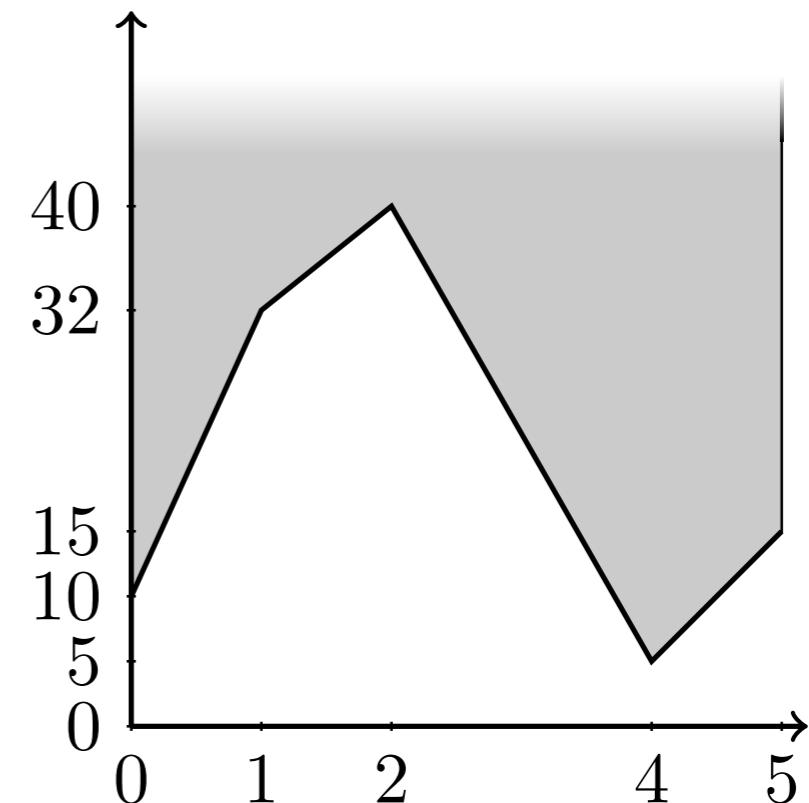
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# Modeling Function = Epigraph

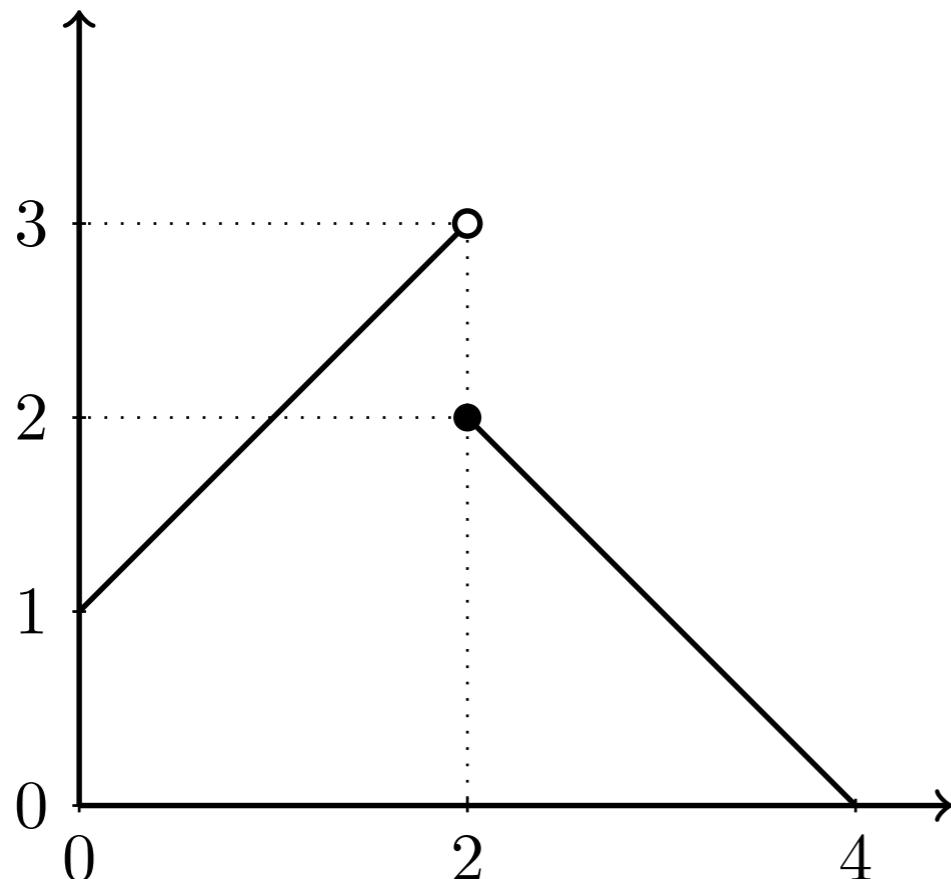
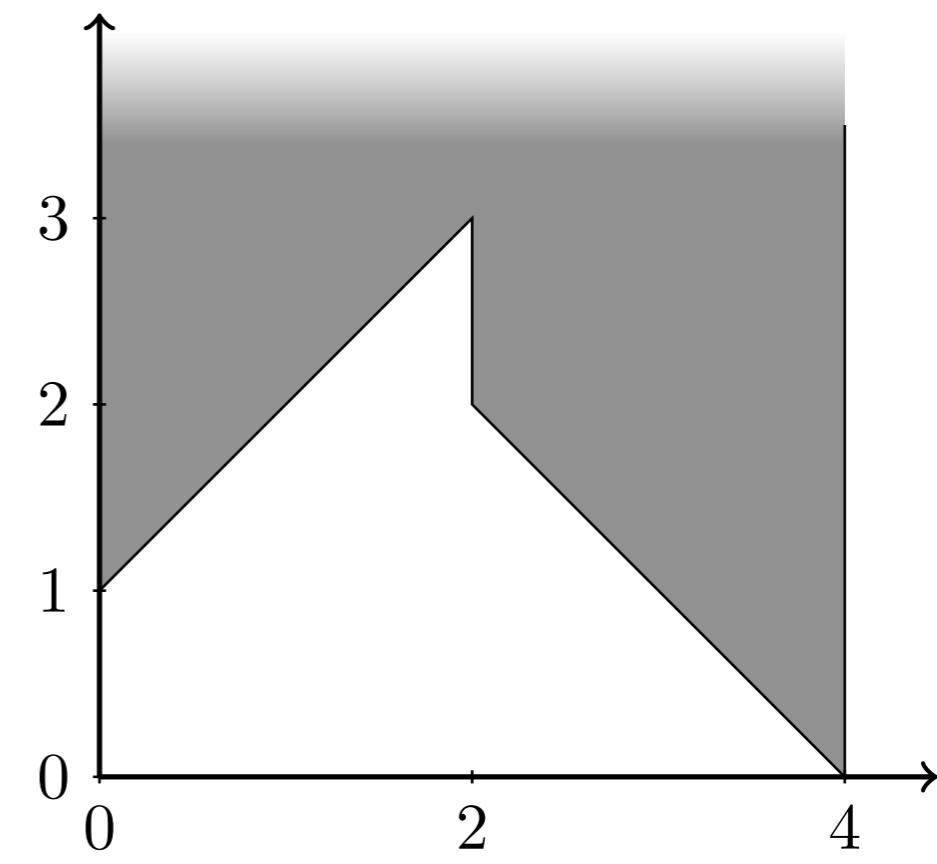
- $\text{epi}(f) := \{(x, z) \in D \times \mathbb{R} \subset \mathbb{R}^n \times \mathbb{R} : f(x) \leq z\}$ .

(a)  $f$ .(b)  $\text{epi}(f)$ .

- Example:  $f(x) \leq 0 \Leftrightarrow (x, z) \in \text{epi}(f), z \leq 0$

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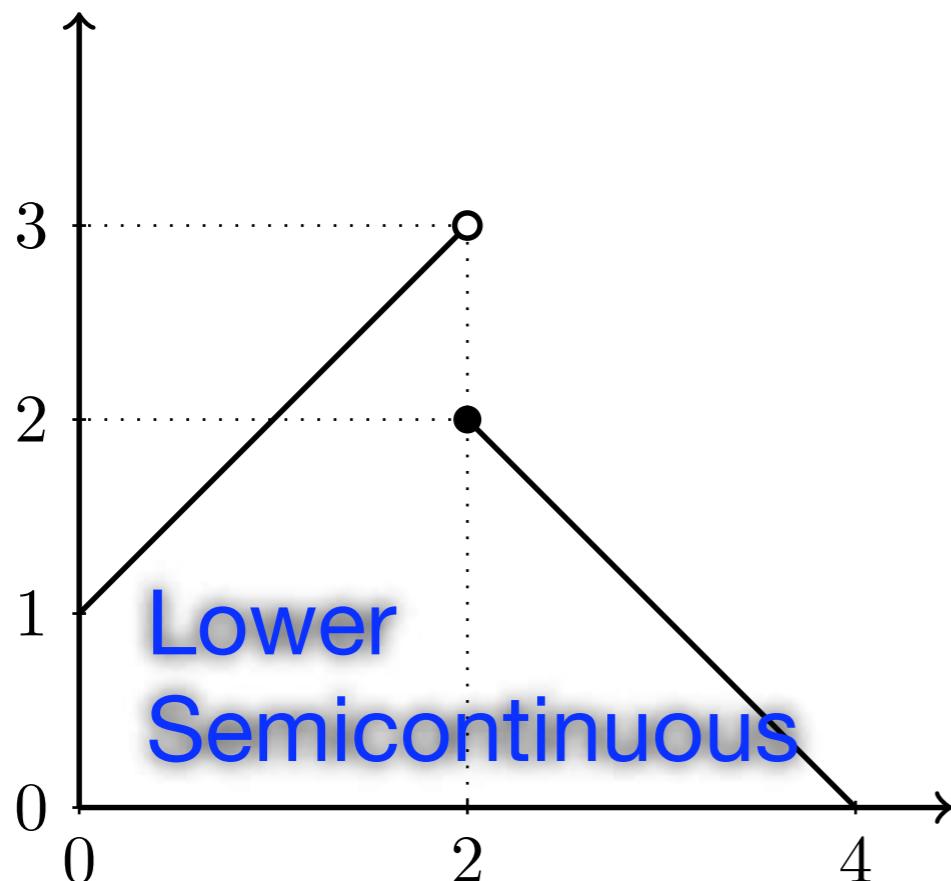
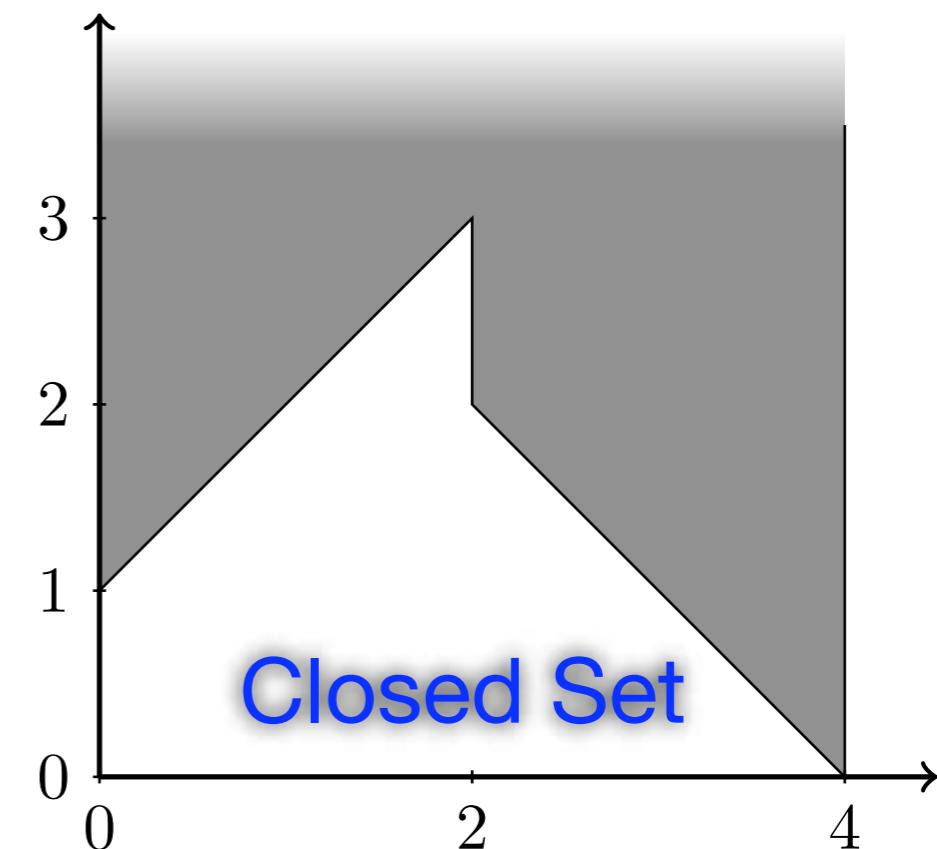
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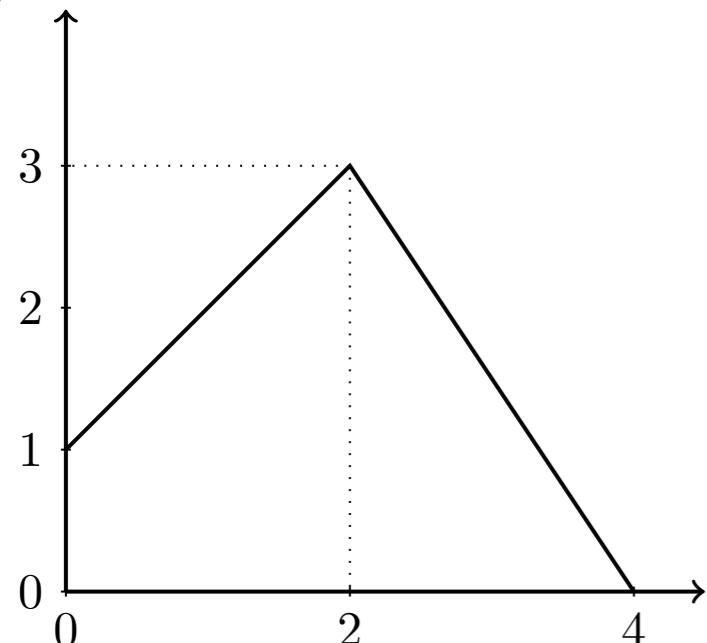
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# Convex Comb. Univariate (Dantzig, 1960)

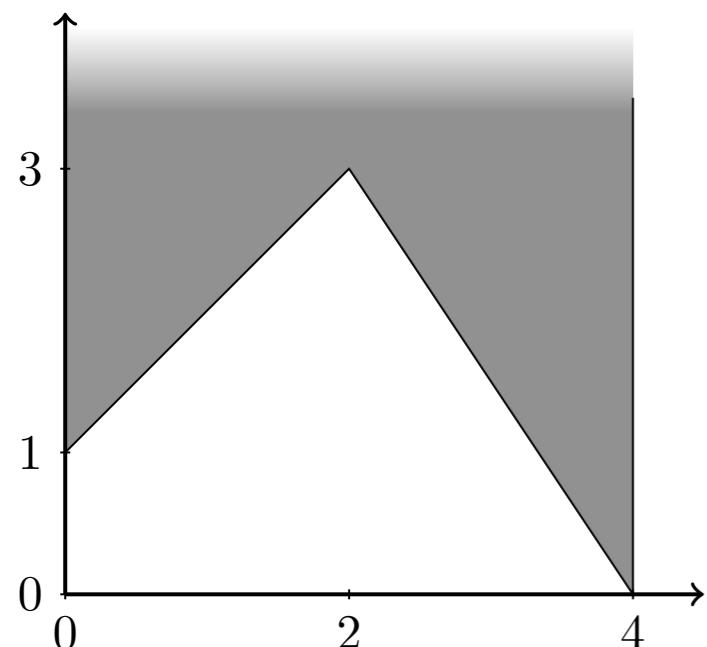


$$f(x) := \begin{cases} x + 1 & x \in [0, 2] \leftarrow P_1 \\ 6 - 3/2x & x \in [2, 4] \leftarrow P_2 \end{cases}$$

