

Two-Period Relaxations on Big-Bucket Production Planning Problems

Kerem Akartunali

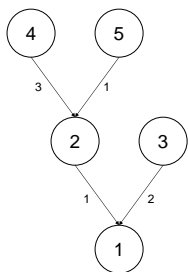
Department of Mathematics and Statistics
The University of Melbourne

Joint work with Andrew J. Miller

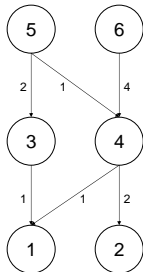
4 August 2008

Problem Description

- Multiple items and levels (BOM structure)
 - Assembly (a) or general (b) structures



(a)



(b)

- Demands
- Big-bucket capacities (items share resources)
- Extensions possible, e.g. overtime and backlogging
- Production plan minimizing total cost to be determined

Basic Formulation

$$\min \sum_{t=1}^{NT} \sum_{i=1}^{NI} f_t^i y_t^i + \sum_{t=1}^{NT} \sum_{i=1}^{NI} h_t^i s_t^i \quad (1)$$

$$\text{s.t. } x_t^i + s_{t-1}^i - s_t^i = d_t^i \quad t \in [1, NT], i \in \text{endp} \quad (2)$$

$$x_t^i + s_{t-1}^i - s_t^i = \sum_{j \in \delta(i)} r^{ij} x_t^j \quad t \in [1, NT], i \notin \text{endp} \quad (3)$$

$$\sum_{i=1}^{NI} (a_k^i x_t^i + ST_k^i y_t^i) \leq C_t^k \quad t \in [1, NT], k \in [1, NK] \quad (4)$$

$$x_t^i \leq M_t^i y_t^i \quad t \in [1, NT], i \in [1, NI] \quad (5)$$

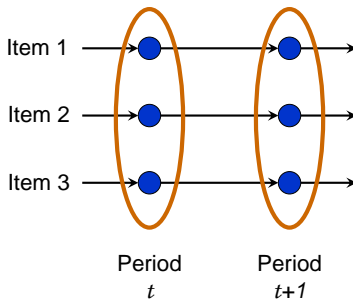
$$y \in \{0, 1\}^{NT \times NI} \quad (6)$$

$$x \geq 0 \quad (7)$$

$$s \geq 0 \quad (8)$$

What do we know?

- Many of the test problems remain challenging
- We do not have an adequate approximation of the convex hull of the **multi-item, single-machine, single-level capacitated problems!**



- Generalizes the “bottleneck flow” model of Atamtürk and Muñoz [2003]

The Model to Study

$$x_{t'}^i \leq \tilde{M}_{t'}^i y_{t'}^i \quad i = [1, \dots, NI], t' = 1, 2 \quad (9)$$

$$x_{t'}^i \leq \tilde{d}_{t'}^i y_{t'}^i + s^i \quad i = [1, \dots, NI], t' = 1, 2 \quad (10)$$

$$x_1^i + x_2^i \leq \tilde{d}_1^i y_1^i + \tilde{d}_2^i y_2^i + s^i \quad i = [1, \dots, NI] \quad (11)$$

$$x_1^i + x_2^i \leq \tilde{d}_1^i + s^i \quad i = [1, \dots, NI] \quad (12)$$

$$\sum_{i=1}^{NI} (a^i x_{t'}^i + ST^i y_{t'}^i) \leq \tilde{C}_{t'} \quad t' = 1, 2 \quad (13)$$

$$x, s \geq 0, y \in \{0, 1\}^{2 \times NI} \quad (14)$$

- Let $X^{2PL} = \{(x, y, s) | (9) - (14)\}$

Separation Over the 2-Period Convex Hull: Details

From the LPR of the original problem, we obtain $(\bar{x}, \bar{y}, \bar{s})$

$$\min z = \sum_i [(\Delta_s^-)^i + \sum_{t'=1}^2 (\Delta_x^+)^i_{t'} + (\Delta_x^-)^i_{t'} + (\Delta_y^+)^i_{t'} + (\Delta_y^-)^i_{t'}]$$

$$\text{s.t. } \bar{x}_{t'}^i = \sum_k \lambda_k (x_k)^i_{t'} + (\Delta_x^+)^i_{t'} - (\Delta_x^-)^i_{t'} \quad \forall i, t' = 1, 2 \quad (\alpha_{t'}^i)$$

$$\bar{y}_{t'}^i = \sum_k \lambda_k (y_k)^i_{t'} + (\Delta_y^+)^i_{t'} - (\Delta_y^-)^i_{t'} \quad \forall i, t' = 1, 2 \quad (\beta_{t'}^i)$$

$$\bar{s}^i \geq \sum_k \lambda_k (s_k)^i - (\Delta_s^-)^i \quad \forall i \quad (\gamma^i)$$

$$\sum_k \lambda_k \leq 1 \quad (\eta)$$

$$\lambda_k \geq 0, \Delta \geq 0$$

Separation Over the 2-Period Convex Hull: Details (cont'd)

