

**Topic(s):** determining whether a critical point is a local minimum, local maximum, or a saddle point

Suppose  $\mathbf{a}$  is a critical point of  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , i.e.  $\nabla f(\mathbf{a}) = \mathbf{0}$ . Then, the quadratic approximation near  $\mathbf{a}$  is

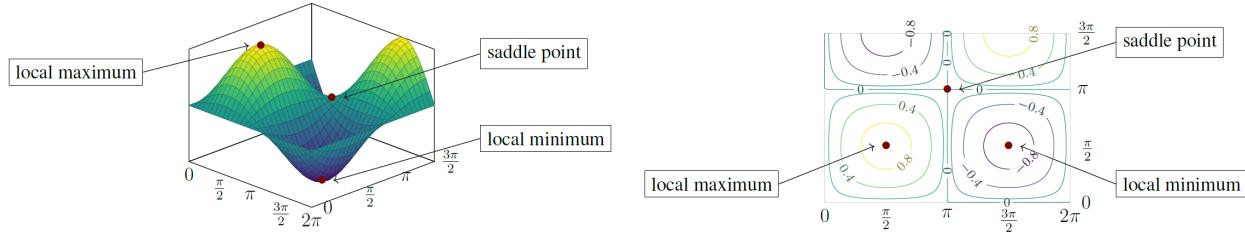
$$f(\mathbf{a} + \mathbf{h}) \approx f(\mathbf{a}) + \frac{1}{2} \mathbf{h}^\top Hf(\mathbf{a}) \mathbf{h} = f(\mathbf{a}) + \frac{1}{2} q_{Hf(\mathbf{a})}(\mathbf{h}).$$

Note that if  $f$  is a function from  $\mathbb{R}$  to  $\mathbb{R}$ , then

$$f(a+h) \approx f(a) + \frac{1}{2} f''(a)h^2.$$

Hence, if  $f''(a) > 0$ , then  $f(a+h) > f(a)$  for small values of  $h$ , making  $a$  a local minimum, and if  $f''(a) < 0$ , then  $f(a+h) < f(a)$  for small values of  $h$ , making  $a$  a local maximum.

Similarly, if  $q_{Hf(\mathbf{a})}$  is positive-definite, then  $f(\mathbf{a} + \mathbf{h}) > f(\mathbf{a})$ , making  $\mathbf{a}$  a local minimum, and if  $q_{Hf(\mathbf{a})}$  is negative-definite, then  $f(\mathbf{a} + \mathbf{h}) < f(\mathbf{a})$ , making  $\mathbf{a}$  a local maximum. On the other hand, if  $q_{Hf(\mathbf{a})}$  is indefinite, then there are small  $\mathbf{h}_1$  and  $\mathbf{h}_2$  for which  $q_{Hf(\mathbf{a})}(\mathbf{h}_1) > 0$  and  $q_{Hf(\mathbf{a})}(\mathbf{h}_2) < 0$ . In other words,  $f(\mathbf{a} + \mathbf{h}_2) < f(\mathbf{a}) < f(\mathbf{a} + \mathbf{h}_1)$ , making  $\mathbf{a}$  a saddle point.



For  $n > 1$  and a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , a critical point  $\mathbf{a} \in \mathbb{R}^n$  is called a **saddle point of  $f$**  if there are different lines  $L$  and  $L'$  in  $\mathbb{R}^n$  through  $\mathbf{a}$  so that  $f$  evaluated just on the line  $L$  has a local maximum at  $\mathbf{a}$ , and  $f$  evaluated just on the line  $L'$  has a local minimum at  $\mathbf{a}$ .

**Theorem 26.1.5. (Second Derivative Test, Version 1)** Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a function and let  $\mathbf{a} \in \mathbb{R}^n$  be a critical point of  $f$ .

- (a) If the Hessian  $Hf(\mathbf{a})$  is positive-definite, then  $\mathbf{a}$  is a local minimum of  $f$ .
- (b) If the Hessian  $Hf(\mathbf{a})$  is negative-definite, then  $\mathbf{a}$  is a local maximum of  $f$ .
- (c) If the Hessian  $Hf(\mathbf{a})$  is indefinite, then  $\mathbf{a}$  is a saddle point of  $f$  and is neither a local minimum nor a local maximum of  $f$ .

If the Hessian  $Hf(\mathbf{a})$  is positive-semidefinite or negative-semidefinite (without being definite), then we need more information, just as the single-variable second derivative test is inconclusive when the second derivative vanishes.

It is easy to check the definiteness of diagonal matrices.