$V(P)$  = vertices of  $P$ .

$$\mathcal{V}(\mathcal{P}) := V(P_1) \cup V(P_2) = \{0, 2, 4\}.$$

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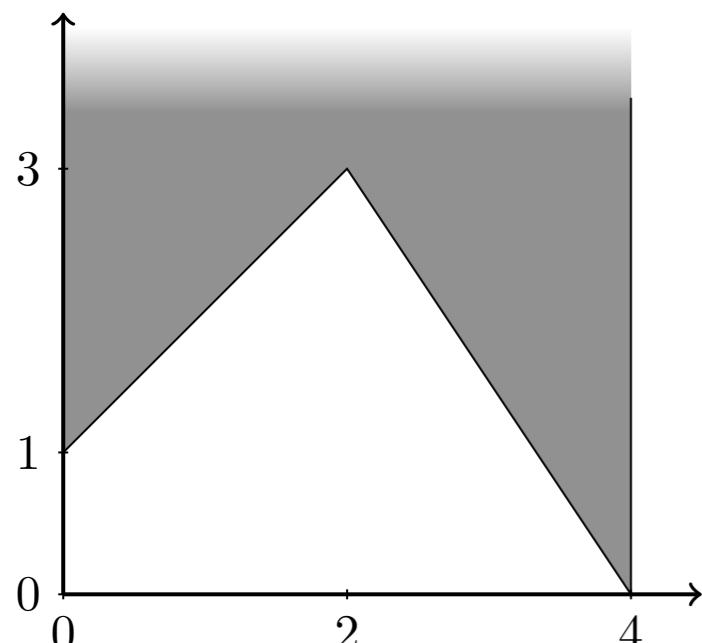


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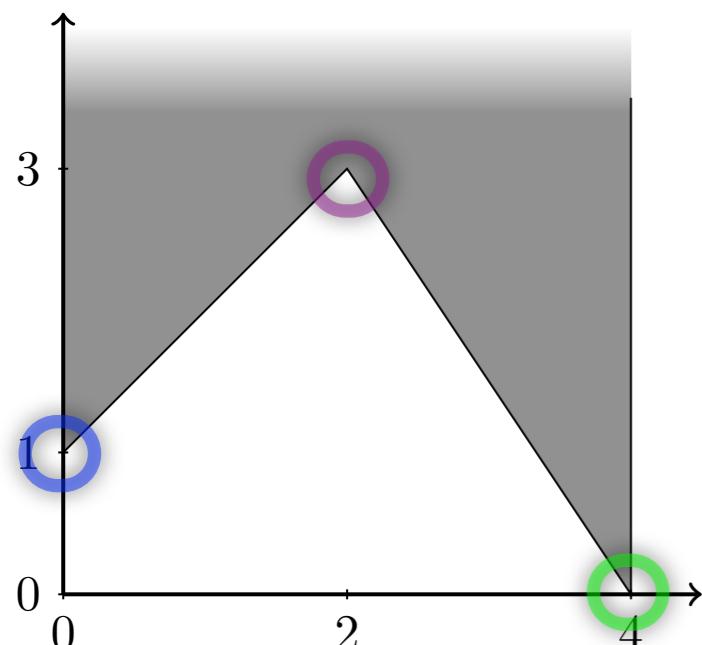
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idea: write  $(x, y) \in \text{epi}(f)$   
as convex combination of  
 $(v, f(v))$  for  $v \in \mathcal{V}(\mathcal{P})$ .

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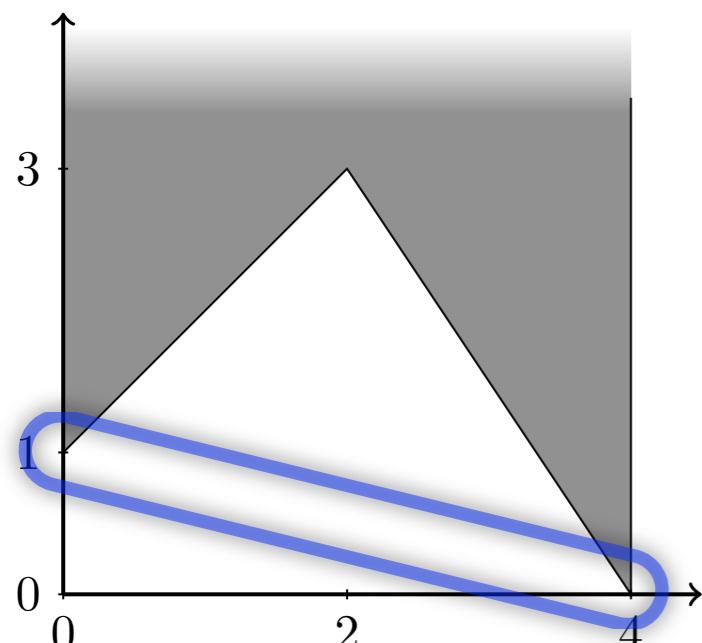
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$$x = 0\lambda_0 + 2\lambda_2 + 4\lambda_4$$

$$z \geq 1\lambda_0 + 3\lambda_2 + 0\lambda_4$$

$$1 = \lambda_0 + \lambda_2 + \lambda_4, \quad \lambda_0, \lambda_2, \lambda_4 \geq 0$$

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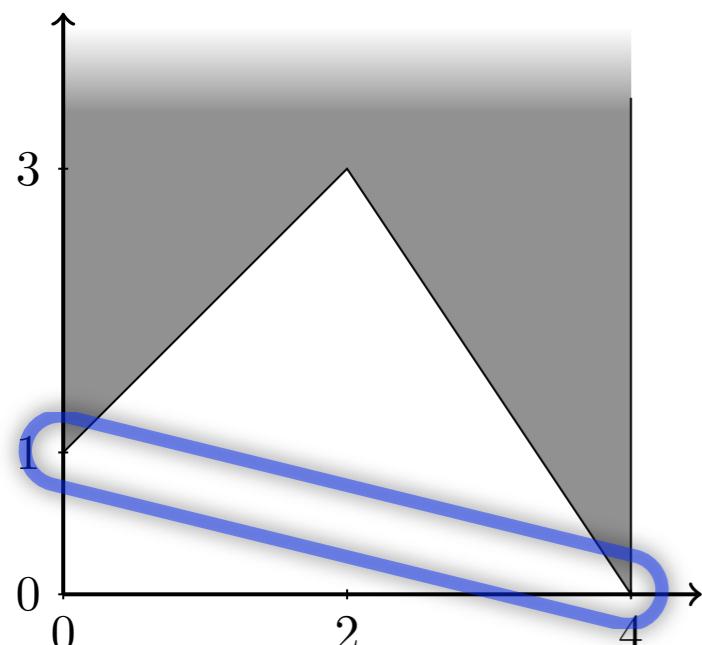
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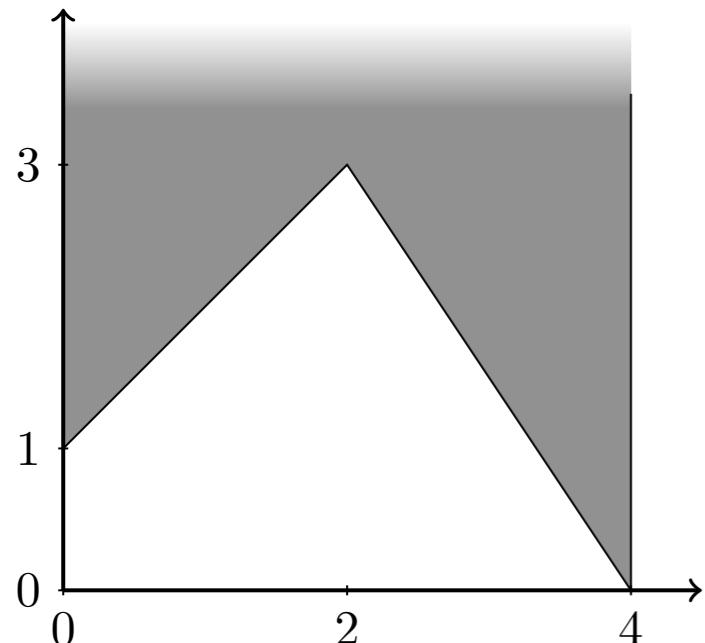
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$$\lambda_0 \leq y_{P_1}, \quad \lambda_2 \leq y_{P_1} + y_{P_2}, \quad \lambda_4 \leq y_{P_2}$$

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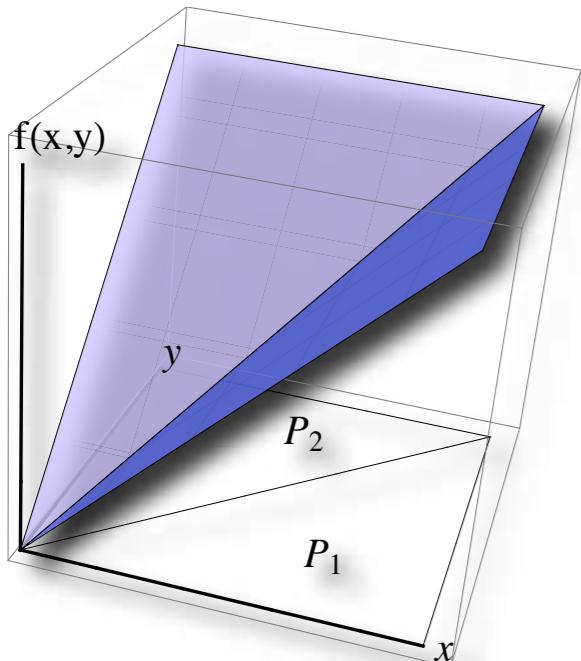
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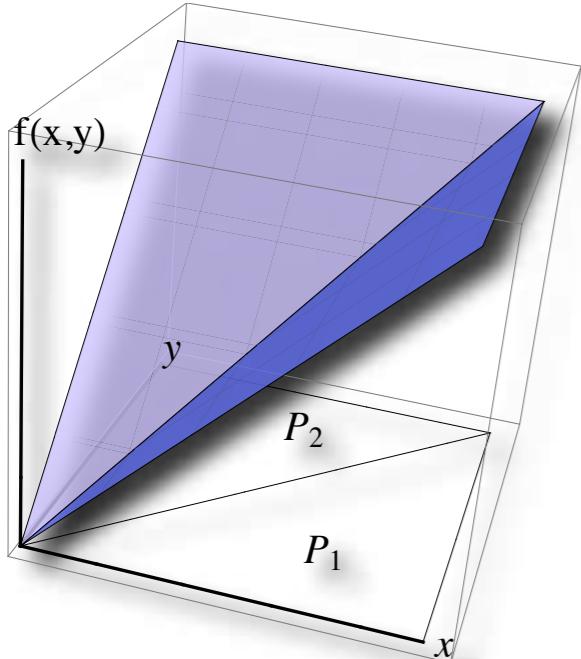
# CC Multivariate (Lee and Wilson, 2001)



$$f(x, y) := \begin{cases} x & (x, y) \in P_1 \\ y & (x, y) \in P_1 \end{cases}$$

$$P_1 := \{(x, y) : x \leq 1, 0 \leq y \leq x\}, \quad P_2 := \dots$$

$$\mathcal{V}(\mathcal{P}) := \{(0, 0), (1, 0), (0, 1), (1, 1)\}.$$

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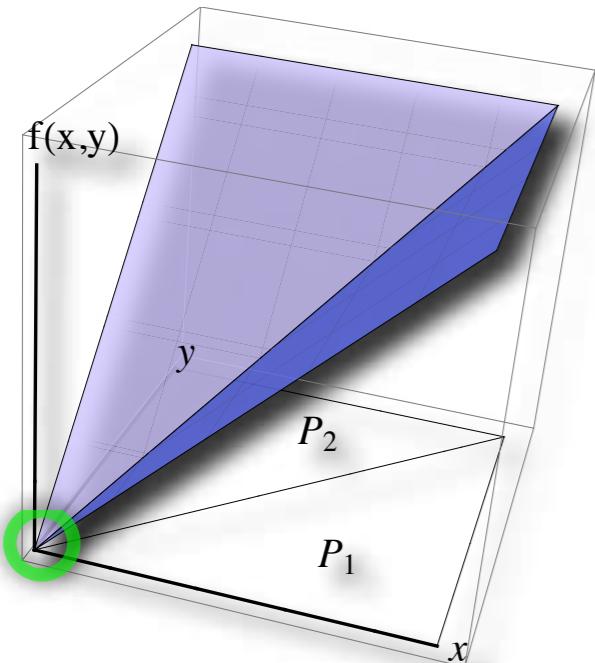
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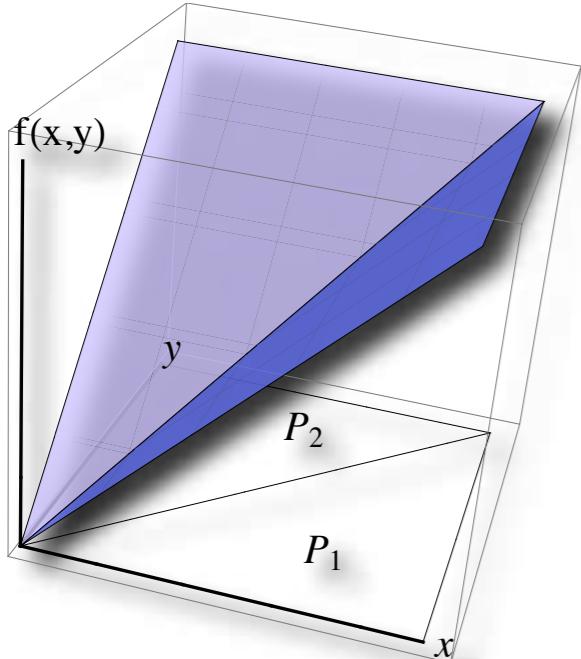
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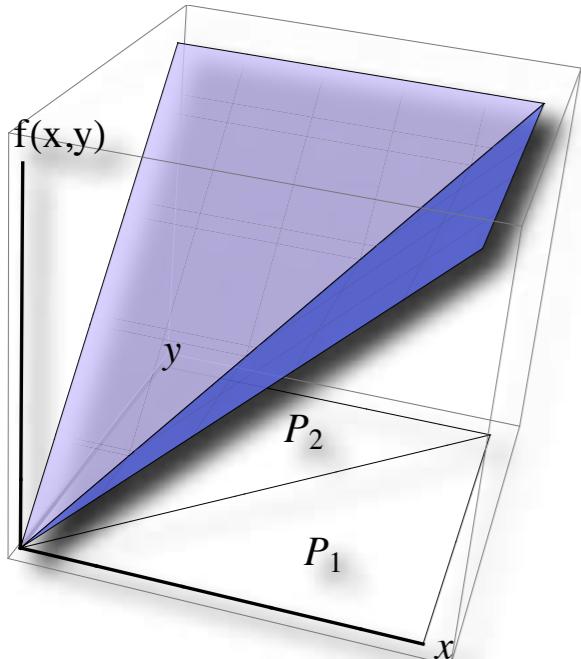
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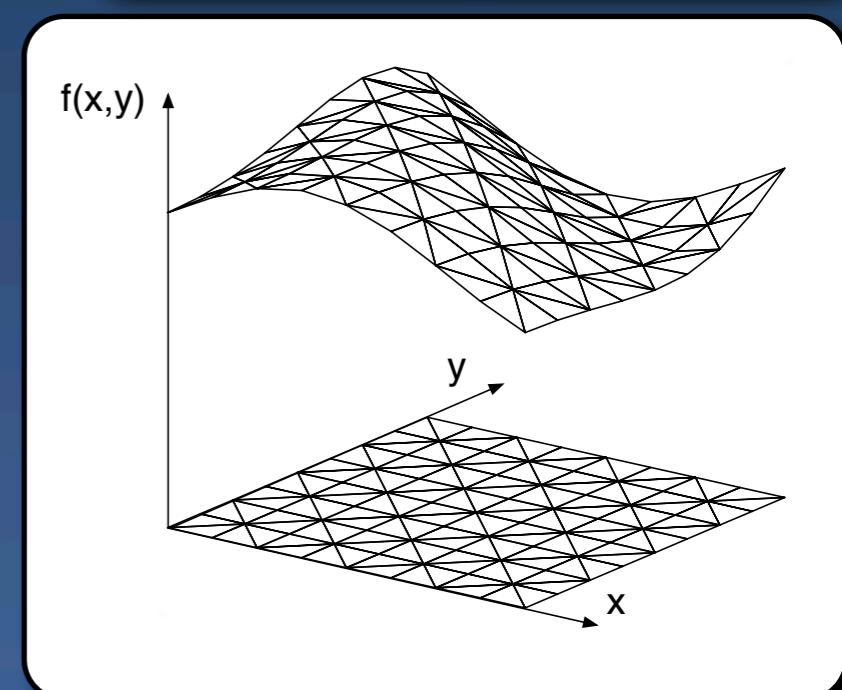
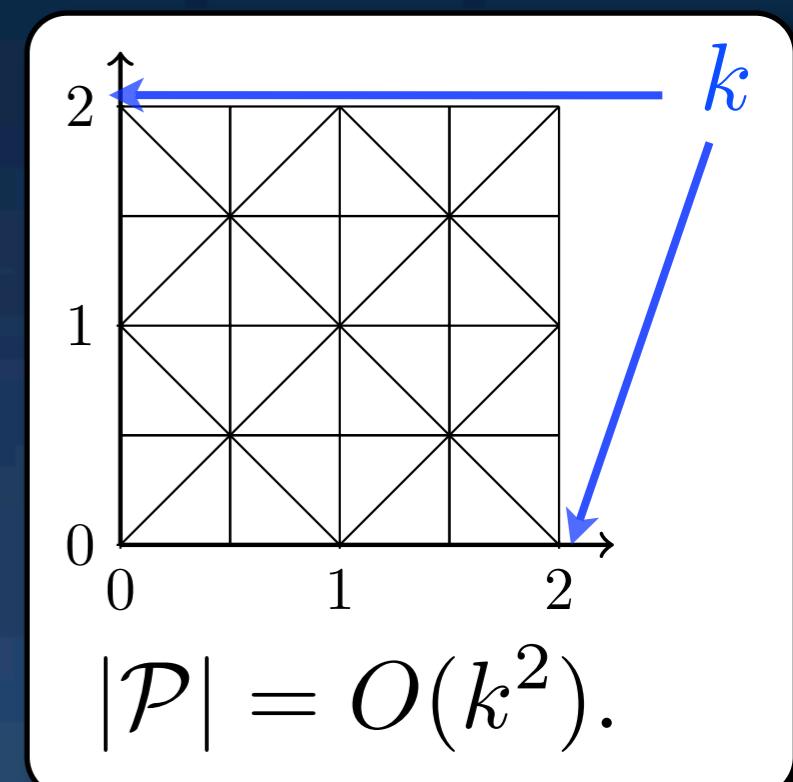
$$1 = \lambda_{(0,0)} + \lambda_{(1,0)} + \lambda_{(0,1)} + \lambda_{(1,1)}, \quad \lambda_{(0,0)}, \dots, \lambda_{(1,1)} \geq 0$$

