Abstract

In this work the stabilization and control of delayed recycling systems is addressed. Recycling systems are characterized by possessing to main paths named the direct (feedforward) and the recycling (feedback) paths. Such class of systems reuse the energy and/or the partially processed matter increasing the efficiency of the overall process. It is proposed a control methodology for the stabilization and control of this kind of systems. The particular class of systems addressed here contains one unstable pole, \(m\) stable poles, a delay term and possible \(p\) left half plane zeros (LHP) in the direct path and a delayed stable subsystem in the recycling path. The strategy is based on an asymptotic observer-predictor to estimate the required internal signals. Necessary and sufficient conditions are stated in order to guarantee the stabilization of the proposed scheme, achieving step tracking and step disturbance rejection.

1 INTRODUCTION

Recycling systems are commonly found in chemical industry, for instance, in a typical plant formed by reactor/separator process, where reactants are recycled back to the reactor [1]. They reuse the energy and the partially processed matter increasing the efficiency of the overall process. In recycling systems a partial feedback of the process output to the input induces a positive feedback, which can give rise to some undesirable effects. Luyben [3] studied the effects of recycle path on dynamics process and their implications to plant-wide control. Taiwo [4], discussed the robust control for recycling plants and proposed the concept of recycle compensation to recuperate inherent process dynamics, i.e. dynamics without recycle. Scali and Ferrari [5] analyzed the problem under same idea. Similar approaches were extended by Lakshminarayanan and Takada [6], and Kwok et. al [7].
On the other hand, stability analysis of systems with time delay (and without recycle) together simple controllers is not an easy task, see for instance [8]. Therefore, when significant transport delay is present in recycling systems the control problem becomes more involved. It is known that when recycle path and time delays occur, exponential terms appear in the direct and the recycling paths. In the state space recycling systems with time delay correspond to systems with delays in the input and the state variables. Model approximations have been proposed to remove the exponential terms from the transfer function denominator of a delayed system, such as the method of moments [9] or Pade approximations [10]. Other techniques, such as Taylor series expansion [11], and the seasonal time-series model [7], have been proposed to obtain an approximate model to represent recycling systems. In [12] an approximated model to represent recycling systems by using discrete-time approach is proposed. In turn, such approximated models can be used for stability analysis or control design [13], [5], [14], [15]. A system with time delay and open-loop unstable poles is notably more difficult to control than a system with only open-loop stable poles. For instance, the classical Smith Predictor cannot be used in the case of delayed unstable processes. Introducing recycle in such system would lead to a more difficult (although interesting) problem. Existing modified schemes to Smith Predictor cannot directly be applied to this kind of systems. This is, because the open-loop recycling system is not a system with only time-delay in the direct path but it is a system with internal delay term, i.e., an open loop infinite dimensional system. In order to tackle this problem, we consider here recycling systems composed of a plant with one unstable pole, $m$ stable poles, a delay term and $p$ possible left half plane zeros (LHP) in the direct path and a delayed stable subsystem of order $n$ in the recycling path. The present work intends to extend the results presented in [16] and [17] where recycling systems restricted to unstable first order plants in the direct path are considered. A first attempt to extend these results was presented in [18], dealing with high order plant in the direct path with one unstable pole but without including zeros and without achieving step tracking or step disturbance rejection.

In this work, an observer-predictor based control scheme is proposed in order to stabilize and control recycling systems with delay term in both paths, a stable system in the recycling path and a high order system with one unstable and $m$ stable poles in the direct path. The proposed scheme allows achieving the step tracking and step disturbance rejection. Then, a modification to the scheme is proposed in order to deal with a more general family of plants including $p$ LHP zeros in the direct path. Necessary and sufficient conditions are given for the existence of the proposed stabilizing scheme. It is important to note that, the problem of the stabilization and control of delayed unstable plants even without recycle path is not completely solved. For instance, recent works [21],[22] deal with the stabilization and control of delayed systems with only one unstable pole.

The outline of the paper is as follows. In Section 2 the problem is formulated and the class of systems considered in this work is precised. The general idea of the solutions is also outlined in this section, namely the need of an
observer-predictor. Section 3 some preliminary stability results are presented. The Section 4 presents the main results. This section is divided in three parts. Firstly in the Subsection 4.1 an observer-predictor scheme to estimate some internal signals of the system is proposed. Then Subsection 4.2 presents the proposed control approach and in the Subsection 4.3, stabilization of recycling systems by considering a high order with an unstable pole and LHP zeros in the direct path is provided. Some simulation results are described in Section 5. Such results illustrate the performance of the proposed control strategy. Finally Section 6 presents some conclusions.

