

1_3_Controller Design With Small Delay

1. Basic Design Flow

Ideal case with $\tau=0$

Non-ideal case

2. New discrete-time model: Sampled-data Model

Signals model

Signals

States

Input Signals

Sampling Period Analysis

(Sampled-data) State-space model with $\tau < h$

Approximation

State Augmentation

3. Controller design with $\tau < h$

Model

Process

4. Alternative Design

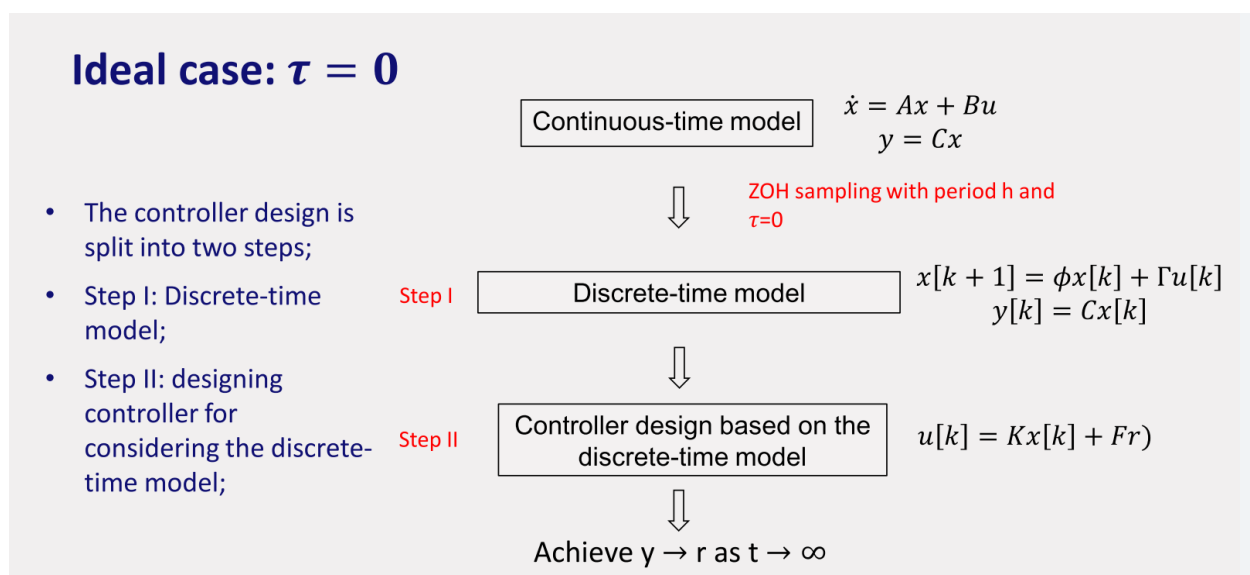
4.1. Linear Quadratic Regulator (LQR) Design

