

# 12. Extreme Values and Optimization Part 2

In this lecture, we will discuss

- Extreme Values of Functions of Two Variables
  - Local and Absolute Extreme Values of  $z = f(x, y)$
  - Critical points
  - Second Derivatives Test
  - Extreme Value Theorem for Functions of Two Variables
- Constrained Optimization: Lagrange Multipliers
  - Method of Lagrange Multipliers

## Extreme Values of Functions of Two Variables

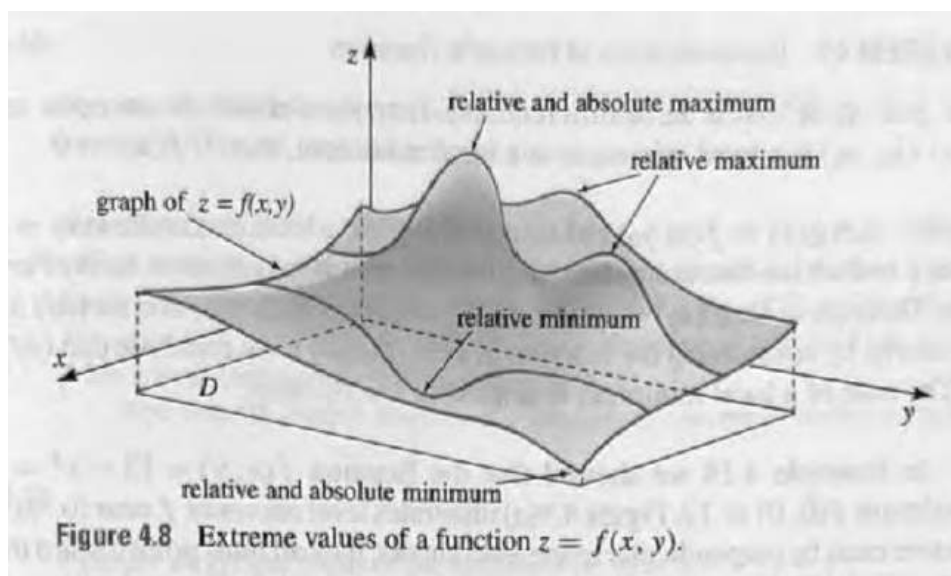
### Definition Local and Absolute Extreme Values of $z = f(x, y)$

A function of two variables has a **local maximum** at  $(a, b)$  if  $f(x, y) \leq f(a, b)$  when  $(x, y)$  is near  $(a, b)$ . [This means that  $f(x, y) \leq f(a, b)$  for all points  $(x, y)$  in some open ball with center  $(a, b)$ .]

The number  $f(a, b)$  is called a **local maximum** value. If  $f(x, y) \geq f(a, b)$  when  $(x, y)$  is near  $(a, b)$ , then  $f(a, b)$  is a local minimum value.

If the inequalities in the above definition hold for all points  $(x, y)$  in the domain of  $f$ , then  $f$  has an **absolute maximum** (or **absolute minimum**) at  $(a, b)$ .

Minimum and maximum values of a function are called **extreme values**.



### Theorem Generalization of Fermat's Theorem

If  $f$  has a local maximum or minimum at  $(a, b)$  and the first-order partial derivatives of  $f$  exist there, then  $f_x(a, b) = 0$  and  $f_y(a, b) = 0$ .

$$\nabla f(a, b) = \vec{0}$$

### Definition Critical Point

A point  $(a, b)$  is called a critical point of  $f$  if  $f_x(a, b) = 0$  and  $f_y(a, b) = 0$ , or if one of these partial derivatives does not exist.