$$\lambda_{(0,0)} \leq y_{P_1} + y_{P_1} \dots \quad y_{P_1} + y_{P_1} = 1, \quad y_{P_1}, y_{P_1} \in \{0, 1\}$$

Polytopes that have  
(0,0) as a vertex.

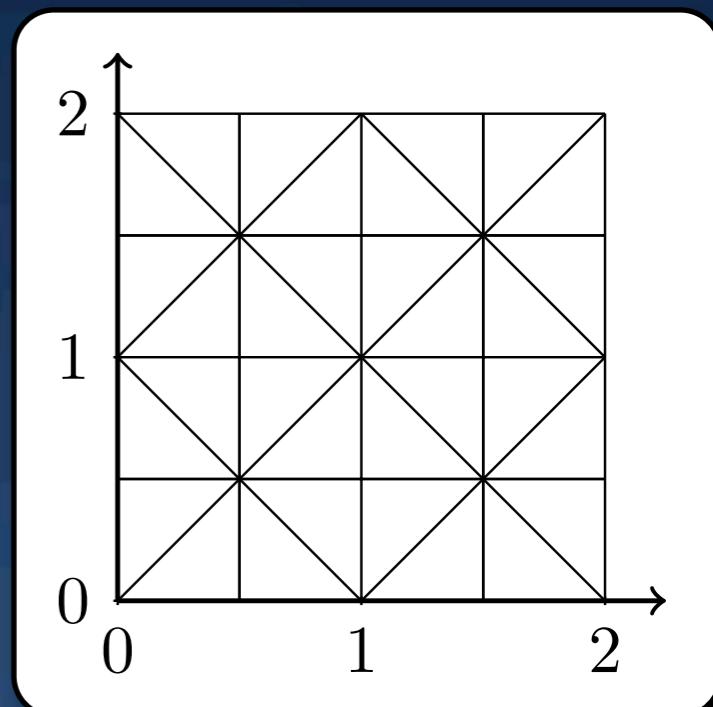
# Other issues: Log models & Strength

- Strength:
  - Popular models are strong.
  - Standard MILP techniques can yield weak models.
- Size of existing models is linear in  $|\mathcal{P}|$ :
  - We can get models with “size” logarithmic in  $|\mathcal{P}|$  (Vielma and Nemhauser 2008, Vielma et al. 2009).



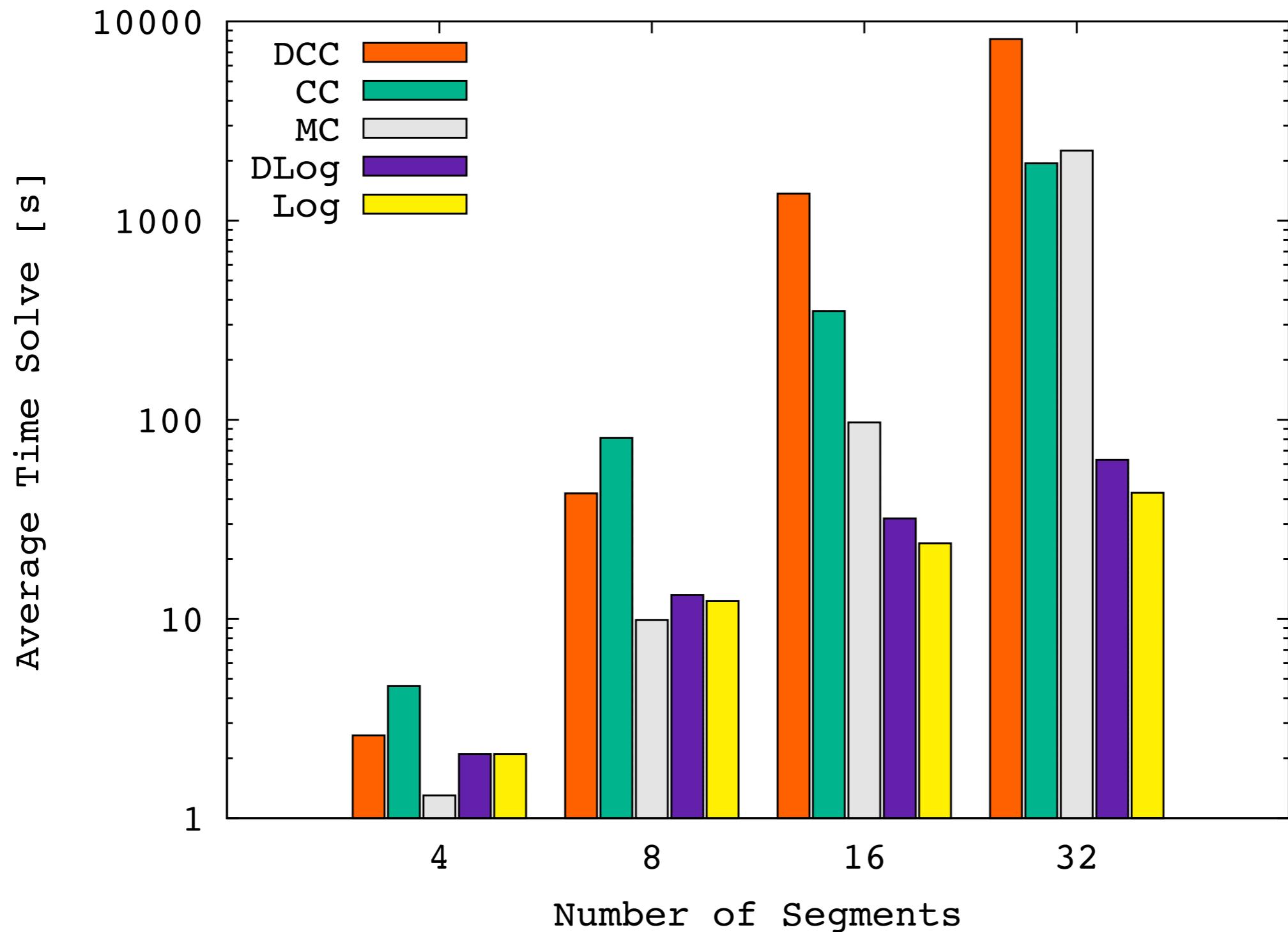
# Computation: Transportation Problems

- Minimization Problems:
  - Univariate Objective: Sum of 100 univariate functions, each affine in  $k$  segments.
  - Multivariate Objective: Sum of 10 bivariate functions, each affine in a  $k \times k$  grid.
- Solver: CPLEX 11 on 2.4Ghz machine.

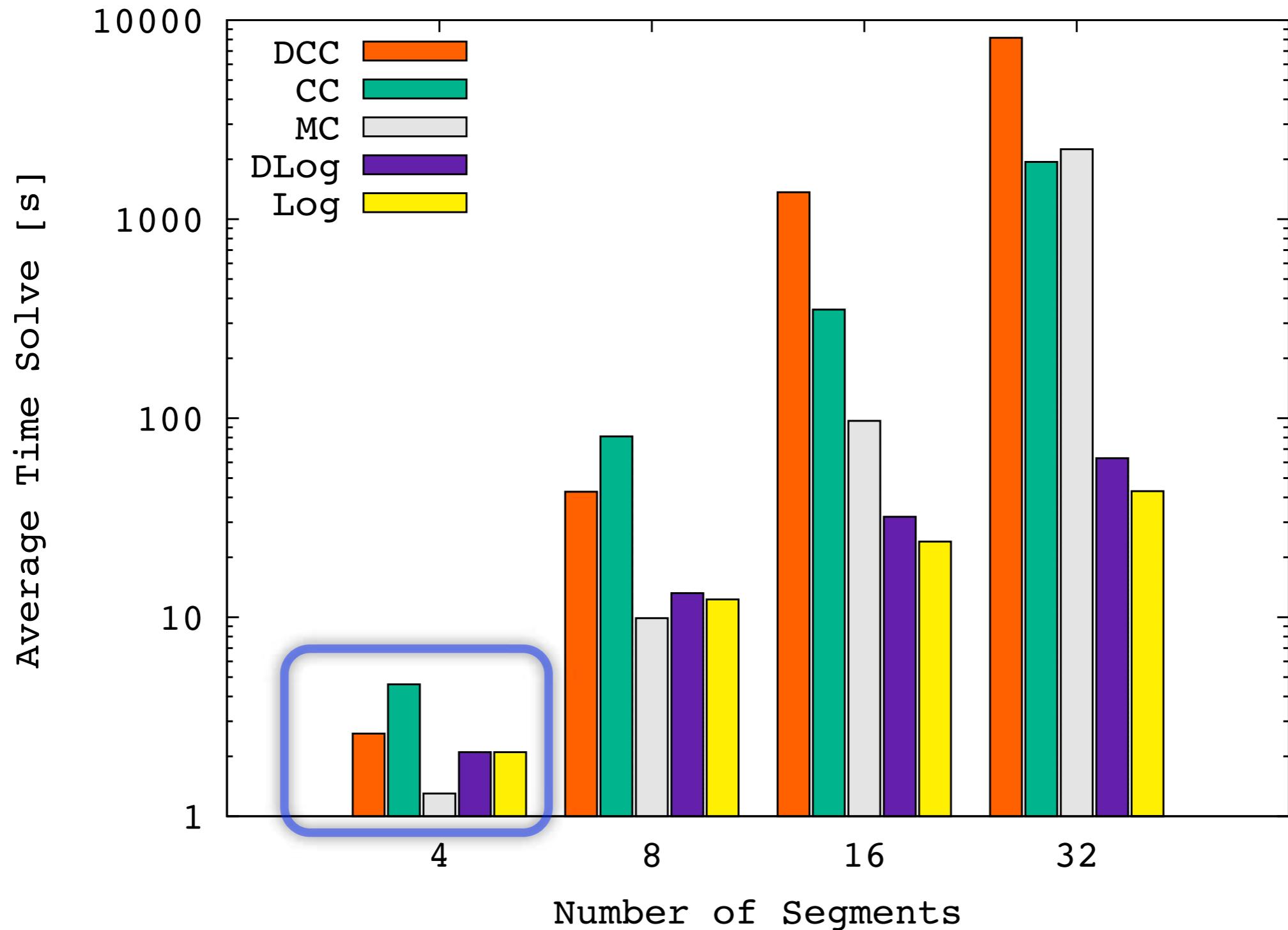


$(x, y) \rightarrow g(\|(x, y)\|)$   
Concave PLF  $g(\cdot)$

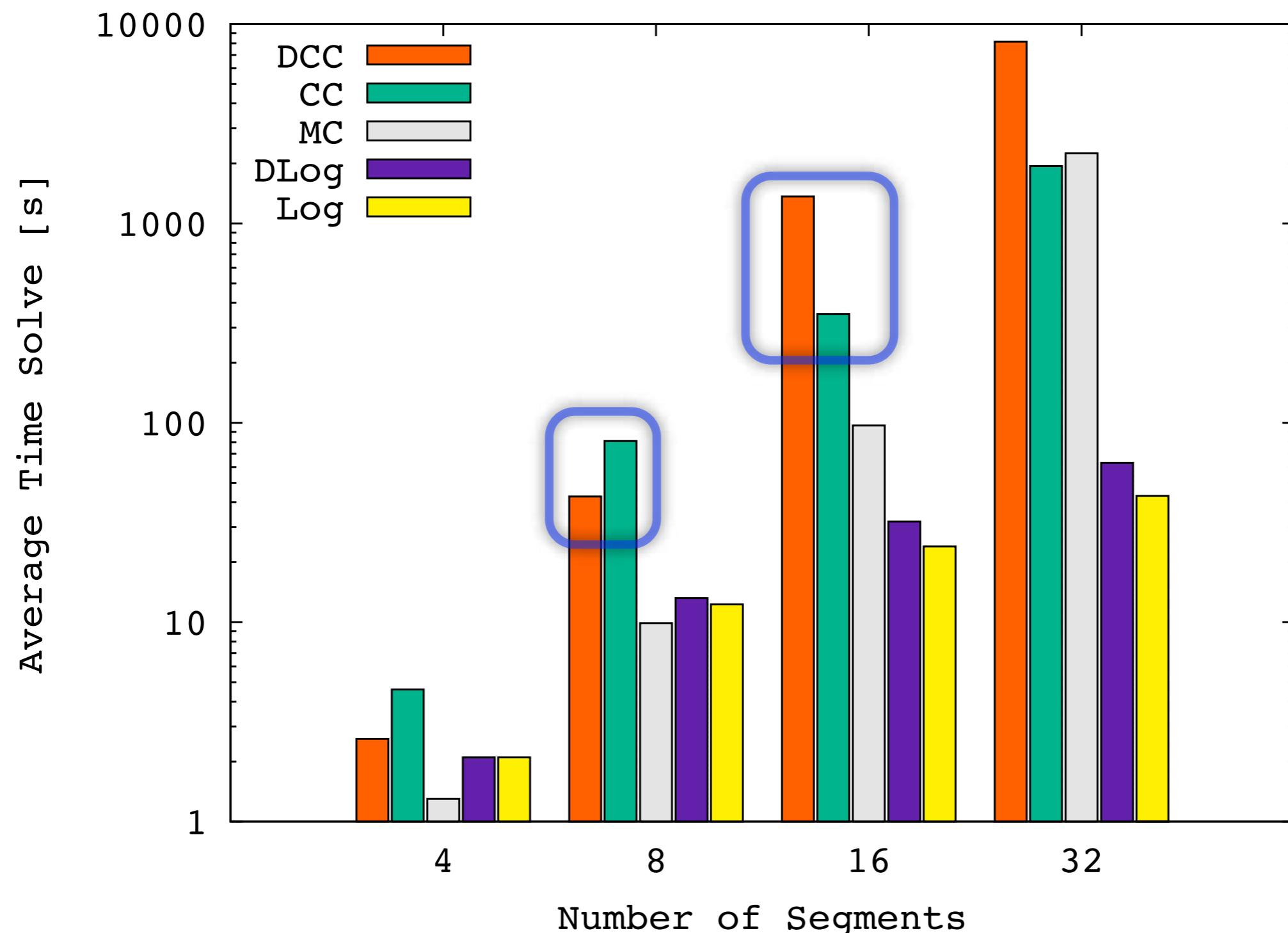
# Univariate Case (Separable)



