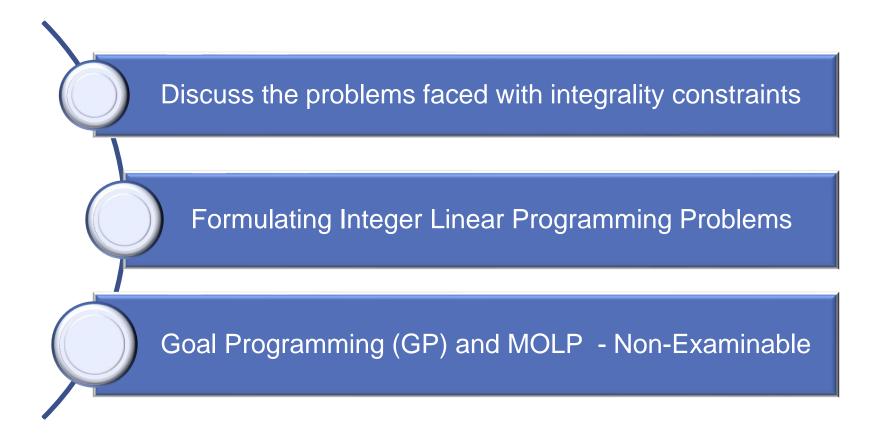
FIT3158 Business Decision Modelling

SEMESTER 2, 2022

Lecture 4

Integer Linear Programming (ILP)

Topics Covered:





Introduction

- Integer Linear Programming (ILP)
 - When one or more variables in an LP problem must assume an integer value
- ILPs occur frequently...
 - Scheduling workers
 - Manufacturing products
- Integer variables also allow us to build more accurate models for a number of common business problems.
 - Quantity discounts
 - Setup and lump sum costs
 - Batch size restrictions



Integrality Conditions

MAX:
$$350X_1 + 300X_2$$
 } profit
S.T.: $1X_1 + 1X_2 <= 200$ } pumps
 $9X_1 + 6X_2 <= 1566$ } labor
 $12X_1 + 16X_2 <= 2880$ } tubing
 $X_1, X_2 >= 0$ } non-negativity
 X_1, X_2 must be integers } integrality

Integrality conditions are easy to state but make the problem much more difficult (and sometimes impossible) to solve.

Relaxation

Original ILP

MAX: $2X_1 + 3X_2$

S.T.: $X_1 + 3X_2 \le 8.25$

 $2.5X_1 + X_2 \le 8.75$

 $X_1, X_2 >= 0$

X₁, X₂ must be integers

This constraint is dropped in LP Relaxation

LP Relaxation

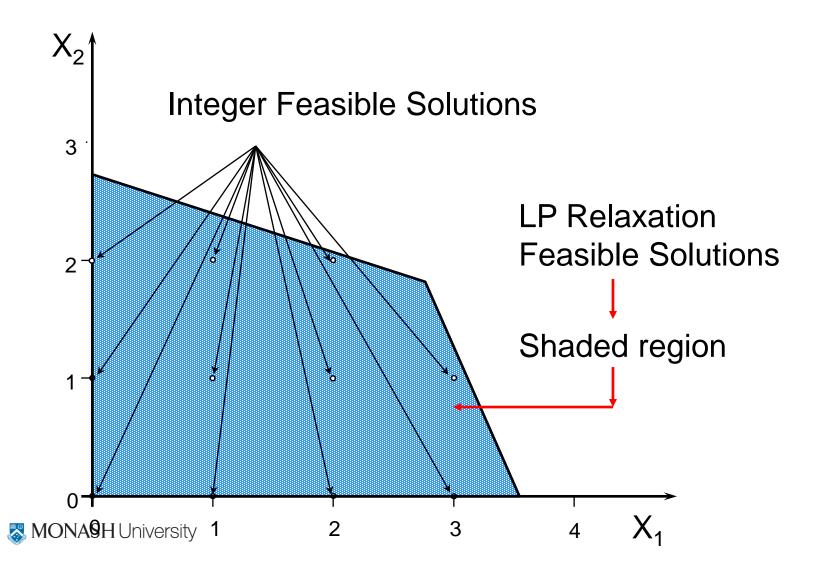
MAX: $2X_1 + 3X_2$

S.T.: $X_1 + 3X_2 \le 8.25$

 $2.5X_1 + X_2 \le 8.75$

 $X_1, X_2 >= 0$

Integer Feasible vs. LP Feasible Region



Solving ILP Problems

- When solving an LP relaxation, sometimes you "get lucky" and obtain an integer feasible solution.
- Example: Blue Ridge Hot Tubs

```
MAX: 350X_1 + 300X_2 } profit

S.T.: 1X_1 + 1X_2 <= 200 } pumps

9X_1 + 6X_2 <= 1566 } labor

12X_1 + 16X_2 <= 2880 } tubing

X_1, X_2 >= 0 } non-negativity

Optimal solution: X_1 = 122 and X_2 = 78
```



