Unconstrained optimization problem. Let $f : \mathbb{R}^n \to \mathbb{R}$. An unconstrained optimization problem is to minimize/maximize $f(\mathbf{x})$, where $\mathbf{x} = (x_1, \dots, x_n) \in \mathbb{R}^n$.

Necessary condition: If $\mathbf{x}^* \in \mathbb{R}^n$ is a local minimum or local maximum point of f over \mathbb{R}^n , then

$$(\nabla f)(\mathbf{x}^*) = 0$$
, that is, $\frac{\partial f}{\partial x_1}(\mathbf{x}^*) = \frac{\partial f}{\partial x_2}(\mathbf{x}^*) = \dots = \frac{\partial f}{\partial x_n}(\mathbf{x}^*) = 0$.

Sufficient condition: If $\mathbf{x}^* \in \mathbb{R}^n$ satisfies $(\nabla f)(\mathbf{x}^*) = 0$ and f is strictly convex (strictly concave) function in a neighbourhood of \mathbf{x}^* , then \mathbf{x}^* is a local minimum (local maximum) point of f over \mathbb{R}^n .

Consider the Hessian matrix

$$(\nabla^2 f)(\mathbf{x}^*) = \left[\left(\frac{\partial^2 f}{\partial x_i \partial x_j} \right) (\mathbf{x}^*) \right]_{n \times n}.$$

If the Hessian matrix is positive definite (negative definite), then f is strictly convex (strictly concave).

Note: The points \mathbf{x}^* satisfying $(\nabla f)(\mathbf{x}^*) = 0$ are called critical points.

n=1. Let $f: \mathbb{R} \to \mathbb{R}$. An unconstrained optimization problem is to minimize/maximize f(x), where $x \in \mathbb{R}$.

Necessary condition: If $x^* \in \mathbb{R}$ is a local minimum or local maximum point of f over \mathbb{R} , then

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

Sufficient condition: Let $f'(x^*) = f''(x^*) = \cdots = f^{(k-1)}(x^*) = 0$, but $f^{(k)}(x^*) \neq 0$. The point x^* is a

- (a) local minimum point if $f^{(k)}(x^*) > 0$ and k is even,
- (b) local maximum point if $f^{(k)}(x^*) < 0$ and k is even,
- (c) neither a local maximum or local minimum if k is odd.
- n=2. Let $f:\mathbb{R}^2\to\mathbb{R}$. An unconstrained optimization problem is to minimize/maximize f(x,y), where $(x,y)\in\mathbb{R}^2$.

Necessary condition: If $(x^*, y^*) \in \mathbb{R}^2$ is a local minimum or local maximum point of f over \mathbb{R}^2 , then

$$\left(\frac{\partial f}{\partial x}\right)(x^*, y^*) = \left(\frac{\partial f}{\partial y}\right)(x^*, y^*) = 0.$$

Sufficient condition: The point $(x^*, y^*) \in \mathbb{R}^2$ satisfying the necessary condition is a local minimum (local maximum) point if

$$\begin{bmatrix}
\left(\frac{\partial^2 f}{\partial x^2}\right)(x^*, y^*) & \left(\frac{\partial^2 f}{\partial x \partial y}\right)(x^*, y^*) \\
\left(\frac{\partial^2 f}{\partial y \partial x}\right)(x^*, y^*) & \left(\frac{\partial^2 f}{\partial y^2}\right)(x^*, y^*)
\end{bmatrix}$$
(Hessian matrix)

is positive definite (negative definite).

If (x^*, y^*) is a critical point, and if for every r > 0, there exist $(\alpha_1, \alpha_2), (\beta_1, \beta_2) \in \mathbb{R}^2$ satisfying $(\alpha_1 - x_1^*)^2 + (\alpha_2 - x_2^*)^2 < r^2$ and $(\beta_1 - x_1^*)^2 + (\beta_2 - x_2^*)^2 < r^2$ such that $f(\alpha_1, \alpha_2) > f(x_1^*, x_2^*)$ and $f(\beta_1, \beta_2) < f(x_1^*, x_2^*)$, then (x_1^*, x_2^*) is called a saddle point. If the determinant of Hessian matrix is negative, then the point (x^*, y^*) is a saddle point.

Note: A real symmetric matrix A of order n is said to be positive definite (P.D.) if $\mathbf{x}^T A \mathbf{x} > 0$ for all $\mathbf{x} \in \mathbb{R}^n$ with $\mathbf{x} \neq 0$. A real symmetric matrix A of order n is said to be negative definite (N.D.) if $\mathbf{x}^T A \mathbf{x} < 0$ for all $\mathbf{x} \in \mathbb{R}^n$ with $\mathbf{x} \neq 0$.

A real symmetric matrix A is positive definite if and only if all eigenvalues of A are positive. A real symmetric matrix A is negative definite if and only if all eigenvalues of A are negative.

