

Math 5B

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Suppose  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  is a smooth function.

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix} : \mathbb{R}^n \rightarrow \mathbb{R}$$

$\vec{c} = (c_1, \dots, c_n)$  is a critical point for  $f$  if  $\nabla f(\vec{c}) = \vec{0}$

Example:  $f(x, y) = \frac{1}{2}y^2 - \cos x$

$$\nabla f = \begin{pmatrix} \sin x \\ y \end{pmatrix} = \vec{0} \Rightarrow \begin{cases} \sin x = 0 \\ y = 0 \end{cases}$$

$\sin x = 0 \Rightarrow x = n\pi$ , where  $n \in \mathbb{Z} = \{\dots, -2, -1, 0, 1, 2, \dots\}$

CRITICAL POINTS ARE CANDIDATES FOR MAXIMA AND MINIMA.

SECOND DERIVATIVE TEST?

If  $\vec{c} = (c_1, \dots, c_n)$  is a critical point for  $f$ , the Hessian

matrix at  $\vec{c}$  is

$$A = \begin{pmatrix} \frac{\partial}{\partial x_1} \left( \frac{\partial f}{\partial x_1} \right) & \dots & \frac{\partial}{\partial x_n} \left( \frac{\partial f}{\partial x_1} \right) \\ \vdots & & \vdots \\ \frac{\partial}{\partial x_1} \left( \frac{\partial f}{\partial x_n} \right) & \dots & \frac{\partial}{\partial x_n} \left( \frac{\partial f}{\partial x_n} \right) \end{pmatrix} (\vec{c})$$

Example:  $f(x, y) = \frac{1}{2}y^2 - \cos x$

$$\nabla f = \begin{pmatrix} -\sin x \\ y \end{pmatrix}$$

Critical points  $(n\pi, 0)$ ,  $n \in \mathbb{Z}$

$$\begin{aligned} A &= \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix} (n\pi, 0) = \begin{pmatrix} -\cos x & 0 \\ 0 & 1 \end{pmatrix} (n\pi, 0) \\ &= \begin{pmatrix} (-1)^n & 0 \\ 0 & 1 \end{pmatrix} \end{aligned}$$

Equality of mixed partials  $\Rightarrow \frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i} \Rightarrow A = A^T$

THEOREM. If  $A$  is an  $n \times n$  matrix such that  $A^T = A$ , then  $A$  has  $n$  REAL eigenvalues  $\lambda_1, \dots, \lambda_n$  counted with multiplicity. Moreover  $\exists$  an  $n \times n$  matrix  $B$  such that  $B^T = B^{-1}$  and

$$B^{-1}AB = \begin{pmatrix} \lambda_1 & 0 & & 0 \\ 0 & \lambda_2 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & & \lambda_n \end{pmatrix}$$

SECOND DERIVATIVE TEST:

Let  $\vec{c}$  be a critical point for  $f$ ,  $A$  its Hessian

matrix at  $\vec{c}$ .

If  $\det A \neq 0$ ,  $\vec{c}$  is a nondegenerate critical point.

Otherwise it is degenerate.

Suppose  $\vec{c}$  is a nondegenerate critical point

$\vec{c}$  is a local minimum if eigenvalues of  $A$  are positive

$\vec{c}$  " " local maximum if " " " " " negative.

$\vec{c}$  is a saddle point of index  $k$ ,  $0 < k < n$

if  $k$  eigenvalues are  $< 0$   
 $n-k$  eigenvalues are  $> 0$

Example:  $f(x, y) = \frac{1}{2}y^2 - cx^2$

Critical points:  $(n\pi, 0)$ ,  $n \in \mathbb{Z}$

$$A = \begin{pmatrix} (-1)^n & 0 \\ 0 & -1 \end{pmatrix}$$

$(n\pi, 0)$  is a local minimum when  $n$  even

$(n\pi, 0)$  is a saddle point when  $n$  odd.

EXAMPLE :  $f(x, y) = 2x^2 + 5xy + 2y^2 - 4x - 5y$

CRITICAL POINTS ?

$$\begin{cases} \frac{\partial f}{\partial x} = 4x + 5y - 4 = 0 \\ \frac{\partial f}{\partial y} = 5x + 5y - 5 = 0 \end{cases}$$

$$\begin{cases} 4x + 5y = 4 \\ 5x + 4y = 5 \end{cases}$$

CRAMER'S RULE :  $x = \frac{\begin{vmatrix} 4 & 5 \\ 5 & 4 \end{vmatrix}}{\begin{vmatrix} 4 & 5 \\ 5 & 4 \end{vmatrix}} = 1$

$$y = \frac{\begin{vmatrix} 4 & 4 \\ 5 & 5 \end{vmatrix}}{\begin{vmatrix} 4 & 5 \\ 5 & 4 \end{vmatrix}} = 0$$

$$A = \begin{pmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{pmatrix} = \begin{pmatrix} 4 & 5 \\ 5 & 4 \end{pmatrix}$$

Eigenvalues ?

$$\begin{vmatrix} 4-\lambda & 5 \\ 5 & 4-\lambda \end{vmatrix} = 0 \quad (4-\lambda)^2 - 5^2 = 0 \quad \lambda - 4 = \pm 5$$

$$\lambda = 9 \text{ OR } -1$$

$\therefore (1, 0)$  is a saddle point.