

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

midterm: differentiability of multi-variable function.
midterm ~ : " vector-valued multi-variable functions

$f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

differentiability = linear approximation
multi-variable function; $L(\vec{x}) = f(\vec{a}) + \nabla f(\vec{a}) \cdot (\vec{x} - \vec{a})$
gradient of f

vector-valued multi-variable function: \hookrightarrow Jacobian matrix

Recall matrix multiplication

$A: m \times n$ matrix, $B: n \times k$ matrix

$\Rightarrow AB$ $m \times k$ matrix

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} e & f & g \\ h & i & j \end{pmatrix} = \begin{pmatrix} ae+bh & af+bi & ag+bj \\ ce+dh & cf+di & gj+dj \end{pmatrix}$$

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} = \begin{pmatrix} \vec{a}_1 \\ \vdots \\ \vec{a}_m \end{pmatrix}$$

$\begin{matrix} m \times n \\ \uparrow \\ m \text{ rows} \end{matrix}$ $\begin{matrix} \text{columns} \\ \uparrow \end{matrix}$

$$\text{If } \vec{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix} = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \in \mathbb{R}^n \quad (\text{written as a column vector})$$

$$A\vec{b} = \begin{pmatrix} -\vec{a}_1 \\ \vdots \\ -\vec{a}_m \end{pmatrix} \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix} = \begin{pmatrix} a_{11}b_1 + \cdots + a_{1n}b_n \\ \vdots \\ a_{m1}b_1 + \cdots + a_{mn}b_n \end{pmatrix}$$

$$= \begin{pmatrix} \vec{a}_1 \cdot \vec{b} \\ \vdots \\ \vec{a}_m \cdot \vec{b} \end{pmatrix} \in \mathbb{R}^m$$

$m \times 1$

Similarly,

$$(A \vec{b}) = \begin{pmatrix} 1 & \cdots & 1 \\ b_1 & \cdots & b_k \end{pmatrix} = (\vec{a} \cdot \vec{b}_1, \dots, \vec{a} \cdot \vec{b}_k)$$

$(m \times n) \quad n \times k \quad 1 \times k$

$$\underset{m \times n}{A} \underset{n \times k}{B} = A \begin{pmatrix} 1 & \cdots & 1 \\ \vec{b}_1 & \cdots & \vec{b}_k \end{pmatrix} = \begin{pmatrix} 1 & \cdots & 1 \\ A\vec{b}_1 & \cdots & A\vec{b}_k \end{pmatrix}$$

$$= \begin{pmatrix} -\vec{a}_1 \\ \vdots \\ -\vec{a}_m \end{pmatrix} B = \begin{pmatrix} -\vec{a}_1 B \\ \vdots \\ -\vec{a}_m B \end{pmatrix}$$

$$\text{eq } A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \quad B = \begin{pmatrix} 5 & 6 & 7 \\ 8 & 9 & 10 \end{pmatrix}$$

$$AB = \begin{pmatrix} 21 & 24 & 27 \\ 47 & 54 & 61 \end{pmatrix}$$

$$A \begin{pmatrix} 5 \\ 8 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} 5 \\ 8 \end{pmatrix} = \begin{pmatrix} 5+16 \\ 15+32 \end{pmatrix} = \begin{pmatrix} 21 \\ 47 \end{pmatrix}$$

$$A \begin{pmatrix} 6 \\ 7 \end{pmatrix} = \begin{pmatrix} 24 \\ 54 \end{pmatrix} \quad A \begin{pmatrix} 7 \\ 10 \end{pmatrix} = \begin{pmatrix} 27 \\ 61 \end{pmatrix}$$

$$(1 \ 2) B = (21 \ 24 \ 27)$$

$$(3 \ 4) B = (47 \ 54 \ 61)$$

Vector-valued functions

$$\vec{f}: \Omega (\subseteq \mathbb{R}^n) \rightarrow \mathbb{R}^m$$

$$\vec{f}(\vec{x}) = \begin{pmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{pmatrix} \quad \text{vector-valued.}$$

Each $f_i(\vec{x})$ is a multi-variable function.

