

Gradiente

Conjuntos de nivel

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Matemáticas para
la ciencia de datos

Definición (Derivada de un campo escalar)

Sea $f : \mathbb{R}^n \rightarrow \mathbb{R}$ un campo escalar. Definimos la derivada de f a partir del punto $a \in \mathbb{R}^n$ en la dirección $y \in \mathbb{R}^n$, como

$$f'(a; y) = \lim_{h \rightarrow 0} \frac{f(a + hy) - f(a)}{h},$$

Diagrama de un punto a en \mathbb{R}^n con un vector dirección y .

$$a = (a_1, \dots, a_n) \quad (1)$$

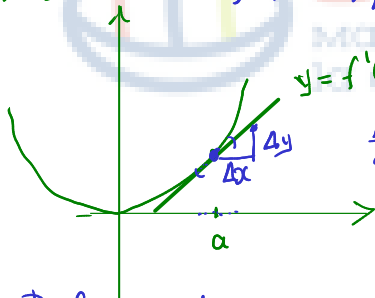
$$y = (y_1, \dots, y_n)$$

siempre que dicho límite exista.

Nota: Si el vector dirección y es de norma 1, la derivada anterior se llama **derivada direccional**, en la dirección de y y el cálculo se simplifica con el gradiente.

$$f(x) = x^2$$

$$f: \mathbb{R} \rightarrow \mathbb{R}$$



$$y = f'(a)(x - a) + f(a)$$

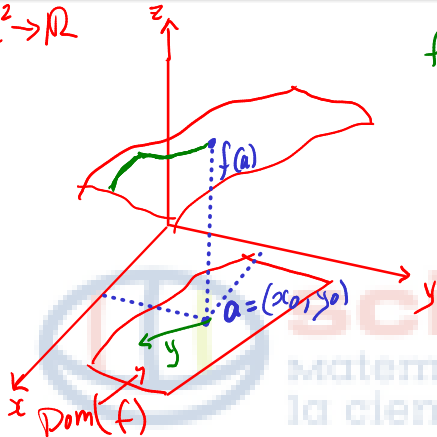
$$\frac{\Delta y}{\Delta x}$$

Ecuación de la recta tangente a f en $x = a$.

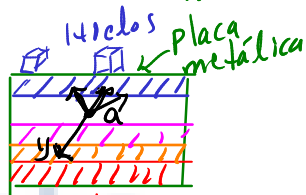
$f'(a) \mapsto$ Razón de cambio o pendiente de la recta tangente.

Diferencial. "Aproximación lineal"

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$



$$f'(a; y) = \lim_{h \rightarrow 0} \frac{f(a + hy) - f(a)}{h}$$



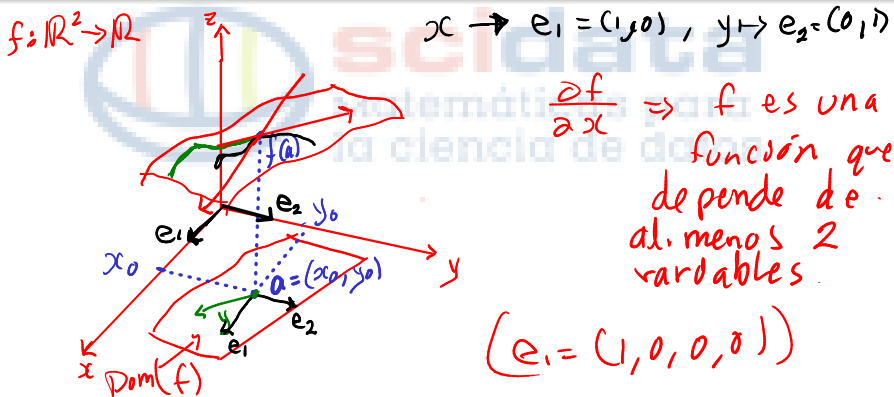
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Definición (Derivadas parciales)

