

Computational Experience with Logmip Solving Linear and Nonlinear Disjunctive Programming Problems

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Motivation

- ❑ Modeling framework for facilitating the formulation of models that can be expressed in terms of algebraic, disjunctive and symbolic logic expressions
- ❑ Language compiler within GAMS for disjunctions expressions, logic constraints and logic propositions.
- ❑ Solution algorithms and techniques for solving linear /nonlinear disjunctive programming problems.

Generalized Disjunctive Programming (GDP)

- Raman and Grossmann (1994)

OR operator \longrightarrow

$$\begin{aligned} \min \quad & Z = \sum_k c_k + f(x) && \text{Objective Function} \\ \text{s.t.} \quad & r(x) \leq 0 && \text{Common Constraints} \\ & \bigvee_{j \in J_k} \begin{bmatrix} Y_{jk} \\ g_{jk}(x) \leq 0 \\ c_k = \gamma_{jk} \end{bmatrix}, k \in K && \text{Disjunction Constraints} \\ & \Omega(Y) = \text{true} && \text{Fixed Charges} \\ & x \in R^n, c_k \in R^1 && \text{Logic Propositions} \\ & Y_{jk} \in \{ \text{true}, \text{false} \} && \text{Continuous Variables} \\ & && \text{Boolean Variables} \end{aligned}$$

Relaxation?

Big-M MINLP (BM)

- MINLP reformulation of GDP

$$\min \mathbf{Z} = \sum_{k \in K} \sum_{j \in J_k} \gamma_{jk} \mathbf{y}_{jk} + f(\mathbf{x})$$

$$s.t. \quad r(\mathbf{x}) \leq 0$$

$$\mathbf{g}_{jk}(\mathbf{x}) \leq M_{jk}(1 - \mathbf{y}_{jk}), \quad j \in J_k, k \in K$$

Big-M Parameter

$$\sum_{j \in J_k} \mathbf{y}_{jk} = 1, \quad k \in K$$

$$A\mathbf{y} \leq \mathbf{a}$$

Logic constraints

$$\mathbf{x} \geq 0, \mathbf{y}_{jk} \in \{0,1\}$$

NLP Relaxation

$$0 \leq \mathbf{y}_{jk} \leq 1$$

Convex Hull Formulation

- Consider **Disjunction** $k \in K$

$$\forall_{j \in J_k} \begin{bmatrix} Y_{jk} \\ g_{jk}(x) \leq 0 \\ c = \gamma_{jk} \end{bmatrix}$$

- ◆ Theorem: Convex Hull of Disjunction k (Lee, Grossmann, 2000)

- **Disaggregated variables** v^j

$$\{(x, c) \mid x = \sum_{j \in J_k} v^{jk}, \quad c = \sum_{j \in J_k} \lambda_{jk} \gamma_{jk},$$

$$0 \leq v^{jk} \leq \lambda_{jk} U_{jk}, \quad j \in J_k$$

=> **Convex Constraints**

$$\sum_{j \in J_k} \lambda_{jk} = 1, \quad 0 < \lambda_{jk} \leq 1,$$

$$\lambda_{jk} g_{jk}(v^{jk} / \lambda_{jk}) \leq 0, \quad j \in J_k \}$$

- λ_j - weights for linear combination

- Generalization of Stubbs and Mehrotra (1999)

Convex Relaxation Problem (CRP)

CRP:

$$\min Z = \sum_{k \in K} \sum_{j \in J_k} \gamma_{jk} \lambda_{jk} + f(x)$$

$$s.t. \quad r(x) \leq 0$$

$$x = \sum_{j \in J_k} v^{jk}, k \in K$$

$$0 \leq v^{jk} \leq \lambda_{jk} U_{jk}, j \in J_k, k \in K$$

$$\sum_{j \in J_k} \lambda_{jk} = 1, k \in K$$

$$\lambda_{jk} g_{jk}(v^{jk} / \lambda_{jk}) \leq 0, j \in J_k, k \in K$$

$$A\lambda \leq a$$

$$x, v^{jk} \geq 0, 0 \leq \lambda_{jk} \leq 1, j \in J_k, k \in K$$

**Convex Hull
Formulation**

Logic constraints

- ◆ **Property:** *The NLP (CRP) yields a lower bound to optimum of (GDP).*
- ◆ **Remark:** *MINLP reformulation by setting $\lambda_{jk} = 0, 1$*

Big-M MINLP (BM)

- ◆ **Theorem:** *The relaxation of (CRP) yields a lower bound that is greater than or equal to the lower bound that is obtained from the relaxation of problem (BM):*

