

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$(x_1, y_1) \mapsto f(x_1, y_1)$$

$$Df = \nabla f = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right\rangle$$

General chain rule:

$$\left[ \frac{\partial f_i}{\partial x_j} \right] \left[ \frac{\partial g_j}{\partial y_k} \right]$$

$$f: \mathbb{R} \rightarrow \mathbb{R}^2$$

$$t \mapsto \langle x(t), y(t) \rangle$$

$$Df = \frac{df}{dt} = \begin{bmatrix} \frac{dx}{dt} \\ \frac{dy}{dt} \end{bmatrix}$$

$$\begin{array}{ccccc} & & f(g(\underline{x})) & & \\ & a & \xrightarrow{g} & b & \xrightarrow{f} c \\ \mathbb{R} & \xrightarrow{a} & \mathbb{R} & \xrightarrow{b} & \mathbb{R} \\ \underline{x} & \xrightarrow{g(\underline{x})} & & \xrightarrow{f(g(\underline{x}))} & \\ Dg(\underline{x}) & & & Df(g(\underline{x})) & \\ bx_a & & & cx_b & \\ \text{matrix} & & & \text{matrix} & \end{array}$$

$$D(f(g(\underline{x}))) = Df(g(\underline{x})) Dg(\underline{x})$$

↑  
matrix multiplication

Example

$$\textcircled{1} \quad \mathbb{R} \xrightarrow{g} \mathbb{R}^2 \xrightarrow{f} \mathbb{R}$$

$$t \quad (x_1, y_1)$$

$$\begin{aligned} Dg &= g'(t) & Df &= \nabla f \\ &= \begin{bmatrix} x'_1(t) \\ y'_1(t) \end{bmatrix} & &= \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right\rangle \end{aligned}$$

$$D(f(g(t))) = Df(g(t)) \cdot Dg(t)$$

$$= \nabla f(g(t)) \cdot g'(t)$$

$$= \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right\rangle \cdot \left\langle \frac{dx}{dt}, \frac{dy}{dt} \right\rangle$$

$$(\text{or} \quad \left[ \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right] \left[ \begin{bmatrix} \frac{dx}{dt} \\ \frac{dy}{dt} \end{bmatrix} \right]).$$

$$= \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt}$$

$$\textcircled{2} \quad \mathbb{R}^2 \xrightarrow{g} \mathbb{R}^2 \xrightarrow{f} \mathbb{R}^2$$

$$x_1, x_2 \quad y_1, y_2$$

$$g(\underline{x}) = (g_1(x_1, x_2), g_2(x_1, x_2))$$

$$f(y_1, y_2) = (f_1(y_1, y_2), f_2(y_1, y_2))$$

$$Dg = \begin{bmatrix} \frac{\partial g_1}{\partial x_1} & \frac{\partial g_1}{\partial x_2} \\ \frac{\partial g_2}{\partial x_1} & \frac{\partial g_2}{\partial x_2} \end{bmatrix}$$

$$Df = \begin{bmatrix} \frac{\partial f_1}{\partial y_1} & \frac{\partial f_1}{\partial y_2} \\ \frac{\partial f_2}{\partial y_1} & \frac{\partial f_2}{\partial y_2} \end{bmatrix}$$

$$D(f(g(\underline{x}))) = Df(g(\underline{x})) \cdot Dg(\underline{x})$$

$$\text{Df Pg} = \begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{bmatrix} \begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{bmatrix} = \begin{bmatrix} \frac{\partial g_1}{\partial x_1} \frac{\partial f_1}{\partial y_1} + \frac{\partial g_1}{\partial x_2} \frac{\partial f_2}{\partial y_2} & \dots & \dots \\ \dots & \dots & \dots \end{bmatrix}$$

Mnemonic  $f(x_1, \dots, x_n)$      $x_i(y_1, y_2, \dots, y_m)$

to find  $\frac{\partial f_i}{\partial y_j} = \frac{\partial f}{\partial x_1} \frac{\partial x_1}{\partial y_j} + \frac{\partial f}{\partial x_2} \frac{\partial x_2}{\partial y_j} + \dots + \frac{\partial f}{\partial x_n} \frac{\partial x_n}{\partial y_j}$

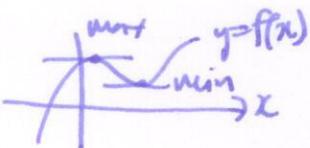
"differentiate  $f$  wrt all variables, multiply by  $\frac{\partial x_k}{\partial y_j}$  and add".

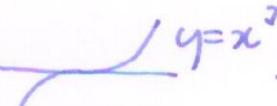
Example  $f(x_1, y_1, z) = xy + z^3$      $x = s+t$      $g: \mathbb{R}^2 \rightarrow \mathbb{R}$   
 $f: \mathbb{R}^3 \rightarrow \mathbb{R}$      $y = s-t$      $(s, t) \mapsto (x_1, y_1, z)$   
 $z = st$

so  $f(g(s, t))$  makes sense.