Sea $f : \mathbb{R}^n \rightarrow \mathbb{R}$ un campo escalar. Definimos la derivada parcial de f en x_i , $i = 1, 2, \dots, n$, en punto $a \in \mathbb{R}^n$ como

$$\frac{\partial f}{\partial x_i} = \lim_{h \rightarrow 0} \frac{f(a + h e_i) - f(a)}{h}, \quad (2)$$

siempre que dicho límite exista. El vector e_i es el vector coordenado donde todas las componentes son cero, excepto en la coordenada i que es 1.



$$\text{Ej. } f(x, y) = x^2 + 5y^3$$

$$(f(x) \mapsto f'(x))$$

$$\frac{\partial f}{\partial x} = 2x, \quad \frac{\partial f}{\partial y} = 15y^2$$

$$\frac{\partial f(x, y)}{\partial x}$$

partial derivative x^2+5y^3

 NATURAL LANGUAGE

 MATH INPUT

Input interpretation

partial derivatives

$$x^2 + 5y^3$$

Results

$$\frac{\partial}{\partial x}(x^2 + 5y^3) = 2x$$

$$\frac{\partial}{\partial y}(x^2 + 5y^3) = 15y^2$$

3D plots

 Enlarge

$$\frac{\partial f(1,0)}{\partial x} = 2(1) = 2.$$


$$\frac{\partial f(1,0)}{\partial y} = 0$$

diff

$$f(x, y, z) = e^{-x^2 y} + \sin(y z)$$

partial derivative $e^{-(x^2 y)} + \sin(y z)$

 NATURAL LANGUAGE

 MATH INPUT

 EXTENDED

Input interpretation

partial derivatives

$$e^{-x^2 y} + \sin(y z)$$

Results

Approximate forms

$$\frac{\partial}{\partial x} (e^{-x^2 y} + \sin(y z)) = -2 x y e^{-x^2 y} = \frac{\partial f}{\partial x}$$

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

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

Differential

$$\frac{\partial f}{\partial x}(1, -1, 0) = 2e$$

$$\frac{\partial f}{\partial y}(1, -1, 0) = -e$$

$$\frac{\partial f}{\partial z}(1, -1, 0) = -1$$

Definición (Gradiente)

Sea $f : \mathbb{R}^n \rightarrow \mathbb{R}$ un campo escalar. Definimos el gradiente de f en el punto \underline{a} , denotado por $\nabla f(\underline{a})$, como

$$\nabla f(\underline{a}) = \left(\frac{\partial f(\underline{a})}{\partial x_1}, \frac{\partial f(\underline{a})}{\partial x_2}, \dots, \frac{\partial f(\underline{a})}{\partial x_n} \right). \quad \underline{a} \in \mathbb{R}^n \quad (3)$$

↑
nabla, pero
se lee gradiente de.

$f(x_1, x_2, \dots, x_n)$


$$\nabla f(\underline{x}) = \left(\frac{\partial f(\underline{x})}{\partial x_1}, \dots, \frac{\partial f(\underline{x})}{\partial x_n} \right)$$

$\underline{x} \in \mathbb{R}^n$

campo vectorial.
campos conservativos


campos escalares




campos vectoriales

$$f(x,y) = x + y + 5$$

$$\frac{\partial f}{\partial x} = 1, \frac{\partial f}{\partial y} = 1 \Rightarrow \nabla f(x,y) = (1, 1)$$



gradient of $x+y+5$

```
: plot_vector_field((1,1),(x,-3,3),(y,-3,3))
```

 NATURAL LANGUAGE


 MATH INPUT

Input interpretation

$\text{grad}(x + y + 5)$

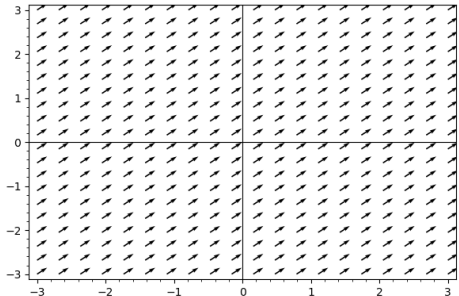
Del operator form


$\nabla(x + y + 5)$

Result in 2D Cartesian coord  Enlarge 

$\text{grad}(x + y + 5) = (1, 1)$

(x: first Cartesian coordinate | y: second Cartesian coordinate)