RBM:

$$\min Z = \sum_{k \in K} \sum_{j \in J_k} \gamma_{jk} y_{jk} + f(x)$$

$$s.t. \quad r(x) \leq 0$$

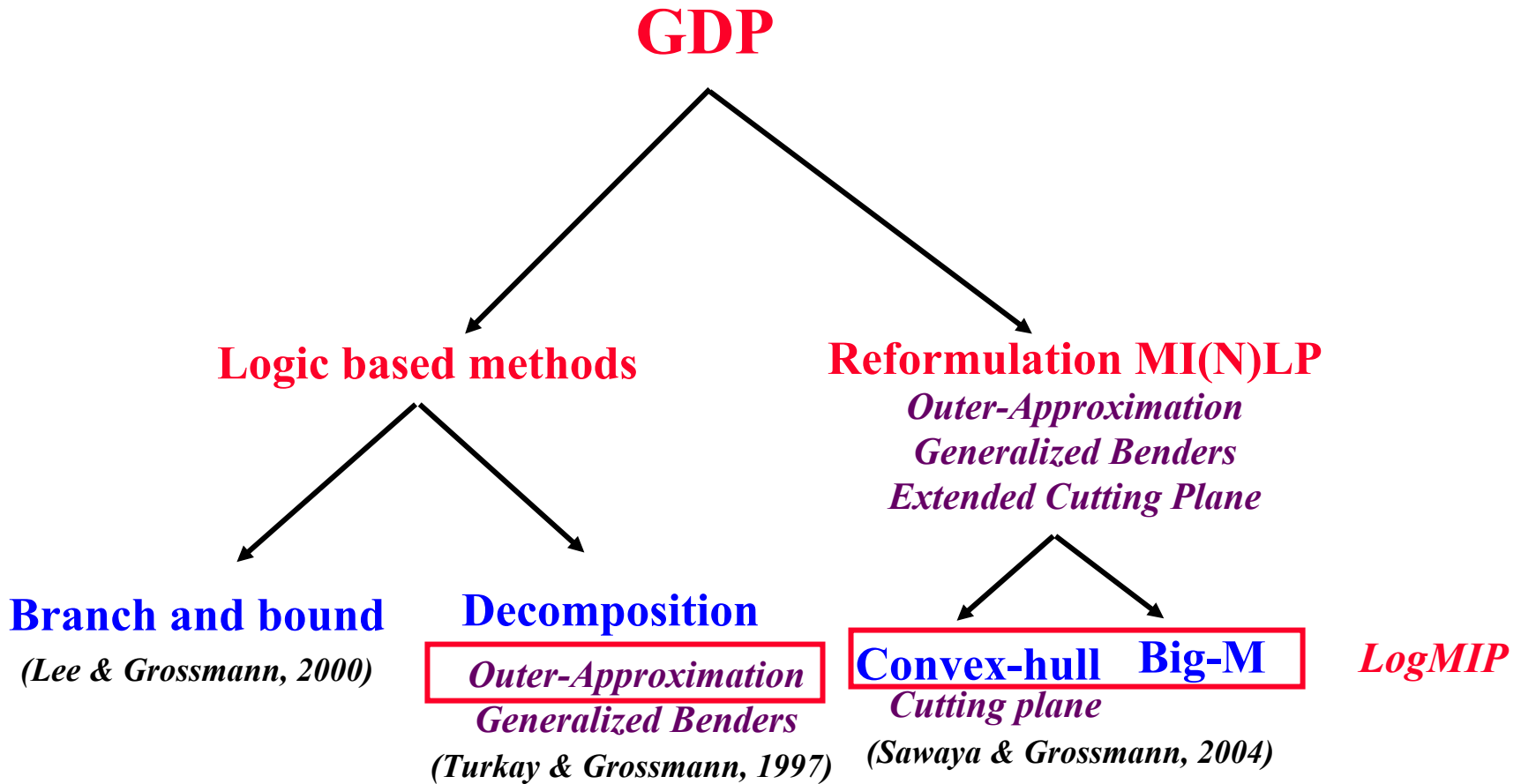
$$g_{jk}(x) \leq M_{jk}(1 - y_{jk}) \quad j \in J_k, k \in K$$

$$\sum_{j \in J_k} y_{jk} = 1, \quad k \in K$$

$$Ay \leq a$$

$$x \geq 0, \quad 0 \leq y_{jk} \leq 1$$

Methods Generalized Disjunctive Programming



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Properties of LogMIP language

- Simple syntax
- Semantic close to disjunction expression to model blocks, exclusive between them
- The syntax must be known for a regular user
- The syntax construction must allow the definition of embeded disjunctions

