

ASEN 6020 - HW 2
Spring 2025
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Problem 1

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HW #2

Problem 1 →

$$f(x) = \sqrt{x \cdot x} \quad x \in \mathbb{R}^n$$

let $n=1 \rightarrow x \in \mathbb{R}^1$ (x is a scalar) $\rightarrow f(x) = \sqrt{x^2}$

$$df_+(x, u) = \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} [f(x + \lambda u) - f(x)] = \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} [\sqrt{(x + \lambda u)^2} - \sqrt{x^2}]$$

only when $x=0 \rightarrow \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} [\sqrt{(\lambda u)^2}] = \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} |\lambda u|$ (can cancel λ because approaching from +ve side)

$$df_+(x, u) = |u| \geq 0$$

let $n \geq 1 \rightarrow x \in \mathbb{R}^n \rightarrow f(x) = \sqrt{x \cdot x} = \sqrt{|x|^2} = |x|$ — where $|x|$ represents the magnitude of x and is ≥ 0 always

$$df_+(x, u) = \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} [|x + \lambda u| - |x|]$$

If $|x|$ is non-zero $\rightarrow df_+(x, u) = \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} [|x + \lambda u| - |x|]$

$$= \lim_{\lambda \rightarrow 0^+} \frac{|x + \lambda u|}{\lambda} - \lim_{\lambda \rightarrow 0^+} \frac{|x|}{\lambda}$$

$\lim_{\lambda \rightarrow 0^+} \frac{|x|}{\lambda}$ — diverges

If $|x|$ is 0 $\rightarrow df_+(x, u) = \lim_{\lambda \rightarrow 0^+} \frac{1}{\lambda} |\lambda u| = |u| \geq 0 = df_+(x, u)$

(can cancel λ because λ approaches 0 from +ve side)

$\therefore df_+(x, u) \geq 0$ for all u when $x = \vec{0}$ (for $x \in \mathbb{R}^n$), $x = \vec{0}$ is minimum

Problem 2

Problem 2 →

$$f(x) = \sum_{i,j=1}^n q_{ij} x_i x_j = \vec{x}^T Q \vec{x}$$

$$a) f(x) = \sum_{i,j=1}^n q_{ij} x_i x_j = \frac{1}{2} \cdot 2 \sum_{i,j=1}^n q_{ij} x_i x_j = \frac{1}{2} \left(\sum_{i,j=1}^n q_{ij} x_i x_j + \sum_{i,j=1}^n q_{ji} x_j x_i \right)$$

$$\sum_{i,j=1}^n q_{ij} x_i x_j \rightarrow \text{Replace } i,j \rightarrow \sum_{i,j=1}^n q_{ji} x_j x_i = \sum_{i,j=1}^n q_{ij} x_i x_j \quad (\text{both representations are the same})$$

$$\therefore f(x) = \frac{1}{2} \left(\sum_{i,j=1}^n q_{ij} x_i x_j + \sum_{i,j=1}^n q_{ji} x_j x_i \right) = \frac{1}{2} (\vec{x}^T Q \vec{x} + \vec{x}^T Q^T \vec{x}) = \boxed{\frac{1}{2} \vec{x}^T (Q + Q^T) \vec{x}}$$

Demonstrating further with a 2D example: (this can be expanded to

$$\text{Let } \vec{x} = [a, b]^T, Q = \begin{bmatrix} c & d \\ e & f \end{bmatrix}, a, b, c, d, e, f \in \mathbb{R} \quad n)$$

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$$f(x) = \vec{x}^T Q \vec{x} = [a \ b] \begin{bmatrix} c & d \\ e & f \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = [a \ b] \begin{bmatrix} ac + bd \\ ae + bf \end{bmatrix} = \boxed{a^2c + abd + abe + b^2f}$$

$$\frac{1}{2} \vec{x}^T (Q + Q^T) \vec{x} = \frac{1}{2} [a \ b] \begin{bmatrix} 2c & d+e \\ d+e & 2f \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \frac{1}{2} [a \ b] \begin{bmatrix} 2ac + bd + be \\ ad + ae + 2bf \end{bmatrix} = \frac{a^2c}{2} + \frac{abd + abe}{2} + \frac{abd + abe}{2} + \frac{b^2f}{2} = \boxed{a^2c + abd + abe + b^2f = f(x)}$$

$$\therefore f(x) = \frac{1}{2} \vec{x}^T (Q + Q^T) \vec{x}$$

$$b) \frac{\partial f}{\partial \mathbf{x}} = \frac{d}{dx} \left(\sum_{i,j=1}^n q_{ij} x_i x_j \right) = \sum_{i,j=1}^n q_{ij} (x_j + x_i) \quad \left| \quad \sum_{i,j=1}^n x_i + x_j \text{ is essentially adding the } 2 \text{ vectors} \right.$$

$$= 2 \sum_{i,j=1}^n q_{ij} x_i = 2 \sum_{i,j=1}^n x_i$$

$$\text{(from part a)} \rightarrow q_{ij} = \frac{1}{2} (q_{ji} + q_{ij})$$

$$= \sum_{i,j=1}^n (q_{ij} + q_{ji}) x_i = \boxed{(Q + Q^T) \vec{x} = \frac{\partial f}{\partial \mathbf{x}}}$$

