

620-362
Applied Operations Research
Mid-Semester Test Revision

Department of Mathematics and Statistics
The University of Melbourne

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Some contents of this presentation are adapted from year 2005 course notes for 620-362 Applied Operations Research, Department of Mathematics and Statistics, The University of Melbourne (compiled by Prof Natasha Boland and Dr Renata Sotirov)

Overview: What to revise?

- Mixed integer program modelling
 - Piecewise linear functions
 - Big-M constraints
 - Other “creative” use of binary/integer variables
 - “Good” and “bad” models
- Branch-and-bound
 - The branch-and-bound procedure.
 - Use graphical method to solve LP.
- The Gomory cutting plane method
 - Proofs
 - Derive Gomory cuts
 - The dual simplex method
- Column generation & Dantzig-Wolfe Reformulation
 - Pricing subproblem
 - D-W reformulation
- Modelling quadratic models
 - Determine covariance matrix
 - Model problem

MIP Modelling

Integer Programming Modelling

- Knapsack Problems (Capital Budgeting)
- Set Covering/Partitioning/Packing (Human resource planning)
- Facility Location Problems
- Lot-sizing Problems
- Logic and Disjunctive Constraints
- Piecewise Linear Functions
- Quadratic Assignment Problems
- Hub Location Problems (Postal Service Planning)
- Vehicle Routing/Travelling Salesperson

Modelling Logic And Disjunctive Constraints

Logical statements, such as implications, can be modelled using binary variables.

Facility location problem: no more than 10 new facilities to be constructed

$$\sum_{i=1}^m y_i \leq 10$$

Can put facilities at both sites 1 and 2 or at neither

$$y_1 = y_2$$

Cannot put facilities at both sites 3 and 4

$$y_3 + y_4 \leq 1$$

If put facility at site 5, must also put facility at site 6

$$y_5 \leq y_6$$

Disjunctive Inequalities

Machine maintenance scheduling:

- Schedule time t to perform maintenance on machine
- Machine will be in use between time T_1 and T_2
- Deadline on maintenance is time T

Disjunctive constraint:

$$t \leq T_1 \quad \text{or} \quad t \geq T_2$$

$$y = \begin{cases} 1, & t \leq T_1 \text{ is enforced} \\ 0, & t \geq T_2 \text{ is enforced} \end{cases}$$

$$t \leq T_1 + (T - T_1)(1 - y)$$

$$t \geq T_2 - T_2 y$$

Modelling Logical Relations

Select at most one (packing): $\sum_{j \in S} x_j \leq 1$

Select precisely one (partitioning): $\sum_{j \in S} x_j = 1$

Select at least one (covering): $\sum_{j \in S} x_j \geq 1$

Special conditions:

$$x_1 = 1 \Rightarrow x_2 = 1$$

$$x_1 - x_2 \leq 0$$

$$x_1 = 1 \Rightarrow x_2 = 0$$

$$x_1 + x_2 \leq 1$$

$$x_1 = 1 \wedge x_2 = 1 \Rightarrow x_3 = 1$$

$$x_1 + x_2 - x_3 \leq 1$$

Production Scheduling

$$\min \max_{t=2,\dots,12} |l_t - l_{t-1}|$$

$$s.t. \quad l_t = b_1 x_{1t} + b_2 x_{2t}, \quad \forall t = 1, \dots, 12$$

$$y_1 = x_{11} + x_{21} + I_0 - d_1$$

$$y_t = x_{1t} + x_{2t} + y_{t-1} - d_t, \quad \forall t = 2, \dots, 12$$

$$0 \leq x_{1t} \leq a_1, \quad \forall t = 1, \dots, 12$$

$$0 \leq x_{2t} \leq a_2, \quad \forall t = 1, \dots, 12$$

$$0 \leq l_t \leq L_t, \quad \forall t = 1, \dots, 12$$

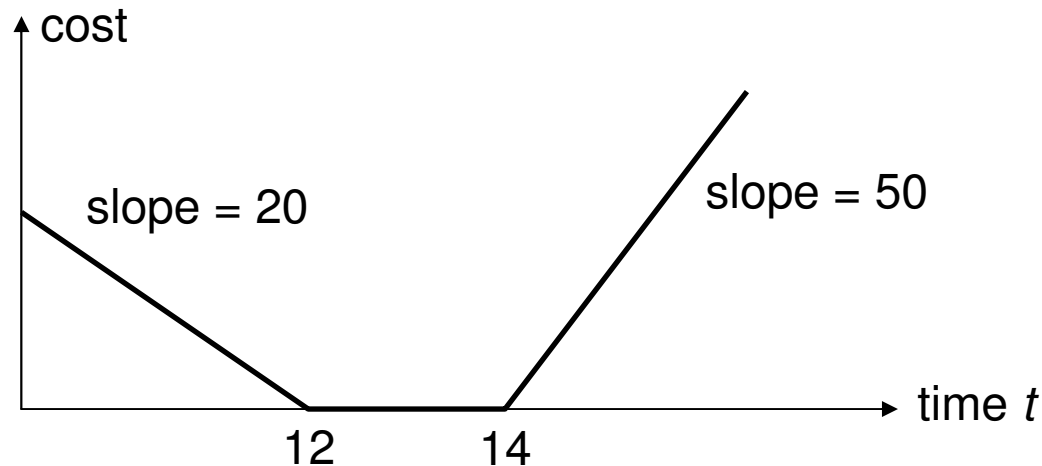
$$y_t \geq 0, \quad \forall t = 1, \dots, 12$$

$$\left\{ \begin{array}{ll} \min & m \\ s.t. & m \geq |l_t - l_{t-1}|, \quad \forall t = 2, \dots, 12 \\ & l_t = b_1 x_{1t} + b_2 x_{2t}, \quad \forall t = 1, \dots, 12 \\ & \vdots \\ & etc \end{array} \right.$$

$$\left\{ \begin{array}{ll} \min & m \\ s.t. & m \geq l_t - l_{t-1} \\ & m \geq l_{t-1} - l_t \end{array} \right\}, \quad \forall t = 2, \dots, 12$$

$$\left\{ \begin{array}{ll} & l_t = b_1 x_{1t} + b_2 x_{2t}, \quad \forall t = 1, \dots, 12 \\ & \vdots \\ & etc \end{array} \right.$$

