

1) Linear functions.

- a) Show that an inner product function, $f(x) = a^T x$, is linear.
- b) Show that any scalar-valued linear function $f(x)$ satisfying superposition can be expressed as an inner product function, say $f(x) = a^T x$. Explicitly state the elements of a in terms of f .

$$a). f(\alpha) = \alpha^T \mathbf{1}.$$

$$f(\mathbf{x} + \mathbf{y}) = \mathbf{a}^T (\mathbf{x} + \mathbf{y}) = \mathbf{a}^T \mathbf{x} + \mathbf{a}^T \mathbf{y} = f(\mathbf{x}) + f(\mathbf{y})$$

$$f(\mathbf{x} k) = \mathbf{a}^T (k \mathbf{x}) = k \cdot \mathbf{a}^T \mathbf{x} = k f(\mathbf{x}).$$

$\therefore f(\mathbf{x})$ is linear func.

$$b). \mathbf{x} \in \sum_{i=1}^n k_i e_i \Rightarrow f(\mathbf{x}) = f\left(\sum_{i=1}^n k_i e_i\right)$$

$$\Rightarrow f(\mathbf{x}) = \sum_{i=1}^n f(k_i e_i) = \sum_{i=1}^n (k_i f(e_i)) \quad (\because \text{linear. func.})$$

$$\mathbf{x} \in (k_1 e_1) \dots (k_n e_n) \Rightarrow f(\mathbf{x}) = \mathbf{a}^T \mathbf{x}.$$

2) Affine functions.

a) Show that an inner product function plus a shift, $f(x) = a^T x + b$, is affine.

b) Show that any scalar-valued affine function $f(x)$ satisfying the restricted superposition (superposition defined for linear combination with coefficients that sum to 1) can be expressed as an inner product function plus a shift, say $f(x) = a^T x + b$. Explicitly state the elements of a and b in terms of f .

$$a) \forall \alpha, \beta \in \mathbb{R}, \Rightarrow f(\alpha x + \beta y) = \alpha f(x) + \beta f(y).$$

$$\Rightarrow \alpha a^T x + \alpha b + \beta a^T y + \beta b. (\beta b + \alpha b = b).$$

$$\Rightarrow \alpha a^T x + \beta a^T y + b = a^T(\alpha x + \beta y) + b. \quad \square$$

$$b) f(x) = f\left(\sum \lambda_i e_i\right) = f\left(\sum \lambda_i e_i + \sum \left(\frac{1}{n} - \lambda_i\right)x_0\right).$$

$$= \sum \{f(\lambda_i e_i) - f(0)\} + \frac{1}{n} n x_0 f(0).$$

$$\therefore f(x) = a^T x + b.$$

$$b = f(0).$$

$$a = (f(e_i) - f(0)). \text{ n-vector.}$$

- 3) *Cauchy-Schwarz inequality.* Show that any two vectors $a, b \in \mathbb{R}^n$ satisfy the following.
 Also state the condition under which the inequality is tight.

$$|a^T b| \leq \|a\| \|b\|.$$

$\forall \lambda \in \mathbb{R}$. $\|(a-\lambda b)\|^2 \geq 0$.

$$\Rightarrow \lambda^2 - 2\lambda(a^T b) + \|b\|^2 \geq 0.$$

$$f(\lambda) \triangleq \|(a-\lambda b)\|^2 \Rightarrow \frac{d f(\lambda)}{d\lambda} = -2(a^T b) + 2\lambda\|b\|^2 = 0.$$

$$d\lambda = \frac{a^T b}{\|b\|^2} \quad f(\lambda) \leq f(a) \quad \forall \lambda \in \mathbb{R},$$

$$\Rightarrow \lambda^2 - 2 \cdot \frac{(a^T b)^2}{\|b\|^2} + \left(\frac{a^T b}{\|b\|^2} \right)^2 \cdot \|b\|^2 \geq 0.$$

$$\Rightarrow \|a\|^2 \|b\|^2 \geq (a^T b)^2 \quad \Leftrightarrow |a^T b| \geq \|a\| \|b\|.$$

where $a = kb$ or $b = ka$. for $\exists k \in \mathbb{R}$.