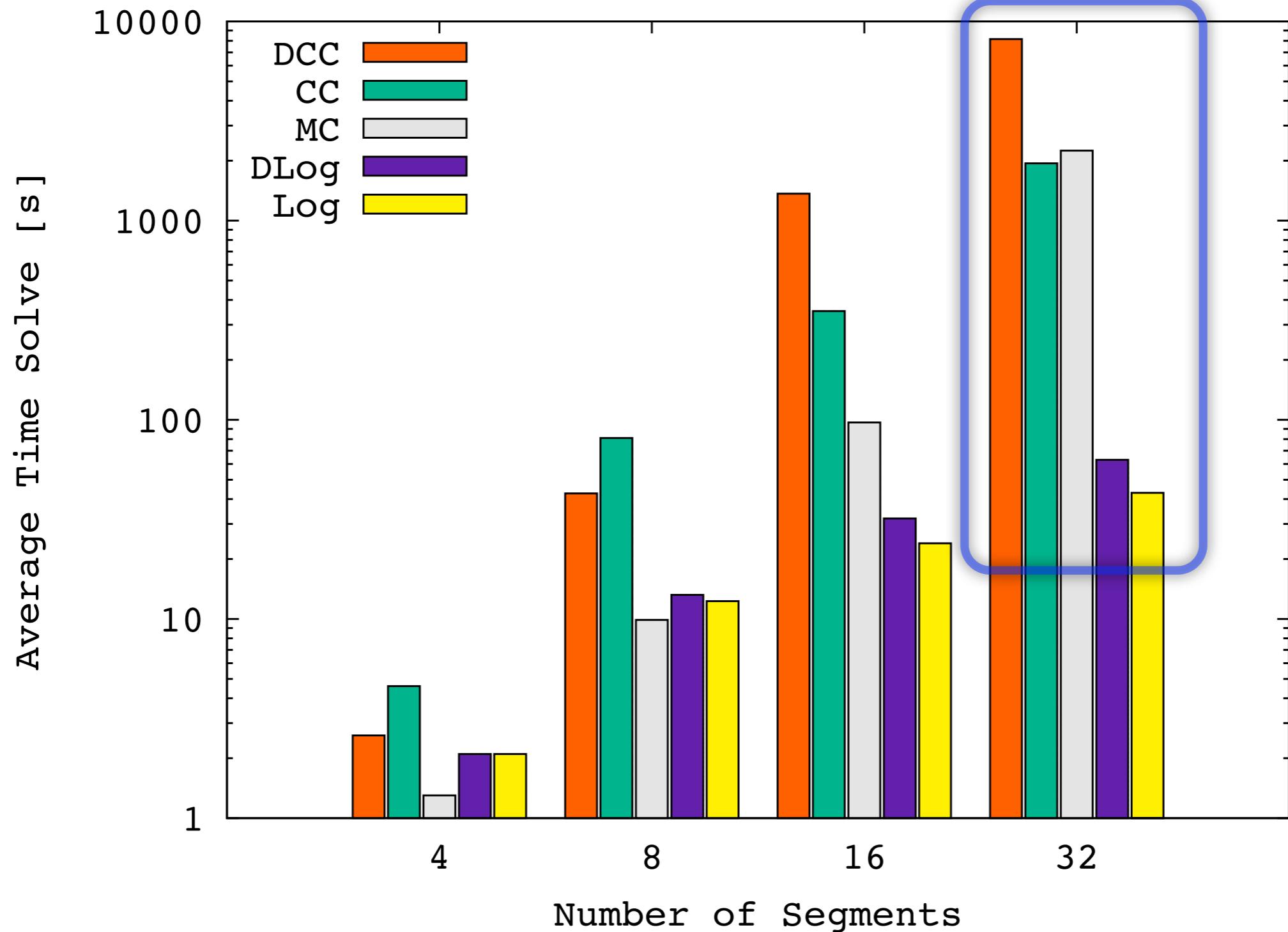
# Univariate Case (Separable)



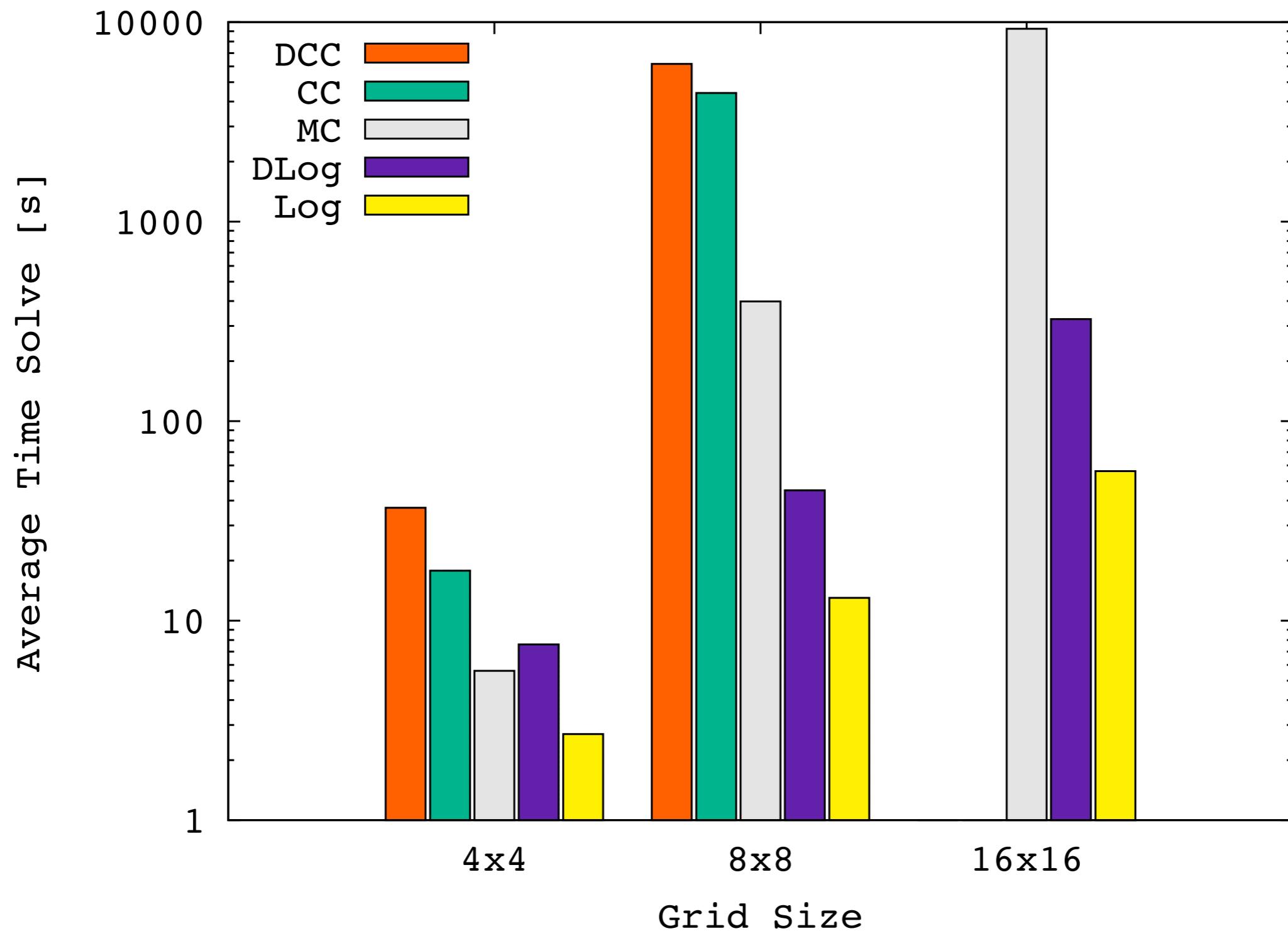
# Univariate Case (Separable)



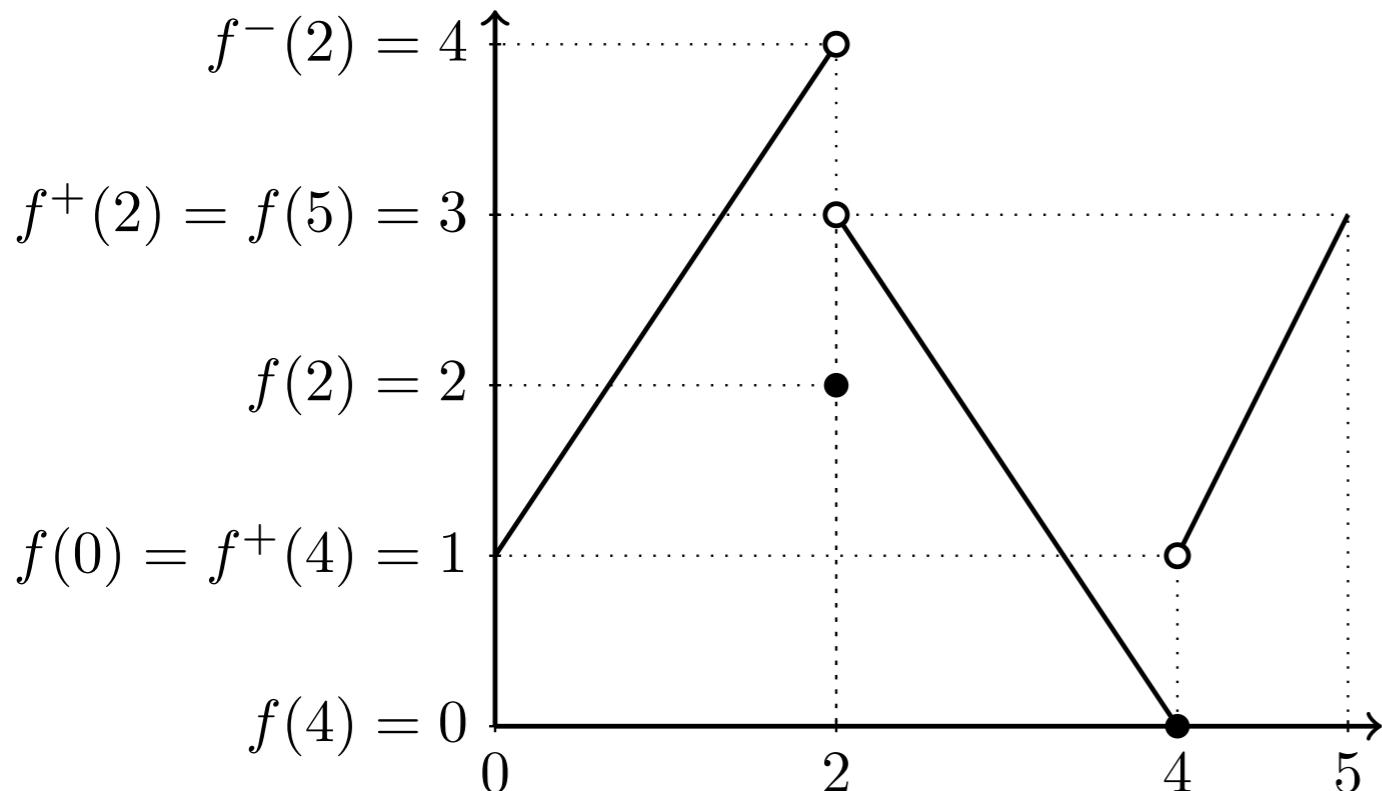
# Univariate Case (Separable)



# Multivariate Case (Non-Separable)



# Lower Semicontinuity (LSC)



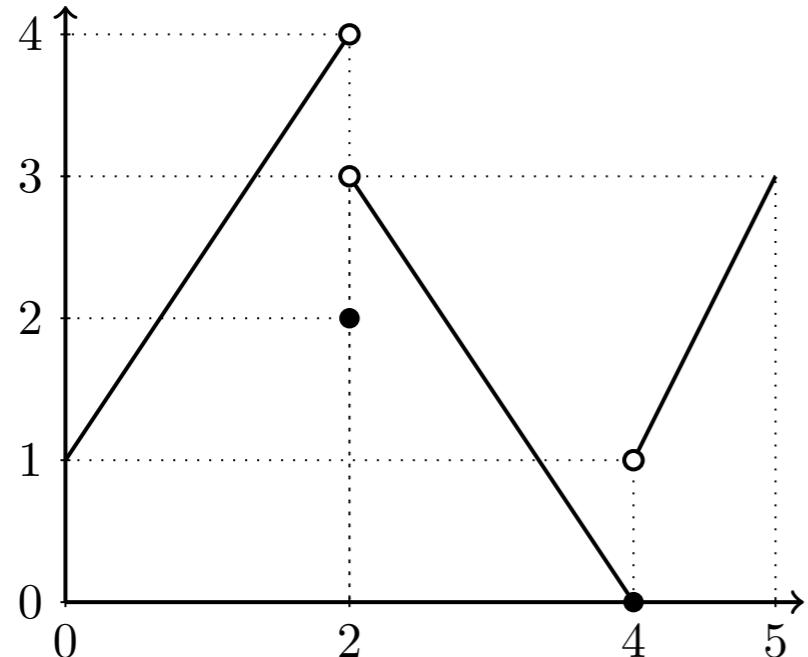
$$f^-(d) = \lim_{\substack{x \rightarrow d \\ x \leq d}} f(x)$$

$$f^+(d) = \lim_{\substack{x \rightarrow d \\ x \geq d}} f(x).$$

- Lower Semicontinuity:

$f(x) \leq f^-(x), f^+(x) \Leftrightarrow \text{epi}(f)$  is closed.