Penalties on Soft Constraints



To model as an LP, introduce two new variables

$y = \# \text{ hours early}$

$z = \# \text{ hours late}$

via

$y \geq 12 - t, y \geq 0$

$z \geq t - 14, z \geq 0$

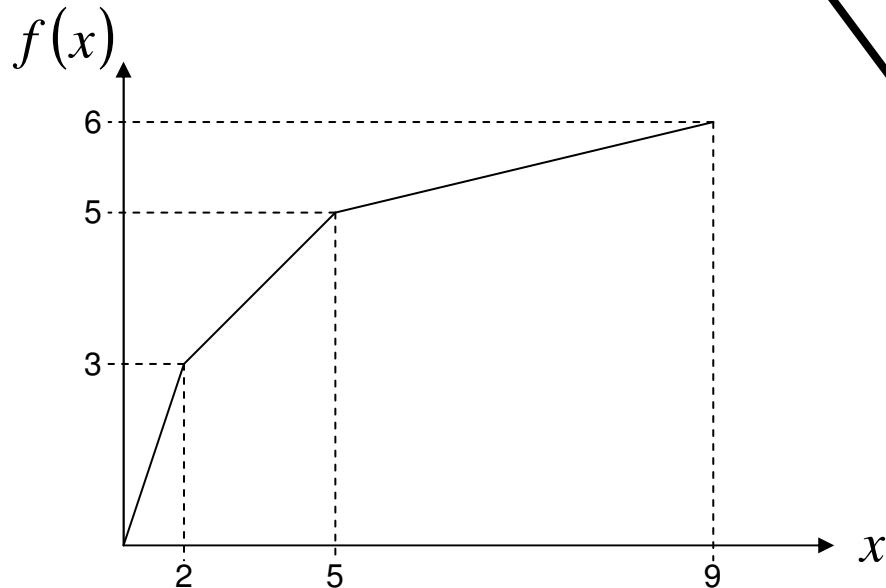
with objective function

$\min \dots + 20y + 50z + \dots$

Max. a Piecewise Linear Concave Function

Example:

$$\left\{ \begin{array}{l} \max \quad f(x_1) + 2x_2 \\ \text{s.t.} \quad x_1 + 3x_2 \leq 15 \\ \quad \quad x_1 + x_2 \leq 9 \\ \quad \quad x_1, x_2 \geq 0 \end{array} \right. \quad \text{where } f(x) = \begin{cases} \frac{3}{2}x, & 0 \leq x \leq 2 \\ \frac{1}{3}(2x+5), & 2 < x \leq 5 \\ \frac{1}{4}(x+15), & 5 < x \leq 9 \end{cases}$$



$$\left\{ \begin{array}{l} \max \quad \frac{3}{2}y_1 + \frac{2}{3}y_2 + \frac{1}{4}y_3 + 2x_2 \\ \text{s.t.} \quad y_1 + y_2 + y_3 + 3x_2 \leq 15 \\ \quad \quad y_1 + y_2 + y_3 + x_2 \leq 9 \\ \quad \quad 0 \leq y_1 \leq 2 \\ \quad \quad 0 \leq y_2 \leq 3 \\ \quad \quad 0 \leq y_3 \leq 4 \\ \quad \quad x_2 \geq 0 \end{array} \right.$$

Example: Piecewise Linear Function

To complete the MIP model of the piecewise linear cost function c , we note that c can be expressed as

$$\begin{aligned}c(x) &= \sum_{k=1}^4 \lambda_k c(t_k) \\ &= 0\lambda_1 + 9000\lambda_2 + 16200\lambda_3 + 22200\lambda_4 \\ &= 9000\lambda_2 + 16200\lambda_3 + 22200\lambda_4\end{aligned}$$

$$c(x) = \begin{cases} 15x, & 0 \leq x \leq 600 \\ 4200 + 8x & 600 \leq x \leq 1500 \\ 20x - 13800 & 1500 \leq x \leq 1800 \end{cases}$$

s.t.

$$\lambda_1 \leq y_1$$

$$\lambda_2 \leq y_1 + y_2$$

$$\lambda_3 \leq y_2 + y_3$$

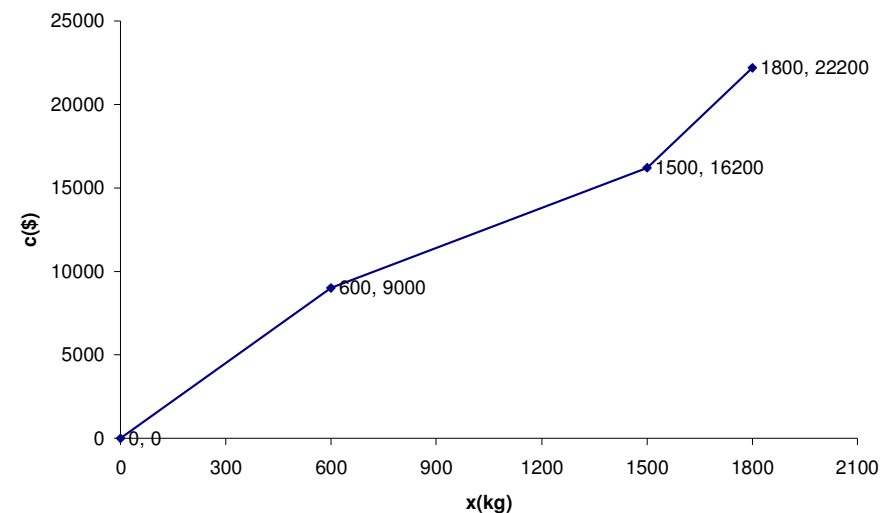
$$\lambda_4 \leq y_3$$

$$y_1 + y_2 + y_3 = 1$$

$$\lambda_1 + \lambda_2 + \lambda_3 + \lambda_4 = 1$$

$$\lambda_i \geq 0, \quad \forall i = 1 \dots 4$$

$$y_i \text{ binary}, \quad \forall i = 1 \dots 3$$



Linearising the QAP Model

$$x_{ij} = \begin{cases} 1, & \text{flight } i \text{ is assigned to gate } k \\ 0, & \text{otherwise} \end{cases}$$