2 PROBLEM FORMULATION

Consider the class of recycling system shown in Fig. 1, which can be described as,

\[ Y(s) = \frac{[G_d(s) \ G_r(s)]}{Y(s)} U(s) \]

with,

\[ G_d(s) = G_1(s)e^{-\tau_1 s} = \frac{\alpha}{(s-a)(s+b_1)...(s+b_m)}e^{-\tau_1 s}, \quad (2a) \]
\[ G_r(s) = G_2(s)e^{-\tau_2 s} = \frac{N(s)}{D(s)}e^{-\tau_2 s}, \quad (2b) \]

where \( G_d(s) \), and \( G_r(s) \) are transfer functions of the direct and the recycling paths, respectively; \( \tau_1, \tau_2 \geq 0 \) are the time delays associated to \( G_d(s) \), and \( G_r(s) \). \( a, b_i \in \mathbb{R}^+ \) with \( i = 1, 2, ..., m \) i.e., \( G_d(s) \) is unstable. \( N(s) \) and \( D(s) \) are polynomials on the complex variable \( s \), with the roots of \( D(s) \) in LHP, i.e., \( G_r(s) \) is stable. \( U(s) \) is the process input and \( Y(s) \) is the process output.

![Figure 1: A process with recycle](image)

The transfer function of the recycling system (1) is given by,

\[ G_t(s) = \frac{Y(s)}{U(s)} = \frac{D(s)\alpha e^{-\tau_1 s}}{H_{aux}(s)D(s) - \alpha N(s)e^{-(\tau_1+\tau_2)s}}. \quad (3) \]

where \( H_{aux}(s) = (s-a)(s+b_1)...(s+b_m) \). Note that exponential terms appear explicitly in the numerator and the denominator of \( G_t(s) \). Stability of (3) is determined by the roots of its characteristic quasi-polynomial,

\[ Q(s) = H_{aux}(s)D(s) - \alpha N(s)e^{-(\tau_1+\tau_2)s}. \quad (4) \]
Note that the transcendental term in $Q(s)$ induces an infinite number of roots. Then, for this kind of plants it is not an easy task to conclude about the dynamical behaviour (stability for instance) even in the uncontrolled plant case. Obviously, the related transfer function when the system is controlled with an output feedback becomes more complicated involving more than one transcendental term.

Let us to describe some ideas behind the proposed methodology. With reference to Fig. 1, if the signal $\omega_2(s)$ is measured, then we could set,

$$U(s) = R_1(s) - \omega_2(s),$$  \hspace{1cm} (5)

obtaining the system shown in Fig. 2. Then it would be possible to design $R_1(s)$ as $R_1(s) = J(s)(R(s) - \omega_1(s))$ as in Fig. 3, where $R(s)$ is the input reference. Since $\omega_1(s)$ and $\omega_2(s)$ are unmeasured internal system signals, an observer-predictor scheme to estimate these variables is developed in the Section 4. In the following Section some preliminary stability results are presented.

$$G(s) = \frac{Y(s)}{U(s)} = G_{df} e^{-s}$$  \hspace{1cm} (6)

where

$$G_{df} = \gamma \frac{1}{(s-a)(s+b_1)(s+b_2)...(s+b_m)}$$

$a, b_i \in \mathbb{R}$, with $a, b_i > 0 \ \forall i = 1, 2, ..., m; \frac{1}{\gamma} - \sum_{i=1}^{m} \frac{1}{b_i} > 0$ and the static control law,

$$U(s) = [R(s) - Y(s)]k,$$  \hspace{1cm} (7)
with \( R(s) \) as the new input reference.

The following result states the stability condition when a proportional controller is used for unstable delayed systems including one unstable pole, \( m \) stable poles and a time delay.

**Lemma 1** [20] [21] There exists a gain \( k \) such that the closed loop system,

\[
Y(s) = \frac{k\gamma e^{-\tau s}}{R(s)} = \frac{k\gamma e^{-\tau s}}{(s - a)(s + b_1)(s + b_2)\ldots(s + b_m) + k\gamma e^{-\tau s}},
\]

is stable if and only if

\[
\tau < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i}.
\]

From condition (9) the size of the delay in order to achieve the closed-loop stability is restricted by the position of the unstable pole and the stable poles in the real axis of the plane \( s \).

In the following procedure the set of stabilizing gains \( k \) related to Lemma 1 is obtained. First the phase crossover frequency \( \omega_c \) is given by,

\[
-\tau \omega_c + \arctan \left( \frac{1}{a \omega_c} \right) - \sum_{i=1}^{m} \arctan \left( \frac{1}{b_i \omega_c} \right) = 0,
\]

where \( \omega_c > 0 \). Then, the stabilizing gain \( k \) is within the range of,

\[
\frac{a}{\gamma} \prod_{i=1}^{m} b_i < k < \frac{a}{\gamma} \prod_{i=1}^{m} b_i M
\]

with,

\[
M = \sqrt{\left( 1 + \left( \frac{1}{a} \right)^2 \omega_c^2 \right) \prod_{i=1}^{m} \left( 1 + \left( \frac{1}{b_i} \right)^2 \omega_c^2 \right)}.
\]