A sufficient condition for a real symmetric matrix A to be positive definite is that all leading principal minors are positive, that is,

$$\begin{vmatrix} a_{11} > 0, & \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} > 0, & \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} > 0, \cdots, \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} > 0.$$

A sufficient condition for a real symmetric matrix A to be negative definite is that all leading principal minors alternate in sign starting from negative, that is,

$$a_{11} < 0, \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} > 0, \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} < 0, \begin{vmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{vmatrix} > 0, \cdots$$

Example. Find the natures of the extreme points of the function

$$f(x) = 12x^5 - 45x^4 + 40x^3 + 5.$$

Solution. The necessary condition for extreme points is

$$f'(x) = 60x^4 - 180x^3 + 120x^2 = 60x^2(x-1)(x-2) = 0.$$

The extreme points are x = 0, 1, 2.

$$f''(x) = 240x^3 - 540x^2 + 240x.$$

Now f''(1) = -60 < 0. So x = 1 is a local maximum point. Again f''(2) = 240 > 0. So x = 2 is a local minimum point. Since f''(0) = 0, we require

$$f'''(x) = 720x^2 - 1080x + 240.$$

Since $f'''(0) = 240 \neq 0$ and 3 is odd, x = 0 is neither a local minimum point nor a local maximum point.

Example. Find the natures of extreme points of the function

$$f(x,y) = x^3 + y^3 + 2x^2 + 4y^2 + 6.$$

Solution. The necessary conditions for extreme points are

$$\frac{\partial f}{\partial x} = 3x^2 + 4x = 0$$
, and $\frac{\partial f}{\partial y} = 3y^2 + 8y = 0$.

The extreme points are (0,0), (0,-8/3), (-4/3,0), (-4/3,-8/3). The Hessian matrix is

$$H = \begin{bmatrix} 6x + 4 & 0 \\ 0 & 6y + 8 \end{bmatrix}.$$

The leading principal minors of H are $H_1 = 6x + 4$ and $H_2 = (6x + 4)(6y + 8)$.

- 1. For (0,0), $H_1 = 4$, $H_2 = 32$. So H is positive definite and hence (0,0) is a local minimum point.
- 2. For (0, -8/3), det H = -32 < 0. So (0, -8/3) is a saddle point.
- 3. For (-4/3, 0), det H = -32 < 0. So (-4/3, 0) is a saddle point.
- 4. For (-4/3, 8/3), $H_1 = -4$, $H_2 = 32$. So H is negative definite and hence (-4/3, -8/3) is a local maximum point.

Note: We have $f\left(-\frac{4}{3}, r\right) - f\left(-\frac{4}{3}, 0\right) = r^3 + 4r^2 > 0$ for all r > 0. Again for every 0 < r < 2,

$$f\left(-\frac{4}{3}+r,0\right) - f\left(-\frac{4}{3},0\right) = \left(-\frac{4}{3}+r\right)^3 + 2\left(-\frac{4}{3}+r\right)^2 - \left(-\frac{4}{3}\right)^3 - 2\left(-\frac{4}{3}\right)^2$$

$$= r\left[\left(-\frac{4}{3}+r\right)^2 - \frac{4}{3}\left(-\frac{4}{3}+r\right) + \left(-\frac{4}{3}\right)^2\right] + 2r\left(-\frac{8}{3}+r\right)$$

$$= r\left(\frac{16}{9} - \frac{8}{3}r + r^2 + \frac{16}{9} - \frac{4}{3}r + \frac{16}{9} - \frac{16}{3} + 2r\right)$$

$$= r(r^2 - 2r) < 0.$$

Therefore every neighbourhood of $\left(-\frac{4}{3},0\right)$ contain two points such that the value of f at one point is larger than $f\left(-\frac{4}{3},0\right)$, and the value of f at other point is smaller than $f\left(-\frac{4}{3},0\right)$. Therefore $\left(-\frac{4}{3},0\right)$ is a saddle point. Similar calculations can be done to show that $\left(0,-\frac{8}{3}\right)$ is a saddle point.

Exercises.

- 1. Find the critical points and their natures for the following functions.
 - (a) $f(x) = x^2 6x^2 + 9x + 5$.
 - (b) $f(x) = 2 + (x 1)^4$.
- 2. Find two numbers whose difference is 100 and whose product is a minimum.
- 3. Find two positive numbers whose product is 100 and whose sum is a minimum.
- 4. Find the point on the parabola $y^2 = 2x$ that is closest to the point (1,4).
- 5. Find the area of the largest rectangle that can be inscribed in a semicircle of radius 5. (The word inscribed means that the rectangle has two vertices on the semicircle and two vertices on the diameter.)
- 6. Find the critical points and their natures for the following functions.
 - (a) $f(x,y) = x^3 y^3 2xy + 6$.
 - (b) $f(x,y) = x^4 2x^2 + y^3 3y$.
 - (c) $f(x,y) = x^2 + y^4 + 2xy$.
 - (d) $f(x,y) = y \cos x$.