Notation

Symbol	Meaning
< >	The phrase enclosed is a syntactic rule
Word	Token
[]	Optional expression
{ }	Expression that can be repeated
()	Expression enclosed can be grouped
::=	Define like
	OR

Rules describing LogMIP syntax

```

<LOGMIP Model> ::=
  <Disjunction Declaration> { ; <LogMIP Model> }
  | <Disjunction Definition> { ; <LogMIP Model> }
  ;
<Disjunction Declaration> ::=
  disjunction identifier { , identifier } ;
  ;
<Disjunction Definition> ::=
  disjunction identifier is <If Sentence>
  ;
  
```

```

<If Sentence> ::=
  if <condition> then <components> else
  <components> end if ;
  | if <condition> then <components> { elsif
  <condition> then <components> } end if
  ;
  <components> ::=
  entity { ; entity } [ ; <If Sentence> ]
  ;
  
```

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$$\begin{bmatrix} \textit{True} \\ \textit{Constraint 1} \end{bmatrix} \vee \begin{bmatrix} \textit{False} \\ \textit{Constraint 2} \end{bmatrix}$$

Modeling two terms disjunction

IF (condition₁) THEN

Constraints to be considered when condition₁ is True

ELSE

Constraints to be considered when condition₁ is False

END IF

$$\begin{bmatrix} 1 \\ \textit{Constraints 1} \end{bmatrix} \vee \begin{bmatrix} 2 \\ \textit{Constraints 2} \end{bmatrix} \vee \dots \vee \begin{bmatrix} N \\ \textit{Constraints N} \end{bmatrix}$$

Modeling a multi-term disjunction

IF (condition₁) THEN

Constraints to be considered when condition₁ is True

ELSIF (condition₂) THEN

Constraints to be considered when condition₁ is True

ELSIF (condition₃) THEN

...

ELSIF (condition_N) THEN

Constraints to be considered when condition₁ is True

END IF

Small Example

$$\min c + 2x_1 + x_2$$

s.a.:

$$\left[\begin{array}{c} Y_1 \\ -x_1 + x_2 + 2 \leq 0 \\ c = 5 \end{array} \right] \vee \left[\begin{array}{c} Y_2 \\ 2 - x_2 \leq 0 \\ c = 7 \end{array} \right]$$

$$\left[\begin{array}{c} Y_3 \\ x_1 - x_2 \leq 1 \end{array} \right] \vee \left[\begin{array}{c} \neg Y_3 \\ x_1 = 0 \end{array} \right]$$

$$Y_1 \wedge \neg Y_2 \Rightarrow \neg Y_3$$

$$\neg(Y_2 \wedge Y_3)$$

$$0 \leq x_1 \leq 5, 0 \leq x_2 \leq 5, c \geq 0$$

$$Y_j \in \{true, false\}, j = 1, 2, 3$$

Small Example

```

SET I /1*3/;
SET J /1*2/;
BINARY VARIABLES Y(I);
POSITIVE VARIABLES X(J), C;
VARIABLE Z;
EQUATIONS EQUAT1, EQUAT2, EQUAT3,
EQUAT4, EQUAT5, EQUAT6,
    DUMMY, OBJECTIVE;

EQUAT1.. X('2')- X('1') + 2 =L= 0;
EQUAT2.. C =E= 5;
EQUAT3.. 2 - X('2') =L= 0;
EQUAT4.. C =E= 7;
EQUAT5.. X('1')-X('2') =L= 1;
EQUAT6.. X('1') =E= 0;
DUMMY.. SUM(I, Y(I)) =G= 0;

OBJECTIVE.. Z =E= C + 2*X('1') + X('2');
X.UP(J)=20;
C.UP=7;

```

```

$ONTEXT BEGIN LOGMIP

```

```

DISJUNCTION D1, D2;

```

```

D1 IS

```

```

IF Y('1') THEN

```

```

    EQUAT1;

```

```

    EQUAT2;

```

```

ELSIF Y('2') THEN

```

```

    EQUAT3;

```

```

    EQUAT4;

```

```

ENDIF;

```

```

D2 IS

```

```

IF Y('3') THEN

```

```

    EQUAT5;

```

```

ELSE

```

```

    EQUAT6;

```

```

ENDIF;

```

```

Y('1') and not Y('2') -> not Y('3');

```

```

Y('2') -> not Y('3');

```

```

Y('3') -> not Y('2');

```

```

$OFFTEXT END LOGMIP

```

```

OPTION MIP=LOGMIPM;

```

```

MODEL PEQUE /ALL/;

```

```

SOLVE PEQUE USING MIP MINIMIZING Z;

```

Recent developments

-Disjunctions specified with IF Then ELSE statements

DISJUNCTION D1(I,K,J);