- 4) Angle between two vectors. Show that any two vectors $a, b \in \mathbb{R}^n$ satisfy the following, where θ is the angle between a and b . You may provide a proof for the two-dimension case, which easily generalizes to general n -dimension cases.

$$a^T b = \|a\| \|b\| \cos \theta.$$

Pf.) For 2D. $a \triangleq (a_1, a_2)$, $b \triangleq (b_1, b_2)$.

$$\|a\| = \sqrt{a_1^2 + a_2^2}, \quad \|b\| = \sqrt{b_1^2 + b_2^2},$$

$$\|a\| = (\|a\| (\cos \alpha, \sin \alpha)), \quad \|b\| = (\|b\| (\cos \beta, \sin \beta)).$$

where $\alpha = \angle(a, e_1)$, $\beta = \angle(b, e_1)$.

$$a^T b = (\|a\| \|b\| (\cos \alpha \cos \beta + \sin \alpha \sin \beta))$$

$$\cos(\alpha + \beta) = \cos \alpha \cos \beta + \sin \alpha \sin \beta. \text{ Let } \theta = \alpha + \beta.$$

$$\Rightarrow a^T b = (\|a\| \|b\| \cos \theta).$$

- 5) *Parallelogram.* Draw two different vectors u and v out from the origin. Complete two more sides to make a parallelogram with diagonals $w = u + v$ and $z = u - v$. Show that $\|w\|^2 + \|z\|^2 = 2\|u\|^2 + 2\|v\|^2$.

$$\begin{aligned} & \|u+v\|^2 + \|u-v\|^2 \\ \Rightarrow & \|u\|^2 + 2(u^Tv) + \|v\|^2 + \|u\|^2 - 2(u^Tv) + \|v\|^2 \\ \doteq & \|u\|^2 + \|v\|^2 \\ \therefore & \|w\|^2 + \|z\|^2 = \|u\|^2 + \|v\|^2. \end{aligned}$$

a) 2.3 Motion of a mass in response to applied force.

$$S(t_0) = S(0) + \int_0^{t_0} F(u) du = S(0) + \sum f_i$$

$$\Delta t \triangleq 1, h \triangleq S(t), F = [f_1, f_2, \dots, f_n]$$

$$\Rightarrow \bar{f} + h$$

$$S(t_0) = S(0) + \int_0^{t_0} S(u) du$$

$$= S + (0 \cdot S(0)) + \frac{1}{2} f_1 + \dots + \frac{1}{2} f_{t_0}$$

$$C \triangleq \left(\frac{1}{2}, \frac{1}{2}, \dots, \frac{1}{2} \right), d \triangleq S(0) + (0 \cdot S(0))$$

$$\bar{f} + d$$

b) 2.12 Price change to maximize profit.

$$\hat{P} = \beta_1 d_7 + P$$

(a). $\beta_3 < 0$: negative-rel.

$$(b). \Rightarrow \beta_7 > 0 \Rightarrow d_7 = 0.01 \Rightarrow 0.01 = \frac{P_7^{\text{new}} - P_7}{P_7} \Rightarrow P_7^{\text{new}} = 1.01 P_7$$

$$\Rightarrow \beta_3 < 0 \Rightarrow d_3 = -0.01 \Rightarrow -0.01 = \frac{P_3^{\text{new}} - P_3}{P_3} \Rightarrow P_3^{\text{new}} = 0.99 P_3$$

$$\therefore \hat{P} = (\beta_3 d_3 + \beta_7 d_7) + P = 0.01 (P_3 + P_7)$$

$$(c). |\beta_{3a}|, |\beta_{3b}| \geq \beta_7 \quad (7 \neq a, b)$$

$$\hat{P} = 0.01 (|\beta_{3a}| + 0.01 (|\beta_{3b}| + P) = 0.01 (|\beta_{3a}| + |\beta_{3b}|) + P$$

c) 3.12 Nearest point to a line.

$$\|P-a\|^2 = \|(1-\theta)(a+\theta(b-a))\|^2 = (a-\lambda + \theta(b-a))^2.$$

$$\frac{d}{d\theta} (\|a-\lambda + \theta(b-a)\|^2) = 2(a-\lambda)^T(b-a) + \theta^2 \|b-a\|^2$$

$$= 2\theta(b-a)^T + 2(a-\lambda)^T(b-a), \Rightarrow \theta = \frac{(b-a)^T(b-a)}{\|b-a\|^2} \quad (b \neq a).$$

$$P = a + \frac{(b-a)^T(b-a)}{\|b-a\|^2}(b-a)$$

$$(P-a)^T(a-b) = a + \frac{(b-a)^T(b-a)}{\|b-a\|^2}(b-a) - a^T(a-b)$$

$$= (a-\lambda)^T(b-a) - \frac{(b-a)^T(b-a)}{\|b-a\|^2} \cdot (b-a)^T(b-a)$$

$$= 0,$$