SYSEN 6000: Foundations of Complex Systems

Deterministic Global Optimization

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Integer Programming

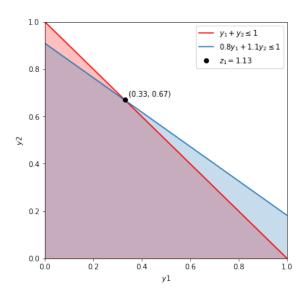
Given that:

$$\max y_1 + 1.2y_2$$
s.t. $y_1 + y_2 \le 1$

$$0.8y_1 + 1.1y_2 \le 1$$

$$y_1, y_2 = \{0, 1\}$$
(1)

When the binary variables are relaxed to continuous ones, such that $y_1, y_2 \in [0, 1]$, the contours of the objective and the feasible region can be exhibited as:



From inspection, the optimal solution to the relaxed problem, such that $y_1, y_2 \in [0, 1]$, is:

$$y_1 = \frac{1}{3}, y_2 = \frac{2}{3}$$

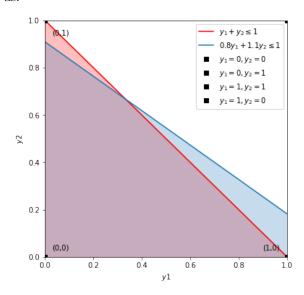
$$z_1 = 1.13$$
(2)

Enumerating all y_1, y_2 integer combinations [0, 1] the solution space can be expressed as:

$$y_1 = 0, y_2 = 0$$

 $y_1 = 0, y_2 = 1$
 $y_1 = 1, y_2 = 1$
 $y_1 = 1, y_2 = 0$ (3)

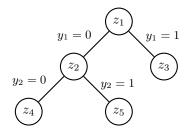
which can be exhibited as:



Therefore, the optimal solution is:

$$y_1 = 1, y_2 = 0 (4)$$

The branch and bound method for solving LP subproblems can be exhibited as:



where:

$$z_1 = 1.13$$
 $z_2 = 1.091$
 $z_3 = 1$
 $z_4 = 0$
 $z_5 = \cdot$
(5)

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Therefore, the optimal solution is:

$$y_1 = 1, y_2 = 0$$

$$z_3 = 1$$
(6)

Gomory Cut

Given the inequality:

$$y_2 + y_3 + 2y_4 \le 6 \tag{7}$$

and the region:

$$X = \{ y \in Z_4 : 4y_1 + 5y_2 + 9y_3 + 12y_4 \le 34 \}$$
(8)

It is provably valid by:

$$\frac{4}{5}y_1 + \frac{5}{5}y_2 + \frac{9}{5}y_3 + \frac{12}{5}y_4 \le \frac{34}{5} \Longrightarrow 0 + \frac{4}{5}y_1 + y_2 + 1 + \frac{4}{5}y_3 + 2 + \frac{2}{5}y_4 \le \frac{34}{5}$$
 (9)

when considering:

$$\sum_{j} a_{ij}^* y_j \le b_i, y_j \in \{0, 1\}$$
(10)

which is given by:

$$\sum_{j} \lfloor a_{ij} \rfloor y_j \le \lfloor b_i \rfloor \tag{11}$$

and when applied is:

$$0 \cdot y_1 + y_2 + 1 \cdot y_3 + 2 \cdot y_4 \le 6 \tag{12}$$

which results in the given inequality:

$$y_2 + y_3 + 2y_4 \le 6 \tag{13}$$

Mixed-Integer Non-Linear Programming

Given that:

min
$$f = y_1 + 1.5y_2 + 0.5y_3 + x_1^2 + x_2^2$$

s.t. $(x_1 - 2)^2 - x_2 \le 0$
 $x_1 - 2y_1 \ge 0$
 $x_1 - x_2 - 4(1 - y_1) \le 0$
 $x_1 - (1 - y_1) \ge 0$
 $x_2 - y_2 \ge 0$
 $x_1 + x_2 \ge 3y_3$
 $y_1 + y_2 + y_3 \ge 1$
 $0 \le x_1 \le 4, 0 \le x_2 \le 4$
 $y_1, y_2, y_3 = \{0, 1\}$ (14)

the inequalities can be reformulated as:

s.t.
$$(x_1 - 2)^2 - x_2 \le 0$$

 $-(x_1 - 2y_1) \le 0$
 $x_1 - x_2 - 4(1 - y_1) \le 0$
 $-x_1 - (1 - y_1) \le 0$
 $-(x_2 + y_2) \le 0$
 $-(x_2 + y_2 - 3y_3) \le 0$
 $-(y_1 + y_2 + y_3 - 1) \le 0$ (15)

