

Taylor series in multi-variables

Recall for single variable function $f(x)$, it has a Taylor series, expanded around 0,

~~$f(x)$~~ \Rightarrow

$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(0)}{n!} x^n$$

Taylor series.
Maclaurin series.

where $f^{(n)}$ is the n^{th} derivative $\frac{d^n f}{dx^n}$

which is valid for some neighborhood of 0, which could be all of \mathbb{R} for certain functions.

For partial finite sums of the series.

$$f(x) \approx \sum_{n=0}^k \frac{f^{(n)}(0)}{n!} x^n$$

we get an approximation to f , getting better with $k \rightarrow \infty$.

What about for multi-variable functions?

Consider a function $f(x, y)$ and its behaviour near (a, b)

we'll write $h := \Delta x$ and $k := \Delta y$, for the small variations in x, y . Then the Taylor series for f about (a, b) will be

$$f(a+h, b+k) = \sum_{n=0}^{\infty} \frac{1}{n!} (D^n f)(a, b)$$

where D is the differential operator

$$D = h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y}$$

and by $D^n f$ we mean

$$D^n f = \underbrace{D(D(D(\dots(Df)\dots))\dots)}_{n \text{ copies of } D}$$

n copies of D .

$$\text{So } D^0 f = f$$

$$Df = h \frac{\partial f}{\partial x} + k \frac{\partial f}{\partial y}$$

$$\begin{aligned} D^2 f &= D(Df) \\ &= \left(h \frac{\partial}{\partial x} + k \frac{\partial}{\partial y} \right) \left(h \frac{\partial f}{\partial x} + k \frac{\partial f}{\partial y} \right) \end{aligned}$$

around $(\pi/2, 0) = (a, b)$ to get an approximation.

$$f(\pi/2 + h, k) \approx \text{poly in } h, k.$$

$$(D^0 f)(\pi/2, 0) = f(\pi/2, 0)$$

$$= 1$$

$$(D^1 f)(\pi/2, 0) = \left(h \frac{\partial f}{\partial x} + k \frac{\partial f}{\partial y} \right) \Big|_{(x,y)=(\pi/2,0)}$$

$$= \left[\left(h(\cos(x+3y) - 3\sin(3x+y)) \right) + k(\underbrace{3\cos(x+3y) - \sin(3x+y)}_{\text{at } (\pi/2,0)}) \right]_{(\pi/2,0)}$$

$$= 3h + k$$

For $D^2 f$, we'll need

at $(\pi/2, 0)$

$$\frac{\partial^2 f}{\partial x^2} = -\sin(x+3y) - 9\cos(3x+y) = -1$$

$$\frac{\partial^2 f}{\partial y^2} = -9\sin(x+3y) - \cos(3x+y) = -9$$

$$\frac{\partial^2 f}{\partial x \partial y} = -3\sin(x+3y) - 3\cos(3x+y) = -3$$

$$\text{So } (D^2 f)(\pi/2, 0) = -h^2 - 9k^2 - 6hk$$

So putting this together gives the order 2 approximation.

$$f\left(\frac{\pi}{2} + h, k\right) \approx 1 + 3h + k - \frac{1}{2}h^2 - \frac{9}{2}k^2 - 3hk$$

Let's visualize this with computer.

This all extends to 3 or more variables in "obvious" way

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

then

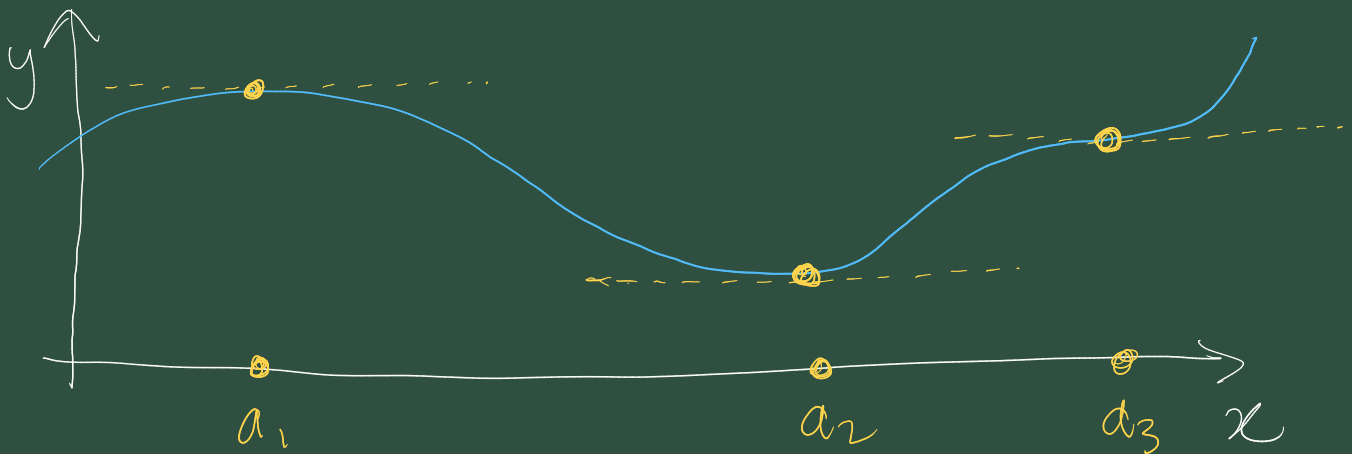
$$f(a_1 + h_1, \dots, a_n + h_n) = \sum_{n=0}^{\infty} \frac{1}{n!} (D^n f)(a_1, \dots, a_n)$$

where
$$D = \sum_{j=1}^n h_j \frac{\partial}{\partial x_j}$$

Optimisation in 2 variables.