**Example 1.** Find the critical points of the function  $f(x, y) = x^2 + y^2 + 2x - 6y + 5$ .

ANS: By def. we compute

$$f_x = \frac{\partial f}{\partial x} = 2x + 2$$

$$f_y = \frac{\partial f}{\partial y} = 2y - 6$$

To find the critical point(s) we set

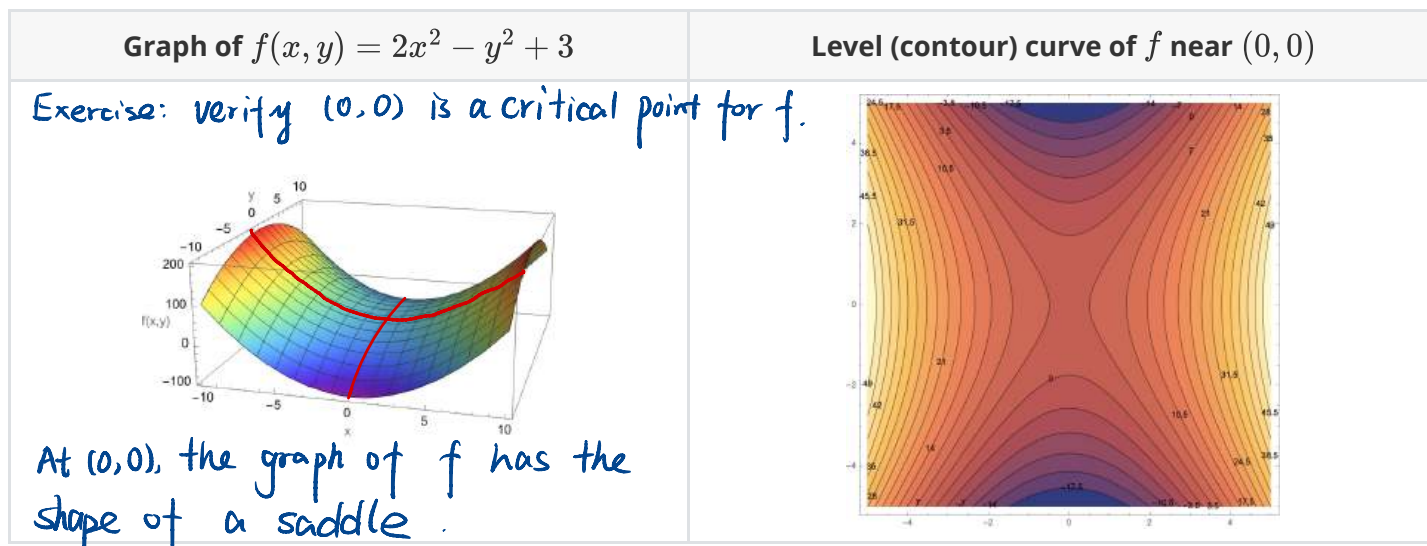
$$\begin{cases} f_x = 2x + 2 = 0 \\ f_y = 2y - 6 = 0 \end{cases} \Rightarrow \begin{cases} x = -1 \\ y = 3 \end{cases}$$

Thus the critical point for the give function

$f(x, y)$  is  $(-1, 3)$ .

## Definition. Saddle Point

A critical point that is neither a local minimum point nor a local maximum point is called a **saddle point**.



## Second Derivatives Test

Suppose the second partial derivatives of  $f$  are continuous on an open ball with center  $(a, b)$ , and suppose that  $f_x(a, b) = 0$  and  $f_y(a, b) = 0$  [that is,  $(a, b)$  is a critical point of  $f$ ]. Let

$$D = D(a, b) = f_{xx}(a, b)f_{yy}(a, b) - [f_{xy}(a, b)]^2$$

- (a) If  $D > 0$  and  $f_{xx}(a, b) > 0$ , then  $f(a, b)$  is a local minimum.
- (b) If  $D > 0$  and  $f_{xx}(a, b) < 0$ , then  $f(a, b)$  is a local maximum.
- (c) If  $D < 0$ , then  $f(a, b)$  is not a local maximum or minimum (i.e., it is a saddle point).

## Remarks.

1. If  $D = 0$ , the test gives no information:  $f$  could have a local maximum or local minimum at  $(a, b)$ , or  $(a, b)$  could be a saddle point of  $f$ .
2. To remember the formula for  $D$  it's helpful to write it as a determinant:

$$D = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix} = f_{xx}f_{yy} - (f_{xy})^2$$

The following matrix is called Hessian matrix

$$Hf(x, y) = \begin{bmatrix} f_{xx}(x, y) & f_{xy}(x, y) \\ f_{yx}(x, y) & f_{yy}(x, y) \end{bmatrix}$$

**Example 2.** Find the critical point of the function  $f(x, y) = 4x - 4y^2 - \ln(|x + y|)$ . Then use the Second Derivative Test to examine the critical point.