Let $y_1 = y_2 = y_3 = 1$, so that:

min
$$f = 3 + x_1^2 + x_2^2$$

s.t. $(x_1 - 2)^2 - x_2 \le 0$
 $-(x_1 - 2 \cdot 1) \le 0$
 $x_1 - x_2 - 4(1 - 1) \le 0$
 $-x_1 - (1 - 1) \le 0$
 $-(x_2 + 1) \le 0$
 $-(x_2 + 1 - 3 \cdot 1) \le 0$
 $-(1 + 1 + 1 - 1) \le 0$
 $0 \le x_1 \le 4, 0 \le x_2 \le 4$
 $y_1, y_2, y_3 = \{0, 1\}$

which results in $x_1^* = 2, x_2^* = 2, UB_1 = 5$, where we can continue to solve for:

$$\nabla f = \begin{bmatrix} 2 \cdot 2 \\ 2 \cdot 2 \end{bmatrix} \Longrightarrow \nabla f^* = \begin{bmatrix} 4 \\ 4 \end{bmatrix} \tag{17}$$

$$f_{Lin} = f(x^*) + \nabla f(x^*) + (x - x^*)$$

$$= 8 + \begin{bmatrix} 4 \\ 4 \end{bmatrix} \begin{bmatrix} 2 - 2 \\ 2 - 2 \end{bmatrix}$$

$$= 8 + 4(2 - 2) + 4(2 - 2) +$$

$$(y_1 + 1.5y_2 + 0.5y_3)$$
(18)

$$\nabla g = \begin{bmatrix} 2(2-2) \\ -1 \end{bmatrix} \Longrightarrow \nabla g^* = \begin{bmatrix} 0 \\ -1 \end{bmatrix} \tag{19}$$

$$g_{Lin} = g(x^*) + \nabla g(x^*) + (x - x^*)$$

$$= -2 + \begin{bmatrix} 0 \\ -1 \end{bmatrix} \begin{bmatrix} 2 - 2 \\ 2 - 2 \end{bmatrix}$$

$$= -2 - (2 - 2) \le 0$$
(20)

Hence $x_1^* = 1, x_2^* = 1, y_1 = 0, y_2 = 1, y_3 = 0, LB_1 = 3.5$, when min α , s.t. $\alpha \ge f_L in, g_L in \le 0$, and the reformulated constraints, where the $UB_1 > LB_1$.

Now let $y_1 = 0, y_2 = 1, y_3 = 0$, so that:

$$\begin{aligned} & \min \quad f = 1.5 + x_1^2 + x_2^2 \\ & \text{s.t.} \quad (x_1 - 2)^2 - x_2 \le 0 \\ & - x_1 \le 0 \\ & x_1 - x_2 - 4 \le 0 \\ & - x_1 - 1 \le 0 \\ & - x_2 + 1 \le 0 \\ & - x_2 + 1 \le 0 \\ & - x_1 - x_2 \le 0 \\ & - x_2 \le 0 \\ & x_1 - 4 \le 0 \\ & x_2 - 4 \le 0 \\ & 0 \le x_1 \le 4, 0 \le x_2 \le 4 \\ & y_1, y_2, y_3 = \{0, 1\} \end{aligned}$$

Hence $x_1^*=1, x_2^*=1, y_1=0, y_2=1, y_3=0, UB_2=3.5$, when min α , s.t. $\alpha \geq f_L in, g_L in \leq 0$, and the reformulated constraints, where $UB_2=LB_1$.