Finding and classifying critical/stationary points on surface defined by $f(x, y)$

1-variable case $y = f(x)$



Three types of critical point local maximums (a_1), local minimums (a_2), inflection point (a_3), all characterised

by $\frac{df}{dx} = 0$ and were classified

by $\frac{d^2f}{dx^2}$ as

$$\left. \frac{d^2 f}{dx^2} \right|_{a_1} < 0 \Rightarrow \text{local max.}$$

$$\left. \frac{d^2 f}{dx^2} \right|_{a_2} > 0 \Rightarrow \text{local min.}$$

$$\left. \frac{d^2 f}{dx^2} \right|_{a_3} = 0 \Rightarrow \text{inflection point}$$

For the surface $z = f(x, y)$

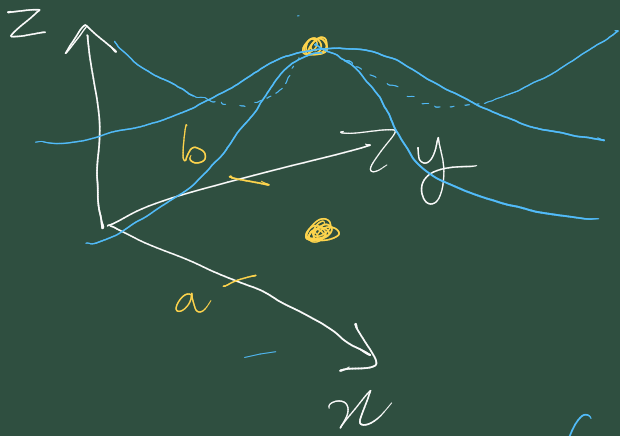
Define a critical point is a point (a, b) where both partial derivatives vanish

$$\text{i.e. } \frac{\partial f}{\partial x}(a, b) = 0 \quad \& \quad \frac{\partial f}{\partial y}(a, b) = 0.$$

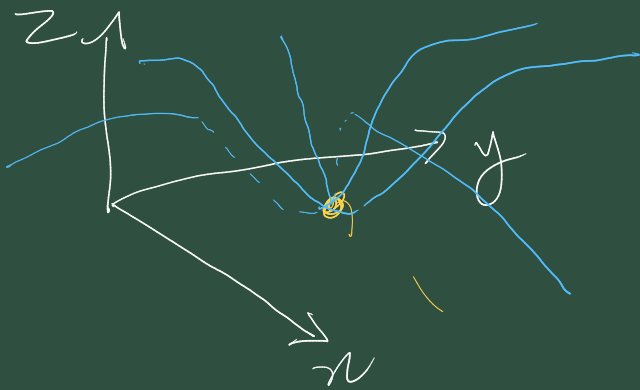
So geometrically the tangent plane to the surface at (a, b) is horizontal.

Again they come in three types

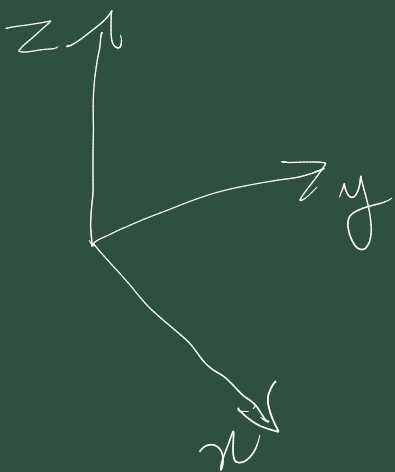
local max. ("top of a hill")



local min. ("bottom of a crater")



saddle point (mixture of a min/max)



Again these are classified using
second-order derivatives.
The Hessian determinant is

$$D = \begin{vmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{vmatrix}$$

$$= \frac{\partial^2 f}{\partial x^2} \frac{\partial^2 f}{\partial y^2} - \left(\frac{\partial^2 f}{\partial x \partial y} \right)^2$$

And if

$D(a,b) < 0$ then (a,b) is a saddle

if $D(a,b) > 0$

then $\begin{cases} \frac{\partial^2 f}{\partial x^2}(a,b) > 0 \Rightarrow \text{local min.} \\ \frac{\partial^2 f}{\partial x^2}(a,b) < 0 \Rightarrow \text{local max.} \end{cases}$

we can justify by analysing the 2nd order Taylor poly.

Example

Consider

