

# Convex approximations for multistage stochastic mixed-integer programs

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Joint work with Jinting Lin, Niels van der Laan, and Ruben van Beesten

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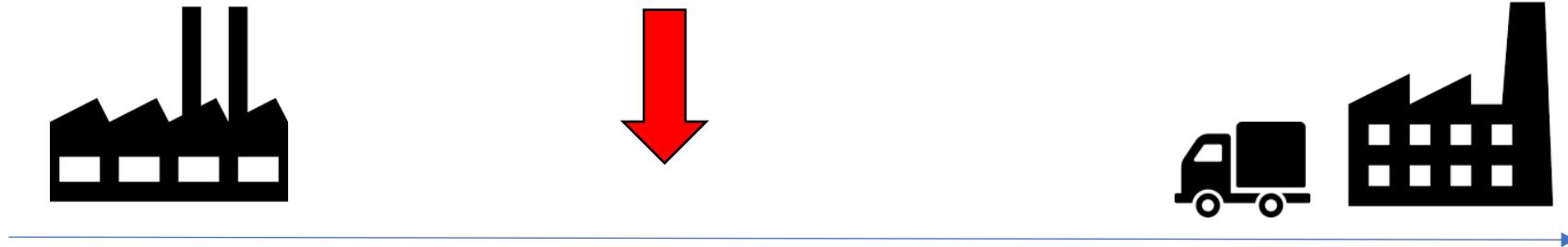
June 3, 2025

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# Production planning example

- Timeline:

Observe demand  $\omega$



Produce  $x \in \mathbb{R}_+$

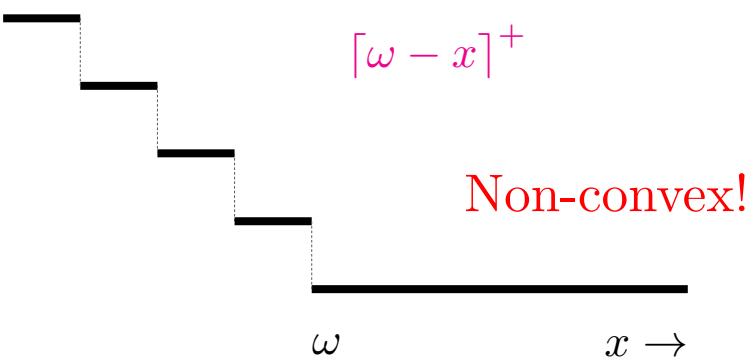
If  $\omega > x$ , then buy  $y \in \mathbb{Z}_+$   
from competitor in batches of size 1

- Minimize total expected costs:

$$\min_{x \geq 0} cx + q \mathbb{E}_{\omega} [\lceil \omega - x \rceil^+]$$

↑  
production costs       $y = \max\{\lceil \omega - x \rceil, 0\}$   
expected future purchasing costs

- For fixed demand  $\omega$ :



# Non-convex objective function?

- Minimize total **expected** costs:

$$\min_{x \geq 0} cx + q \mathbb{E}_{\omega} \left[ [\omega - x]^+ \right]$$

$\underbrace{\hspace{10em}}$   
 $Q(x)$

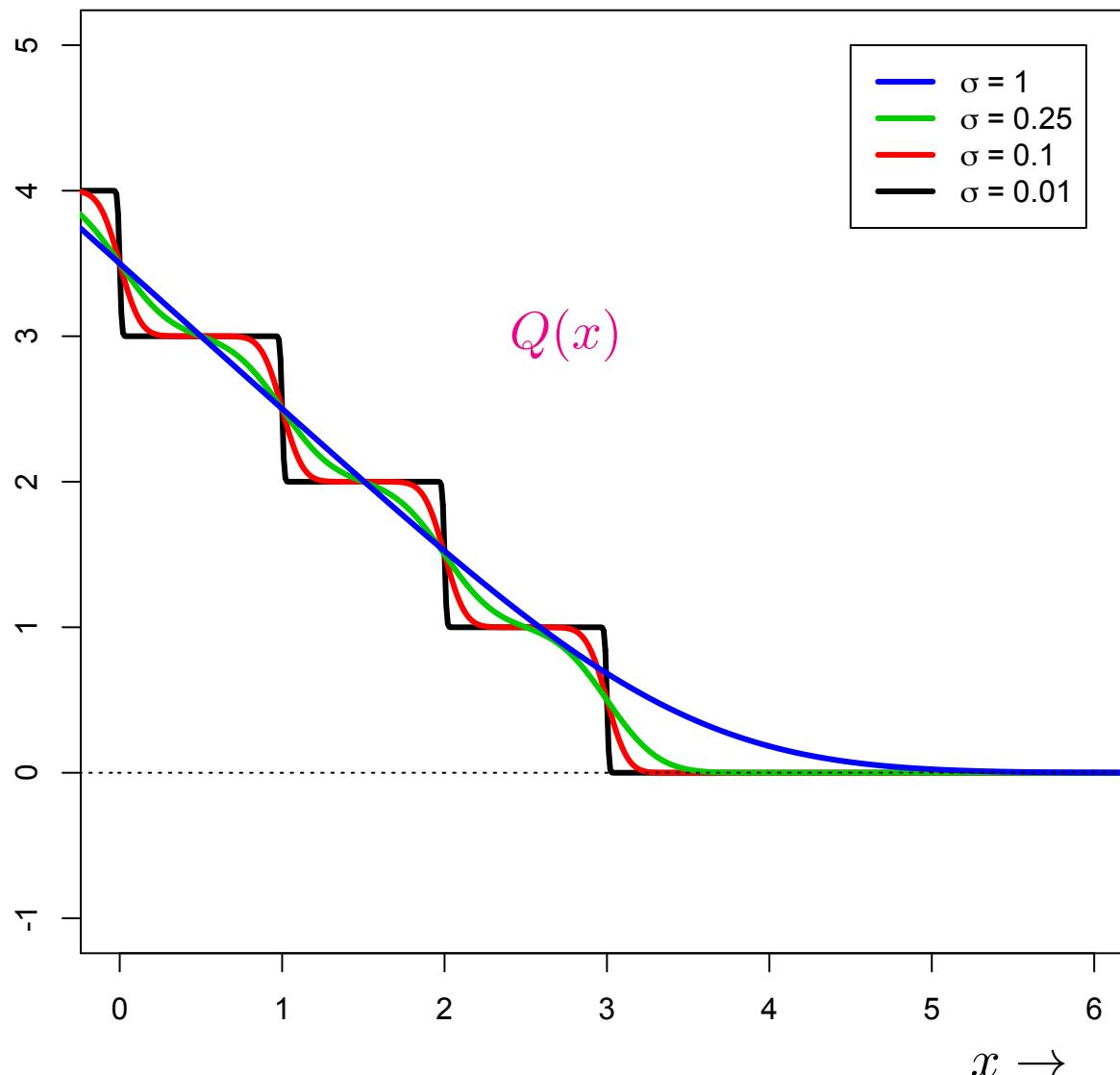
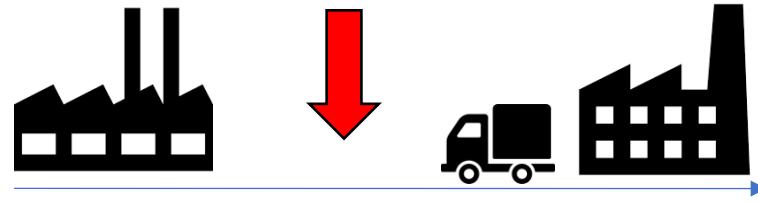
- Numerical example:

$$\omega \sim N(\mu, \sigma^2)$$

$$\mu = 3$$

$$q = 1$$

Observation:  $Q$  “convexer”  
if  $\sigma$  increases



# Bounds on the expectation of periodic functions

- Convex approximation  $\hat{Q}$

$$\hat{Q}(x) = \mathbb{E}_\omega \left[ (\omega + 1/2 - x)^+ \right]$$

- Approximation error

$$\hat{Q}(x) - Q(x) = \mathbb{E}_\omega [\varphi_x(\omega)]$$

- Error bound

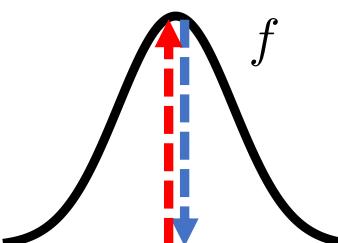
For any  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  and random variable  $\omega$  with pdf  $f$