4.2. Linear Quadratic Integral (LQI) Design

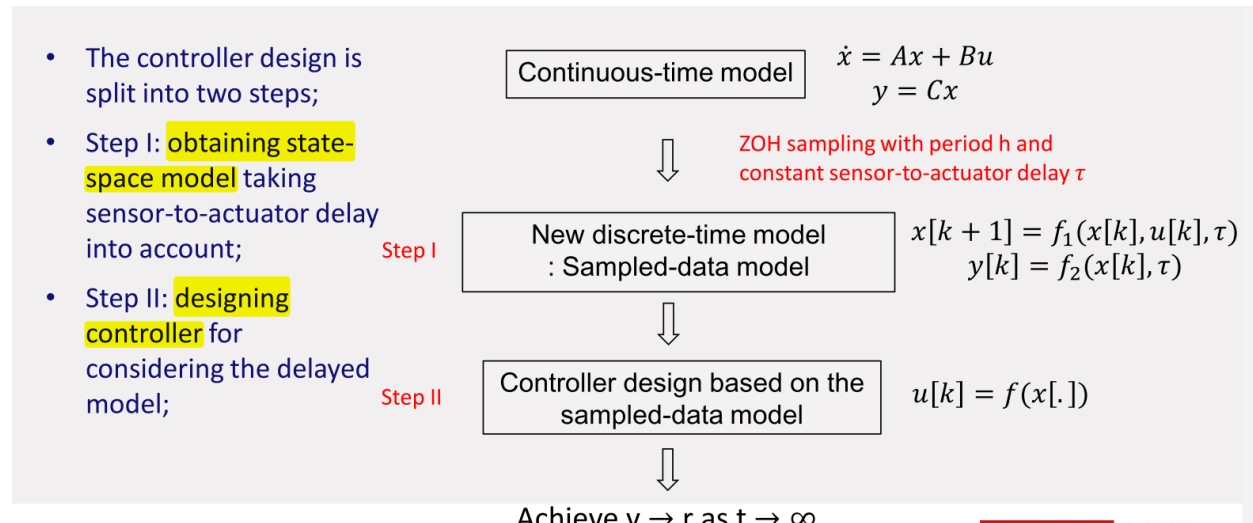
Summary

1. Basic Design Flow

Ideal case with $\tau=0$

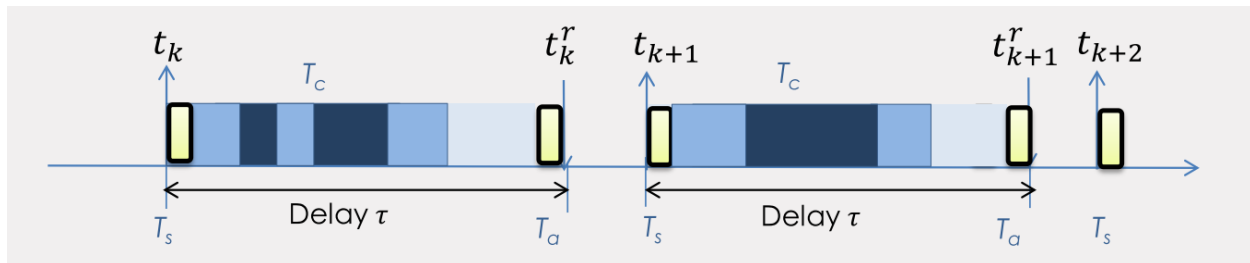


Non-ideal case



2. New discrete-time model: Sampled-data Model

Signals model



Constant sensor-to-actuator delay τ

$$t_k^r = t_k + \tau$$

$$t_{k+1}^r = t_{k+1} + \tau$$

Sampling period h

$$t_{k+1} = t_k + h$$

$$t_{k+2} = t_{k+1} + h$$

Signals

- Measurement is done in every sampling instant. So, we can assume **state unchanged** between measurements:

$$x(t) = x(t_k) = x[k], t_k \leq t \leq t_{k+1}$$

- The **input signal** is held **constant** for one sampling interval

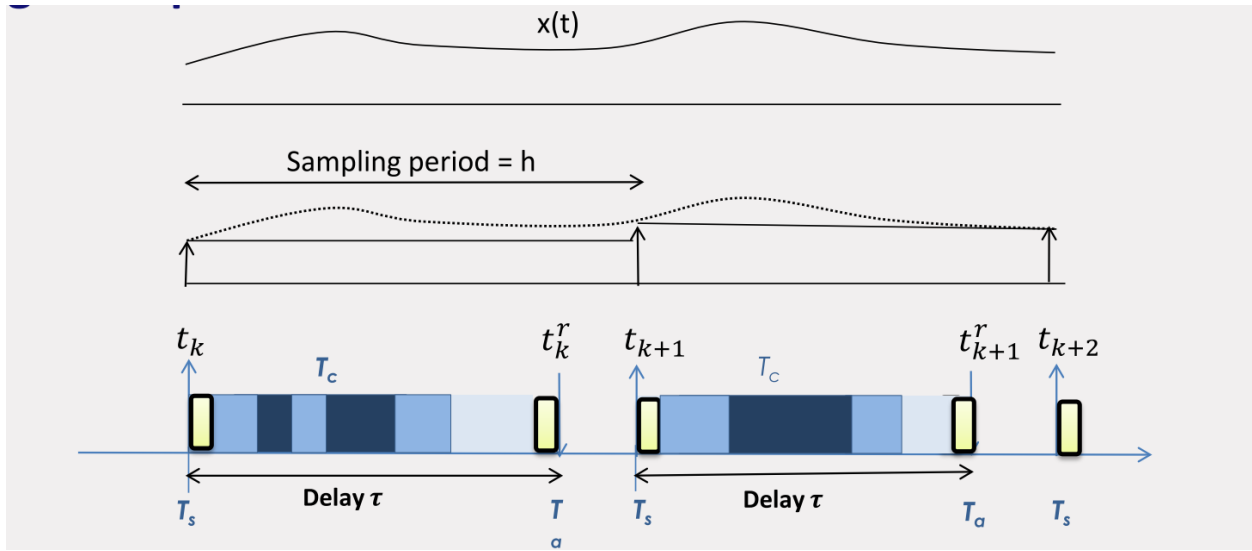
$$u(t) = u(t_k) = u[k], t_k^r \leq t \leq t_{k+1}^r$$