# Lower Semicontinuous PLFs

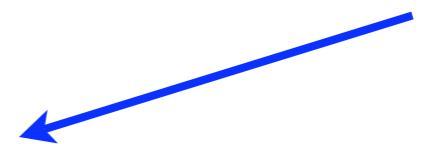


$$f(x) := \begin{cases} 1.5x + 1 & x \in [0, 2) \\ 2 & x \in [2, 2] \\ -1.5x + 6 & x \in (2, 4] \\ 2x - 7 & x \in (4, 5] \end{cases}$$

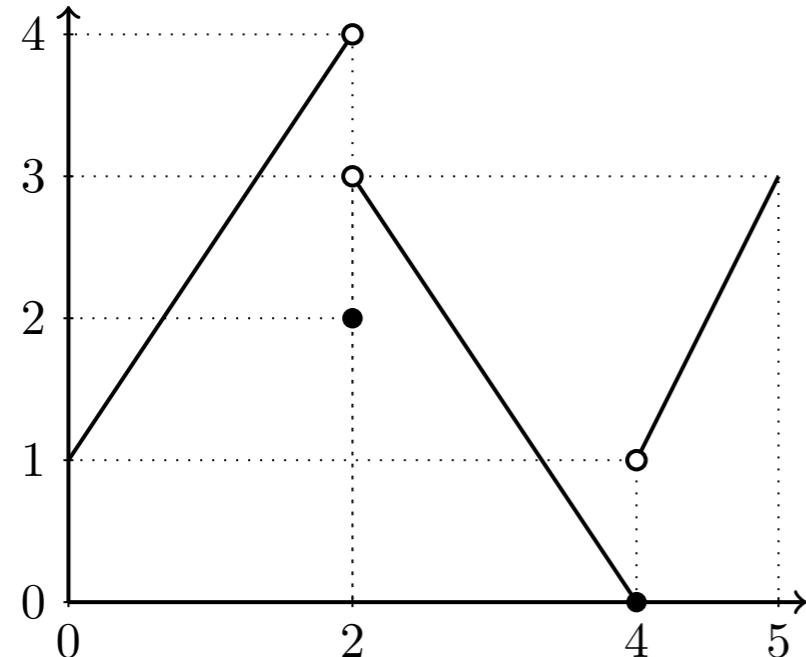
$$f(x) := \begin{cases} m_P x + c_P & x \in P \quad \forall P \in \mathcal{P} \end{cases}$$

Finite family of  
copolytopes

$$P = \{x \in \mathbb{R}^n : a_i x \leq b_i \quad \forall i \in \{1, \dots, p\}, \\ a_i x < b_i \quad \forall i \in \{p, \dots, m\}\}$$



# Lower Semicontinuous PLFs

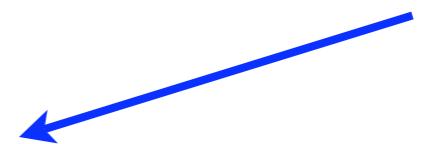


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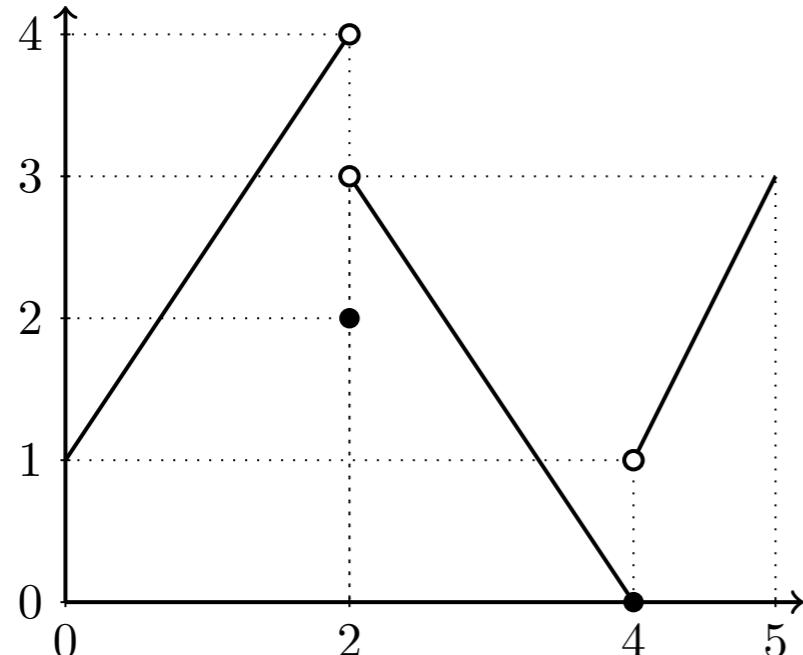
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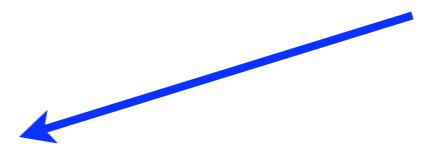


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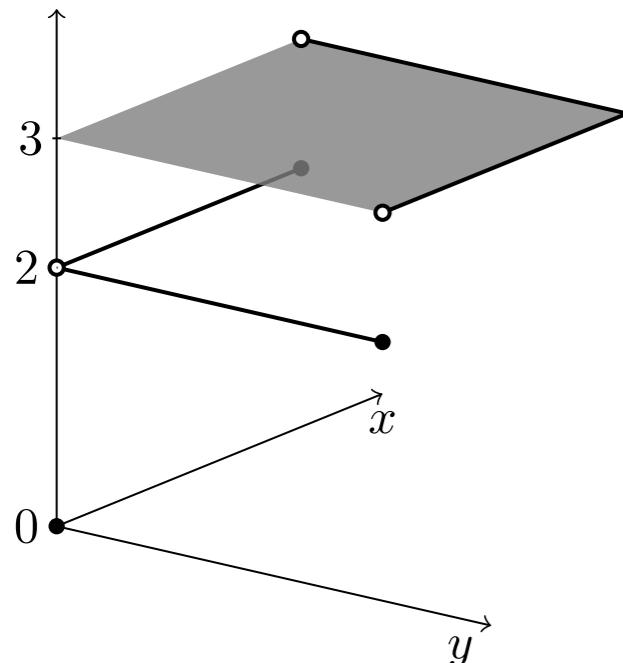
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# Lower Semicontinuous PLFs

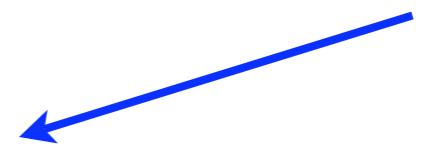


$$f(x, y) := \begin{cases} 3 & (x, y) \in (0, 1]^2 \\ 2 & (x, y) \in \{(x, y) \in \mathbb{R}^2 : x = 0, y > 0\} \\ 2 & (x, y) \in \{(x, y) \in \mathbb{R}^2 : y = 0, x > 0\} \\ 0 & (x, y) \in \{(0, 0)\}. \end{cases}$$

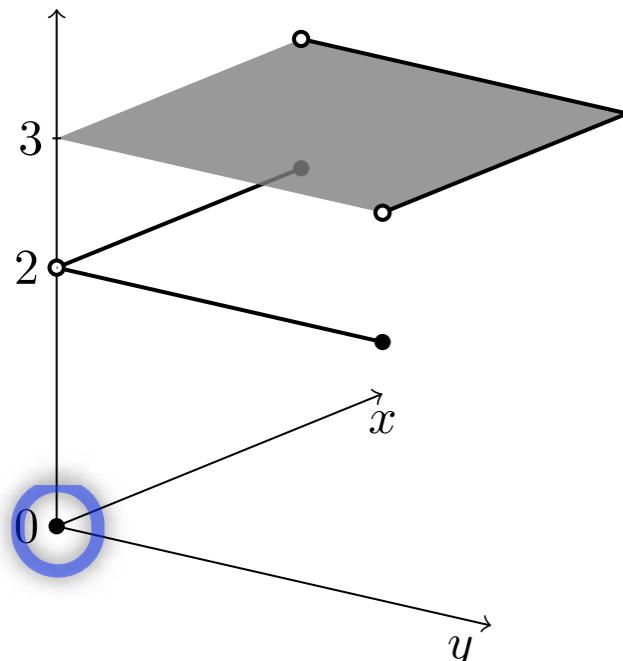
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# Lower Semicontinuous PLFs

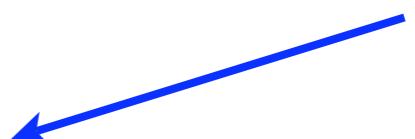


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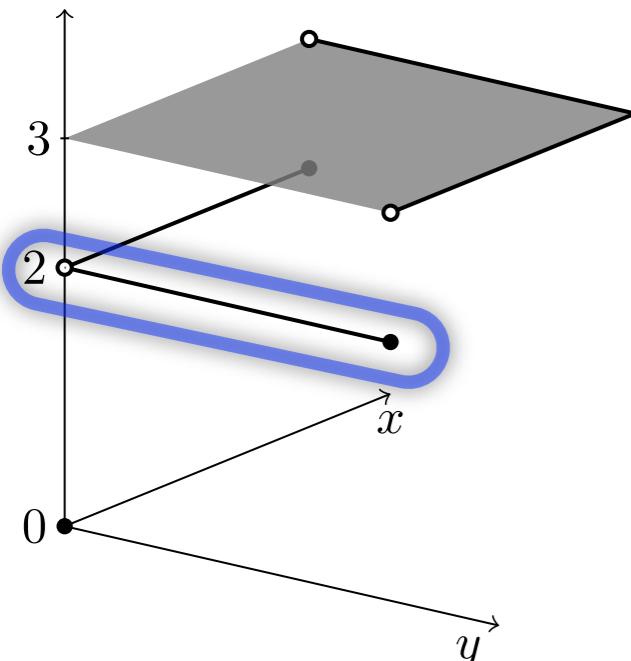
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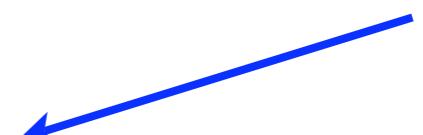


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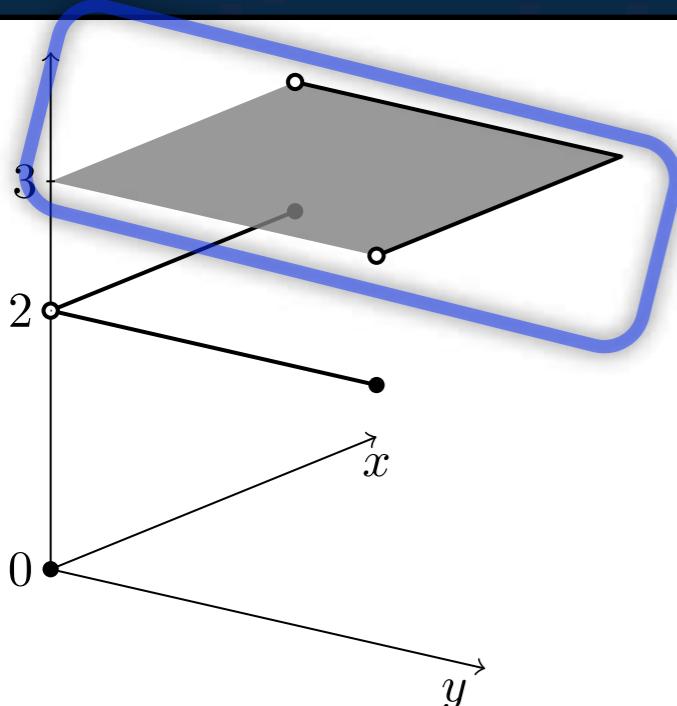
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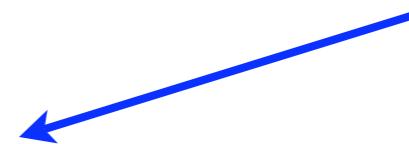


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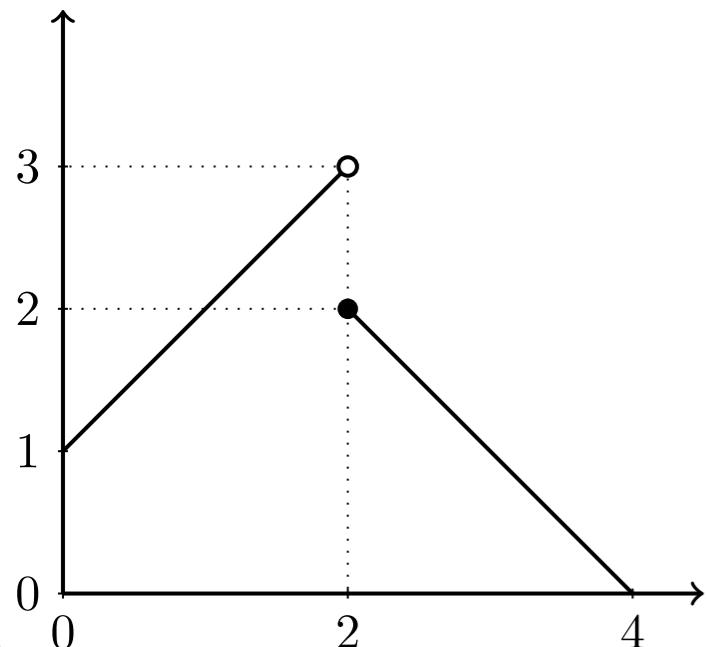
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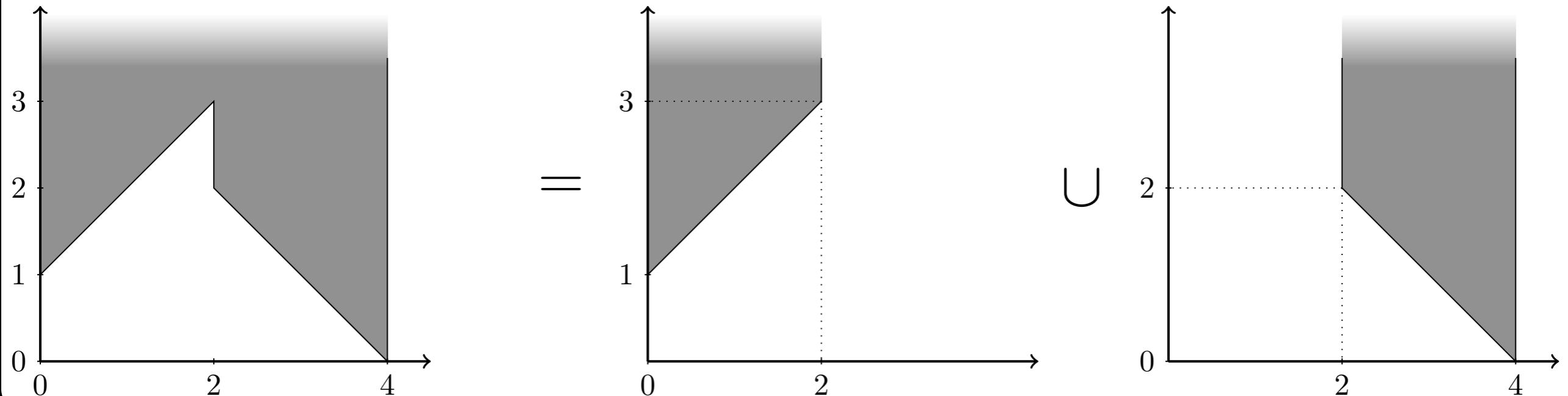
# Disaggregated CC (Jeroslow and Lowe, 1984)



(DCC)

$$f(x) := \begin{cases} x + 1 & x \in [0, 2) \\ 4 - x & x \in [2, 4] \end{cases}$$

