

GERMAN UNIVERSITY IN CAIRO

Lectures 18

Math301

Fall 2020

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1 Lecture 3

Rule 1. Clairaut's Thm.

if the functions F_{xy} and F_{yx} are continuous on an open Disk D

$$F_{xy} = F_{yx} \quad (1)$$

Note 1. Please note that Clairaut's thm. holds for any number of variables, that is $f_{zh} = f_{hz}$. So it's true for any two vars. (2nd partial derivative)

Rule 2. Direction vector from a slope

Recall that if m = slope of the tangent line, then:

$$v = (1, m) \quad (2)$$

Rule 3. Same goes with partial derivatives:

$$V1 = (1, 0, \frac{\partial f}{\partial x}) \quad V2 = (0, 1, \frac{\partial f}{\partial y}) \quad (3)$$

Note 2. the zero in the aforementioned vectors comes from differentiating with respect to a constant :) as we all know that in order to partial differentiate you fix one variable and diff wrt to the other.

Rule 4. Find the partial derivatives:

in these questions we find f_x, f_y, f_z and so on.

Rule 5. Find the mixed partial derivatives:

now we find $f_{xx}, f_{yy}, f_{xy}, f_{yx}$

Rule 6. The meaning/interpretation of $\frac{\partial f}{\partial w}$, where w is any variable x, y, z .

The instantaneous change of f with respect to w while holding all the other variables fixed.

Rule 7. Find a tangent plane to a surface:

We know that a two variable function's graph is a surface or a shape in space. We also know how to get two tangent lines/vectors to surface at any point.

So Assuming that the surface is **Smooth** and has no peaks: a normal to the surface at that same point can be obtained:

$$n = V1 \times V2 = \begin{vmatrix} i & j & k \\ 1 & 0 & \frac{\partial f}{\partial x} \\ 0 & 1 & \frac{\partial f}{\partial y} \end{vmatrix} = (-f_x, -f_y, 1) \quad (4)$$

and now the equation of the plane is

$$PM.n = 0 \quad (5)$$

that makes sense because n is normal to any vector in the plane. M is any point that belongs to the plane so $M = (x, y, z)$

and finally the equation of the plane is:

$$z = f(a, b) + f_x(a, b) * (x - a) + f_y(a, b) * (y - b) \quad (6)$$

Rule 8. find Linear Approximation of a two variable function:

Tangent planes that we studied now provide a good approximation of the original function at that point, same as we did with linear approximation in one var functions; the tangent line assumes a near value to the function.

Hence, we use tangent planes to find Linear approximations:

$$L(a, b) = f(a, b) + f_x(a, b) * (x - a) + f_y(a, b) * (y - b) \quad (7)$$

For $(x, y) \approx (a, b)$ (very close to)

Rule 9. Using differentials and linearization

we use them together in order to find the difference between two close values of the real function. smth like finding the error of calculating volume of box, The inputs (x, y, z) has two values $x_1, x_2, y_1, y_2, z_1, z_2$.

However, using $L(a, b, c) = f(a, b, c) + f_x(a, b, c) * (x - a) + f_y(a, b, c) * (y - b) + f_z(a, b, c) * (z - c)$ **together with** $\Delta f = f(x, y, z) - f(a, b, c)$ **and** $f(x, y, z) \approx L(x, y, z)$

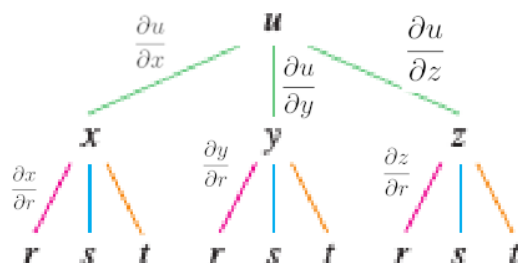
We find that;

$$\Delta f \approx df = f_x(x - a) + f_y(y - b) + f_z(z - c) \quad (8)$$

it's the same as L but without the function at that point $f(a, b, c)$

2 Lecture 4

Rule 10. Solving problems including chain rule is just drawing a tree diagram:



after drawing the dependency tree as show, solve!

for example:

Exercise 1. Find $\frac{\partial u}{\partial r}$

Solution 1.

$$\frac{\partial u}{\partial r} = \frac{\partial u}{\partial x} * \frac{\partial x}{\partial r} + \frac{\partial u}{\partial y} * \frac{\partial y}{\partial r} + \frac{\partial u}{\partial z} * \frac{\partial z}{\partial r} \quad (9)$$

Note 3. please note that

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial}{\partial x} \left[\frac{\partial f}{\partial y} \right] = f_{yx} \quad (10)$$

The notation is just inverted

Rule 11. Directional Derivatives ?

When we get $\frac{\partial f}{\partial x}$ we try to find the derivative of f with respect to x or in the x direction.

but what if we want to get the Directional derivative in any other direction $u = (u_1, u_2)$ or wrt to any other variable ? Note that that variable is a combination of the original variables.

However:

$$D_u(f(a, b)) = \frac{d}{ds} \Big|_{s=0} f(a + su_1, b + su_2) \quad (11)$$

this rule is a bit amibgious so lets look at the easy form of it

$$D_u(P_0) = \nabla \mathbf{f} \cdot \mathbf{u} \quad (12)$$

Rule 12. Gradient Vectors

$$\nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right) = \frac{\partial f}{\partial x} i + \frac{\partial f}{\partial y} j \quad (13)$$

and for the 3d case it's the same but we add another variable, hence another term in the vector in k direction

So the directional derivative is simply a **dot product** of the gradient vector and a unit vector in the direction we need.