The dual of the distance problem:

$$\max \sum_{i=1}^{NI} \sum_{t'=1}^2 (\bar{x}_{t'}^i \alpha_{t'}^i + \bar{y}_{t'}^i \beta_{t'}^i) + \sum_{i=1}^{NI} \bar{s}_k^i \gamma^i + \eta \quad (15)$$

$$\text{s.t.} \quad \sum_{i=1}^{NI} \sum_{t'=1}^2 ((x_k)_{t'}^i \alpha_{t'}^i + (y_k)_{t'}^i \beta_{t'}^i) + \sum_{i=1}^{NI} (s_k)^i \gamma^i + \eta \leq 0 \quad \forall k \quad (16)$$

$$-1 \leq \alpha_{t'}^i \leq 1 \quad \forall i, t' \quad (17)$$

$$-1 \leq \beta_{t'}^i \leq 1 \quad \forall i, t' \quad (18)$$

$$-1 \leq \gamma^i \leq 0 \quad \forall i \quad (19)$$

$$\eta \leq 0 \quad (20)$$

Separation Over the 2-Period Convex Hull: Details (cont'd)

Theorem

Let $z > 0$ for $(\bar{x}, \bar{y}, \bar{s})$, and $(\bar{\alpha}, \bar{\beta}, \bar{\gamma}, \bar{\eta})$ optimal dual values. Then,

$$\sum_{i=1}^{NI} \sum_{t'=1}^2 (\bar{\alpha}_{t'}^i x_{t'}^i + \bar{\beta}_{t'}^i y_{t'}^i) + \sum_i \bar{\gamma}^i s^i + \bar{\eta} \leq 0 \quad (21)$$

is a valid inequality for $\text{conv}(X^{2PL})$ that cuts off $(\bar{x}, \bar{y}, \bar{s})$.

Proof.

Validity: Using (16), $\bar{\gamma} \leq 0$ and $\lambda \geq 0$.

Violation for $(\bar{x}, \bar{y}, \bar{s})$: Using (15) □

Separation Over the 2-Period Convex Hull: Extreme Points

- How to generate (x_k, y_k, s_k) ?
 - Using column generation
 - Solve the minimum reduced cost problem using the optimal dual values $(\bar{\alpha}, \bar{\beta}, \bar{\gamma}, \bar{\eta})$:

$$\begin{aligned} \max_{x,y,s} z_P &= \sum_i \sum_{t'=1}^2 (\bar{\alpha}_{t'}^i x_{t'}^i + \bar{\beta}_{t'}^i y_{t'}^i) + \sum_i \bar{\gamma}^i s^i + \bar{\eta} \\ \text{s.t. } (x, y, s) &\in X^{2PL} \end{aligned}$$

- If $z_P > 0$, then the solution is an extreme point of X^{2PL} ; otherwise, generating extreme points is done

Separation Algorithm

repeat

Solve the distance problem for $\text{conv}(X^{2PL})$

if $z = 0$ **then** break

else Solve column generation problem

if $z_P \leq 0$ **then** break

else Add new extreme point

until $z = 0$ **or** $z_P \leq 0$

if $z=0$ **then** $(\bar{x}, \bar{y}, \bar{s}) \in \text{conv}(X^{2PL})$

else Add the violated cut (21)

Note: The same is valid for \mathcal{L}_∞

Using Euclidean Distance (\mathcal{L}_2)

We can also apply the same framework with Euclidean distance

$$\min_{\Delta, \lambda} z = \sum_i \left[[(\Delta_s)^i]^2 + \sum_{t'=1}^2 ([(\Delta_x)_{t'}^i]^2 + [(\Delta_y)_{t'}^i]^2) \right]$$

$$\text{s.t. } \bar{x}_{t'}^i = \sum_k \lambda_k (x_k)_{t'}^i + (\Delta_x)_{t'}^i \quad \forall i, t' = 1, 2 \quad (\alpha_{t'}^i)$$

$$\bar{y}_{t'}^i = \sum_k \lambda_k (y_k)_{t'}^i + (\Delta_y)_{t'}^i \quad \forall i, t' = 1, 2 \quad (\beta_{t'}^i)$$

