

Example Session for:

Week 10 A: Connections between Total / Directional / Partial Derivatives

Week 10 B: Chain Rule, Gradient, Higher-order Derivatives, Taylor Expansion

Jacobian, Hessian, Taylor

Let us consider $f: \mathbb{R}^2 \rightarrow \mathbb{R}, (x_1, x_2) \mapsto f(x_1, x_2) = x_1 e^{x_1 x_2}$.

We find:

• The Jacobian is

$$\begin{aligned} \mathcal{D}f|_x = \nabla f(x) &= \begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \end{pmatrix} f(x_1, x_2) = \begin{pmatrix} \frac{\partial}{\partial x_1} (x_1 e^{x_1 x_2}) \\ \frac{\partial}{\partial x_2} (x_1 e^{x_1 x_2}) \end{pmatrix} = \begin{pmatrix} e^{x_1 x_2} + x_1 x_2 e^{x_1 x_2} \\ x_1^2 e^{x_1 x_2} \end{pmatrix} \\ &= \begin{pmatrix} 1 + x_1 x_2 \\ x_1^2 \end{pmatrix} e^{x_1 x_2} \end{aligned}$$

• Furthermore:
$$\begin{aligned} \frac{\partial^2 f}{\partial x_1^2} &= \frac{\partial}{\partial x_1} \left((1 + x_1 x_2) e^{x_1 x_2} \right) = x_2 e^{x_1 x_2} + (1 + x_1 x_2) x_2 e^{x_1 x_2} \\ &= (2x_2 + x_1 x_2^2) e^{x_1 x_2} \end{aligned}$$

$$\begin{aligned} \frac{\partial^2 f}{\partial x_2 \partial x_1} &= \frac{\partial}{\partial x_2} \left((1 + x_1 x_2) e^{x_1 x_2} \right) = x_1 e^{x_1 x_2} + (1 + x_1 x_2) x_1 e^{x_1 x_2} \\ &= (2x_1 + x_1^2 x_2) e^{x_1 x_2} \end{aligned}$$

$$\begin{aligned}\frac{\partial^2 f}{\partial x_1 \partial x_2} &= \frac{\partial}{\partial x_1} (x_1^2 e^{x_1 x_2}) = 2x_1 e^{x_1 x_2} + x_1^2 x_2 e^{x_1 x_2} \\ &= (2x_1 + x_1^2 x_2) e^{x_1 x_2}\end{aligned}$$

$$\frac{\partial^2 f}{\partial x_2^2} = \frac{\partial}{\partial x_2} (x_1^2 e^{x_1 x_2}) = x_1^2 e^{x_1 x_2}$$

Hence, the Hessian is $H_f(x) = \begin{pmatrix} 2x_2 + x_1 x_2^2 & 2x_1 + x_1^2 x_2 \\ 2x_1 + x_1^2 x_2 & x_1^2 \end{pmatrix} e^{x_1 x_2}$.

Let us consider f near $x_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$. We find:

$$f(1,1) = e$$

$$\nabla f|_{(1,1)} = \begin{pmatrix} 2e \\ e \end{pmatrix}$$

$$H_f|_{(1,1)} = \begin{pmatrix} 3e & 3e \\ 3e & e \end{pmatrix}$$

Hence, the second-order Taylor expansion of f around $(1,1)$ reads:

$$\begin{aligned}f\left(\begin{pmatrix} 1 \\ 1 \end{pmatrix} + h\right) &= f(1,1) + \langle \nabla f|_{(1,1)}, h \rangle + \frac{1}{2} \langle h, H_f|_{(1,1)} h \rangle + r(h) \\ &= e + (2e, e) \begin{pmatrix} h_1 \\ h_2 \end{pmatrix} + \frac{1}{2} (h_1, h_2) \begin{pmatrix} 3e & 3e \\ 3e & e \end{pmatrix} \begin{pmatrix} h_1 \\ h_2 \end{pmatrix} + r(h)\end{aligned}$$

Leibniz integral rule

Consider $I(x) = \int_a^b f(x,t) dt = F(x,b) - F(x,a)$ (where $\frac{\partial F}{\partial t} = f$). keep x fixed

i.e. F is the antiderivative of f w.r.t. the t variable

$$\text{Then } \frac{dI(x)}{dx} = \frac{\partial F(x,b)}{\partial x} - \frac{\partial F(x,a)}{\partial x}$$

FTC \rightarrow

$$= \int_a^b \frac{\partial}{\partial t} \left(\frac{\partial F(x,t)}{\partial x} \right) dt$$

Clairaut/Schwarz \rightarrow

$$= \int_a^b \frac{\partial}{\partial x} \underbrace{\frac{\partial F(x,t)}{\partial t}}_{= f(x,t)} dt$$

$$\Rightarrow \boxed{\frac{d}{dx} \int_a^b f(x,t) dt = \int_a^b \frac{\partial f(x,t)}{\partial x} dt}$$

Without proof, we note the following:

Theorem: This formula is true if $f(x,t)$ and $\frac{\partial f(x,t)}{\partial x}$ are continuous.

Example: $I(x) = \int_a^b \frac{\sin(xt)}{t} dt$, for $0 < a < b$ (s.t. conditions of theorem hold).

$I(x)$ is not so easy to integrate. But we can easily compute:

$$\frac{dI(x)}{dx} = \int_a^b \frac{\partial}{\partial x} \left(\frac{\sin(xt)}{t} \right) dt = \int_a^b \cos(tx) dt = \frac{\sin(tx)}{x} \Big|_a^b = \frac{\sin(bx)}{x} - \frac{\sin(ax)}{x}$$