On the other hand, consider the transfer function given by,

\[
G(s) = \frac{\alpha(s + z_1)\ldots(s + z_p)}{(s - a)(s + b_1)\ldots(s + b_m)e^{-\tau_1 s}},
\]

with \( p \leq m \) and \( z_1, \ldots, z_p > 0 \), i.e., \( p \) left half plane (LHP) zeros in the direct path of the recycling system and a partition of \( G(s) \) as follows,

\[
G(s) = G_a(s)e^{-\tau_1 s}G_b(s)
\]

where,

\[
G_a(s) = \frac{\alpha}{(s - a)(s + b_1)\ldots(s + b_{m-p})},
\]

\[
G_b(s) = \frac{(s + z_1)\ldots(s + z_p)}{(s + b_{m-p+1})\ldots(s + b_m)},
\]

and a state space representation of \( G_b(s) \) as \((A_b, B_b, C_b, D_b)\). The following result states the stability condition for the output injection shown in Fig. 4.
Lemma 2 Consider the system given by (12), the partition (13) and the output injection scheme shown in Fig. 4. There exists $L$ and $k$ such that the closed-loop system is BIBO (Bounded Input-Bounded Output) stable if and only if

$$\tau < \frac{1}{a} - \sum_{i=1}^{m-p} \frac{1}{b_i}.$$  

Proof. Consider the partial output injection shown in Fig. 4 by means of the vector $L$. Then, $L$ should be computed such that,

$$\sigma(sI - A_b + LC_b) = \{-z_1, -z_2, \ldots, -z_p\},$$  

where $\sigma(.)$ represents the eigenvalues of, and $-z_1, -z_2, \ldots, -z_p$ are the zeros of $G_b(s)$. Notice that the vector $L$ satisfying (15) involves zero-pole cancellations. Thus, the cancellations are valid since they take place in the left half plane (LHP). With this, we see that the output injection shown in Fig. 4 is reduced to a system of the form (6)-(7), where its stability properties are analyzed in Lemma 1. Hence the result follows.

Lemma 3 Consider the delay free model $G_{df}$ in (6) and a PI/PID controller with a unitary output feedback. Then there exists a stabilizing controller if and only if

$$0 < \frac{1}{a} - \sum_{i=1}^{m-p} \frac{1}{b_i}$$  

for PI, or

$$0 < \sqrt{\frac{m}{a} + \sum_{i=1}^m \frac{1}{b_i} + \frac{1}{a} - \sum_{i=1}^m \frac{1}{b_i}}$$  

for PID controller.

Proof. This result is a particular case of the Theorem 1/2, presented in Lee et. al [21], by considering the delay term equal to zero.

4 MAIN RESULTS

4.1 Prediction Strategy

In order to estimate $\omega_1(s)$ and $\omega_2(s)$ in Fig. 1 we propose the observer-predictor depicted in Fig. 5. Its convergence is established in the following result.

Theorem 4 Consider the observer-predictor scheme shown in Fig. 5, assuming $G_r(s)$ is a stable transfer function and $G_d(s)$ defined as in (2a). There exists a constant $k$ such that

$$\lim_{t \to \infty} [\omega_j - \hat{\omega}_j] = 0,$$  

(16)
for $j = 1, 2, \ldots$, if and only if

$$\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i},$$

for $i = 1, 2, \ldots, m$.

**Proof.** A state space representation of the observer-predictor scheme shown in Fig. 5 is

$$\dot{x}(t) = Ax(t) + A_1 x(t - \tau_1) + A_2 x(t - \tau_2) + Bu(t) \quad (17)$$

$$y(t) = Cx(t - \tau_1) \quad (18)$$

with, $x(t) = \begin{bmatrix} x_d(t) & x_r(t) & \hat{x}_d(t) & \hat{x}_r(t) \end{bmatrix}^T$, $y(t) = \begin{bmatrix} y(t) & \hat{y}(t) \end{bmatrix}^T$, $B = \begin{bmatrix} B_d & 0 & 0 \end{bmatrix}^T$,

\[ A = \begin{bmatrix} A_d & 0 & 0 & 0 \\ 0 & A_r & 0 & 0 \\ 0 & 0 & A_d & 0 \\ 0 & 0 & 0 & A_r \end{bmatrix}, \]

\[ A_1 = \begin{bmatrix} B_r C_d & 0 & 0 & 0 \\ 0 & B_r k C_d & 0 & -B_d k C_d \\ 0 & B_d C_r & 0 & 0 \\ 0 & 0 & B_d C_r & 0 \end{bmatrix}, \]

\[ A_2 = \begin{bmatrix} 0 & B_d C_r & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & B_d C_r \\ 0 & 0 & 0 & 0 \end{bmatrix}, \]

