

Optimization Homework 1

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- I have already completed the equivalent courses at Southern University of Science and Technology (SUSTech), including the following courses.

Introduction to Computer Science (WashU) \iff Introduction to Computer Science (SUSTech)

Matrix Algebra (WashU) \iff Advanced Linear Algebra I and II (SUSTech)

Engineering Mathematics A \iff Mathematical Analysis I, II, III (SUSTech)

- There are 5000 units of wood and 1500 units of labor.

nightstand : number — x_1 , price — \$100

chairs : number — x_2 , price — \$150

bookshelves : number — x_3 , price — \$200

dining tables : number — x_4 , price — \$400

$$\Rightarrow P(\text{Profit of all products}) = 100x_1 + 150x_2 + 200x_3 + 400x_4$$

and, $x_1 \geq 0, x_2 \geq 0, x_3 \geq 0, x_4 \geq 0$ (non-negative)

So, the objective function is to maximize P

$$\text{Maximize } P = \max(100x_1 + 150x_2 + 200x_3 + 400x_4)$$

Constraint Condition :

A nightstand \rightarrow 10 units of wood, 2 units of labor

A chair \rightarrow 12 units of wood, 4 units of labor

A bookshelf \rightarrow 25 units of wood, 8 units of labor

A dining table \rightarrow 20 units of wood, 2 units of labor

\Rightarrow Wood constraint: $10x_1 + 12x_2 + 25x_3 + 20x_4 \leq 5000$

Labor constraint: $2x_1 + 4x_2 + 8x_3 + 12x_4 \leq 1500$

$\Rightarrow \text{Max } P = \text{Max} (100x_1 + 150x_2 + 200x_3 + 400x_4)$

$$\begin{cases} 10x_1 + 12x_2 + 25x_3 + 20x_4 \leq 5000 \\ 2x_1 + 4x_2 + 8x_3 + 12x_4 \leq 1500 \\ x_1, x_2, x_3, x_4 \geq 0 \end{cases}$$

This problem falls into the class of Linear Programming problems.

And, it is linear optimization problem because both the objective function and the constraints are linear functions.

Finally, I applied MATLAB to find the solution as follows

```
编辑器 - C:\Users\Zhy20\Desktop\MATLAB\Optimization\Homework01.m
Homework01.m  +
1 % Define the coefficients of the objective function (need to be inverted to convert into a minimization problem)
2 f = [100 150 200 400];
3
4 % Define the coefficient matrix and the right-hand side values of the inequality constraints
5 A = [10 12 25 20; 2 4 8 12];
6 b = [5000; 1500];
7
8 % Define the limits of the variables (here the lower limit of all variables is 0, and the upper limit defaults to infinity)
9 lb = [0 0 0 0];
10 ub = [];
11
12 % Set variable types for integer programming, assuming all variables are integers
13 intcon = 1:4;
14
15 % Call intlinprog function to solve
16 [x, fval, exitflag, output] = intlinprog(f, intcon, A, b, [], [], lb, ub);
17
18 disp(x)
19
```

The solution is $x_1 = 372$, $x_2 = 3$, $x_3 = 0$, $x_4 = 62$

Therefore, the max sale profit is

$$372 \times 100 + 3 \times 150 + 0 \times 200 + 62 \times 400 = \$ 62450$$

3. Let the center of the circle be (a, b) and the radius be r .

The goal is to find the smallest radius r such that all point (x_i, y_i) are within or on the circle.

$$\Rightarrow \text{①} \min r \text{ subject to } (x_i - a)^2 + (y_i - b)^2 \leq r^2 \text{ for all } i = 1, \dots, N$$

$$\Rightarrow \text{②} \min r^2 \text{ subject to } (x_i - a)^2 + (y_i - b)^2 \leq r^2 \text{ for all } i = 1, \dots, N$$

Also, we can have a $2 \times N$ matrix $A = \begin{pmatrix} x_1 & x_2 & \cdots & x_N \\ y_1 & y_2 & \cdots & y_N \end{pmatrix}$

According a paper call The Minimum Covering Sphere Problem, by

D.J. Elzinga and D.W. Hearn, let $m = N$, $n = 2$, this smallest circle

problem can be formulated as a quadratic programming problem.