$$y_{ijkl} = \begin{cases} 1, & i \text{ assigned to } k \text{ and } j \text{ to } l \\ 0, & \text{otherwise} \end{cases}$$

$$\min \sum_{i \in F} \sum_{j \in F} \sum_{k \in G} \sum_{l \in G} t_{ij} d_{kl} x_{ik} x_{jl}$$

s.t.

$$\sum_{k \in G} x_{ik} = 1, \quad \forall i \in F$$

$$\sum_{k \in F} x_{ik} = 1, \quad \forall k \in G$$

$$x_{ij} \in \{0,1\}, \quad \forall i \in F, k \in G$$

$$\min \sum_i \sum_j \sum_k \sum_l t_{ij} d_{kl} y_{ijkl}$$

s.t.

$$\sum_i \sum_j \sum_k \sum_l y_{ijkl} = n^2$$

$$x_{ik} + x_{jl} \geq 2y_{ijkl}, \quad \forall i, j, k, l$$

$$y_{ijkl} \geq x_{ik} + x_{jl} - 1$$

Project Management

$$\min \sum_{i \in S} C_i x_i + \sum_{k \in H} \sum_{i \in J_k} W_k l_{ik}$$

s.t.

$$t_i + D_i - x_i \leq T, \quad \forall i \in S$$

$$t_i \geq t_j + D_j - x_j, \quad \forall i \in S, j \in P_i$$

$$x_i \leq D_i, \quad \forall i \in S$$

$$\sum_{\substack{k \in H \\ \text{s.t. } i \in J_k}} y_{ik} = 1, \quad \forall i \in S$$

$$l_{ik} \geq D_i - x_i - M(1 - y_{ik}), \quad \forall k \in H, i \in J_k$$

$$t_i, x_i \geq 0, \quad \forall i \in S$$

$$y_{ik} \in \{0,1\}, l_{ik} \geq 0, \quad \forall k \in H, i \in J_k$$

Employee shift covering

$$\min \sum_{q \in S} \sum_{t=1}^T C_t x_{tq}$$

s.t.

$$n_t = \sum_{q \in S} \sum_{r=t-L_q+1}^t x_{rq}, \quad \forall t = 1, \dots, T$$

$$D_t \leq n_t \leq W, \quad \forall t = 1, \dots, T$$

$$\sum_{q \in S} \sum_{t=1}^T L_q x_{tq} \leq H \leftarrow \text{What is the use of this constraint?}$$

A Formal Definition

- Model M_1 is at least as good as model M_2 if

$$R_1 \subseteq R_2$$

- Model M_1 is better than model M_2 if

$$R_1 \subset R_2$$

- Model M_1 is better than model M_2 with respect to objective $\mathbf{c}x$ if

$$\max_{x \in R_1} \mathbf{c}x < \max_{x \in R_2} \mathbf{c}x$$

Uncapacitated Facility Location (UFL)

$$\begin{aligned} \min \quad & \sum_{i \in I, j \in J} c_{ij} y_{ij} + \sum_{i \in I} f_i x_i \\ \text{s.t.} \quad & \sum_{i \in I} y_{ij} = 1, \quad \forall j \in J \\ & \sum_{j \in J} y_{ij} \leq m x_i, \quad \forall i \in I \\ & y_{ij} \geq 0, \quad \forall i \in I, j \in J \\ & x_i \in \{0,1\}, \quad \forall i \in I \end{aligned}$$

VS

$$\begin{aligned} \min \quad & \sum_{i \in I, j \in J} c_{ij} y_{ij} + \sum_{i \in I} f_i x_i \\ \text{s.t.} \quad & \sum_{i \in I} y_{ij} = 1, \quad \forall j \in J \\ & y_{ij} \leq x_i, \quad \forall i \in I, j \in J \\ & y_{ij} \geq 0, \quad \forall i \in I, j \in J \\ & x_i \in \{0,1\}, \quad \forall i \in I \end{aligned}$$

Lot-Sizing Problem

$$\min \sum_{t=1}^T (p_t y_t + h_t s_t + c_t x_t)$$

s.t.

$$y_1 = d_1 + s_1$$

$$s_{t-1} + y_t = d_t + s_t, \quad t = 2, \dots, T-1$$

$$s_{T-1} + y_T = d_T$$

$$y_t \leq Mx_t, \quad t = 1, \dots, T$$

$$y_t, s_t \geq 0, x_t \in \{0,1\}, \quad t = 1, \dots, T$$

VS

$$\min \sum_{t=1}^T \left[\left[\sum_{i=1}^t \left(p_i + \sum_{k=i}^{t-1} h_k \right) q_{it} \right] + c_t x_t \right]$$

s.t.

