Definition 22 [Directional derivative]: The directional derivative of $f(\mathbf{x})$ at \mathbf{x} in the direction of the unit vector \mathbf{v} is

$$D_{\mathbf{v}}f(\mathbf{x}) = \lim_{h \to 0} \frac{f(\mathbf{x} + h\mathbf{v}) - f(\mathbf{x})}{h}$$
(4.12)

provided the limit exists.

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http://www.cse.iitb.ac.in/~cs709/notes/BasicsOfConvexOptimiz ation.pdf

As a special case, when $\mathbf{v} = \mathbf{u}^k$ the directional derivative reduces to the partial derivative of f with respect to x_k .

$$D_{\mathbf{u}^k}f(\mathbf{x}) = \frac{\partial f(\mathbf{x})}{\partial x_k} + \frac{\partial f(\mathbf{x})}{\partial x_k} \frac{$$

Theorem 57 If $f(\mathbf{x})$ is a differentiable function of $\mathbf{x} \in \mathbb{R}^n$, then f has a directional derivative in the direction of any unit vector \mathbf{v} , and

differentiable function of
$$\mathbf{x} \in \mathbb{R}^n$$
, then f has a differentiable function of \mathbf{a} \mathbf{x} \mathbf{x}

Definition 23 [Gradient Vector]: If f is differentiable function of $\mathbf{x} \in \mathbb{R}^n$, then the gradient of $f(\mathbf{x})$ is the vector function $\nabla f(\mathbf{x})$, defined as:

$$\nabla f(\mathbf{x}) = [f_{x_1}(\mathbf{x}), f_{x_2}(\mathbf{x}), \dots, f_{x_n}(\mathbf{x})]$$

The directional derivative of a function f at a point x in the direction of a unit vector \mathbf{v} can be now written as

Theorem 58 Suppose f is a differentiable function of $\mathbf{x} \in \mathbb{R}^n$. The maximum value of the directional derivative $D_{\mathbf{v}}f(\mathbf{x})$ is $||\nabla f(\mathbf{x})||$ and it is so when \mathbf{v} has the same direction as the gradient vector $\nabla f(\mathbf{x})$.

What does the gradient $\nabla f(\mathbf{x})$ tell you about the function $f(\mathbf{x})$? We will illustrate with some examples. Consider the polynomial $f(x,y,z) = x^2y + z \sin xy$ and the unit vector $\mathbf{v}^T = \frac{1}{\sqrt{3}}[1,1,1]^T$. Consider the point $p_0 = (0,1,3)$. We will compute the directional derivative of f at p_0 in the direction of \mathbf{v} . To do this, we first compute the gradient of f in general: $\nabla f = \begin{bmatrix} 2xy + yz \cos xy, & x^2 + xz \cos xy, & \sin xy \end{bmatrix}$ Evaluating the gradient at a specific point p_0 , $\nabla f(0,1,3) = \begin{bmatrix} 3, 0, 0 \end{bmatrix}^T$. The directional derivative at p_0 in the direction \mathbf{v} is $D_{\mathbf{v}}f(0,1,3) = \begin{bmatrix} 3, 0, 0 \end{bmatrix} \cdot \frac{1}{\sqrt{3}}[1,1,1]^T = \sqrt{3}$. This directional derivative is the rate of change of f at p_0 in the direction \mathbf{v} ; it is positive indicating that the function f increases at f0 in the direction f1. All our ideas about first and second derivative in the case of a single variable carry over to the directional derivative.

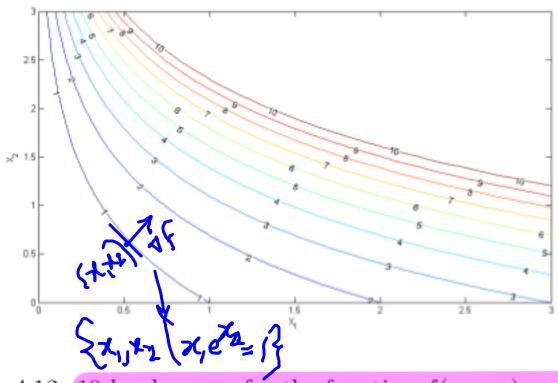


Figure 4.12: 10 level curves for the function $f(x_1, x_2) = x_1 e^{x_2}$.