Similar to part a, let's further verify this with a 2D example:

$$\text{let } \rightarrow \vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, Q = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, x_1, x_2, a, b, c, d \in \mathbb{R}$$

$$f(x) = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} ax_1 + bx_2 \\ cx_1 + dx_2 \end{bmatrix} = ax_1^2 + bx_1x_2 + cx_1x_2 + dx_2^2$$

$$\frac{\partial f}{\partial \mathbf{x}} = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 2ax_1 + bx_2 + cx_2 \\ bx_1 + cx_1 + 2dx_2 \end{bmatrix}$$

$$(Q + Q^T) \mathbf{x} = \begin{bmatrix} 2a & b+c \\ b+c & 2d \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2ax_1 + bx_2 + cx_2 \\ bx_1 + cx_1 + 2dx_2 \end{bmatrix}$$

$$\boxed{\frac{\partial f}{\partial \mathbf{x}} = (Q + Q^T) \vec{x}}$$

Problem 3

Problem 3 →

$$f(x) = -\sin(x) \cos(y)$$

∵ $\sin(x)$ & $\cos(y)$ are continuous, $f(x)$ is continuous

∵ $\sin(x)$ & $\cos(y)$ are smooth, $f(x)$ is smooth

To get local minimizer → $\frac{\partial f}{\partial x} \Big|_{\vec{x}^*} = 0$ & $\frac{\partial^2 f}{\partial x^2} \Big|_{\vec{x}^*} = f^*$ has to be ^{pos-} definite

$$\frac{\partial f}{\partial x} = \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix} = \begin{bmatrix} -\cos(x) \cos(y) \\ \sin(x) \sin(y) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

The following candidates emerge → $\vec{x}^* = \begin{bmatrix} 0 \\ \pi/2 \end{bmatrix}, \begin{bmatrix} 0 \\ 3\pi/2 \end{bmatrix}, \begin{bmatrix} \pi/2 \\ 0 \end{bmatrix}, \begin{bmatrix} \pi/2 \\ \pi \end{bmatrix}, \begin{bmatrix} \pi/2 \\ 3\pi/2 \end{bmatrix}, \begin{bmatrix} 3\pi/2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3\pi/2 \\ \pi \end{bmatrix}$

$$\frac{\partial^2 f}{\partial x^2} = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix} = \begin{bmatrix} \sin(x) \cos(y) & \cos(x) \sin(y) \\ \cos(x) \sin(y) & \sin(x) \cos(y) \end{bmatrix} = f^*$$

for the definite matrix → $|f^*| > 0$

$$|f^*| = \sin^2(x) \cos^2(y) - \cos^2(x) \sin^2(y) > 0$$

Testing the candidates →

$$\vec{x}_1^* = [0, \pi/2] \rightarrow |f^*| = -1, \quad \vec{x}_2^* = [\pi/2, 0] \Rightarrow |f^*| = 1$$

$$\vec{x}_3^* = [0, 3\pi/2] \rightarrow |f^*| = -1, \quad \vec{x}_4^* = [3\pi/2, 0] \Rightarrow |f^*| = 1$$

$$\vec{x}_5^* = [0, -\pi/2] \rightarrow |f^*| = -1, \quad \vec{x}_6^* = [-\pi/2, 0] \Rightarrow |f^*| = 1$$

$$\vec{x}_7^* = [0, -3\pi/2] \rightarrow |f^*| = -1, \quad \vec{x}_8^* = [-3\pi/2, 0] \Rightarrow |f^*| = 1$$

$\vec{x}_1^*, \vec{x}_3^*, \vec{x}_5^*, \vec{x}_7^*$ meet 2nd order necessary condition.

∴ $\vec{x}_1^* = \begin{bmatrix} 0 \\ \pi/2 \end{bmatrix}, \begin{bmatrix} 0 \\ 3\pi/2 \end{bmatrix}, \begin{bmatrix} 0 \\ -\pi/2 \end{bmatrix}, \begin{bmatrix} 0 \\ -3\pi/2 \end{bmatrix}$ are local minimizers

The domain of $f(x,y)$ can be reduced to be $x \rightarrow [0, \pi]$ and $y \rightarrow [-1, 1]$ and then $[\pi/2, 0]$ will be the unique global minimizer.