D1(I,K,J)

with (L(I,K,J)) IS

IF Y(I,K,J) THEN

NOCLASH1(I,K,J);

ELSE

NOCLASH2(I,K,J);

ENDIF;

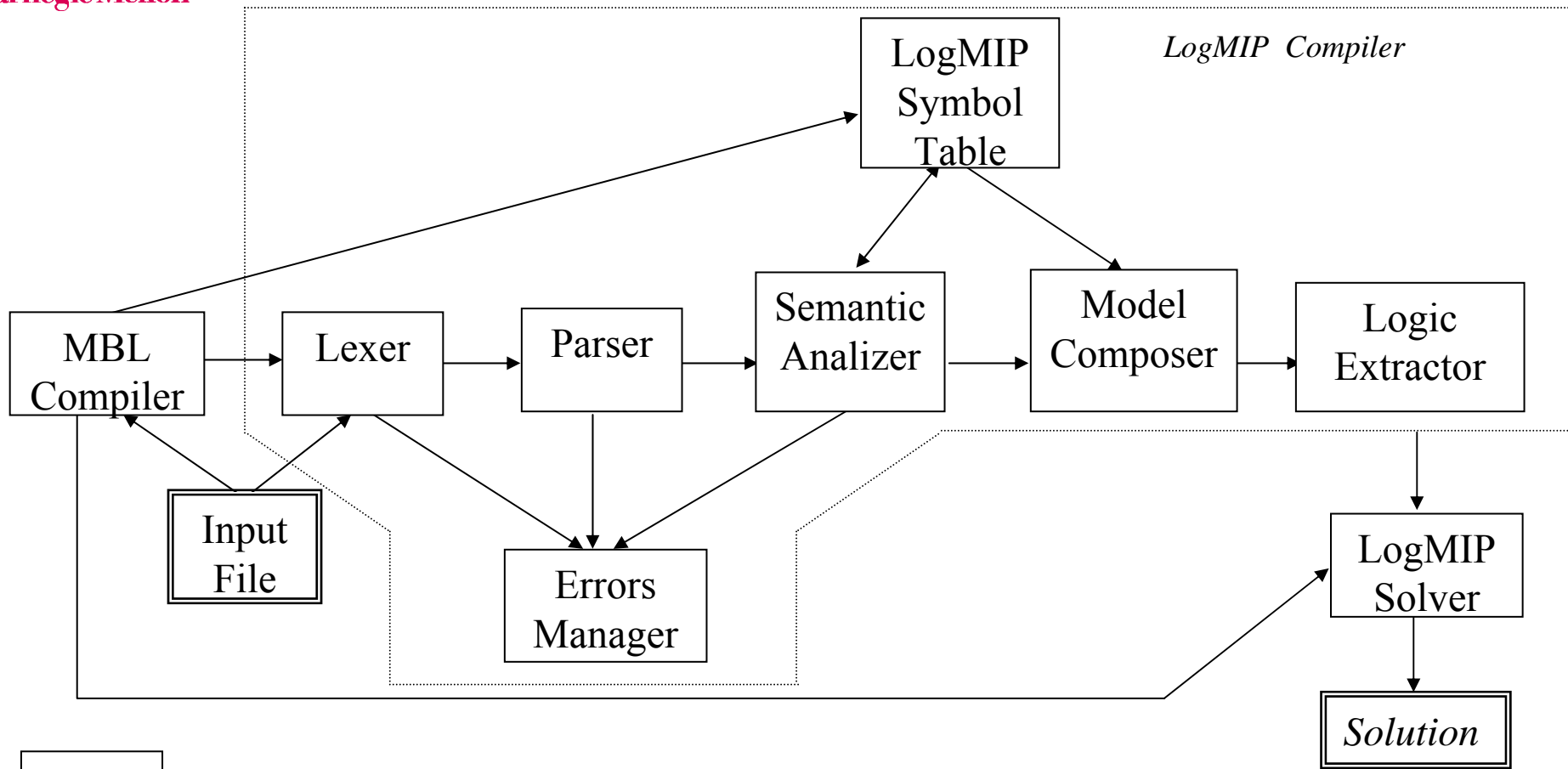
-Logic can be specified in symbolic form (\Rightarrow , OR, AND, NOT)

or special operators (ATMOST, ATLEAST, EXACTLY)

