

CSCI 567 - LINEAR ALGEBRA DISCUSSION II

① DERIVATIVES, GRADIENTS AND JACOBIAN

$$f: \mathbb{R}^1 \rightarrow \mathbb{R}^1$$

e.g. $f(x)$

$$\nabla f = \frac{df}{dx} = \underbrace{\frac{\partial f}{\partial x}}_{1 \times 1} = f_x$$

derivative

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}^1$$

e.g. $f(x, y)$

$$\nabla f = \left[\begin{array}{cc} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{array} \right]_{1 \times 2}$$

gradient (row vector)

$$\vec{f}: \mathbb{R}^1 \rightarrow \mathbb{R}^2$$

e.g. $\begin{bmatrix} a(x) \\ b(x) \end{bmatrix}$

$$\nabla f = \begin{bmatrix} \frac{\partial a}{\partial x} \\ \frac{\partial b}{\partial x} \end{bmatrix}_{2 \times 1} = \begin{bmatrix} a_x \\ b_x \end{bmatrix}$$

Jacobian

$$\vec{f}: \mathbb{R}^2 \rightarrow \mathbb{R}^3$$

e.g. $\begin{bmatrix} a(x, y) \\ b(x, y) \\ c(x, y) \end{bmatrix}$

$$\nabla f = \begin{bmatrix} \frac{\partial a}{\partial x} & \frac{\partial a}{\partial y} \\ \frac{\partial b}{\partial x} & \frac{\partial b}{\partial y} \\ \frac{\partial c}{\partial x} & \frac{\partial c}{\partial y} \end{bmatrix}_{3 \times 2}$$

Jacobian

functions
are
always
column
vectors.

* The usual derivative is a special case of gradients which itself is a special case of jacobians.

* We always represent gradients as row vectors.
This is the numerator layout.

Exercise: Find the dimension of the jacobian/gradient

- i) $f: \mathbb{R} \rightarrow \mathbb{R}$ 1×1
- ii) $f: \mathbb{R}^5 \rightarrow \mathbb{R}$ 1×5
- iii) $\vec{f}: \mathbb{R} \rightarrow \mathbb{R}^3$ 3×1
- iv) $\vec{f}: \mathbb{R}^7 \rightarrow \mathbb{R}^{10}$ 10×7

② CHAIN RULE

a) $f: \mathbb{R} \rightarrow \mathbb{R}$
 $g: \mathbb{R} \rightarrow \mathbb{R}$
 $h = f \circ g : \mathbb{R} \rightarrow \mathbb{R}$

} Single variable

$$h(x) = f(g(x))$$

$$\nabla h = \frac{dh}{dx} = \frac{df}{dg} \cdot \frac{dg}{dx} = f'(g) g'$$

$$\nabla h = \nabla f \cdot \nabla_x g$$

(b) $f: \mathbb{R} \rightarrow \mathbb{R} \longrightarrow 1 \times 1$

$g: \mathbb{R}^2 \rightarrow \mathbb{R} \xrightarrow{\text{e.g. } g(x,y)} 1 \times 2$

$h: f \circ g: \mathbb{R}^2 \rightarrow \mathbb{R} \longrightarrow$

$h(x,y) = f(g(x,y))$

$\nabla h = [h_x \ h_y] = \left[\frac{\partial h}{\partial x} \quad \frac{\partial h}{\partial y} \right]$

$$\frac{\partial h}{\partial x} = \underline{\underline{\frac{\partial f}{\partial g} \cdot \frac{\partial g}{\partial x}}} \quad \text{and} \quad \frac{\partial h}{\partial y} = \underline{\underline{\frac{\partial f}{\partial g} \cdot \frac{\partial g}{\partial y}}}$$

$$\nabla h = \left[\begin{array}{c|c} \frac{\partial f}{\partial g} & \frac{\partial g}{\partial x} \\ \hline \frac{\partial f}{\partial g} & \frac{\partial g}{\partial y} \end{array} \right]_{1 \times 2} = \left[\begin{array}{c|c} \frac{\partial f}{\partial g} & \end{array} \right]_{1 \times 1} \left[\begin{array}{c|c} \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \end{array} \right]_{1 \times 2}$$

$$\rightarrow \nabla h = \nabla_g f \cdot \nabla_{(x,y)} g$$

$$h' = \overline{f'(g)} \ g'$$

(c) $\vec{f}: \mathbb{R} \rightarrow \mathbb{R}^2 \quad \text{e.g. } \vec{f}(x) = \begin{bmatrix} a(x) \\ b(x) \end{bmatrix}$

