

Multi-Variable Calculus Notes

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The Gradient:

- Suppose $f: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$ is a function that takes as input a matrix A of size $m \times n$ and returns a real value. Then, the **gradient** of f with respect to $A \in \mathbb{R}^{m \times n}$ is the matrix of partial derivatives.

$$\text{I.e. } \nabla f(A) = \begin{bmatrix} \frac{\partial f(A)}{\partial A_{11}} & \frac{\partial f(A)}{\partial A_{12}} & \dots & \frac{\partial f(A)}{\partial A_{1n}} \\ \vdots & & & \vdots \\ \frac{\partial f(A)}{\partial A_{m1}} & \dots & \dots & \frac{\partial f(A)}{\partial A_{mn}} \end{bmatrix}$$

- Find the gradient in each of the next examples.

E.g. 1 $f(x, y, z) = xy + z$

Soln:

$$\frac{\partial f}{\partial x} = y \quad \frac{\partial f}{\partial y} = x \quad \frac{\partial f}{\partial z} = 1$$

$$\therefore \nabla f(x, y, z) = [y, x, 1]$$

E.g. 2 $f(x, y, z) = xy + yz + xz$

Soln:

$$\frac{\partial f}{\partial x} = y + z, \quad \frac{\partial f}{\partial y} = x + z, \quad \frac{\partial f}{\partial z} = y + x$$

$$\therefore \nabla f(x, y, z) = [y + z, x + z, x + y]$$

- Some properties of gradients:

1. $\nabla(f(x) + g(x)) = \nabla f(x) + \nabla g(x)$
2. For $t \in \mathbb{R}$, $\nabla(tf(x)) = t(\nabla f(x))$

The Hessian:

- Suppose that $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a function that takes a vector in \mathbb{R}^n and returns a real number. The **Hessian matrix** with respect to x , denoted as $\nabla_x^2 f(x)$ or H , is the $n \times n$ matrix of partial derivatives.

I.e.

$$H = \begin{bmatrix} \frac{\partial^2 f(x)}{\partial x_1^2} & \frac{\partial^2 f(x)}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f(x)}{\partial x_1 \partial x_n} \\ \vdots & & & \vdots \\ \frac{\partial^2 f(x)}{\partial x_n \partial x_1} & \dots & \dots & \frac{\partial^2 f(x)}{\partial x_n^2} \end{bmatrix}$$

- Find the Hessian matrix for each of the examples below:

E.g. 1 $f(x, y) = x^3 - 2xy - y^6$

Soln:

$$f_x = 3x^2 - 2y$$

$$f_y = -2x - 6y^5$$

$$f_{xx} = 6x$$

$$f_{yy} = -30y^4$$

$$f_{xy} = f_{yx} = -2$$

$$\therefore H = \begin{bmatrix} 6x & -2 \\ -2 & -30y^4 \end{bmatrix}$$

E.g. 2 $f(x,y) = y^4 + x^3 + 3x^2 + 4y^2 - 4xy - 5y + 8$

Soln:

$$f_x = 3x^2 + 6x - 4y$$

$$f_y = 4y^3 + 8y - 4x - 5$$

$$f_{xx} = 6x + 6$$

$$f_{yy} = 12y^2 + 8$$

$$f_{xy} = f_{yx} = -4$$

$$\therefore H = \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix} = \begin{bmatrix} 6x+6 & -4 \\ -4 & 12y^2+8 \end{bmatrix}$$

E.g. 3 $f(x,y) = x^2y + y^2x$

Soln:

$$f_x = 2xy + y^2$$

$$f_y = x^2 + 2yx$$

$$f_{xx} = 2y$$

$$f_{yy} = 2x$$

$$f_{xy} = f_{yx} = 2x + 2y$$

$$\left. \begin{array}{l} f_x = 2xy + y^2 \\ f_y = x^2 + 2yx \\ f_{xx} = 2y \\ f_{yy} = 2x \\ f_{xy} = f_{yx} = 2x + 2y \end{array} \right\} H = \begin{bmatrix} 2y & 2x+2y \\ 2x+2y & 2x \end{bmatrix}$$

- **Note:** The Hessian matrix is always symmetrical.

Gradients and Hessians of Quadratic and Linear Functions:

- $\nabla b^T x = b$
- $\nabla x^T A x = 2Ax$ if A is symmetric
- $\nabla^2 x^T A x = 2A$ if A is symmetric

Least Squares

- Let $A \in \mathbb{R}^{m \times n}$ have a full rank.

Let $b \in \mathbb{R}^m$ s.t. $b \in R(A)$

In this situation, we want to find a vector $x \in \mathbb{R}^n$ s.t. Ax is as close to b as possible, as measured by the square of the Euclidean norm $(\|Ax - b\|_2)^2$.

Using the fact $(\|x\|_2)^2 = x^T x$, we have:

$$\begin{aligned} (\|Ax - b\|_2)^2 &= (Ax - b)^T (Ax - b) \\ &= x^T A^T A x - 2b^T A x + b^T b \end{aligned}$$

Taking the gradient of both sides we get:

Taking the gradient w.r.t x , we have:

$$\begin{aligned} &\nabla_x (x^T A^T A x - 2b^T A x + b^T b) \\ &= \nabla_x (x^T A^T A x) - \nabla_x (2b^T A x) + \nabla_x (b^T b) \\ &= 2A^T A x - 2A^T b \end{aligned}$$

Setting the last expression to 0, and solving for x , we get:

$$x = (A^T A)^{-1} A^T b$$

Gradients of the Determinant:

- Let $A \in \mathbb{R}^{n \times n}$. We want to find $\nabla_A |A|$.

