

Quick Review from previous lecture

- Orientations in a closed surface:
 - **Outward pointing normal**: normal points outward.
 - **Inward pointing normal**: normal points inward.

- **The divergence (Gauss) theorem** says that

Let F be a smooth vector field on W . Then

$$\int \int \int_W (\nabla \cdot F) dV = \int \int_{\partial W} \overbrace{F \cdot d\mathbf{S}}^{\text{flux}}, \quad (1)$$

where W has boundary ∂W , oriented with outward pointing normal.

This means

“The total expansion of the fluid inside 3D region W ” equals

“the total flux of the fluid out of the boundary of W ”

Quiz 9: 8.4, 3.1, 3.2.

3.1 Iterated partial derivatives

$f(x, y)$ scalar valued.

If $f : \mathbb{R}^2 \rightarrow \mathbb{R}^1$, then $\frac{\partial f}{\partial x}$, $\frac{\partial f}{\partial y}$ are also functions of two variables. The partial derivatives of $\frac{\partial f}{\partial x}$, $\frac{\partial f}{\partial y}$ are

$$f_{xx} = \frac{\partial^2 f}{\partial x^2} = \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial x} \right), \quad f_{yy} = \frac{\partial^2 f}{\partial y^2} = \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial y} \right)$$
$$f_{xy} = \frac{\partial^2 f}{\partial y \partial x} = \frac{\partial}{\partial y} \left(\frac{\partial f}{\partial x} \right), \quad f_{yx} = \frac{\partial^2 f}{\partial x \partial y} = \frac{\partial}{\partial x} \left(\frac{\partial f}{\partial y} \right)$$

Example 1. Let $f(x, y) = e^{xy} + y \cos(x)$. Find f_{xx} , f_{xy} , f_{yx} , f_{yy} .

$$f_x = y e^{xy} + (-y \sin x)$$

$$f_y = x e^{xy} + \cos x.$$

$$f_{xx} = \frac{\partial}{\partial x} (f_x) = \frac{\partial}{\partial x} (y e^{xy} - y \sin x) = y^2 e^{xy} - y \cos x.$$

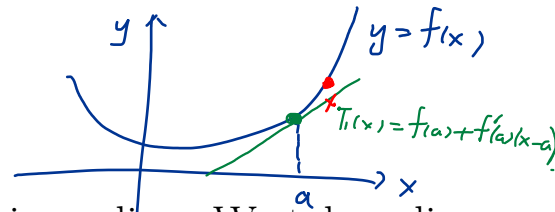
$$f_{xy} = \frac{\partial}{\partial y} (f_x) = \underline{e^{xy} + xy e^{xy} - \sin x}.$$

$$f_{yy} = \frac{\partial}{\partial y} (f_y) = \frac{\partial}{\partial y} (x e^{xy} + \cos x) = x^2 e^{xy} + 0.$$

$$f_{yx} = \frac{\partial}{\partial x} (f_y) = \underline{e^{xy} + xy e^{xy} - \sin x}.$$

we saw that $f_{xy} = f_{yx}$. H

3.2 Taylor's Theorem



Recall “what is Linear approximation”:
 That is, we want to approximate $f(x)$ near $x = a$ by using a line. We take a line through the point $(a, f(a))$ with slope $f'(a)$:
1 variable.

$$T_1(x) = f(a) + f'(a)(x - a).$$

We call it the **first order Taylor polynomial** (or Linear approximation) of f near a .
(approximation)

In Calculus 2, you also learned the **second order Taylor polynomial** (or quadratic approximation) of f near a :

$$T_2(x) = f(a) + f'(a)(x - a) + \frac{f''(a)}{2!}(x - a)^2. \quad (2)$$

n-th Taylor polynomial $T_n(x) =$  $+ \dots + \frac{1}{n!} f^{(n)}(a)(x-a)^n$

Now let's start section 3.2.

We want to generalize the Taylor polynomial to functions of multiple variables.

Fact. (Taylor Polynomials for $f : \mathbb{R}^n \rightarrow \mathbb{R}^1$)

We consider a C^2 function $f : \mathbb{R}^n \rightarrow \mathbb{R}^1$ with n variables. Let f be differentiable at a . Denote

$$x = (x_1, x_2, \dots, x_n),$$

$$a = (a_1, a_2, \dots, a_n).$$

- Then the **first order Taylor polynomial (approximation)** of f at a is

$$T_1(x) = f(a) + \overbrace{\mathbf{D}f(a)}^{\text{matrix of partial derivatives}}(x - a),$$

that is,

$$T_1(x) = f(a) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(a)(x_i - a_i).$$