$$\left| \mathbb{E}_\omega [\varphi(\omega)] \right| \leq \frac{|\Delta|f}{2} \sup_{s \leq t} \left| \int_s^t \varphi(u) du \right|$$

with  $|\Delta|f$  denoting the total variation of  $f$

- Total variation

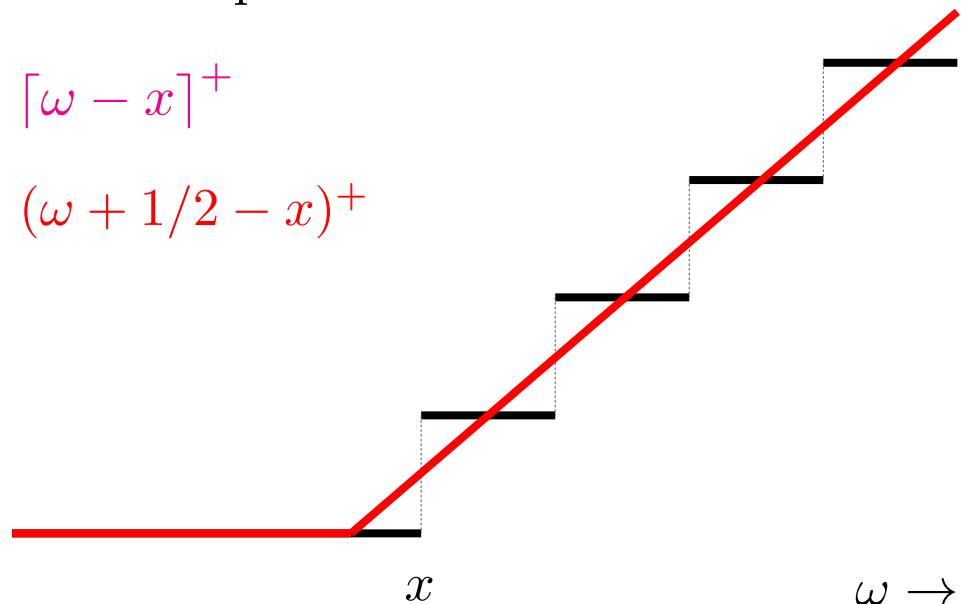
$$|\Delta|f = \text{total increase} + \text{total decrease}$$



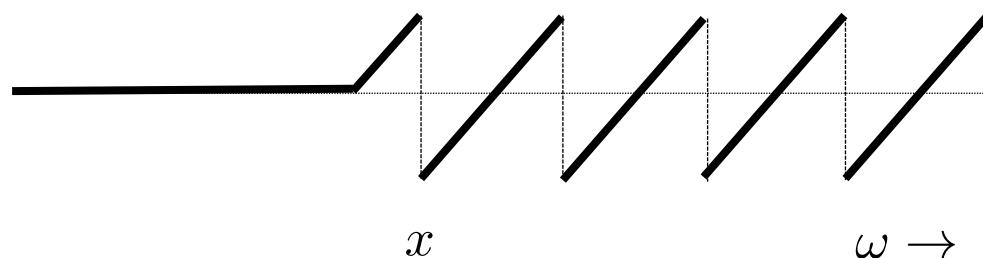
- For a fixed production level  $x$ :

$$[\omega - x]^+$$

$$(\omega + 1/2 - x)^+$$



- Periodic difference function  $\varphi_x(\omega)$



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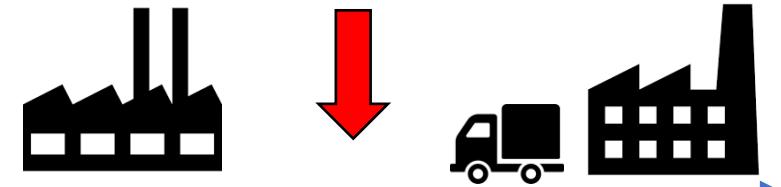
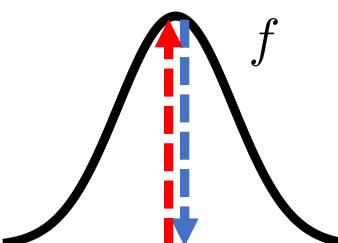
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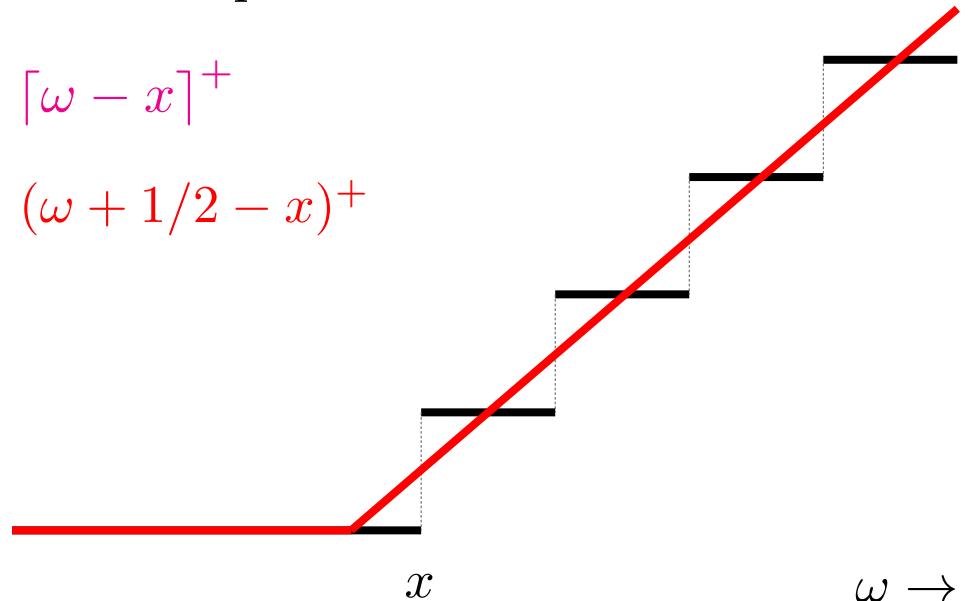
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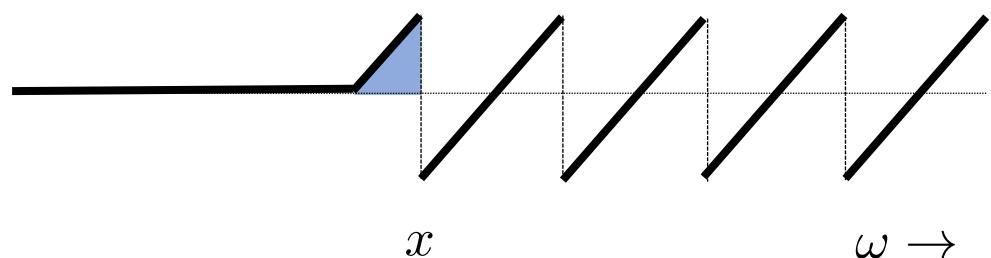
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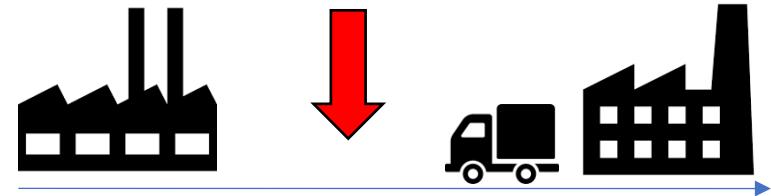
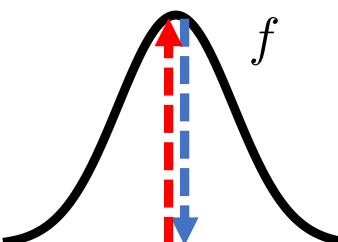
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- Total variation

$$|\Delta|f = \text{total increase} + \text{total decrease}$$



## Goal

- Production planning example is a stochastic MIP
- Prove asymptotic convexity for multistage stochastic MIPs

