

# <<Event-triggered Luenberger observer>>

Notation  $y(t) \in \mathbb{R}^m = \begin{bmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_m(t) \end{bmatrix}$

$$Y(t) \in \mathbb{R}^{Tm} = \begin{bmatrix} y_1(t-T+1) \\ \vdots \\ y_1(t) \\ \vdots \\ y_m(t-T+1) \\ \vdots \\ y_m(t) \end{bmatrix}$$

Be careful!

The time-horizon model in this paper is defined a little different from the usual form.

$$Y_i(t) \in \mathbb{R}^m = \begin{bmatrix} y_{i(t-T+1)} \\ \vdots \\ y_{i(t-1)} \\ y_{i(t)} \end{bmatrix}$$

## System model development

Model (1):

$$\begin{aligned} X(t) &= AX(t-1) \\ y(t) &= CX(t) + e(t) \end{aligned}$$

$$X(t) \in \mathbb{R}^n, y(t), e(t) \in \mathbb{R}^m$$

Model (2):  $Y(t) = HX(t-T+1) + E(t)$

$$Y(t), E(t) \in \mathbb{R}^{Tm}$$

$$H = \begin{bmatrix} C_1 \\ C_1 A \\ \vdots \\ C_1 A^T \\ \vdots \\ C_m \\ C_m A \\ \vdots \\ C_m A^T \end{bmatrix}$$

Model (3):  $Y(t) = [H \ I] \begin{bmatrix} X(t-T+1) \\ E(t) \end{bmatrix} \triangleq QZ(t)$

$$Z(t) = \begin{bmatrix} X(t-T+1) \\ E(t) \end{bmatrix} \in \mathbb{R}^{n+Tm}$$

Model (4): Given shift matrices  $S$ ,  $C_i$ ,  $N_i$

$$S = \begin{bmatrix} 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \\ 0 & 0 & \cdots & 0 \end{bmatrix}$$

$$C_i = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ -C_i A^T \end{bmatrix}$$

$$N_i = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ b_i^T \end{bmatrix}$$

( $b_i$  is  $i$ th natural basis vector)

$$\begin{bmatrix} X(t-T+1) \\ E_1(t) \\ \vdots \\ E_m(t) \end{bmatrix} = \begin{bmatrix} A & 0 & \cdots & 0 \\ C_1 & S & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ C_m & 0 & \cdots & S \end{bmatrix} \begin{bmatrix} X(t-T) \\ E_1(t-1) \\ \vdots \\ E_m(t-1) \end{bmatrix} + \begin{bmatrix} 0 \\ N_1 \\ \vdots \\ N_m \end{bmatrix} y(t)$$

Model (5): → This is what we will use for estimation scheme.

$$Z(t) = \bar{A} Z(t-1) + \bar{B} y(t)$$

$$Y(t) = Q Z(t)$$

where,  $Z(t) = \begin{bmatrix} X(t-T+1) \\ E(t) \end{bmatrix}$      $\bar{A} = \begin{bmatrix} A & 0 & \cdots & 0 \\ C_1 & S & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ C_m & 0 & \cdots & S \end{bmatrix}$      $\bar{B} = \begin{bmatrix} 0 \\ N_1 \\ \vdots \\ N_m \end{bmatrix}$

$$Q = [H \quad I]$$

### projection operator

**Def.** The closest sparse solution measured by 2-norm

$\Pi$  is the projection operator such that:

$$\|\Pi(z) - z\|_2 \leq \|z' - z\|_2$$

$$z = \begin{bmatrix} x \\ E \end{bmatrix} \in \mathbb{R}^{n+rm}, \quad \Pi(z), z' \in \mathbb{R}^{n+rk} \quad k < m$$

**Calculation**

$$\Pi(z) = \Pi(x, E) = (x, \Pi(E))$$

$$E \in \mathbb{R}^{Tm} \xrightarrow{\text{Reshape}} \tilde{E} \in \mathbb{R}^{m \times T} \quad (1)$$

$$\text{Then, } \tilde{E}_i = \|\tilde{E}\|_2 \quad (\tilde{E}_i \text{ means } i\text{th row of } \tilde{E})$$

Next, set  $(m-r)$  smallest entries of  $\tilde{E}_i$  as zero,

which leads the corresponding entries in  $\Pi(E)$  be zero.

(please refer to Example 4.2 in paper)

## Algorithm

① Initialization:  $\hat{z}_0 = 0$ ,  $\bar{A}$ ,  $\bar{B}$ ,  $Q$ ,  $k$  (number of attacks),  $T$  (horizon time)

② Time update:  $\hat{z}_T(t) = \bar{A} \hat{z}(t-1) + \bar{B} y(t)$

③ solve projection operator:

$$\hat{z}_{\pi}(t) = \Pi(\hat{z}_T(t)) \quad \left\{ \begin{array}{l} 1^{\circ} \hat{E}_T(t) = \hat{z}_T(t)[n+1: \text{end}] \in \mathbb{R}^{Tm} \\ 2^{\circ} \text{Reshape } \hat{E}_T(t) \text{ to } \tilde{E}_T(t) \in \mathbb{R}^{m \times T} \\ 3^{\circ} \text{calculate 2-norm of each row of } \tilde{E}_T(t) \rightarrow \bar{E}_T(t) \in \mathbb{R}^m \\ 4^{\circ} \text{set } (m-k) \text{ smallest entries of } \bar{E}_T(t) \text{ be zero,} \\ \text{collect the indexes in set "idx"} \\ 5^{\circ} \Pi(\hat{E}_T(t)) = \hat{E}_T(t)[\sim \text{idx}] \\ \quad [\text{set idx of } \hat{E}_T(t) \text{ be zero, then return } \hat{E}_T(t) \text{ to } \Pi(\hat{E}_T(t))] \\ 6^{\circ} \Pi(\hat{z}_T(t)) = \begin{bmatrix} \hat{z}_T(t)[1:n] \\ \Pi(\hat{E}_T(t)) \end{bmatrix} \end{array} \right.$$

$$\hat{z}(t) = \hat{z}_{\pi}(t)$$

④ Lyapunov-based Event trigger:

$$\text{Define } V(\hat{z}(t)) = \frac{1}{2} \|Y(t) - Q\hat{z}(t)\|_2^2$$

while  $V(\hat{z}_{\pi}(t)) \geq (1-\alpha) V(\hat{z}_{\pi}(t-1))$  do: (set  $\alpha \in [0, 1]$ , the author did not talk about it, let's try  $\alpha=0.01$  firstly)

$$m=0$$

$\hat{z}_{\pi}(t) = \Pi(\hat{z}(t))$  (Do the ③ process again)

$$\hat{z}^{(0)}(t) = \hat{z}_{\pi}(t)$$

$$V_{\text{temp}} = V(\hat{z}_{\pi}(t))$$

while  $V_{\text{temp}} \geq (1-\alpha) V(\hat{z}_{\pi}(t))$  do:

$$\hat{z}^{(m+1)}(t) = \hat{z}^{(m)}(t) + L(Y(t) - Q\hat{z}^{(m)}(t))$$

(where,  $L = Q^T Z$ , choose a positive definite matrix  $Z$ )

$$V_{\text{temp}} = V[\Pi(\hat{z}^{(m+1)}(t))] \quad \text{(Do ③ again and calculate } V)$$

$$m=m+1$$

end while

$$\hat{z}(t) = \hat{z}^{(m)}(t)$$

end while