- Then the **second order Taylor polynomial (approximation)** of f at a is

$$T_2(x) = f(a) + \mathbf{D}f(a)(x - a) + \frac{1}{2!}(x - a)^T \mathbf{H}f(a)(x - a),$$

that is,

$$T_2(x) = f(a) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(a)(x_i - a_i) + \frac{1}{2!} \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(a)(x_i - a_i)(x_j - a_j).$$

$$Df(a) = \left[\frac{\partial f}{\partial x_1}(a) \quad \dots \quad \frac{\partial f}{\partial x_n}(a) \right]$$

$$\begin{aligned} Df(a) [x - a] &= \left[\frac{\partial f}{\partial x_1}(a) \quad \dots \quad \frac{\partial f}{\partial x_n}(a) \right] \begin{bmatrix} x_1 - a_1 \\ x_2 - a_2 \\ \vdots \\ x_n - a_n \end{bmatrix} \\ &= \frac{\partial f}{\partial x_1}(a)(x_1 - a_1) + \frac{\partial f}{\partial x_2}(a)(x_2 - a_2) + \dots + \frac{\partial f}{\partial x_n}(a)(x_n - a_n). \end{aligned}$$

Hessian matrix of f :

$$Hf(a) = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1} & \frac{\partial^2 f}{\partial x_1 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2 \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_2 \partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \frac{\partial^2 f}{\partial x_n \partial x_2} & \dots & \frac{\partial^2 f}{\partial x_n \partial x_n} \end{bmatrix}_{n \times n}$$

$$(x - a)^T Hf(a) (x - a) = [x_1 - a_1 \quad \dots \quad x_n - a_n] \begin{bmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1} & \dots & \frac{\partial^2 f}{\partial x_1 \partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1} & \dots & \frac{\partial^2 f}{\partial x_n \partial x_n} \end{bmatrix}_{n \times n} \begin{bmatrix} x_1 - a_1 \\ \vdots \\ x_n - a_n \end{bmatrix}$$

Example 2. Find 1st and 2nd order Taylor approximation of

$$f(x, y) = 2x^2 + xy + 4y^2 - 1$$

at the point $(x_0, y_0) = (1, 2)$.

$$f(1, 2) = 19$$

$$f_x = 4x + y, \quad f_x(1, 2) = 6$$

$$f_y = x + 8y, \quad f_y(1, 2) = 17$$

$$f_{xx} = \frac{\partial}{\partial x} (4x + y) = 4.$$

$$f_{yy} = \frac{\partial}{\partial y} (x + 8y) = 8.$$

$$f_{xy} = \frac{\partial}{\partial y} (4x + y) = 1.$$

$$\begin{aligned} T_1(x, y) &= f(1, 2) + f_x(1, 2)(x-1) + f_y(1, 2)(y-2) \\ &= 19 + 6(x-1) + 17(y-2). \end{aligned}$$

$$\begin{aligned} T_2(x, y) &= f(1, 2) + f_x(1, 2)(x-1) + f_y(1, 2)(y-2) \\ &\quad + \frac{1}{2} \left[f_{xx}(1, 2)(x-1)^2 + f_{yy}(1, 2)(y-2)^2 \right. \\ &\quad \left. + 2f_{xy}(1, 2)(x-1)(y-2) \right]. \end{aligned}$$

$$= 19 + 6(x-1) + 17(y-2) + \frac{1}{2} \left[4(x-1)^2 + 8(y-2)^2 + 2(x-1)(y-2) \right]$$

Example 3. Consider the function $f(x, y, z) = (x^2 + y^2 + z^2)^{1/2}$.

1. Find a linear approximation of f near $(4, 4, 2)$.
 (1st Taylor polynomial)

$$f(4, 4, 2) = (4^2 + 4^2 + 2^2)^{1/2} = (36)^{1/2} = 6.$$

$$f_x = \frac{1}{2}(x^2 + y^2 + z^2)^{-1/2} 2x, \quad f_x(4, 4, 2) = \frac{1}{2}(16 + 16 + 4)^{-1/2} 8$$

$$f_y = \frac{1}{2}(x^2 + y^2 + z^2)^{-1/2} 2y, \quad f_y(4, 4, 2) = \frac{1}{2}(16 + 16 + 4)^{-1/2} 8 = \frac{1}{2} \cdot \frac{1}{6} \cdot 8 = \frac{2}{3}.$$

$$f_z = \frac{1}{2}(x^2 + y^2 + z^2)^{-1/2} 2z, \quad f_z(4, 4, 2) = \frac{1}{3}.$$

$$T_1(x, y, z) = 6 + \frac{2}{3}(x-4) + \frac{2}{3}(y-4) + \frac{1}{3}(z-2). \quad \#$$

2. Estimate the value $(4.01^2 + 3.99^2 + 2.03^2)^{1/2}$ by using the linear approximation you found in (a).
 ↗ real value (calculator: we get 6.01008...)

$$\left((4.01)^2 + (3.99)^2 + (2.03)^2 \right)^{1/2}$$

$$\sim T_1(4.01, 3.99, 2.03) = 6 + \frac{2}{3}(0.01) + \frac{2}{3}(-0.01) + \frac{1}{3}(0.03)$$

$$= 6.01.$$

(approximated value)