$$u(t) = u(t_{k+1}) = u[k+1], t_{k+1}^r \leq t \leq t_{k+2}^r$$

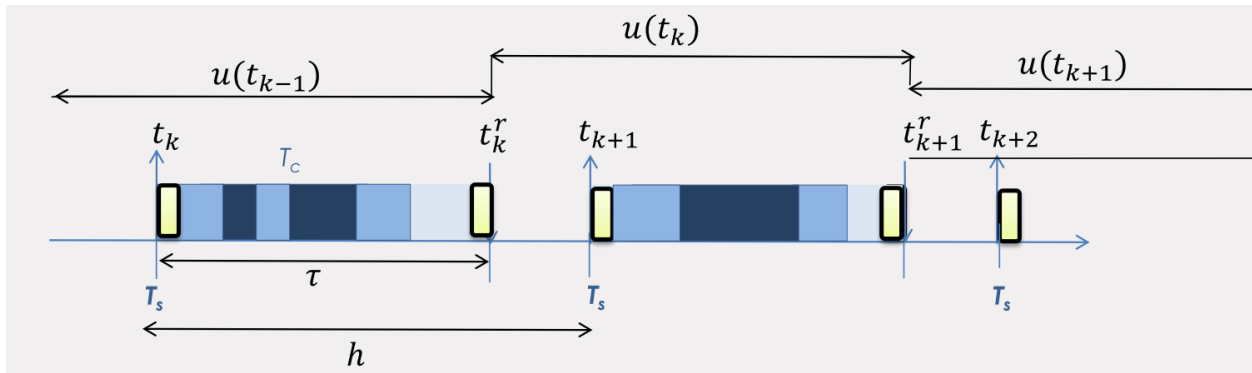
- Control input **updated once** in every sampling interval

$$t_{k+1}^r - t_k^r = h$$

States



Input Signals

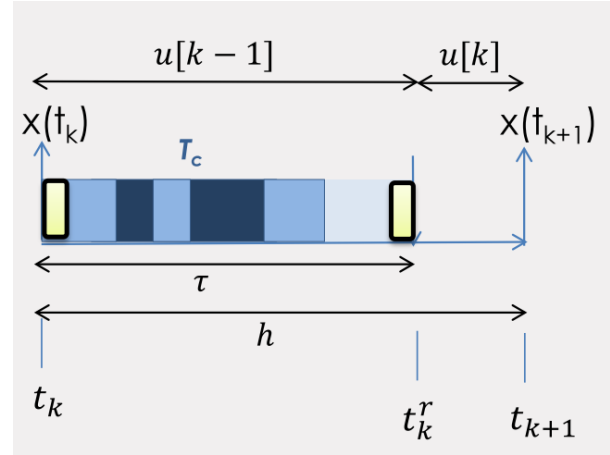


- The $u(t_k)$ is computed based on the latest measurement $x(t_k)$, i.e., $u(t_k) = f(x(t_k))$;
- $u(t_k)$ is applied at $t = (t_k + \tau) = t_k^r$
- Between $t_{k-1}^r \leq t \leq t_k^r$, the previous control input is held, i.e. $u(t) = u(t_{k-1}) = u[k-1]$

Sampling Period Analysis

For $t_k \leq t \leq t_k + \tau$, $u(t) = u[k-1]$

For $t_k + \tau < t \leq t_{k+1}$, $u(t) = u[k]$



(Sampled-data) State-space model with $\tau < h$

$$x(t_{k+1}) = e^{A(t_{k+1}-t_k)} x(t_k) + \int_{t_k}^{t_{k+1}} e^{A(t_{k+1}-\tau)} B u(\tau) d\tau$$

Considering Signals Characteristics

$$x[k+1] = e^{Ah} x[k] + \int_{t_k}^{t_k^r} e^{A(t_{k+1}-\tau)} B d\tau \cdot u[k-1] + \int_{t_k^r}^{t_{k+1}} e^{A(t_{k+1}-\tau)} B d\tau \cdot u[k]$$

It can be written as:

$$x[k+1] = \phi x[k] + \Gamma_1(\tau) u[k-1] + \Gamma_0(\tau) u[k]$$

$$\phi = e^{Ah}$$

$$\Gamma_1(\tau) = \int_{h-\tau}^h e^{As} B ds$$

$$\Gamma_0(\tau) = \int_0^{h-\tau} e^{As} B ds$$

Approximation

when τ is very “short” compared to h and A matrix needs to be invertible:

$$x[k+1] \approx \phi x[k] + \tau B u[k-1] + (h - \tau) B u[k]$$

Proof

$$\begin{aligned}
\phi &= e^{Ah} \approx I + Ah \quad (h \text{ Taylor, } h \text{ when is small}) \\
\Gamma_1(\tau) &= \int_{h-\tau}^h e^{As} B ds = A^{-1} (e^{Ah} - e^{A(h-\tau)}) B \\
&\approx A^{-1} (I + Ah - I - A(h-\tau)) B = \tau B \\
\Gamma_0(\tau) &= (h - \tau) B
\end{aligned}$$

