

Maintained Zones in CSTR with Recycle

Eskild Schroll-Fleischer

September 27, 2020

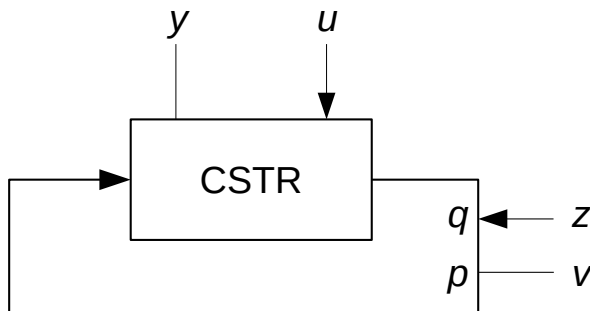


Figure 1: CSTR with PFR recycle. Two measurements y and v and two manipulated variables u and z .

Consider a CSTR of volume V with a recycle through a PFR of length L and cross-sectional area A at constant superficial velocity v , see figure 1. Measurement points are placed inside the CSTR and at the beginning of the PFR. Inlets are placed in the CSTR and just before the measurement point in the PFR. The objective is to simultaneously maintain one measurement value y in the CSTR by manipulating u and another measurement value v in the PFR by manipulating z . The measurement could for example be a concentration of a component or a temperature.

Point v is placed at axial position pL relative to the outlet to the PFR and correspondingly z is placed at axial position qL . In this terminology $0 < p < q < 1$. The residence time in the CSTR is $\tau_c = V(vA)^{-1}$ and the residence time in the PFR is $\tau_p = Lv^{-1}$.

The sensor in the PFR behaves as a first order system even though plug flow conditions prevail in the PFR. The time constant of this first order behaviour is τ_s .

DDE Model

The system is modeled by equations (1) to (4) where one unit of time t corresponds to one residence time τ_p in the PFR.

$$\frac{dx_1(t)}{dt} = \tau_c^{-1} (u(t) + x_2(t-p) - x_1(t)), \text{ for } t > 0, x_1(t \leq 0) = 0 \quad (1)$$

$$\frac{dx_2(t)}{dt} = \tau_s^{-1} (z(t-(q-p)) + x_1(t-(1-p)) - x_2(t)), \text{ for } t > 0, x_2(t \leq 0) = 0 \quad (2)$$

The measurements are then $y = x_1$ and $v = x_2$. Both u and z are PI-controllers:

$$u(t) = K_{p,u} \left(-x_1(t) + \tau_{I,u}^{-1} I_u \right), \quad \frac{dI_u(t)}{dt} = -x_1(t), \quad \text{for } t > 0, \quad I_u(t \leq 0) = 0 \quad (3)$$

$$z(t) = K_{p,z} \left((x_{\text{PFR}} - x_2(t)) + \tau_{I,z}^{-1} I_z \right), \quad \frac{dI_z(t)}{dt} = x_{\text{PFR}} - x_2(t), \quad \text{for } t > 0, \quad I_z(t \leq 0) = 0 \quad (4)$$

It is desired to move the time dependence of z from equation (2) to equation (4) to ease implementation of the model in MATLAB® R2018a. This is accomplished as follows:

$$\frac{dx_2(t)}{dt} = \tau_s^{-1} (z(t) + x_1(t - (1 - p)) - x_2(t)), \quad \text{for } t > 0, \quad x_2(t \leq 0) = 0 \quad (5)$$

$$z(t) = K_{p,z} \left((x_{\text{PFR}} - x_2(t - (q - p))) + \tau_{I,z}^{-1} I_z \right), \quad \frac{dI_z(t)}{dt} = x_{\text{PFR}} - x_2(t - (q - p)) \quad (6)$$

The points in history which are required are then: $t - p$, $t - (q - p)$ and $t - (1 - p)$.

Controller tuning

The PI-controllers are tuned according to the Skogestad SIMC rules which dictate that for a unit gain process with time constant τ and time delay θ :

$$K_p = \frac{\tau}{\theta + \tau_c} \quad \text{and} \quad \tau_I = \min(\tau, 4(\theta + \tau_c)) \quad \text{for} \quad \tau_c = \frac{\tau}{5} + \theta \quad (7)$$

The time constant in the CSTR is τ_c and for the PFR τ_s . There is a time delay in the PFR and this amounts to $(q - p)$ which is the temporal distance between manipulation and measurement.

Simulation

The DDE model in equations (1), (3), (5) and (6) is implemented in MATLAB® R2018a, see appendix A.

A simulation of a closed loop experiment is presented in figure 2. The simulation shows that the two zones are maintained.