Consider the function $f(x_1, x_2) = x_1e^{x_2}$. Figure 4.12 shows 10 level curves for this function, corresponding to $f(x_1, x_2) = c$ for c = 1, 2, ..., 10. The idea behind a level curve is that as you change \mathbf{x} along any level curve, the function value remains unchanged, but as you move \mathbf{x} across level curves, the function value changes.

Theorem 59 Let $f: \mathcal{D} \to \Re$ with $\mathcal{D} \in \Re^n$ be a differentiable function. The gradient ∇f evaluated at \mathbf{x}^* is orthogonal to the tangent hyperplane (tangent

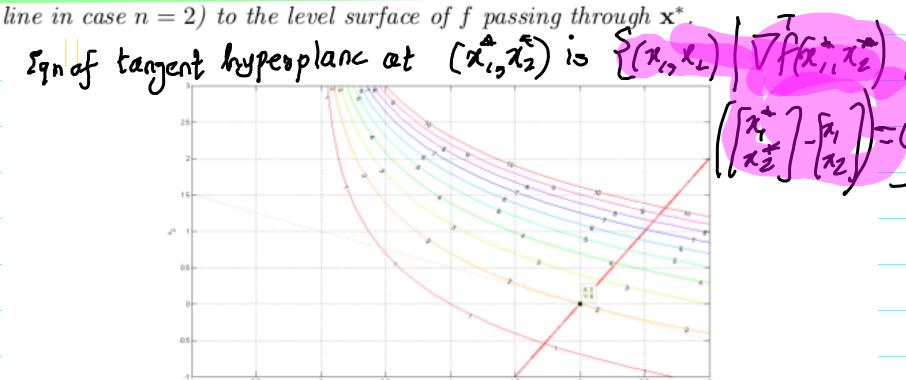


Figure 4.13: The level curves from Figure 4.12 along with the gradient vector at (2,0). Note that the gradient vector is perpenducular to the level curve $x_1e^{x_2} = 2$ at (2,0).

Consider the same plot as in Figure 4.12 with a gradient vector at (2,0) as shown in Figure 4.13. The gradient vector $[1, 2]^T$ is perpendicular to the tangent hyperplane to the level curve $x_1e^{x_2} = 2$ at (2,0). The equation of the tangent hyperplane is $(x_1 - 2) + 2(x_2 - 0) = 0$ and it turns out to be a tangent line.

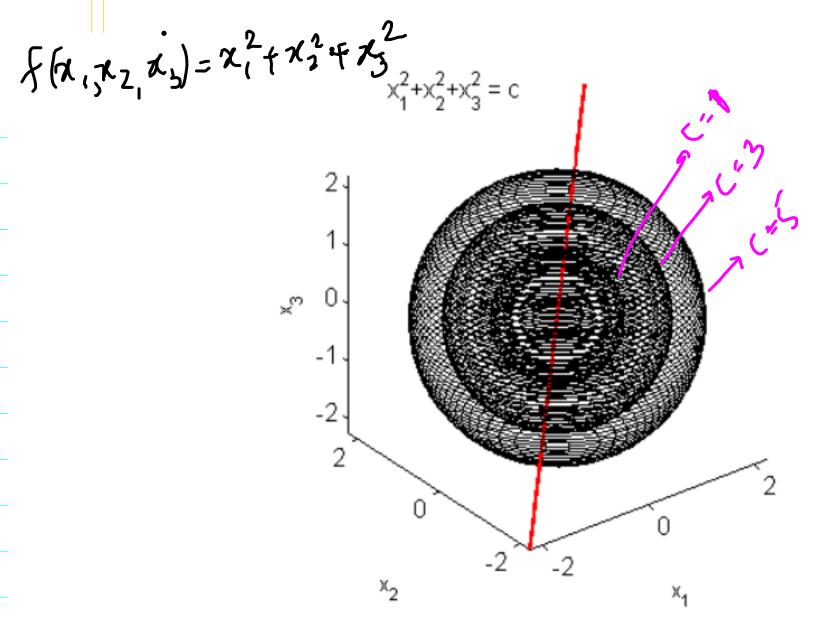


Figure 4.14: 3 level surfaces for the function $f(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2$ with c = 1, 3, 5. The gradient at (1, 1, 1) is orthogonal to the level surface $f(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2 = 3$ at (1, 1, 1).