$$\sum_{i=1}^T q_{it} = d_t, \quad t = 1, \dots, T$$

$$q_{it} \leq d_t x_i, \quad i = 1, \dots, T, t = 1, \dots, T$$

$$q_{it} \geq 0, \quad i = 1, \dots, T, t = 1, \dots, T$$

$$x_i \in \{0,1\}, \quad i = 1, \dots, T$$

$$M = \sum_{k=1}^T d_k$$

A tighter bound:

$$M_t = \sum_{k=t}^T d_k$$

HRP – Assignment Model

$$\begin{aligned} \min \quad & \sum_{j=1}^m w_j \\ \text{s.t.} \quad & \\ & \sum_{j=1}^m z_{ij} = 1, \quad \forall i = 1, \dots, n \\ & \sum_{i=1}^n h_i z_{ij} \leq H w_j, \quad \forall j = 1, \dots, m \\ & z_{ij} \in \{0,1\}, \quad \forall i = 1, \dots, n; j = 1, \dots, m \\ & w_j \in \{0,1\}, \quad \forall j = 1, \dots, m \end{aligned}$$

VS

$$\begin{aligned} \min \quad & \sum_{s \in J} x_s \\ \text{s.t.} \quad & \\ & \sum_{s \in J: i \in S} x_s = 1, \quad \forall i = 1, \dots, n \\ & x_s \in \{0,1\}, \quad \forall s \in J \end{aligned}$$

...might have other restrictions, e.g. some jobs cannot be done by the same worker.

Any problems?

SYMMETRY – every worker is identical

(May also have symmetry if several jobs require same length of time.)

HRP – Assignment Model

How can we break symmetry?

One way is to include the following set of constraints:

$$\sum_{i=1}^n z_{i(j-1)} \geq \sum_{i=1}^n z_{ij}, \quad \forall j = 2, \dots, m$$

Considers only solutions where the number of jobs assigned to a worker is non-decreasing with increasing worker “index”.

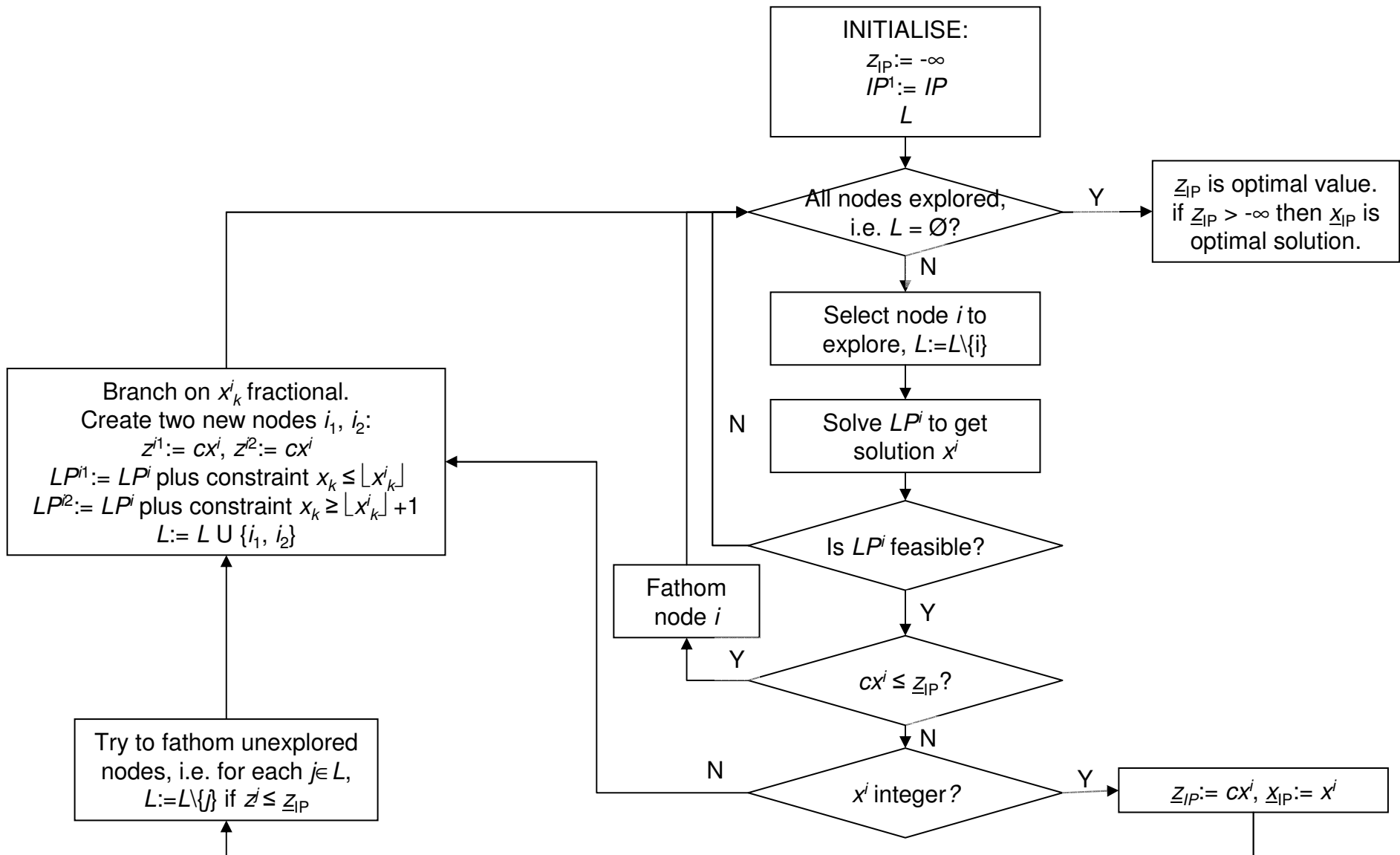
- This is reasonable since any feasible solution obtained for this problem can be arranged in this order.

Also...

review other MIP modelling
problems (lectures and practice
classes)

The Branch-and-Bound Method

Branch-and-Bound Algorithm (Max. prob.)