Solving ILP Problems

- But what if we reduce the amount of labor available to 1520 hours and the amount of tubing to 2650 feet?
- See file <u>Lecture 4.xlsm</u> (Blue Ridge)

_				
	Blue Ridg	e Hot Tubs		
	Aqua-Spas	Hydro-Luxes		
Number to Make	116.944444	77.91666667	Total Profit	
Unit Profits	\$350	\$300	\$64,306	
Constraints			Used	Available
Pumps Req'd	1	1	195	200
Labor Req'd	9	6	1520	1520
Tubing Req'd	12	16	2650	2650



		Blue Ridg	e Hot Tubs		
		Aqua-Spas	Hydro-Luxes		
	Number to Make	116.9444444	77.91666667	Total Profit	
Daunda	Unit Profits	\$350	\$300	\$64,306	
Bounds					
	Constraints			Used	Available
	Pumps Req'd	1	1	195	200
	Labor Req'd	9	6	1520	1520
	Tubing Req'd	12	16	2650	2650

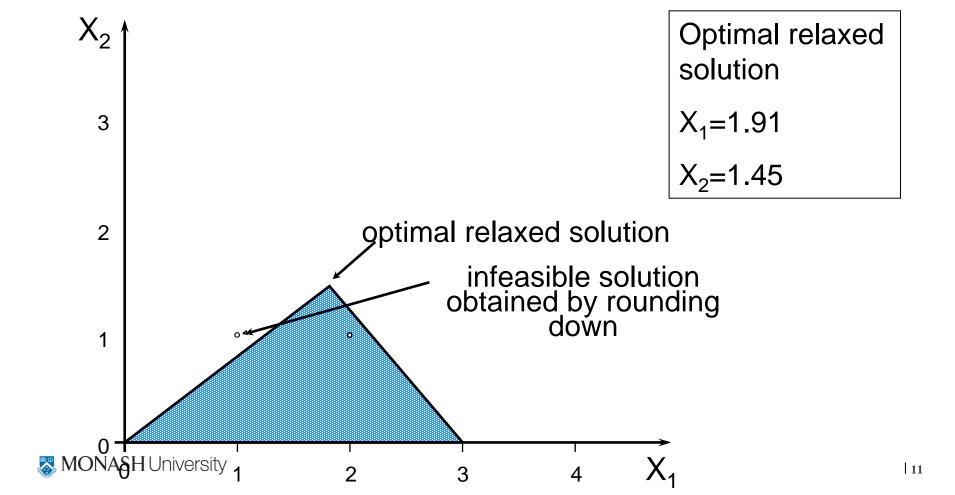
- The optimal solution to an LP relaxation of an ILP problem gives us a bound on the optimal objective function value.
- For maximization problems, the optimal relaxed objective function values is an <u>upper bound</u> on the optimal integer value.
- For minimization problems, the optimal relaxed objective function values is a <u>lower bound</u> on the optimal integer value.



Rounding

- It is tempting to simply round a fractional solution to the closest integer solution.
- In general, this does not work reliably:
 - The rounded solution may be infeasible.
 - The rounded solution may be suboptimal.

How Rounding Down Can Result in an Infeasible Solution



Rounding Up

LP

solution:

Blue Ridge	e Hot Tubs		
Aqua-Spas	Hydro-Luxes		
116.944444	77.91666667	Total Profit	
\$350	\$300	\$64,306	
		Used	Available
1	1	195	200
9	6	1520	1520
12	16	2650	2650
	Aqua-Spas 116.944444 \$350 1 9	116.9444444 77.91666667 \$350 \$300 1 1 1 9 6	Aqua-Spas Hydro-Luxes 116.9444444 77.91666667 Total Profit \$350 \$300 \$64,306 Used 1 1 1 195 9 6 1520

Round up - Infeasible \

_				
	Blue Ridg			
	Aqua-Spas	Hydro-Luxes		
Number to Make	117	78	Total Profit	
Unit Profits	\$350	\$300	\$64,350	
Constraints			Used	Available
Pumps Req'd	1	1	195	200
Labor Req'd	9	6	1521	1520
Tubing Req'd	12	16	2652	2650

Rounding Down

LP

solution

	Blue Ridg	e Hot Tubs		
	Aqua-Spas	Hydro-Luxes		
Number to Make	116.944444	77.91666667	Total Profit	
Unit Profits	\$350	\$300	\$64,306	
Constraints			Used	Available
Pumps Req'd	1	1	195	200
Labor Req'd	9	6	1520	1520
Tubing Req'd	12	16	2650	2650

Round down – Feasible …..but

	Blue Ridg	e Hot Tubs		
	Aqua-Spas	Hydro-Luxes		
Number to Make	116	77	Total Profit	
Unit Profits	\$350	\$300	\$63,700	
Constraints			Used	Available
Pumps Req'd	1	1	193	200
Labor Req'd	9	6	1506	1520
Tubing Req'd	12	16	2624	26 5 0