**Example 1.**

$$A = \begin{bmatrix} 11 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 4 \end{bmatrix} \quad B = \begin{bmatrix} -3 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -3 \end{bmatrix} \quad C = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & -11 \end{bmatrix}$$

$q_A(x, y, z) = 11x^2 + 3y^2 + 4z^2$  is positive definite.

$q_B(x, y, z) = -3x^2 - y^2 - 3z^2$  is negative definite.

$q_C(x, y, z) = 3x^2 - 4y^2 - 11z^2$  is indefinite.

**Example 2.** Let  $f(x, y) = x^4 + y^4$ ,  $g(x, y) = x^4 - y^4$ , and  $h(x, y) = -x^4 - y^4$ . What type of critical point is  $(0, 0)$  for  $f$ ,  $g$ , and  $h$ ?

$$\text{Since } Hf = \begin{bmatrix} 12x^2 & 0 \\ 0 & 12y^2 \end{bmatrix}, \quad Hg = \begin{bmatrix} 12x^2 & 0 \\ 0 & -12y^2 \end{bmatrix}, \quad \text{and } Hh = \begin{bmatrix} -12x^2 & 0 \\ 0 & -12y^2 \end{bmatrix},$$

$$Hf(0, 0) = Hg(0, 0) = Hh(0, 0) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix},$$

which is not definite.

For  $f$ ,  $(0, 0)$  is a local minimum since  $f(x, y) = x^4 + y^4 \geq 0 = f(0, 0)$  for all  $(x, y) \in \mathbb{R}^2$ .

For  $g$ , on  $x=0$ ,  $g(0, y) = -y^4$  and on  $y=0$ ,  $g(x, 0) = x^4$ . Hence,  $(0, 0)$  is a saddle point as it is a local minimum on  $y=0$  and a local maximum on  $x=0$ .

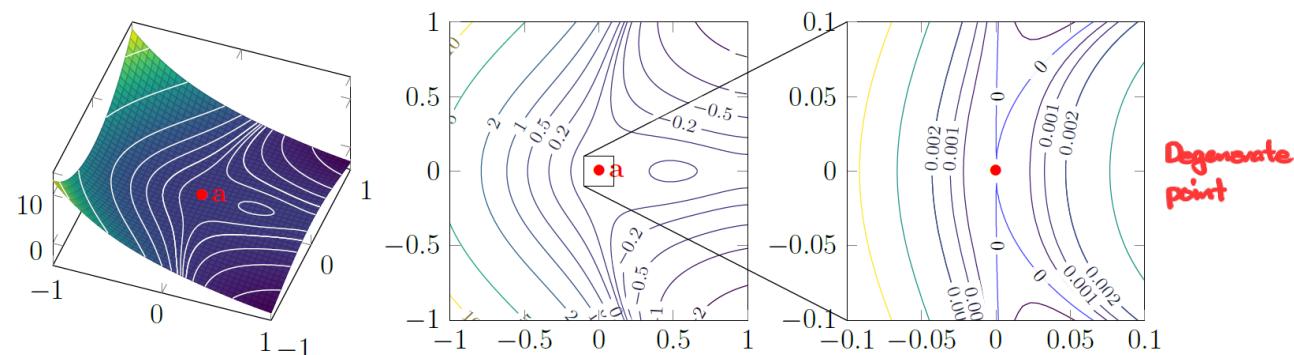
For  $h$ ,  $(0, 0)$  is a local maximum since  $h(x, y) = -x^4 - y^4 \leq 0 = h(0, 0)$  for all  $(x, y) \in \mathbb{R}^2$ .

**Example 3.** Consider the function  $f(x, y) = (x^2 + y^2 - 3x)^2 + 4x^2(x - 2)$ . What happens at  $(0, 0)$ ?

We compute  $\nabla f = \begin{bmatrix} 4x^3 - 6x^2 + 4xy^2 + 2x - 6y \\ 4x^2y - 12xy + 4y^3 \end{bmatrix}$ , and so,  $(0, 0)$  is a critical point.

Next,  $Hf = \begin{bmatrix} 12x^2 - 12x + 4y^2 + 2 & 8xy - 12y \\ 8xy - 12y & 4x^2 - 12x + 12y^2 \end{bmatrix}$  and  $Hf(0, 0) = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$ .

$Hf(0, 0)$  has eigenvectors  $\vec{w}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\vec{w}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  with eigenvalues  $\lambda_1 = 2$  and  $\lambda_2 = 0$ , respectively.  $Hf(0, 0)$  is positive semidefinite.