ANS: Step 1. Find the critical point(s):

$$\text{We set } \begin{cases} f_x = \frac{\partial f}{\partial x} = 4 - \frac{1}{x+y} = 0 \\ f_y = \frac{\partial f}{\partial y} = -8y - \frac{1}{x+y} = 0 \end{cases}$$

To solve this eqn, first note  $\frac{1}{x+y} = 4$  from the 1st eqn. Plug it into the 2nd one,

We have 
$$-8y - 4 = 0 \Rightarrow y = -\frac{1}{2}$$

Then plug  $y = -\frac{1}{2}$  into  $\frac{1}{x+y} = 4$  we have

$$4 = \frac{1}{x - \frac{1}{2}} \Rightarrow x - \frac{1}{2} = \frac{1}{4} \Rightarrow x = \frac{3}{4}$$

Thus the critical point is  $(\frac{3}{4}, -\frac{1}{2})$ .

Note  $f_x$  and  $f_y$  are not defined at points  $x+y=0$ . But we don't need to worry about them because  $f(x, y)$  is not defined on  $x+y$  since  $\ln|x+y|$  appears in  $f(x, y)$ .

Step 2. We use the second derivative test to classify the critical point.



Recall

$$D = \begin{vmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{vmatrix} = f_{xx}f_{yy} - f_{xy}^2$$

$\uparrow = f_{xy}$

We need to compute .

$$f_{xx}(x, y) = \frac{\partial}{\partial x} \left( 4 - \frac{1}{x+y} \right) = \frac{1}{(x+y)^2}$$

$$f_{yy}(x, y) = \frac{\partial}{\partial y} \left( -(8y + \frac{1}{x+y}) \right) = \frac{1}{(x+y)^2} - 8$$

$$f_{xy}(x, y) = \frac{\partial}{\partial y} \left( 4 - \frac{1}{x+y} \right) = \frac{1}{(x+y)^2}$$

Thus the discriminant

$$D(x, y) = f_{xx}f_{yy} - f_{xy}^2 = \frac{1}{(x+y)^2} \left( \frac{1}{(x+y)^2} - 8 \right) - \frac{1}{(x+y)^4} = -\frac{8}{(x+y)^2}$$

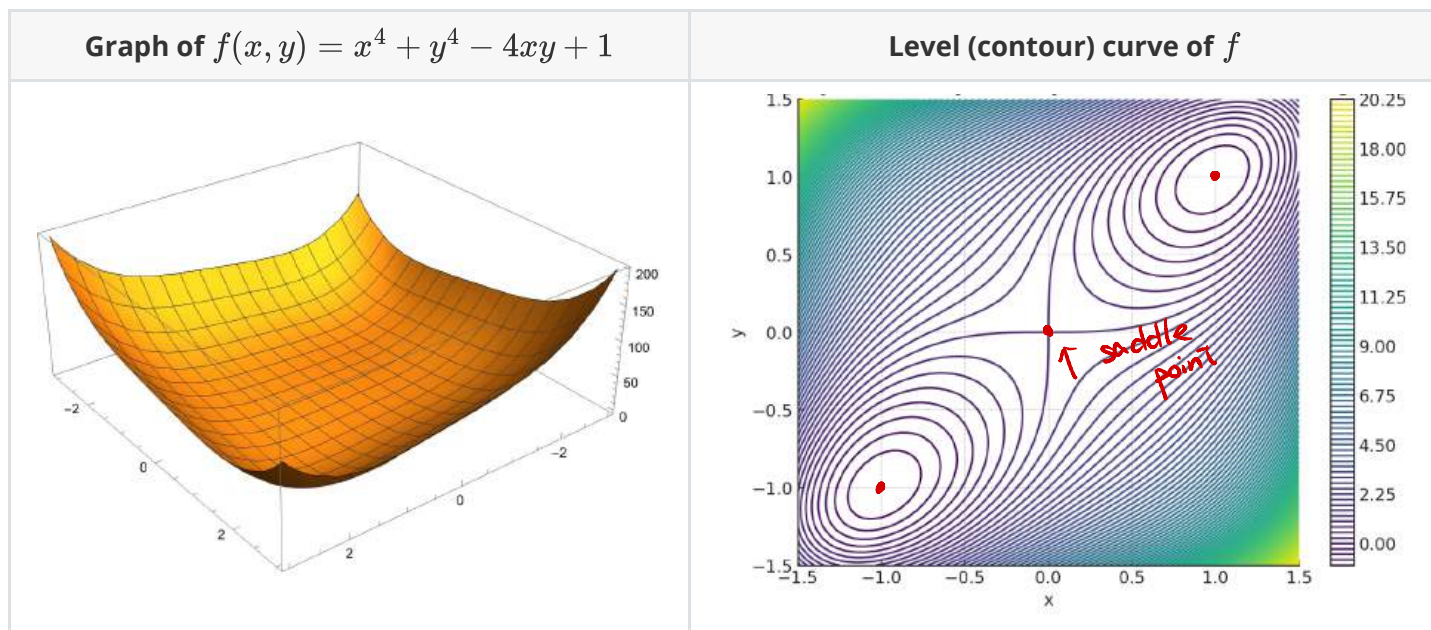
Applying the second derivative test, we have