The level surfaces for $f(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2$ are shown in Figure 4.14. The gradient at (1, 1, 1) is orthogonal to the tangent hyperplane to the level surface $f(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2 = 3$ at (1, 1, 1). The gradient vector at (1, 1, 1) is $[2, 2, 2]^T$ and the tanget hyperplane has the equation $2(x_1 - 1) + 2(x_2 - 1) + 2(x_3 - 1) = 0$, which is a plane in 3D. On the other hand, the dotted line in Figure 4.15 is not orthogonal to the level surface, since it does not coincide with the gradient.

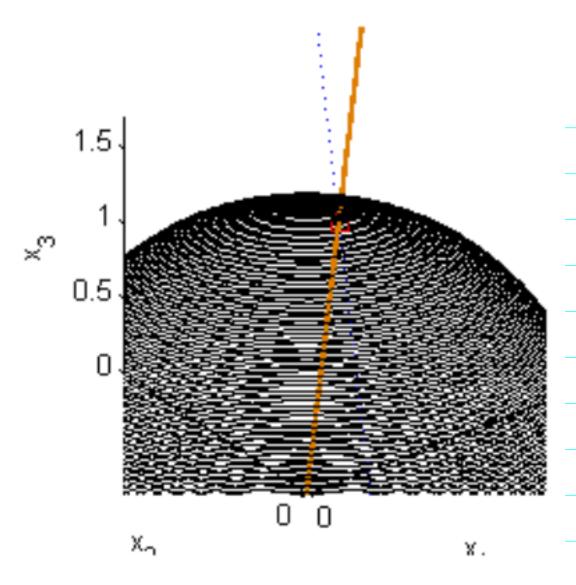


Figure 4.15: Level surface $f(x_1, x_2, x_3) = x_1^2 + x_2^2 + x_3^2 = 3$. The gradient at (1, 1, 1), drawn as a bold line, is perpendicular to the tangent plane to the level surface at (1, 1, 1), whereas, the dotted line, though passing through (1, 1, 1) is not perpendicular to the same tangent plane.

- 3. Let $f(x_1, x, x_3) = x_1^2 x_2^3 x_3^4$ and consider the point $\mathbf{x}^0 = (1, 2, 1)$. We will find the equation of the tangent plane to the level surface through \mathbf{x}^0 . The level surface through \mathbf{x}^0 is determined by setting f equal to its value evaluated at \mathbf{x}^0 ; that is, the level surface will have the equation $x_1^2 x_2^3 x_3^4 = 1^2 2^3 1^4 = 8$. The gradient vector (normal to tangent plane) at
 - (1,2,1) is $\nabla f(x_1,x_2,x_3)|_{(1,2,1)} = [2x_1x_2^3x_3^4, 3x_1^2x_2^2x_3^4, 4x_1^2x_2^3x_3^3]^T|_{(1,2,1)} = [16, 12, 32]^T$. The equation of the tangent plane at \mathbf{x}^0 , given the normal vector $\nabla f(\mathbf{x}^0)$ can be easily written down: $\nabla f(\mathbf{x}^0)^T \cdot [\mathbf{x} \mathbf{x}^0] = 0$ which turns out to be $16(x_1 1) + 12(x_2 2) + 32(x_3 1) = 0$, a plane in 3D.