$g: \mathbb{R} \rightarrow \mathbb{R} \quad \text{e.g. } g(x)$

$h = f \circ g: \mathbb{R} \rightarrow \mathbb{R}^2 \longrightarrow 2 \times 1$

$h(x) = f(g(x)) = \begin{bmatrix} a(g(x)) \\ b(g(x)) \end{bmatrix}$

$$\nabla h = \begin{bmatrix} \frac{\partial a}{\partial x} \\ \frac{\partial b}{\partial x} \end{bmatrix}_{2 \times 1} = \begin{bmatrix} \frac{\partial a}{\partial g} & \frac{\partial a}{\partial x} \\ \frac{\partial b}{\partial g} & \frac{\partial b}{\partial x} \end{bmatrix}_{2 \times 1}$$

$$= \begin{bmatrix} \frac{\partial a}{\partial g} \\ \frac{\partial b}{\partial g} \end{bmatrix}_{2 \times 1} \begin{bmatrix} \frac{\partial g}{\partial x} \end{bmatrix}_{1 \times 1}$$

$$\nabla h = \begin{bmatrix} \nabla_g f \\ \nabla_x g \end{bmatrix}_{2 \times 1} \cdot \begin{bmatrix} \nabla_x g \end{bmatrix}_{1 \times 1}$$

d) $f: \mathbb{R}^3 \rightarrow \mathbb{R} \rightarrow 1 \times 3$
 $\vec{g}: \mathbb{R}^2 \rightarrow \mathbb{R}^3 \rightarrow 3 \times 2$
 $h = f \circ g: \mathbb{R}^2 \rightarrow \mathbb{R} \rightarrow \underline{1 \times 2}$

$$f = f(p, q, r) \quad \vec{g} = g(x, y) = \begin{bmatrix} p(x, y) \\ q(x, y) \\ r(x, y) \end{bmatrix}$$

$$\frac{\partial h}{\partial x} = \frac{\partial f}{\partial p} \frac{\partial p}{\partial x} + \frac{\partial f}{\partial q} \frac{\partial q}{\partial x} + \frac{\partial f}{\partial r} \frac{\partial r}{\partial x}$$

$$\frac{\partial h}{\partial y} = \frac{\partial f}{\partial p} \frac{\partial p}{\partial y} + \frac{\partial f}{\partial q} \frac{\partial q}{\partial y} + \frac{\partial f}{\partial r} \frac{\partial r}{\partial y}$$

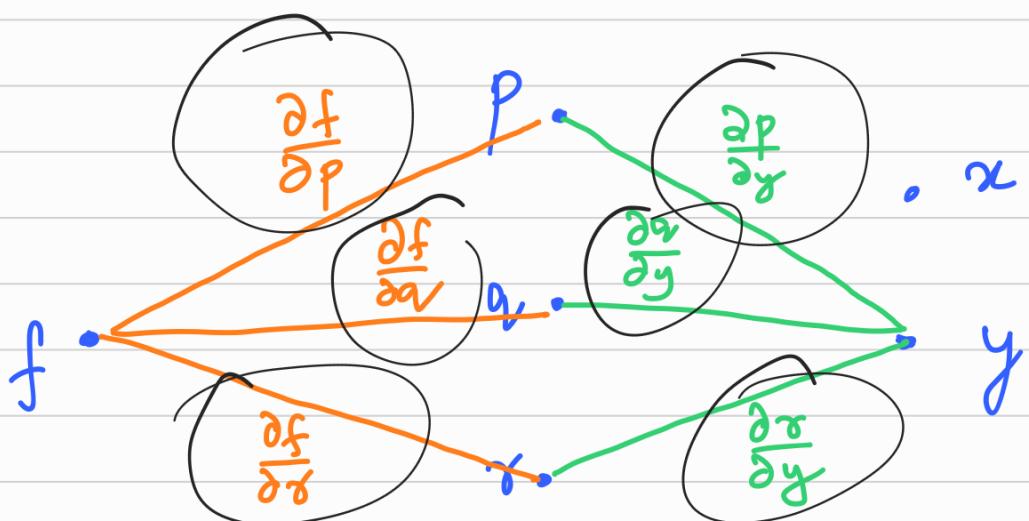
$$\nabla h = \begin{bmatrix} \frac{\partial h}{\partial x} \\ \frac{\partial h}{\partial y} \end{bmatrix}_{1 \times 2}$$