Suppose $\frac{\partial f_i}{\partial x_j}(\vec{a})$ for each i, j .

For each $1 \leq i \leq m$

$$f_i(\vec{x}) = f_i(\vec{a}) + \nabla f_i(\vec{a}) \cdot (\vec{x} - \vec{a}) + \epsilon_i(\vec{x})$$

$$1 \times 1 \quad 1 \times 1 \quad (x_1 \quad n \times 1) \quad 1 \times 1$$

regard as regarded as
 $\begin{matrix} n \\ \text{row vector} \end{matrix}$ $\begin{matrix} n \\ \text{column} \\ \text{vector} \end{matrix}$

$$\begin{pmatrix} f_1(\vec{x}) \\ \vdots \\ f_m(\vec{x}) \end{pmatrix} = \begin{pmatrix} f_i(\vec{a}) \\ \vdots \\ f_m(\vec{a}) \end{pmatrix} + \begin{pmatrix} -\nabla f_1(\vec{a}) - \\ \vdots \\ -\nabla f_m(\vec{a}) - \end{pmatrix} \begin{pmatrix} x_1 - a_1 \\ \vdots \\ x_n - a_n \end{pmatrix}$$

$$m \times n \text{ matrix} \quad + \begin{pmatrix} \epsilon_1(\vec{x}) \\ \vdots \\ \epsilon_m(\vec{x}) \end{pmatrix}$$

Def ① The Jacobian matrix of $\vec{f}: \mathbb{R}^n \rightarrow \mathbb{R}^m$ at \vec{a} is $D\vec{f}(\vec{a}) = \begin{pmatrix} -\nabla f_1(\vec{a}) - \\ \vdots \\ -\nabla f_m(\vec{a}) - \end{pmatrix}$ ($m \times n$ matrix)

② The linearization of \vec{f} at \vec{a} to be

$$\Sigma(\vec{x}) = \vec{f}(\vec{a}) + D\vec{f}(\vec{a}) \cdot (\vec{x} - \vec{a})$$

③ \vec{f} is differentiable at \vec{a} if Error term

$$\vec{\xi}(\vec{x}) = \vec{f}(\vec{x}) - \vec{L}(\vec{x}) \quad \text{satisfies}$$

$$\lim_{\vec{x} \rightarrow \vec{a}} \frac{\|\vec{\xi}(\vec{x})\|}{\|\vec{x} - \vec{a}\|} = 0.$$

Rmk ① $[D\vec{f}(\vec{a})]_{ij} = \frac{\partial f^i}{\partial x_j}(\vec{a})$

$$② \vec{f}(\vec{x}) = \vec{f}(\vec{a}) + D\vec{f}(\vec{a}) \cdot (\vec{x} - \vec{a}) + \vec{\xi}(\vec{x})$$

$m \times 1$ $m \times 1$ $\underbrace{m \times n}_{m \times 1}$ $n \times 1$ $m \times 1$

③ If f is real-valued (i.e. $m=1$)

$$D\vec{f}(\vec{a}) = Df(\vec{a})$$

④ $\|\vec{\xi}(\vec{x})\|$: length in \mathbb{R}^m

$\|\vec{x} - \vec{a}\|$: length in \mathbb{R}^n .

$$⑤ \lim_{\vec{x} \rightarrow \vec{a}} \frac{\|\vec{\xi}(\vec{x})\|}{\|\vec{x} - \vec{a}\|} = 0 \Leftrightarrow \lim_{\vec{x} \rightarrow \vec{a}} \frac{\xi_i(\vec{x})}{\|\vec{x} - \vec{a}\|} = 0 \quad \text{for } i=1,\dots,m$$

$\therefore \vec{f}$ is differentiable at $\vec{a} \Leftrightarrow f_i$ is differentiable

at a_i for $i=1,\dots,m$

$$\text{Approximation} \quad \tilde{f}(\vec{x}) \approx L(\vec{x}) = \tilde{f}(\vec{a}) + D\tilde{f}(\vec{a}) \cdot (\vec{x} - \vec{a})$$

$$\Rightarrow \underbrace{\tilde{f}(\vec{x}) - \tilde{f}(\vec{a})}_{\Delta \tilde{f}} \approx \underbrace{D\tilde{f}(\vec{a}) \cdot (\vec{x} - \vec{a})}_{\Delta \vec{x}}$$

$\Delta \tilde{f}$ = change of \tilde{f}

$\Delta \vec{x}$ = change of \vec{x} .