Problem 4

Problem 4 → $f(x) = a + b^T x + x^T Q x$

$f(x)$ is continuous and smooth. \therefore Necessary condition for $f(x)$ using 2nd order Taylor's series expansion

$$\frac{\partial f}{\partial x} \Big|_{\bar{x}^*} = \bar{b}, \quad \frac{\partial^2 f}{\partial x^2} = 0 + \bar{x} + (Q + Q^T) \bar{x} = \bar{b} + 2Q\bar{x} \quad (\because Q^T = Q)$$

$$\therefore \boxed{\bar{b} + 2Q\bar{x} = 0} \quad \text{--- (A)}$$

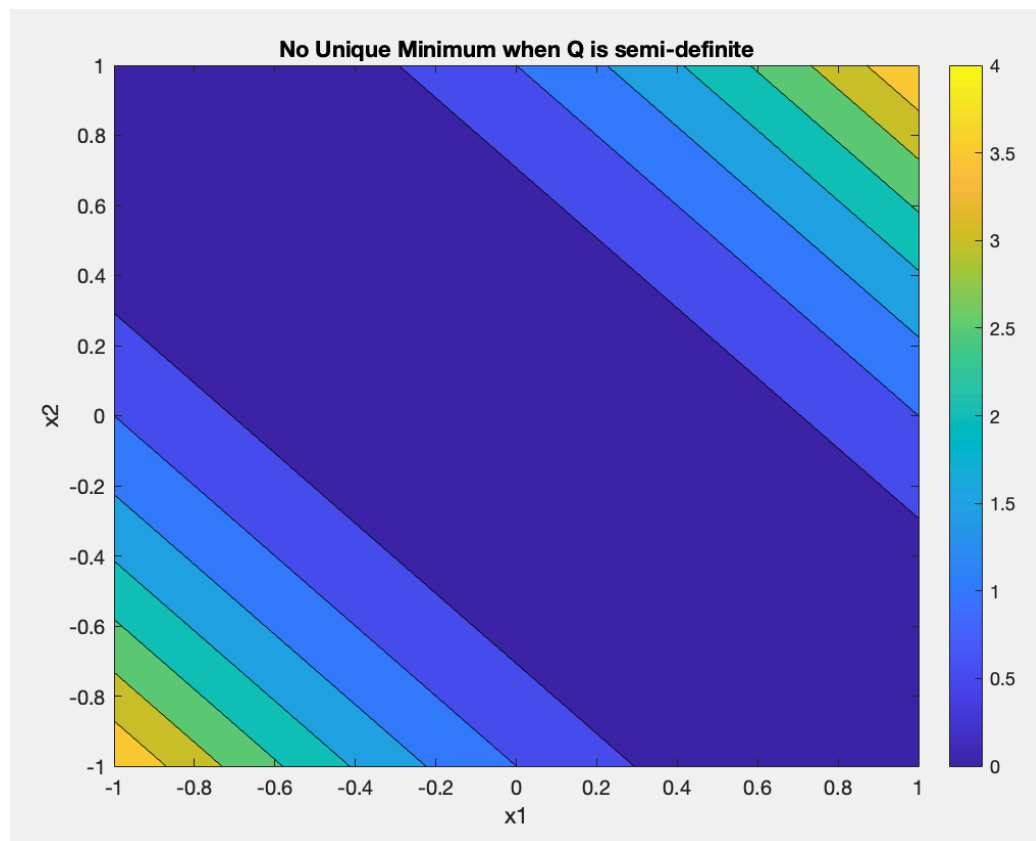
$$\frac{\partial^2 f}{\partial x^2} \Big|_{\bar{x}^*} \geq 0 \rightarrow \frac{\partial^2 f}{\partial x^2} = 2Q \geq 0 \rightarrow \boxed{Q \geq 0}$$

[B1] If Q is positive semi-definite, \bar{x}^* is a local minimizer

$Q > 0$ --- **[B2]** If Q is positive definite, \bar{x}^* is unique local minimizer

Sufficient condition $\rightarrow \bar{x}^*$ satisfies **[A]** & **[B2]**

If Q is $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$, then it is semi-definite and the cost function looks like this:



As seen above the plot has a lot of minimizers and no unique minimizers.

