Non–Interacting Fuzzy Control System Design for Distillation Columns

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Abstract: - This paper proposes a fuzzy multiloop control design for a distillation column. The interaction that occurs between the strategic distillation column variables, which represent a major constraint to construct the rules base when the fuzzy multivariable control of the distillation control is considered, are reduced by introducing a compensator in cascade with the distillation column. The introduced compensator is designed using an interaction analysis method. Thus, a complete fuzzy rule base is generated for each loop independently without domain experts. The main advantage of the proposed fuzzy multiloop strategy is that the determination of the rule base is simplified and facilitated; besides a weak level of interaction and the control stability are guaranteed. Consequently, in comparison with the conventional multiloop control based on classical PI controllers, the fuzzy multiloop control achieves better control performance.

Key Words: - Multiloop control, interaction analysis, control configuration, compensator, Direct Nyquist Array (DNA), Relative Gain Array (RGA), PI fuzzy controller.

1 Introduction

For many reasons, distillation remains the most important separation technique in the chemical process industries around the world. For these reasons, improved distillation control can have a significant impact on reducing energy consumption, improving product quality and protecting environmental resources. However, distillation control is a challenging problem, due to the following factors [12]:

- process nonlinearity;
- substantial coupling of manipulated variables;
- severe disturbances; and
- nonstationary behavior.

Accordingly, much research and development in both the private and public sector has focused on control methods that use modern computing power to cope with these control related difficulties [15].

Many methods have been suggested to design non–interacting or decoupling multivariable control systems [1], [7]. Weischedel and coworkers have studied decoupling of conventionally (energy balance) controlled distillation towers, and conclude that complete decoupling is not feasible for many (high purity) columns due to sensitivity to model error [17].

Ryskamp suggested using implicit decoupling rather than explicit decoupling [13]. Ryskamp argued that by the proper selection of process measurements one could obtain a naturally decoupled system. Georgakis developed the extensive variable control method to decompose the process system into slow and fast dynamic modes, which are related to the system or to the total energy or mass content of subsystem [3]. A simple structure for neural network control is proposed by Ramchandran & Rhinehart [11]. Viel et al. developed a Lyapunov–based controller for the composition control of binary distillation columns [16].

Most advanced control techniques are generally grounded in the use of nonlinear multivariable models [11]. The linear models generally tend to become rigorous and computationally intensive as the process behavior become more complex. While control success has been demonstrated, it is often at the expense of computational power, operator–friendly interaction or ease of controller development and maintenance.

Fuzzy logic offers an alternative approach to the control of processes, as they do not require a priori knowledge of the process phenomena. Fuzzy logic is capable of handling complex
and nonlinear problems, process information rapidly and can reduce the engineering effort required in controller model development [2], [10], [18].

Fuzzy logic has been successfully applied to variety of industrial processes such as cemetery, paper production, control of the chemical reactor temperature and ethylene production. In the area of process control, a few applications have been reported.

In this paper, the fuzzy control of distillation column is presented and the performances of this control approach are compared in simulation with that provided by the classical PID controllers. The outline of this paper is as follows. Section 2 deals with fuzzy logic and its relevance to process control and the encountered difficulties when it is applied for distillation control whereas section 3 presents the considered control strategy for the distillation column. A comparison between the use of fuzzy and the classical PID controllers in the multiloop strategy considered is reported in section 4 followed by a final conclusion.

2 Fuzzy logic and its relevance to process control

The fuzzy logic control is interesting for the following reasons:

• Simple to realize, and adaptable to the working production conditions;
• The synthesis of several expert acknowledge is easily feasible;
• It’s a robust control technique;
• The users judge that it permits a high accuracy and generally the energy economy;
• It has proven its efficiency in many applications [10].

The application of the fuzzy logic to control the distillation column stumbles to the following problems:

• The choices of the fuzzified variables, since, there are several control variables and variables to be controlled;
• Modeling difficulties of the operators acknowledge, so the determinations of the rules base, since the distillation phenomenon is complex;
• Time consuming to generate the control actions seen the number of the rules.
• Power integrity.

In addition the interaction that occurs between the distillation column variables poses a serious problem to design a fuzzy control system.

In this work, we design a fuzzy multiloop control for a distillation column. This control strategy simplifies and facilitates the design of the fuzzy control since the rules base is determined for each loop independently. Besides the fuzzy multiloop design, also a comparison with the multiloop control using the classical controllers is presented to show the contribution of the fuzzy control.

