

HW #5

$$\text{problem 1} \rightarrow \ddot{\theta} + 2\xi\omega_n\dot{\theta} + \omega_n^2 \sin \theta = 0$$

8.3 a)  $\theta_0 = 0, \vec{\theta} = [\theta, \dot{\theta}]^T$

$$\dot{\vec{\theta}} = \begin{bmatrix} \dot{\theta} \\ \ddot{\theta} \end{bmatrix} = \vec{f}(\vec{\theta}) = \begin{bmatrix} \dot{\theta} \\ -2\xi\omega_n\dot{\theta} - \omega_n^2 \sin \theta \end{bmatrix}, [A] = \frac{\partial \vec{f}}{\partial \vec{\theta}} = \begin{bmatrix} \partial f_1 / \partial \theta & \partial f_1 / \partial \dot{\theta} \\ \partial f_2 / \partial \theta & \partial f_2 / \partial \dot{\theta} \end{bmatrix}$$

$$[A] = \begin{bmatrix} 0 & 1 \\ -\omega_n^2 \cos \theta & -2\xi\omega_n \end{bmatrix} \rightarrow \delta \dot{\vec{\theta}} \approx \begin{bmatrix} 0 & 1 \\ -\omega_n^2 \cos \theta & -2\xi\omega_n \end{bmatrix} \delta \vec{\theta}$$

$$\text{Reference} \rightarrow \theta_0 = 0 \rightarrow \ddot{\vec{\theta}} \approx [A]\vec{\theta} + [B]\vec{u}^{\neq 0} = \begin{bmatrix} 0 & 1 \\ -\omega_n^2 & -2\xi\omega_n \end{bmatrix} \vec{\theta} = \dot{\vec{\theta}}$$

b)  $\xi = 0 \rightarrow \dot{\vec{\theta}} = \begin{bmatrix} 0 & 1 \\ -\omega_n^2 & 0 \end{bmatrix} \vec{\theta} = [A]\vec{\theta}$

$$\text{Calculate eigenvalues} \rightarrow |A - \lambda I| = \begin{vmatrix} 0 & 1 \\ -\omega_n^2 & 0 \end{vmatrix} - \lambda \begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = \begin{vmatrix} -\lambda & 1 \\ -\omega_n^2 & -\lambda \end{vmatrix} = \lambda^2 + \omega_n^2 = 0$$

$\lambda^2 = -\omega_n^2 \rightarrow \lambda = \pm \sqrt{-\omega_n^2}, \omega_n^2 > 0 \rightarrow \lambda$  is purely imaginary.  $\therefore$  linearized system is marginally stable. Small motions about  $\theta_0$  are marginally stable.

c)  $v(\vec{\theta}) = \vec{\theta}^T [P] \vec{\theta} \leftarrow \text{candidate Lyapunov function}$

$$\dot{v}(\vec{\theta}) = \dot{\vec{\theta}}^T [P] \vec{\theta} + \vec{\theta}^T [P] \dot{\vec{\theta}} \quad (\dot{\vec{\theta}} = [A]\vec{\theta})$$

$$= ([A]\vec{\theta})^T [P] \vec{\theta} + \vec{\theta}^T [P][A]\vec{\theta} = \vec{\theta}^T ([A]^T [P] + [P][A]) \vec{\theta} = -\vec{\theta}^T [R] \vec{\theta}$$

$$\begin{bmatrix} 0 & -\omega_n^2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} + \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -\omega_n^2 & 0 \end{bmatrix} = \begin{bmatrix} R_{11} & R_{12} \\ R_{21} & R_{22} \end{bmatrix}$$

$$\begin{bmatrix} -P_{12}\omega_n^2 - P_{22}\omega_n^2 \\ P_{11} \end{bmatrix} + \begin{bmatrix} -P_{12}\omega_n^2 & P_{11} \\ -P_{22}\omega_n^2 & P_{12} \end{bmatrix} = \begin{bmatrix} R_{11} & R_{12} \\ R_{21} & R_{22} \end{bmatrix}$$

$$R_{11} = +2P_{12}\omega_n^2, R_{12} = P_{22}\omega_n^2 - P_{11}, R_{22} = -2P_{12}$$

$$R = \begin{bmatrix} 2P_{12}\omega_n^2 & P_{22}\omega_n^2 - P_{11} \\ P_{22}\omega_n^2 - P_{11} & -2P_{12} \end{bmatrix} \rightarrow |R| = -4P_{12}^2\omega_n^2 - (P_{22}\omega_n^2 - P_{11})^2$$

For  $R$  to be positive definite,  $|R| > 0$

$$\therefore -4P_{12}^2\omega_n^2 \geq (P_{22}\omega_n^2 - P_{11})^2 \rightarrow \text{so, } |R| \text{ is not } > 0 \rightarrow [R] \text{ is not positive definite.}$$

Theorem 8.6 does not hold,  $\because [A]$  is not stable, it makes sense that 8.6 does not hold.