-Linear case: MILP reformulation big-M, convex hull

-Nonlinear: Logic-based OA

LogMIP Compiler Architecture



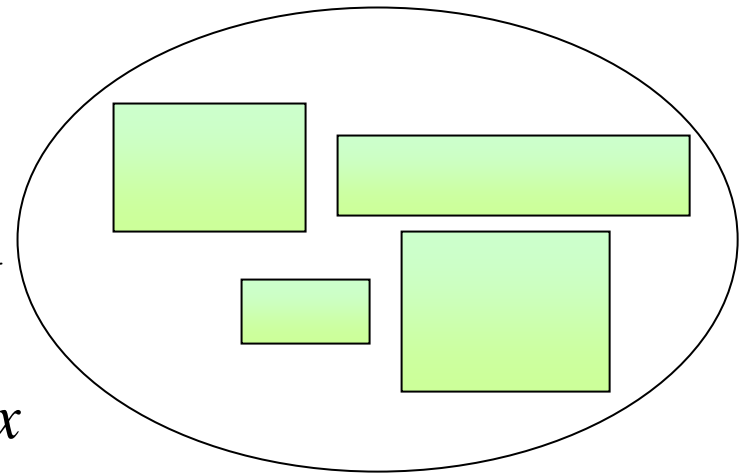
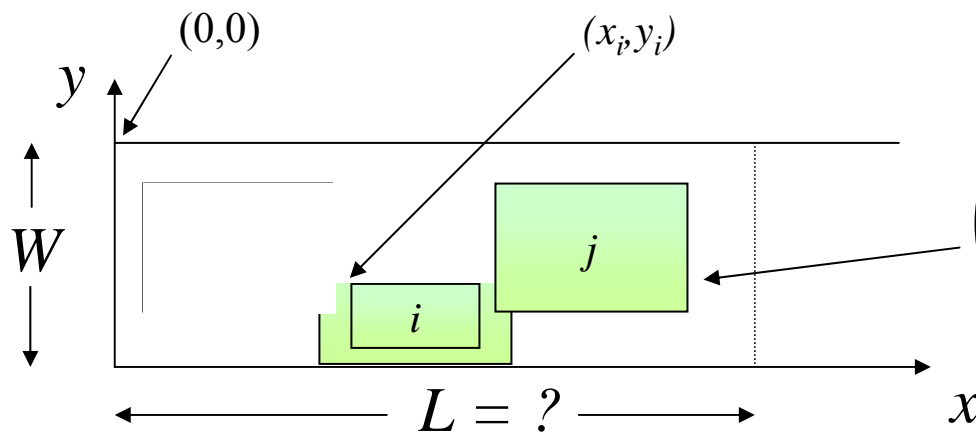
Filter

→ Pipe

Strip-packing Problem

Problem statement: *Hifi (1998)*

- Given a set of small rectangles with width w_i and length l_i .
- Large rectangular strip of fixed width W and unknown length L .
- Objective is to fit small rectangles onto strip without overlap and rotation while **minimizing length L of the strip.**



Set of small rectangles

GDP Model For Strip-packing Problem

<i>Min</i>	lt	
<i>st.</i>	$lt \geq x_i + L_i$	$\forall i \in N$
$\left[\begin{array}{c} Y_{ij}^1 \\ x_i + L_i \leq x_j \end{array} \right] \vee \left[\begin{array}{c} Y_{ij}^2 \\ x_j + L_j \leq x_i \end{array} \right] \vee \left[\begin{array}{c} Y_{ij}^3 \\ y_i - H_i \geq y_j \end{array} \right] \vee \left[\begin{array}{c} Y_{ij}^4 \\ y_j - H_j \geq y_i \end{array} \right]$		$\forall i, j \in N, i < j$
$x_i \leq UB_i - L_i$		$\forall i \in N$
$H_i \leq y_i \leq W$		$\forall i \in N$
$lt, x_i, y_i \in \mathbf{R}_+^1, Y_{ij}^1, Y_{ij}^2, Y_{ij}^3, Y_{ij}^4 \in \{True, False\}$		$\forall i, j \in N, i < j$

Objective function

Minimize length

Disjunctive constraints

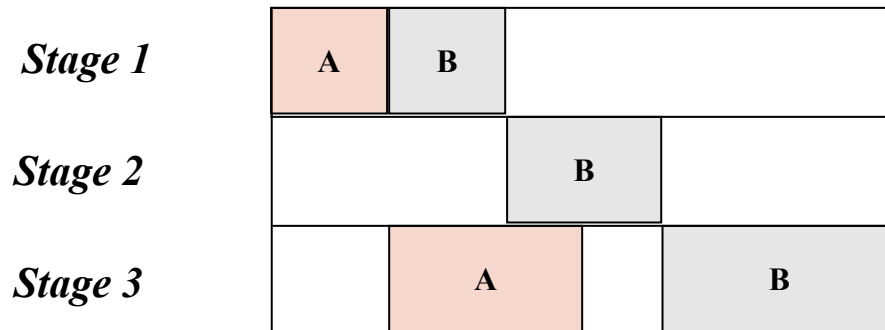
No overlap between rectangles

Bounds on variables

Zero-wait Job-shop Scheduling Problem

Problem Statement: *Raman R. & Grossmann I.E. (1994)*

- Set of jobs $i \in I$ must be processed sequentially on a set of consecutive stages $j \in J$.
 - All jobs can be sequenced on a subset of stages $j \in J(i)$.
 - Zero-wait transfer is assumed between stages.
- Objective is to obtain a schedule that **minimizes makespan**.