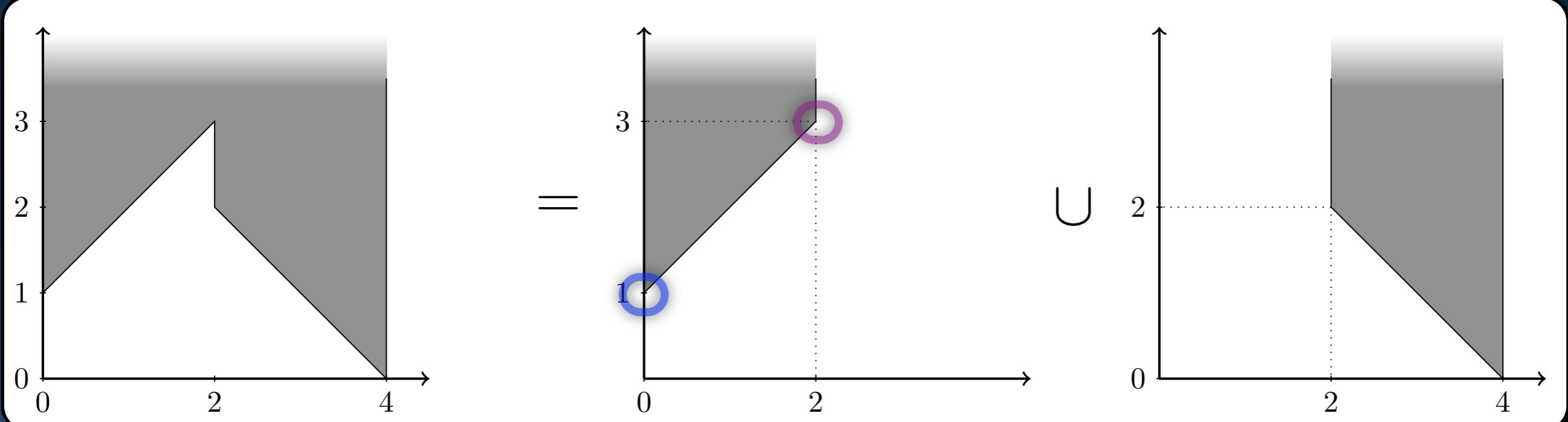
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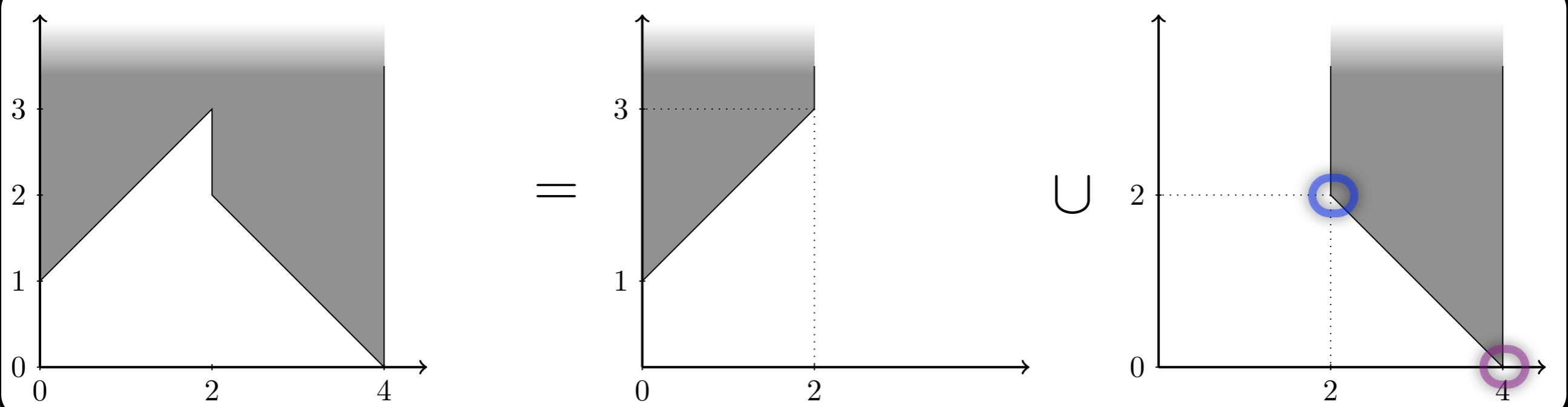
(DCC)

 $P_1$   
 $\downarrow$ 

$$f(x) := \begin{cases} x + 1 & x \in [0, 2) \\ 4 - x & x \in [2, 4] \end{cases}$$

$$\begin{aligned} x &= 0\lambda_{P_1,0} + 2\lambda_{P_1,2} \\ z &\geq 1\lambda_{P_1,0} + 3\lambda_{P_1,2} \\ 1 &= \lambda_{P_1,0} + \lambda_{P_1,2}, \quad \lambda_{P_1,0}, \lambda_{P_1,2} \geq 0 \end{aligned}$$

# Disaggregated CC (Jeroslow and Lowe, 1984)



(DCC)

$$f(x) := \begin{cases} x + 1 & x \in [0, 2) \\ 4 - x & x \in [2, 4] \end{cases}$$

$\uparrow$   
 $P_2$

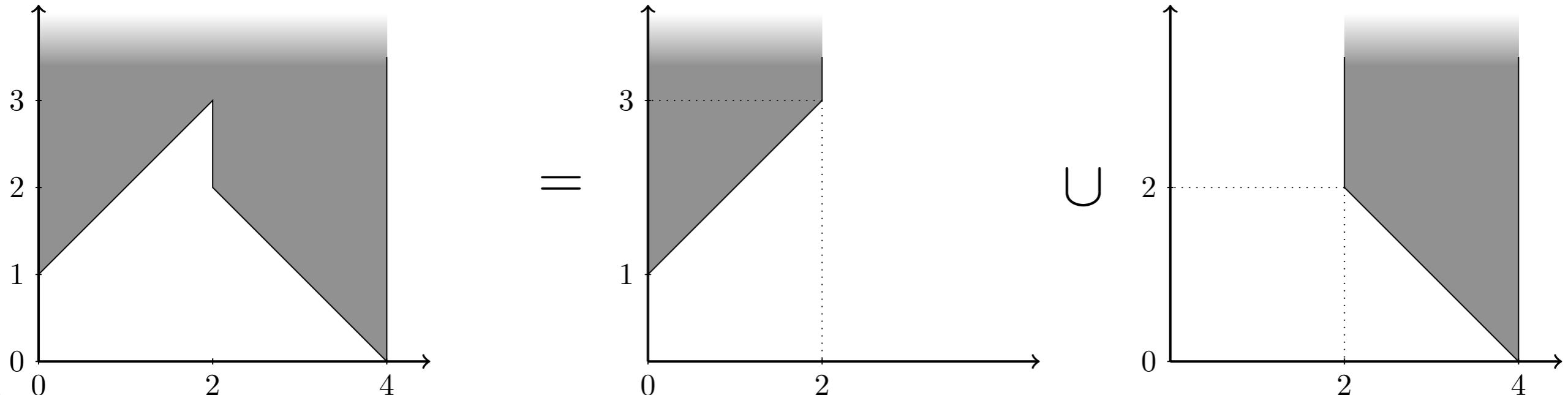
$$x =$$

$$z \geq$$

$$1 = \lambda_{P_2,2} + \lambda_{P_2,4}, \quad \lambda_{P_2,2}, \lambda_{P_2,4} \geq 0$$

$$\begin{aligned} & 2\lambda_{P_2,2} + 4\lambda_{P_2,4} \\ & 2\lambda_{P_2,2} + 0\lambda_{P_2,4} \end{aligned}$$

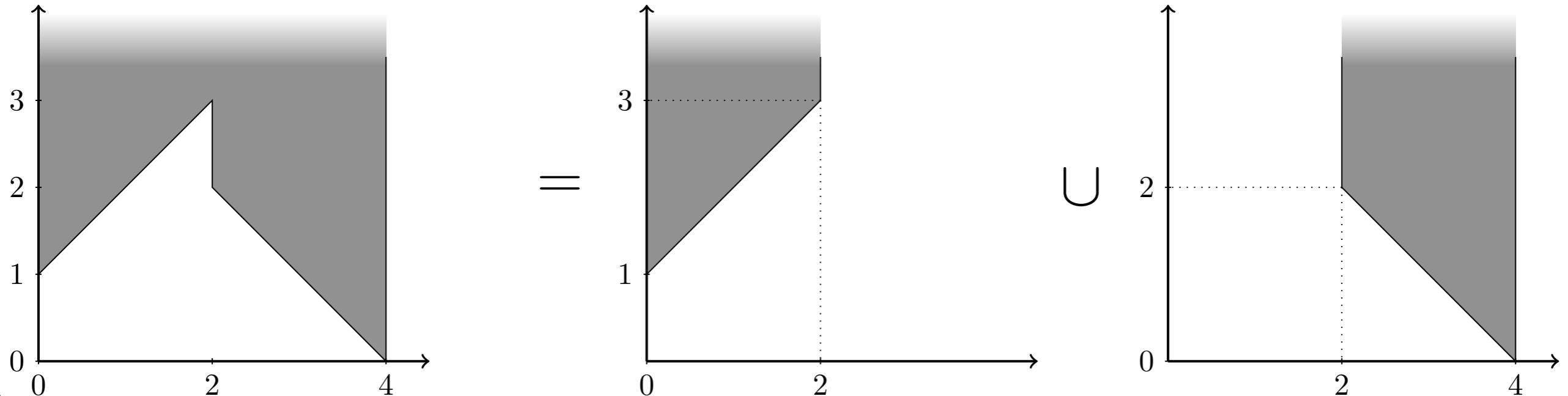
# Disaggregated CC (Jeroslow and Lowe, 1984)



(DCC)

$$\begin{aligned}
 x &= 0\lambda_{P_1,0} + 2\lambda_{P_1,2} + 2\lambda_{P_2,2} + 4\lambda_{P_2,4} \\
 z &\geq 1\lambda_{P_1,0} + 3\lambda_{P_1,2} + 2\lambda_{P_2,2} + 0\lambda_{P_2,4} \\
 f(x) := \begin{cases} x+1 & x \in [0, 2) \\ 4-x & x \in [2, 4] \end{cases} & 1 = \lambda_{P_1,0} + \lambda_{P_1,2}, \quad \lambda_{P_1,0}, \lambda_{P_1,2} \geq 0 \\
 & 1 = \lambda_{P_2,2} + \lambda_{P_2,4}, \quad \lambda_{P_2,2}, \lambda_{P_2,4} \geq 0
 \end{aligned}$$

# Disaggregated CC (Jeroslow and Lowe, 1984)



(DCC)

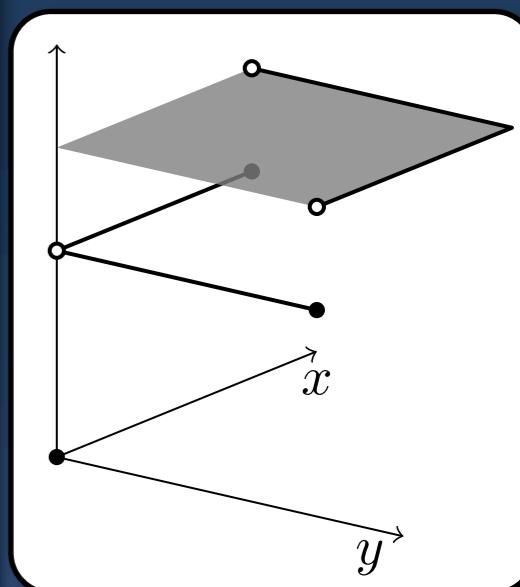
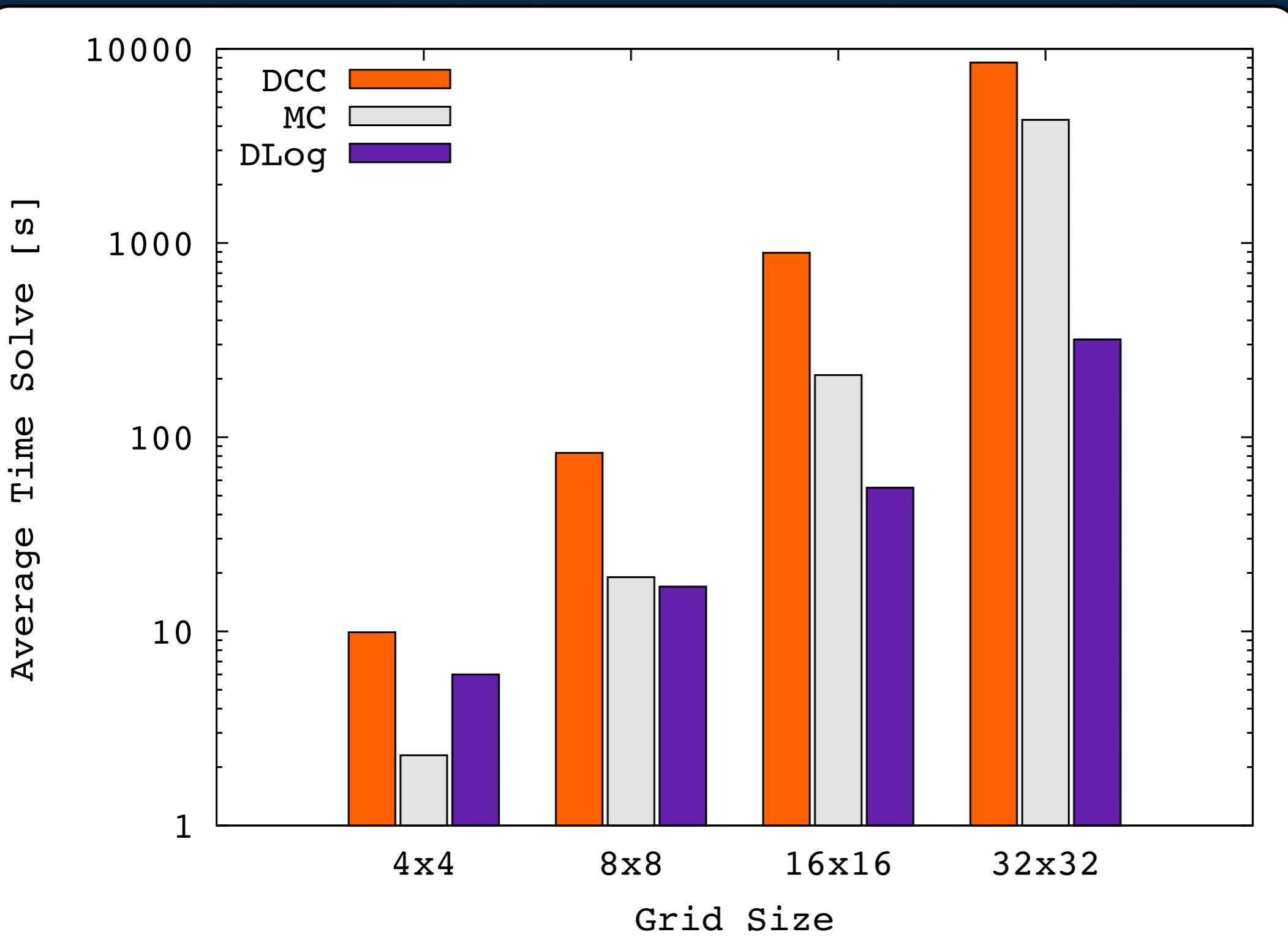
$$x = 0\lambda_{P_1,0} + 2\lambda_{P_1,2} + 2\lambda_{P_2,2} + 4\lambda_{P_2,4}$$

$$z \geq 1\lambda_{P_1,0} + 3\lambda_{P_1,2} + 2\lambda_{P_2,2} + 0\lambda_{P_2,4}$$