- 4. Consider the function $f(x,y,z) = \frac{x}{y+z}$. The directional derivative of f in the direction of the vector $\mathbf{v} = \frac{1}{\sqrt{14}}[1,\ 2,\ 3]$ at the point $x^0 = (4,1,1)$ is $\nabla^T f\big|_{(4,1,1)} \cdot \frac{1}{\sqrt{14}}[1,\ 2,\ 3]^T = \left[\frac{1}{y+z},\ -\frac{x}{(y+z)^2},\ -\frac{x}{(y+z)^2}\right]\big|_{(4,1,1)} \cdot \frac{1}{\sqrt{14}}[1,\ 2,\ 3]^T = \left[\frac{1}{2},\ -1,\ -1\right] \cdot \frac{1}{\sqrt{14}}[1,\ 2,\ 3]^T = -\frac{9}{2\sqrt{14}}$. The directional derivative is negative, indicating that the function decreases along the direction of \mathbf{v} . Based on theorem 58, we know that the maximum rate of change of a function at a point \mathbf{x} is given by $||\nabla f(\mathbf{x})||$ and it is in the direction $\frac{\nabla f(\mathbf{x})}{||\nabla f(\mathbf{x})||}$. In the example under consideration, this maximum rate of change at \mathbf{x}^0 is $\frac{3}{2}$ and it is in the direction of the vector $\frac{2}{3}\left[\frac{1}{2},\ -1,\ -1\right]$.
- 5. Let us find the maximum rate of change of the function $f(x,y,z) = x^2y^3z^4$ at the point $\mathbf{x}^0 = (1,1,1)$ and the direction in which it occurs. The gradient at \mathbf{x}^0 is $\nabla^T f\big|_{(1,1,1)} = [2, 3, 4]$. The maximum rate of change at \mathbf{x}^0 is therefore $\sqrt{29}$ and the direction of the corresponding rate of change is $\frac{1}{\sqrt{29}}[2, 3, 4]$. The minimum rate of change is $-\sqrt{29}$ and the corresponding direction is $-\frac{1}{\sqrt{29}}[2, 3, 4]$.

6. Let us determine the equations of (a) the tangent plane to the paraboloid $\mathcal{P}: x_1 = x_2^2 + x_3^2 + 2$ at $(-1,1,0)$ and (b) the normal line to the tangent plane. To realize this as the level surface of a function of three variables, we define the function $f(x_1, x_2, x_3) = x_1 - x_2^2 - x_3^2$ and find that the paraboloid	
$[1,-2,0]^T$ and its parametric equation is $[x_1, x_2, x_3] = [-1+t, 1-2t, 0]$. The equation of the tangent plane is therefore $(x_1+1)-2(x_2-1)=0$.	
	equation of the temperature is energiate (w1 + 1) 2(w2 1) of

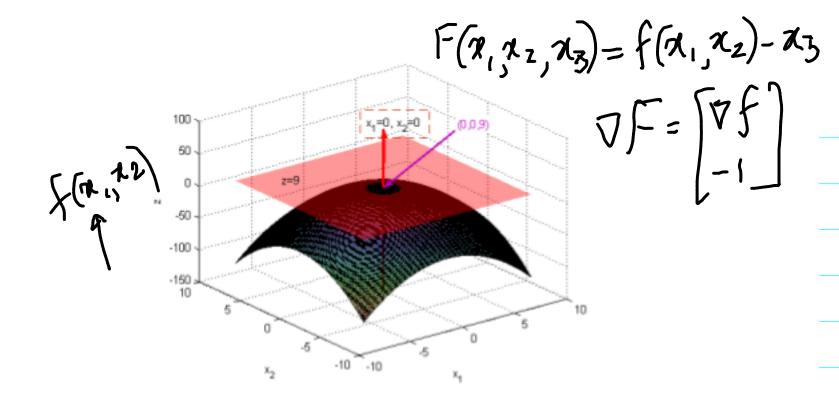


Figure 4.17: The paraboloid $f(x_1, x_2) = 9 - x_1^2 - x_2^2$ attains its maximum at (0,0). The tanget plane to the surface at (0,0,f(0,0)) is also shown, and so is the gradient vector ∇F at (0,0,f(0,0)).

We can embed the graph of a function of n variables as the 0-level surface of a function of n+1 variables. More concretely, if $f: \mathcal{D} \to \Re$, $\mathcal{D} \subseteq \Re^n$ then we define $F: \mathcal{D}' \to \Re$, $\mathcal{D}' = \mathcal{D} \times \Re$ as $F(\mathbf{x}, z) = f(\mathbf{x}) - z$ with $\mathbf{x} \in \mathcal{D}'$. The function f then corresponds to a single level surface of F given by $F(\mathbf{x}, z) = 0$. In other words, the 0-level surface of F gives back the graph of f. The gradient of F at any point (\mathbf{x}, z) is simply, $\nabla F(\mathbf{x}, z) = [f_{x_1}, f_{x_2}, \dots, f_{x_n}, -1]$ with the first n components of $\nabla F(\mathbf{x}, z)$ given by the n components of $\nabla f(\mathbf{x})$. We note that the level surface of F passing through point $(\mathbf{x}^0, f(\mathbf{x}^0))$ is its 0-level surface, which is essentially the surface of the function $f(\mathbf{x})$. The equation of the tangent hyperplane to the 0-level surface of F at the point $(\mathbf{x}^0, f(\mathbf{x}^0))$ (that is, the tangent hyperplane to $f(\mathbf{x})$ at the point \mathbf{x}_0), is $\nabla F(\mathbf{x}^0, f(\mathbf{x}^0))^T \cdot [\mathbf{x} - \mathbf{x}^0, z - f(\mathbf{x}^0)]^T = 0$. Substituting appropriate expression for $\nabla F(\mathbf{x}^0)$, the equation of the tangent plane can be written as