\[ C = \begin{bmatrix} C_d & 0 & 0 & 0 \\ 0 & 0 & C_d & 0 \end{bmatrix}. \]

where $x \in \mathbb{R}^p$ ($p = n + m + 1$) is the state vector, $u \in \mathbb{R}$ is the input, $y \in \mathbb{R}^2$ is the output, $\tau_1 \geq 0$ and $\tau_2 \geq 0$ are the time delays in the system. $A_d \in \mathbb{R}^{(n+m+1) \times (n+m+1)}$.
\( \mathbb{R}^{(m+1) \times (m+1)} \), \( B_d \in \mathbb{R}^{(m+1) \times 1} \), and \( C_d \in \mathbb{R}^{1 \times (m+1)} \) are matrices and vectors parameters that correspond to the direct path in the process \((G_1(s))\), given by,

\[
A_d = \begin{bmatrix}
    a & 0 & 0 & \cdots & 0 \\
    1 & -b_1 & 0 & \cdots & 0 \\
    0 & 1 & -b_1 & \cdots & 0 \\
    \vdots & \vdots & \vdots & \ddots & \vdots \\
    0 & 0 & 0 & \cdots & -b_m
\end{bmatrix},
\]

\[
B_d = \begin{bmatrix}
    \alpha \\
    0 \\
    0 \\
    \vdots \\
    0
\end{bmatrix},
C_d = \begin{bmatrix}
    0 & 0 & 0 & \cdots & 1
\end{bmatrix},
\]

and \( A_r \in \mathbb{R}^{n \times n} \), \( B_r \in \mathbb{R}^{n \times 1} \), and \( C_r \in \mathbb{R}^{1 \times n} \) are matrices and vectors parameters that corresponds to the recycling path in the process, \( \hat{x}(t) \) is the estimation of \( x(t) \).

From Fig. 5 we are strictly interested in the convergence of \( e_{\omega_i} = \hat{\omega}_i - \omega_i \) (for \( i = 1, 2 \)). However, as \( \omega_1(s) \) and \( \omega_2(s) \) are a static functions of the states \( x_d(t) \) and \( x_r(t) \), we can equivalently analyze the convergence of \( e_{x_d}(t) \) and \( e_{x_r}(t) \) defined in what follows. Defining the state prediction errors \( e_{x_d}(t) = \hat{x}_d(t) - x_d(t) \), \( e_{x_r}(t) = \hat{x}_r(t) - x_r(t) \), and the output estimation \( e_y(t) = \hat{y}(t) - y(t) \), it is possible to describe the behavior of the error signals as,

\[
\begin{bmatrix}
    \dot{e}_{x_d}(t) \\
    \dot{e}_{x_r}(t) \\
    e_y(t + \tau_1) \\
    e_{\omega_2}(t + \tau_2)
\end{bmatrix} = A_p \begin{bmatrix}
    e_{x_d}(t) \\
    e_{x_r}(t) \\
    e_y(t) \\
    e_{\omega_2}(t)
\end{bmatrix},
\]

with,

\[
A_p = \begin{bmatrix}
    A_d & 0 & -B_d k & B_d \\
    0 & A_r & 0 & 0 \\
    C_d & 0 & 0 & 0 \\
    0 & C_r & 0 & 0
\end{bmatrix}.
\]

Note that \( e_y(t) = C_d e_{x_d}(t - \tau_1) \) and that \( e_{\omega_2}(t) = C_r e_{x_r}(t - \tau_2) \). Then, system (19) can be rewritten as,

\[
\begin{align*}
    \dot{e}_{x_d}(t) &= A_d e_{x_d}(t) - B_d k C_d e_{x_d}(t - \tau_1) + B_d C_r e_{x_r}(t - \tau_2), \\
    \dot{e}_{x_r}(t) &= A_r e_{x_r}(t).
\end{align*}
\]

As \( A_r \) is a Hurwitz matrix, the stability of system (20) can be analyzed by considering the reduced dynamics

\[
\dot{e}_{x_d}(t) = A_d e_{x_d}(t) - B_d k C_d e_{x_d}(t - \tau_1),
\]

(21)
or equivalently,
\[
\begin{bmatrix}
\dot{e}_x(t) \\
\dot{e}_y(t + \tau_1)
\end{bmatrix} =
\begin{bmatrix}
A_d & -B_d k \\
C_d & 0
\end{bmatrix}
\begin{bmatrix}
e_x(t) \\
e_y(t)
\end{bmatrix}.
\tag{22}
\]

