

# Positive Definite Matrices (正定矩阵)

## Lecture 27

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# Minima, Maxima, and Saddle Points

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# Minimum Point, Positive Definiteness

Questions:

1. Given a matrix, how could you determine the signs of the eigenvalues?
2. What test can guarantee that all of the eigenvalues of a matrix are positive?
3. How to find a minimum point efficiently? This arises throughout science and engineering and every problem of optimization. The mathematical problem is to move the second derivative test  $F'' > 0$  into  $n$  dimensions.

# Examples

Here are two examples:

$$F(x, y) = 7 + 2(x + y)^2 - y \sin y - x^3$$

$$f(x, y) = 2x^2 + 4xy + y^2$$

Does either  $F(x, y)$  or  $f(x, y)$  have a minimum at the point  $x = y = 0$ ?

## Remark 1

The zero-order terms  $F(0,0) = 7$  and  $f(0,0) = 0$  have no effect on the answer. They simply raise or lower the graphs of  $F$  and  $f$ .

The linear terms of  $F$  give a necessary condition:

To have any chance of a minimum, the first derivatives must vanish at  $x = y = 0$ :

$$\begin{aligned}\frac{\partial F}{\partial x} &= 4(x+y) - 3x^2 = 0 \quad \text{and} \\ \frac{\partial F}{\partial y} &= 4(x+y) - y \cos y - \sin y = 0\end{aligned}$$

## Remark 2

The linear terms of  $f$  give a necessary condition:

To have any chance of a minimum, the first derivatives must vanish at  $x = y = 0$ :

$$\begin{aligned}\frac{\partial f}{\partial x} &= 4(x+y) = 0 \quad \text{and} \\ \frac{\partial f}{\partial y} &= 4x - 2y = 0. \quad \text{All zero.}\end{aligned}$$

Thus  $(x, y) = (0, 0)$  is a stationary point for both functions. The surface  $z = F(x, y)$  is tangent to the horizontal plane  $z = 7$ , and the surface  $z = f(x, y)$  is tangent to the plane  $z = 0$ . The question is whether the graphs go above those planes or not, as we move away from the tangency point  $x = y = 0$ .

## Remark 3

The second derivatives at  $(0,0)$  are decisive:

$$\frac{\partial^2 F}{\partial x^2} = 4 - 6x = 4$$

$$\frac{\partial^2 F}{\partial x \partial y} = \frac{\partial^2 F}{\partial y \partial x} = 4$$

$$\frac{\partial^2 F}{\partial y^2} = 4 + y \sin y - 2 \cos y = 2$$

$$\frac{\partial^2 f}{\partial x^2} = 4$$

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial^2 f}{\partial y \partial x} = 4$$

$$\frac{\partial^2 f}{\partial y^2} = 2.$$

These second derivatives 4, 4, 2 contain the answer. Since they are the same for  $F$  and  $f$ , they must contain the same answer. The two functions behave in exactly the same way near the origin.  $F$  has a minimum if and only if  $f$  has a minimum. We are going to show that those functions don't!

## Remark 4

For  $F(x, y)$ :

- (a) The higher-degree terms in  $F$  have no effect on the question of a local minimum, but they can prevent it from being a global minimum.
- (b) In our example the term  $-x^3$  must sooner or later pull  $F$  toward  $-\infty$ .

For  $f(x, y)$  with no higher terms, all the action is at  $(0, 0)$ :

- (a) Every quadratic form  $f(x, y) = ax^2 + 2bxy + cy^2$  has a stationary point at the origin, where  $\frac{\partial f}{\partial x} = \frac{\partial f}{\partial y} = 0$ .
- (b) A local minimum would also be a global minimum.
- (c) The surface  $z = f(x, y)$  will then be shaped like a bowl, resting on the origin.