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 POWERED BY THE WOLFRAM LANGUAGE

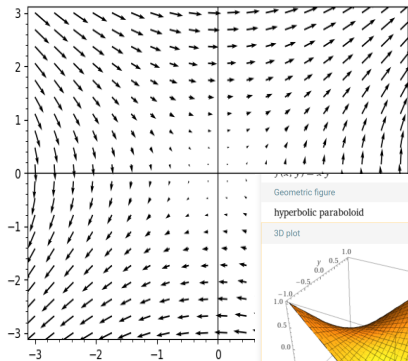
$$f(x,y) = xy$$

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

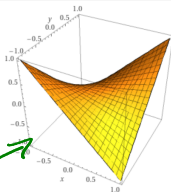
$$\Rightarrow \nabla f(x,y) = (y, x)$$

gradient of xy

```
plot_vector_field((y,x),(x,-3,3),(y,-3,3))
```



Geometric figure
hyperbolic paraboloid
3D plot



Input interpretation

$\text{grad}(x\ y)$

Del operator form

$\nabla(x\ y)$

Coordinate-free result

$\text{grad}(x\ y) = y(\text{grad } x) + x(\text{grad } y)$
assuming x is a scalar-valued function and y is a scalar-valued function

Result in 2D Cartesian coordinates

$\text{grad}(x\ y) = (y, x)$

(x : first Cartesian coordinate | y : second Cartesian coordinate)

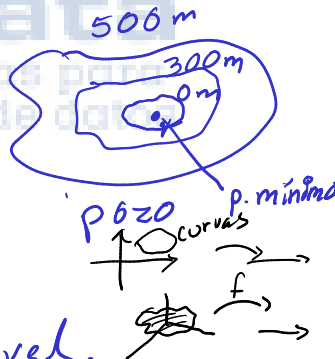
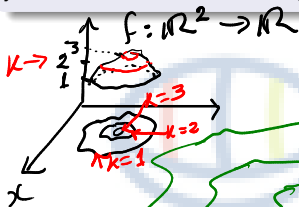
Gráfica de $f \rightarrow$

Definición (Conjuntos de nivel)

Dado un campo escalar $f: \mathbb{R}^n \rightarrow \mathbb{R}$, definimos el conjunto de nivel, al nivel $k \in \mathbb{R}$, como el conjunto siguiente:

$$CN_k = \{x \in \mathbb{R}^n : f(x) = k\}.$$

Si $n = 2$, dichos conjuntos de nivel se conocen como **curvas de nivel** (y $CN_2 = CN$), $CN \subseteq \mathbb{R}^2$, y si $n = 3$ se conocen como **superficies de nivel** ($CN_3 = SN$), $SN \subseteq \mathbb{R}^3$.



$f: \mathbb{R}^2 \rightarrow \mathbb{R}$ curvas de nivel

$f: \mathbb{R}^3 \rightarrow \mathbb{R}$ superficies de nivel.

$$f(x,y) = x + y$$

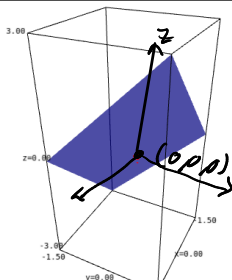
$$z = f(x,y)$$

$$z = x + y \Rightarrow x + y - z = 0 \quad \text{E. lineal.}$$

plano

1. Gráfica del plano $f(x,y) = x + y$.

```
var('x,y,z')
f(x,y)=x+y
p=plot3d(f(x,y),(x,-1.5,1.5),(y,-1.5,1.5),color='blue',opacity=0.7)
show(p)
```