$$\begin{bmatrix} \frac{\partial f}{\partial p} \frac{\partial p}{\partial x} + \frac{\partial f}{\partial v} \frac{\partial v}{\partial x} & \frac{\partial f}{\partial p} \frac{\partial p}{\partial y} + \frac{\partial f}{\partial v} \frac{\partial v}{\partial y} + \frac{\partial f}{\partial x} \frac{\partial x}{\partial y} \end{bmatrix}_{1 \times 2}$$

$$= \begin{bmatrix} \frac{\partial f}{\partial p} & \frac{\partial f}{\partial v} & \frac{\partial f}{\partial x} \end{bmatrix}_{1 \times 3} \begin{bmatrix} \frac{\partial p}{\partial x} & \frac{\partial p}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \\ \frac{\partial x}{\partial x} & \frac{\partial x}{\partial y} \end{bmatrix}_{3 \times 2}$$

$$\nabla h = \nabla f(p, v) \cdot \nabla \vec{g}(u, y)$$

$1 \times 3 \quad 3 \times 2$



$$\frac{\partial h}{\partial y} = \left(\frac{\partial f}{\partial p} \frac{\partial p}{\partial y} \right) + \left(\frac{\partial f}{\partial q} \frac{\partial q}{\partial y} \right) + \left(\frac{\partial f}{\partial s} \frac{\partial s}{\partial y} \right)$$

(e)

$$\vec{f}: \mathbb{R}^2 \rightarrow \mathbb{R}^3, \rightarrow 3 \times 2$$

$$\vec{g}: \mathbb{R}^3 \rightarrow \mathbb{R}^2, \rightarrow 2 \times 1$$

$$\vec{h} = \vec{f} \circ \vec{g}: \mathbb{R}^3 \rightarrow \mathbb{R}^3, \rightarrow 3 \times 3$$

e.g. $\vec{f} = f(p, q) = \begin{bmatrix} a(p, q) \\ b(p, q) \\ c(p, q) \end{bmatrix}$

$$\vec{g} = g(x, y, z) = \begin{bmatrix} p(x, y, z) \\ q(x, y, z) \end{bmatrix}$$

$$h = \begin{bmatrix} a(p(x, y, z), q(x, y, z)) \\ b(p(x, y, z), q(x, y, z)) \\ c(p(x, y, z), q(x, y, z)) \end{bmatrix}$$

$$\nabla h(x, y, z) = \begin{bmatrix} \frac{\partial a}{\partial x} & \frac{\partial a}{\partial y} & \frac{\partial a}{\partial z} \\ \frac{\partial b}{\partial x} & \frac{\partial b}{\partial y} & \frac{\partial b}{\partial z} \\ \frac{\partial c}{\partial x} & \frac{\partial c}{\partial y} & \frac{\partial c}{\partial z} \end{bmatrix}_{3 \times 3}$$

$$= \begin{bmatrix} \frac{\partial a}{\partial p} \frac{\partial p}{\partial x} + \frac{\partial a}{\partial q} \frac{\partial q}{\partial x} & \frac{\partial a}{\partial p} \frac{\partial p}{\partial y} + \frac{\partial a}{\partial q} \frac{\partial q}{\partial y} & \frac{\partial a}{\partial p} \frac{\partial p}{\partial z} + \frac{\partial a}{\partial q} \frac{\partial q}{\partial z} \\ \frac{\partial b}{\partial p} \frac{\partial p}{\partial x} + \frac{\partial b}{\partial q} \frac{\partial q}{\partial x} & \frac{\partial b}{\partial p} \frac{\partial p}{\partial y} + \frac{\partial b}{\partial q} \frac{\partial q}{\partial y} & \frac{\partial b}{\partial p} \frac{\partial p}{\partial z} + \frac{\partial b}{\partial q} \frac{\partial q}{\partial z} \\ \frac{\partial c}{\partial p} \frac{\partial p}{\partial x} + \frac{\partial c}{\partial q} \frac{\partial q}{\partial x} & \frac{\partial c}{\partial p} \frac{\partial p}{\partial y} + \frac{\partial c}{\partial q} \frac{\partial q}{\partial y} & \frac{\partial c}{\partial p} \frac{\partial p}{\partial z} + \frac{\partial c}{\partial q} \frac{\partial q}{\partial z} \end{bmatrix}_{3 \times 3}$$