Problem 5

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Problem 5 →

Plane $P \rightarrow \vec{b} \cdot \vec{x} + a = 0$, $\vec{b} \in \mathbb{R}^3$, $\vec{x} \in \mathbb{R}^3$, $a \in \mathbb{R}$, Let $\vec{b} = [c, d, e]^T$
 $J = \frac{1}{2} \vec{x} \cdot \vec{x}$, $g(\vec{x}) = cx + dy + ez + a = 0$
 $\vec{x} = [x, y, z]^T$

$$J = \frac{1}{2}(x^2 + y^2 + z^2), \quad g(\vec{x}) = cx + dy + ez + a = 0$$

$$L = \frac{1}{2}(x^2 + y^2 + z^2) + \lambda(cx + dy + ez + a)$$

$$\begin{cases} \frac{\partial L}{\partial x} = x + c\lambda = 0 \\ \frac{\partial L}{\partial y} = y + d\lambda = 0 \\ \frac{\partial L}{\partial z} = z + e\lambda = 0 \end{cases} \rightarrow \begin{cases} x = -c\lambda \\ y = -d\lambda \\ z = -e\lambda \end{cases}$$

$$\frac{\partial L}{\partial \lambda} = cx + dy + ez + a = 0 \rightarrow -c^2\lambda - d^2\lambda - e^2\lambda + a = 0 \rightarrow \lambda = \frac{a}{c^2 + d^2 + e^2}$$

$$x^* = \frac{-ac}{c^2 + d^2 + e^2}, \quad y^* = \frac{-ad}{c^2 + d^2 + e^2}, \quad z^* = \frac{-ae}{c^2 + d^2 + e^2}$$

Now, use $J = \sqrt{\vec{x} \cdot \vec{x}} = \sqrt{x^2 + y^2 + z^2} = m$, $g(\vec{x})$ is the same

$$L = \sqrt{x^2 + y^2 + z^2} + \lambda(cx + dy + ez + a)$$

$$\begin{cases} \frac{\partial L}{\partial x} = \frac{x}{m} + \lambda c = 0 \\ \frac{\partial L}{\partial y} = \frac{y}{m} + \lambda d = 0 \\ \frac{\partial L}{\partial z} = \frac{z}{m} + \lambda e = 0 \end{cases} \rightarrow \begin{cases} x = -\lambda cm \\ y = -\lambda dm \\ z = -\lambda em \end{cases} \quad \begin{aligned} x^2 + y^2 + z^2 = m^2 &= \lambda^2 m^2 (c^2 + d^2 + e^2) \\ \lambda &= \frac{1}{\sqrt{c^2 + d^2 + e^2}} \end{aligned}$$

$$\frac{\partial L}{\partial \lambda} = cx + dy + ez + a = 0 \rightarrow -c^2 m \lambda - d^2 m \lambda - e^2 m \lambda + a = 0$$

$$m \lambda (c^2 + d^2 + e^2) = a \rightarrow \lambda m = \frac{a}{c^2 + d^2 + e^2} \rightarrow m = \frac{1}{\lambda} \cdot \frac{a}{c^2 + d^2 + e^2} = \frac{a}{\sqrt{c^2 + d^2 + e^2}} = m^*$$

$$x^* = \frac{-ac}{c^2 + d^2 + e^2}, \quad y^* = \frac{-ad}{c^2 + d^2 + e^2}, \quad z^* = \frac{-ae}{c^2 + d^2 + e^2}$$

Problem 6

Problem 6 →

$$\vec{x} = [x, y, z]^T, \vec{b} = [c, d, e]^T, h(\vec{x}) = cx + dy + ez + a \geq 0$$

$$J = \frac{1}{2} \vec{x} \cdot \vec{x} = \frac{1}{2} (x^2 + y^2 + z^2)$$

Use \vec{x}^* from Problem 5 and evaluate $h(\vec{x}^*)$ to determine answer.

$$h(\vec{x}^*) = \frac{-ac^2}{c^2+d^2+e^2} - \frac{ad^2}{c^2+d^2+e^2} - \frac{ae^2}{c^2+d^2+e^2} + a = -a \left[\frac{c^2+d^2+e^2}{c^2+d^2+e^2} \right] + a = -a + a = 0$$