$$D\left(\frac{3}{4}, -\frac{1}{2}\right) = -\frac{8}{\text{positive number}} < 0$$

Thus by the 2nd derivative test (case (c)),

we know  $f\left(\frac{3}{4}, -\frac{1}{2}\right)$  is a saddle point.

**Example 3.** Find the local maximum and minimum values and saddle points of  $f(x, y) = x^4 + y^4 - 4xy + 1$ .



ANS: We first compute the critical points. Let

$$\begin{cases} f_x = \cancel{4}x^3 - \cancel{4}y = 0 \\ f_y = \cancel{4}y^3 - \cancel{4}x = 0 \end{cases}$$

$$\Rightarrow \begin{cases} x^3 - y = 0 \Rightarrow y = x^3 \\ y^3 - x = 0 \end{cases}$$

We plug  $y = x^3$  into the 2nd eqn, then

$$x^9 - x = 0 \Rightarrow x(x^8 - 1) = x(x^4 + 1)(x^4 - 1) = x(x^4 + 1)(x^2 + 1)(x^2 - 1) = 0$$

Note  $x^2 + 1$  and  $x^4 + 1$  cannot be 0 for any  $x \in \mathbb{R}$ .

Thus the solutions are  $x = 0, 1, -1$ .

Then  $y = x^3 = 0, 1, -1$ .

Therefore the critical points are  
 $(0, 0), (1, 1), (-1, -1)$

Next, we compute the 2nd partial derivatives to get  $D(x, y)$ :

$$f_{xx} = 12x^2, \quad f_{xy} = -4, \quad f_{yy} = 12y^2.$$

Thus

$$D(x, y) = f_{xx}f_{yy} - f_{xy}^2 = 144x^2y^2 - 16$$

Then we examine each critical points:

- $(0, 0)$ :  $D(0, 0) = -16 < 0$ , thus by the second derivative test,  $(0, 0)$  is a saddle point.

- $(1, 1)$ :  $D(1, 1) > 0$  and  $f_{xx}(1, 1) = 12 > 0$ .

Thus from case (a) of the second derivative test,  $f(1, 1) = -1$  is a local minimum.

- $(-1, -1)$ :  $D(-1, -1) > 0$  and  $f_{xx}(-1, -1) = 12 > 0$ .

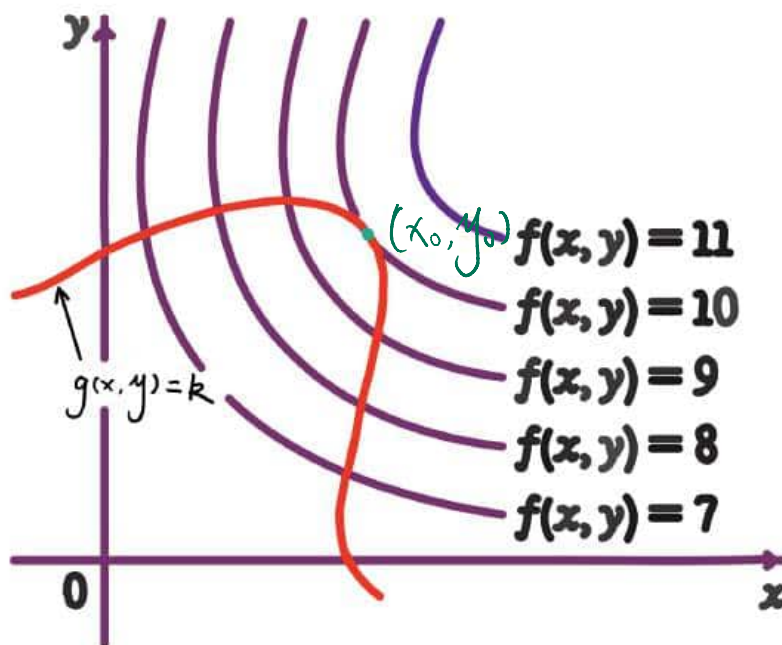
Thus  $f(-1, -1) = -1$  is also a local minimum.