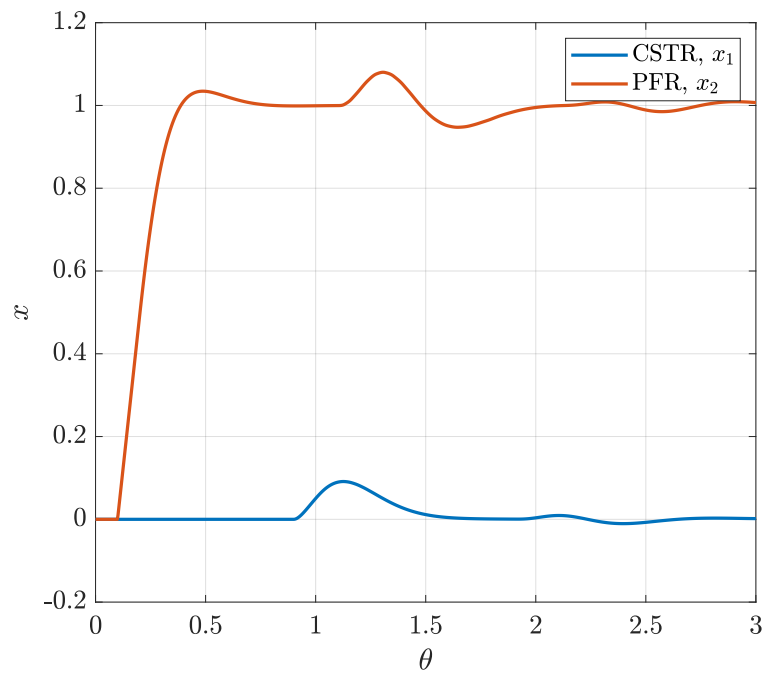


Figure 2: Simulation for $\tau_c = 0.2$, $\tau_s = 0.02$, $q = 0.9$, $p = 0.8$ and initial history $x_1(t \leq 0) = x_2(t \leq 0) = 0$.

A maintained_zones.m

```
1 set(0,'defaultlinewidth',1.7,'DefaultAxesFontSize',14,'DefaultAxesXGrid','on','DefaultAxesYGrid','
2 on','defaultTextInterpreter','latex','DefaultFigurePosition',[0 0 600 500])
3 set(groot,'defaultAxesTickLabelInterpreter','latex','defaultLegendInterpreter','latex');
4 clear;clc;close all;
5 %% Parameters
6 P.tau_cstr = 0.2; % [theta], time constant of CSTR
7 P.tau_pfr = 0.02; % [theta], time constant of sensor in PFR
8 P.p = 0.8; % Distance from end of PFR to measured variable
9 P.q = 0.9; % Distance from end of PFR to manipulated variable
10 tau_c_cstr = P.tau_cstr/5;
11 P.Kpu = P.tau_cstr/tau_c_cstr; % CSTR controller gain
12 P.tauIu = min(P.tau_cstr,4*tau_c_cstr); % [theta], CSTR controller integral time
13 tau_c_pfr = P.tau_pfr/5+(P.q-P.p);
14 P.Kpz = P.tau_pfr/((P.q-P.p)+tau_c_pfr); % PFR controller gain
15 P.tauIz = min(P.tau_pfr,4*((P.q-P.p)+tau_c_pfr)); % [theta], PFR controller integral time
16 n_theta = 3; % Simulation duration [theta]
17 %% Solve DDE model
18 sol = dde23(@model_dde,[P.p, (1-P.p), (P.q-P.p)],@model_dde_history,[0, n_theta],[],P);
19 % Extract solution
20 theta = linspace(0,n_theta,1000);
21 y = deval(sol, theta);
22 x1 = y(1,:);
23 x2 = y(2,:);
24 %% Plot of solutions side by side
25 figure
26 plot(theta,x1)
27 hold on
28 plot(theta,x2)
29 ylabel('$x_1$')
30 xlabel('$\theta$')
31 legend('CSTR, $x_1$', 'PFR, $x_2$')
32 %print('plot','-dpdf')
33 %% DDE history function
34 function [x] = model_dde_history(t,P)
35     x = zeros(4,1);
36 end
37 %% DDE model function
38 function [dx] = model_dde(t,x,xh,P)
39     % Pick relevant states
40     x1 = x(1);
41     x2 = x(2);
42     Iu = x(3);
43     Iz = x(4);
44     x2ptheta = xh(2,1); % x2(t-p)
45     x1lptheta = xh(1,2); % x1(t-(1-p))
46     x2qptheta = xh(2,3); % x2(t-(q-p))
47     % Integral states are calculated:
48     dIudt = -x1;
49     if t > (P.q-P.p)
50         dIzdt = 1-x2qptheta;
51     else
52         dIzdt = 0;
53     end
54     % Control action
55     u = P.Kpu*(dIudt + Iu/P.tauIu);
56     z = P.Kpz*(dIzdt + Iz/P.tauIz);
57     % DDEs
58     dx1dt = (u+x2ptheta-x1)/P.tau_cstr;
59     dx2dt = (z+x1lptheta-x2)/P.tau_pfr;
60     dx = [dx1dt; dx2dt; dIudt; dIzdt];
61 end
```