This means that  $\dot{\mathbf{x}} = -\mathbf{R}^T \mathbf{R} \mathbf{x}$  is not strictly negative definite.  $\therefore$  The system cannot be guaranteed to be hyperbolic stable or asymptotically stable.

d)  $A = \begin{bmatrix} 0 & 1 \\ -w_n^2 & -2\zeta w_n \end{bmatrix}$ , calculate evals  $|A - \lambda I| = 0 \rightarrow \begin{vmatrix} -\lambda & 1 \\ -w_n^2 - 2\zeta w_n - \lambda & +w_n^2 \end{vmatrix} = (-\lambda)(-2\zeta w_n - \lambda)$

$$\lambda^2 + 2\zeta w_n \lambda + w_n^2 = 0 \rightarrow \lambda = \frac{-2\zeta w_n \pm \sqrt{4\zeta^2 w_n^2 - 4w_n^2}}{2} = -\zeta w_n \pm w_n \sqrt{\zeta^2 - 1}$$

$2w_n\sqrt{\zeta^2 - 1}$  will be imaginary.  $-2\zeta w_n$  will be  $< 0 \rightarrow \therefore \operatorname{Re}(\lambda) < 0$

$\therefore [A]$  is strictly stable ( $\because \operatorname{Re}(\lambda) < 0$ )

e)  $[A]^T [P] + [P] [A] = -[\mathbf{R}] \rightarrow \begin{bmatrix} 0 & -w_n^2 \\ 1 & -2\zeta w_n \end{bmatrix} \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} + \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ w_n^2 - 2\zeta w_n \end{bmatrix}$

$$= \begin{bmatrix} -P_{12}w_n^2 & -P_{22}w_n^2 \\ P_{11} - 2P_{12}\zeta w_n & P_{12} - 2P_{22}\zeta w_n \end{bmatrix} + \begin{bmatrix} -P_{12}w_n^2 & P_{11} - 2P_{12}\zeta w_n \\ -P_{22}w_n^2 & P_{12} - 2P_{22}\zeta w_n \end{bmatrix} = -\begin{bmatrix} R_{11} & R_{12} \\ R_{21} & R_{22} \end{bmatrix}$$

$$R_{11} = 2P_{12}w_n^2, \quad R_{12} = P_{22}w_n^2 + 2P_{12}\zeta w_n - P_{11}, \quad R_{22} = 4P_{22}\zeta w_n - 2P_{12}$$

f)  $R = \begin{bmatrix} R_{11} & 0 \\ 0 & R_{22} \end{bmatrix}, \quad A = \begin{bmatrix} 0 & 1 \\ -w_n^2 - 2\zeta w_n \end{bmatrix} \rightarrow R_{12} = 0 = P_{11} = P_{22}w_n^2 + 2P_{12}\zeta w_n$   
 Eval of  $[\mathbf{R}] = R_{11}, R_{22} \rightarrow R_{11} > 0, R_{22} > 0$  ( $\because R \geq 0$  pos. def.)

$$R_{11} > 0 \rightarrow 2P_{12}w_n^2 > 0 \rightarrow P_{12} > 0$$

$$R_{22} > 0 \rightarrow P_{22}^2 \zeta^2 w_n^2 > 2P_{12}^2 \zeta^2 w_n^2 \rightarrow 2P_{22}\zeta w_n > P_{12} \rightarrow P_{22} > 0 \quad (\because w_n > 0 \text{ volume reduced frequency is true.})$$

$$P = \begin{bmatrix} P_{22}w_n^2 + 2P_{12}\zeta w_n & P_{12} \\ P_{12} & P_{22} \end{bmatrix}$$

$$|P| = P_{22}^2 w_n^2 + \underbrace{2P_{12}P_{22}\zeta w_n}_{\zeta > 0} - P_{12}^2$$

$$\therefore 2P_{22}\zeta w_n > P_{12} \rightarrow 2P_1 P_{22}\zeta w_n > P_{12}^2 \rightarrow \boxed{2P_1 P_{22}\zeta > P_{12}^2}$$

And,  $P_{22}^2 w_n^2 > 0$  (square of 2 terms)

