

Today: 12.3, 12.4 - Partial derivatives

~~Quiz~~ Recall:

Def $f(x,y)$ - a function of 2 variables.

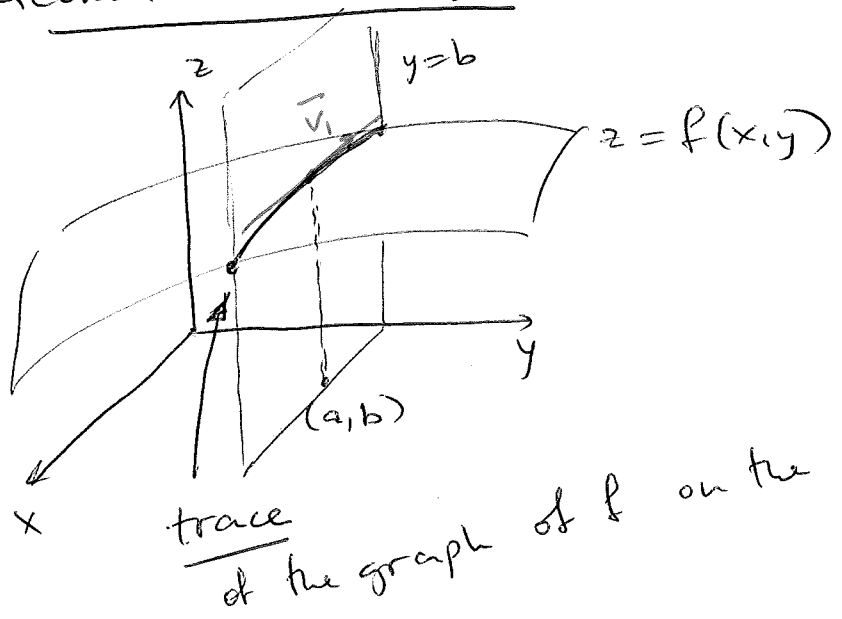
$$f_x(a,b) = \frac{\partial f}{\partial x} \Big|_{(a,b)} = \lim_{h \rightarrow 0} \frac{f(a+h, b) - f(a,b)}{h}$$

↑
the book uses $f_x(a,b)$.
I will not.

↑
x-partial derivative of f at the point (a,b)

$$\frac{\partial f}{\partial y} \Big|_{(a,b)} = \lim_{h \rightarrow 0} \frac{f(a, b+h) - f(a,b)}{h}$$

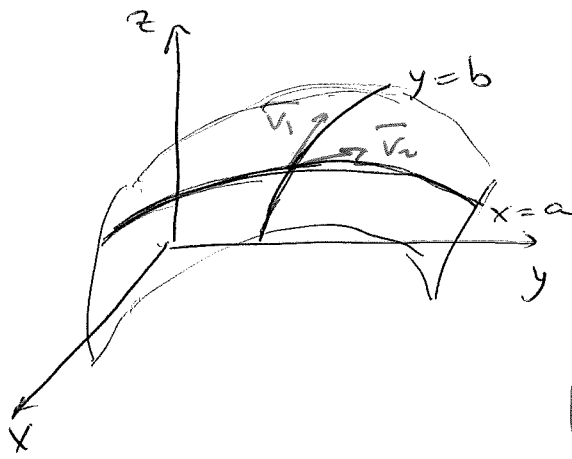
Geometric meaning



$\frac{\partial f}{\partial x} \Big|_{(a,b)}$ is the slope of the tangent line to the trace on $y=b$ plane

↑
in the sense of $\frac{\Delta z}{\Delta x}$

Fix $x=a$,
 $\frac{\partial f}{\partial y} \Big|_{(a,b)}$ is the slope of the tangent line to the trace on the plane $x=a$.



The tangent vectors to the traces span the tangent plane to the graph at the point $(a, b, f(a, b))$

- if the function is smooth enough

- will come back to it, see weird examples.

Assume f is "smooth enough"

How to find the equation of the tangent plane to the graph of f ?

2 ways:

① " f is smooth enough" should mean that we can write a reasonable linear approximation to $f(x, y)$ in a neighbourhood of (a, b) :

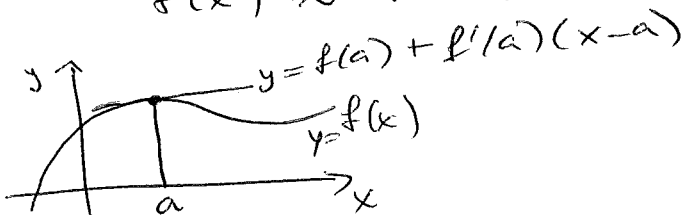
$$f(x, y) \approx f(a, b) + \frac{\partial f}{\partial x} \Big|_{(a, b)} (x-a) + \frac{\partial f}{\partial y} \Big|_{(a, b)} (y-b)$$

linear approximation to $f(x, y)$ near (a, b) . Its graph is a plane. This is the tangent plane at (a, b) .

