

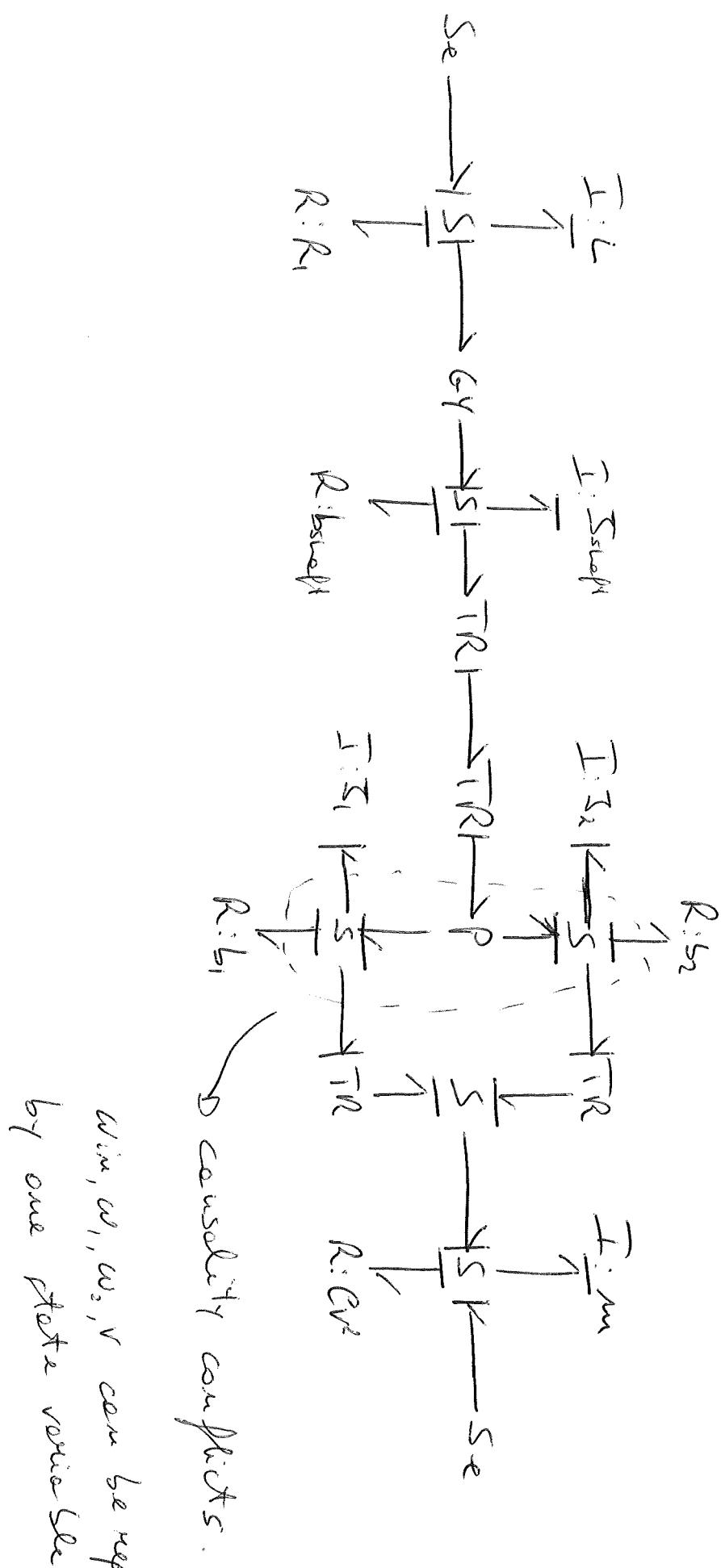
SOLUTION OF THE
MODELING AND SIMULATION
EXAM (ESS 101)

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(1)

Ex 1. Q)



w_1, w_2, v can be replaced
by one state variable.