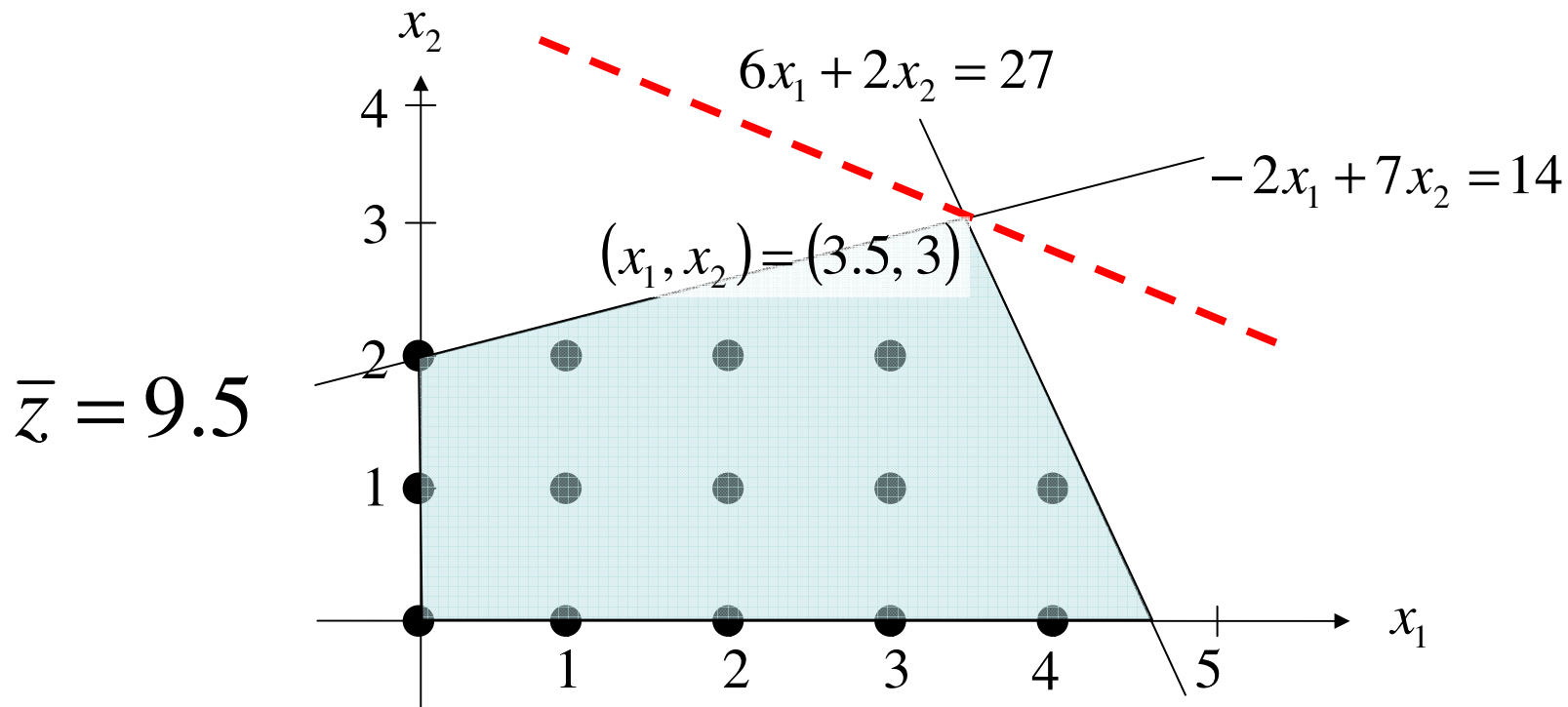


LP-based Branch-and-Bound

Example 1

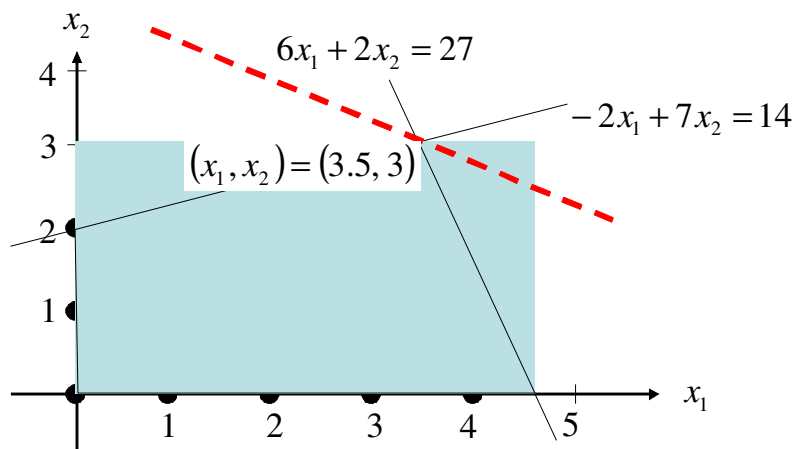
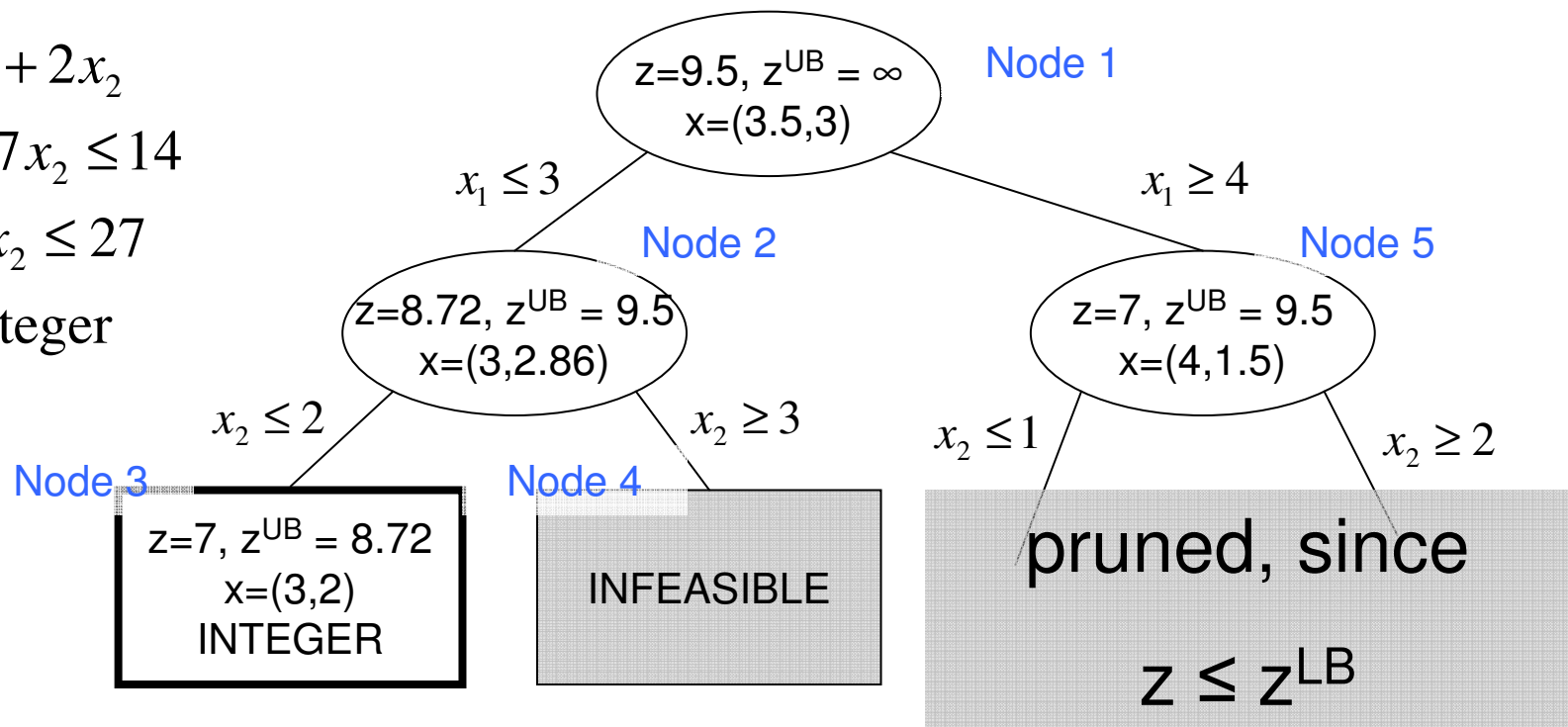
$$\begin{aligned} \max z &= x_1 + 2x_2 \\ \text{s.t. } -2x_1 + 7x_2 &\leq 14 \\ 6x_1 + 2x_2 &\leq 27 \\ x_1, x_2 &\text{ integer} \end{aligned}$$

Best objective = *NONE*



Example 1: B&B Tree

$$\begin{aligned} \max z &= x_1 + 2x_2 \\ \text{s.t. } -2x_1 + 7x_2 &\leq 14 \\ 6x_1 + 2x_2 &\leq 27 \\ x_1, x_2 &\text{ integer} \end{aligned}$$



z^{LB}
initial: $-\infty$
update: 7

The Gomory Cutting Plane Method

The Gomory Cutting Plane Method

Proposition 1.

If $\alpha x \leq \beta$ is valid for P then

$\sum_{i=1}^n \lfloor \alpha_i \rfloor x_i \leq \lfloor \beta \rfloor$ is also valid for P .

The Chavátal-Gomory Procedure

...is determined by the following proposition

Proposition 2.

Let $P = \{x \in Z^n : Ax \leq b, x \geq 0\}$, where $A = (a_{ij}) \in \mathfrak{R}^{m \times n}$, and $u \in \mathfrak{R}_+^m$.

Then

$$\sum_{j=1}^n \left\lfloor \sum_{i=1}^m u_i a_{ij} \right\rfloor x_j \leq \left\lfloor \sum_{i=1}^m u_i b_i \right\rfloor$$

or equivalently

$$\sum_{j=1}^n \left\lfloor u^T a_j \right\rfloor x_j \leq \left\lfloor u^T b \right\rfloor$$

where a_j is j^{th} column of A , is valid constraint for P .