$$\bar{s}^i \geq \sum_k \lambda_k (s_k)^i - (\Delta_s)^i \quad \forall i \quad (\gamma^i)$$

$$\sum_k \lambda_k \leq 1 \quad (\eta)$$

$$\lambda_k \geq 0, \Delta_s \geq 0, \Delta_x, \Delta_y \text{ free}$$

Using Euclidean Distance: Dual

$$\begin{aligned}
 \max_{\Delta, \alpha, \beta, \gamma} z_D &= - \sum_i \left[[(\Delta_s)^i]^2 + \sum_{t'=1}^2 [(\Delta_x)_{t'}^i]^2 + [(\Delta_y)_{t'}^i]^2 \right] \\
 &\quad - \left(\sum_{i=1}^{NI} \sum_{t'=1}^2 (\bar{x}_{t'}^i \alpha_{t'}^i + \bar{y}_{t'}^i \beta_{t'}^i) + \sum_{i=1}^{NI} \bar{s}^i \gamma^i + \eta \right) \\
 \text{s.t.} \quad &\sum_{i=1}^{NI} \sum_{t'=1}^2 ((x_k)_{t'}^i \alpha_{t'}^i + (y_k)_{t'}^i \beta_{t'}^i) + \sum_{i=1}^{NI} (s_k)^i \gamma^i + \eta \geq 0 \quad \forall k \\
 &\alpha_{t'}^i = -2(\Delta_x)_{t'}^i \quad \forall i, t' \\
 &\beta_{t'}^i = -2(\Delta_y)_{t'}^i \quad \forall i, t' \\
 &-\gamma^i \geq -2(\Delta_s)^i \quad \forall i \\
 &\gamma \geq 0, \eta \geq 0, \Delta_s \geq 0, \alpha, \beta, \Delta_x, \Delta_y \text{ free}
 \end{aligned}$$

Using Euclidean Distance: Theory

Theorem

Let $z > 0$ for $(\bar{x}, \bar{y}, \bar{s})$, with optimal primal values $(\bar{\Delta}_x, \bar{\Delta}_y, \bar{\Delta}_s, \bar{\lambda})$, and $(\bar{\alpha}, \bar{\beta}, \bar{\gamma}, \bar{\eta})$ be the associated optimal dual values. Then,

$$\sum_i \sum_{t'=1}^2 (\bar{\alpha}_{t'}^i x_{t'}^i + \bar{\beta}_{t'}^i y_{t'}^i) + \sum_i \bar{\gamma}^i s^i + \bar{\eta} \geq 0 \quad (22)$$

is a valid inequality for $\text{conv}(X^{2PL})$ that cuts off $(\bar{x}, \bar{y}, \bar{s})$.

Proof.

Using a similar approach to the previous proof and also using the strong duality theorem for QP. □

Defining 2-Period Subproblems

- **Question 1:** On which two periods to run the separation?
 - We can look at all the 2-period problems ($NT - 1$ of them)
- **Question 2:** Which period's stock is represented by s^i ?
 - Let $\phi(i) \in [t + 1, \dots, NT]$ be the horizon parameter for each i
 - Obvious choice: $t + 1$, i.e., $s^i = s_{t+1}^i$
 - Then, parameters are defined as follows ($\forall i, t' = 1, 2$):
 - $\tilde{M}_{t'}^i = M_{t+t'-1}^i, \tilde{C}_{t'}^i = C_{t+t'-1}^i$
 - $\tilde{d}_{t'}^i = d_{t+t'-1, t+1}^i$, i.e., $\tilde{d}_1^i = d_{12}^i$ and $\tilde{d}_2^i = d_2^i$.
- **Observation 1:** If a number of periods following $t + 1$ have no setups, their demands should be incorporated
- **Observation 2:** If a setup occurs after $t + 1$, (ℓ, S) inequalities will be weakened if that period's demand is in

2-Period Convex Hull Closure Framework

- Following Miller, Nemhauser, Savelsbergh (2000)

$$\phi(i) = \max\{t' \mid t' \geq t + 1, \sum_{t''=t+1}^{t'} y_{t''}^i \leq y_{t+1}^i + \Theta\}$$

where $\Theta \in (0, 1]$ is a random number

- Let X_t^{2PL} be $X_t^{2PL}(\phi(1), \phi(2), \dots, \phi(NI))$

Solve LPR of the original problem

→ $(\bar{x}, \bar{y}, \bar{s})$

for $t=1$ **to** $NT-1$

Define X_t^{2PL}

Apply 2-period convex hull separation algorithm

Computational Results: 2-Period Problems

- *2PCLS* instances: 20 problems with two periods only and with two to six items
 - **cdd** might provide the full description of the convex hull
 - Generate all the extreme points and rays of the LPR
 - Eliminate all the fractional extreme points
 - Using these integral extreme points, generate all the facets of the integral polyhedron
 - The more items share a resource, the more the structure tends to resemble that of an uncapacitated problem
- The separation algorithm implemented in Mosel (Mosel version 2.0.0, Xpress 2007 package)