∴ Inequality is active,

$$L = \frac{1}{2} (x^2 + y^2 + z^2) + \sigma (cx + dy + ez + a)$$

$$\begin{cases} \frac{\partial L}{\partial x} = x + c\sigma = 0 \\ \frac{\partial L}{\partial y} = y + d\sigma = 0 \\ \frac{\partial L}{\partial z} = z + e\sigma = 0 \end{cases} \rightarrow \begin{cases} x = -c\sigma \\ y = -d\sigma \\ z = -e\sigma \end{cases} \rightarrow \vec{x} = -\vec{b}\sigma$$

$$h(\vec{x}) \geq 0 \rightarrow cx + dy + ez + a \geq 0 \rightarrow -c^2\sigma - d^2\sigma - e^2\sigma + a \geq 0$$

$$a \geq \sigma(c^2 + d^2 + e^2) \rightarrow \frac{a}{c^2 + d^2 + e^2} \geq \sigma \rightarrow \sigma \leq \frac{a}{\vec{b} \cdot \vec{b}}$$

$$\vec{x}^* \leq -\vec{b} \cdot \frac{a}{\vec{b} \cdot \vec{b}}$$

Problem 7

Problem 7 →

$$f(\vec{x}) = \frac{1}{2} \vec{x}^T \cdot Q \cdot \vec{x}, \quad g_1(\vec{x}) = \vec{a}_1 \cdot \vec{x} + b_1 = 0, \quad g_2(\vec{x}) = \vec{a}_2 \cdot \vec{x} + b_2 = 0$$

$$\vec{x} = [x_1, x_2, x_3]^T, \quad \vec{a}_1 = [a, b, c]^T, \quad \vec{a}_2 = [d, e, f]^T, \quad \vec{\lambda} = [\lambda_1, \lambda_2]^T$$

$$f(\vec{x}) = \frac{1}{2} [x_1, x_2, x_3] \begin{bmatrix} 1 & -1 & 1 \\ 1 & 1 & -1 \\ -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \frac{1}{2} [x_1, x_2, x_3] \begin{bmatrix} x_1 - x_2 + x_3 \\ x_1 + x_2 - x_3 \\ -x_1 + x_2 + x_3 \end{bmatrix} = \frac{1}{2} [x_1^2 - x_1 x_2 + x_1 x_3 + x_1 x_2 + x_2^2 - x_2 x_3 - x_1 x_3 + x_2 x_3 + x_3^2]$$

$$= \frac{1}{2} [x_1^2 + x_2^2 + x_3^2]$$

$$L = \frac{1}{2} [x_1^2 + x_2^2 + x_3^2] + \lambda_1 (a x_1 + b x_2 + c x_3 + b_1) + \lambda_2 (d x_1 + e x_2 + f x_3 + b_2)$$

a) We want $\frac{\partial b_1}{\partial \vec{x}}$ and $\frac{\partial b_2}{\partial \vec{x}}$ to be linearly independent:

$$\frac{\partial b_1}{\partial \vec{x}} = \vec{a}_1 + b_1, \quad \frac{\partial b_2}{\partial \vec{x}} = \vec{a}_2 + b_2$$

$m(\vec{a}_1 + b_1) + n(\vec{a}_2 + b_2) \neq 0$ where $m, n \in \mathbb{R}$

b)

$$\begin{aligned} \frac{\partial L}{\partial x_1} &= x_1 + a \lambda_1 + d \lambda_2 = 0 \\ \frac{\partial L}{\partial x_2} &= x_2 + b \lambda_1 + e \lambda_2 = 0 \\ \frac{\partial L}{\partial x_3} &= x_3 + c \lambda_1 + f \lambda_2 = 0 \end{aligned} \rightarrow \begin{cases} x_1 = -a \lambda_1 - d \lambda_2 \\ x_2 = -b \lambda_1 - e \lambda_2 \\ x_3 = -c \lambda_1 - f \lambda_2 \end{cases} = -\vec{a}_1 \lambda_1 - \vec{a}_2 \lambda_2$$

$$\frac{\partial L}{\partial \lambda_1} = a x_1 + b x_2 + c x_3 + b_1 = 0 = -a^2 \lambda_1 - a d \lambda_2 - b^2 \lambda_1 - b e \lambda_2 - c^2 \lambda_1 - c f \lambda_2 + b_1 = 0$$

$$b_1 = \lambda_1 (a^2 + b^2 + c^2) + \lambda_2 (a d + b e + c f) \rightarrow \lambda_1 = \frac{b_1 - \lambda_2 (a d + b e + c f)}{a^2 + b^2 + c^2}$$

$$\frac{\partial L}{\partial \lambda_2} = d x_1 + e x_2 + f x_3 + b_2 = 0 = -a d \lambda_1 - d^2 \lambda_2 - b e \lambda_1 - e^2 \lambda_2 - f c \lambda_1 - f^2 \lambda_2 + b_2 = 0$$

$$b_2 = \lambda_1 (a d + b e + f c) + \lambda_2 (d^2 + e^2 + f^2)$$

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$$b_2 = \frac{b_1 - \lambda_2(ad+be+cf)^2}{a^2+b^2+c^2} + \lambda_2(d^2+e^2+f^2) = \frac{b_1 - \lambda_2(\hat{q}_1 \cdot \hat{q}_2)^2}{\hat{q}_1 \cdot \hat{q}_1} + \lambda_2(\hat{q}_2 \cdot \hat{q}_2)$$