Q find  $\frac{\partial f}{\partial s} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial s} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial s} + \frac{\partial f}{\partial z} \frac{\partial z}{\partial s}$

### § 14.7 Optimization

recall: 1d  max, min  $\Rightarrow \frac{dy}{dx} = 0$

$\frac{dy}{dx} \not\Rightarrow$  max, min 

2d local max  local min   $\rightarrow$  flat tangent plane  $z = \text{const.}$

recall: tangent plane to  $z = f(x, y)$  at  $(a, b)$  is

$$z = f(a, b) + f_x(a, b)(x-a) + f_y(a, b)(y-b)$$

flat tangent plane  $\Rightarrow f_x(a, b) = 0$  and  $f_y(a, b) = 0$ .

warning:  $f_x(a, b) = 0$  and  $f_y(a, b) = 0 \not\Rightarrow$  local max or min.

Defn A critical point  $(a, b)$  is a point s.t.  $\frac{\partial f}{\partial x}(a, b) = 0$  and  $\frac{\partial f}{\partial y}(a, b) = 0$  or at least one of  $f_x(a, b), f_y(a, b)$  does not exist.

Thm If  $f(x, y)$  has a local max or min at  $(a, b)$ , then  $(a, b)$  is a critical point.

Example finding critical points

$$f(x, y) = x^2 - 2xy + 2y^2 + 3y + 1$$

$$\begin{aligned} \frac{\partial f}{\partial x} &= 2x - 2y = 0 \quad \text{①} \\ \frac{\partial f}{\partial y} &= -2x + 4y + 3 = 0 \quad \text{②} \end{aligned} \quad \left. \begin{array}{l} \text{①+②: } 2y + 3 = 0 \quad y = -\frac{3}{2} \\ \text{exactly one critical point at } (-\frac{3}{2}, -\frac{3}{2}) \end{array} \right\}$$

Types of critical point

contours/level sets



local max



local min



saddle

more complicated...  
monkey saddle



Q: how to tell which one? A: look at 2nd order (quadratic approximation).

1 var:  $f(x) \approx f(a) + f'(a)(x-a) + \frac{1}{2} f''(a)(x-a)^2 + \dots$

$\stackrel{=0 \text{ if}}{\text{critical point}}$

$f'(a) > 0$		$\text{min}$
$f''(a) < 0$		$\text{max}$
$f''(a) = 0$		no information!

2 vars: Thm (2nd derivative test). Let  $(a, b)$  be a critical point for  $f(x, y)$

let  $D(a, b) = f_{xx}(a, b)f_{yy}(a, b) - f_{xy}(a, b)^2$ , then:

- 1) if  $D > 0$  then  $f(a, b)$  is a minimum if  $f_{xx}(a, b) > 0$   
maximum if  $f_{xx}(a, b) < 0$
- 2) if  $D < 0$  then  $f(a, b)$  is a saddle
- 3) if  $D = 0$  no information

Why does this work?

$$\text{Quadratic approximation is } f(x,y) = f(a,b) + f_x(a,b)(x-a) + [x-a] \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix} \begin{bmatrix} x-a \\ y-b \end{bmatrix} + f_y(a,b)(y-b)$$

quadratic terms are:  $f_{xx}(a,b)(x-a)^2 + 2f_{xy}(a,b)(x-a)(y-b) + f_{yy}(a,b)(y-b)^2$

this is a quadratic form, like:  $x^2y^2 \quad -x^2y^2 \quad xy + x^2y^2$

up to change of coordinates, a symmetric matrix looks like

$\begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$	$\lambda_1 x^2 + \lambda_2 y^2$	$\lambda_1, \lambda_2 > 0$	local min
		$\lambda_1, \lambda_2 < 0$	local max
		different signs at least one 0	saddle
			don't know.

D = det of matrix = product of eigenvalues

so  $D > 0$  same sign

$D < 0$  different sign.

Example find the critical points of  $f(x,y) = (x^2+y^2)e^{-2x} = x^2e^{-2x} + y^2e^{-2x}$ .

$$f_x = 2xe^{-2x} + x^2(-e^{-2x}) - 2y^2e^{-2x}$$

$$f_y = 2ye^{-2x}$$

$$\begin{array}{l} \text{solve } f_x = 0 \\ f_y = 0 \end{array} \left. \begin{array}{l} e^{-2x}(2x - 2x^2 - 2y^2) \\ e^{-2x}(2y) = 0 \end{array} \right\} \Rightarrow \begin{array}{l} 2(x-x^2) = 0 \\ x(x-1) = 0 \end{array} \Rightarrow x = 0, 1. \quad (0,0) \quad (1,0).$$

$$f_{xx} = 2x(-2e^{-2x}) + 2e^{-2x} + 2x \quad \left. \right\} D = f_{xx}f_{yy} - (f_{xy})^2$$

$$f_{xy} = -4ye^{-2x}$$

$$f_{yy} = 2e^{-2x}$$

$$D(0,0), D(1,0) \text{ etc. ...}$$