Here the stopping criteria UB = LB has been satisfied, where the optimal solution is:

$$x_1 = 1, x_2 = 1$$

 $y_1 = 0, y_2 = 1, y_3 = 0$
 $z_3 = 3.5$ (22)

Mixed-Integer Linear Fractional Programming

Given that:

$$\max \frac{2x + 4y_1 + 3y_2}{x + y_1 + 1}$$
s.t.
$$4x - y_1 - 4y_2 \le 1$$

$$5x + 4y_1 + 2y_2 \le 3$$

$$x \ge 0$$

$$y_1, y_2 \in \{0, 1\}$$
(23)

The parametric algorithm applies q, which reformulates into:

$$\max (2x + 4y_1 + 3y_2) - q \cdot (x + y_1 + 1)$$
s.t. $x + y_1 + 1 > 0$

$$4x - y_1 - 4y_2 \le 1$$

$$5x + 4y_1 + 2y_2 \le 3$$

$$x \ge 0$$

$$y_1, y_2 \in \{0, 1\}$$
(24)

that when solved, the optimal solution is:

$$x = 0$$

$$y_1 = 0, y_2 = 1$$
 (25)

The reformulation-linearization algorithm applies g and z, which reformulates into:

$$\max 2 \cdot z + 4 \cdot g \cdot y_1 + 3 \cdot g \cdot y_2$$
s.t. $z + g \cdot y_1 + 1 \cdot g = 1$

$$4 \cdot z - g \cdot y_1 - 4 \cdot g \cdot y_2 \le 1$$

$$5 \cdot z + 4 \cdot g \cdot y_1 + 2 \cdot g \cdot y_2 \le 3$$

$$z \ge 0, g \ge 0$$

$$y_1, y_2 \in \{0, 1\}$$
(26)

and with w_1 and w_2 , further into:

$$\max 2 \cdot z + 4 \cdot w_1 + 3 \cdot w_2$$
s.t. $z + w_1 + 1 \cdot g = 1$

$$4 \cdot z - w_1 - 4 \cdot w_2 - g \le 0$$

$$5 \cdot z + 4 \cdot w_1 + 2 \cdot w_2 - 3 \cdot g \le 0$$

$$0 \le w_1 \le M \cdot y_1$$

$$0 \le w_2 \le M \cdot y_2$$

$$w_1 \le g$$

$$w_2 \le g$$

$$w_2 \le g$$

$$w_1 \ge g - M(1 - y_1)$$

$$w_2 \ge g - M(1 - y_2)$$

$$z \ge 0$$

$$g \ge 0$$

$$w_1 \ge 0$$

$$w_2 \ge 0$$

$$y_1, y_2 \in \{0, 1\}$$

where:

$$z = g \cdot x$$

$$g = \frac{1}{x + y_1 + 1}$$

$$w_1 = g \cdot y_1, w_2 = g \cdot y_2$$
(28)

that when solved, the optimal solution is:

$$z = 0$$

 $g = 1$
 $w_1 = 0, w_2 = 1$
 $y_1 = 0, y_2 = 1$ (29)

Appendix

1. Integer Programming

```
## librarires
2 import numpy as np
3 from scipy.optimize import linprog
4 import matplotlib.pyplot as plt
6 ## minimize objective
7 ## (inversed original maximize)
8 \text{ obj } = [-1.0, -1.2]
10 ## constraints
## left side inequalities
12 lft_ieq = [
      [1.0, 1.0],
[0.8, 1.1]
13
14
15 ]
16
17 ## right side inequalities
18 rgt_ieq = [
    1.0,
19
      1.0
20
21 ]
23 ## global bounds
gbl_bds = [
     [0.0, 1.0], ## y1
      [0.0, 1.0] ## y2
27 ]
29 ## optimize
30 opt = linprog(
c = obj
      A_ub = lft_ieq,
32
      b_ub = rgt_ieq,
     bounds = gbl_bds,
method = 'simplex'
34
35
36 )
37
38 ## solution
39 y1 = round(
number = opt.x[0],
      ndigits = 2
41
42 )
43
y2 = round(
number = opt.x[1],
      ndigits = 2
46
47 )
48
49 z = round(
     number = -opt.fun,
51
      ndigits = 2
54 print('y1 = {x}'.format(x = y1))
55 print('y2 = {x}'.format(x = y2))
print('z = \{x\}'.format(x = z))
```