One can consider $D\tilde{f}(\vec{a})$ as a linear map

$$D\tilde{f}(\vec{a}) : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$\Delta \vec{x} \mapsto D\tilde{f}(\vec{a}) \Delta \vec{x}.$$

e.g. $\tilde{f}(x,y) = \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} = \begin{pmatrix} (y+1) \ln x \\ x^2 - \sin y + 1 \end{pmatrix} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$

① What is $D\tilde{f}(1,0)$?

$$Df_1 = \begin{pmatrix} \frac{y+1}{x} & \ln x \end{pmatrix}, Df_2 = \begin{pmatrix} 2x & -\cos y \end{pmatrix}$$

$$\therefore D\tilde{f}(x,y) = \begin{pmatrix} \frac{y+1}{x} & \ln x \\ 2x & -\cos y \end{pmatrix}$$

$$D\tilde{f}(1,0) = \begin{pmatrix} 1 & 0 \\ 2 & -1 \end{pmatrix}$$

② Linearization of \tilde{f} at $(1,0)$ $\vec{x} = \vec{\xi}$

$$\begin{aligned}\tilde{L}(x,y) &= \tilde{f}(1,0) + D\tilde{f}(1,0) \cdot \begin{pmatrix} x \\ y \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 2 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ 2 & -1 \end{pmatrix} \begin{pmatrix} x-1 \\ y \end{pmatrix}\end{aligned}$$

③ Approximation of $\tilde{f}(0.9, 0.1)$

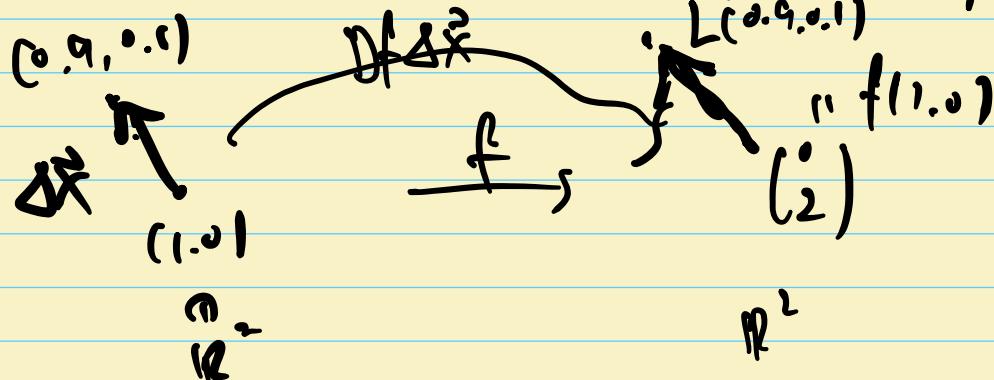
$$\tilde{f}(0.9, 0.1) \approx \tilde{L}(0.9, 0.1)$$

$$= \begin{pmatrix} 0 \\ 2 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ 2 & -1 \end{pmatrix} \left| \begin{pmatrix} 0.9-1 \\ 0.1 \end{pmatrix} \right.$$

$$\underbrace{\quad}_{D\tilde{f}(\vec{\alpha})} \underbrace{\Delta \vec{x}}_{\begin{pmatrix} 0.9-1 \\ 0.1 \end{pmatrix}}$$

$$= \begin{pmatrix} 0 \\ 2 \end{pmatrix} + \begin{pmatrix} -0.1 \\ -0.2 -0.1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ 2 \end{pmatrix} + \begin{pmatrix} -0.1 \\ -0.3 \end{pmatrix} = \begin{pmatrix} -0.1 \\ 1.7 \end{pmatrix}$$