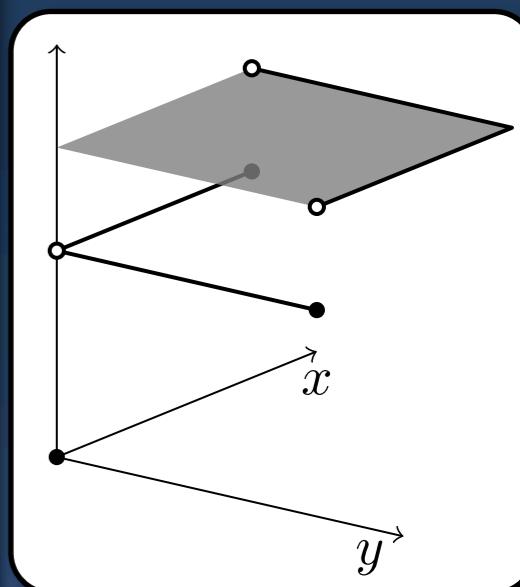
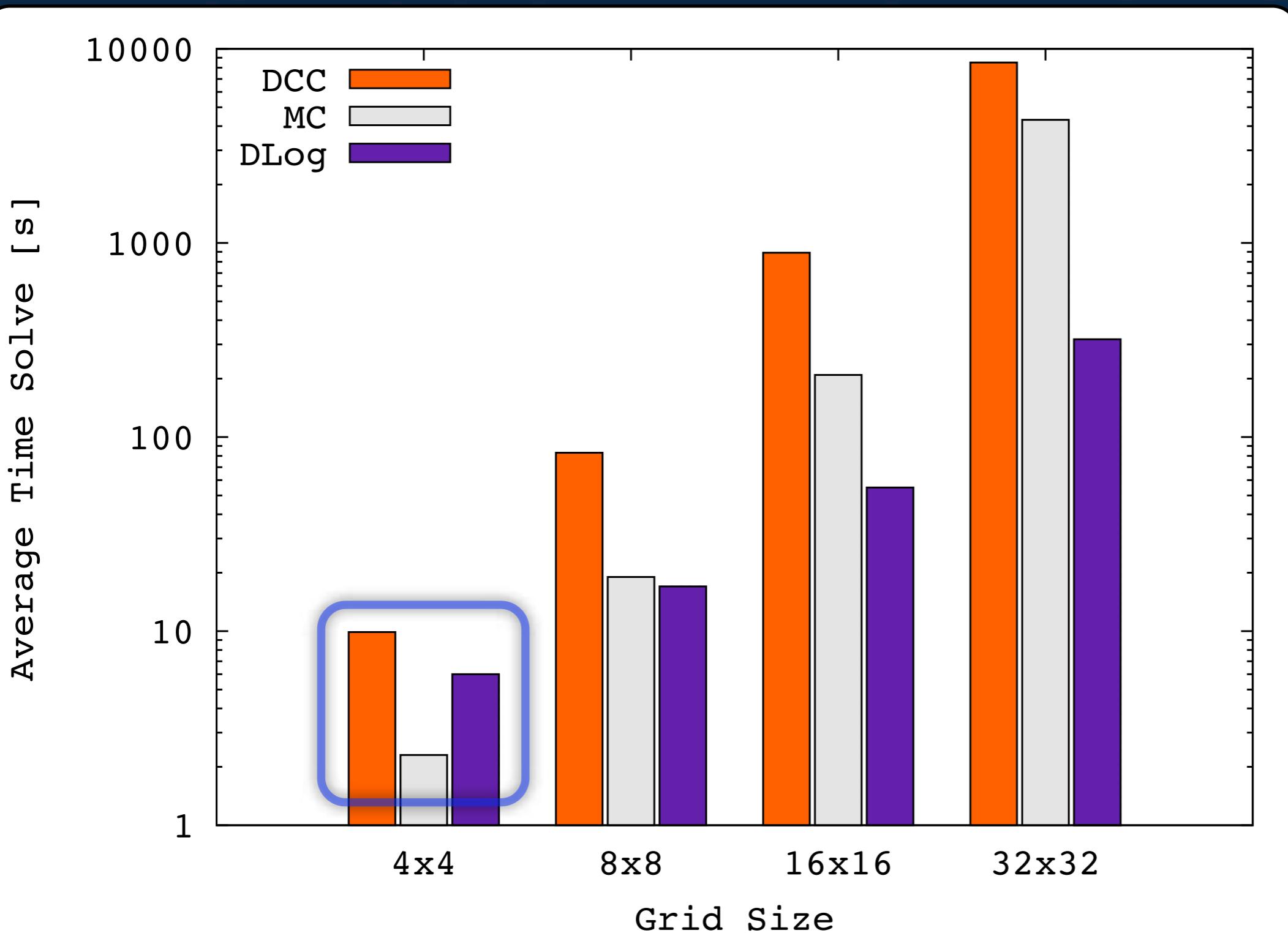
$$f(x) := \begin{cases} x+1 & x \in [0, 2) \\ 4-x & x \in [2, 4] \end{cases} \quad \begin{aligned} y_{P_1} &= \lambda_{P_1,0} + \lambda_{P_1,2}, & \lambda_{P_1,0}, \lambda_{P_1,2} &\geq 0 \\ y_{P_2} &= \lambda_{P_2,2} + \lambda_{P_2,4}, & \lambda_{P_2,2}, \lambda_{P_2,4} &\geq 0 \end{aligned}$$

$$1 = y_{P_1} + y_{P_2}, \quad y_{P_1}, y_{P_2} \in \{0, 1\}$$

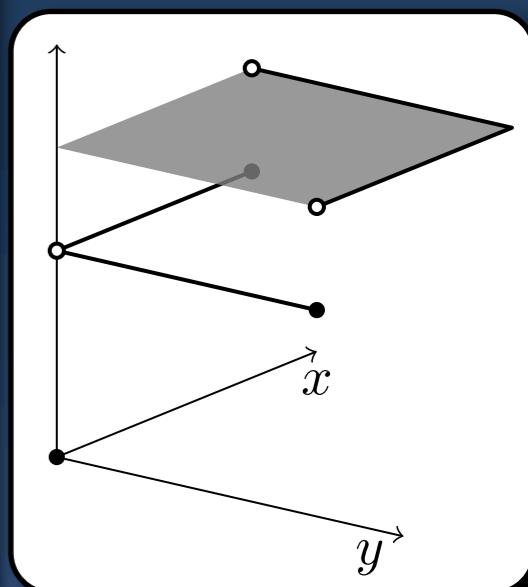
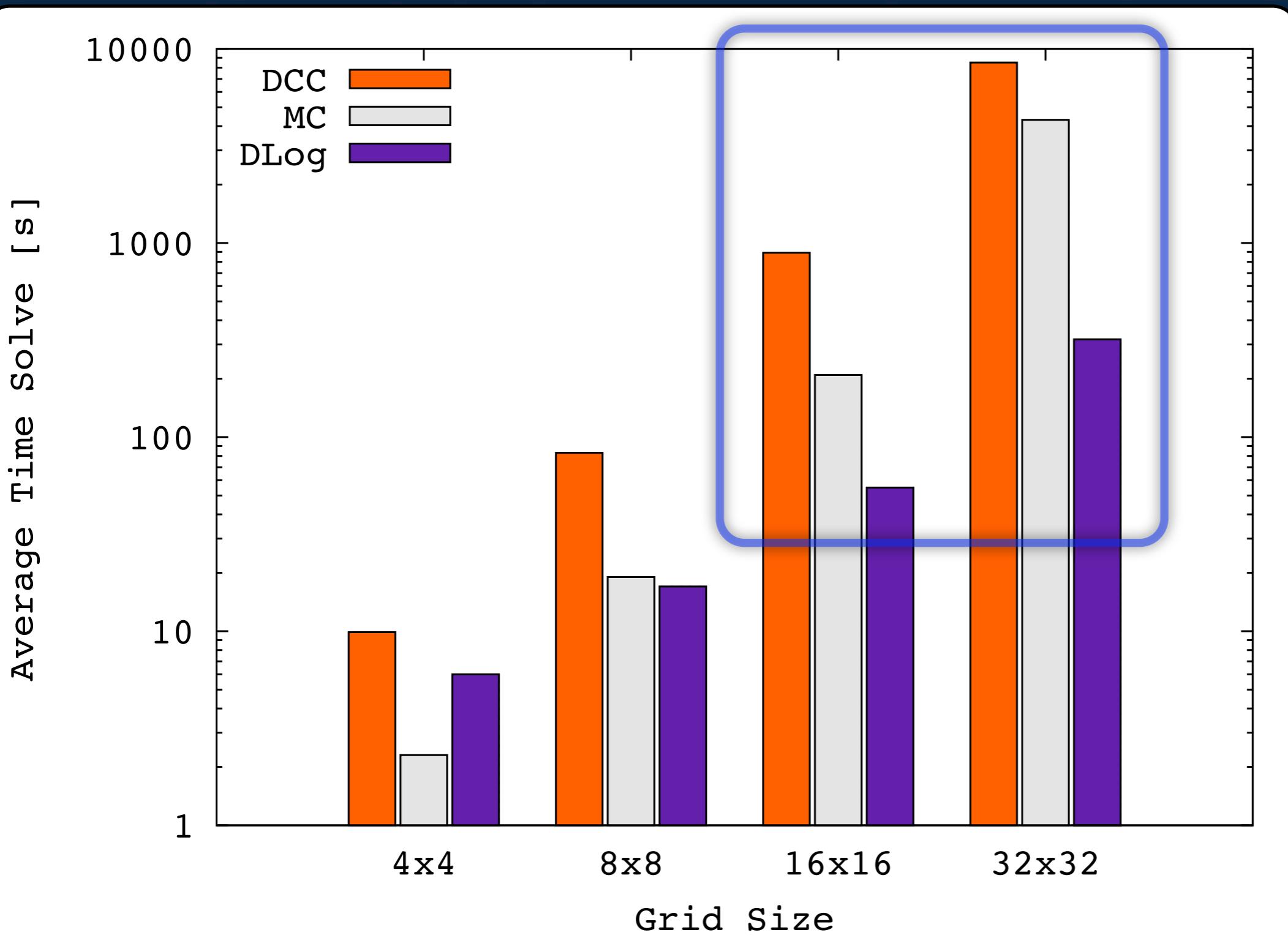
# Multivariate Lower Semicontinuous



# Multivariate Lower Semicontinuous



# Multivariate Lower Semicontinuous

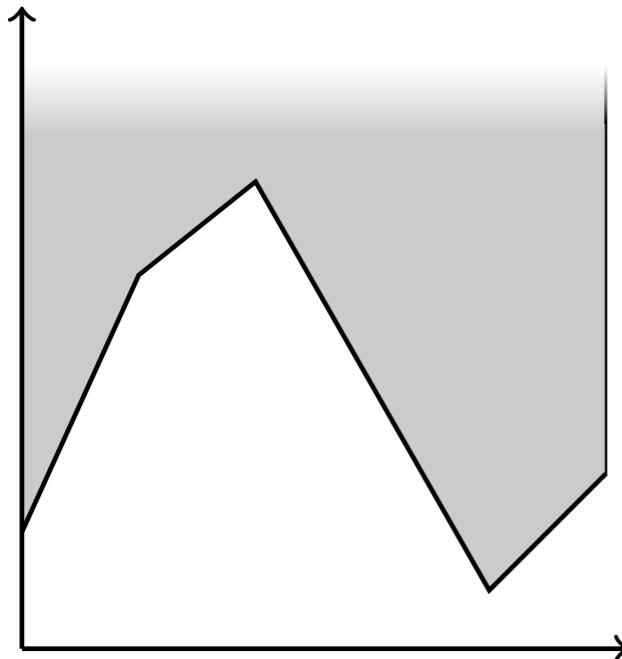


# Final Remarks

- Many MILP models for PLF: Most are simple.
- Popular models are strong.
- Caution: Standard MILP techniques can weak models.
- Best model varies: e.g. Log's best for fine grids.
- Advertisement: Papers at  
<http://www2.isye.gatech.edu/~jvielma>

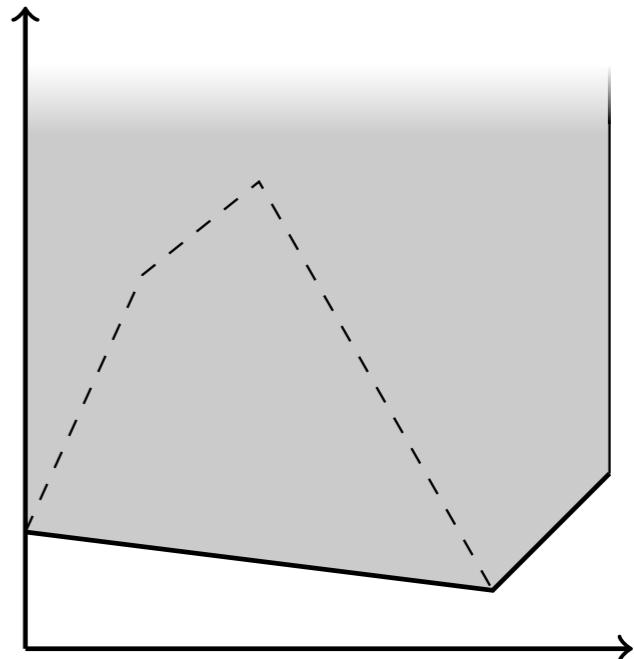
# Strength of LP Relaxations

- Sharp Models:  $\text{LP} = \text{lower convex envelope.}$



(a)  $\text{epi}(f).$

LP relaxation  
→

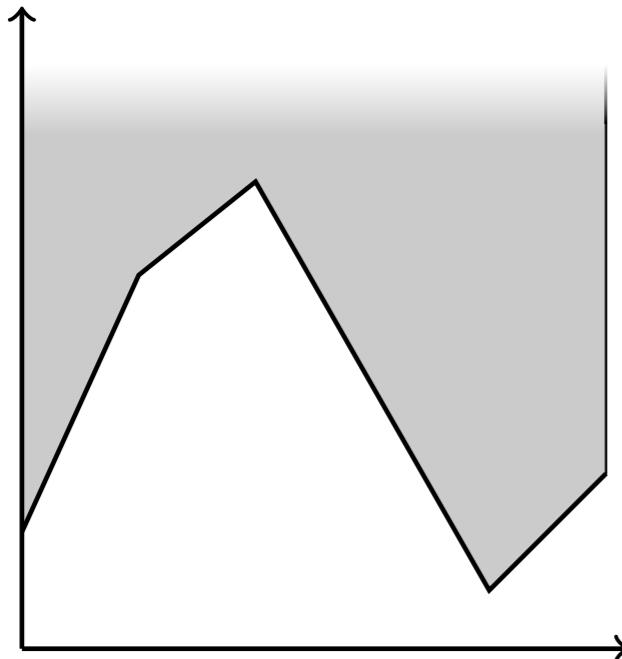
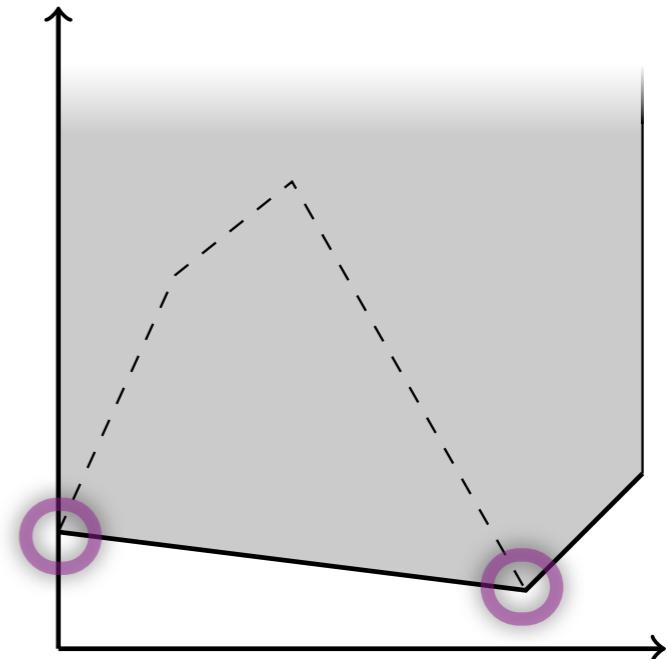


(b)  $\text{conv}(\text{epi}(f)).$

- All popular models are sharp.
- Locally Ideal:  $\text{LP} = \text{Integral}$  (All but CC, even Log).
- Locally ideal implies Sharp.

# Strength of LP Relaxations

- Sharp Models: LP = lower convex envelope.

(a)  $\text{epi}(f)$ .(b)  $\text{conv}(\text{epi}(f))$ .

- All popular models are sharp.
- Locally Ideal: LP = Integral (All but CC, even Log).
- Locally ideal implies Sharp.