Property

$$\begin{aligned}
&\text{when } \tau \ll h \\
&\Gamma_1(\tau) \approx 0 \\
&\Gamma_0 \approx \Gamma
\end{aligned}$$

When τ is not very short, We are not suitable to only maintain two component in Taylor Series

State Augmentation

This model has the term with old input $u[k-1]$, the state augmentation can help us refine it;

We define

$$z[k] = \begin{bmatrix} x[k] \\ u[k-1] \end{bmatrix}$$

Then we have

$$\begin{aligned}
z[k+1] &= \begin{bmatrix} \phi & \Gamma_1(\tau) \\ 0 & 0 \end{bmatrix} z[k] + \begin{bmatrix} \Gamma_0(\tau) \\ 1 \end{bmatrix} u[k] \\
y[k] &= \begin{bmatrix} C & 0 \end{bmatrix} z[k]
\end{aligned}$$

We can denote:

$$\phi_{aug} = \begin{bmatrix} \phi & \Gamma_1(\tau) \\ 0 & 0 \end{bmatrix}, \Gamma_{aug} = \begin{bmatrix} \Gamma_0(\tau) \\ 1 \end{bmatrix}, C_{aug} = \begin{bmatrix} C & 0 \end{bmatrix}$$

3. Controller design with $\tau < h$

Model

We have

$$z[k+1] = \phi_{aug} z[k] + \Gamma_{aug} u[k]; y[k] = C_{aug} z[k]$$

And we aim to design

$$u[k] = Kz[k] + Fr$$

Our goal is:

$$y[k] \rightarrow r \text{ as } k \rightarrow \infty$$

Process

1. **Check for controllability** of ϕ_{aug}, Γ_{aug} ; if they are controllable, then design K & F in the following way
2. **K** is designed using **pole placement** or any other advanced technique
3. **F** is designed using $F = \frac{1}{C_{aug}(I - \phi_{aug} - \Gamma_{aug}K)^{-1}\Gamma_{aug}}$;

4. Alternative Design

4.1. Linear Quadratic Regulator (LQR) Design

Consider system $z[k+1] = \phi_{aug}z[k] + \Gamma_{aug}u[k]$

LQR should minimize:

$$J = \sum_{k=0}^{\infty} (z^T[k]Qz[k] + u^T[k]Ru[k] + 2z^T[k]Nu[k])$$

The solution will have format $u[k] = -Kz[k]$

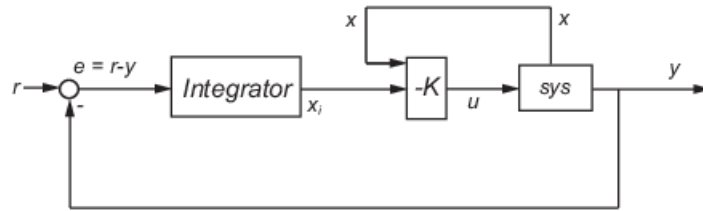
where

$$K = (\Gamma_{aug}^T P \Gamma_{aug} + R)^{-1} (\Gamma_{aug}^T P \phi_{aug} + N^T)$$

$$P = \phi_{aug}^T P \phi_{aug} - (\Gamma_{aug}^T P \phi_{aug} + N) (\Gamma_{aug}^T P \Gamma_{aug} + R^{-1} (\Gamma_{aug}^T P \phi_{aug} + N^T) + Q$$

4.2. Linear Quadratic Integral (LQI) Design

LQI design is capable of **disturbance rejection**



For system

$$\frac{dx}{dt} = Ax + Bu$$

$$y = Cx + Du$$

The state-feedback control is of the form

$$u = -K[x, x_i]$$

Minimizes the following cost functions:

- for continuous time: $J(u) = \int_0^\infty \{z^T Q z + u^T R u + 2z^T N u\} dt$
- for discrete time: $J(u) = \sum_{n=0}^\infty \{z^T Q z + u^T R u + 2z^T N u\}$

In discrete time, the LQI controller computes the **integrator output(integral tracking error)** x_i using the forward Euler formula

$$x_i[n+1] = x_i[n] + Ts(r[n] - y[n])$$

The augmented state-space is:

$$\begin{bmatrix} z[k+1] \\ x_i[k+1] \end{bmatrix} = \begin{bmatrix} \phi_{aug} & 0 \\ C_{aug} & 1 \end{bmatrix} \begin{bmatrix} z[k] \\ x_i[k] \end{bmatrix} + \begin{bmatrix} \Gamma_{aug} \\ 0 \end{bmatrix} u[k]$$

$$y[k] = \begin{bmatrix} C & 0 \end{bmatrix} x[k]$$

Summary

- Consideration when design controller for system with τ and h : mainly for $\tau < h$
 - augmented state-space
 - approximation when $\tau \ll h$
- Three ways of designing controller:
 - pole placement
 - LQR controller
 - LQI controller