Proposition 3

If $\alpha x = \beta$ is valid for P then

$$\sum_{i=1}^n (\alpha_i - \lfloor \alpha_i \rfloor) x_i \geq \beta - \lfloor \beta \rfloor$$

is also valid for P .

Proof. $\alpha x \leq \beta$ is obviously valid, so by Proposition 1,

$$\sum_{i=1}^n \lfloor \alpha_i \rfloor x_i \leq \lfloor \beta \rfloor, \quad \forall x \in P$$

$$\Leftrightarrow \sum_{i=1}^n \alpha_i x_i - \sum_{i=1}^n \lfloor \alpha_i \rfloor x_i \geq \beta - \lfloor \beta \rfloor, \quad \forall x \in P$$

$$\Leftrightarrow \sum_{i=1}^n (\alpha_i - \lfloor \alpha_i \rfloor) x_i \geq \beta - \lfloor \beta \rfloor, \quad \forall x \in P$$

The Dual Simplex Method

Initial tableau:

| <i>Eqn</i> | x_N | x_B | <i>RHS</i> |
|------------|-------|-------|------------|
| 1 | | | |
| \vdots | y | I | r |
| m | | | |
| z | y_0 | 0 | r_0 |

Assumption: $y_0 \geq 0$ (if max. problem); $y_0 \leq 0$ (if min. problem)

1. If $r \geq 0$ the STOP: basic feasible solution is optimal.
Else select row i such that $r_i = \min_j \{r_j\}$ (i.e. select row with the most negative RHS).
2. If $y_{ij} \geq 0$ for all j then STOP: dual is unbounded, so primal is infeasible.
Else choose $j \in N$ such that $j = \arg \min \{|y_{0j}/y_{ij}| : j \in N, y_{ij} < 0\}$
3. Pivot so that j enters the basis and i leaves the basis. Go to Step 1.