Ex 1.5)

$$i = -\frac{R_1}{L} i - K\omega$$

$$\dot{\omega}_{im} = -\frac{b_{im}}{J_{im}} + K_i - \bar{T}_{im}$$

$$\ddot{\omega}_1 = \frac{\bar{T}_{im}\tau}{2J_1} - \frac{b_1}{J_1} \omega_1 - \bar{F}_1 \tau_1$$

$$\ddot{\omega}_2 = \frac{\bar{T}_{im}\tau}{2J_2} - \frac{b_2}{J_2} \omega_2 - \bar{F}_2 \tau_2$$

$$\ddot{v} = \frac{\bar{F}_1 + \bar{F}_2}{m} - \frac{Cv^2}{m} - g \sin \alpha$$

$$V = \omega_1 \tau_1 = \omega_2 \tau_2$$

$$\omega_{im} = \frac{\tau}{2} (\omega_1 + \omega_2)$$

$$\bar{F}_1 = \bar{F}_2 = \frac{\tau}{2} \bar{T}_{im}$$

Ex 2.9)

$\alpha \rightarrow 0 = 0.6$ and $\beta \rightarrow 0.3$ as $N \rightarrow \infty$

$$\begin{aligned} y(t) &= \theta_0^\top \varphi(t) + e(t) && \text{with } \theta_0 = \begin{bmatrix} 0.6 \\ 0.3 \end{bmatrix} \quad \varphi(t) = \begin{bmatrix} 1(t) \\ u(t) \end{bmatrix} \\ y(t+1\delta) &= \theta^\top \varphi(t) && \theta = \begin{bmatrix} -\alpha \\ \beta \end{bmatrix} \end{aligned}$$

$$\begin{aligned} \varepsilon(t+1\delta) &= y(t) - y(t+1\delta) = \\ &= (\theta_0 - \theta)^\top \varphi(t) + e(t) \end{aligned}$$

$$\begin{aligned} \text{In PNL } \theta^* &= \underset{\theta}{\operatorname{argmin}} \frac{1}{N} \sum_{t=1}^N \varepsilon^2(t+1\delta) \\ &= \underset{\theta}{\operatorname{argmin}} V_N(\theta) \end{aligned}$$

$$\text{As } N \rightarrow \infty \quad V_N(\theta) \rightarrow \bar{E}[\varepsilon^2(t+1\delta)]$$

$$\text{Hence as } N \rightarrow \infty \quad \theta^* = \underset{\theta}{\operatorname{argmin}} \bar{E}[\varepsilon^2(t+1\delta)]$$

$$\begin{aligned} \bar{E}[\varepsilon^2(t+1\delta)] &= \bar{E}[\varphi^\top (\theta_0 - \theta)^* (\theta_0 - \theta)^\top \varphi(t) + \\ &\quad + 2(\theta_0 - \theta)^\top \varphi(t) e(t) + e^2(t)] = \\ &= \bar{E}[\varphi^\top (\theta_0 - \theta)(\theta_0 - \theta)^\top \varphi(t)] + \bar{E}[e^2(t)] \end{aligned}$$

The smallest $\bar{E}[\varepsilon^2(t+1\delta)]$ is $\bar{E}[e^2(t)]$ and is achieved with $\theta^* = \theta_0$. That is $\begin{bmatrix} -\alpha \\ \beta \end{bmatrix} = \begin{bmatrix} 0.6 \\ 0.3 \end{bmatrix}$

Ex 2.5)

The least squares formula is:

$$\hat{g}^* = R_n^{-1} f_n \text{ with } R_n = \frac{1}{N} \sum_{t=1}^N \begin{bmatrix} y^2(t-1) & u(t-1)y(t-1) \\ y(t-1)u(t-1) & u^2(t-1) \end{bmatrix}$$

$$f_n = \frac{1}{N} \sum_{t=1}^N \begin{bmatrix} y(t-1)y(t) \\ u(t-1)y(t) \end{bmatrix}$$

$$R_n \text{ invertible} \Rightarrow \mathbb{E}[u^2(t-1)] = R_n(0) \neq 0$$

Ex 3

Set $H(\omega) = \frac{1}{2+2\cos\omega t}$. Since $H(\omega) \simeq \overline{\Phi}_y(\omega)$

it holds that

$$H(\omega) \simeq |G(j\omega)|^2 \overline{\Phi}_u(\omega) \text{ with } \overline{\Phi}_u(\omega) = 0.2T$$

Hence:

$$|G(j\omega)|^2 = \frac{1}{(2+2\cos\omega T)0.2T}$$

$$G(j\omega) \text{ is of the type } G(j\omega) = \frac{1 + \sum_{k=0}^{M_{\text{num}}} \alpha_k e^{-j\omega T}}{\left(1 + \sum_{k=0}^{M_{\text{den}}} \beta_k e^{-j\omega T}\right) 0.2T}$$

Setting $M_{\text{num}} = 0$, $\alpha_0 = 0$, $M_{\text{den}} = 1$, $\beta_0 = 1$ solves the problem.

Ex 4

1) T

2) F

3) F

4) T

5) F