# Convex approximations for multistage stochastic mixed-integer programs

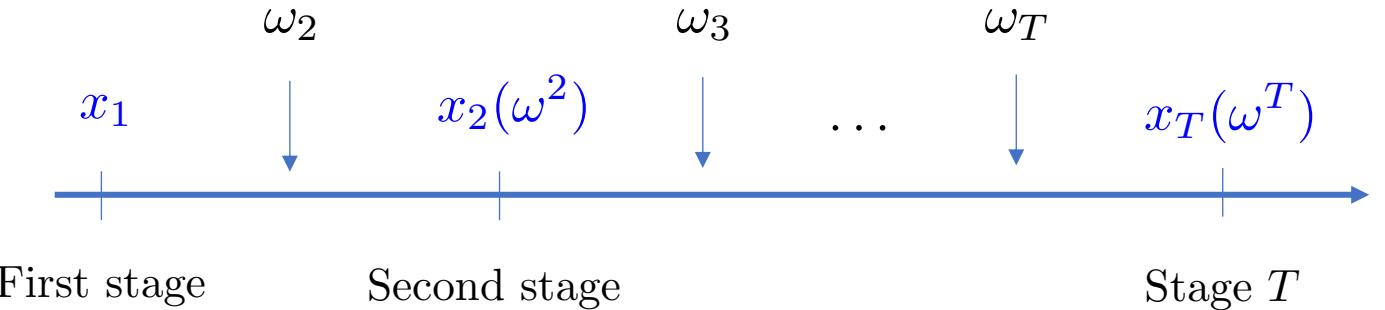
## Outline

- Definition multistage stochastic mixed-integer programs (M-SMIPs)
  - Nested formulation
  - Known results for two-stage SMIPs
- Asymptotic periodicity for mixed-integer value functions
- Construction convex approximations for M-SMIPs
- Asymptotic total variation error bound

# Multistage stochastic mixed-integer programs (M-SMIPs)

- Notation:

- Time horizon  $t = 1, \dots, T$
- Decision variables  $x_t(\omega^t)$
- Random parameters  $\omega_t$   
with history  $\omega^t := (\omega_1, \dots, \omega_t)$



- Nested formulation:

$$\eta := \min_{x_1 \in X_1} \left\{ c_1^\top x_1 + \underbrace{Q_1(x_1)}_{\substack{\downarrow \\ Q_t(x_t) = \mathbb{E}_{\omega_{t+1}} \left[ v_{t+1}(\omega_{t+1}, x_t) \right], \\ t = 1, \dots, T-1 \quad \text{with } Q_T \equiv 0}} : W_1 x_1 = \omega_1 \right\}$$

Fixed

↑

$v_t(\omega_t, x_{t-1}) = \min_{x_t \in X_t} \left\{ c_t^\top x_t + Q_t(x_t) : T_t x_{t-1} + W_t x_t = \omega_t \right\}$

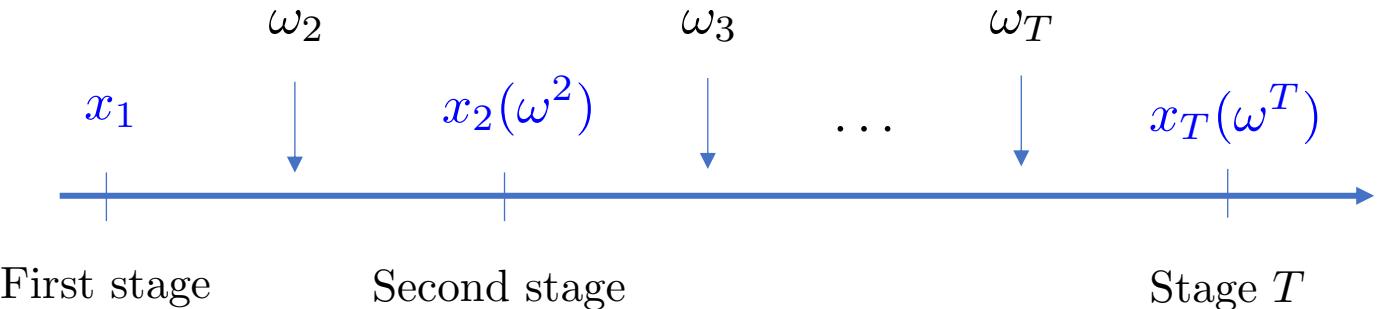
↓

Integrality restrictions

# Multistage stochastic mixed-integer programs (M-SMIPs)

- Assumptions:

- Sufficiently expensive and complete recourse  
→  $v_t$  is always finite
- Stagewise independent  $\omega_t$
- $\mathbb{E}_{\omega_t} \|\omega_t\| < +\infty \rightarrow Q_t$  is finite
- The matrix of coefficients  $W_t$  is integer



- Nested formulation:

$$\eta := \min_{x_1 \in X_1} \left\{ c_1^\top x_1 + \underbrace{Q_1(x_1)}_{\text{Fixed}} : W_1 x_1 = \omega_1 \right\}$$

↑

$$Q_t(x_t) = \mathbb{E}_{\omega_{t+1}} \left[ v_{t+1}(\omega_{t+1}, x_t) \right], \quad t = 1, \dots, T-1 \quad \text{with } Q_T \equiv 0$$

↓

$$v_t(\omega_t, x_{t-1}) = \min_{x_t \in X_t} \left\{ c_t^\top x_t + Q_t(x_t) : T_t x_{t-1} + W_t x_t = \omega_t \right\}$$

↓

Integrality restrictions

## Two-stage stochastic MIPs

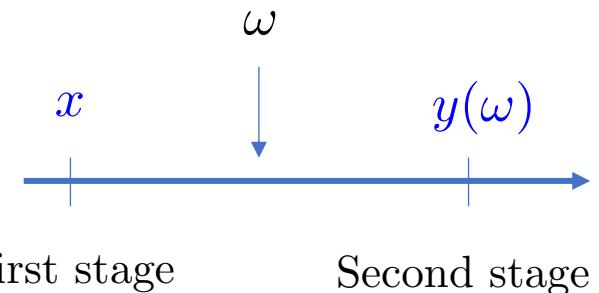
- Stage- $t$  value function

$$v_t(\omega_t, x_{t-1}) = \min_{x_t \in X_t} \left\{ c_t^\top x_t + Q_t(x_t) : T_t x_{t-1} + W_t x_t = \omega_t \right\}$$

- Second-stage value function:

$$v(\omega, x) = \min_y \left\{ q^\top y : W y = \omega - T x, y \in \underbrace{\mathbb{Z}_+^{n_2} \times \mathbb{R}_+^{n_3}}_{\text{2nd-stage feasible region}} \right\}$$

↑                      ↑  
 1<sup>st</sup>-stage variables    2<sup>nd</sup>-stage variables



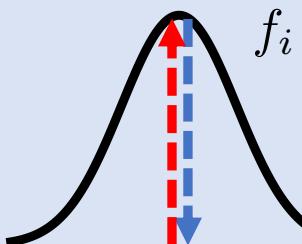
- First-stage expected value function:

$$Q(x) = \mathbb{E}_\omega[v(\omega, x)]$$

- Theorem (R., Schultz, van der Vlerk and Klein Haneveld, 2016)

- There exists a convex approximation  $\hat{Q}(x) = \mathbb{E}_\omega[\hat{v}(\omega, x)]$  of  $Q(x)$  and a constant  $C$  such that for all independent random vectors  $\omega$  with marginal density functions  $f_i$

$$\|Q - \hat{Q}\|_\infty \leq C \sum_{i=1}^m |\Delta| f_i$$



## Generic continuous value function

- Second-stage value function:

$$v(\omega, x) = \min_y \left\{ q^\top y : Wy = \omega - Tx, y \in \mathbb{Z}_+^{n_2} \times \mathbb{R}_+^{n_3} \right\}$$