$\therefore \boxed{|P| > 0} \rightarrow P \text{ is positive definite}$

HW #5

problem 2 →

$$V = 2K \ln(1 + \sigma^T \sigma)$$

$$\sigma^2 = \sigma^T \sigma$$

$${}^B\vec{w}^T \sigma = \sigma^T {}^B\vec{w}$$

$$\dot{V}(\sigma) = \frac{\partial V}{\partial \sigma} \cdot \dot{\sigma} = \frac{2K}{1 + \sigma^2} \cdot (2\vec{\sigma}) \cdot \dot{\sigma} = \frac{4K\sigma^T}{1 + \sigma^2} \cdot \dot{\sigma}$$

$$\dot{\sigma} = \frac{1}{4} [(1 - \sigma^2) I_{3 \times 3} + 2[\vec{\sigma}] + 20\sigma^T] {}^B\vec{w}$$

$$\dot{V}(\sigma) = \frac{4K\sigma^T}{1 + \sigma^2} \cdot \frac{1}{4} [(1 - \sigma^2) I_{3 \times 3} + 2[\vec{\sigma}] + 20\sigma^T] {}^B\vec{w}$$

$$= \frac{K}{1 + \sigma^2} [\sigma^T {}^B\vec{w} - \sigma^T \sigma^2 {}^B\vec{w} + \cancel{2\sigma^T [\vec{\sigma}] {}^B\vec{w}} + 2\sigma^T \sigma \sigma^T {}^B\vec{w}]$$

$$\sigma^T \sigma^2 {}^B\vec{w} \text{ is the same as } \sigma^T \sigma \sigma^T {}^B\vec{w} \rightarrow \therefore -\sigma^T \cancel{\sigma^2} {}^B\vec{w} + 2\sigma^T \sigma \sigma^T {}^B\vec{w} = \sigma^T \sigma^2 {}^B\vec{w}$$

$$[\vec{\sigma}] {}^B\vec{w} = \sigma \times {}^B\vec{w} \rightarrow \text{Resulting vector is } \perp \text{ to } \sigma, {}^B\vec{w} \rightarrow \therefore \sigma^T [\vec{\sigma}] {}^B\vec{w} = \vec{0}$$

$$\dot{V}(\sigma) = \frac{K}{1 + \sigma^2} [\sigma^T {}^B\vec{w} + \sigma^T \sigma^2 {}^B\vec{w}] = \frac{K}{1 + \sigma^2} (\sigma^T {}^B\vec{w}) \cancel{[\sigma^2]} = K \sigma^T {}^B\vec{w} = [{}^B\vec{w}^T (K\sigma)] = \dot{V}(\sigma)$$

Alternate derivation →

$$\dot{V}(\sigma) = \frac{2K}{1 + \sigma^2} (\sigma^T \dot{\sigma} + \dot{\sigma}^T \sigma) = \frac{2K}{1 + \sigma^2} \left[ \sigma^T \cdot \frac{1}{4} [(1 - \sigma^2) I_{3 \times 3} + 2[\vec{\sigma}] + 20\sigma^T] {}^B\vec{w} \right]$$

$$+ {}^B\vec{w}^T \left[ \frac{1}{4} [(1 - \sigma^2) I_{3 \times 3} - 2[\vec{\sigma}] + 20\sigma^T] \sigma \right]$$

$$= \frac{2K}{1 + \sigma^2} \cdot \frac{1}{4} \left[ \sigma^T (1 - \sigma^2) {}^B\vec{w} + \cancel{\sigma^T \cdot 2[\vec{\sigma}] {}^B\vec{w}} + 2\sigma^T \sigma \sigma^T {}^B\vec{w} + {}^B\vec{w}^T (1 - \sigma^2) \sigma - \cancel{2\vec{w}^T [\vec{\sigma}] \sigma} + 2\vec{w}^T \sigma^T \sigma \right]$$

$$= \frac{K}{2(1 + \sigma^2)} \left[ 2{}^B\vec{w}^T (1 - \sigma^2) \sigma + 4{}^B\vec{w}^T \sigma^2 \sigma \right] = \frac{K}{2(1 + \sigma^2)} \left[ 2{}^B\vec{w}^T \sigma - 2{}^B\vec{w}^T \sigma^2 \sigma + 4{}^B\vec{w}^T \sigma^2 \sigma \right]$$

$$= \frac{K}{1 + \sigma^2} \left[ {}^B\vec{w}^T \sigma + {}^B\vec{w}^T \sigma^2 \sigma \right] = \frac{K}{1 + \sigma^2} \left[ {}^B\vec{w}^T \sigma (1 + \sigma^2) \right]$$

$$= K {}^B\vec{w}^T \sigma = [{}^B\vec{w}^T (K\sigma)] = \dot{V}(\sigma)$$