## Lagrange Multipliers

Now we present Lagrange's method for maximizing or minimizing a general function  $f(x, y, z)$  subject to a constraint (or side condition) of the form  $g(x, y, z) = k$ .

Let's consider the geometric basis of Lagrange's method for functions of two variables.

- We want to find the extreme values of  $f(x, y)$  subject to a constraint of the form  $g(x, y) = k$ . That is, we seek the extreme values of  $f(x, y)$  when the point  $(x, y)$  is restricted to lie on the level curve  $g(x, y) = k$ .
- The following figure shows this curve together with several level curves of  $f$ .



- These have the equations  $f(x, y) = c$ , where  $c = 7, 8, 9, 10, 11$ .
- To maximize  $f(x, y)$  subject to  $g(x, y) = k$  is to find the largest value of  $c$  such that the level curve  $f(x, y) = c$  intersects  $g(x, y) = k$ .
- It appears from the figure that this happens when these curves just touch each other, that is, when they have a common tangent line. (Otherwise, the value of  $c$  could be increased further.)
- This means that the normal lines at the point  $(x_0, y_0)$  where they touch are identical. So the gradient vectors are parallel; that is,  $\nabla f(x_0, y_0) = \lambda \nabla g(x_0, y_0)$  for some scalar  $\lambda$ .

### Theorem Lagrange Multipliers for Functions of Two Variables

Let  $f, g : U \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be functions with continuous first derivatives. If the function  $f(x, y)$  has a local maximum or a local minimum subject to the constraint  $g(x, y) = k$  at  $\mathbf{x}_0 = (x_0, y_0)$ , and if  $\nabla g(\mathbf{x}_0) \neq \mathbf{0}$ , then  $\nabla f(x_0, y_0) = \lambda \nabla g(x_0, y_0)$ , for some real number  $\lambda$ .

**Remark.** The number in the above theorem is called a **Lagrange multiplier**.

We can generalize the above discussion to functions with 3 variables

### Method of Lagrange Multipliers for Functions of Three Variables

To find the maximum and minimum values of  $f(x, y, z)$  subject to the constraint  $g(x, y, z) = k$  [assuming that these extreme values exist and  $\nabla g \neq \mathbf{0}$  on the surface  $g(x, y, z) = k$ ]:

(a) Find all values of  $x, y, z$ , and  $\lambda$  such that

$$\begin{aligned}\nabla f(x, y, z) &= \lambda \nabla g(x, y, z) \\ g(x, y, z) &= k\end{aligned}$$

and

(b) Evaluate  $f$  at all the points  $(x, y, z)$  that result from step (a). The largest of these values is the maximum value of  $f$ ; the smallest is the minimum value of  $f$ .

### Example 7.

A company manufactures  $x$  units of one item and  $y$  units of another. The total cost in dollars,  $C$ , of producing these two items is approximated by the function

$$C = 5x^2 + 2xy + 3y^2 + 500.$$

(a) If the production quota for the total number of items (both types combined) is 30, find the minimum production cost.

$$g(x, y) = x + y = 30$$

(b) Estimate the additional production cost or savings if the production quota is raised to 31 or lowered to 29.

ANS: We want to minimize  $C(x, y)$  subject to

$$g(x, y) = x + y = 30$$

We compute

$$\nabla C(x, y) = \left( \frac{\partial C}{\partial x}, \frac{\partial C}{\partial y} \right) = (10x + 2y, 2x + 6y)$$

$$\nabla g(x, y) = \left( \frac{\partial g}{\partial x}, \frac{\partial g}{\partial y} \right) = (1, 1)$$

Use Lagrange multiplier method.