Rounding Down Causes Sub-Optimality

	Blue Ridg	e Hot Tubs		
	Aqua-Spas	Hydro-Luxes	5	
Number to Make	116	77	Total Profit	
Unit Profits	\$350	\$300	\$63,700	
Constraints			Used	Available
Pumps Req'd	1	1	193	200
Labor Req'd	9	6	1506	1520
Tubing Req'd	12	16	2624	2650

 A better integer solution exists (i.e. better than the above sub-optimal solution):

	Blue Ridg	e Hot Tubs		
	A qua-Spas	Hydro-Luxes		
Number to Make	118	76	Total Pr	ofit
Unit Profits	\$350	\$300	\$64,100	0
Constraints			Used	Available
Pumps Req'd	1	1	194	200
Labor Req'd	9	6	1518	1520
Tubing Req'd	12	16	2632	2650

Branch-and-Bound

- The Branch-and-Bound (B&B) algorithm can be used to solve ILP problems.
- Requires the solution of a series of LP problems termed "candidate problems".
- Theoretically, this can solve any ILP.
- Practically, it often takes LOTS of computational effort (and time).

Stopping Rules

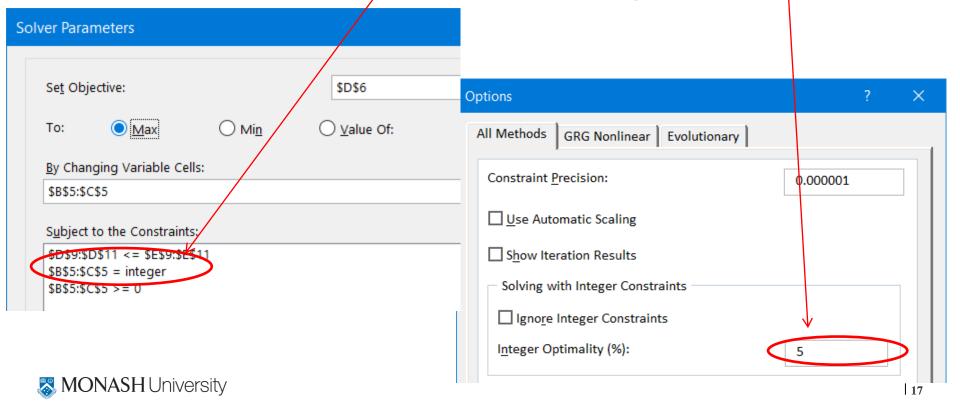
- Because B&B can take so long, most ILP packages allow you to specify a sub-optimality tolerance factor.
- This allows you to stop once an integer solution is found that is within some % of the global optimal solution.
- Bounds obtained from LP relaxations are helpful here.
 - Example
 - LP relaxation has an optimal obj. value of \$64,306.
 - 95% of \$64,306 is \$61,090.
 - Thus, an integer solution with obj. value of \$61,090 or better must be within 5% of the optimal solution.



Using Solver

Let's see how to specify integrality conditions and sub-optimality tolerances using Solver...

See file Lecture 4.xlsm (Blue Ridge – ILP)



An Employee Scheduling Problem: Air-Express

- An express shipping company guarantees o/night delivery
- Various hubs across the country shipments go to hubs, then on to their destination
- Manager of Baltimore hub is concerned about labour costs and wants to investigate the most effective way of scheduling of workers
- Hub open 7 days per week
- # packages varies from 1 day to the next
- An estimate of the number of workers needed on each day of the week has been calculated using historical data



An Employee Scheduling Problem: Air-Express

Day of Week	Workers Needed
Sunday	18
Monday	27
Tuesday	22
Wednesday	26
Thursday	25
Friday	21
Saturday	19

Shift	Days Off	Wage
1	Sun & Mon	\$680
2	Mon & Tue	\$705
3	Tue & Wed	\$705
4	Wed & Thr	\$705
5	Thr & Fri	\$705
6	Fri & Sat	\$680
7	Sat & Sun	\$655



Defining the Decision Variables

 X_1 = the number of workers assigned to shift 1

 X_2 = the number of workers assigned to shift 2

 X_3 = the number of workers assigned to shift 3

 X_4 = the number of workers assigned to shift 4

 X_5 = the number of workers assigned to shift 5

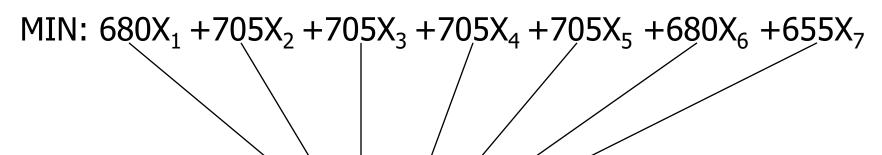
 X_6 = the number of workers assigned to shift 6

 X_7 = the number of workers assigned to shift 7



Defining the Objective Function

Minimize the total wage expense.