$$\left(\sum_{i=1}^n f_{x_i}(\mathbf{x}^0)(x_i-x_i^0)\right) - \left(z-f(\mathbf{x}^0)\right) = 0$$
 or equivalently as,
$$\left(\sum_{i=1}^n f_{x_i}(\mathbf{x}^0)(x_i-x_i^0)\right) + f(\mathbf{x}^0) = z$$

As an example, consider the paraboloid, $f(x_1, x_2) = 9 - x_1^2 - x_2^2$, the corresponding $F(x_1, x_2, z) = 9 - x_1^2 - x_2^2 - z$ and the point $x^0 = (\mathbf{x}^0, z) = (1, 1, 7)$ which lies on the 0-level surface of F. The gradient $\nabla F(x_1, x_2, z)$ is $[-2x_1, -2x_2, -1]$, which when evaluated at $x^0 = (1, 1, 7)$ is [-2, -2, -1]. The equation of the tangent plane to f at x^0 is therefore given by $-2(x_1 - 1) - 2(x_2 - 1) + 7 = z$.

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Definition 25 [Local maximum]: A function f of n variables has a local

Definition 25 [Local maximum]: A function f of n variables has a local maximum at \mathbf{x}^0 if $\exists \epsilon > 0$ such that $\forall ||\mathbf{x} - \mathbf{x}^0|| < \epsilon$. $f(\mathbf{x}) \leq f(\mathbf{x}^0)$. In other words, $f(\mathbf{x}) \leq f(\mathbf{x}^0)$ whenever \mathbf{x} lies in some circular disk around \mathbf{x}^0 .

Definition 26 [Local minimum]: A function f of n variables has a local minimum at \mathbf{x}^0 if $\exists \epsilon > 0$ such that $\forall ||\mathbf{x} - \mathbf{x}^0|| < \epsilon$. $f(\mathbf{x}) \geq f(\mathbf{x}^0)$. In other words, $f(\mathbf{x}) \geq f(\mathbf{x}^0)$ whenever \mathbf{x} lies in some circular disk around \mathbf{x}^0 .

Definition 29 [Global maximum]: A function f of n variables, with domain $\mathcal{D} \subseteq \Re^n$ has an absolute or global maximum at \mathbf{x}^0 if $\forall \mathbf{x} \in \mathcal{D}$, $f(\mathbf{x}) \leq f(\mathbf{x}^0)$.

Definition 30 [Global minimum]: A function f of n variables, with domain $\mathcal{D} \subseteq \mathbb{R}^n$ has an absolute or global minimum at \mathbf{x}^0 if $\forall \mathbf{x} \in \mathcal{D}$, $f(\mathbf{x}) \geq f(\mathbf{x}^0)$.

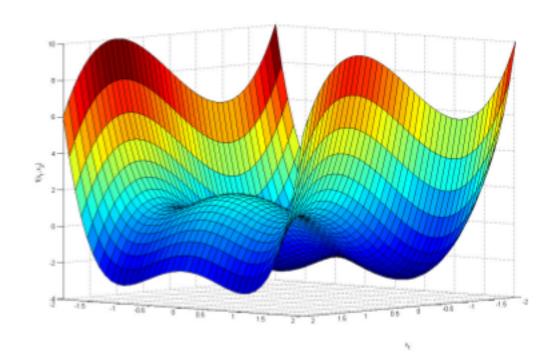


Figure 4.16: Plot of $f(x_1, x_2) = 3x_1^2 - x_1^3 - 2x_2^2 + x_2^4$, showing the various local maxima and minima of the function.