Example 3 (repeat example)

$$\max \quad z = 7x_1 + 9x_2$$

$$s.t. \quad -x_1 + 3x_2 \leq 6$$

$$7x_1 + x_2 \leq 35$$

$$x_1, x_2 \in Z_+$$

Final tableau after Simplex method:

| eqn | x1 | x2 | s1 | s2 | RHS | Ratio test |
|-----|----|----|--------|--------|-------|--------------|
| 1 | 0 | 1 | 7/22 | 1/22 | 3 1/2 | R1=R1+R2'/3 |
| 2 | 1 | 0 | - 1/22 | 3/22 | 4 1/2 | R2'=R2*3/22 |
| z | 0 | 0 | 2 6/11 | 1 4/11 | 63 | Rz=Rz+10*R2' |

Gomory cut due to row 1 is $7/22s_1 + 1/22s_2 \geq 1/2$

Example 3 (repeat example)

Add cut $7/22s_1 + 1/22s_2 \geq 1/2$ to tableau.

Applying dual simplex to the new tableau:

| eqn | x1 | x2 | s1 | s2 | s3 | RHS |
|------------|----|----|--------|--------|----|-------|
| 1 | 0 | 1 | 7/22 | 1/22 | 0 | 3 1/2 |
| 2 | 1 | 0 | - 1/22 | 3/22 | 0 | 4 1/2 |
| 3 | 0 | 0 | - 7/22 | - 1/22 | 1 | - 1/2 |
| z | 0 | 0 | 2 6/11 | 1 4/11 | 0 | 63 |
| Ratio Test | | | 8 | 30 | | |

| eqn | x1 | x2 | s1 | s2 | s3 | RHS |
|------------|----|----|----|-----|--------|-------|
| 1 | 0 | 1 | 0 | 0 | 1 | 3 |
| 2 | 1 | 0 | 0 | 1/7 | - 1/7 | 4 4/7 |
| 3 | 0 | 0 | 1 | 1/7 | -3 1/7 | 1 4/7 |
| z | 0 | 0 | 0 | 1 | 8 | 59 |
| Ratio Test | | | | | | |

$R1=R1-R3'*7/22$
 $R2=R2-R3'/22$
 $R3'=R3*-22/7$
 $Rz=Rz-R3'28/11$

Optimal solution: $x_1 = 32/7$, $x_2 = 3$, $z = 59$

Gomory cut due to row 2 is $1/7s_2 + 6/7s_3 \geq 4/7$

Example 3 (repeat example)

Add cut $1/7s_2 + 6/7s_3 \geq 4/7$ to tableau.

Applying dual simplex to the new tableau:

| eqn | x1 | x2 | s1 | s2 | s3 | s4 | RHS |
|------------|----|----|----|-------|-----------|----|-------|
| 1 | 0 | 1 | 0 | 0 | 1 | 0 | 3 |
| 2 | 1 | 0 | 0 | 1/7 | - 1/7 | 0 | 4 4/7 |
| 3 | 0 | 0 | 1 | 1/7 | -3 1/7 | 0 | 1 4/7 |
| 4 | 0 | 0 | 0 | - 1/7 | - 6/7 | 1 | - 4/7 |
| z | 0 | 0 | 0 | 1 | 8 | 0 | 59 |
| Ratio Test | | | | 7 | 9.3333333 | | |

| eqn | x1 | x2 | s1 | s2 | s3 | s4 | RHS | |
|-----|----|----|----|----|----|----|-----|---------------|
| 1 | 0 | 1 | 0 | 0 | 1 | 0 | 3 | R1=R1 |
| 2 | 1 | 0 | 0 | 0 | -1 | 1 | 4 | R2=R2-1/7*R4' |
| 3 | 0 | 0 | 1 | 0 | -4 | 1 | 1 | R3=R3-1/7*R4' |
| 4 | 0 | 0 | 0 | 1 | 6 | -7 | 4 | R4'=R4*-7 |
| z | 0 | 0 | 0 | 0 | 2 | 7 | 55 | Rz=Rz-R4' |

Optimal solution: $x_1 = 4$, $x_2 = 3$, $z = 55$ (integer!!)

Column Generation and Dantzig-Wolfe Decomposition

Algorithm to solve the Master LP

Step 0:

Choose an initial set of columns A^s .

Step 1:

Solve the Restricted LP to obtain optimal basic solution y and corresponding duals u .

Step 2:

Note that x given by

$$x_i = y_i, \text{ if column } i \text{ is in } A^s; 0, \text{ otherwise}$$

is a basic feasible point for the Master LP and is optimal if
 $uA \leq c$.

Solve column generation subproblem:

$$z^{\text{sub}} = \min\{c_i - ua_i; a_i \text{ is a column of } A\}$$

If $z^{\text{sub}} \geq 0$ then STOP: x is optimal for Master LP.

Otherwise add a_i such that $c_i - ua_i = z^{\text{sub}}$ to A^s and go to Step 1.

Human Resource Planning (HRP)

Tasks to be performed: 1, 2, ..., n

h_i = time required for task i (hours)

H = max number of hours worker can work in a day

What is the minimum workforce required?

