

WEEK 3 – CLASS EXERCISES

EXERCISE 1: Suppose that at the point $(1, 2)$ the directional derivative of the function $z = f(x, y)$ in the direction $3\vec{i} + 4\vec{j}$ is equal to 2 and the directional derivative of $z = f(x, y)$ in the direction $5\vec{i} + 12\vec{j}$ is equal to 1 .

(a) Determine the gradient of f at the point $(1, 2)$.

The directional derivative of function f at some point (x_0, y_0) in the direction of a vector \vec{v} is defined as:

$$\partial_{\vec{v}} f = \langle \nabla f, \vec{v} \rangle = v_1 f_x(x_0, y_0) + v_2 f_y(x_0, y_0)$$

Therefore, we can write the following system of equations:

$$\begin{cases} 3f_x(1, 2) + 4f_y(1, 2) = 2 & (\alpha) \\ 5f_x(1, 2) + 12f_y(1, 2) = 1 & (\beta) \end{cases}$$

Now, by $(\beta) - 3(\alpha)$:

$$5f_x(1, 2) + 12f_y(1, 2) - 9f_x(1, 2) - 12f_y(1, 2) = 1 - 6 \Rightarrow -4f_x(1, 2) = -5 \Rightarrow f_x(1, 2) = \frac{5}{4}$$

and by $5(\alpha) - 3(\beta)$:

$$15f_x(1, 2) + 20f_y(1, 2) - 15f_x(1, 2) - 36f_y(1, 2) = 10 - 3 \Rightarrow -16f_y(1, 2) = 7 \Rightarrow f_y(1, 2) = -\frac{7}{16}$$

Therefore, the gradient of the function f at the point $(1, 2)$ will be:

$$\nabla f(1, 2) = \left(\frac{5}{4}, -\frac{7}{16} \right)$$

Another way of interpreting this exercise would be to consider the normalised vectors. In this case, we would have the following system of equations:

$$\begin{cases} \frac{3}{5} f_x(1, 2) + \frac{4}{5} f_y(1, 2) = 2 \\ \frac{5}{13} f_x(1, 2) + \frac{12}{13} f_y(1, 2) = 1 \end{cases} \Rightarrow \begin{cases} 3f_x(1, 2) + 4f_y(1, 2) = 10 & (\alpha) \\ 5f_x(1, 2) + 12f_y(1, 2) = 13 & (\beta) \end{cases}$$

Now, by $(\beta) - 3(\alpha)$:

$$5f_x(1, 2) + 12f_y(1, 2) - 9f_x(1, 2) - 12f_y(1, 2) = 13 - 30 \Rightarrow -4f_x(1, 2) = -17 \Rightarrow f_x(1, 2) = \frac{17}{4}$$

and by $5(\alpha) - 3(\beta)$:

$$15f_x(1, 2) + 20f_y(1, 2) - 15f_x(1, 2) - 36f_y(1, 2) = 50 - 39 \Rightarrow -16f_y(1, 2) = 11 \Rightarrow f_y(1, 2) = -\frac{11}{16}$$

Therefore, the gradient of the function f at the point $(1, 2)$ will be:

$$\nabla f(1, 2) = \left(\frac{17}{4}, -\frac{11}{16} \right)$$

Note that both interpretations of the exercise are totally valid!

(b) What is the maximum rate of change of the function f at the point $(1, 2)$? And in what direction does this maximum occur?

Note that as in the previous exercise, we can have the two different interpretations. We will now solve it considering the non-normalised vectors, but the other approach is as valid!

The maximum rate of change of the function f at the point $(1, 2)$ corresponds to the greatest directional derivative at such point. The directional derivative of a function in the direction of some vector \vec{v} can be calculated as:

$$\partial_{\vec{v}} f = \langle \nabla f, \vec{v} \rangle = \|\nabla f\| \|\vec{v}\| \cos \alpha$$

Since \vec{v} is a unit vector, that is $\|\vec{v}\| = 1$, and the maximum value for $\cos \alpha$ is 1, the maximum rate of change will correspond to the norm of the gradient:

$$\|\nabla f\| = \sqrt{\left(\frac{5}{4}\right)^2 + \left(-\frac{7}{16}\right)^2} = \sqrt{\frac{5^2 * 16 + 7^2}{256}} = \sqrt{\frac{449}{256}} = \frac{\sqrt{449}}{16}$$