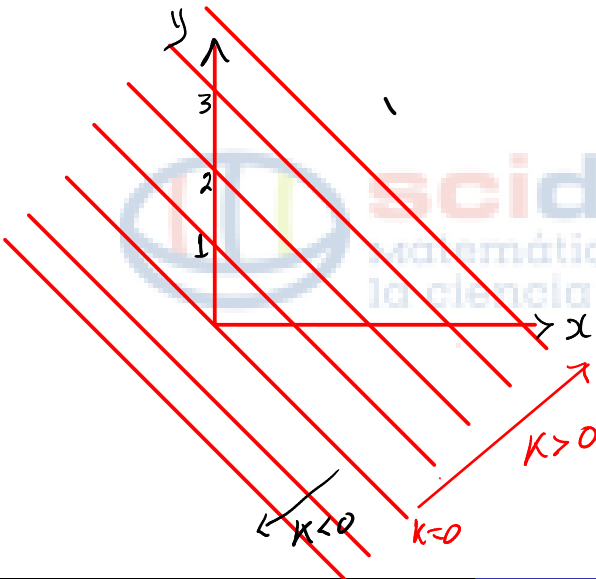


$$z = f(x, y) = x + y = k$$

plano

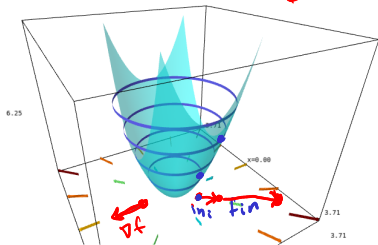
$$y = -x + k$$

$$k \in \mathbb{R}$$



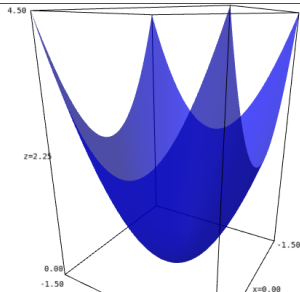
$$f(x,y) = x^2 + y^2 = K$$

cónica · $c = (0,0)$ $r = \sqrt{K}$, $K \geq 0$



gráfica de $f(x,y) = -x^2 - y^2$.

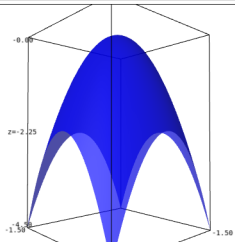
`1.5),(y,-1.5,1.5),color='blue',opacity=0.7)`



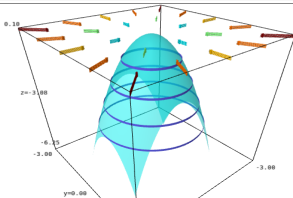
$$f(x,y) = -x^2 - y^2 = k \Rightarrow x^2 + y^2 = -k$$

$$C = (0,0) \quad r = \sqrt{-k} \quad k \leq 0$$

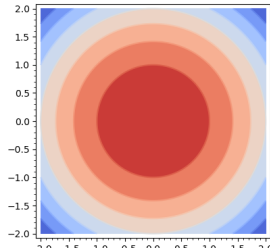
```
var('x,y,z')
f(x,y)=-x^2-y^2
p=plot3d(f(x,y),(x,-1.5,1.5),(y,-1.5,1.5),color='blue',opacity=0.7)
show(p)
```



```
var('x,y,z')
f(x,y)=-x^2-y^2
g(x,y,z)=(-2*x,-2*y,0)
levels=[-1,-2,-3,-4]
epsilon=0.1
p=plot3d(f(x,y),(x,-2,2),(y,-1.5,1.5),color='cyan',opacity=0.5)
for h in levels:
    p+=implicit_plot3d(f(x,y)==h,(x,-2,2),(y,-2,2),(z,h+epsilon))
q=plot_vector_field3d(g,(x,-3,3),(y,-3,3),(z,0,0.1))
show(q)
```

