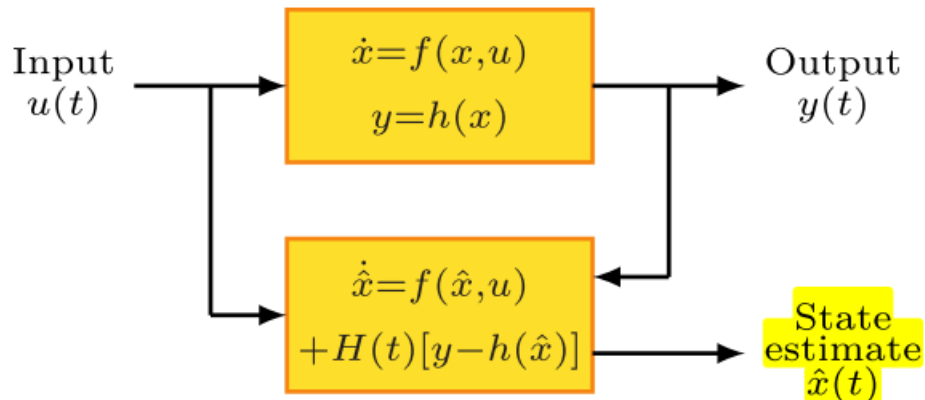


# 09\_Nonlinear\_Observers

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## 1. Introduction



State is often not directly available, controllers need estimates system state.

### Goal:

Choose an observer gain  $H(t)$  such that the observer's state  $\hat{x}(t)$  converges towards the state of the observed system,  $x(t)$ .

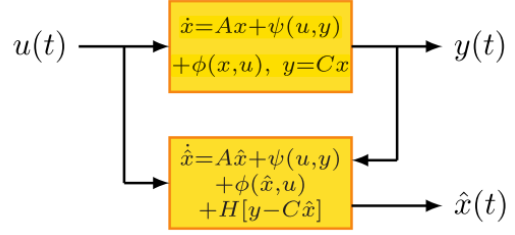
That is, the state reconstruction error  $\tilde{x}(t)$  should vanish over time:

$$\tilde{x}(t) := x(t) - \hat{x}(t) \rightarrow 0, \quad t \rightarrow \infty.$$

## 2. Global Observer

### Target:

Choose a **constant observer gain**  $H(t) \equiv H$  such that  
 $\tilde{x}(t) := x(t) - \hat{x}(t) \rightarrow 0, \quad t \rightarrow \infty$



## Error Dynamics

$$\begin{aligned} \dot{\tilde{x}} = \dot{x} - \dot{\hat{x}} &= [Ax + \psi(u, y) + \phi(x, u)] - [A\hat{x} + \psi(u, y) + \phi(\hat{x}, u) + H(y - C\hat{x})] \\ &= (A - HC)\tilde{x} + \phi(x, u) - \phi(\hat{x}, u). \end{aligned}$$

## Theorem For Global Observers

### Theorem For Global Observers:

We consider the system  $\dot{x} = Ax + \psi(u, y) + \phi(x, u), y = Cx$  with

$$\psi : \mathbb{R}^m \times \mathbb{R}^p \rightarrow \mathbb{R}^n, \quad \phi : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^p \quad \text{locally Lipschitz}$$

Let  $x(t)$  and  $u(t)$  exist at all times. Assume that

1. there exists a Lipschitz constant  $L > 0$  such that (depend on system dynamics)

$$\|\phi(x, u) - \phi(z, u)\| \leq L\|x - z\|, \quad \forall x, z, u$$

2. the positive definite solution  $P$  of the Lyapunov equation (depend on designer)

$$P(A - HC) + (A - HC)^T P = -I$$

$$\text{satisfies } L < \frac{1}{2\|P\|}.$$

Then there exist  $k, \lambda > 0$ , independent of  $x(t)$  and  $u(t)$ , such that

$$\|\tilde{x}(t)\| \leq k \|\tilde{x}(t_0)\| e^{-\lambda(t-t_0)}, \quad \forall t \geq t_0, \tilde{x}(t_0) \in \mathbb{R}^n$$

**Note:**

The  $V(z) = z^T P z$  is a candidate Lyapunov function

## 3. High Gain Observers

### Model:

- **System Dynamics:**  $w$  is state/parameter we do not care about or we have uncertainty on it

$$\begin{cases} \dot{w} = f_0(w, x, u) \\ \dot{x}_i = x_{i+1} + \psi_i(x_1, \dots, x_i, u), \quad i \in \{1, \dots, \rho-1\} \\ \dot{x}_\rho = \phi(w, x, u) \\ y = x_1 \end{cases}$$

• **Observer Dynamics:**

$$\begin{cases} \dot{\hat{x}}_i = \hat{x}_{i+1} + \psi_i(\hat{x}_1, \dots, \hat{x}_i, u) + \frac{\alpha_i}{\varepsilon^i} (y - \hat{x}_1) \\ \dot{\hat{x}}_\rho = \phi_0(\hat{x}, u) + \frac{\alpha_\rho}{\varepsilon^\rho} (y - \hat{x}_1) \end{cases}$$