We can also use the diagonalization formula to determine the nature of a critical point. If  $Hf(\mathbf{a})$  has orthogonal eigenvectors  $\mathbf{w}_1$  and  $\mathbf{w}_2$  (with corresponding eigenvalues  $\lambda_1$  and  $\lambda_2$ , respectively), then

$$f(\mathbf{a} + h_1\mathbf{w}_1 + h_2\mathbf{w}_2) \approx f(\mathbf{a}) + \frac{1}{2}q_{Hf(\mathbf{a})}(h_1\mathbf{w}_1 + h_2\mathbf{w}_2) = f(\mathbf{a}) + \frac{1}{2}\lambda_1(\mathbf{w}_1 \cdot \mathbf{w}_1)h_1^2 + \frac{1}{2}\lambda_2(\mathbf{w}_2 \cdot \mathbf{w}_2)h_2^2.$$

From this, we can tell that the definiteness of the Hessian at  $\mathbf{a}$  is determined by the sign of the eigenvalues.

**Theorem 26.3.1. (Second Derivative Test, Version 2)** Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be a function and let  $\mathbf{a} \in \mathbb{R}^n$  be a critical point of  $f$ .

- (a) If the Hessian  $Hf(\mathbf{a})$  has all eigenvalues positive, then  $\mathbf{a}$  is a local minimum of  $f$ .
- (b) If the Hessian  $Hf(\mathbf{a})$  has all eigenvalues negative, then  $\mathbf{a}$  is a local maximum of  $f$ .
- (c) If the Hessian  $Hf(\mathbf{a})$  has both a positive eigenvalue and a negative eigenvalue (some other eigenvalue may be 0), then  $\mathbf{a}$  is a saddle point of  $f$ .

In the  $2 \times 2$  case, there is a quick way to determine the signs of the eigenvalues:

**Theorem 26.3.3.** Consider a symmetric  $2 \times 2$  matrix  $A = \begin{bmatrix} a & b \\ b & d \end{bmatrix}$ , so the eigenvalues are the roots of the characteristic polynomial  $P_A(\lambda) = \lambda^2 - \text{tr}(A)\lambda + \det(A)$ . For its two roots  $\lambda_1$  and  $\lambda_2$ ,

- (a)  $\lambda_1$  and  $\lambda_2$  have opposite signs precisely when the product  $\det(A) = \lambda_1\lambda_2$  is negative; so the indefinite case occurs exactly when  $ad - b^2 < 0$ .
- (b)  $\lambda_1$  and  $\lambda_2$  are either both positive or negative precisely when their product  $\det(A) = \lambda_1\lambda_2$  is positive.
- (c) In case (b), the common sign of  $\lambda_1$  and  $\lambda_2$  is the same as that of their sum  $\lambda_1 + \lambda_2 = a + d$ .

**Example 4.** Determine the nature of the quadratic form  $10x_1^2 - 14x_1x_2 + 5x_2^2$ .

The unique symmetric matrix associated is  $A = \begin{bmatrix} 10 & -7 \\ -7 & 5 \end{bmatrix}$ . Since  $\text{tr}(A) = 15$  and  $\det(A) = 1$ ,  $A$  is positive-definite.

**Example 5.** Analyze the behavior of  $f(x, y) = x^3 - 3x^2 - 6xy + 9x + 3y^2$  at each of its critical points.

We get  $\nabla f = \begin{bmatrix} 3x^2 - 6x - 6y + 9 \\ -6x + 6y \end{bmatrix}$  and  $Hf = \begin{bmatrix} 6x - 6 & -6 \\ -6 & 6 \end{bmatrix}$ . Hence, the critical points are  $(1, 1)$  and  $(3, 3)$ . At  $(1, 1)$ ,  $Hf = \begin{bmatrix} 0 & -6 \\ -6 & 6 \end{bmatrix}$  has trace 6 and determinant  $-36$ . Hence,  $Hf(1, 1)$  is indefinite and  $(1, 1)$  is a saddle point. At  $(3, 3)$ ,  $Hf = \begin{bmatrix} 12 & -6 \\ -6 & 6 \end{bmatrix}$  has trace 18 and determinant 36. So,  $Hf(3, 3)$  is positive-definite and  $(3, 3)$  is a local minimum.

**Example 6.** Consider the function  $f(x, y) = x^3 - 4x^2 + x^2y - 3xy$ .

- (a) Determine the eigenvalues of the Hessian of  $f$  at the critical points, and compute an eigenvector for each eigenvalue for each critical point.