/ Recall: single variable: linear approx.

$$f(x) \approx f(a) + f'(a)(x-a) + \frac{f''(a)}{2}(x-a)^2 + \dots$$

higher-order terms.



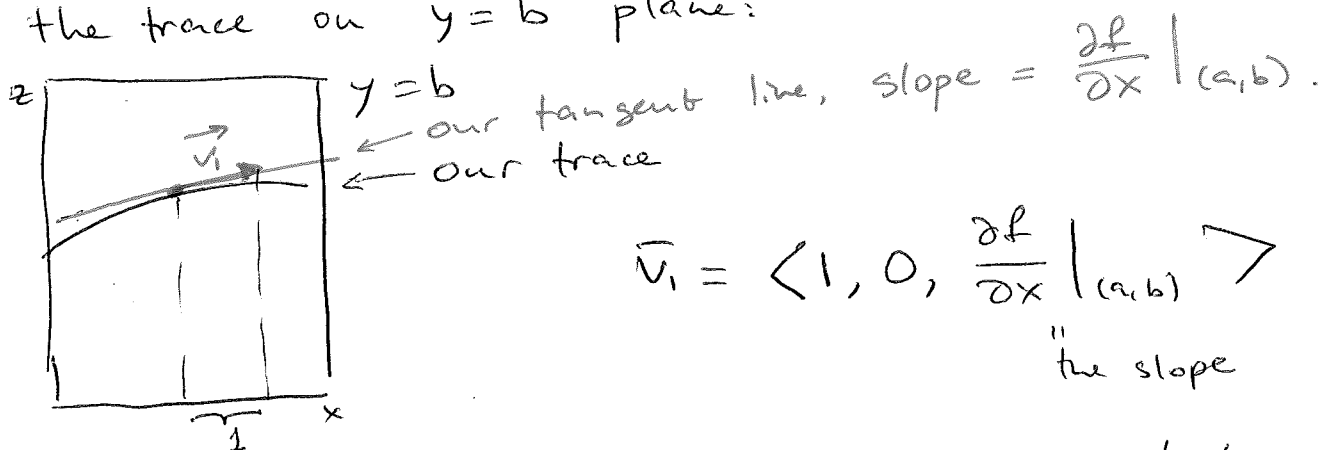
② Second way of thinking about the tangent plane:

Write an equation of ^{the} plane containing the two tangent lines to the traces on $x=a$, $y=b$.

Normal: cross product of the two tangent vectors to these traces.

The tangent vectors to the traces:

the trace on $y=b$ plane:



In the same way, the direction vector of the tangent line to the trace on $x=a$ plane

is: $\vec{v}_2 = \left\langle 0, 1, \frac{\partial f}{\partial y} \Big|_{(a,b)} \right\rangle$

$$\vec{n} = \begin{vmatrix} \hat{i} & \hat{j} & \hat{k} \\ 1 & 0 & \frac{\partial f}{\partial x} \Big|_{(a,b)} \\ 0 & 1 & \frac{\partial f}{\partial y} \Big|_{(a,b)} \end{vmatrix} = \left\langle -\frac{\partial f}{\partial x} \Big|_{(a,b)}, -\frac{\partial f}{\partial y} \Big|_{(a,b)}, 1 \right\rangle$$

~~Another~~ Note: if we use this vector and the point $(a, b, f(a,b))$, we get exactly the same equation as before!

Note: The first way works in any dimension, but this only works for a function of 2 variables.

Def: For $f(x_1, \dots, x_n)$: linear approx. at
the point (a_1, \dots, a_n) is

$$\begin{aligned} f(x_1, \dots, x_n) \approx & f(a_1, \dots, a_n) + \frac{\partial f}{\partial x_1} \Big|_{(a_1, \dots, a_n)} (x_1 - a_1) \\ & + \frac{\partial f}{\partial x_2} \Big|_{(a_1, \dots, a_n)} (x_2 - a_2) + \dots \\ & + \frac{\partial f}{\partial x_n} \Big|_{(a_1, \dots, a_n)} (x_n - a_n) \end{aligned}$$

One can prove that this is the best possible
of all linear approximations.

Read: 12.4.