$$= \frac{b_1 - \lambda_2(\hat{q}_1 \cdot \hat{q}_2)^2 + \lambda_2(\hat{q}_2 \cdot \hat{q}_2)(\hat{q}_1 \cdot \hat{q}_1)}{\hat{q}_1 \cdot \hat{q}_1} \rightarrow b_2 \cdot \hat{q}_1 \cdot \hat{q}_1 - b_1 = \lambda_2[(\hat{q}_2 \cdot \hat{q}_2)(\hat{q}_1 \cdot \hat{q}_1) - (\hat{q}_1 \cdot \hat{q}_2)^2]$$

$$\lambda_2 = \frac{b_2 \cdot \hat{q}_1 \cdot \hat{q}_1 - b_1}{(\hat{q}_2 \cdot \hat{q}_2)(\hat{q}_1 \cdot \hat{q}_1) - (\hat{q}_1 \cdot \hat{q}_2)^2}$$

$$\rightarrow \lambda_1 = \frac{b_1 - \lambda_2(\hat{q}_1 \cdot \hat{q}_2)}{\hat{q}_1 \cdot \hat{q}_1}$$

$$= \frac{b_1 - [b_2(\hat{q}_1 \cdot \hat{q}_1) - b_1] \hat{q}_1 \cdot \hat{q}_2}{(\hat{q}_2 \cdot \hat{q}_2)(\hat{q}_1 \cdot \hat{q}_1) - (\hat{q}_1 \cdot \hat{q}_2)^2}$$

$$\vec{z}^* = -\hat{q}_1 \lambda_1 - \hat{q}_2 \lambda_2$$

$$\vec{z}^* = -\hat{q}_1 \left[\frac{b_2(\hat{q}_1 \cdot \hat{q}_1) - b_1}{(\hat{q}_2 \cdot \hat{q}_2)(\hat{q}_1 \cdot \hat{q}_1) - (\hat{q}_1 \cdot \hat{q}_2)^2} \right] - \hat{q}_2 \left[\frac{b_1 - \frac{[b_2(\hat{q}_1 \cdot \hat{q}_1) - b_1] \hat{q}_1 \cdot \hat{q}_2}{(\hat{q}_2 \cdot \hat{q}_2)(\hat{q}_1 \cdot \hat{q}_1) - (\hat{q}_1 \cdot \hat{q}_2)^2}}{\hat{q}_1 \cdot \hat{q}_1} \right]$$

Problem 8

Problem 8 \rightarrow $GM(\vec{x}) = (x_1 x_2 \dots x_n)^{1/n} = g$ $AM(\vec{x}) = \frac{1}{n} [x_1 + x_2 + \dots + x_n] = J$

$$g(\vec{x}) = (x_1 x_2 \dots x_n)^{1/n} - g = 0, \quad J = \frac{1}{n} [x_1 + x_2 + \dots + x_n]$$

$$L = \frac{1}{n} [x_1 + x_2 + \dots + x_n] + \lambda [(x_1 x_2 \dots x_n)^{1/n} - g]$$

$$\frac{\partial L}{\partial x_1} = \frac{1}{n} + \frac{\lambda}{n} (x_1 x_2 \dots x_n)^{\frac{1}{n}-1} (x_2 x_3 \dots x_n) = 0 = \frac{1}{n} + \frac{\lambda}{n} \frac{(x_1 x_2 \dots x_n)^{1/n}}{(x_1 x_2 \dots x_n)} \cdot (x_2 x_3 \dots x_n)$$

$$\rightarrow \frac{1}{g} = -\frac{\lambda}{g} \frac{g}{x_1} \rightarrow x_1 = -\lambda g \rightarrow \lambda = -\frac{x_1}{g}$$

$$\frac{\partial L}{\partial x_2} = 0 \rightarrow \text{should lead to } \lambda = -\frac{x_2}{g}, \quad \frac{\partial L}{\partial x_n} \rightarrow \lambda = -\frac{x_n}{g}$$

$$\lambda \text{ s are all equal} \rightarrow \therefore \lambda = -\frac{x_1}{g} = -\frac{x_2}{g} = \dots = -\frac{x_n}{g} \rightarrow x_1 = x_2 = \dots = x_n$$

