

Taylor Theorem (single-variable) Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is of class C^k and $x_0 \in \mathbb{R}$. Then

$$f(x_0+h) = f(x_0) + \underbrace{\sum_{i=1}^k \frac{1}{i!} f^{(i)}(x_0) h^i}_{\text{Taylor polynomial}} + R(x_0, h)$$

where $\frac{R(x_0, h)}{h^k} \rightarrow 0$ as $h \rightarrow 0$. ▣

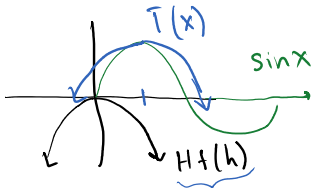
Ex $f(x) = \sin x$ on $[0, \pi]$. Locate all local extrema.

Find critical points: $f'(x) = \cos x = 0 \Rightarrow \boxed{x = \frac{\pi}{2}}$

Compute Taylor Polynomial at $x_0 = \frac{\pi}{2}$:

Set $x = x_0 + h = \frac{\pi}{2} + h$ to get

$$\begin{aligned} f(x) &\approx f\left(\frac{\pi}{2}\right) + f'\left(\frac{\pi}{2}\right)(x - \frac{\pi}{2}) + \frac{1}{2} f''\left(\frac{\pi}{2}\right)(x - \frac{\pi}{2})^2 \\ &= \underbrace{1 - \frac{1}{2}(x - \frac{\pi}{2})^2}_T \quad \text{Define } \underbrace{Hf(h) = -\frac{1}{2}h^2}_{\text{Hessian}} \end{aligned}$$



according to the Taylor Thm,

$\frac{\pi}{2}$ is a max for $\sin x$ since the best quadratic approximation is concave down. We want to generalize this idea to C^k maps $f: \mathbb{R}^n \rightarrow \mathbb{R}$. ▣

Taylor Theorem (second-degree multivariable case) Suppose $f: \mathbb{R}^n \rightarrow \mathbb{R}$ of class C^2 . Let $\vec{h} = (h_1, \dots, h_n)$ and let $\vec{x}_0 \in \mathbb{R}^n$. Then

$$f(\vec{x}_0 + \vec{h}) = f(\vec{x}_0) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\vec{x}_0) h_i + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\vec{x}_0) h_i h_j + R_2(\vec{x}_0, \vec{h})$$

where $\frac{\|R_2(\vec{x}_0, \vec{h})\|}{\|\vec{h}\|^2} \rightarrow 0$ as $\vec{h} \rightarrow \vec{0}$.

Proof (sketch) Consider the line $c(t) = \vec{x}_0 + t\vec{h}$. Compose w/ f :

$$g(t) = f(c(t)).$$

By the single-variable Taylor theorem:

$$g(t) = g(0) + g'(0)t + \frac{1}{2}g''(0)t^2 + R$$

$$\Rightarrow f(\vec{x}_0 + t\vec{h}) = g(t) = g(0) + g'(0)t + \frac{1}{2}g''(0)t^2 + R$$

compute using chain rule:

$$\begin{aligned} g'(t) &= (f \circ c)'(t) = \nabla f(c(t)) \cdot c'(t) & c'(t) &= h = (h_1, \dots, h_n) \\ &= \nabla f(\vec{x}_0 + t\vec{h}) \cdot c'(t) \leftarrow \\ &= \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\vec{x}_0 + t\vec{h}) h_i \end{aligned}$$

$$\begin{aligned} g''(t) &= \sum_{i=1}^n \frac{d}{dt} \left(\frac{\partial f}{\partial x_i}(\vec{x}_0 + t\vec{h}) \right) h_i \\ &= \sum_{i=1}^n \left(\sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\vec{x}_0 + t\vec{h}) h_j \right) h_i \end{aligned}$$

Evaluate at $t=0$ to get

$$\begin{aligned} f(\vec{x}_0 + t\vec{h}) &= g(t) = g(0) + g'(0)t + \frac{1}{2}g''(0)t^2 + R \\ &= f(\vec{x}_0) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\vec{x}_0) h_i + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(\vec{x}_0) h_i h_j + R \end{aligned}$$

We skip the statement about the remainder R . ▣

Ex Taylor Polynomial for $f(x,y) = e^{x+y}$ at $x_0 = (0,0)$.