Computational Results: 2-Period Problems (cont'd)

Inst.	NI	XLP	IP	# cuts (\mathcal{L}_1)	# cuts (\mathcal{L}_∞)	# cuts (\mathcal{L}_2)
2pcls01	3	17.033	25	11	8	19
2pcls02	3	12.6253	19	13	7	7
2pcls03	3	76.5345	104	5	3	1
2pcls04	2	14.7674	19	4	2	1
2pcls05	3	38.39	52	8	6	4
2pcls06	3	117.375	173	5	6	5
2pcls07	2	36.5	43	2	1	1
2pcls08	2	21.45	26	7	2	2
2pcls09	2	129	153	2	3	3
2pcls10	3	17.6539	24	1	3	1

Computational Results: 2-Period Problems (cont'd)

Inst.	NI	XLP	IP	# cuts (\mathcal{L}_1)	# cuts (\mathcal{L}_∞)	# cuts (\mathcal{L}_2)
2pcls11	3	71.7209	102	4	1	1
2pcls12	3	46.68	69	4	1	2
2pcls13	4	85.6256	113	7	7	9
2pcls14	4	70.2961	81	6	8	5
2pcls15	4	54.1848	74	6	3	1
2pcls16	4	34.0844	39	6	7	4
2pcls17	5	164.858	211	39	19	14
2pcls18	5	57.0825	97	34	10	6
2pcls19	6	115.131	150	11	6	1
2pcls20	6	59.2412	89	34	11	5

Computational Results: Multi-Period Problems

tr6-15 detailed results (30 iterations):

	\mathcal{L}_1		\mathcal{L}_2		
	lim=100 $\phi(i)$	lim=150 $\phi(i)$	lim=100 $t + 1$	lim=150 $t + 1$	lim=150 $\phi(i)$
2PL	37,234.6	37,298.8	37,306.7	37,306.8	37,331.2

Trigeiro instances results with Euclidean approach:

	tr6-15	tr6-30	tr12-15	tr12-30
XLP	37,201	60,946	73,848	130,177
2PL	37,344	61,090	73,932	130,215
OPT	37,721	61,746	74,634	130,596

Computational Results: Multi-Period Problems (cont'd)

TDS instances results with \mathcal{L}_∞ approach:

	AK501131	AK501132	AK501141	AK502131	AK502132
LB (I,S)	96,968	101,699	134,805	93,369	96,312
LB (2PL)	114,788	108,846	150,954	107,477	110,542
Best Sol.	123,366	123,473	170,897	115,819	118,319
Gap was	27.22%	21.41%	26.77%	24.04%	22.85%
Gap now	7.47%	13.44%	13.21%	7.76%	7.04%

	BK511131	BK511141	BK521131	BK521142	BK522142
LB (I,S)	92,602	125,307	92,350	124,988	119,559
LB (2PL)	101,565	134,997	106,302	139,090	131,552
Best Sol.	120,303	172,762	118,217	154,258	148,471
Gap was	29.91%	37.87%	28.01%	23.42%	24.18%
Gap now	18.45%	27.97%	11.21%	10.91%	12.86%

Conclusions

- Study of a stronger relaxation
 - A new framework independent from defining families of valid inequalities or reformulations a priori, although expected output is to define new valid inequalities using the results from the framework
 - To our knowledge, this is an original approach in production planning literature
- Different norms useful to generate cuts and improve lower bounds significantly
 - Euclidean and \mathcal{L}_∞ approaches computationally much more efficient than Manhattan approach
 - Observed both on the efficiency of cuts and on the number of extreme points generated in column generation before termination

Future Directions

- Immediate directions
 - Achieving more computational results on realistic size problems
 - Applying lifting on the generated inequalities
 - Possibility to obtain facets instead of faces
- Near-future directions
 - Careful analysis of the inequalities generated by the framework and the facets from **cdd**
 - Significant potential to identify new families of valid inequalities
- Possibilities for extending results to other MIP techniques
 - *Example:* Disjunctive cuts
 - Normalization or Manhattan distance used by researchers
 - Euclidean has potential to provide more efficient cuts
- Extension to other MIP problems possible
 - “Local Cuts” of Cook, Espinoza and Chvátal [2006]