Step (a). We need to find all  $x, y, \lambda$  such that.

$$\begin{cases} \nabla C = \lambda \nabla g \\ g(x, y) = 30 \end{cases} \Rightarrow \begin{cases} (10x + 2y, 2x + 6y) = \lambda(1, 1) \\ x + y = 30 \end{cases}$$

$$\begin{cases} 10x + 2y = \lambda \\ 2x + 6y = \lambda \\ x + y = 30 \rightarrow y = 30 - x \end{cases}$$

$$\Rightarrow \begin{cases} 10x + 2(30 - x) = \lambda \\ 2x + 6(30 - x) = \lambda \end{cases}$$

$$\Rightarrow 8x + 60 = -4x + 180 \Rightarrow 12x = 120 \Rightarrow x = 10$$

Then  $y = 30 - 10 = 20$ . And  $\lambda = 2x + 6y = 140$

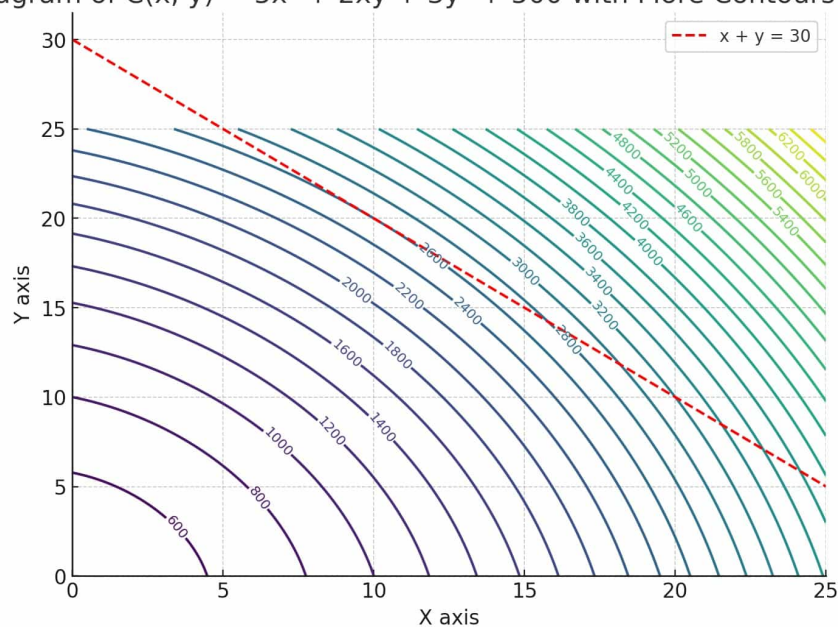
$$C(10, 20) = 5 \cdot (10)^2 + 2 \cdot (10) \cdot (20) + 3 \cdot (20)^2 + 500 = 2600 \text{ dollars.}$$

We can confirm  $(10, 20)$  is the min cost by either plug in a different value (say  $x=30, y=0$ ) and compare

$$\text{with } C(10, 20) = \$2600.$$

We can also draw the graph of  $C(x, y)$  using software

or website. Contour Diagram of  $C(x, y) = 5x^2 + 2xy + 3y^2 + 500$  with More Contours and  $x + y = 30$



(b). If  $g(x, y) = 31$ , we have

$$\begin{cases} 10x + 2y = \lambda \\ 2x + 6y = \lambda \\ x + y = 31 \end{cases} \quad \left. \begin{array}{l} \text{Note this part is roughly the same} \\ \text{Thus } \lambda \approx 140 \text{ as before.} \end{array} \right\}$$

Then  $\nabla C(x, y) \approx \lambda \nabla g(x, y)$ . this part is the same as (a)  $= \lambda(1, 1)$

Thus  $\nabla C(x, y) \approx (\lambda, \lambda)$

Therefore, the rate of change of  $C$  is roughly  $\lambda = 140$ .

Thus increasing production by 1, will cause cost increase by approximately \$140.

Similarly, decreasing production by 1, will save approx. \$140.

Comment. Another method is to solve explicitly the eqn.

$$\begin{cases} 10x + 2y = \lambda \\ 2x + 6y = \lambda \\ x + y = 31 \end{cases} \Rightarrow \begin{cases} x = \frac{31}{3} \\ y = \frac{62}{3} \end{cases} \quad \text{Then compute}$$

the corresponding  $C$ , and compare with (a).

We will get the cost increase by \$142.33.

Similarly, changing  $x + y = 29$ ,  $C$  will decrease by \$137.67.

Therefore, our previous estimation is good enough.