

§13.1 Functions of Several Variables

1. Definition: A function like $f(x, y)$, $f(x, y, z)$, $g(s, t)$ etc.
2. Definition of the graph of a function of two variables and classic examples like: Plane, paraboloid, cone, parabolic sheet, hemisphere.
3. Definition of level curve for $f(x, y)$ and level surface for $f(x, y, z)$.
4. Graphs of surfaces which are not necessarily functions: Sphere, ellipsoid, cylinder sideways parabolic sheet like $y = x^2$, double-cone.

§13.2 Limits and Continuity

1. Nothing much said other than $\lim_{(x,y) \rightarrow (x_0,y_0)} f(x, y)$ asks what $f(x, y)$ approaches as (x, y) gets closer to (x_0, y_0) .

§13.3 Partial Derivatives

1. Defn: We can define *the partial derivative of $f(x, y)$ with respect to x* , denoted $\frac{\partial f}{\partial x}$ or f_x , as the derivative of f treating all variables other than x as constant. Similarly for any variable for any function.
2. For $f(x, y)$ it turns out f_x and f_y give the slopes of the lines tangent to the graph of $f(x, y)$ at the point (x, y) in the positive x and positive y directions respectively. A picture can clarify.
3. Higher derivatives will also be used but there are some points to note:
 - (a) f_{xy} means $(f_x)_y$ so first take the derivative with respect to x and then y .
 - (b) $\frac{\partial^2 f}{\partial x \partial y}$ means $\frac{\partial}{\partial x} \left(\frac{\partial f}{\partial y} \right)$ so first take the derivative with respect to y and then x .
 - (c) $\frac{\partial^2 f}{\partial x^2}$ means x both times.
 - (d) It turns out that 99% of the time the order doesn't matter so for example $f_{xy} = f_{yx}$.

§13.4 The Chain Rule

1. Consider: For example if f is a function of x and y which are both functions of s and t then really f is a function of s and t and so $\frac{\partial f}{\partial s}$ and $\frac{\partial f}{\partial t}$ make sense. How to find them?
2. The chain rule says:
 - (a) First draw the tree diagram.
 - (b) For each route from the starting to ending variable write down the product of the derivatives along that path.
 - (c) Add the paths.
3. The chain rule is good for related rates problems when multiple rates are given and one rate is needed.

§13.5 Directional Derivative

1. Intro: We saw that f_x and f_y (for example) give derivatives in specific directions (the \hat{i} and \hat{j} directions) and so what if we asked for the derivative (slope) in another direction?
2. Defn: If $\bar{u} = a\hat{i} + b\hat{j}$ is a unit vector then *the directional derivative of f in the direction of \bar{u}* is $D_{\bar{u}}f = af_x + bf_y$. If we have 3D then $+cf_z$ on the end. Sometimes we use the term “directional derivative” when the direction is not a unit vector so we must make it a unit vector first.
3. A good analogy is that $f(x, y, z)$ is temperature and $D_{\bar{u}}f$ gives us temperature change (slope) in a specific direction.

§13.6 The Gradient

1. Defn: The gradient of f , denoted $\text{grad } f$ or ∇f , is defined as $\nabla f = f_x\hat{i} + f_y\hat{j} + f_z\hat{k}$ in 3D.
2. Properties:
 - (a) For any \bar{u} we see $D_{\bar{u}}f = \bar{u} \cdot \nabla f$.
 - (b) Since $D_{\bar{u}}f = \bar{u} \cdot \nabla f = \|\bar{u}\|\|\nabla f\|\cos\theta = \|\nabla f\|\cos\theta$ we see that the directional derivative is maximum when $\theta = 0$ which shows that the gradient points in the direction of maximum directional derivative.
 - (c) It also shows that the actual value of the maximum directional derivative is $\|\nabla f\|$.
 - (d) In the 2D case ∇f is perpendicular to the level curve for $f(x, y)$ at (x, y) . If we want a vector perpendicular to the graph of a function $f(x)$ we need to rewrite as $y = f(x)$ then $f(x) - y = 0$ and then the graph of the function is the level curve for $g(x, y) = f(x) - y$ and we use ∇g .
 - (e) In the 3D case ∇f is perpendicular to the level surface for $f(x, y, z)$ at (x, y, z) . If we want a vector perpendicular to the graph of a function $f(x, y)$ we need to rewrite as $z = f(x, y)$ then $f(x, y) - z = 0$ and then the graph of the function is the level surface for $g(x, y, z) = f(x, y) - z$ and we use ∇g .

§13.8 Extreme Values

1. Defn: Relative maximum/minimum/extremum for $f(x, y)$. Method:
 - (a) First find where both f_x and f_y are zero or one is undefined. Those are the *critical points*.
 - (b) Find the discriminant $D(x, y) = f_{xx}f_{yy} - (f_{xy})^2$ and then for each critical point:
 - If $D(x, y) < 0$ then (x, y) is a saddle point.
 - If $D(x, y) > 0$ and $f_{xx}(x, y) < 0$ then (x, y) is a relative maximum.
 - If $D(x, y) > 0$ and $f_{xx}(x, y) > 0$ then (x, y) is a relative minimum.

Good examples: $f(x, y) = x^2 + 2y^2 - 6x + 8y + 1$ and $f(x, y) = 3x^2 - 3xy^2 + y^3 + 3y^2$.

2. Defn: Absolute m/m/e of $f(x, y)$ on a closed and bounded region R . Method:
 - (a) Find all CP for $f(x, y)$ which are inside the region. Take f of those.
 - (b) Find the maximum and minimum of f on the edge of the region. Usually this involves combining f with the equation for the region (sometimes part by part) and then getting f in a form where we can see what the max and min would be.
 - (c) Pick out the largest and smallest values from the previous two steps.

Good examples: $f(x, y) = x^2 - y^2$ with $\frac{x^2}{4} + y^2 \leq 1$ and $f(x, y) = 3x - y$ on the triangle with vertices $(0, 0)$, $(0, 3)$ and $(6, 0)$.

§13.9 Lagrange Multipliers

1. Idea: If (x, y) are constrained by a level curve $g(x, y) = c$ and we want to find the maximum of $f(x, y)$ how do we do it?
2. Thm: If a max/min occurs at (x, y) then $\nabla f = \lambda \nabla g$ at that point so the method is:
 - (a) We set those equal and solve those along with the constraint. In other words we solve the system: $f_x = \lambda g_x$, $f_y = \lambda g_y$ and $g(x, y) = c$.
 - (b) The result are potential winners. We take each (x, y) we get and plug it into f , picking out the largest and smallest.

Good Examples: $f(x, y) = 2x + 3y$ with $x^2 + y^2 = 9$, $f(x, y) = xy$ with $(x - 1)^2 + y^2 = 1$ and $f(x, y) = x^2 + y^2$ with $2x + 6y = 10$.