$$= \begin{bmatrix} \frac{\partial a}{\partial p} & \frac{\partial a}{\partial q} \\ \frac{\partial b}{\partial p} & \frac{\partial b}{\partial q} \\ \frac{\partial c}{\partial p} & \frac{\partial c}{\partial q} \end{bmatrix}_{3 \times 2} \begin{bmatrix} \frac{\partial p}{\partial x} & \frac{\partial p}{\partial y} & \frac{\partial p}{\partial z} \\ \frac{\partial q}{\partial x} & \frac{\partial q}{\partial y} & \frac{\partial q}{\partial z} \end{bmatrix}_{2 \times 3}$$

$$\Rightarrow \vec{\nabla} \vec{h}(x, y, z) = \vec{\nabla} \vec{f}(p, q) \cdot \vec{\nabla} \vec{g}(x, y, z)$$

3×3 3×2 2×3

5.5 Consider the following functions:

$$f_1(\mathbf{x}) = \sin(x_1) \cos(x_2), \quad \mathbf{x} \in \mathbb{R}^2 \quad \leftarrow 1 \times 2$$

$$f_2(\mathbf{x}, \mathbf{y}) = \mathbf{x}^\top \mathbf{y}, \quad \mathbf{x}, \mathbf{y} \in \mathbb{R}^n \quad 1 \times 2n$$

$$\underline{f_3(\mathbf{x}) = \mathbf{x}\mathbf{x}^\top}, \quad \mathbf{x} \in \mathbb{R}^n \quad (n \times n) \times \underline{n}$$

Draft (2024-01-15) of "Mathematics for Machine Learning". Feedback: <https://mml-book.com>.

Exercises

171

- a. What are the dimensions of $\frac{\partial f_i}{\partial \mathbf{x}}$?
- b. Compute the Jacobians.

(a)

$$f_1: \mathbb{R}^2 \rightarrow \mathbb{R}^1, \quad \nabla f_1 \in \mathbb{R}^{1 \times 2}$$

$$f_2: \mathbb{R}^{2n} \rightarrow \mathbb{R}^1, \quad \nabla f_2 \in \mathbb{R}^{1 \times 2n}$$

$$f_3: \mathbb{R}^n \rightarrow \mathbb{R}^{n \times n}, \quad \nabla f_3 \in \mathbb{R}^{(n \times n) \times n}$$

$\underbrace{\hspace{10em}}$
3D matrix OR

3rd order tensor

$$\textcircled{b} \quad f_1(x_1, x_2) = \sin(x_1) \cos(x_2)$$

$$\nabla f_1 = \left[\frac{\partial f_1}{\partial x_1} \quad \frac{\partial f_1}{\partial x_2} \right]$$

$$= \begin{bmatrix} \cos(x_1) \cos(x_2) & -\sin(x_1) \sin(x_2) \end{bmatrix}$$

$$f_2 = \mathbf{x}^T \mathbf{y} . \quad \text{Let } \mathbf{x}, \mathbf{y} \in \mathbb{R}^2$$

$$f_2(x_1, x_2, y_1, y_2) = [x_1, x_2] \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

$$= x_1 y_1 + x_2 y_2$$

$$\nabla f_2 = \left[\frac{\partial f_2}{\partial x_1} \quad \frac{\partial f_2}{\partial x_2} \quad \frac{\partial f_2}{\partial y_1} \quad \frac{\partial f_2}{\partial y_2} \right]$$

$$= \begin{bmatrix} y_1 & y_2 & x_1 & x_2 \end{bmatrix}$$

$$\therefore \text{in general, } \nabla f_2 = [\vec{y} : \vec{x}]$$

$$f_3 = \mathbf{x}\mathbf{x}^T, \text{ let } \mathbf{x} \in \mathbb{R}^2$$

$$f_3 = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{2 \times 1} \begin{bmatrix} x_1 & x_2 \end{bmatrix}$$