- LP-relaxation:

$$v_{LP}(\omega, x) = \min_y \left\{ q^\top y : Wy = \omega - Tx, y \geq 0 \right\}$$

- For any basis matrix  $B$ :

$$v_{LP}(\omega, x) = \min_{y_B, y_N} \left\{ q_B^\top y_B + q_N^\top y_N : By_B + Ny_N = \omega - Tx, y_B \geq 0, y_N \geq 0 \right\}$$

- Substituting  $y_B = B^{-1}(\omega - Tx) - B^{-1}Ny_N$ :

$$\begin{aligned} v_{LP}(\omega, x) &= q_B^\top B^{-1}(\omega - Tx) + \min_{y_N} \bar{q}_N^\top y_N \\ \text{s.t. } &B^{-1}(\omega - Tx) - B^{-1}Ny_N \geq 0 \\ &y_N \geq 0 \end{aligned}$$

- with reduced costs  $\bar{q}_N^\top := q_B^\top - q_N^\top B^{-1}N \geq 0$

# Generic continuous value function

$$v_{LP}(\omega, x) = q_B^\top B^{-1}(\omega - Tx) + \min_{y_N} \bar{q}_N^\top y_N$$

s.t.     $B^{-1}(\omega - Tx) - B^{-1}Ny_N \geq 0$

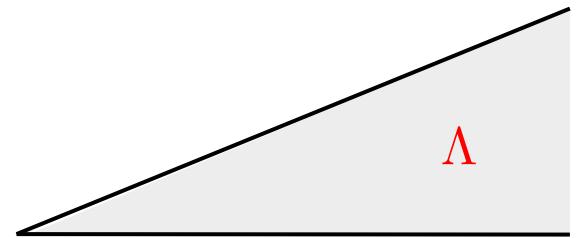
$y_N \geq 0$

with reduced costs  $\bar{q}_N^\top := q_B^\top - q_N^\top B^{-1} N \geq 0$

- Observation:

$$y_N^*(\omega, x) = 0 \text{ if } \underbrace{B^{-1}(\omega - Tx) \geq 0}_{\omega - Tx \in \Lambda := \left\{ s \in \mathbb{R}^m : B^{-1}s \geq 0 \right\}}$$

↑  
closed convex cone



- Holds for all basis matrices  $B$  with reduced costs  $\bar{q}_N^\top \geq 0$

→ Basis Decomposition Theorem

## Basis decomposition theorem

$$\begin{aligned}
 v_{LP}(\omega, x) &= q_B^\top B^{-1}(\omega - Tx) + \min_{y_N} \bar{q}_N^\top y_N \\
 \text{s.t. } &B^{-1}(\omega - Tx) - B^{-1}Ny_N \geq 0 \\
 &y_N \geq 0
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with reduced costs  $\bar{q}_N^\top := q_B^\top - q_N^\top B^{-1}N \geq 0$

- Walkup and Wets (1969)

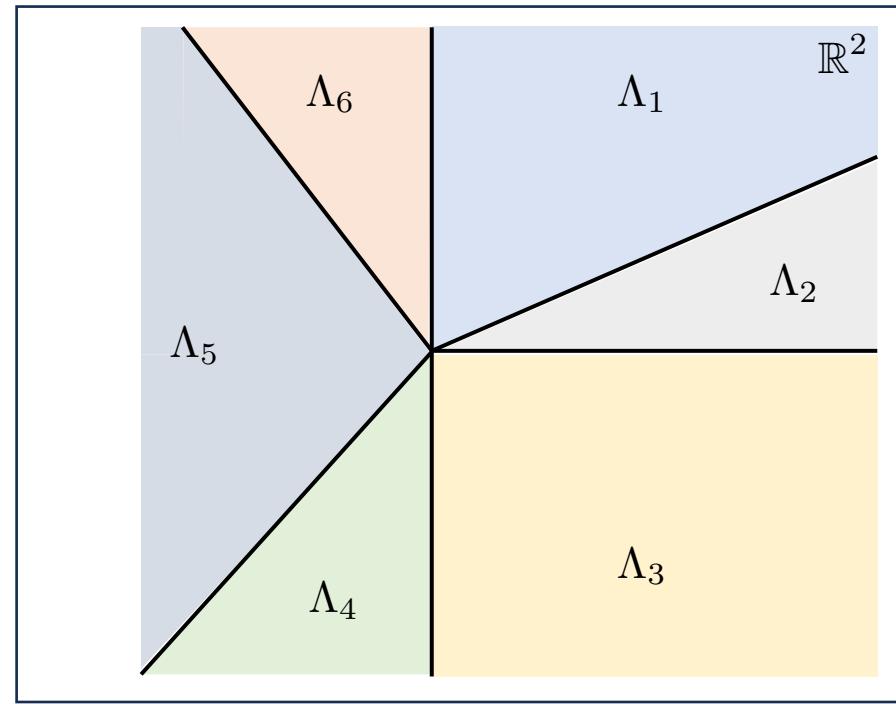
- There exist

- dual feasible **basis matrices**  $B^k$
- corresponding closed convex **cones**  $\Lambda_k$

$$\text{with } \bigcup_{k=1}^K \Lambda_k = \mathbb{R}^m \quad \text{and } \text{int}(\Lambda_i) \cap \text{int}(\Lambda_j) = \emptyset$$

- such that

$$v_{LP}(\omega, x) = \underbrace{q_{B^k}^\top (B^k)^{-1}(\omega - Tx)}_{\text{affine}} \quad \text{if } \omega - Tx \in \Lambda_k$$



## Mixed-integer value function

$$v(\omega, x) = q_B^\top B^{-1}(\omega - Tx) + \min_{y_N} \bar{q}_N^\top y_N$$

s.t.  $B^{-1}(\omega - Tx) - B^{-1}Ny_N \in \mathbb{Z}_+^{n_B} \times \mathbb{R}_+^{m-n_B}$

$$y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}$$

with reduced costs  $\bar{q}_N^\top := q_B^\top - q_N^\top B^{-1}N \geq 0$

- Gomory (1969) and R. et al. (2016)

- There exist

- dual feasible **basis matrices**  $B^k$
- corresponding closed convex **cones**  $\Lambda_k$
- distances**  $d_k$
- periodic functions**  $\psi_k$

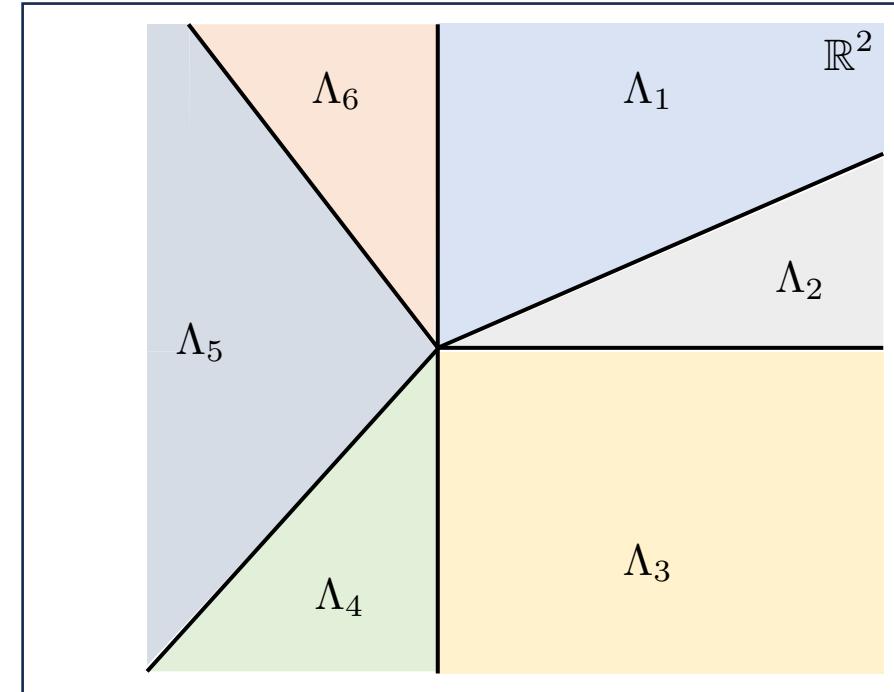
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with  $\bigcup_{k=1}^K \Lambda_k = \mathbb{R}^m$  and  $\text{int}(\Lambda_i) \cap \text{int}(\Lambda_j) = \emptyset$