Python 3: Integer Programming Solution to the Relaxed Problem

2. Continuous Contours

```
1 ## librarires
2 import numpy as np
3 import matplotlib.pyplot as plt
5 ## space
6 x = np.linspace(0, 1, 100)
8 ## constraints
9 y1 = (1 - x)
y2 = (0.8 / -1.1) * x + (-1.0 / -1.1)
12 ## plot
plt.figure(figsize=(6,6))
plt.plot(x, y1,
      color = 'red',
      label = r'$y_1 + y_2 \leq 1$'
16
17 )
18 plt.plot(
19
     x, y2,
      label = r'$0.8y_1 + 1.1y_2 \leq 1$'
20
21 )
22
23 plt.plot(
      0.33, 0.67, 'o',
24
      color = 'black',
      label = r'$z_1 = 1.13$'
26
27 )
29 plt.annotate(
s = '(0.33, 0.67)',
     xy = (0.33, 0.67),
textcoords = "offset points",
31
32
      xytext = (5, 5),
33
      ha = 'left'
34
35 )
36
37 plt.fill_between(
38
    x, y1,
      color = 'red',
39
      alpha = 0.25
40
41 )
42
43 plt.fill_between(
44
      x, y2,
      alpha = 0.25
45
46 )
48 plt.xlabel(r'$y1$')
49 plt.ylabel(r'$y2$')
50 plt.xlim(0, 1)
51 plt.ylim(0, 1)
53 plt.legend(
      bbox_to_anchor = (1, 1),
      loc = 0,
55
      borderaxespad = 0.5
56
57 )
59 plt.show()
```

Python 3: Contours of the Relaxed Problem

3. Enumerated Combinations

```
1 ## librarires
2 import numpy as np
3 import matplotlib.pyplot as plt
5 ## space
6 x = np.linspace(0, 1, 100)
8 ## constraints
9 y1 = (1 - x)
y2 = (0.8 / -1.1) * x + (-1.0 / -1.1)
12 ## plot
plt.figure(figsize=(6,6))
14 plt.plot(
15
      x, y1,
      color = 'red',
16
      label = r' y_1 + y_2 \leq 1 
17
18 )
19
20 plt.plot(
x, y2,
22
      label = r'$0.8y_1 + 1.1y_2 \leq 1$'
23 )
24
25 plt.plot(
     0, 0, 's',
26
27
       color = 'black',
      label = r' y_1 = 0, y_2 = 0;
28
29 )
31 plt.plot(
32 0, 1, 's', color = 'black',
      label = r' y_1 = 0, y_2 = 1;
34
35 )
36
37 plt.plot(
38 1, 1, 's', color = 'black',
      label = r' y_1 = 1, y_2 = 1;
40
41 )
42
43 plt.plot(
     1, 0, 's',
      color = 'black',
45
      label = r' y_1 = 1, y_2 = 0;
46
47 )
48
49 plt.annotate(
s = '(0,0)',
     xy = (0, 0),
textcoords = "offset points",
51
52
     xytext = (10, 10),
53
54
      ha = 'left'
55 )
56
57 plt.annotate(
s = (0,1),
     xy = (0, 1),
59
     textcoords = "offset points",
xytext = (10, -20),
60
61 xytext = (1
62 ha = 'left'
63 )
```

```
65 plt.annotate(
66 s = '(1,0)',
67 xy = (1, 0),
68 textcoords = "offset points",
      xytext = (-10, 10),
69
      ha = 'right'
70
71 )
72
73 plt.fill_between(
      x, y1,
color = 'red',
75
       alpha = 0.25
76
77 )
78
79 plt.fill_between(
80
     x, y2,
81
       alpha = 0.25
82 )
83
84 plt.xlabel(r'$y1$')
85 plt.ylabel(r'$y2$')
86 plt.xlim(0, 1)
87 plt.ylim(0, 1)
89 plt.legend(
      bbox_to_anchor = (1, 1),
loc = 0,
90
       borderaxespad = 0.5
92
93 )
94
95 plt.show()
97 ## happy halloween!
```

Python 3: Enumerated Combinations for the Optimal Solution