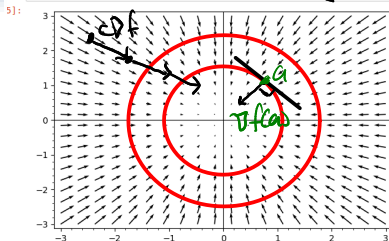


```
4]: var('x,y')
p = contour_plot(-x^2-y^2, (x,-2,2), (y,-2,2), cmap="coolwarm")
show(p)
```



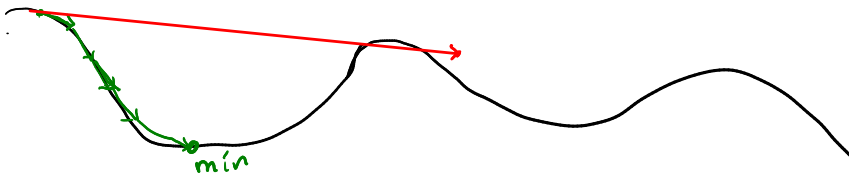
max
min
u →

```
5]: plot_vector_field((-2*x,-2*y),(x,-3,3),(y,-3,3))
```



Todos los elementos juntos de $f(x,y) = -x^2 - y^2$.





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Definición (Campo escalar diferenciable)

Sea $f : \mathbb{R}^n \rightarrow \mathbb{R}$ un campo escalar. Decimos que f es diferenciable en $a \in \mathbb{R}^n$, si existe una transformación lineal $T_a : \mathbb{R}^n \rightarrow \mathbb{R}$, y una función escalar E_a tal que

$$f(a+v) = f(a) + T_a(v) + E_a$$

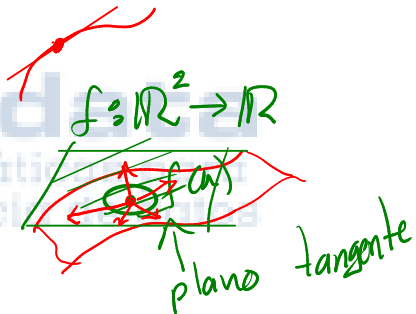
$$f'(a) = \lim_{h \rightarrow 0} \frac{f(a+h) - f(a)}{h}$$

$$f'(a) \approx \frac{f(a+h) - f(a)}{h}$$

$$f'(a) \cdot h \approx f(a+h) - f(a)$$

$$f(a+h) = f(a) + f'(a) \cdot h + e.$$

Fórmula de Taylor de 1^{er} orden



Diferenciable \Rightarrow derivable \Rightarrow continua
Transformación lineal número

Teorema (Condición suficiente de diferenciabilidad)

Sea $f: \mathbb{R}^n \rightarrow \mathbb{R}$ un campo escalar. Decimos que f es diferenciable en a con diferencial T_a , si $\frac{\partial f(a)}{\partial x_i}$ es continua para toda $i = 1, 2, \dots, n$ y $T_a(v) = \nabla f(a) \cdot v$.

Extra. $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$f(x_1, \dots, x_n) = (f_1, f_2, \dots, f_m)$$

$$A_T \in M_{m \times n}$$

$$A = J = \begin{bmatrix} \nabla f_1 \\ \nabla f_2 \\ \vdots \\ \nabla f_m \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \dots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