Since the greatest directional derivative is:

$$\partial_{\vec{v}} f = \nabla f \cdot \vec{v} = \|\nabla f\|$$

we can easily determine the direction \vec{v} as:

$$\vec{v} = \frac{\nabla f(1, 2)}{\|\nabla f(1, 2)\|} = \frac{16}{\sqrt{449}} \left(\frac{5}{4}, -\frac{7}{16} \right) = \frac{1}{\sqrt{449}} (20, -7)$$

EXERCISE 2: Consider the function $f(x, y)$ defined by:

$$f(x, y) = \sin(xe^y) - x + 3$$

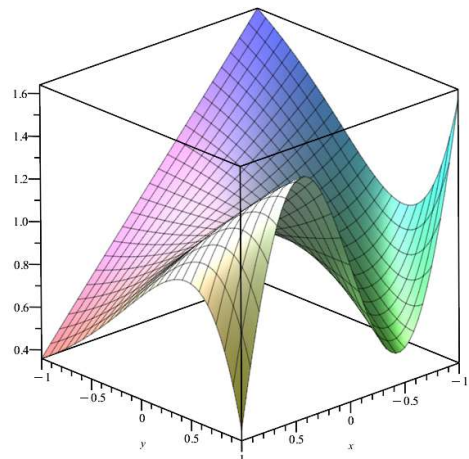
(a) Compute all the first and second order partial derivatives.

$$\frac{\partial}{\partial x} f(x, y) = \cos(xe^y) * e^y - 1$$

$$\frac{\partial}{\partial y} f(x, y) = \cos(xe^y) * xe^y$$

$$\frac{\partial^2}{\partial x^2} f(x, y) = -\sin(xe^y) * e^{2y}$$

$$\frac{\partial^2}{\partial y^2} f(x, y) = -\sin(xe^y) * x^2 e^{2y} + \cos(xe^y) * xe^y$$



$$\frac{\partial^2}{\partial y \partial x} f(x, y) = -\sin(xe^y) * xe^{2y} + \cos(xe^y) * e^y$$

As the function is twice differentiable and its derivatives are continuous, its second order mixed partial derivatives are equal.

$$\frac{\partial^2}{\partial x \partial y} f(x, y) = -\sin(xe^y) * xe^{2y} + \cos(xe^y) * e^y$$

(b) A given fact is that $f(0.2, 0.1) = 3.01924$ accurate to 5 decimal places. Taking $(0, 0)$ as the reference point, use linear approximation to find an approximation of $f(0.2, 0.1)$.

The linear approximation of the function at the point $(0, 0)$ is given by:

$$T_1(x, y) = f(0, 0) + \nabla f(0, 0) \cdot (x, y) = f(x, y) + x f_x(0, 0) + y f_y(0, 0)$$

The gradient of f at the point $(0, 0)$ can be calculated by substituting in the derivatives we calculated in **(a)**.

$$\nabla f(0, 0) = (f_x(0, 0), f_y(0, 0)) = (0, 0)$$

and therefore:

$$T_1(0.2, 0.1) = f(0, 0) + (0, 0) \cdot (0.2, 0.1) = f(0, 0) = 3$$

(c) By referring to the idea of a tangent plane, explain why you obtained the answer you did in part **(b)**.

Since the gradient of the function at $(0, 0)$ is the $(0, 0)$ vector, the tangent plane will be horizontal, so the linear approximation corresponds to the value of the function at that point.

(d) Use a 2nd order Taylor polynomial to find an approximation of $f(0.2, 0.1)$. Is this approximation better or worse than the linear approximation?

The second order Taylor polynomial centred at $(0, 0)$ can be calculated as:

$$T_2(x, y) = T_1(x, y) + \frac{1}{2} [x \ y] H(0,0) \begin{bmatrix} x \\ y \end{bmatrix} = T_1(x, y) + \frac{1}{2} [x \ y] \begin{bmatrix} f_{xx}(0,0) & f_{xy}(0,0) \\ f_{yx}(0,0) & f_{yy}(0,0) \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Let's first calculate the second order partial derivatives at the point $(0, 0)$.

$$f_{xx}(0, 0) = f_{yy}(0, 0) = 0 \quad \text{and} \quad f_{xy}(0, 0) = 1$$

Therefore, the new approximation for the function at the point $(0, 0)$ gives us:

$$T_2(x, y) = T_1(x, y) + \frac{1}{2} [x \ y] \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = T_1(x, y) + \frac{1}{2} [x \ y] \begin{bmatrix} y \\ x \end{bmatrix} = T_1(x, y) + \frac{2xy}{2} = T_1(x, y) + xy$$

Substituting with the point $(0.2, 0.1)$:

$$T_2(0.2, 0.1) = T_1(0.2, 0.1) + 0.2 * 0.1 = 3 + 0.02 = 3.02$$

We can clearly see by comparing both approximations that the second order Taylor polynomial gives a better approximation than the linear one.