$$1 \times 2 = \begin{bmatrix} ? \end{bmatrix}_{2 \times 2}$$

$$= \begin{bmatrix} x_1^2 \\ x_2 x_1 \\ x_2^2 \end{bmatrix}_{2 \times 2} \quad \begin{bmatrix} x_1 & x_2 \\ x_2 & x_2^2 \end{bmatrix}_{2 \times 2}$$

$$\nabla f_3 = \begin{bmatrix} \frac{\partial x_1^2}{\partial x_2} & \frac{\partial(x_1 x_2)}{\partial x_2} \\ \frac{\partial(x_2 x_1)}{\partial x_2} & \frac{\partial x_2^2}{\partial x_2} \\ \frac{\partial x_1^2}{\partial x_1} & \frac{\partial(x_1 x_2)}{\partial x_1} \\ \frac{\partial(x_2 x_1)}{\partial x_1} & \frac{\partial x_2^2}{\partial x_1} \end{bmatrix}_{2 \times 2 \times 2}$$

$$= \begin{bmatrix} 0 & x_1 \\ x_1 & 2x_2 \end{bmatrix} \\ \begin{bmatrix} 2x_1 & x_2 \\ x_2 & 0 \end{bmatrix}$$

Can be generalized when $x \in \mathbb{R}^n$

(we will skip this)

RESULTS TO REMEMBER

① $f(\vec{x}) = \underline{\underline{a^T x}}$

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

$$x \in \mathbb{R}^n$$

$$f(x) = [a_1 \ a_2] \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \underbrace{a_1 u_1 + a_2 u_2}_{\text{gradient always a row vector}} \quad \begin{bmatrix} a_1 \ a_2 \end{bmatrix}$$

$$\nabla f(x) = \underline{\underline{a^T}}$$

gradient always a row vector

② $f(\vec{x}) = \underline{\underline{x^T a}}$

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

$$x \in \mathbb{R}^n$$

$$\nabla f(x) = \underline{\underline{a^T}} \quad a^T$$

③ $\vec{f}(\vec{x}) = Ax$

$$\nabla f = A$$

$$A \in \mathbb{R}^{m \times n}$$

$$x \in \mathbb{R}^n$$

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

Ex: $A = \begin{bmatrix} a & b \\ c & d \\ e & f \end{bmatrix}$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

2×1

3×2

$$f(x) = Ax = \begin{bmatrix} ax_1 + bx_2 \\ cx_1 + dx_2 \\ ex_1 + fx_2 \end{bmatrix} \quad 3 \times 1$$

$$\nabla f = \begin{bmatrix} \frac{\partial}{\partial x_1}(ax_1 + bx_2) & b \\ c & \frac{\partial}{\partial x_2}(cx_1 + dx_2) \\ e & f \end{bmatrix}$$

3×2

$$= A$$

$$\text{COROLLARY} \quad \bar{f}(\vec{x}) = \vec{x} = I\vec{x}$$

$$\text{then } \nabla f = I$$

$$x = I^n x$$

$$A = I$$

$$A^n \rightarrow A = I$$

$$\textcircled{4} \quad \vec{f}(x) = x^T A \text{ is a row vector}$$

But vector functions are always

column vectors by convention.

So we convert $\vec{f}(x)$ into a column vector first.

$$\vec{f}(x) = (x^T A)^T = A^T x$$

$$\nabla f = A^T$$



$$\cancel{x^T a} \rightarrow a^T$$

$$\cancel{x^T A x} = A^T$$

$$\textcircled{5} \quad f(x) = x^T A x \in \text{Quadratic Form}$$

$$\text{Let } g(x) = Ax, \quad \nabla g = A$$

$$f(x) = x^T g$$

$$\nabla f = g^T + x^T \nabla g$$

(product rule)

$$= (Ax)^T + x^T A \quad (fg)' = f'g + f.g' \checkmark$$

$$= x^T A^T + x^T A$$

$$\cancel{f'g + g'f} \times$$

$$= \underline{\underline{x^T (A^T + A)}} \leftarrow \text{row vector}$$

COROLLARY

$$\underline{\|x\|^2} = x^T x = x^T I x$$

$$\nabla \|x\|^2 = \underline{x^T (I + I^T)} = \underline{2x^T}$$

Q3 $f(\omega) = \frac{1}{2} \|X\omega - y\|_2^2 + w^T M \omega, \omega \in \mathbb{R}^d$