Rank Total differential can be written in matrix form

$$\tilde{f}: \Omega (\subseteq \mathbb{R}^n) \rightarrow \mathbb{R}^m$$

$$\tilde{f} = \begin{pmatrix} f_1 \\ \vdots \\ f_m \end{pmatrix}$$

$$d\tilde{f} = \begin{pmatrix} df_1 \\ \vdots \\ df_m \end{pmatrix}$$

$$= \left(\sum_{j=1}^n \frac{\partial f_i}{\partial x_j} dx_j \right)$$
$$\vdots$$
$$\sum_{j=1}^n \frac{\partial f_m}{\partial x_j} dx_j$$

$$= \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \dots & \frac{\partial f_m}{\partial x_n} \end{pmatrix} \begin{pmatrix} dx_1 \\ \vdots \\ dx_n \end{pmatrix}$$

$$\underbrace{\qquad}_{D\tilde{f}}$$

$$= D\tilde{f}(\bar{a}) \cdot d\tilde{x}$$

Chain rule: differential of composition of two functions.

Recall in one-variable

$$\begin{array}{ccc} \mathbb{R} \xrightarrow{f} \mathbb{R} \xrightarrow{g} \mathbb{R} & (g \circ f)'(x) \\ \xrightarrow{\text{g} \circ f} & = g'(f(x)) \cdot f'(x) \end{array}$$

$$\begin{array}{ll} w = g(u) & \frac{dw}{dx} = \frac{du}{du} \cdot \frac{du}{dx} \\ u = f(x) & \end{array}$$

$$\begin{array}{ll} \text{eg } w = 2u+1 & \frac{dw}{dx} = \frac{du}{du} \cdot \frac{du}{df} \\ u = x^2 & \end{array}$$

$$= 2 \cdot 2x = 4x.$$

$$\begin{array}{ccc} \mathbb{R}^n \xrightarrow{f} \mathbb{R} \xrightarrow{g} \mathbb{R} & \nabla(g \circ f)(\vec{a}) \\ \xrightarrow{g \circ f} & = g'(f(\vec{a})) \cdot \nabla f(\vec{a}) \end{array}$$

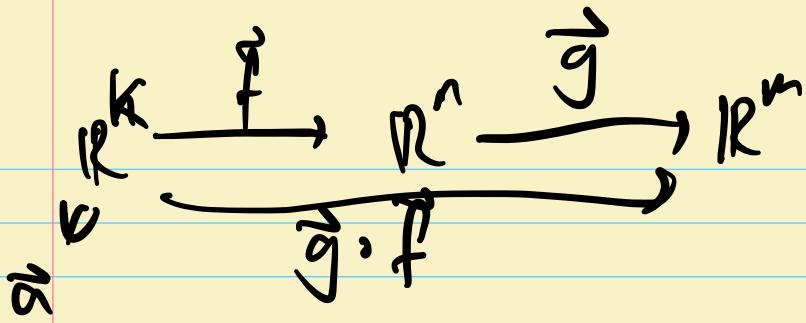
For vector-valued multi-variable functions,

Thm (Chain rule)

Let $\vec{f}: \mathbb{R}_1 (\subseteq \mathbb{R}^k) \rightarrow \mathbb{R}^n$, $\vec{g}: \mathbb{R}_2 (\subseteq \mathbb{R}^m) \rightarrow \mathbb{R}^m$

Suppose \vec{f} is differentiable at $\vec{a} \in \mathbb{R}_1$.

... \vec{g} ... - $\vec{g} = \vec{f}(\vec{a}) \in \mathbb{R}_2$



$\vec{g} \circ \vec{f}$ is differentiable at \vec{a} and

$$D(\vec{g} \circ \vec{f})(\vec{a}) = D\vec{g}(\vec{f}(\vec{a})) \cdot D\vec{f}(\vec{a})$$

$m \times k \quad m \times n \quad n \times k$

Rank For simplicity, we omit \rightarrow for vectors from now on. $\vec{f} = f$, $\vec{x} = x$ etc.