We get  $\nabla f = \begin{bmatrix} 3x^2 - 8x + 2xy - 3y \\ x^2 - 3x \end{bmatrix}$  and  $Hf = \begin{bmatrix} 6x - 8 + 2y & 2x - 3 \\ 2x - 3 & 0 \end{bmatrix}$ . Hence, the critical points are  $(0, 0)$  and  $(3, -1)$ .

$Hf(0, 0) = \begin{bmatrix} -8 & -3 \\ -3 & 0 \end{bmatrix}$  has eigenvalues  $\lambda_1 = -9$  and  $\lambda_2 = 1$  with corresponding eigenvectors  $\vec{w}_1 = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$  and  $\vec{w}_2 = \begin{bmatrix} -1 \\ 3 \end{bmatrix}$ , respectively.

$Hf(3, -1) = \begin{bmatrix} 8 & 3 \\ 3 & 0 \end{bmatrix}$  has eigenvalues  $\lambda_1 = -1$  and  $\lambda_2 = 9$  with corresponding eigenvectors  $\vec{w}_1 = \begin{bmatrix} 1 \\ -3 \end{bmatrix}$  and  $\vec{w}_2 = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$ , respectively.

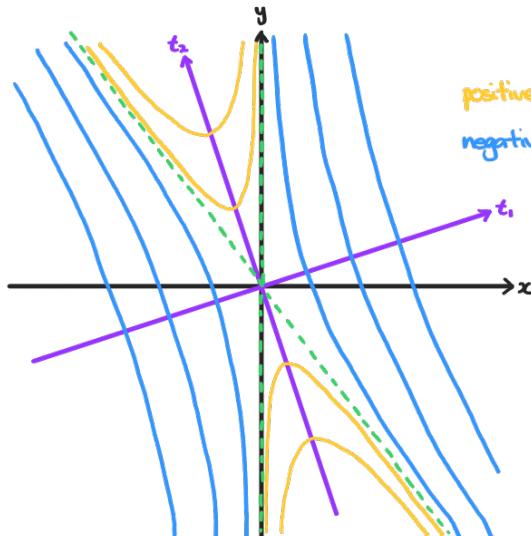
- (b) Use the information from part (a) to sketch the contour plot of  $f$  near each critical point.

$(0, 0)$

$$\vec{w}'_1 = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ 1 \end{bmatrix}, \quad \vec{w}'_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} -1 \\ 3 \end{bmatrix}.$$

If  $\vec{v} = t_1 \vec{w}'_1 + t_2 \vec{w}'_2$ , then

$$g_{Hf(0,0)}(\vec{v}) = -9t_1^2 + t_2^2.$$

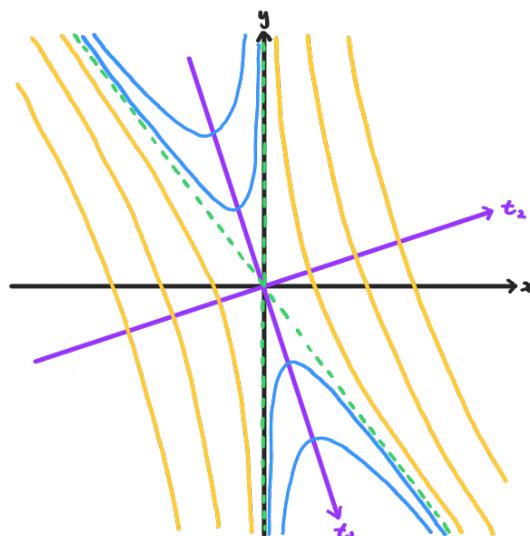


$(3, -1)$

$$\vec{w}'_1 = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ -3 \end{bmatrix}, \quad \vec{w}'_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

If  $\vec{v} = t_1 \vec{w}'_1 + t_2 \vec{w}'_2$ , then

$$g_{Hf(3,-1)}(\vec{v}) = -t_1^2 + 9t_2^2.$$



**Example 7.** Let  $f(x, y) = x^3 + 3x^2 + y^2 - 4y$ .

- (a) Find all critical points of  $f(x, y)$ .

We compute  $\nabla f = \begin{bmatrix} 3x^2 + 6x \\ 2y - 4 \end{bmatrix}$ , and so, the critical points are  $(0, 2)$  and  $(-2, 2)$ .

- (b) Compute the Hessian matrix at each of these points, and for each, determine if it is a local maximum, local minimum, or a saddle point.