Points in  $\Lambda_k$  with at least distance  $d_k$  to the boundary of  $\Lambda_k$

if  $\omega - Tx \in \Lambda_k(d_k)$



## Mixed-integer value function

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$$y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}$$

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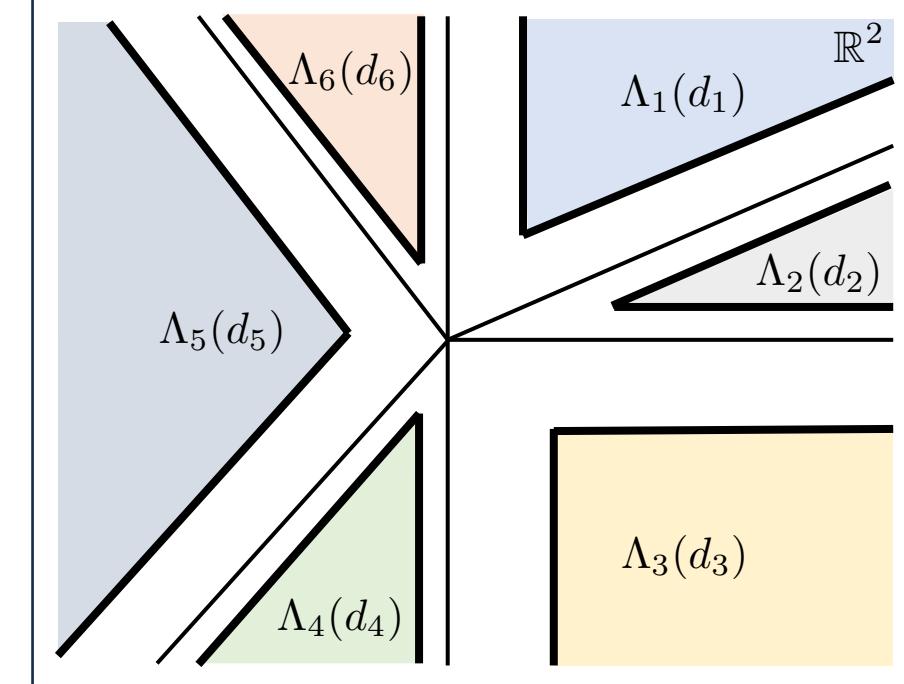
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- such that

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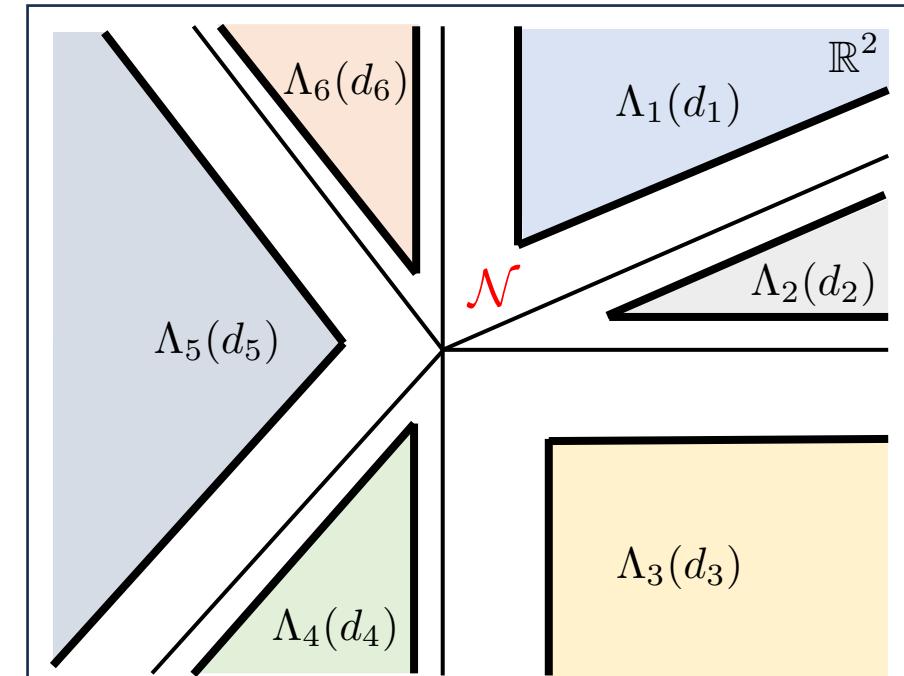
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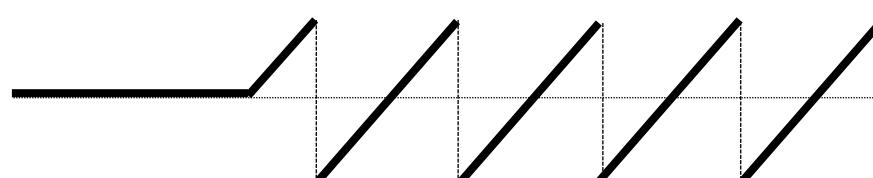
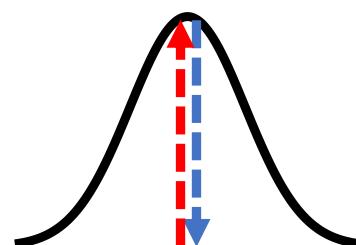
$$\text{s.t. } B^{-1}(\omega - Tx) - B^{-1}Ny_N \in \mathbb{Z}_+^{n_B} \times \mathbb{R}_+^{m-n_B}$$

$$y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}$$

with reduced costs  $\bar{q}_N^\top := q_B^\top - q_N^\top B^{-1}N \geq 0$



- Intuition asymptotic convexity  $Q$ 
  - Area  $\mathcal{N} := \bigcup_{k=1}^K (\Lambda_k \setminus \Lambda_k(d_k))$  is “small”
  - For every  $\Lambda_k(d_k)$ :
    - Total variation error bound for periodic functions

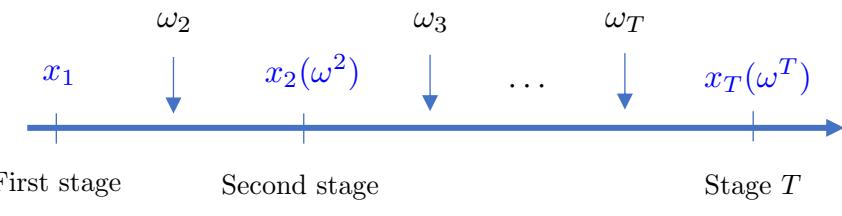


# Multistage SMIPs

- Generic value function:

$$v(\omega, x) = \min_y \left\{ q^\top y + \underbrace{Q(y)}_{\substack{\text{1st-stage variables} \\ \text{2nd-stage variables}}} : W y = \omega - T x, y \in \underbrace{\mathbb{Z}_+^{n_2} \times \mathbb{R}_+^{n_3}}_{\text{2nd-stage feasible region}} \right\}$$

Expected cost-to-go function

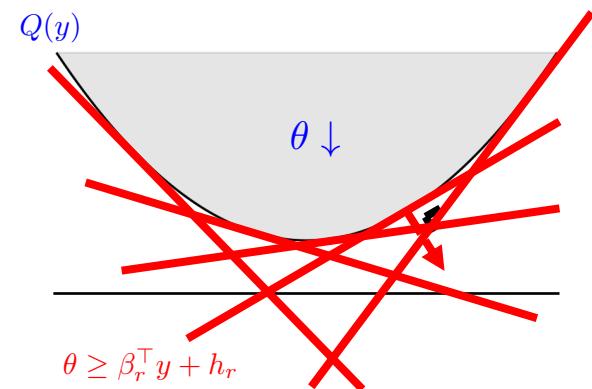


- Assumption:

- Expected cost to-go-function  $Q$  is convex polyhedral
  - Holds if  $x_\tau$  is continuous and  $\omega_\tau$  discrete for  $\tau = t + 1, \dots, T$
  - Holds if  $Q$  can be approximated well by a convex approximation  $\hat{Q}$

$$\rightarrow Q(y) = \min_{\theta} \theta \quad \text{does not depend on } \omega \text{ and } x$$

s.t.  $\theta \geq \beta_r^\top y + h_r \quad \forall r = 1, \dots, R$



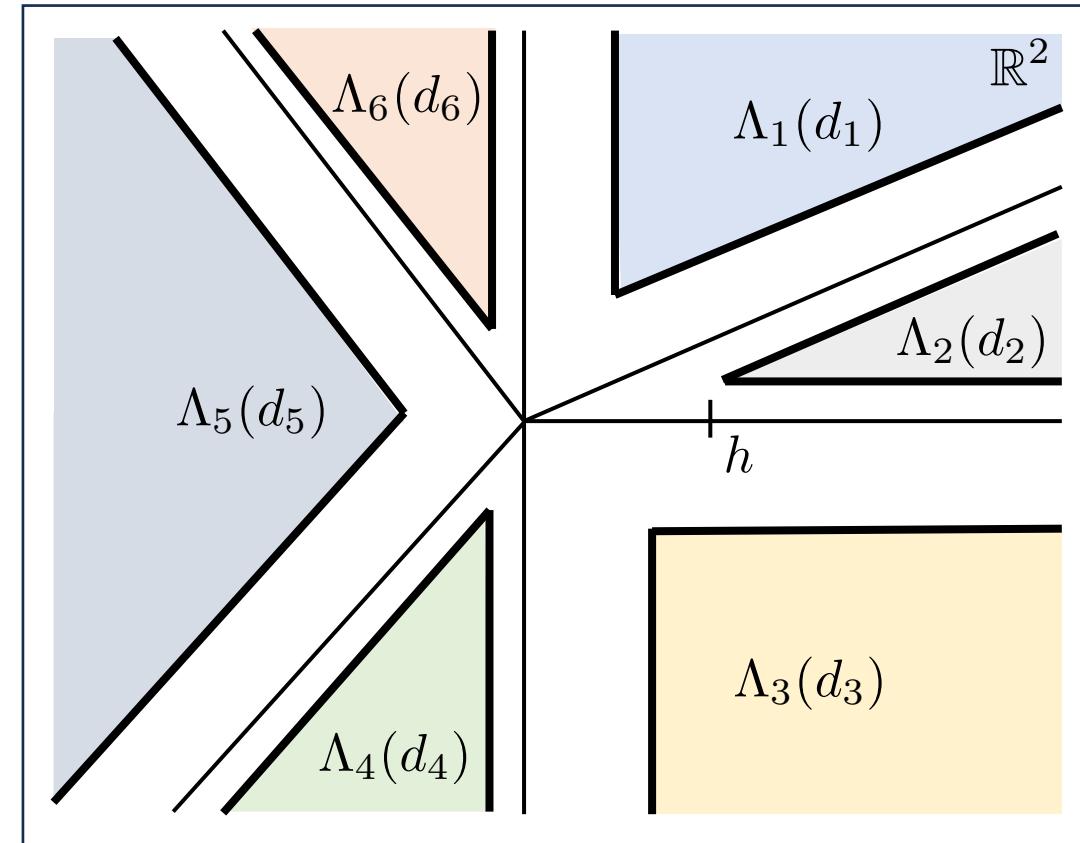
- Generic mixed-integer value function:

$$v(\omega, x) = \min_y \left\{ q^\top y : W y = \begin{pmatrix} h \\ \omega - Tx \end{pmatrix}, y \in \mathbb{Z}_+^{n_2} \times \mathbb{R}_+^{n_3} \right\}$$

## Generic mixed-integer value function

- Definition:

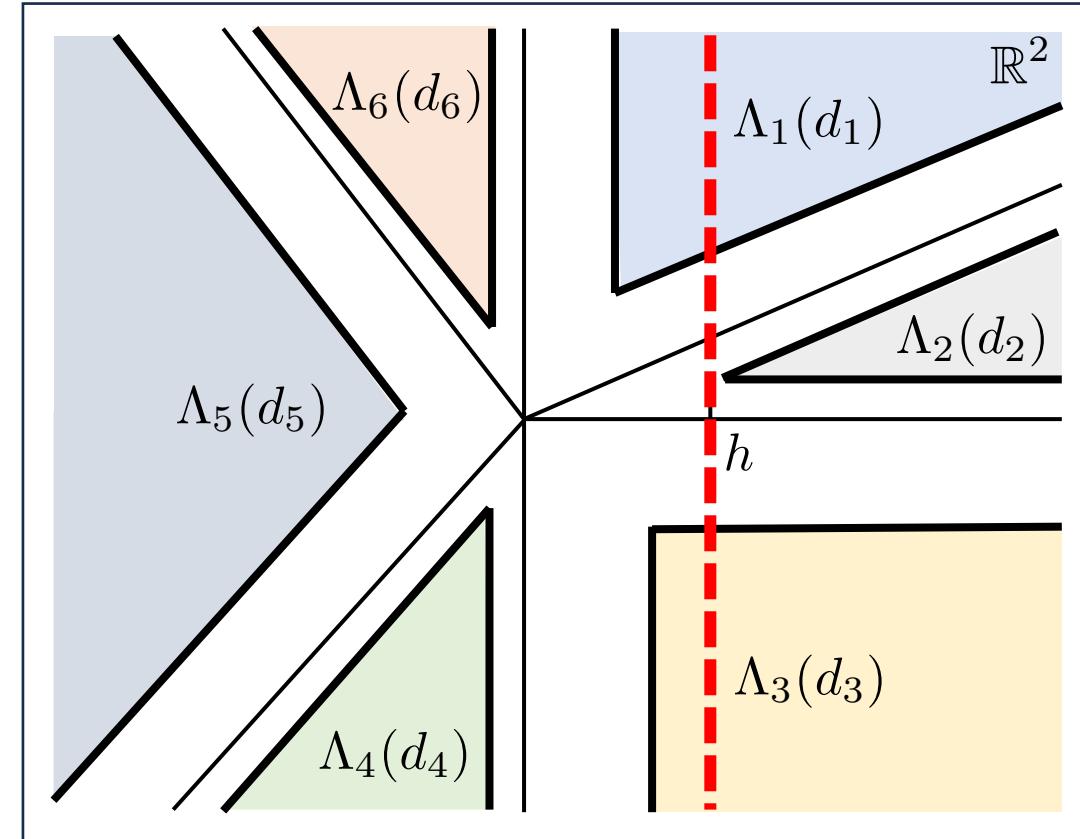
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For fixed  $x$  and  $h$