Wage per shift

Shift	Days Off	Wage
1	Sun & Mon	\$680
2	Mon & Tue	\$705
3	Tue & Wed	\$705
4	Wed & Thr	\$705
5	Thr & Fri	\$705
6	Fri & Sat	\$680
7	Sat & Sun	\$655



Defining the Constraints

Workers required each day

$$0X_{1}+1X_{2}+1X_{3}+1X_{4}+1X_{5}+1X_{6}+0X_{7}>=18$$
 } Sunday $0X_{1}+0X_{2}+1X_{3}+1X_{4}+1X_{5}+1X_{6}+1X_{7}>=27$ } Monday $1X_{1}+0X_{2}+0X_{3}+1X_{4}+1X_{5}+1X_{6}+1X_{7}>=22$ } Tuesday $1X_{1}+1X_{2}+0X_{3}+0X_{4}+1X_{5}+1X_{6}+1X_{7}>=26$ } Wednesday $1X_{1}+1X_{2}+1X_{3}+0X_{4}+0X_{5}+1X_{6}+1X_{7}>=25$ } Thursday $1X_{1}+1X_{2}+1X_{3}+1X_{4}+0X_{5}+0X_{6}+1X_{7}>=21$ } Friday $1X_{1}+1X_{2}+1X_{3}+1X_{4}+0X_{5}+0X_{6}+0X_{7}>=19$ } Saturday

Non-negativity & integrality conditions

$$X_i >= 0$$
 and integer for all i

Implementing the Model

See file <u>Lecture 4.xlsm</u> (*AirExpress*)

	Α	В	С	D	Е	F	G	Н		J
1				Λ:						
2				AIF-	Expres	SS				
3			D	ays On	=1, Da	ys Off-	=0		Workers	Wages per
4	Shift	Sun	Mon	Tues	Wed	Thur	Fri	Sat	Scheduled	Worker
5	1	0	0	1	1	1	1	1	6	\$680
6	2	1	0	0	1	1	1	1	0	\$705
- 7	3	1	1	0	0	1	1	1	5	\$705
8	4	1	1	1	0	0	1	1	1	\$705
9	5	1	1	1	1	0	0	1	7	\$705
10	6	1	1	1	1	1	0	0	5	\$680
11	7	0	1	1	1	1	1	0	9	\$655
12	Available <	18	27	28	27	25	21	19	> Total	\$22,540
13	Required	18	27	22	26	25	21	19		
14										

At least as many as required

Binary Variables

- Binary variables are integer variables that can assume only two values: 0 or 1.
- These variables can be useful in a number of practical modeling situations...

A Capital Budgeting Problem: CRT Technologies

Expected NPV		Capital (in \$000s) Required in					
Project	(in \$000s)	Year 1	Year 2	Year 3	Year 4	Year 5	
1	\$141	\$75	\$25	\$20	\$15	\$10	
2	\$187	\$90	\$35	\$0	\$0	\$30	
3	\$121	\$60	\$15	\$15	\$15	\$15	
4	\$83	\$30	\$20	\$10	\$5	\$5	
5	\$265	\$100	\$25	\$20	\$20	\$20	
6	\$127	\$50	\$20	\$10	\$30	\$40	

- The company has \$250,000 available to invest in new projects. It has budgeted \$75,000 for continued support for these projects in year 2 and \$50,000 per year for years 3, 4, and 5.
- Unused funds in any year cannot be carried over.

Defining Decision Variables & Objective Function

$$\mathbf{X}_{i} = \begin{cases} 1, & \text{if project } i \text{ is selected} \\ 0, & \text{otherwise} \end{cases} i = 1, 2, \dots, 6$$

Maximize total NPV of selected projects

MAX:
$$141X_1 + 187X_2 + 121X_3 + 83X_4 + 265X_5 + 127X_6$$

Expected NPV (\$000s)

Expected NPV

Project yr	(in \$000s)	Year 1	Year 2	Year 3	Year 4	Year 5
1	\$141	\$75	\$25	\$20	\$15	\$10
2	\$187	\$90	\$35	\$0	\$0	\$30
3	\$121	\$60	\$15	\$15	\$15	\$15
4	\$83	\$30	\$20	\$10	\$5	\$5
5	\$265	\$100	\$25	\$20	\$20	\$20
6	\$127	\$50	\$20	\$10	\$30	\$40

Defining Constraints

Capital Constraints

must ensure for each year that the selected projects do not require more capital than is available

e.g. year 2, \$75,000 is available, so:

$$25X_1 + 35X_2 + 15X_3 + 20X_4 + 25X_5 + 20X_6 \le 75$$

E	Expected NPV					
Project yr	(in \$000s)	Year 1	Year 2	Year 3	Year 4	Year 5
1	\$141	\$75	\$25	\$20	\$15	\$10
2	\$187	\$90	\$35	\$0	\$0	\$30
3	\$121	\$60	\$15	\$15	\$15	\$15
4	\$83	\$30	\$20	\$10	\$ 5	\$5
5	\$265	\$100	\$25	\$20	\$20	\$20
MONASH Univ	ersity\$127	\$50	\$20	\$10	\$30	\$40