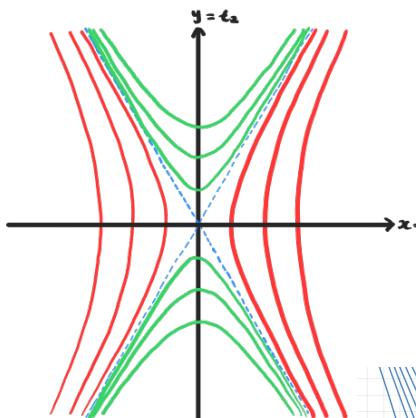
We compute  $H_f = \begin{bmatrix} 6x+6 & 0 \\ 0 & 2 \end{bmatrix}$ ,  $H_f(0, 2) = \begin{bmatrix} 6 & 0 \\ 0 & 2 \end{bmatrix}$  and  $H_f(-2, 2) = \begin{bmatrix} -6 & 0 \\ 0 & 2 \end{bmatrix}$ .

Hence,  $(0, 2)$  is a local minimum, and  $(-2, 2)$  is a saddle point.

- (c) Sketch the contour plot of  $f$  near each critical point.

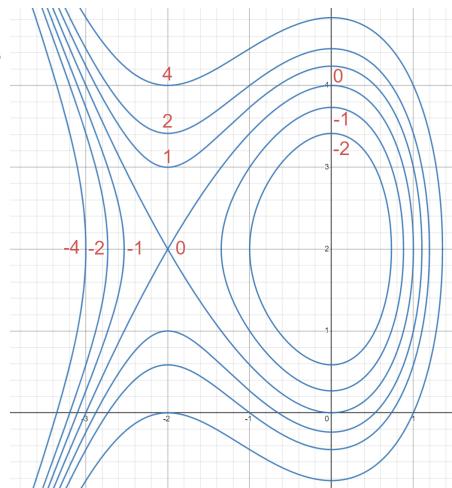
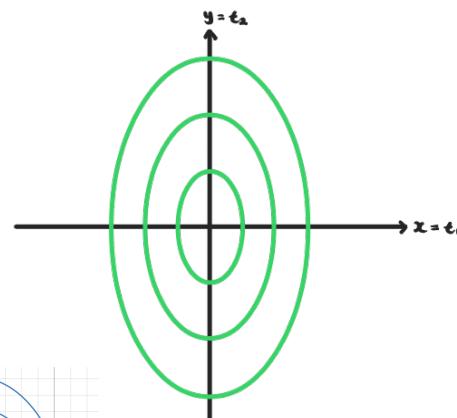
$H_f(-2, 2)$  has eigenvectors  $\vec{w}_1' = \vec{e}_1$  and  $\vec{w}_2' = \vec{e}_2$  with respective eigenvalues  $\lambda_1 = -6$  and  $\lambda_2 = 2$ . For  $\vec{v} = t_1 \vec{w}_1' + t_2 \vec{w}_2'$ ,

$$g_{H_f(-2, 2)}(\vec{v}) = -6t_1^2 + 2t_2^2.$$



$H_f(0, 2)$  has eigenvectors  $\vec{w}_1' = \vec{e}_1$  and  $\vec{w}_2' = \vec{e}_2$  with respective eigenvalues  $\lambda_1 = 6$  and  $\lambda_2 = 2$ . For  $\vec{v} = t_1 \vec{w}_1' + t_2 \vec{w}_2'$ ,

$$g_{H_f(0, 2)}(\vec{v}) = 6t_1^2 + 2t_2^2.$$



**Example 8.** Find the critical points of  $g(x, y) = 3x^2y + 2y^3 - xy$  and classify each as a local maximum, local minimum, or a saddle point.

We compute  $\nabla g = \begin{bmatrix} 6xy - y \\ 3x^2 + 6y^2 - x \end{bmatrix}$  and  $Hg = \begin{bmatrix} 6y & 6x - 1 \\ 6x - 1 & 12y \end{bmatrix}$ . Setting  $\nabla g = \vec{0}$  gives us critical points  $P(0, 0)$ ,  $Q(\frac{1}{3}, 0)$ ,  $R(\frac{1}{6}, \frac{1}{6\sqrt{2}})$ ,  $S(\frac{1}{6}, -\frac{1}{6\sqrt{2}})$ .

$Hg(P) = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$  has eigenvalues 1 and -1. P is a saddle point.

$Hg(Q) = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$  has eigenvalues 1 and -1. Q is a saddle point.

$Hg(R) = \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$  has eigenvalues  $\frac{1}{2}$  and  $\frac{1}{2}$ . R is a local minimum.

$Hg(S) = \begin{bmatrix} -\frac{1}{2} & 0 \\ 0 & -\frac{1}{2} \end{bmatrix}$  has eigenvalues  $-\frac{1}{2}$  and  $-\frac{1}{2}$ . S is a local maximum.