Consider now a state space realization of system (6). This dynamics can be written in state space form as,
\[
\begin{bmatrix}
\dot{x}(t) \\
y(t - \tau)
\end{bmatrix} =
\begin{bmatrix}
A & -Bk \\
C & 0
\end{bmatrix}
\begin{bmatrix}
x(t) \\
y(t)
\end{bmatrix}
+ \begin{bmatrix} B \\ 0 \end{bmatrix} u(t)
\tag{23}
\]

with,
\[
A = \begin{bmatrix}
a & 0 & 0 & \cdots & 0 \\
1 & -b_1 & 0 & \cdots & 0 \\
0 & 1 & -b_1 & \cdots & 0 \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
0 & 0 & 0 & \cdots & -b_m
\end{bmatrix}, B = \begin{bmatrix} \gamma \\ 0 \end{bmatrix}
\]
\[
C = \begin{bmatrix}
0 & 0 & 0 & \cdots & 1
\end{bmatrix}
\]

Comparing (23) and (22) it is clear that Lemma 1 can be applied to system (22). Hence the result of the theorem follows.

4.2 Proposed Control Scheme

Using the signals \(\hat{\omega}_1(s)\) and \(\hat{\omega}_2(s)\) from the proposed observer-predictor scheme, the ideas depicted in Section 2 can be implemented, i.e., \(\hat{\omega}_2(s)\) is used in order to decouple the recycling path while \(\hat{\omega}_1(s)\) can be used to design a controller \(J(s)\) that assures the closed-loop stability.

In order to achieve the above purposes, the general control law can be set as,
\[
U(s) = J(s)(R(s) - \hat{\omega}_1(s)) - \hat{\omega}_2(s).
\tag{24}
\]

This is illustrated in Fig. 6. Notice that the controller \(J(s)\) should be designed based on the delay free model \(G_1(s)\) due to the use of the estimated signal \(\omega_1(s)\).

**Remark 5** Notice that any controller \(J(s)\) that stabilizes the delay free model \(G_1(s)\) with a simple unitary feedback allows stabilizing the closed-loop system shown in Fig. 6 since the convergence of the signals \(\omega_1(s)\) and \(\omega_2(s)\) has been assured.

On the other hand, from the design controller \((J(s))\) viewpoint there is not restriction on the magnitude of the delay term \(\tau_1\), however the estimation of the variable \(\omega_1(s)\) is possible when \(\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i}\) (Theorem 4), therefore this is the only restriction on \(\tau_1\).

The previous ideas can be summarized in the following result.
Theorem 6 Consider the observer-based control scheme depicted in Figure 6. Then, there exists a controller $J(s)$, PI/PID, such that the closed-loop system is stable if and only if

$$
\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i},
$$

for $i = 1, 2, \ldots, m$.

Proof. Consider the observer scheme shown in Figure 6. From Theorem 4, an adequate estimation $\hat{\omega}_j(s)$, of the signals $\omega_j$ are assured if and only if

$$
\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i},
$$

for $i = 1, 2, \ldots, m$. On the other hand, by Lemma 3, it is possible to find a controller, PI/PID, such that the closed-loop system is stable if and only if

$$
0 < \frac{1}{a} - \sum_{i=1}^{m-p} \frac{1}{b_i} \quad \text{for PI}, \quad \text{or} \quad 0 < \sqrt{\frac{1}{b^2} + \sum_{i=1}^{m} \frac{1}{b_i}^2 + \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i}} \quad \text{for PID controller}.
$$

Then we can conclude that the closed-loop system of Figure 6 is stable if and only if

$$
\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i}.
$$

Notice that the observer and the control are designed separately then the closed-loop system (Figure 6) with the dynamic measurement feedback is stable. Such assertion is valid since the separation principle holds for the proposed linear observer based control. In this case the control and observer systems are stable, satisfying the above mentioned condition.

The following result is concerned with the tracking and disturbance rejection problem for step inputs by considering a PI or PID controller. Here it is assumed that the used PI/PID controller stabilizes the delay free model $G_1(s)$ (see Theorem 6). In this way, the general control law (24) is modified as,

$$
U(s) = J(s)(R(s) - \hat{\omega}_1(s) + E_y(s)) - \hat{\omega}_2(s) \quad (25)
$$

where $E_y(t) = \hat{y}(t) - y(t)$ and $J(s)$ is a PI/PID controller.
Corollary 7 Consider the observer scheme shown in Fig. 7. Then, if the condition of Theorem 6 is satisfied the tracking of step input references and rejecting step disturbances properties are achieved i.e.,

$$\lim_{t \to \infty} y(t) = \rho,$$

where $R(s) = \rho/s$ is the step input reference and $T(s) = 0$, and

$$\lim_{t \to \infty} y(t) = 0,$$

where $R(s) = 0$ and $T(s)$ is a step input disturbance.