Rule 13. The magnitude of change and its significance

$$|D_u| = |\nabla f| * |u| * \cos(\theta) \quad (14)$$

1. At any point P_0 the function f increases most rapidly in the direction where $\theta = 0$ or equally in the direction of the gradient vector.
this also means that $u/\|\nabla f\|$ and $|D_u| = \|\nabla f\|$
2. Similarly, f decreases most rapidly in the direction where $\theta = \pi$ so that $|D_u| = -\|\nabla f\|$
3. lastly f faces no change when $\theta = \frac{\pi}{2}$ so that $|D_u| = 0$ and ∇f is perpendicular to u

Rule 14. How to find the normal to any given vector ?

$$\text{let } v \cdot u = 0 \quad (15)$$

and solve it algebraically then you find v

Rule 15. Gradient Vectors are normal to **Level Curves**

The proof is simple; we know that level curves are obtained when $f = k$ so when f is constant. Constant means there is no change and hence from the above rule and third point in the list, We conclude that the grad. vector is perp. to the curve which is a level curve.

So **Gradient Vectors** are also perp. to **Level Surfaces** in case of a three variable function.

Note 4. A function of one variable can be thought of as a level curve to another function of two variables. HOW?

$$y = x^2 \text{ This is a normal function of one var} \quad (16)$$

$$f = y - x^2 = 0 \text{ This is a level curve of another function } f=k \text{ where } k=0 \quad (17)$$

Same goes with a function of two variables can be thought of as a level surface of another function.

$$z^2 = x^2 + y^2 \quad (18)$$

$$f = x^2 + y^2 - z^2 = 0 \quad (19)$$

why we did that ? in order to find a unit normal to the graph of z we can get it using $n = u_1 \times u_2$ and find u_1, u_2 using the rule no. 7

or simply we can make it a level surface and then finding ∇f is just that easy.

Rule 16. unit vectors to level surfaces point downward or upward?

you know that by looking at the components of the gradient vector or simply draw that gradient/unit normal in a 3d plane and look at it

3 Lecture 5

Rule 17. Function optimization (finding local and absolute extrema points)

To find the extrema points we have two methods

1. First derivative test
2. Lagrange Multiplier

Rule 18. First Derivative Test

you find solutions of the following equations :

$$\nabla f = 0 \tag{20}$$

Which is equally

$$f_x = 0 \tag{21}$$

$$f_y = 0 \tag{22}$$

$$f_z = 0 \tag{23}$$

The solution of those equations gives us **Critical** points

Rule 19. Checking if critical points are local max/min using Second deriv. Test

$$D(a, b) = f_{xx}(a, b) * f_{yy}(a, b) - [f_{xy}(a, b)]^2 \tag{24}$$

1. if $D > 0$
 - (a) if $f_{xx} < 0$ f has a local maximum point at (a,b)
 - (b) if $f_{xx} > 0$ f has a local minimum extrema at (a,b)
2. $D < 0$ Then the point is a saddle point (not local max nor min)
3. $D = 0$ Then the test fails

Rule 20. What is D ? it's called the **Hessian** or the **Discriminant** of the function f

$$D = \begin{vmatrix} f_{xx} & f_{yy} \\ f_{yx} & f_{yy} \end{vmatrix} \tag{25}$$

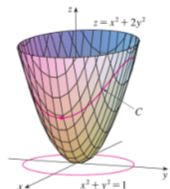
Note 5. This first derivative test is used to find critical points which are inside the domain of f, not on the boundary of that domain. At the boundary you either use another method or find it by substitution and some math **please look at the following example**

Example

Find the absolute extreme values of the function $z = f(x, y) = x^2 + 2y^2$ in the domain $x^2 + y^2 \leq 1$.

Solution

The absolute extreme values of $f(x, y)$ are either attained in the interior of the domain or at points on the boundary of the domain.



In the interior of the domain (the open disk $\{(x, y) | x^2 + y^2 < 1\}$), an extreme value is attained at a critical point (as implied by the first derivative theorem for extreme values).

Critical points are given by:

$$f_x = 0 \text{ and } f_y = 0 \iff 2x = 0 \text{ and } 4y = 0.$$

Thus, $(0, 0)$ is the only critical point. One may apply the second derivative test in order to characterize this point but it is straight forward to see that $f(0, 0) = 0$ is an absolute minimum of the function over the domain, thus in particular, over its interior.

On the boundary (which is the circle $x^2 + y^2 = 1 \iff y^2 = 1 - x^2$), the function $f(x, y)$ reduces to a single variable function

$$z = x^2 + 2(1 - x^2) = 2 - x^2, \text{ where } -1 \leq x \leq 1.$$

Obviously, z attains its absolute maximum of 2 at $x = 0$ (which is the unique critical point of z in the interval $(-1, 1)$).

Consequently, $f(x, y)$ attains an absolute maximum at the points where $x = 0$ and $y^2 = 1 - x^2$. That is, at the points $(0, \pm 1)$.

Conclusion, in the domain $x^2 + y^2 \leq 1$ the function $f(x, y)$ has absolute minimum of 0 attained at $(0, 0)$, an absolute maximum of 2 attained at $(0, \pm 1)$.

Rule 21. Method of lagrange multipliers

This method is used when you have a constraint (like the previous boundary problem)

$$\nabla f(x, y) = \lambda \nabla g(x, y) \tag{26}$$

$$g(x, y) = k \tag{27}$$

Solving the two equations together and finding λ will yield solutions for the critical points your goal is to find all the values of x, y, λ that satisfies the equations and then substitute with those points

1. the point that yields the largest value of f is a maximum value
2. same with the smallest is the minimum value