**Target:**

Choose the constants  $\alpha_1, \dots, \alpha_\rho$  and  $\varepsilon > 0$  s.t.  $\tilde{x}(t) \rightarrow 0$

## Error Dynamics

$$\begin{aligned} \dot{\tilde{x}}_i &= \dot{x}_i - \dot{\hat{x}}_i = [x_{i+1} + \psi_i(x_1, \dots, x_i, u)] - \left[ \hat{x}_{i+1} + \psi_i(\hat{x}_1, \dots, \hat{x}_i, u) + \frac{\alpha_i}{\varepsilon^i} (y - \hat{x}_1) \right] \\ &= \tilde{x}_{i+1} + [\psi_i(x_1, \dots, x_i, u) - \psi_i(\hat{x}_1, \dots, \hat{x}_i, u)] - \frac{\alpha_i}{\varepsilon^i} \tilde{x}_1 \end{aligned}$$

It can be transformed to:

$$\underbrace{\varepsilon \frac{\dot{\tilde{x}}_i}{\varepsilon^{\rho-i}}}_{=\dot{\eta}_i} = \underbrace{\frac{\tilde{x}_{i+1}}{\varepsilon^{\rho-(i+1)}}}_{=\eta_{i+1}} + \varepsilon \underbrace{\frac{\psi_i(x_1, \dots, x_i, u) - \psi_i(\hat{x}_1, \dots, \hat{x}_i, u)}{\varepsilon^{\rho-i}}}_{=:\delta_i(x, \tilde{x}, u)} - \underbrace{\frac{\alpha_i}{\varepsilon^i \varepsilon^{\rho-(i+1)}} \tilde{x}_1}_{=\alpha_i \eta_1}.$$

Then it can be wrote to a perturbed linear system

$$\dot{\eta} = \underbrace{\frac{1}{\varepsilon} \begin{bmatrix} -\alpha_1 & 1 & & & \\ -\alpha_2 & 0 & 1 & & \\ \vdots & & 0 & \ddots & \\ -\alpha_{\rho-1} & & & \ddots & 1 \\ -\alpha_\rho & & & & 0 \end{bmatrix}}_{=:F} \eta + \delta(w, x, \tilde{x}, u), \quad \delta_\rho := \phi(w, x, u) - \phi_0(\hat{x}, u).$$

Then the matrix  $F$  can be made Hurwitz.

- When the gain  $\frac{1}{\varepsilon}$  is high, the perturbation  $\delta$  is negligible

## Theorem for High Gain Observers

**Theorem for High Gain Observers:**

Let there be **compact sets**  $W, X, U$  **such that**  $w(t) \in W, x(t) \in X$  **and**  $u(t) \in U$  for all times and initial states / inputs of interest. Assume that

1. Over the domain of interest, the functions  $f_0, \psi_1, \dots, \psi_{\rho-1}$  and  $\phi$  are locally Lipschitz in their arguments;
2. The functions  $\phi_0, \psi_1, \dots, \psi_{\rho-1}$  are Lipschitz in  $x$ , uniformly in  $u$  :

$$\exists L_0 > 0 \text{ s.t. } |\phi_0(x, u) - \phi_0(z, u)| \leq L_0 \sum_{k=1}^{\rho} |x_k - z_k|$$

$$\exists L_i > 0 \text{ s.t. } |\psi_i(x_1, \dots, x_i, u) - \psi_i(z_1, \dots, z_i, u)| \leq L_i \sum_{k=1}^i |x_k - z_k|$$

3. the matrix  $F$  is Hurwitz (see the previous slide);
4. there are constants  $L, M > 0$  such that

$$\|\phi(w, x, u) - \phi_0(z, u)\| \leq L\|x - z\| + M, \quad \forall w \in W, x, z \in X, u \in U.$$

Then, there exist constants  $a, b, c, \varepsilon^* > 0$  such that

$$0 < \varepsilon \leq \varepsilon^* \Rightarrow \|\tilde{x}^i(t)\| \leq \max \left\{ \frac{b}{\varepsilon^{i-1}} e^{-at/\varepsilon}, \varepsilon^{\rho+1-i} cM \right\}, \forall t \geq t_0.$$

#### Pros

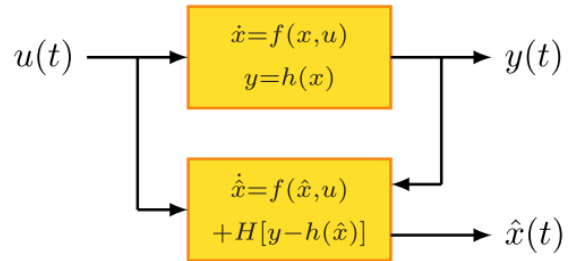
We do not have to know the exact system dynamics in order to design an observer. The function  $\phi_0$  only has to approximate  $\phi$ .