Proof. Consider the observer scheme shown in Fig. 7 and the output response expressed as,

$$Y(s) = G_{in}R(s) + G_{pert}(s)T(s)$$

where, $G_{in}$ and $G_{pert}$ are transfer functions of the output with respect to the input reference and disturbance input, respectively. First, the tracking step reference property is addressed. Thus, consider

$$G_{in} = \frac{G_dJ(1 + kG_1e^{-s\tau_1})}{J(G_d + G_1(1 - e^{-s\tau_1} + kG_d)) + kG_1e^{-s\tau_1} + 1}$$

The stability of the transfer function (28) can be assured by satisfying the condition of Theorem 6, i.e.,

$$\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i}.$$

Therefore the Final Value Theorem with $R(s) = \rho/s$ to the transfer function (28) can be applied, obtaining,

$$\lim_{t \to \infty} y(t) = \lim_{s \to 0} sY(s) = \lim_{s \to 0} s \frac{N_1(s)}{D_1(s)} \frac{\rho}{s}$$
with,

\[ N_1(s) = G_dJ(1 + kG_1 e^{-s\tau_1}) \]  \hspace{1cm} (30)

\[ D_1(s) = J(G_d + G_1(1 - e^{-s\tau_1} + kG_d)) + kG_1 e^{-s\tau_1} + 1 \]  \hspace{1cm} (31)

Substituting \( G_d(s), \ G_1(s) \) given in (2a) and \( J(s) \) as a PI or PID controller, the result follows

\[ \lim_{t \to \infty} y(t) = \lim_{s \to 0} sY(s) = \rho \]  \hspace{1cm} (32)

Now the disturbance rejection property is analyzed. Consider

\[ G_{\text{pert}} = \frac{G_d + G_1 JG_d + kG_1 G_d e^{-s\tau_1} - G_1 JG_d e^{-s\tau_1}}{J(G_d + G_1(1 - e^{-s\tau_1} + kG_d)) + kG_1 e^{-s\tau_1} + 1} \]  \hspace{1cm} (33)

The stability of the transfer function (33) can be assured by satisfying the condition of Theorem 6, i.e.,

\[ \tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i}. \]

Therefore the Final Value Theorem with \( T(s) = \eta/s \), to the transfer function (33) can be applied, obtaining,

\[ \lim_{t \to \infty} y(t) = \lim_{s \to 0} sY(s) = \lim_{s \to 0} s \frac{N_2(s) \eta}{D_2(s) s} \]  \hspace{1cm} (34)

with

\[ N_2(s) = G_d + G_1 JG_d + kG_1 G_d e^{-s\tau_1} - G_1 JG_d e^{-s\tau_1} \]  \hspace{1cm} (35a)

\[ D_2(s) = J(G_d + G_1(1 - e^{-s\tau_1} + kG_d)) + kG_1 e^{-s\tau_1} + 1. \]  \hspace{1cm} (35b)

Substituting \( G_d(s), \ G_1(s) \) given in (2a) and \( J(s) \) as a PI or PID controller, we have,

\[ \lim_{t \to \infty} y(t) = \lim_{s \to 0} sY(s) = 0 \]  \hspace{1cm} (36)

Now, from the previous discussions and results, the proposed methodology can be summarized as follows.

**Procedure 8** To obtain the complete control scheme proposed in Fig. 6:

1. Check assumptions of Theorem 4, i.e., \( G_r(s) \) a stable transfer function and \( G_d(s) \) of the form (2a)

2. Check that the necessary and sufficient condition in Theorem 4 i.e, \( \tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i} \) is satisfied.

3. Choose the parameter \( k \). This can be done by considering (10) and (11).
4. Design the controller $J(s)$ based on the delay free model of the direct path, $G_1(s)$. A PI/PID control based strategy can be considered (see Theorem 6 and Corollary 7). Notice that as the internal states are estimated, it is also possible to use an estimated state feedback instead of the controller $J(s)$.

4.3 Application of the method to the direct path with possible LHP zeros.

This subsection presents how to apply the methodology developed in previous section in order to deal with recycling systems when the direct path includes LHP (stable) zeros. In particular the stabilization problem is solved for this class of systems. Therefore, $G_d(s)$ and $G_r(s)$ are now defined as,

$$G_d(s) = e^{-\tau_1 s} \frac{\alpha(s + z_1) \ldots (s + z_p)}{(s - a)(s + b_1) \ldots (s + b_m)},$$

$$G_r(s) = G_2(s)e^{-\tau_2 s} = \frac{N(s)}{D(s)} e^{-\tau_2 s},$$

with $p \leq m$ and $z_1, \ldots, z_p > 0$, i.e., left half plane (LHP) zeros in the direct path of the recycling system. Before providing the observation strategy designed for system (37) some further ideas are depicted. Consider the partition of $G_d(s)$ as,

$$G_d(s) = G_{1a}(s)e^{-\tau_1 s}G_{1b}(s)$$

where,

$$G_{1a}(s) = \frac{\alpha}{(s - a)(s + b_1) \ldots (s + b_{m-p})},$$

$$G_{1b}(s) = \frac{(s + z_1) \ldots (s + z_p)}{(s + b_{m-p+1}) \ldots (s + b_m)}.$$  

Now, Fig. 8 shows how to extend the ideas presented in Section 2 to the recycling system given by (37). From Fig. 8, the estimation of $\omega_2(s)$ as well as $\omega_1(s)$ are required in order to stabilize the process. However, from our assumptions, the measure of these variables are not available from the process and an observer to estimate the required signals is used. The proposed observer scheme is shown in Fig. 9.