Find $\underline{w^*} = \arg \min f(\omega)$

Solⁿ Set $\nabla f = 0$ $f \in \mathbb{R}^d \rightarrow \mathbb{R}^1$
 $w \in \mathbb{R}^d$

$$\begin{aligned}\nabla f &= \underline{\nabla \|z\|^2} + \underline{\nabla (w^T M \omega)} \\ &= 2 \underline{z^T z} + \underline{w^T (M + M^T)} \\ &= 2 \underline{z^T X} + \underline{w^T (M + M^T)} \\ &= 2 (X\omega - y)^T x + w^T (M + M^T) \\ &= 2 \underline{(w^T x^T x - y^T x)} + \underline{w^T (M + M^T)}\end{aligned}$$

$$\nabla f = 0 \Rightarrow \underline{w^T (2x^T x + M + M^T)} = \underline{2y^T x}$$

$$\Rightarrow (2x^T x + M^T + M) w = 2x^T y$$

$$\Rightarrow (x^T x + \frac{M^T + M}{2}) w = x^T y$$

$$\Rightarrow w^* = \left(x^T x + \frac{M^T + M}{2} \right)^{-1} x^T y$$

$$\text{sol}^n = \frac{1}{2} \|x_w - y\|_2^2 + w^T M w$$

$$\nabla f = \nabla \frac{1}{2} \|x_w - y\|_2^2 + \nabla (w^T M w)$$

$$= \nabla \frac{1}{2} \|x_w - y\|_2^2 + w^T (M + M^T)$$

$$\|x_w - y\|_2^2 = \underline{(x_w - y)^T (x_w - y)}$$

$$= (w^T x^T - y^T) (x_w - y)$$

$$= w^T x^T x_w - w^T x^T y - y^T x_w + y^T y$$

$$\Rightarrow \nabla \frac{1}{2} \|x_w - y\|_2^2 = w^T (x^T x + x^T x) - y^T x - y^T x + 0$$

$$= 2w^T x^T x - 2y^T x$$

$$\therefore \nabla f = 2w^T x^T x - 2y^T x + w^T (M + M^T)$$

$$\Rightarrow \nabla f = 0 \Rightarrow 2w^T (x^T x + \frac{M + M^T}{2}) = 2y^T x$$

$$\Rightarrow \left(x^T x + \frac{M^T + M}{2} \right) w = x^T y$$

$$\Rightarrow w^* = \left(x^T x + \frac{M^T + M}{2} \right)^{-1} x^T y \quad \checkmark$$

COROLLARY 1: LINEAR REGRESSION

(ORDINARY LEAST SQUARES)

In linear regression, $\hat{y} = X\omega$ and

$$\text{error} = \|\hat{y} - y\|_2^2 = \|X\omega - y\|_2^2$$

Here $M = 0$

$M = 0$

then $f(\omega) = \|X\omega - y\|_2^2$ = error in linear regression

$$\omega^* = \left(X^T X + \frac{M + M^T}{2} \right)^{-1} X^T y$$

$$\boxed{\omega^* = (X^T X)^{-1} X^T y}$$

ω^* = Solution of best parameters for the linear regression model $\hat{y} = X\omega$.

Note : $X^T X \omega = X^T y$ is called the normal equation.

ω^* is a solution of the normal equation.

There could be multiple solutions based on whether $X^T X$ is invertible or not.

COROLLARY 2 : Regularization

If $M = \lambda I$

Then $f(w) = \|Xw - y\|_2^2 + \underline{w^T(\lambda I)w}$

$$= \|Xw - y\|_2^2 + \lambda w^T I w$$

$$= \|Xw - y\|_2^2 + \lambda \underline{w^T w}$$

$$= \|Xw - y\|_2^2 + \lambda \|w\|^2 \quad \text{Regularizer to control model complexity.}$$

Then $w^* = \left[X^T X + \frac{\lambda I + (\lambda I)^T}{2} \right]^{-1} X^T y$