(e) Find a critical point of $f(x, y)$ and determine if it is a local minimum, local maximum or saddle point. You do not need to find all of them.

The critical points of the function $f(x, y)$ are those where the gradient is 0. In part (c), we saw that:

$$\nabla f(0, 0) = (0, 0)$$

which means that the point $(0, 0)$ is a critical point of the function f . We can determine if $(0, 0)$ is a maximum, minimum or saddle point by calculating the determinant of the Hessian matrix $H(0, 0)$:

$$\det(H(0, 0)) = f_{xx}(0, 0) f_{yy}(0, 0) - f_{xy}^2(0, 0) = 0 * 0 - 1^2 = -1$$

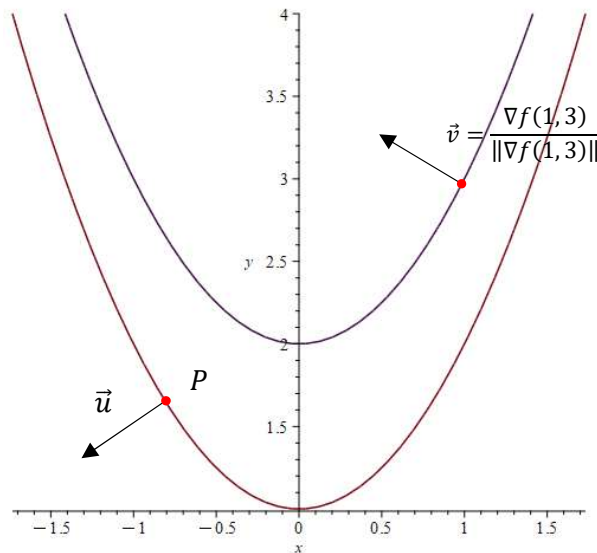
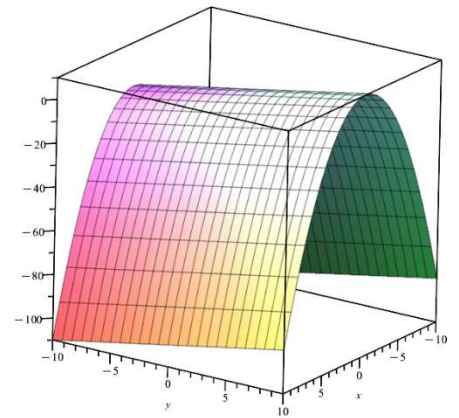
Because $\det(H(0, 0)) < 0$, the point $(0, 0)$ is a saddle point.

EXERCISE 3: Let $f(x, y) = y - x^2$.

(a) Make a sketch of the level curves $f(x, y) = 1$ and $f(x, y) = 2$ on the same axes, that is, make a contour plot with these two level curves.

We can draw the contour plot with Maple:

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with(plots)
contourplot(y - x^2, contours = [1, 2])
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Note that the red curve corresponds to the level curve $f = 1$ and the purple one to $f = 2$.

(b) On the above plot, draw a point P and a vector \vec{u} (with initial point at P) such that the directional derivative $D_{\vec{u}}(P)$ is negative.

Since the gradient vector points to the direction of maximum increase and is orthogonal to the level curve, we can consider any vector \vec{u} pointing away from the level curve given by $f(x, y) = 2$.

(c) At the point $(1, 3)$, find the direction in which $f(x, y)$ is increasing the fastest. Sketch the point and the vector on the above plot.

The directional derivative of a function in the direction of some vector \vec{v} can be calculated as:

$$\partial_{\vec{v}} f = \langle \nabla f, \vec{v} \rangle = \|\nabla f\| \|\vec{v}\| \cos \alpha$$

Since \vec{v} is a unit vector, that is $\|\vec{v}\| = 1$, and the maximum value for $\cos \alpha$ is 1, the maximum directional derivative is given by:

$$\partial_{\vec{v}} f = \nabla f \cdot \vec{v} = \|\nabla f\|$$

Therefore, the direction \vec{v} in which f is increasing the fastest will be:

$$\vec{v} = \frac{\nabla f(1, 3)}{\|\nabla f(1, 3)\|} = \frac{1}{\sqrt{(-2)^2 + 1^2}} (-2, 1) = \frac{1}{\sqrt{5}} (-2, 1) = \left(-\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)$$