Figure 8: Main idea of stabilization.
Theorem 9 Consider the observer-predictor scheme shown in Fig. 9, assuming $G_r(s)$ is a stable transfer function and $G_d(s)$ defined as in (37a). There exists a constant $k$ and a vector $L$ such that

$$\lim_{t \to \infty} [\omega_j - \hat{\omega}_j] = 0,$$

for $j = 1, 2, \ldots$, if and only if

$$\tau_1 < \frac{1}{a} - \sum_{i=1}^{m} \frac{1}{b_i},$$

for $i = 1, 2, \ldots, m$.

**Sketch of the proof.** The proof of this Theorem is similar to the proof of Theorem 4 by using Lemma 2 instead of Lemma 1. Thus, the error dynamics of the scheme shown in Fig. 9 are obtained, which are compared with the dynamics analyzed in Lemma 2. As both dynamics are the same, we conclude that the error convergence conditions are given by the stability conditions of the closed-loop system presented in Lemma 2.

From Theorem 9, we have adequately estimated the internal signals $\hat{\omega}_1(s)$ and $\hat{\omega}_2(s)$. In order to stabilize the recycling system given by (37), the ideas provided previously in Fig. 8 are applied by using the following control law,

$$U(s) = C(s)(R(s) - \hat{\omega}_1(s)) - \hat{\omega}_2(s)$$

Note that in (41), the controller $C(s)$ should be designed for the delay free plant $G_{1a}(s)$ as shown Fig. 8 (for instance, a proportional controller). However, notice that in this case it is not possible to implement $C(s)$ as a PI or PID controller in order to achieve tracking step references, since in the direct path of the recycling plant we have an additional dynamic due to $G_{1b}(s)$. In this way, the control-observer strategy is shown in Fig. 9 and achieves the stabilization of the recycling system when the direct path consider LHP zeros. Now, the application of the method for this class of systems can be summarized in the following result.
Procedure 10  To get the whole control scheme proposed in Fig. 9, is,

1. Check assumptions of Theorem 4 are satisfied, i.e., $G_r(s)$ a stable transfer function and $G_d(s)$ of the form (37a).

2. Check that the necessary and sufficient condition in Theorem 4, i.e., $\frac{1}{a} - \sum_{i=1}^{m-p} \frac{1}{b_i}$ is satisfied.

3. Design of the observer parameter $L$, by considering a traditional output injection to $G_{1a}(s)$. The parameter $L$ should be computed such that the relocated poles of $G_{1b}(s)$ are set at \{-z_1, -z_2, ..., -z_p\}. This can be done by using the commands acker or place from the package MATLAB.

4. Choose observer parameter $k$ from (11). This can be done by taking into account only $G_{1a}(s)e^{-\tau_1s}$ ($G_{1a}(s)$ given by (39a)).

5. Design the controller $C(s)$ based on the delay free model of the direct path, $G_{1a}(s)$, for instance a proportional controller.

5 SIMULATION RESULTS

In this section, two academic examples show the performance of observer based control strategy previously proposed. In examples the procedure given have to be followed namely:

1. Check assumptions of Theorem 4/9 are satisfied.

2. Check that the necessary and sufficient condition in Theorem 4/9.

3. Design of the observer parameter(s).

4. Design of the controller $J(s)/C(s)$ based on the delay free model of the direct path, $G_{1a}(s)/G_{1a}(s)$.

5.1 Example 1.

Consider the delayed recycling system of the form (1) with,

\[ G_d(s) = \frac{1}{(s-1)(s+10)}e^{-0.5s}, \quad G_r(s) = \frac{1}{s+1}e^{-2s}. \] (42)