Defining the Constraints

Capital Constraints

$$75X_{1} + 90X_{2} + 60X_{3} + 30X_{4} + 100X_{5} + 50X_{6} <= 250$$
 } year 1
$$25X_{1} + 35X_{2} + 15X_{3} + 20X_{4} + 25X_{5} + 20X_{6} <= 75$$
 } year 2
$$20X_{1} + 0X_{2} + 15X_{3} + 10X_{4} + 20X_{5} + 10X_{6} <= 50$$
 } year 3
$$15X_{1} + 0X_{2} + 15X_{3} + 5X_{4} + 20X_{5} + 30X_{6} <= 50$$
 } year 4
$$10X_{1} + 30X_{2} + 15X_{3} + 5X_{4} + 20X_{5} + 40X_{6} <= 50$$
 } year 5

Binary Constraints

All X, must be binary

Implementing the Model

See file <u>Lecture 4.xlsm(CRT)</u>

	Α	В	С	D	Е	F	G	Н		
1										
2			C	RT Tech						
3										
4		Select?		Capital Requir				ed in		
5	Project	(0=no, 1=yes)	NPV	Year 1	Year 2	Year 3	Year 4	Year 5		
6	1	1	\$141	\$75	\$25	\$20	\$15	\$10		
7	2	0	\$187	\$90	\$35	\$0	\$0	\$30		
8	3	0	\$121	\$60	\$15	\$15	\$15	\$15		
9	4	1	\$83	\$30	\$20	\$10	\$5	\$5		
10	5	1	\$265	\$100	\$25	\$20	\$20	\$20		
11	6	0	\$127	\$50	\$20	\$10	\$30	\$40		
12		Capita	al Required	\$205	\$70	\$50	\$40	\$35		
13		Capit	al Available	\$250	\$75	\$50	\$50	\$50		
14										
15		Total Net Pre	esent Value	\$489						
16										



Binary Variables & Logical Conditions

- Binary variables are also useful in modeling a number of logical conditions.
 - Of projects 1, 3 & 6, no more than one may be selected: $X_1 + X_3 + X_6 \le 1$
 - Of projects 1, 3 & 6, exactly one must be selected: $X_1 + X_3 + X_6 = 1$
 - Project 4 cannot be selected unless project 5 is also selected: $X_4 X_5 \le 0$

The Fixed-Charge Problem

- Many decisions result in a fixed or lump-sum cost being incurred:
 - The cost to lease, rent, or purchase a piece of equipment or a vehicle that will be required if a particular action is taken.
 - The setup cost required to prepare a machine or to produce a different type of product.
 - The cost to construct a new production line that will be required if a particular decision is made.
 - The cost of hiring additional personnel that will be required if a particular decision is made.



Example Fixed-Charge Problem: Remington Manufacturing

Hours Required By:

Operation	Prod. 1	Prod. 2	Prod. 3	Hours Available
Machining	2	3	6	600
Grinding	6	3	4	300
Assembly	5	6	2	400
Unit Profit	\$48	\$55	\$50	
Setup Cost	\$1000	\$800	\$900	

Fixed charge for making any quantity of prod 1, prod 2 or prod 3



Defining Decision Variables

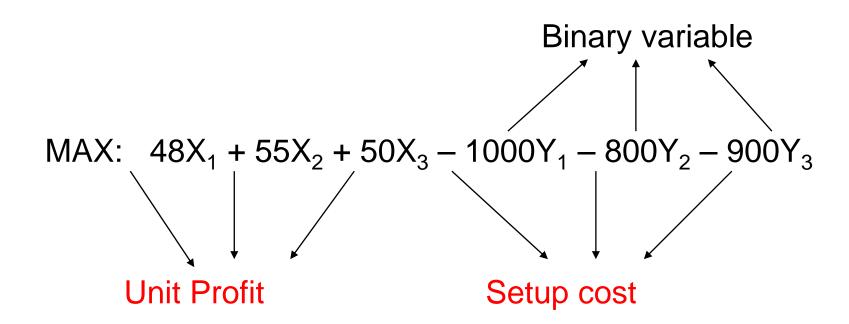
 X_i = the amount of product *i* to be produced, i = 1, 2, 3

$$Y_{i} = \begin{cases} 1, & \text{if } X_{i} > 0 \\ 0, & \text{if } X_{i} = 0 \end{cases} i = 1, 2, 3$$

Y_i are binary variables that will be used to include the fixed charges

Defining the Objective Function

Maximize total profit.



Defining the Constraints

Resource Constraints

$$2X_1 + 3X_2 + 6X_3 \le 600$$
 } machining $6X_1 + 3X_2 + 4X_3 \le 300$ } grinding $5X_1 + 6X_2 + 2X_3 \le 400$ } assembly

Non-negativity & integer conditions

$$X_i >= 0, i = 1, 2,...3$$

 X_i integer, i=1,...3

Binary Constraints

All Y_i must be binary

- Is there a missing link?
- Yes we need to ensure that $Y_i = 1$ if $X_i > 0$

Linking Constraints

Linking Constraints (with "Big M")

$$X_1 \le M_1 Y_1$$
 or $X_1 - M_1 Y_1 \le 0$
 $X_2 \le M_2 Y_2$ or $X_2 - M_2 Y_2 \le 0$
 $X_3 \le M_3 Y_3$ or $X_3 - M_3 Y_3 \le 0$