Based on Kuhn-Tucker conditions, which are both necessary and

sufficient for ②, assure the existence of multipliers v_i , $i = 1, \dots, n$

such that $\sum_{i=1}^N v_i = 1$, $v_i \geq 0$, $i = 1, \dots, N$ and $\bar{V} = (v_1, v_2, \dots, v_n)$

$$\Rightarrow \min \left[\bar{V}^T (A^T A) \bar{V} - \sum_{i=1}^N v_i (x_i^2 + y_i^2) \right] = \min f(\bar{V})$$

subject to $\sum_{i=1}^N v_i = 1$, $v_i \geq 0$, for $i = 1, \dots, N$

If $\bar{V} = (\bar{v}_1, \bar{v}_2, \dots, \bar{v}_n)$ is the solution of this quadratic programming problem, then the center of the circle is located at (a, b) where

$$\begin{cases} a = \bar{v}_1 x_1 + \bar{v}_2 x_2 + \cdots + \bar{v}_n x_n \\ b = \bar{v}_1 y_1 + \bar{v}_2 y_2 + \cdots + \bar{v}_n y_n \end{cases}$$

and the radius of the circle is given by $r = \sqrt{-f(\bar{V})}$

And, there is another solution method call Welzl's Algorithm.

The following part is the pseudocode of this Algorithm.

```
01. def welzl(points):
02.     """
03.     compute the minimum bounding sphere for a given set of points.
04.     points: point set
05.     return value: center and radius of the smallest enclosing circle
06.     """
07.
08.     # If the point set is empty, an empty bounding sphere is returned.
09.     if not points:
10.         return None
11.
12.     # If the point set has only one point, then that point is the center of the smallest enclosing sphere, with a radius of 0.
13.     if len(points) == 1:
14.         return Ball(points[0], 0)
15.
16.     # A point is randomly chosen as the center of the minimum bounding sphere.
17.     center = random.choice(points)
18.
19.     # Divide the remaining points into two groups: points inside the minimum bounding sphere and points outside the minimum bounding sphere.
20.     inside = []
21.     outside = []
22.     for point in points:
23.         if distance(point, center) <= radius:
24.             inside.append(point)
25.         else:
26.             outside.append(point)
27.
28.     # If all points are within the minimum bounding sphere, then the minimum bounding sphere is the final result.
29.     if not outside:
30.         return Ball(center, radius)
31.
32.     # Calculate the minimum enclosing sphere recursively among the remaining points.
33.     ball = welzl(outside)
34.
35.     # If the recursively obtained minimum bounding sphere contains the center point, then the minimum bounding sphere is the final result.
36.     if center in ball:
37.         return ball
38.
39.     # Otherwise, add the center point to the minimum bounding sphere and return the expanded minimum bounding sphere.
40.     else:
41.         return Ball(center, radius + distance(center, ball.center))
```

4. (a) Let $n=1$.

$$f(x) = f(x^*) + f'(x^*)(x - x^*) + \frac{1}{2} f''(z)(x - x^*)^2$$

where z is a point between x and x^*

Since $f'(x^*) = 0$

$$f(x) = f(x^*) + \frac{1}{2} f''(z)(x - x^*)^2$$

$$\Rightarrow f(x) - f(x^*) = \frac{1}{2} f''(z)(x - x^*)^2$$

Since $f''(x) > 0$ for all $x \in I$,

Therefore, $f(x) - f(x^*) > 0$ for all $x \neq x^*$

This inequality implies that $f(x) > f(x^*)$ for all $x \neq x^*$, which means that x^* is a strict global minimizer of f

on the open interval I .