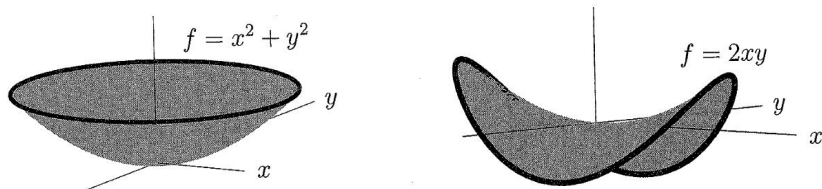
# Positive Definiteness

- (a) If the stationary point of  $F$  is at  $x = \alpha, y = \beta$ , the only change would be to use the second derivatives at  $(\alpha, \beta)$ :

$$f(x,y) = \frac{x^2}{2} \frac{\partial^2 F}{\partial x^2}(\alpha, \beta) + xy \frac{\partial^2 F}{\partial x \partial y}(\alpha, \beta) + \frac{y^2}{2} \frac{\partial^2 F}{\partial y^2}(\alpha, \beta)$$

- (b) This  $f(x,y)$  behaves near  $(0,0)$  in the same way that  $F(x,y)$  behaves near  $(\alpha, \beta)$ .
- (c) The third derivatives are drawn into the problem when the second derivatives fail to give a definite decision. That happens when the quadratic part is singular.
- (d) For a true minimum,  $f$  is allowed to vanish only at  $x = 0, y = 0$ . When  $f(x,y)$  is strictly positive at all other points (the bowl goes up), it is called **positive definite**.

## Figure 6.1



**Figure 6.1:** A bowl and a saddle: Definite  $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  and indefinite  $A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ .

The problem comes down to this: For a function of two variables  $x$  and  $y$ , what is the correct replacement for the condition  $\frac{\partial^2 F}{\partial x^2} > 0$ ? With only one variable, the sign of the second derivative decides between a minimum or a maximum. Now we have three derivatives:  $F_{xx}, F_{xy} = F_{yx}, F_{yy}$ . These three numbers must determine whether or not  $F$  has a minimum.

# Examples

**Example 1**  $f(x,y) = x^2 - 10xy + y^2$ . Here  $a = 1$  and  $c = 1$  are both positive. But  $f$  is not positive definite, because  $f(1,1) = -8$ . The conditions  $a > 0$  and  $c > 0$  ensure that  $f(x,y)$  is positive on the  $x$  and  $y$  axes. But this function is negative on the line  $x = y$ , because  $b = -10$  overwhelms  $a$  and  $c$ .

**Example 2** In our original  $f$  the coefficient  $2b = 4$  was positive. Does this ensure a minimum? Again the answer is no; the sign of  $b$  is of no importance! Even though its second derivatives are positive,  $2x^2 + 4xy + y^2$  is not positive definite. Neither  $F$  nor  $f$  has a minimum at  $(0,0)$  because  $f(1,-1) = 2 - 4 + 1 = -1$ . It is the size of  $b$ , compared to  $a$  and  $c$ , that must be controlled.

# General requirements

1. It is the size of  $b$ , compared to  $a$  and  $c$ , that must be controlled. We now want a necessary and sufficient condition for positive definiteness.
2. The simplest technique is to complete the square:

$$f(x, y) = ax^2 + 2bxy + cy^2 = a \left( x + \frac{b}{a}y \right)^2 + \left( c - \frac{b^2}{a} \right) y^2 \quad (1)$$

The first term on the right is never negative, when the square is multiplied by  $a > 0$ . But this square term must then be positive. That term has coefficient  $\frac{ac-b^2}{a}$ .

3. The last requirement for positive definiteness is that this coefficient must be positive:

If  $ax^2 + 2bxy + cy^2$  stays positive, then necessarily  $ac > b^2$ .

## Test for a minimum

The conditions  $a > 0$  and  $ac > b^2$  are just right. They guarantee  $c > 0$ . The right side of (1) is positive, and we have found a minimum.