GDP Model For Zero-wait Job-shop Scheduling Problem

Min ms

s.t. $ms \geq t_i + \sum_{\forall j \in J(i)} TAU_{ij} \quad \forall i \in I$

$$\left[\begin{array}{c} Y_{ik}^1 \\ t_i + \sum_{\substack{\forall m \in J(i) \\ m \leq j}} TAU_{im} \leq t_k + \sum_{\substack{\forall m \in J(k) \\ m < j}} TAU_{km} \end{array} \right] \vee \left[\begin{array}{c} Y_{ik}^2 \\ t_k + \sum_{\substack{\forall m \in J(k) \\ m \leq j}} TAU_{km} \leq t_i + \sum_{\substack{\forall m \in J(i) \\ m < j}} TAU_{im} \end{array} \right] \quad \forall j \in C_{ik}, \forall i, k \in I, i < k$$

$ms, t_i \in \mathbf{R}_+^1, Y_{ik}^1, Y_{ik}^2 \in \{True, False\} \quad \forall i, k \in I, i < k$

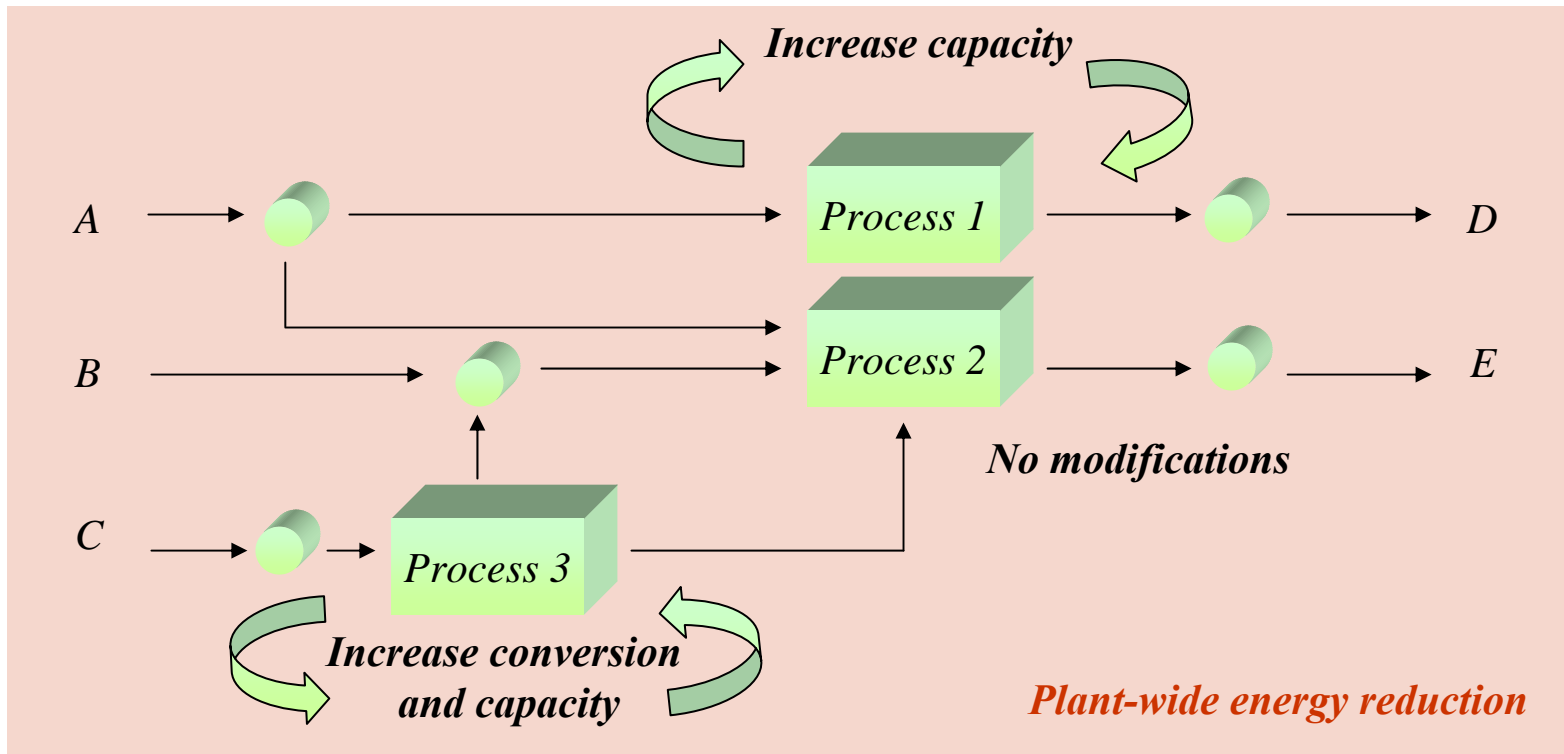
Objective function
Minimize makespan

Disjunctive constraints
No clashes between jobs

Retrofit Planning Problem

Problem Statement: *Jackson J. & Grossmann I.E. (2002)*

- Retrofit: Redesign of existing plant.
 - Improvements such as higher yield, increased capacity, energy reduction.
- Objective is to identify modifications that **maximize economic potential**, given time horizon and limited capital investments.



GDP Model For Retrofit Planning Problem

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$$\sum_{\forall t \in T} \sum_{\forall s \in S_{prod}} PR_s^t m f_s^t - \sum_{\forall t \in T} \sum_{\forall s \in S_{raw}} PR_s^t m f_s^t - \sum_{\forall t \in T} PRST qst^t - \sum_{\forall t \in T} PRWT qwt^t - \sum_{\forall t \in T} \sum_{\forall p \in P} fc_p^t - \sum_{\forall t \in T} ec^t$$

$$\begin{aligned} s.t. \quad & m f_s^t = f_s^t M W_s & \forall s \in S, \forall t \in T \\ & m f_s^t \geq DEM_s^t & \forall s \in S_{prod}, \forall t \in T \\ & m f_s^t \leq SUP_s^t & \forall s \in S_{raw}, \forall t \in T \\ & \sum_{\forall s \in S_{in}} m f_s^t = \sum_{\forall s \in S_{out}} m f_s^t & \forall n \in N, \forall t \in T \\ & \sum_{\forall s \in S_{pin}} m f_s^t = \sum_{\forall s \in S_{pout}} m f_s^t + unrcr_p^t & \forall p \in P, \forall t \in T \end{aligned}$$