- If X_i > 0 these constraints force the associated Y_i to equal 1.
- If $X_i = 0$ these constraints allow Y_i to equal 0 or 1, but the objective will cause Solver to choose 0.
- Note that M_i imposes an upper bounds on X_i.
- It helps to find reasonable values for the M_i. Maximum value of Xi = Mi



But we don't want to constrain X_i any further

Finding Reasonable Values for M1

Consider the resource constraints

$$2X_1 + 3X_2 + 6X_3 \le 600$$
 } machining $6X_1 + 3X_2 + 4X_3 \le 300$ } grinding $5X_1 + 6X_2 + 2X_3 \le 400$ } assembly

■ What is the maximum value X₁ can assume?

Let
$$X_2 = X_3 = 0$$

 $X_1 = MIN(600/2, 300/6, 400/5)$
 $= MIN(300, 50, 80)$
 $= 50$

So we can put M₁

• Maximum values for $X_2 \& X_3$ can be found =50

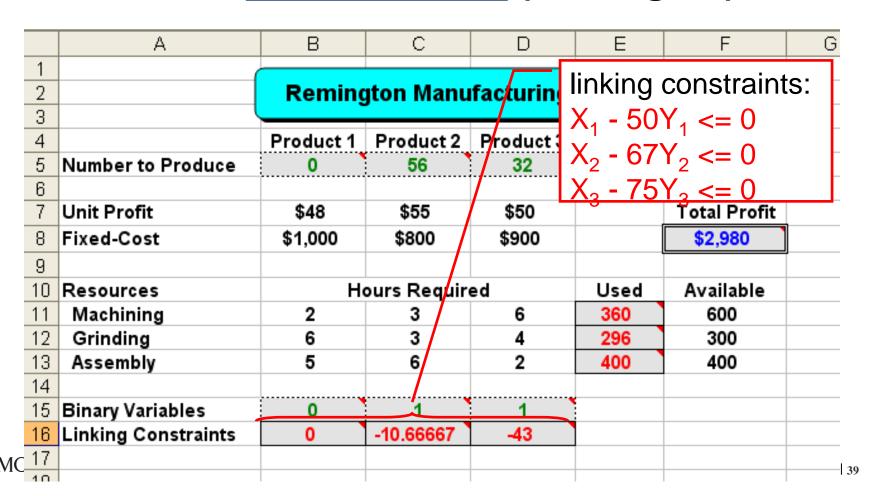
Summary of the Model

MAX:
$$48X_1 + 55X_2 + 50X_3 - 1000Y_1 - 800Y_2 - 900Y_3$$

S.T.: $2X_1 + 3X_2 + 6X_3 <= 600$ } machining $6X_1 + 3X_2 + 4X_3 <= 30$ } grinding $5X_1 + 6X_2 + 2X_3 <= 400$ } assembly $X_1 - 50Y_1 <= 0$ $X_2 - 67Y_2 <= 0$ $X_3 - 75Y_3 <= 0$ All Y_i must be binary $X_i >= 0$, $i = 1, 2, 3$ (= integer)

Implementing Model

See file <u>Lecture 4.xlsm</u> (Remington)



Potential Pitfall

- Do not use IF() functions to model the relationship between the X_i and Y_i.
 - Suppose cell B5 represents X₁
 - Suppose cell B15 represents Y₁
 - You'll want to let B15 = IF(B5>0,1,0)
 - This will not work with Solver!
- Treat the Y_i just like any other variable.
 - Make them changing cells.
 - Use the linking constraints to enforce the proper relationship between the X_i and Y_i.



Minimum Order Size Restrictions

Suppose Remington doesn't want to manufacture any units of product 3 unless it produces at least 40 units...

Consider,

 $X_3 \leq M_3 Y_3$

$$X_3 >= 40 Y_3$$

See <u>Lecture 4.xlsm</u> (*Remington – Min order*)

Use M3 = min(600/6,300/4,400/2)=75

B&G – A Contract Award Problem

B&G Construction has 4 building projects and can purchase cement from 3 companies for the following costs:

	Cost per Delivered Ton of Cement				Max.
	Project 1	Project 2	Project 3	Project 4	Supply
Co. 1	\$120	\$115	\$130	\$125	525
Co. 2	\$100	\$150	\$110	\$105	450
Co. 3	\$140	\$95	\$145	\$165	550
Needs	450	275	300	350	
(tons)					



Defining the Decision Variables

 X_{ij} = tons of cement purchased from company i for project j

Defining the Objective Function

Minimize total cost

MIN:
$$120X_{11} + 115X_{12} + 130X_{13} + 125X_{14}$$

 $+ 100X_{21} + 150X_{22} + 110X_{23} + 105X_{24}$
 $+ 140X_{31} + 95X_{32} + 145X_{33} + 165X_{34}$

A Contract Award Problem

Side constraints:

Co. 1 will not supply orders of less than 150 tons for any project

Co. 2 can supply more than 200 tons to no more than one of the projects

Co. 3 will accept only orders that total 200, 400, or 550 tons

Defining the Constraints

Supply Constraints

$$X_{11} + X_{12} + X_{13} + X_{14} \le 525$$
 } company 1
 $X_{21} + X_{22} + X_{23} + X_{24} \le 450$ } company 2
 $X_{31} + X_{32} + X_{33} + X_{34} \le 550$ } company 3

Demand Constraints

$$X_{11} + X_{21} + X_{31} = 450$$
 } project 1
 $X_{12} + X_{22} + X_{32} = 275$ } project 2
 $X_{13} + X_{23} + X_{33} = 300$ } project 3
 $X_{14} + X_{24} + X_{34} = 350$ } project 4

Defining the Constraints - I

Company 1 Side Constraints

$$X_{11} < =525Y_{11}$$

$$X_{12} < =525Y_{12}$$

$$X_{13} < =525Y_{13}$$

$$X_{14} < = 525Y_{14}$$

$$X_{11} > = 150Y_{11}$$

$$X_{12} > = 150Y_{12}$$

$$X_{13} > = 150Y_{13}$$

$$X_{14} > = 150Y_{14}$$

 Y_{ii} binary

Defining the Constraints- II & III

Company 2 Side Constraints

$$X_{21} <= 200 + 250 Y_{21}$$

 $X_{22} <= 200 + 250 Y_{22}$
 $X_{23} <= 200 + 250 Y_{23}$
 $X_{24} <= 200 + 250 Y_{24}$
 $Y_{21} + Y_{22} + Y_{23} + Y_{24} <= 1$
 Y_{ij} binary

Company 3 Side Constraints

$$X_{31} + X_{32} + X_{33} + X_{34} = 200Y_{31} + 400Y_{32} + 550Y_{33}$$

 $Y_{31} + Y_{32} + Y_{33} \le 1$

Implementing the Transportation Constraints

See file <u>Lecture 4.xlsm(*B&G*</u>)



These are non-examinable but good to know ...

- ☐ Goal Programming
- ☐ Multiple Objective LP (MOLP)



Multiple Objectives

- Most optimisation problems considered to this point have had a single objective.
- Often, more than one objective can be identified for a given problem.
 - Maximize Return or Minimize Risk
 - Maximize Profit or Minimize Pollution
- These objectives often conflict with one another.
- How can such problems be dealt with?

Goal Programming (GP)

- Most LP problems have <u>hard constraints</u> that cannot be violated...
 - There are 1,566 labor hours available.
 - There is \$850,000 available for projects.
- In some cases, hard constraints are too restrictive...
 - You have a maximum price in mind when buying a car (this is your "goal" or target price).
 - If you can't buy the car for this price you'll likely find a way to spend more.
- We use <u>soft constraints</u> to represent such goals or targets we'd like to achieve.



GP Example: Beach Hotel Expansion

- Davis McKeown wants to expand the convention center at his hotel in Myrtle Beach, SC.
- The types of conference rooms being considered are:

	Size (sq ft)	Unit Cost
Small	400	\$18,000
Medium	750	\$33,000
Large	1,050	\$45,150

- Davis would like to add 5 small, 10 medium and 15 large conference rooms.
- He also wants the total expansion to be 25,000 square feet and to limit the cost to \$1,000,000.

Defining Goals

- Goal 1: The expansion should include *approximately* 5 small conference rooms.
- Goal 2: The expansion should include *approximately* 10 medium conference rooms.
- Goal 3: The expansion should include *approximately* 15 large conference rooms.
- Goal 4: The expansion should consist of *approximately* 25,000 square feet.
- Goal 5: The expansion should cost *approximately* \$1,000,000.



Defining Decision Variables

 X_1 = number of small rooms to add

 X_2 = number of medium rooms to add

 X_3 = number of large rooms to add

Deviational Variables:

MONASH University

Amounts by which each goal deviates from its target value

$$d_i^-$$
 and d_i^+

- (-) represents amount of underachievement of each goal's target value
- (+) represents amount of overachievement of each goal's target value

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Defining the Goal Constraints-I

Small Rooms

$$X_1 + d_1^- - d_1^+ = 5$$

Medium Rooms

$$X_2 + d_2^- - d_2^+ = 10$$

Large Rooms

$$X_3 + d_3^- - d_3^+ = 15$$

where

$$\mathbf{d}_{i}^{-},\mathbf{d}_{i}^{+}\geq0$$

If
$$X_1 = 3$$
,

$$d1(-) = 2$$

$$d1(+) = 0$$

If
$$X_2 = 13$$
,

$$d2(-) = 0$$

$$d2(+) = 3$$

Defining the Goal Constraints-II

Total Expansion

$$400X_1 + 750X_2 + 1,050X_3 + d_4^- - d_4^+ = 25,000$$

Total Cost (in \$1,000s)

$$18X_1 + 33X_2 + 45.15X_3 + d_5^- - d_5^+ = 1,000$$

where

$$d_i^-, d_i^+ \ge 0$$

GP Objective Functions – Option 1

- There are numerous objective functions we could formulate for a GP problem.
- Minimize the sum of the deviations:

$$\mathsf{MIN} \sum_{i} \left(d_i^- + d_i^+ \right)$$

- Problem: The deviations measure different things, so what does this objective represent?
- e.g. 7 rooms + 1500 \$ = 1507 of ?