\therefore AM x_i should be equal for J to be minimum

Problem 9

Problem 9 \rightarrow $f(x) = (x_1 - 2)^2 + x_2^2 + x_3^2$, $g_1(x) = x_1^2 + x_2^2 + x_3^2 - 2 = 0$, $g_2(x) = x_1^2 + x_2^2 - 1 = 0$

$$L = (x_1 - 2)^2 + x_2^2 + x_3^2 + \lambda_1 (x_1^2 + x_2^2 + x_3^2 - 2) + \lambda_2 (x_1^2 + x_2^2 - 1)$$

$$\frac{\partial L}{\partial x_1} = 2x_1 + 2\lambda_1 x_1 + 2\lambda_2 x_1 = 0 \quad \left\{ \begin{array}{l} 2x_1 + 2\lambda_1 x_1 + 2\lambda_2 x_1 = 0 \\ 2x_2 + 2\lambda_1 x_2 + 2\lambda_2 x_2 = 0 \end{array} \right. \xrightarrow{\text{add}} 2\lambda_2 (x_1 + x_2) = 0$$

$$\frac{\partial L}{\partial x_2} = 2x_2 + 2\lambda_1 x_2 + 2\lambda_2 x_2 = 0 \quad \left\{ \begin{array}{l} 2x_1 + 2\lambda_1 x_1 + 2\lambda_2 x_1 = 0 \\ 2x_2 + 2\lambda_1 x_2 + 2\lambda_2 x_2 = 0 \end{array} \right. \quad \therefore \lambda_2 \neq 0 \rightarrow \boxed{x_1 = -x_2}$$

$$\frac{\partial L}{\partial x_3} = 2x_3 + 2\lambda_1 x_3 = 0 \rightarrow \boxed{x_3 = -\lambda_1 x_3 + \lambda_1 = -1}$$

$$g_1(x) = x_1^2 + x_2^2 + x_3^2 - 2 = 0 \rightarrow 2x_1^2 + x_3^2 = 2 \rightarrow x_3^2 = 1 \rightarrow \boxed{x_3 = \pm 1}$$

$$g_2(x) = x_1^2 + x_2^2 - 1 = 0 \rightarrow 2x_1^2 = 1 \rightarrow \boxed{x_1 = \pm \frac{1}{\sqrt{2}} \rightarrow x_2 = \mp \frac{1}{\sqrt{2}}}$$

KKT conditions apply: $f(x)$, $g_1(x)$, $g_2(x)$ are C^1 .

$$\frac{\partial^2 f}{\partial x^2} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 x_2} & \frac{\partial^2 f}{\partial x_1 x_3} \\ \frac{\partial^2 f}{\partial x_2 x_1} & \frac{\partial^2 f}{\partial x_2^2} & \frac{\partial^2 f}{\partial x_2 x_3} \\ \frac{\partial^2 f}{\partial x_3 x_1} & \frac{\partial^2 f}{\partial x_3 x_2} & \frac{\partial^2 f}{\partial x_3^2} \end{bmatrix} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \rightarrow \left| \frac{\partial^2 f}{\partial x^2} \right| = 8 > 0$$

$\frac{\partial^2 f}{\partial x^2}$ is positive definite

$\therefore f(x)$ is convex.

$$\vec{g}(x) = \begin{bmatrix} g_1(x) \\ g_2(x) \end{bmatrix} = \begin{bmatrix} x_1 & x_2 & x_3 \\ x_1 & x_2 & 0 \end{bmatrix} + \begin{bmatrix} -2 \\ -1 \end{bmatrix} = 0 = \bar{A}\bar{x} + \bar{b} \rightarrow \therefore \boxed{x^* \text{ is the minimizer}}$$

Problem 10

Problem 10 $\rightarrow \vec{x} = [x, y, z]^T, R^2 = x^2 + y^2 + z^2, f(\vec{x}) = y^2 + z^2, g(\vec{x}) = x^2 + y^2 + z^2 - R^2 = 0$

$h(\vec{x}) = x - R \geq 0$, Assuming $R^2 = x^2 + y^2 + z^2$ & $R \geq 0$.

$g(\vec{x})$ & $h(\vec{x})$ are not linearly dependent i.e. they are linearly independent

because $\rightarrow m \cdot g(\vec{x}) + n \cdot h(\vec{x}) \neq 0$ where $m, n \in \mathbb{R} [m, n \neq 0]$

Similarly $\rightarrow \frac{\partial g}{\partial \vec{x}} = \begin{bmatrix} \frac{\partial g}{\partial x} \\ \frac{\partial g}{\partial y} \\ \frac{\partial g}{\partial z} \end{bmatrix} = \begin{bmatrix} 2x \\ 2y \\ 2z \end{bmatrix}, \frac{\partial h}{\partial \vec{x}} = \begin{bmatrix} \frac{\partial h}{\partial x} \\ \frac{\partial h}{\partial y} \\ \frac{\partial h}{\partial z} \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$

\therefore There does not exist c_1, c_2 (non-zero) $\in \mathbb{R}$ s.t.