$$\begin{aligned} \forall m \in M \quad & \left[\begin{array}{l} Y_{pm}^t \\ f_s^t = f_{p_{in}}^t \left(\frac{GMA^t}{GMA_{p_{in}}^t} \right) ETA_{pm}^t \\ \sum_{\forall s \in S_{pin}} m f_s^t \leq CAP_{pm}^t \end{array} \right] & \forall s \in S_{p_{out}}, \forall p \in P, \forall t \in T \\ \forall m \in M \quad & \left[\begin{array}{l} W_{pm}^t \\ fc_p^t = FC_{pm}^t \end{array} \right] & \forall p \in P, \forall t \in T \end{aligned}$$

$$\begin{aligned} q_{sk}^t &= m f_s^t CP_s (T_{s_{out k}^t} - T_{s_{in k}^t}) & \forall s \in S_{cold}, \forall k \in K, \forall t \in T \\ q_{sk}^t &= m f_s^t CP_s (T_{s_{in k}^t} - T_{s_{out k}^t}) & \forall s \in S_{hot}, \forall k \in K, \forall t \in T \\ \left[\begin{array}{l} X_1^t \\ qst^t = \sum_{\forall k \in K} \sum_{\forall s \in S_{cold}} q_{sk}^t \\ qwt^t = \sum_{\forall k \in K} \sum_{\forall s \in S_{hot}} q_{sk}^t \end{array} \right] & \vee \left[\begin{array}{l} X_2^t \\ r_k^t - r_{k-1}^t - qst_k^t + qwt_k^t = \sum_{\forall s \in S_{hot}} q_{sk}^t - \sum_{\forall s \in S_{cold}} q_{sk}^t \\ qst^t = \sum_{\forall k \in K} qst_k^t \\ qwt^t = r_{|K|}^t + \sum_{\forall k \in K} qwt_k^t \end{array} \right] & \forall k \in K, \forall t \in T \\ \left[\begin{array}{l} V_1^t \\ ec^t = EFC_1^t \end{array} \right] & \vee \left[\begin{array}{l} V_2^t \\ ec^t = EFC_2^t \end{array} \right] & \forall t \in T \end{aligned}$$

$$\sum_{\forall p \in P} fc_p^t + ec^t + \sum_{\forall s \in S_{raw}} PR_s^t m f_s^t + PRST qst^t + PRWT qwt^t \leq INV^t \quad \forall t \in T$$