GP Objective Functions – Option 2

- Minimize the sum of percentage deviations
 - $MIN \sum_{i} \frac{1}{t_i} \left(d_i^- + d_i^+ \right)$
 - where ti represents the target value of goal i
- Problem: Suppose the first goal is underachieved by 1 small room and the fifth goal is overachieved by \$20,000.
 - We underachieve goal 1 by 1/5=20%
 - We overachieve goal 5 by 20,000/1,000,000= 2%
 - This implies being \$20,000 over budget is just as undesirable as having one too few small rooms.



GP Objective Functions – Option 3

- Weights can be used in the previous objectives to allow the decision maker indicate
 - desirable vs. undesirable deviations
 - the relative importance of various goals
- Minimize the weighted sum of deviations

$$\min \left| \sum_{i} \left(w_i^- d_i^- + w_i^+ d_i^+ \right) \right|$$

Minimize the weighted sum of % deviations

$$\sum_{i} \frac{1}{t_i} \left(w_i^- d_i^- + w_i^+ d_i^+ \right)$$

Defining the Objective

- Assume
 - It is undesirable to underachieve (-) any of the first three room goals
 - It is undesirable to overachieve (-, +) or underachieve the 25,000 sq ft expansion goal
 - It is undesirable to overachieve (+) the \$1,000,000 total cost goal

MIN:
$$\frac{w_1^-}{5}d_1^- + \frac{w_2^-}{5}d_2^- + \frac{w_3^-}{5}d_3^- + \frac{w_4^-}{25,000}d_4^- + \frac{w_4^+}{25,000}d_4^+ + \frac{w_5^+}{1,000,000}d_5^+$$

Initially, we will assume all the above weights equal 1.

Implementation - Lecture 4_GP.xlsm



About GP

- GP involves making trade-offs among the goals until the most satisfying solution is found.
- GP objective function values should not be compared because the weights are changed in each iteration.
 Compare the solutions!
- An arbitrarily large weight will effectively change a soft constraint to a hard constraint.
- Hard constraints can be place on deviational variables.



Summary of GP

- 1. Identify the decision variables in the problem.
- Identify any hard constraints in the problem and formulate them in the usual way.
- 3. State the goals of the problem along with their target values.
- 4. Create constraints using the decision variables that would achieve the goals exactly.
- Transform the above constraints into goal constraints by including deviational variables.
- 6. Determine which deviational variables represent undesirable deviations from the goals.
- 7. Formulate an objective that penalizes the undesirable deviations.
- 8. Identify appropriate weights for the objective.
- 9. Solve the problem.
- 10. Inspect the solution to the problem. If the solution is unacceptable, return to step 8 and revise the weights as needed.



Multiple Objective Linear Programming (MOLP)

- An MOLP problem is an LP problem with more than one objective function.
- MOLP problems can be viewed as special types of GP problems where we must also determine target values for each goal or objective.
- Analyzing these problems effectively also requires that we use the MiniMax objective

Summary of MOLP

- 1. Identify the decision variables in the problem.
- 2. Identify the objectives in the problem and formulate them as usual.
- 3. Identify the constraints in the problem and formulate them as usual.
- 4. Solve the problem once for each of the objectives identified in step 2 to determine the optimal value of each objective.
- 5. Restate the objectives as goals using the optimal objective values identified in step 4 as the target values.
- 6. For each goal, create a deviation function that measures the amount by which any given solution fails to meet the goal (either as an absolute or a percentage).
- 7. For each of the functions identified in step 6, assign a weight to the function and create a constraint that requires the value of the weighted deviation function to be less than the MINIMAX variable Q.
- 8. Solve the resulting problem with the objective of minimizing Q.
- 9. Inspect the solution to the problem. If the solution is unacceptable, adjust the weights in step 7 and return to step 8.



End of Lecture 4

Content References:

Ragsdale, C. Spreadsheet Modeling & Decision Analysis: A Practical Introduction to Business Analytics (8e/9e) Cengage Learning: Chapter 6 & 7



Homework

- Go through today's lecture examples and Ragsdale Chapter 6, to:
 - ✓ Familiarise yourself with the ILP formulation and models
 - ✓ Understand the use of "Big M" in linking constraints
- ➤ Concepts and modeling techniques used in GP and MOLP problems are non-examinable and there only for your reference with Ragsdale chapter 7.



Readings for next week Lecture:

Ragsdale, C. Spreadsheet Modeling & Decision Analysis: A Practical Introduction to Business Analytics (8e/9e) Cengage Learning: Chapter 5



Tutorial 3 this week:

- Interpreting Solver reports
 - Answer Report
 - Sensitivity Report
 - Limits report

- Spider plot
- Solver tables