#### Cons:

- can suffer from strong initial peaks in the estimation error
- sensitive to measurement noise

## 4. Local Observer

**Goal:** Choose a **constant observer gain**  $H(t) \equiv H$  such that  $\tilde{x}(t) := x(t) - \hat{x}(t) \rightarrow 0, \quad t \rightarrow \infty$



### Linearization

- Linearize around a constant state  $x_{ss}$  and a constant input  $u_{ss}$  to get  $A$  and  $C$
- Assume that  $(A, C)$  is detectable, then we can find a  $H \in \mathbb{R}^{n \times p}$  that makes  $A - HC$  Hurwitz

## Theorem

We consider the system  $\dot{x} = f(x, u), y = h(x)$  with  $f : D \times U \rightarrow \mathbb{R}^n, \quad h : D \rightarrow \mathbb{R}^p, \quad D \subset \mathbb{R}^n, U \subset \mathbb{R}^m$  domains, with  $f, h$  **twice continuously differentiable** in  $(x, u)$ . Assume that

1.  $x_{ss} \in D$  is an **equilibrium** for a constant input  $u(t) \equiv u_{ss} \in U$  with output zero. That is,  $f(x_{ss}, u_{ss}) = 0$  and  $h(x_{ss}) = 0$ ;
2. for any  $\varepsilon > 0$  there exist constants  $\delta_1, \delta_2 > 0$  such that if

$$\|x(t_0) - x_{ss}\| \leq \delta_1, \|u(t) - u_{ss}\| \leq \delta_2, \forall t \geq t_0$$

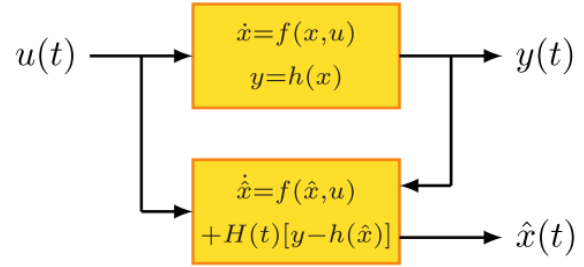
then  $\|x(t) - x_{ss}\| \leq \varepsilon$  for all  $t \geq t_0$ ;

3. the matrix  $H \in \mathbb{R}^{n \times p}$  makes  $A - HC$  Hurwitz.

Then  $\lim_{t \rightarrow \infty} \tilde{x}(t) = 0$  whenever  $\|\tilde{x}(t_0)\|, \|x(t_0) - x_{ss}\|$  and  $\sup_{t \geq t_0} \|u(t) - u_{ss}\|$  are small enough.

## 5. Extended Kalman Filter

**Goal:** Choose a **time-varying observer gain**  $H(t)$  such that  $\tilde{x}(t) := x(t) - \hat{x}(t) \rightarrow 0, \quad t \rightarrow \infty$



## Linearization

- We **linearize** around the current estimated state  $\hat{x}(t)$  and input  $u(t)$  :

$$A(t) = \left[ \frac{\partial f(\hat{x}(t), u(t))}{\partial x} \right] \text{ and } C(t) = \left[ \frac{\partial h(\hat{x}(t))}{\partial x} \right].$$

- The **Kalman gain** is  $H(t) = P(t)C^\top(t)R^{-1}$ , where  $P(t)$  is given by

$$\dot{P} = AP + PA^\top + Q - PC^\top R^{-1}CP, \quad P(t_0) = P_0,$$

and the matrices  $P_0, Q$  and  $R$  are symmetric and positive definite.

## Theorem for the Extended Kalman Filter

We consider the system  $\dot{x} = f(x, u), y = h(x)$  with

$$f : D \times U \rightarrow \mathbb{R}^n, \quad h : D \rightarrow \mathbb{R}^p, \quad D \subset \mathbb{R}^n, U \subset \mathbb{R}^m \text{ bounded}$$

Assume that

1.  $f, h$  are twice continuously differentiable in all arguments;
2. The solution of the Riccati equation exists for all  $t \geq t_0$ , and there are constants  $\alpha_1, \alpha_2 > 0$  such that

$$\alpha_1 l \leq P(t) \leq \alpha_2 l, \quad \forall t \geq t_0$$

Then there exist constants  $c, k, \lambda > 0$  such that

$$\|\tilde{x}(t_0)\| \leq c \quad \Rightarrow \quad \|\tilde{x}(t)\| \leq k e^{-\lambda(t-t_0)}, \forall t \geq t_0 \geq 0.$$

## Summary

- Observers reconstruct the **unknown state** of a system from **inputs and output**
- Observers for **specific classes** of nonlinear systems
  - Global Observers (for systems with linear state updates)
  - High-gain Observers (for systems in a normal form)
- Generic Observers based on **linearization**
  - Local Observers (constant observer gain)
  - Extended Kalman Filter (time-varying observer gain)

There are a lot of theorem in this chapter to tell us the convergence and stability of the observer. These theorem has similar structures:

- They should try to give a bounded to uncertainty/nonlinear part (such as nonlinear part in dynamic and approximation of nonlinear part)
- Part to guarantee the stability and convergence in ideal part of the system