$$\begin{aligned} Y_{pm}^t &\rightarrow \bigwedge_{\tau < t} Y_{pm}^{\tau} & \forall p \in P, \forall t \in T, \forall m \in M \setminus m_i \\ W_{pm}^t &\rightarrow \bigwedge_{\tau < t} W_{pm}^{\tau} & \forall p \in P, \forall t \in T, \forall m \in M \setminus m_i \\ Y_{pi}^t &\rightarrow W_{pi}^t & \forall p \in P, \forall t \in T \\ Y_{pm}^t \wedge \neg Y_{pm}^{\tau} &\rightarrow W_{pm}^{\tau} & \forall p \in P, \forall t \in T, \forall m \in M \setminus m_i \\ X_j^t &\rightarrow \bigwedge_{\tau < t} X_j^{\tau} & \forall t \in T, \forall j \in J \setminus j_i \\ V_j^t &\rightarrow \bigwedge_{\tau < t} V_j^{\tau} & \forall t \in T, \forall j \in J \setminus j_i \\ X_i^t &\rightarrow V_i^t & \forall t \in T \\ X_j^t \wedge \neg X_j^{\tau} &\rightarrow V_j^{\tau} & \forall t \in T, \forall j \in J \setminus j_i \end{aligned}$$

$$\begin{aligned} m f_s^t, f_s^t &\in \mathbf{R}_+^1 & \forall s \in S, \forall t \in T \\ f_{lim,p}^t, unrcr_p^t, fc_p^t &\in \mathbf{R}_+^1 & \forall p \in P, \forall t \in T \\ q_{sk}^t &\in \mathbf{R}_+^1 & \forall s \in S, \forall k \in K, \forall t \in T \\ qst^t, qwt^t, ec^t &\in \mathbf{R}_+^1 & \forall t \in T \\ qst_k^t, qwt_k^t, r_k^t &\in \mathbf{R}_+^1 & \forall k \in K, \forall t \in T \\ Y_m^t, W_m^t &\in \{True, False\} & \forall p \in P, \forall t \in T, \forall m \in M \\ X_j^t, V_j^t &\in \{True, False\} & \forall j \in J, \forall t \in T \end{aligned}$$

Objective function
Maximize economic potential

Common constraints
Mass balances

Disjunctive constraints
Conversion/Capacity scenarios

Disjunctive constraints
Energy reduction scenarios

Common constraint Investment limit

Logic constraints
Connect disjunctions

Computational Results

Linear Disjunctive Problems

Problem	# disc. var.	# var.	# Equat .	CPU BigM (sec.)	iter. BigM	nodes BigM	CPU CH (sec.)	iter CH	nodes CH
cut-1	24	34	30	0.05	64	32	0.11	92	0
cut-2	180	202	236	7.19	29,196	4673	43.78	44,434	873
jobshop-1	12	21	13	0.05	5	0	0.001	9	0
jobshop-2	245	253	78	0.16	341	86	0.93	2155	200
jobshop-3	320	319	219	3.57	10,034	2209	154.45	207,605	20,600
jobshop-4	125	131	106	0.11	288	55	2.25	3035	299
retrofit	72	160	211	0.72	1449	136	0.11	122	0
retrofitNS	336	1685	2935	1810.11	6,995,748	494,291	2.08	3019	18
pipeline	387	1640	3385	327.52	89,820	13,657	940.65	420,574	23,750

Performance of big-M and Convex Hull (CH) is problem dependent

CPLEX 7.5

Computational Results

Nonlinear Disjunctive Problems

Problem	# disc. var.	# var.	# Equat.	# nlp initial.	#nlp total	# master	CPU master (sec.)	CPU nlp (sec.)
8 processes	8	42	70	3	4	1	1.2	0.22
batch-design	54	113	217	1	2	2	0.82	0.18
spectroscopy	30	99	162	1	14	14	1.71	0.74
methanol	17	310	557	2	5	3	0.55	1.27

Logic-based Outer-Approximation (CPLEX 7.5/CONOPT)

Conclusions

1. LogMIP provides unique capability for formulating and solving discrete/continuous optimization problems (GDPs) in GAMS environment:
 - Handling disjunctions
 - Handling symbolic logic propositions

2. Numerical results show following:
 - For linear GDPs performance of big-M and convex hull is problem dependent
 - For nonlinear GDPs robustness increased with logic-based outer-approximation

3. Future work:
 - Extend big-M and convex hull reformulation to nonlinear GDPs
 - Branch and bound for linear and nonlinear GDPs

LogMIP can be downloaded from <http://www.ceride.gov.ar/logmip/>