All partial derivatives at $(0,0)$ are equal to 1

$$T(x,y) = 1 + x + y + \frac{1}{2}(x^2 + xy + yx + y^2)$$
▣

First-Derivative Test: Suppose $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is differentiable and suppose x_0 is a local max/min of f . Then x_0 is a critical point.

Proof Define $g(t) = (f \circ c)(t)$ where $c(t) = \underline{x}_0 + t\vec{h}$ for any $h \in \mathbb{R}^n$.

Then $g(t)$ has a local max/min when $t=0$, so

$$\begin{aligned} 0 &= g'(0) = \nabla f(c(0)) \cdot c'(0) \\ &= \nabla f(x_0) \cdot h & h &= (h_1, \dots, h_n) \\ &= \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x_0) \cdot h_i \end{aligned}$$

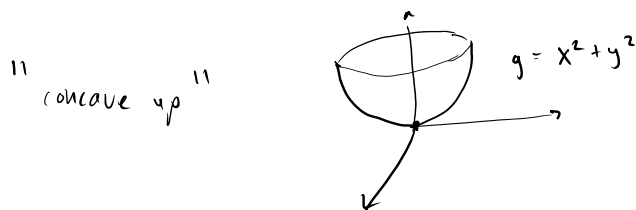
Since h is arbitrary, $\frac{\partial f}{\partial x_i}(x_0) = 0$ for all $i=1, \dots, n$. So $Df(x_0) = 0$. ▣

Def A quadratic function $g: \mathbb{R}^n \rightarrow \mathbb{R}$ of the form

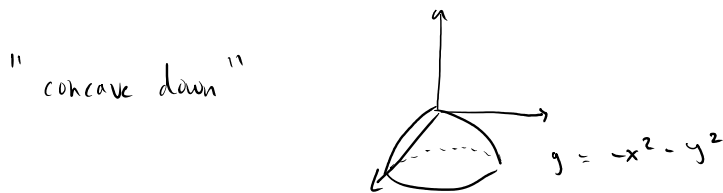
$$g(h_1, \dots, h_n) = \sum_{i=1}^n \sum_{j=1}^n a_{ij} h_i h_j, \quad a_{ij} = a_{ji}$$

$$= [h_1, \dots, h_n] \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} h_1 \\ \vdots \\ h_n \end{bmatrix}$$

(1) g is positive-definite if $g(h) \geq 0$ and $g(h) = 0 \Leftrightarrow h = 0$

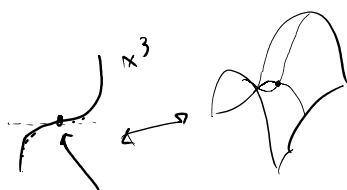


(2) g is negative-definite if $g(h) \leq 0$ and $g(h) = 0 \Leftrightarrow h = 0$.



Def (Hessian) Let $f: \mathbb{R}^n \rightarrow \mathbb{R}$ be of class C^2 at $x_0 \in \mathbb{R}^n$. The Hessian is the quadratic func

$$Hf(x_0)(h_1, \dots, h_n) = \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(x_0) h_i h_j$$



$$\det \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} = 0$$

$$= \frac{1}{2} [h_1, \dots, h_n] \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \dots & \frac{\partial^2 f}{\partial x_n \partial x_n} \end{bmatrix} \begin{bmatrix} h_1 \\ \vdots \\ h_n \end{bmatrix}$$

Now assume x_0 is a critical point. By Taylor theorem,

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$$= f(\vec{x}_0) + \nabla f(\vec{x}_0) \cdot \vec{h} + Hf(\vec{x}_0)(\vec{h}) + R$$

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Second-Der. Test Let $f: \mathbb{R}^n \rightarrow \mathbb{R}$ be of class C^2 and let x_0 be a critical point.

- (1) If $Hf(x_0)$ is positive-definite, then x_0 is a local min
 (2) ——— // ——— negative-definite, ——— // ——— local max.