eg $f: \mathbb{R} \rightarrow \mathbb{R}^2 \quad f(\theta) = \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix}$

$$g: \mathbb{R}^2 \rightarrow \mathbb{R}^2 \quad g(u, v) = \begin{pmatrix} 2uv \\ u^2 - v^2 \end{pmatrix}$$

$$D(g \circ f)(\theta) \quad g \circ f: \mathbb{R} \rightarrow \mathbb{R}^2$$

(Sol) direct computation.

$$(g \circ f)(\theta) = g(\cos \theta, \sin \theta) = \begin{pmatrix} 2 \cos \theta \sin \theta \\ \cos^2 \theta - \sin^2 \theta \end{pmatrix}$$

$$= \begin{pmatrix} \sin 2\theta \\ \cos 2\theta \end{pmatrix}$$

$$\therefore D(g \circ f)(\theta) = \begin{pmatrix} 2 \cos 2\theta \\ -2 \sin 2\theta \end{pmatrix}$$

apply chain rule,

$$Df(\theta) = \begin{pmatrix} f_1'(\theta) \\ f_2'(\theta) \end{pmatrix} = \begin{pmatrix} -\sin \theta \\ \cos \theta \end{pmatrix}$$

$$Dg(u, v) = \begin{pmatrix} 2v & 2u \\ 2u & -2v \end{pmatrix}$$

$$D(g \circ f)(\theta) = Dg(f(\theta)) \cdot I f(\theta)$$

$$= \begin{pmatrix} 2 \sin \theta & 2 \cos \theta \\ 2 \cos \theta & -2 \sin \theta \end{pmatrix} \begin{pmatrix} -\sin \theta \\ \cos \theta \end{pmatrix}$$

$$= \begin{pmatrix} -2 \sin^2 \theta + 2 \cos^2 \theta \\ -2 \cos \theta \sin \theta - 2 \sin \theta \cos \theta \end{pmatrix}$$

$$= \begin{pmatrix} 2 \cos 2\theta \\ -2 \sin 2\theta \end{pmatrix}$$

$$\text{eq} \quad f(x,y) = \begin{pmatrix} x^2 \\ 3xy \\ x+y^2 \end{pmatrix} \quad . \quad g(u,v,w) = \frac{uw}{v}$$

$$: \mathbb{R}^2 \rightarrow \mathbb{R}^3 \quad : \mathbb{R}^3 \rightarrow \mathbb{R}$$

$$\frac{\partial(g \cdot f)}{\partial x}(1,1) ?$$

$$(\text{Sol}) \quad Dg = \nabla g = \left(\frac{w}{v}, -\frac{uw}{v^2}, \frac{u}{v} \right)$$

$$Df = \begin{pmatrix} 2x & 0 \\ 3y & 3x \\ 1 & 2y \end{pmatrix} \quad f(1,1) = \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}$$

$$\begin{aligned} \therefore D(g \cdot f)(1,1) &= Dg(f(1,1)) \cdot Df(1,1) \\ &= \left(\frac{2}{3}, -\frac{2}{9}, \frac{1}{2} \right) \cdot \begin{pmatrix} 2 & 0 \\ 3 & 3 \\ 1 & 2 \end{pmatrix} \end{aligned}$$

$$= (1, 0)$$

↓

$$\frac{\partial(g \cdot f)}{\partial x}(1,1) \quad \therefore \frac{\partial(g \cdot f)}{\partial x}(1,1) = 1$$

Rank In classical notation, we write $\frac{\partial g}{\partial x}$ for $\frac{\partial(g \circ f)}{\partial x}$. Using classical notation,

chain rule can be written as

(in the previous example)

$$\left. \begin{aligned} \frac{\partial g}{\partial x} &= \frac{\partial g}{\partial u} \cdot \frac{\partial u}{\partial x} + \frac{\partial g}{\partial v} \cdot \frac{\partial v}{\partial x} + \frac{\partial g}{\partial w} \cdot \frac{\partial w}{\partial x} \\ \frac{\partial g}{\partial y} &= \frac{\partial g}{\partial u} \cdot \frac{\partial u}{\partial y} + \frac{\partial g}{\partial v} \cdot \frac{\partial v}{\partial y} + \frac{\partial g}{\partial w} \cdot \frac{\partial w}{\partial y} \end{aligned} \right\}$$