### Theorem

$ax^2 + 2bxy + cy^2$  is positive definite if and only if  $a > 0$  and  $ac > b^2$ . Any  $F(x, y)$  has a minimum at a point  $(\alpha, \beta)$  where  $\frac{\partial F}{\partial x}(\alpha, \beta) = \frac{\partial F}{\partial y}(\alpha, \beta) = 0$  with

$$\frac{\partial^2 F}{\partial x^2}(\alpha, \beta) > 0 \text{ and } \left[ \frac{\partial^2 F}{\partial x^2}(\alpha, \beta) \right] \left[ \frac{\partial^2 F}{\partial y^2}(\alpha, \beta) \right] > \left[ \frac{\partial^2 F}{\partial x \partial y}(\alpha, \beta) \right]^2$$

Here

$$a = \frac{1}{2} \frac{\partial^2 F}{\partial x^2}(\alpha, \beta), \quad b = \frac{1}{2} \frac{\partial^2 F}{\partial x \partial y}(\alpha, \beta), \quad c = \frac{1}{2} \frac{\partial^2 F}{\partial y^2}(\alpha, \beta).$$

## Test for a Maximum

Since  $f$  has a maximum whenever  $-f$  has a minimum, we just reverse the signs of  $a, b$ , and  $c$ .

- This actually leaves  $ac > b^2$  unchanged: The quadratic form is negative definite if and only if  $a < 0$  and  $ac > b^2$ .
- The same change applies for a maximum of  $F(x, y)$ .

## Singular case $ac = b^2$

- (a) The second term in equation

$$f(x, y) = ax^2 + 2bxy + cy^2 = a \left( x + \frac{b}{a}y \right)^2 + \left( c - \frac{b^2}{a} \right) y^2$$

disappears to leave the first square—which is either positive semidefinite, when  $a > 0$ , or negative semidefinite, when  $a < 0$ .

- (b) The prefix semi allows the possibility that  $f$  can equal zero, as it will at the point  $x = b, y = -a$ .
- (c) The surface  $z = f(x, y)$  degenerates from a bowl into a valley. For  $f = (x + y)^2$ , the valley runs along the line  $x + y = 0$ .

# Saddle point

- (a) In one dimension,  $F(x)$  has a minimum or a maximum, or  $F'' = 0$ .
- (b) In two dimensions, a very important possibility still remains: The combination  $ac - b^2$  may be negative. This occurred in both examples, when  $b$  dominated  $a$  and  $c$ . It also occurs if  $a$  and  $c$  have opposite signs. Then two directions give opposite results—in one direction  $f$  increases, in the other it decreases.
- (c) It is useful to consider two special cases:

$$f_1(x, y) = 2xy \quad \text{and} \quad f_2(x, y) = x^2 - y^2$$

Saddle points at  $(0, 0)$ .

- (d) The saddles  $2xy$  and  $x^2 - y^2$  are practically the same; if we turn one through  $45^\circ$  we get the other. They are also hard to draw.



# Quadratic Forms

For any symmetric matrix  $A$ , the product  $x^T Ax$  is a pure **quadratic form**  $f(x_1, x_2, \dots, x_n)$ :

$$\begin{aligned} x^T Ax &= \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \cdot & \cdot & \cdot & \cdot \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \\ &= \sum_{i=1}^n \sum_{j=1}^n a_{ij} x_i x_j \end{aligned}$$

There are no higher-order terms or lower-order terms-only second-order. The function is zero at  $x = (0, \dots, 0)$ , and its first derivatives are zero.

# Examples

**Example 3**  $f = 2x^2 + 4xy + y^2$  and  $A = \begin{bmatrix} 2 & 2 \\ 2 & 1 \end{bmatrix} \rightarrow$  saddle point.

**Example 4**  $f = 2xy$  and  $A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \rightarrow$  saddle point.

**Example 5**  $A$  is 3 by 3 for  $2x_1^2 - 2x_1x_2 + 2x_2^2 - 2x_2x_3 + 2x_3^2$ :

$$f = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$f$  has a minimum at  $(0,0,0)$ .

# Test for positive definiteness

For function  $F(x_1, x_2, \dots, x_n)$ :  $F$  has a minimum when the pure quadratic  $x^T Ax$  is positive definite. These second-order terms control  $F$  near the stationary point:

$$F(x) = F(0) + x^T (\text{grad } F) + \frac{1}{2} x^T Ax + \text{higher order terms}$$

The next section contains the tests to decide whether  $x^T Ax$  is positive (the bowl goes up from  $x = 0$ ). Equivalently, the tests decide whether the matrix  $A$  is positive definite—which is the main goal of the chapter.

# Homework Assignment 27

6.1: 1, 2, 5, 12, 14, 17, 19.