# For Multivariate Functions

$$\sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} v = x, \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} (m_P v + c_P) \leq z$$

$$(\text{DCC}) \quad \lambda_{P,v} \geq 0 \quad \forall P \in \mathcal{P}, v \in V(\bar{P}), \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} = 1$$

$$\sum_{v \in V(\bar{P})} \lambda_{P,v} = y_P \quad \forall P \in \mathcal{P}, \quad \sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P}$$

$$\sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v v = x, \quad \sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v (m_P v + c_P) \leq z$$

$$(\text{CC}) \quad \lambda_v \geq 0 \quad \forall v \in \mathcal{V}(\mathcal{P}) := \bigcup_{P \in \mathcal{P}} V(P), \quad \sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v = 1$$

$$\lambda_v \leq \sum_{\{P \in \mathcal{P} : v \in P\}} y_P \quad \forall v \in \mathcal{V}(\mathcal{P}), \quad \sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P}$$



# For Multivariate Functions

$$\sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} v = x, \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} (m_P v + c_P) \leq z$$

$$(DCC) \quad \lambda_{P,v} \geq 0 \quad \forall P \in \mathcal{P}, v \in V(\bar{P}), \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} = 1$$

$$\sum_{v \in V(\bar{P})} \lambda_{P,v} = y_P \quad \forall P \in \mathcal{P}, \quad \sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P}$$

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$$(CC) \quad \lambda_v \geq 0 \quad \forall v \in \mathcal{V}(\mathcal{P}) := \bigcup_{P \in \mathcal{P}} V(P), \quad \sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v = 1$$

$$\lambda_v \leq \sum_{\{P \in \mathcal{P} : v \in P\}} y_P \quad \forall v \in \mathcal{V}(\mathcal{P}), \quad \sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P}$$



# For Multivariate Functions

$$\sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} v = x, \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} (m_P v + c_P) \leq z$$

$$(DCC) \quad \lambda_{P,v} \geq 0 \quad \forall P \in \mathcal{P}, v \in V(\bar{P}), \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(\bar{P})} \lambda_{P,v} = 1$$

$$\boxed{\sum_{v \in V(\bar{P})} \lambda_{P,v} = y_P \quad \forall P \in \mathcal{P}, \quad \sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P}}$$

$$\sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v v = x, \quad \sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v (m_P v + c_P) \leq z$$

$$(CC) \quad \lambda_v \geq 0 \quad \forall v \in \mathcal{V}(\mathcal{P}) := \bigcup_{P \in \mathcal{P}} V(P), \quad \sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v = 1$$

$$\boxed{\lambda_v \leq \sum_{\{P \in \mathcal{P} : v \in P\}} y_P \quad \forall v \in \mathcal{V}(\mathcal{P}), \quad \sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P}}$$



# Logarithmic DCC (DLog)

$$\sum_{P \in \mathcal{P}} \sum_{v \in V(P)} \lambda_{P,v} v = x,$$

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$$\lambda_{P,v} \geq 0 \quad \forall P \in \mathcal{P}, v \in V(P), \quad \sum_{P \in \mathcal{P}} \sum_{v \in V(P)} \lambda_{P,v} = 1$$

$$\sum_{P \in \mathcal{P}^+(B,l)} \sum_{v \in V(P)} \lambda_{P,v} \leq y_l, \quad \sum_{P \in \mathcal{P}^0(B,l)} \sum_{v \in V(P)} \lambda_{P,v} \leq (1 - y_l), \quad y_l \in \{0, 1\} \quad \forall l \in L(\mathcal{P})$$

where  $B : \mathcal{P} \rightarrow \{0, 1\}^{\lceil \log_2 |\mathcal{P}| \rceil}$  is any injective function,  $L(\mathcal{P}) := \{1, \dots, \lceil \log_2 |\mathcal{P}| \rceil\}$ ,

$\mathcal{P}^+(B,l) := \{P \in \mathcal{P} : B(P)_l = 1\}$  and  $\mathcal{P}^0(B,l) := \{P \in \mathcal{P} : B(P)_l = 0\}$ .

- New? Direct from ideas in Ibaraki (1976), Vielma and Nemhauser (2008)



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# Logarithmic Conv. Comb. (Log)

$$\sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v v = x,$$

$$\sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v (m_P v + c_P) \leq z$$

$$\lambda_v \geq 0 \quad \forall v \in \mathcal{V}(\mathcal{P}),$$

$$\sum_{v \in \mathcal{V}(\mathcal{P})} \lambda_v = 1$$

$$\sum_{v \in L_s} \lambda_v \leq y_s,$$

$$\sum_{v \in R_s} \lambda_v \leq (1 - y_s),$$

$$y_s \in \{0, 1\} \quad \forall s \in \mathcal{S}.$$

- Requires Independent Branching Scheme.
- Vielma and Nemhauser (2008).



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# Multiple Choice (MC)

$$\sum_{P \in \mathcal{P}} x^P = x, \quad \sum_{P \in \mathcal{P}} (m_P x^P + c_P y_P) \leq z$$
$$A_P x^P \leq y_P b_P \quad \forall P \in \mathcal{P}$$

$$\sum_{P \in \mathcal{P}} y_P = 1, \quad y_P \in \{0, 1\} \quad \forall P \in \mathcal{P},$$

where  $A_P x \leq b_P$  is the set of linear inequalities describing  $P$ .

- Balakrishnan and Graves (1989), Croxton et al. (2003a), Jeroslow and Lowe (1984) and Lowe (1984)



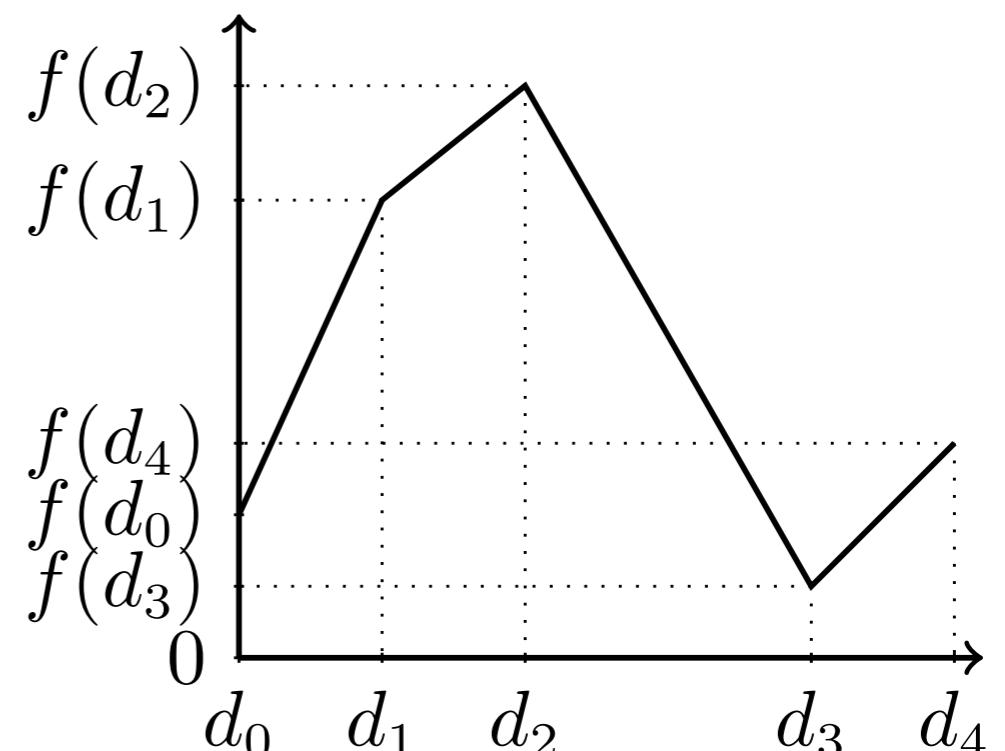
# Incremental or Delta (Inc)

$$d_0 + \sum_{k=1}^K \delta_k (d_k - d_{k-1}) = x$$

$$f(d_0) + \sum_{k=1}^K \delta_k (f(d_k) - f(d_{k-1})) \leq z$$

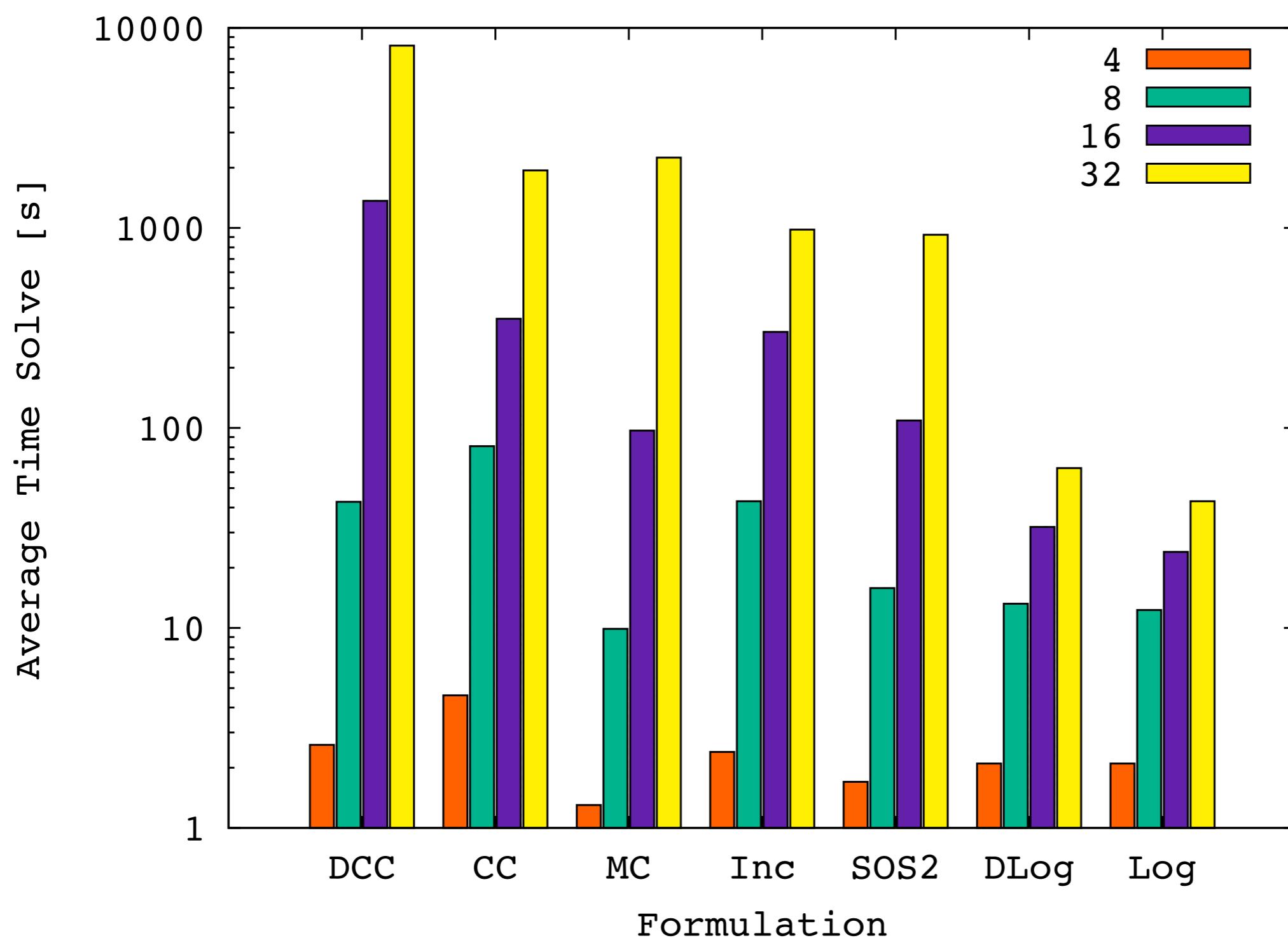
$$\delta_1 \leq 1, \quad \delta_K \geq 0, \quad \delta_{k+1} \leq y_k \leq \delta_k,$$

$$y_k \in \{0, 1\} \quad \forall k \in \{1, \dots, K-1\}.$$

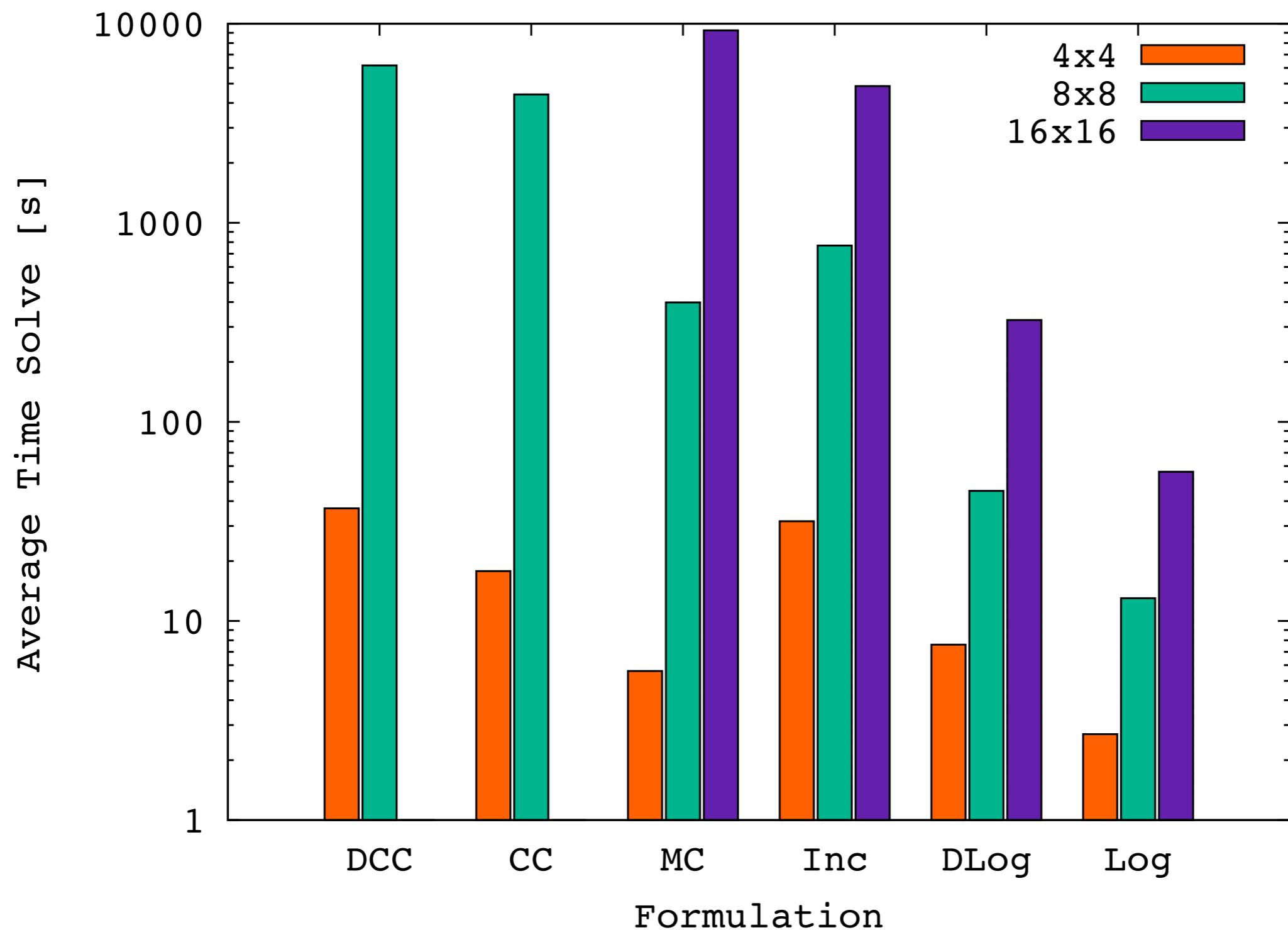


- Similar for multivariate functions.
- Croxton et al. (2003a), Dantzig (1963, 1960), Keha et al. (2004), Markowitz and Manne (1957), Padberg (2000), Sherali (2001), Vajda (1964) and Wilson (1998).

# Univariate Case (Separable)



# Multivariate Case (Non-Separable)



# Multivariate Lower Semicontinuous