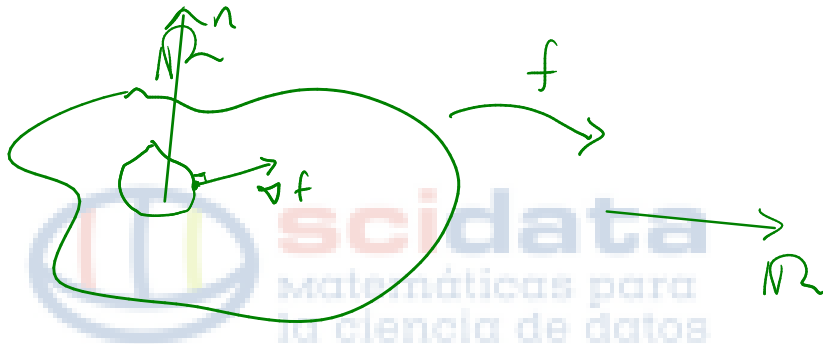
Matriz

Jacobiana

$|J|$

\rightarrow Jacobiano (determinante de J)

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

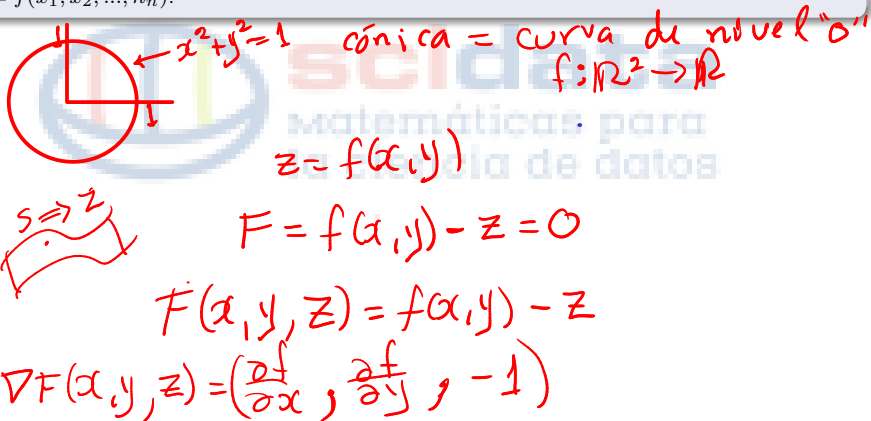


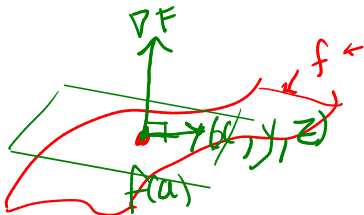
Teorema (Plano tangente)

Sea $f : \mathbb{R}^n \rightarrow \mathbb{R}$ un campo escalar. Si f es diferenciable en $a = (a_1, a_2, \dots, a_n)$, entonces existe un plano tangente en el punto $f(a)$, el cual está dado por

$$\nabla F(a) \cdot x = 0,$$

donde $F : \mathbb{R}^{n+1} \rightarrow \mathbb{R}$ esta dada por $F(a) = (a_1, a_2, \dots, a_n, f(a_1, a_2, \dots, a_n))$,
 $x_{n+1} = f(x_1, x_2, \dots, x_n)$.



