

# COMP 395 -2 Deep Learning: Partial Derivatives and Gradients Practice

Name: \_\_\_\_\_

Date: \_\_\_\_\_

## Part A: Single Partial Derivatives

For each function, compute only the partial derivative indicated.

1. Let  $f(x, y) = 4x^3 - 7y^2 + 5xy$

Find  $\frac{\partial f}{\partial x}$

2. Let  $g(x, y) = x^2 \ln(y) + 3y^4$

Find  $\frac{\partial g}{\partial y}$

3. Let  $h(x, y) = \sin(x) + y \cos(x) + y^3$

Find  $\frac{\partial h}{\partial x}$

## Part B: Both Partial Derivatives and Geometric Interpretation

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

This function describes a surface in 3D space (an elliptic paraboloid).

(a) Find  $\frac{\partial f}{\partial x}$

(b) Find  $\frac{\partial f}{\partial y}$

- (c) Evaluate both partial derivatives at the point  $(1, 2)$ .

- (d) **Geometric interpretation:** In your own words, explain what  $\left. \frac{\partial f}{\partial x} \right|_{(1,2)}$  tells you about the surface at the point  $(1, 2, f(1, 2))$ .

*Hint: Imagine slicing the surface with a plane where  $y = 2$  (constant). What does the partial derivative describe about that slice?*

## Part C: The Gradient Vector

5. Let  $f(x_1, x_2, x_3, x_4) = (x_1 - 2)^2 + (x_2 + 1)^3 + e^{2x_3} + x_4^2$

(a) Find all four partial derivatives:

$$\frac{\partial f}{\partial x_1} = \quad \quad \quad (\text{chain rule})$$

$$\frac{\partial f}{\partial x_2} = \quad \quad \quad (\text{chain rule})$$

$$\frac{\partial f}{\partial x_3} = \quad \quad \quad (\text{chain rule})$$

$$\frac{\partial f}{\partial x_4} =$$

(b) Write the gradient vector  $\nabla f$  as a column vector (in terms of  $x_1, x_2, x_3, x_4$ ):

*Hint: The gradient collects all partial derivatives into one vector.*

$$\nabla f = \begin{pmatrix} \phantom{0} \\ \phantom{0} \\ \phantom{0} \\ \phantom{0} \end{pmatrix}$$

(c) Evaluate the gradient at the point  $(3, 0, 0, 2)$ :

$$\nabla f|_{(3,0,0,2)} = \begin{pmatrix} \phantom{0} \\ \phantom{0} \\ \phantom{0} \\ \phantom{0} \end{pmatrix}$$

(d) **Interpretation:** Answer the following about the gradient you computed:

- i. The gradient vector points in the direction of \_\_\_\_\_.

*Hint: If you were standing on the surface and wanted  $f$  to increase as fast as possible, which way would you walk?*

- ii. If you wanted  $f$  to *decrease* as fast as possible, you would move in the direction of \_\_\_\_\_.

- iii. The *magnitude* of the gradient,  $\|\nabla f\|$ , tells you \_\_\_\_\_.

*Hint: A larger magnitude means the surface is changing more rapidly. What does that correspond to physically?*