$c_1 \begin{bmatrix} 2x \\ 2y \\ 2z \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \therefore \frac{\partial g}{\partial \vec{x}} \& \frac{\partial h}{\partial \vec{x}}$ are linearly independent. $\therefore \lambda_0 = 1$

$L = y^2 + z^2 + \lambda [x^2 + y^2 + z^2 - R^2] + \sigma (x - R)$

$\frac{\partial L}{\partial x} = 2x\lambda + \sigma = 0 \quad \left. \begin{array}{l} \frac{\partial L}{\partial y} = 2y + 2y\lambda = 0 \\ \frac{\partial L}{\partial z} = 2z + 2z\lambda = 0 \end{array} \right\} \begin{array}{l} x = -\frac{\sigma}{2\lambda} \rightarrow x = +\frac{\sigma}{2} \rightarrow \sigma = 2x \\ 2y(1+\lambda) = 0 \\ 2z(1+\lambda) = 0 \end{array} \rightarrow \lambda = -1 \rightarrow \begin{array}{l} y^* = 0 \\ z^* = 0 \end{array}$

$x^2 + y^2 + z^2 - R^2 = 0 \rightarrow x^2 = R^2 \rightarrow \boxed{x^* = R}$ (where $R \geq 0$), $R \in \mathbb{R}$

Problem 11

ASEN 6020
Spring 2021
Josh Brauer

HW #2

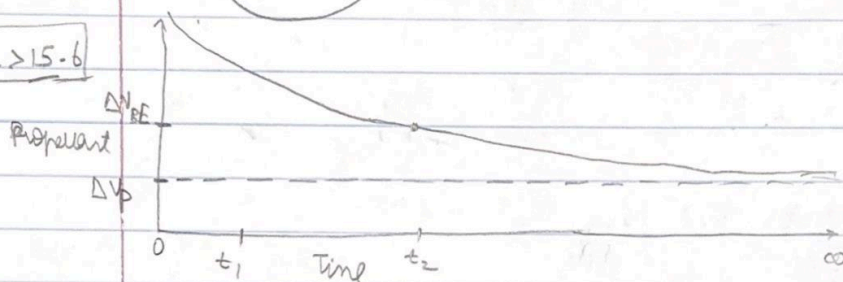
Problem 11 →

$$R \geq 11.94$$

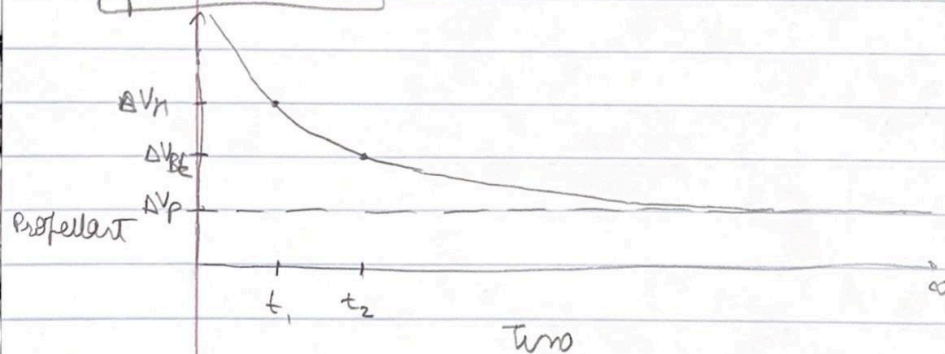


Time to reach the surface
from the center

$$\text{If } R > 15.6$$



$$\text{If } R \leq 15.6 \text{ \& } R \geq 11.94$$



There are two scenarios when $r > 11.94$ for the Pareto front optimal ΔV -transfer time sketch. First is when r is > 11.94 AND $r < 15.6$. The second is when $r > 15.6$. The top plot shows the latter case ($r > 15.6$). When $r > 15.6$, there are only two optimal options - parabolic or bi-elliptic transfers. When time is at t_1 (from the top figure), it's really close to 0 and a hyperbolic transfer is needed. But, at some point time t_2 bi-elliptic transfers are the most optimal and as time increases, parabolic transfers are most optimal. When time is ∞ , parabolic transfer is optimal.

Then, the second case is when $r > 11.94$ AND $r < 15.6$. Here, when time is really small (close to 0), hyperbolic transfers will still be the most optimal. But, as time increases, and reaches some time t_1 (from the bottom figure), hohmann transfer is most optimal (there exists an intermediate ratio I where $J_H < J_{BE}$). Then, as time reaches time t_2 , bi-elliptic transfers are most optimal and same as the previous case, as time reaches ∞ , parabolic transfer is optimal.