$$\begin{matrix} \mathbb{R}^2 & \xrightarrow{f} & \mathbb{R}^3 & \xrightarrow{g} & \mathbb{R} \\ x & & u & & \\ y & & v & & \\ & & w & & \end{matrix}$$

eg $w(x, y, z) = \sqrt{x^2 + y^2 + z^2}$

$$x = 3e^t \sin s$$

$$y = 3e^t \cos s$$

$$z = 4et$$

What is $\frac{\partial w}{\partial t}$ at $s=1 \Rightarrow ?$

$$(c) (1) \frac{\partial w}{\partial t} = \frac{\partial w}{\partial x} \cdot \frac{\partial x}{\partial t} + \frac{\partial w}{\partial y} \cdot \frac{\partial y}{\partial t} + \frac{\partial w}{\partial z} \cdot \frac{\partial z}{\partial t}$$

$$= \frac{x}{\sqrt{x^2+y^2+z^2}} \cdot 3e^t \sin s + \frac{y}{\sqrt{x^2+y^2+z^2}} \cdot (+3e^t \cos s) \\ + \frac{z}{\sqrt{x^2+y^2+z^2}} \cdot 4e^t$$

at $s=t=0$, $(x,y,z) = (0, 3, 4)$

$$\Rightarrow = \frac{0}{5} \cdot 0 + \frac{3}{5} \cdot 3 + \frac{4}{5} \cdot 4$$

$$= \frac{9+16}{5} = 5.$$

eg P is moving with position at time t

$$\text{by } \begin{cases} x(t) = t^2 + 1 \\ y(t) = 2t^2 \end{cases} \quad \text{where altitude is}$$

$$\text{given by } H(x,y) = x^2 - y^2 + 100$$

- ① Is P going up/down at $t=1$?
- ② Which direction should P move at $t=1$ to go down most quickly?

(sol) ① we need to find $\frac{\partial H}{\partial t} \Big|_{t=1}$.

$$\frac{\partial H}{\partial t} = \frac{\partial H}{\partial x} \cdot \frac{\partial x}{\partial t} + \frac{\partial H}{\partial y} \cdot \frac{\partial y}{\partial t}$$

$$x = t^3 + 1$$

$$y = 2t^2$$

$$= (2x) \cdot (3t^2) + (-2y) \cdot (4t)$$

$$= 2(t^3 + 1)(3t^2) + (-2 \cdot 2t^2)(4t)$$

$$= 6t^5 + 6t^2 - 16t^3$$

$$\therefore \frac{\partial H}{\partial t} \Big|_{t=1} = 6t^6 - 16t^3 = -4 < 0.$$

∴ P is moving down at $t=1$.

② Recall that f decrease most rapidly in the direction of $-\nabla f$.

$$\text{At } t=1, (x,y) = (2,2)$$

$$\nabla H = (2x, -2y)$$

$$\nabla H(2,2) = (4, -4)$$

\therefore At $t=1$, H decreases most rapidly in the direction $-\nabla H(2,2) = (-4,4)$

"

$$(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$$



P should move toward $(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$

Rank

$$\frac{\partial H}{\partial t} = \frac{\partial H}{\partial x} \cdot \frac{\partial x}{\partial t} + \frac{\partial H}{\partial y} \cdot \frac{\partial y}{\partial t}$$