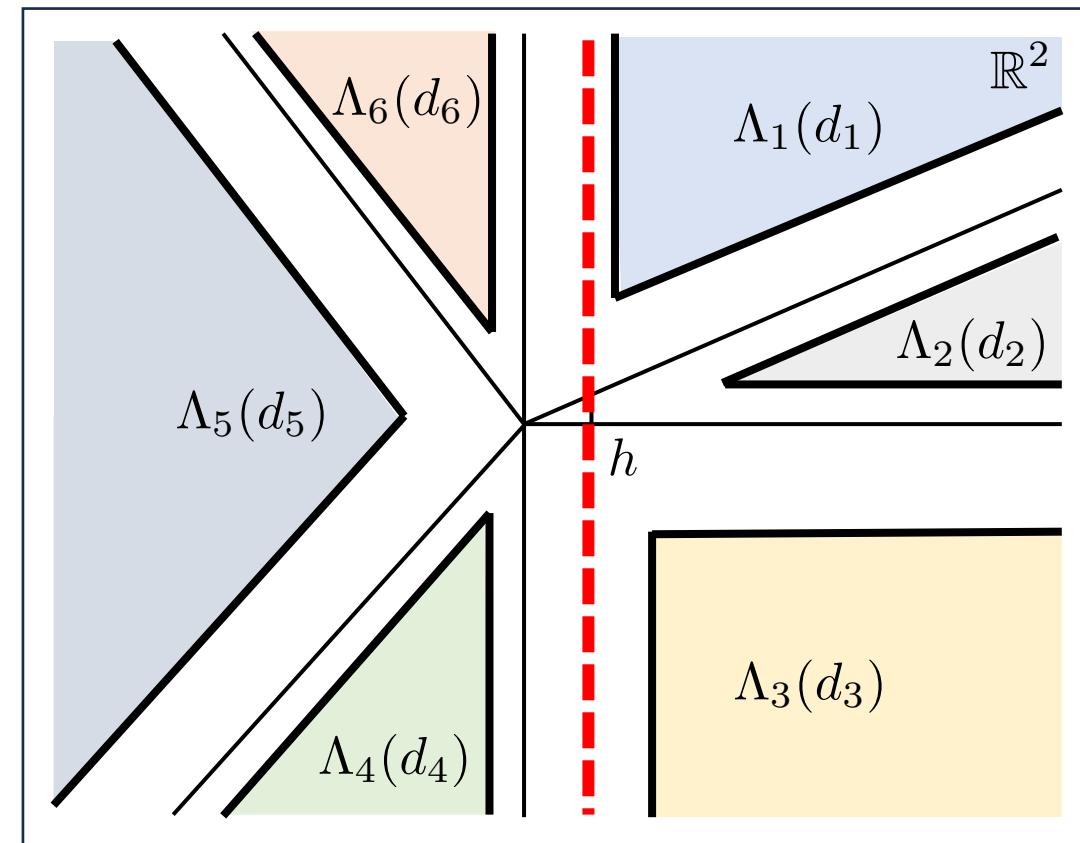
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- Observations:

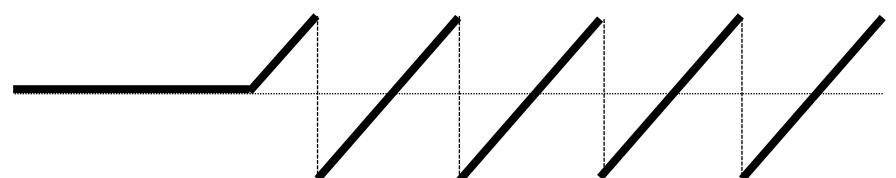
- Need to derive additional properties of  $v(\omega, x)$  when right-hand side is partially uncertain



For fixed  $x$  and  $h$

- Next steps:

- Prove asymptotic periodicity for  $v(\omega, x)$  with partial r.h.s. uncertainty
  - Requires new Adapted Gomory relaxation



## Gomory relaxation

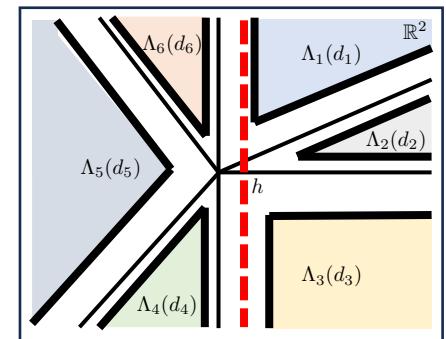
$$\begin{aligned}
 v(\omega, x) = q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \quad & \bar{q}_N^\top y_N \\
 \text{s.t.} \quad & B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} N y_N \in \mathbb{Z}_+^{n_B} \times \mathbb{R}_+^{m-n_B} \\
 & y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$

- Identify critical basic variables  $(y_B)_i$ :

- Write  $B^{-1} = (B_h^{-1} \quad B_\omega^{-1}) \rightarrow y_B = B_h^{-1}h + B_\omega^{-1}(\omega - Tx) - B^{-1}Ny_N$
- Define  $i \in I \Leftrightarrow (B_\omega^{-1})_i = 0$

- Adapted Gomory relaxation  $v_B(\omega, x)$ :

- Relax non-negativity of  $(y_B)_i$  if  $i \notin I$



$$\begin{aligned}
 v_B(\omega, x) = q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \quad & \bar{q}_N^\top y_N \\
 \text{s.t.} \quad & B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} N y_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\
 & e_i^\top B_h^{-1} h - e_i^\top B^{-1} N y_N \geq 0 \quad \forall i \in I \\
 & y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$

$i$ -th unit vector  $\xrightarrow{\hspace{1cm}}$

## Adapted Gomory relaxation

- Adapted Gomory relaxation:

$$\begin{aligned} v_B(\omega, x) &= q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \bar{q}_N^\top y_N \\ \text{s.t. } &B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} N y_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\ &e_i^\top B_h^{-1} h - e_i^\top B^{-1} N y_N \geq 0 \quad \forall i \in I \\ &y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N} \end{aligned}$$

- Properties of optimal solutions  $y_N^*(\omega, x)$ :

- For every  $x$ :
  - $|\det B|$ -periodic in  $\omega$
  - uniformly bounded
  - Optimal for  $v(\omega, x)$  if  $(y_B^*(\omega, x))_i \geq 0$  for all  $i \notin I$

## Periodicity of $y_N^*(\omega, x)$

- Adapted Gomory relaxation:

$$\begin{aligned}
 v_B(\omega, x) = & q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \bar{q}_N^\top y_N \\
 \text{s.t. } & B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} Ny_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\
 & e_i^\top B_h^{-1} h - e_i^\top B^{-1} Ny_N \geq 0 \quad \forall i \in I \\
 & y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$

- Definition

- A function  $\varphi : \mathbb{R}^m \rightarrow \mathbb{R}^n$  is called  **$p$ -periodic** for some  $p \in \mathbb{R}$  if and only if

$$\varphi(s + pl) = \varphi(s) \text{ for all } s \in \mathbb{R}^m \text{ and } l \in \mathbb{Z}^m$$

- $|\det B|$ -periodicity of  $y_N^*(\omega, x)$  in  $\omega$ :

- Let  $\omega' = \omega + |\det B|l$  with  $l \in \mathbb{Z}^m$

$$\rightarrow B^{-1} \begin{pmatrix} h \\ \omega' - Tx \end{pmatrix} = B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \overbrace{B^{-1} \begin{pmatrix} 0 \\ |\det B|l \end{pmatrix}}$$

Since  $W$  is integer

$$\frac{1}{\det B} \text{adj}(B) \begin{pmatrix} 0 \\ |\det B|l \end{pmatrix} \in \mathbb{Z}^m$$

- Fractional values of  $B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix}$  and  $B^{-1} \begin{pmatrix} h \\ \omega' - Tx \end{pmatrix}$  are the same

- Thus,  $y_N^*(\omega, x) = y_N^*(\omega + pl, x)$  for  $p = |\det B|$  and for all  $l \in \mathbb{Z}^m$

## Boundedness of $y_N^*(\omega, x)$

- Adapted Gomory relaxation:

$$\begin{aligned}
 v_B(\omega, x) = & q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \bar{q}_N^\top y_N \\
 \text{s.t. } & B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} Ny_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\
 & e_i^\top B_h^{-1} h - e_i^\top B^{-1} Ny_N \geq 0 \quad \forall i \in I \\
 & y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$

- Observations:

- $\bar{q}_N^\top \geq 0$

- For the LP-relaxation:  $y_N^*(\omega, x) = 0 \quad \text{if } e_i^\top B_h^{-1} h \geq 0 \quad \forall i \in I$   
 $\longrightarrow y_N^*(\omega, x) = 0 \text{ is feasible}$

- Theorem (e.g., Schrijver Thm 17.2 adapted to MIP):

- Optimal solutions of MIPs and LP-relaxations are “close”
- There exists a constant  $D > 0$  such that for all  $\omega$  and  $x$ :  $\|y_N^*(\omega, x)\| \leq D$

## Relation between $v(\omega, x)$ and $v_B(\omega, x)$

- Adapted Gomory relaxation:

$$\begin{aligned}
 v_B(\omega, x) &= q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \bar{q}_N^\top y_N \\
 \text{s.t. } &B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} N y_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\
 &e_i^\top B_h^{-1} h - e_i^\top B^{-1} N y_N \geq 0 \quad \forall i \in I \\
 &y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$

- Optimality condition:

- If  $\underbrace{e_i^\top B_h^{-1} h + e_i^\top B_\omega^{-1} (\omega - Tx) - e_i^\top B^{-1} N y_N^*(\omega, x)}_{(y_B^*(\omega, x))_i} \geq 0$  for all  $i \notin I$

then  $y_N^*(\omega, x)$  is optimal for  $v(\omega, x)$

- Sufficient condition:

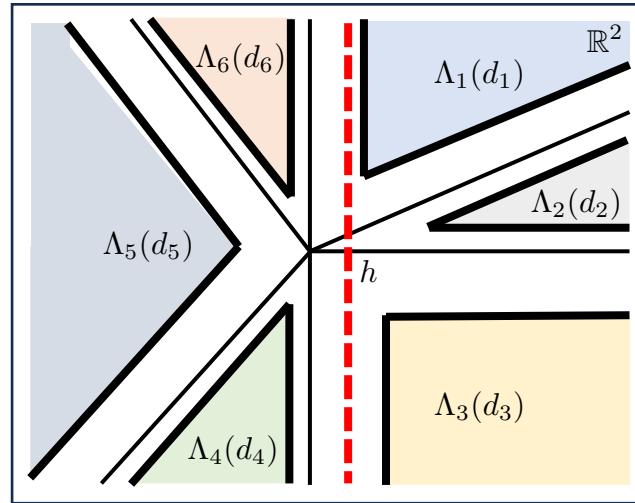
- Use boundedness of  $y_N^*(\omega, x)$
- If  $\omega - Tx \in \mathcal{H}_i(\hat{h}_i)$  for all  $i \notin I$  with  $\mathcal{H}_i(\hat{h}_i) := \left\{ s \in \mathbb{R}^m : e_i^\top B_\omega^{-1} s \geq \hat{h}_i \right\}$

$$\hat{h}_i := \max_{\|y_N\| \leq D} e_i^\top B^{-1} N y_N - e_i^\top B_h^{-1} h$$

# Relation between $v(\omega, x)$ and $v_B(\omega, x)$

- Adapted Gomory relaxation:

$$\begin{aligned}
 v_B(\omega, x) &= q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \bar{q}_N^\top y_N \\
 \text{s.t. } &B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} N y_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\
 &e_i^\top B_h^{-1} h - e_i^\top B_\omega^{-1} (\omega - Tx) \geq 0 \quad \forall i \in I \\
 &y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$



- Optimality condition:

- If  $\underbrace{e_i^\top B_h^{-1} h + e_i^\top B_\omega^{-1} (\omega - Tx) - e_i^\top B^{-1} N y_N^*(\omega, x)}_{(y_B^*(\omega, x))_i} \geq 0$  for all  $i \notin I$

then  $y_N^*(\omega, x)$  is optimal for  $v(\omega, x)$

- Sufficient condition:

- Use boundedness of  $y_N^*(\omega, x)$

- If  $\omega - Tx \in \mathcal{H}_i(\hat{h}_i)$  for all  $i \notin I$  with  $\mathcal{H}_i(\hat{h}_i) := \left\{ s \in \mathbb{R}^m : e_i^\top B_\omega^{-1} s \geq \hat{h}_i \right\}$

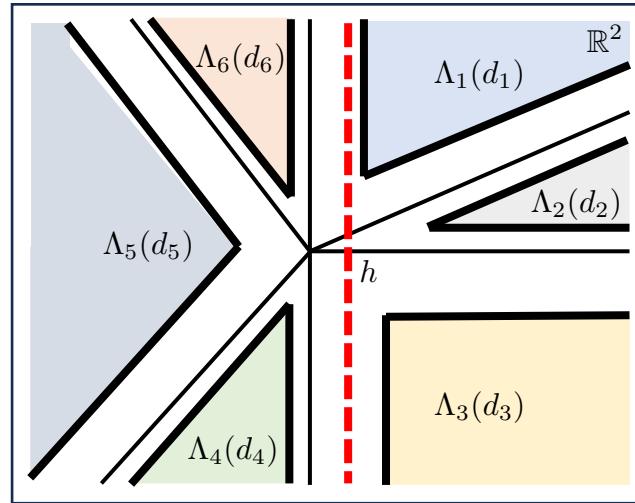
$$\hat{h}_i := \max_{\|y_N\| \leq D} e_i^\top B^{-1} N y_N - e_i^\top B_h^{-1} h$$



# Relation between $v(\omega, x)$ and $v_B(\omega, x)$

- Adapted Gomory relaxation:

$$\begin{aligned}
 v_B(\omega, x) &= q_B^\top B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \min_{y_N} \bar{q}_N^\top y_N \\
 \text{s.t. } &B^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} - B^{-1} N y_N \in \mathbb{Z}^{n_B} \times \mathbb{R}^{m-n_B} \\
 &e_i^\top B_h - e_i^\top B^{-1} N y_N \geq 0 \quad \forall i \in I \\
 &y_N \in \mathbb{Z}_+^{n_N} \times \mathbb{R}_+^{n-n_N}
 \end{aligned}$$

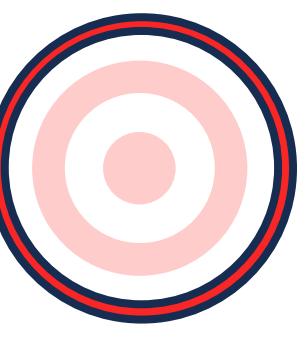


- There exist  $\tilde{\Lambda}_k(\tilde{d}_k) \subset \mathbb{R}^m$  such that

$$v(\omega, x) = \underbrace{q_{B_k}^\top B_k^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix}}_{\text{affine}} + \underbrace{\bar{q}_{N_k}^\top y_{N_k}^*(\omega, x)}_{\text{periodic}} \quad \text{if } \omega - Tx \in \tilde{\Lambda}_k(\tilde{d}_k)$$

- Define  $\psi_k(\omega, x) := \bar{q}_{N_k}^\top y_{N_k}^*(\omega, x)$

# Convex approximation

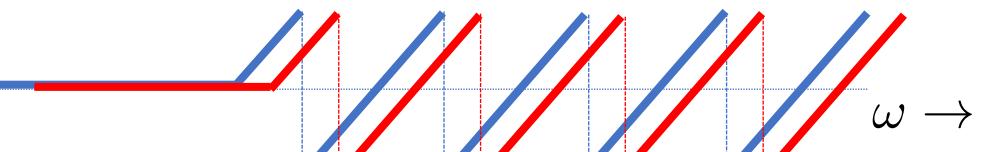


- Idea of convex approximation

$$v(\omega, x) = q_{B_k}^\top B_k^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \psi_k(\omega, x) \quad \text{if } \omega - Tx \in \tilde{\Lambda}_k(\tilde{d}_k)$$

- Replace  $x$  in  $\psi_k(\omega, x)$  by a constant  $\alpha$

$\rightarrow \psi_k(\omega, x) - \psi_k(\omega, \alpha)$  is periodic



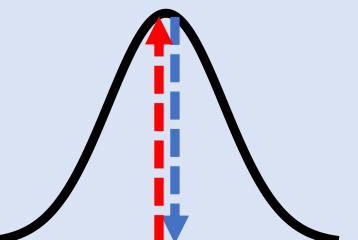
- Definition convex approximation  $\hat{v}$

$$\hat{v}(\omega, x) = \max_{k=1, \dots, K} \left\{ q_{B^k}^\top (B^k)^{-1} \begin{pmatrix} h \\ \omega - Tx \end{pmatrix} + \psi_k(\omega, \alpha) \right\}$$

- Error bound (Independent case)

- There exists a constant  $C > 0$  such that  
for all independent random vectors  $\omega$  with marginal density functions  $f_i$

$$\|Q - \hat{Q}\|_\infty \leq C \sum_{i=1}^m |\Delta| f_i$$



# Conclusion

- We derive a **convex approximation** and **error bound** for **two-stage SMIP** with partially uncertain right-hand side
- Can be applied to **expected cost-to-go functions** of M-SMIPs
  - By using induction over time stages
- Future research direction:
  - Construct **SDDP-like algorithms** based on convex approximation