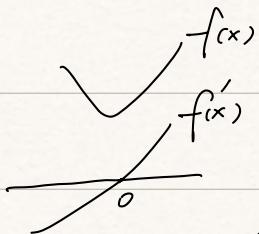
(b) Let $f(x) = e^{x^2}$,

$$\text{we can have } f'(x) = 2x e^{x^2}$$

$$f''(x) = 2e^{x^2} + 4x^2 e^{x^2} = 2e^{x^2}(1+2x^2)$$

Setting $f'(x^*) = 0$

$$2x^* e^{x^{*2}} = 0 \implies x^* = 0$$



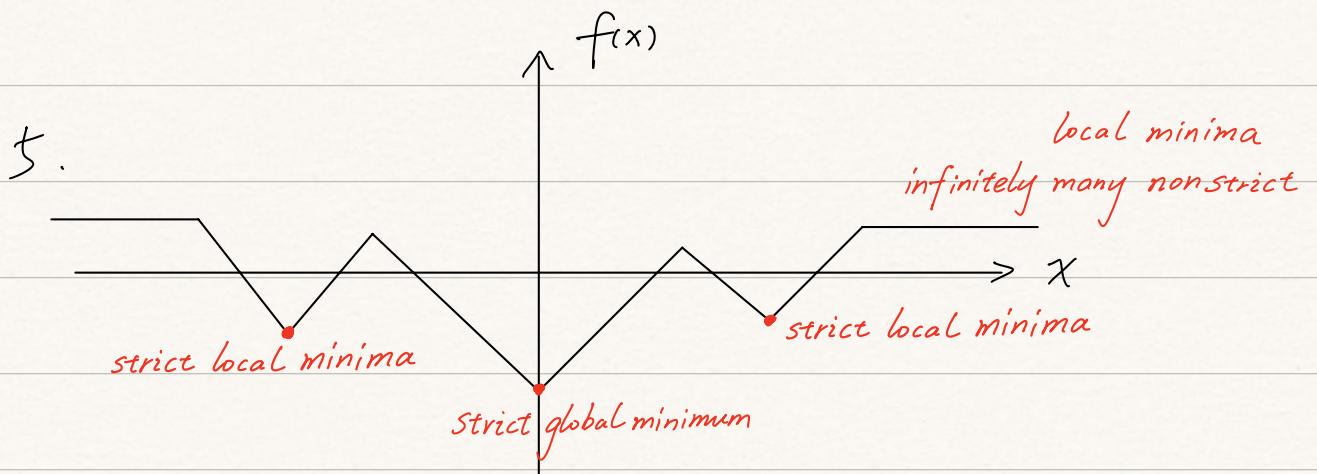
$$f''(x) = 2e^{x^2}(1+2x^2) > 0 \text{ for all } x \in \mathbb{R}$$

$f'(0) = 0$, so $x^* = 0$ is a stationary point

$f''(x) > 0$ and $f'(0) = 0$ implies that $f(x)$ is strictly convex

Therefore, $x^* = 0$ is a strict global minimizer of

$$f(x) = e^{x^2} \text{ on } \mathbb{R}$$



$$6. (a) f: \mathbb{R}^3 \rightarrow \mathbb{R}^2, \quad f(x, y, z) = (xz + y^2, \sin(xyz) + z)$$

$$\text{Suppose } f_1(x, y, z) = xz + y^2,$$

$$\frac{\partial f_1}{\partial x} = z, \quad \frac{\partial f_1}{\partial y} = 2y, \quad \frac{\partial f_1}{\partial z} = x$$

$$\text{Suppose } f_2(x, y, z) = \sin(xyz) + z$$

$$\frac{\partial f_2}{\partial x} = yz \cos(xyz), \quad \frac{\partial f_2}{\partial y} = xz \cos(xyz), \quad \frac{\partial f_2}{\partial z} = xy \cos(xyz) + 1$$

$$J_f(x, y, z) = \begin{pmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} & \frac{\partial f_1}{\partial z} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} & \frac{\partial f_2}{\partial z} \end{pmatrix} = \begin{pmatrix} z & 2y & x \\ yz \cos(xyz) & xz \cos(xyz) & xy \cos(xyz) + 1 \end{pmatrix}$$

$$J_f(0, -1, 1) = \begin{pmatrix} 1 & -2 & 0 \\ -1 & 0 & 1 \end{pmatrix}$$

(b) Suppose that there is a direction $\vec{v} = (v_1, v_2, v_3)^T$ in which the directional derivative of f at $(0, -1, 1)$ is zero.

$$\begin{aligned} D_{\vec{v}} f(0, -1, 1) &= J_f(0, -1, 1) \cdot \vec{v} \\ &= \begin{pmatrix} 1 & -2 & 0 \\ -1 & 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} \\ &= \begin{pmatrix} v_1 - 2v_2 \\ -v_1 + v_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \end{aligned}$$

$$\Rightarrow \begin{cases} v_1 - 2v_2 = 0 \\ -v_1 + v_3 = 0 \end{cases} \Rightarrow \begin{cases} v_1 = 2v_2 \\ v_1 = v_3 \end{cases}$$

Therefore, the direction vector \vec{v} that satisfies this condition is any vector of the form $\vec{v} = (2v_2, v_2, v_2)$

where v_2 can be any real number, for example $(2, 1, 2)$