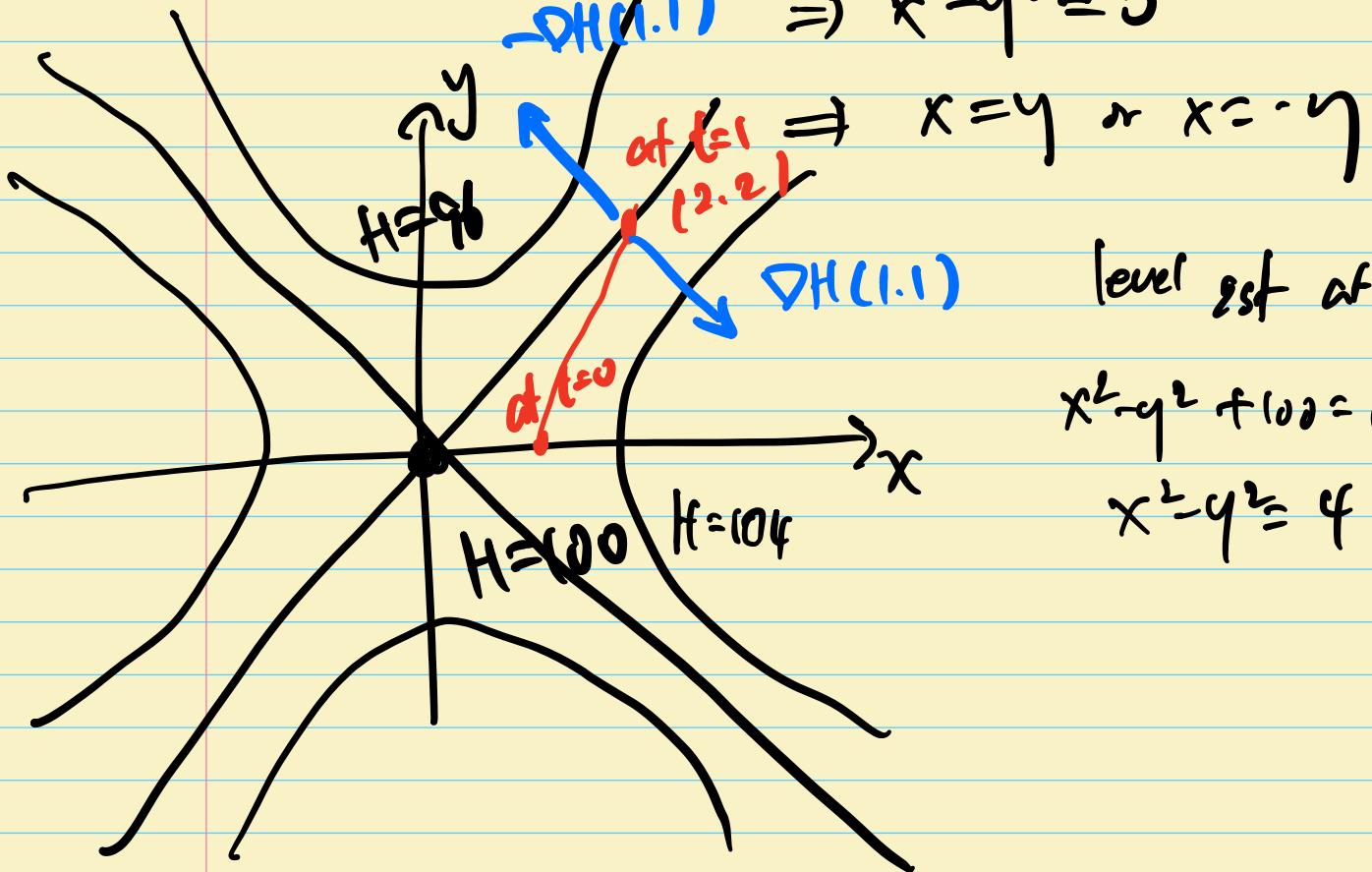
↑ ↑ ↑ ↑
 slope velocity slope velocity
 in x direction in x direction in y direction in y direction

$$= \nabla H \cdot \left(\underbrace{\frac{\partial x}{\partial t}, \frac{\partial y}{\partial t}}_{\text{velocity}} \right)$$

Picture level set of $H(x,y) = x^2 - y^2 + 100$.

$$\text{level set at } 100 \Rightarrow x^2 - y^2 + 100 = 100$$

$$-\nabla H(1,1) \Rightarrow x^2 - y^2 = 0$$



level set of $H(x,y) = x^2 - y^2 + 104$

$$x^2 - y^2 + 104 = 104$$

$$x^2 - y^2 = 0$$