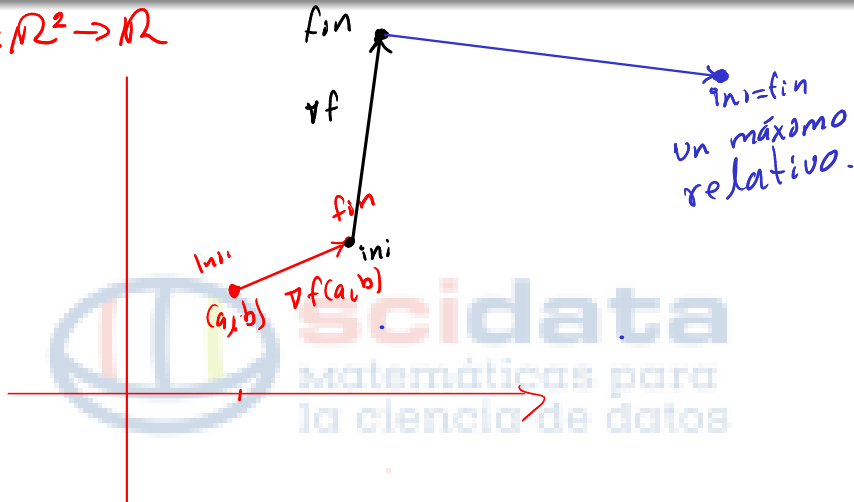


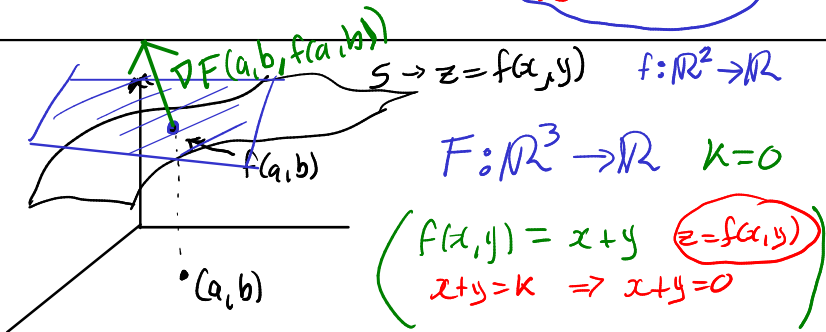
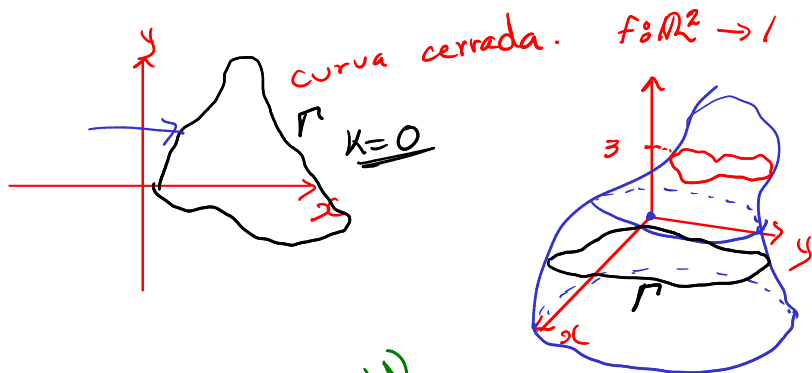
$$\nabla F \cdot (x, y, z) = 0$$

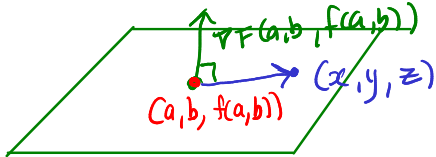


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$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$







$$\nabla F(a, b, f(a, b)) \cdot ((x, y, z) - (a, b, f(a, b))) = 0$$

$\nabla F(a, b, f(a, b))$ E. plano tangente a f en (a, b) .

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

en $(1, 2)$

$$F(x, y, z) = x^2 + y^2 - z$$

$$\nabla F(x, y, z) = (2x, 2y, -1)$$

$$z = f(x, y)$$

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

$$x^2 + y^2 - z = 0 \leftarrow k$$

$$z = (1)^2 + 2^2 = 5$$

$$\Rightarrow \nabla F(1, 2, 5) = (2, 4, -1) \quad \therefore (1, 2, 5)$$

$$\Rightarrow \nabla F(1, 2, 5) \cdot ((x, y, z) - (1, 2, 5)) = 0$$

$$(2, 4, -1) \cdot (x-1, y-2, z-5) = 0$$

$$2(x-1) + 4(y-2) - (z-5) = 0$$

$$2x - 2 + 4y - 8 - z + 5 = 0$$

$$2x + 4y - z - 5 = 0$$

$$z = 2x + 4y - 5$$

$$g(x, y) = 2x + 4y - 5$$