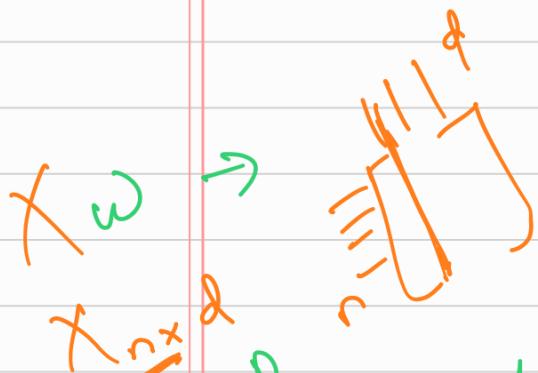
$$= \left[\underline{X^T X + \lambda I} \right]^{-1} \underline{X^T y}$$

GEOMETRIC INTERPRETATION OF OLS / LINEAR REGRESSION SOLUTION

$$\hat{y} = X\omega \leftarrow$$

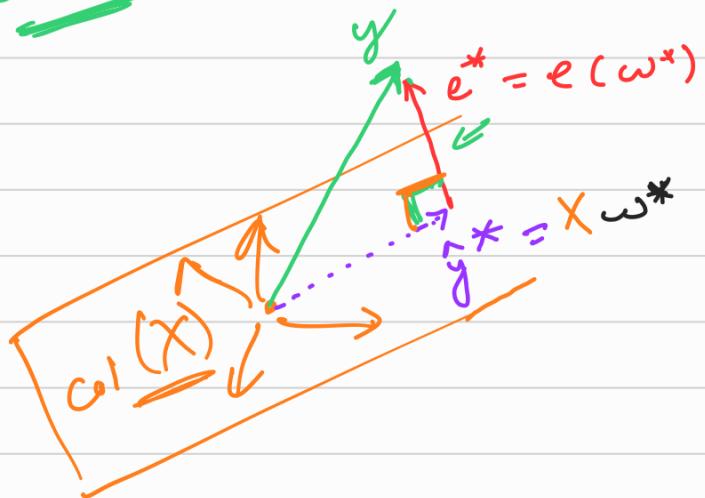
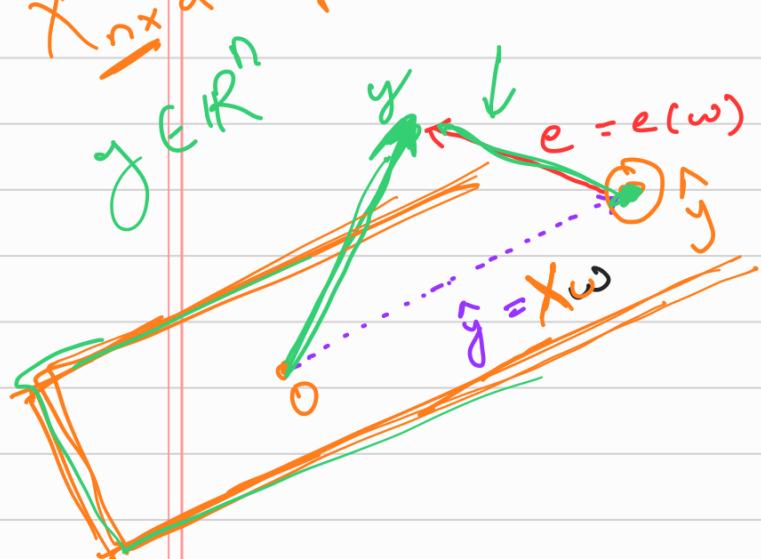
$$\vec{e} = y - \hat{y} = y - X\omega$$

$$\omega^* = \arg \min \|X\omega - y\|_2^2 = \|r\|^2$$



$$X = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

$$y = \begin{bmatrix} 10 \\ 11 \\ 12 \end{bmatrix}$$



Clearly, $e^* \perp \text{ColSpace}(X)$ $(X^T X)^{-1} X^T y$

$\Rightarrow y - \hat{y}^* \perp \text{each column of } X$

$$\Rightarrow X^T(y - \hat{y}^*) = 0$$

$$\Rightarrow X^T(y - X\omega^*) = 0$$

$$\Rightarrow X^T y - X^T X \omega^* = 0$$

$$\Rightarrow X^T X \omega^* = X^T y \quad (\text{Normal eqn})$$

$$\Rightarrow w^* = (X^\top X)^{-1} X^\top y$$