Following the procedure described by Procedure 8, it is verified that $G_r(s)$ is a stable transfer function and $G_d(s)$ is of the form of (2a). Also we see that the condition $0.5 = \tau_1 < \frac{1}{a} - \sum_{i=1}^{m-p} \frac{1}{b_i} = 0.9$ is satisfied. Then, the single observer parameter $k = 12$ is chosen from (11). Finally, the controller $J(s)$ is proposed as
a two degree of freedom PID [24], which is designed based on delay free model $G_1(s)$. Such controller is implemented as,

$$U(s) = R(s)G_{ff}(s) - G_c(s)\widetilde{\omega}_1(s) - \widetilde{\omega}_2(s)$$

with

$$G_{ff}(s) = 160 \left(0.3 + \frac{1.6}{s} + 0.1s\right),$$

$$G_c(s) = 160 \left(1 + \frac{1.6}{s} + 0.1s\right).$$

Now, in order to evaluate the output signal evolution some numerical simulations are presented. It is considered a positive unit step input and initial conditions in $G_1(s)$ of the process and the observer of magnitude 0.4 units and 0.2 units, respectively. Also an unit step disturbance $T(s)$ acting at 15 sec. is considered. In Fig. 10, a solid line shows the output response when it is considered the exact knowledge of the model parameters; a dashed line presents the output signal when the time delays $\tau_1$ and $\tau_2$, are increased by 8% and 15%, respectively. Then, without disturbance $T(s)$ Fig. 11 shows the corresponding output estimation error $e_y(t)$. From Fig. 10 and Fig. 11 it can be seen the observer predictor convergence and the nice behavior of the control based on estimated signals.

**Figure 10:** Output signal with different initial condition in process and observer.

5.2 **Example 2.**

Consider the recycling system (1) where,

$$G_d(s) = \frac{(s + 3)(s + 5)}{(s - 1)(s + 8)(s + 1)(s + 2)}e^{-0.6s}$$

$$G_r(s) = \frac{1}{s + 1}e^{-s}.$$  

As LHP zeros in the direct path of recycling system are considered, the method developed in Subsection 4.3 is used. Following the procedure of Procedure 10,
we see that the assumptions given by Theorem 4 are satisfied, i.e. $G_p(s)$ is a stable transfer function and $G_d(s)$ is of the form of (37a). Thus we define,

$$G_{1a}(s) = \frac{1}{(s-1)(s+8)}$$  \hspace{1cm} (47a)

$$G_{1b}(s) = \frac{(s+3)(s+5)}{(s+1)(s+2)}.$$  \hspace{1cm} (47b)

Then, the established condition in Theorem 4 is also satisfied since $0.6 = \tau_1 < \frac{1}{a} - \sum_{i=1}^{m-p} \frac{1}{b_i} = 0.875$.

Now, in order to design the observer parameter $L$, an output injection to $G_{1b}(s)$ is applied. In this way, consider a possible state space representation of $G_{1b}(s)$ as,

$$A_{1b} = \begin{bmatrix} -3 & -1.4142 \\ 1.4142 & 0 \end{bmatrix}, \quad B_{1b} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad C_{1b} = \begin{bmatrix} 5 & 9.1924 \end{bmatrix} \quad \text{and} \quad D_{1b} = 1.$$

The vector $L$ should be computed such that the poles of $G_{1b}(s)$ are relocated at $\{-3,-5\}$, i.e.,

$$\sigma(sI - A_{1b} - LC_{1b}) = \{-3 - 5\}$$  \hspace{1cm} (48)

where $\sigma(.)$ represents the eigenvalues of. As a result we obtain $L = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.

The second observer parameter $k$ can be obtained from (11) with $G_{1a}(s)e^{-\tau_1 s}$, in such case we have $8 < k < 12.8$, to simulation $k = 10$ is used. Finally, as mentioned before a proportional controller $C(s)$ can be designed with a Nyquist or Root Locus diagrams based on delay free model $G_{1a}(s)$. In this way, we consider $C(s) = K_1$ as a static controller. Therefore, it is obtained $K_1 > 8$ in order to stabilize, and $K_1 = 20$ is chosen to the simulation experiments.

In order to evaluate the output signal performance, an initial condition in the states of $G_{1a}(s)G_{1b}(s)$ of magnitude 0.4 units is regarded. In Fig. 12 a solid line shows the output response when it is considered the exact knowledge of the model parameters; a dashed line presents the output signal when the process

![Figure 11: Estimation error $e_y(t)$ by considering different initial condition in process and observer.](image-url)
time delays $\tau_1$ and $\tau_2$, are increased by 5%. Also in Fig. 13 the estimation error at the output $e_y(t)$ is presented.

Figure 12: Stabilization of recycling system, Example 2.

Figure 13: Output error signal, Example 2.

6 CONCLUSIONS

Using recycle in unstable processes with significant time delay leads to a challenging control problem. In this work this problem has been addressed for the case of one unstable pole, $m$ stable poles, and possibly $p$ LHP zeros with significant delay in the direct path, which is a more general case than the one analyzed in [16] and [17]. Results with respect to step tracking reference and step disturbance rejection are also presented. Explicit conditions for the construction of a stabilizing observer based controller scheme for such class of systems are presented. The observer-prediction strategy is used to estimate some internal variables of the process that are used to: i) remove the dynamics of recycling path in the process and ii) design a stabilizing control law for the free delay model of the direct path. Simulation results have been presented in order to show the performance of the method under different circumstances.
References