$$\frac{\partial}{\partial A_{k\ell}} |A| = \frac{\partial}{\partial A_{k\ell}} \sum_{i=1}^n (-1)^{i+j} A_{ij} |A_{\setminus i, \setminus j}|$$

$$= (-1)^{k+\ell} |A_{\setminus k, \setminus \ell}|$$

$$= (\text{adj}(A))_{\ell k}$$

$$= (\text{adj}(A))^T$$

$$= |A| A^{-T}$$

Eigenvalues as Optimization:

- If we want to optimize (min or max) $f(x, y, z)$ subject to the constraint $g(x, y, z) = k$, we can do:

1. $\nabla f(x, y, z) = \lambda \nabla g(x, y, z)$

2. $g(x, y, z) = k$

3. Plug all solns into $f(x, y, z)$ and find the max and min.

- $L(x, \lambda) = f(x) - \lambda g(x)$ is called Lagrange function.

- λ is called Lagrange Multiplier.

- Note: $\nabla g \neq 0$ at the point

- E.g.

1. Find the max and min of $f(x,y) = 5x - 3y$ subject to the constraint $x^2 + y^2 = 136$.

Soln:

$$\nabla f(x,y) = \begin{bmatrix} 5 \\ -3 \end{bmatrix} \quad \lambda \nabla g(x,y) = \begin{bmatrix} 2\lambda x \\ 2\lambda y \end{bmatrix}$$

$$5 = 2\lambda x$$

$$-3 = 2\lambda y$$

$$x^2 + y^2 = 136$$

$$x = \frac{5}{2\lambda}, \quad y = \frac{-3}{2\lambda}$$

$$\left(\frac{5}{2\lambda}\right)^2 + \left(\frac{-3}{2\lambda}\right)^2 = 136$$

$$\frac{25}{4\lambda^2} + \frac{9}{4\lambda^2} = 136$$

$$\frac{34}{4\lambda^2} = 136$$

$$\frac{17}{2\lambda^2} = 136$$

$$17 = 272\lambda^2$$

$$\frac{17}{272} = \lambda^2$$

$$\frac{1}{16} = \lambda^2$$

$$\lambda = \pm \frac{1}{4}$$

When $\lambda = \frac{1}{4}$:

$$\rightarrow x = 10$$

$$\rightarrow y = -6$$

When $\lambda = -\frac{1}{4}$:

$$\rightarrow x = -10$$

$$\rightarrow y = 6$$

$$f(10, -6) = 68 \quad \text{Max}$$

$$f(-10, 6) = -68 \quad \text{Min}$$

2. Find the max and min of $f(x, y, z) = x + z$ subject to the constraint $x^2 + y^2 + z^2 = 1$.

Soln:

$$1 = 2\lambda x$$

$$0 = 2\lambda y$$

$$1 = 2\lambda z$$

$$x^2 + y^2 + z^2 = 1$$

$$x = \frac{1}{2\lambda}, \quad y = 0, \quad z = \frac{1}{2\lambda}$$

$$\left(\frac{1}{2\lambda}\right)^2 + (0)^2 + \left(\frac{1}{2\lambda}\right)^2 = 1$$

$$\frac{2}{4\lambda^2} = 1$$

$$\frac{1}{2\lambda^2} = 1$$

$$\lambda^2 = \frac{1}{2}$$

$$\lambda = \pm \frac{1}{\sqrt{2}}$$

$$x = \pm \frac{\sqrt{2}}{2} = \pm \frac{1}{\sqrt{2}} = z, \quad y = 0$$

$(-\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}})$ and $(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}})$ are the 2 points.

$$f\left(\frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}}\right) = \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} = \frac{2}{\sqrt{2}} \quad \text{Max}$$

$$f\left(-\frac{1}{\sqrt{2}}, 0, -\frac{1}{\sqrt{2}}\right) = -\frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}} = -\frac{2}{\sqrt{2}} \quad \text{Min}$$

3. Find the max and min of $f(x, y, z) = x - y + z$ subject to the constraint $x^2 + y^2 + z^2 = 2$

$$\begin{aligned} 1 &= 2\lambda x \\ -1 &= 2\lambda y \\ 1 &= 2\lambda z \\ x^2 + y^2 + z^2 &= 2 \end{aligned}$$

$$x = \frac{1}{2\lambda}, \quad y = \frac{-1}{2\lambda}, \quad z = \frac{1}{2\lambda}$$

$$3\left(\frac{1}{2\lambda}\right)^2 = 2$$

$$\frac{1}{4\lambda^2} = \frac{2}{3}$$

$$\begin{aligned} \lambda^2 &= \frac{3}{8} \\ \lambda &= \pm \sqrt{\frac{3}{8}} \end{aligned}$$

$$\text{When } \lambda = \sqrt{\frac{3}{8}}$$

$$x = \frac{1}{2\lambda}$$

$$= \frac{1}{2\sqrt{3/8}}$$

$$= \frac{1}{\sqrt{12/8}}$$

$$= \frac{1}{\sqrt{3/2}}$$

$$= \sqrt{2/3}$$

$$y = -\sqrt{2/3}$$

$$z = \sqrt{2/3}$$

$$\text{When } \lambda = -\sqrt{3/8}$$

$$x = -\sqrt{2/3}$$

$$y = \sqrt{2/3}$$

$$z = -\sqrt{2/3}$$

$$f(\sqrt{2/3}, -\sqrt{2/3}, \sqrt{2/3}) = \sqrt{2/3} \quad \text{Max}$$

$$f(-\sqrt{2/3}, \sqrt{2/3}, -\sqrt{2/3}) = -\sqrt{2/3} \quad \text{Min}$$