Master IP:

$$z = \min \sum_{s \in J'} x_s$$

s.t.

$$\sum_{s \in J': i \in s} x_s = 1, \quad \forall i = 1, \dots, n$$

$$x_s \in \{0,1\}, \quad \forall s \in J'$$

Subproblem:

$$z^{sub} = 1 - \left\{ \begin{array}{l} \max \sum_{i=1}^n u_i y_i \\ \text{s.t.} \\ \sum_{i=1}^n h_i y_i \leq H \\ y_i \in \{0,1\}, \quad \forall i = 1, \dots, n \end{array} \right.$$

DW Decomposition...

$$z = \max \quad \mathbf{C}^1 \mathbf{x}^1 + \mathbf{C}^2 \mathbf{x}^2$$

s.t.

$$\mathbf{D}^1 \mathbf{x}^1 \leq \mathbf{d}^1$$

$$\mathbf{D}^2 \mathbf{x}^2 \leq \mathbf{d}^2$$

$$\mathbf{A}^1 \mathbf{x}^1 + \mathbf{A}^2 \mathbf{x}^2 \leq \mathbf{b}$$

\equiv

$$z = \max \quad \mathbf{C}^1 \left(\sum_{t=1}^{|\mathbf{T}_1|} \lambda_t^1 \mathbf{x}_t^1 \right) + \mathbf{C}^2 \left(\sum_{t=1}^{|\mathbf{T}_2|} \lambda_t^2 \mathbf{x}_t^2 \right)$$

s.t.

$$\sum_{t=1}^{|\mathbf{T}_1|} \lambda_t^1 = 1$$

$$\sum_{t=1}^{|\mathbf{T}_2|} \lambda_t^2 = 1$$

$$\mathbf{A}^1 \left(\sum_{t=1}^{|\mathbf{T}_1|} \lambda_t^1 \mathbf{x}_t^1 \right) + \mathbf{A}^2 \left(\sum_{t=1}^{|\mathbf{T}_2|} \lambda_t^2 \mathbf{x}_t^2 \right) \leq \mathbf{b}$$

$$\lambda_t^1 \geq 0, \quad t = 1, \dots, |\mathbf{T}_1|$$

$$\lambda_t^2 \geq 0, \quad t = 1, \dots, |\mathbf{T}_2|$$

Pricing subproblems...

Restricted Master LP

$$z = \max \quad \mathbf{C}^1 \left(\sum_{t=1}^{|\hat{\mathbf{T}}_1|} \lambda_t^1 \mathbf{x}_t^1 \right) + \mathbf{C}^2 \left(\sum_{t=1}^{|\hat{\mathbf{T}}_2|} \lambda_t^2 \mathbf{x}_t^2 \right)$$

s.t.

$$\sum_{t=1}^{|\hat{\mathbf{T}}_1|} \lambda_t^1 = 1 \quad (\text{dual} : \pi_1)$$

$$\sum_{t=1}^{|\hat{\mathbf{T}}_2|} \lambda_t^2 = 1 \quad (\text{dual} : \pi_2)$$

$$\mathbf{A}^1 \left(\sum_{t=1}^{|\hat{\mathbf{T}}_1|} \lambda_t^1 \mathbf{x}_t^1 \right) + \mathbf{A}^2 \left(\sum_{t=1}^{|\hat{\mathbf{T}}_2|} \lambda_t^2 \mathbf{x}_t^2 \right) \leq \mathbf{b} \quad (\text{dual} : \rho)$$

$$\lambda_t^1 \geq 0, \quad t = 1, \dots, |\hat{\mathbf{T}}_1|$$

$$\lambda_t^2 \geq 0, \quad t = 1, \dots, |\hat{\mathbf{T}}_2|$$

Pricing subproblem for \mathbf{x}^1

$$\max \quad z_1 = \mathbf{C}^1 \mathbf{x}^1 - \rho \mathbf{A}^1 \mathbf{x}^1 - \pi_1$$

s.t.

$$\mathbf{D}^1 \mathbf{x}^1 \leq \mathbf{d}^1$$

$$\mathbf{x}^1 \geq 0$$

Pricing subproblem for \mathbf{x}^2

$$\max \quad z_2 = \mathbf{C}^2 \mathbf{x}^2 - \rho \mathbf{A}^2 \mathbf{x}^2 - \pi_2$$

s.t.

$$\mathbf{D}^2 \mathbf{x}^2 \leq \mathbf{d}^2$$

$$\mathbf{x}^2 \geq 0$$

Quadratic Models

Portfolio Optimisation: Xpress^{MP} Example

An investor is evaluating ten different securities ('shares').

He estimates the return on investment for a period of one year.

He further wishes to invest at least half of his capital in North-American shares and at most a third in shares.

How should the capital be divided among the shares to minimize the risk whilst obtaining a certain target yield?

The investor adopts the **Markowitz** idea of getting estimates of the variance/covariance matrix of estimated returns on the securities.

Portfolio Optimisation: Xpress^{MP} Example

Given

I^{\max} = max. investment per share (%)

I^{\min} = min. investment into North American share (%)

S = set of all shares

A = set of North American shares

T = target yield (%)

R_p = estimated return in investment for share p (%)

Cov = covariance matrix of estimated return

Variable:

x_p = fraction of capital invested in share p

Small Example

Budget = \$2,000 (spend all)

Minimum return required = 5%

$$\left\{ \begin{array}{l} \min \quad (x_1 \quad x_2) \begin{bmatrix} 26.25 & 2.1 \\ 2.1 & 36.04 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ s.t. \\ 4.5x_1 + 5.6x_2 \geq 10000 \\ x_1 + x_2 = 2000 \\ x_1, x_2 \geq 0 \end{array} \right.$$

$$\frac{(4.5x_1 + 5.6x_2) / 100}{2000} \geq \frac{5}{100}$$

Portfolio Optimisation: Xpress^{MP} Example

$$\min \sum_{p \in S} \sum_{q \in S} \text{cov}(p, q) x_p x_q$$

s.t.

$$\sum_{p \in A} x_p \geq I^{\min}$$

$$\sum_{p \in S} x_p = 1$$

$$x_p \leq I^{\max}, \quad \forall p \in S$$

$$\sum_{p \in S} R_p x_p \geq T$$

$$x_p \geq 0, \quad \forall p \in S$$