3 Control strategy

In designing a multiloop control, the key decision is the selection of the best control configuration. The most encountered problem while designing the multiloop control is the interactions that occur between system variables. Many techniques to analyze interaction have been developed to design more effective control system [5], [9].

However there are many cases where the interaction measures show the absence of an adequate control configuration (the system is highly interactive). The Decoupling method (simplified, ideal or inverted decoupling) is relatively complex task since all techniques have their advantages and limitations [8], [9]. Seen the interaction that occurs between the distillation control variables, the multiloop control do not provide the desired control performance for expected disturbances and set point changes.

In order to apply the fuzzy control of distillation columns, the control strategy given in Figure 1 is considered. As shown in Figure 1, a compensator is introduced in cascade with the distillation column that reduces the interaction

![Fig.1 The considered multiloop control structure.](image-url)
between the considered control configuration loops.

4 Fuzzy and classical multiloop control of distillation column

4.1 Distillation column model

The example consists of a two-product distillation tower separating a binary feed. Both top and bottom product compositions are of equal importance, and the major disturbance is a change in feed composition. The considered process shown in Figure 2, has the distillate and reflux pairings interchanged; this called material balance and has the following model [8]:

\[
\begin{bmatrix}
    x_d \\
    x_b
\end{bmatrix} = G(s) \begin{bmatrix}
    f_d \\
    f_v
\end{bmatrix} + G_d(s)x_f. \tag{1}
\]

Where

\[
G(s) = \begin{bmatrix}
    -0.0747 e^{-2s} & 0.0080 e^{-2s} \\
    10 s + 1 & 5 s + 1 \\
    0.1173 e^{-2s} & 0.0080 e^{-2s} \\
    9 s + 1 & 3 s + 1
\end{bmatrix}
\]

\[
G_d(s) = \begin{bmatrix}
    0.70 e^{-5s} \\
    14.4 s + 1 \\
    1.30 e^{-3s} \\
    12.0 s + 1
\end{bmatrix}
\]

Where \(x_d\) is the distillate, \(x_b\) is the bottoms, \(f_d\) is the distillate flow, \(f_v\) is the reboiled vapor and \(x_f\) is the feed composition.

Note that the reflux flow \(f_v\) are potential manipulated variables, and the feed composition \(x_f\) is a disturbance, because it depends on upstream operations and is assumed not free to adjust.

The units are mole fractions of the light key components for the composition, K mole/min for the flows, and min for time.

4.2 Interactions analysis

The Direct Nyquist Array in Figure 3 is used to analyze the interaction in the considered distillation column. Thus, Figure 4 gives the diagonal elements superposed by the Gershgorin circles [5], [9]. It shows that the manipulated variable \(f_d\) affects strongly the controlled variable \(x_d\) and \(x_b\). Therefore, the two possible control configurations are interactive. Thus, the control strategy given in Figure 1 is indicated to control correctly the distillation column using the multiloop control system.

4.3 Compensator design

To reduce the interaction between the distillation column variables, the compensator \(K(s)\) is introduced. The compensator \(K(s)\) is designed using the method proposed in [6] to control highly interactive system. The application of the Direct Nyquist Array (DNA) as an interaction

![Fig. 2. Schematic diagram of material balance energy.](image)

![Fig. 3. The Direct Nyquist Array of the study distillation column](image)

![Fig. 4. The diagonal elements of the distillation column superposed by the Gershgorin circles.](image)
analysis method gives the following compensator $K(s)$ (see Appendix).

$$K = \begin{bmatrix} 0.3891 & -0.1488 \\ -5.7046 & -0.3891 \end{bmatrix}. \quad (2)$$

The contribution of the compensator is illustrated by Figure 5, which shows that the cascade (compensator–distillation column) is much more diagonally dominant. Therefore, interactions between the control loop configuration defined by the cascade diagonally elements $[f_j^c - x_j]$; $[f_i^c - x_i]$ are insignificant, which permits to apply multiloop control.

4.4 Stability analysis

The stability of the considered control strategy is guaranteed if the control configuration loops verify the Bristol’s condition, that means, the correspondent relative gain to the control configuration pairs must be positive [5], [8-9], [14]. The values of the cascade’s RGA below, indicate that the stability of this column using the control strategy shown in Figure 1 is assured, since the diagonally elements corresponds to the considered control configuration pairs are positives.

$$RGA = \begin{bmatrix} 0.5692 & 0.4308 \\ 0.4308 & 0.5692 \end{bmatrix}. \quad (3)$$

4.5 Conventional and fuzzy PI controllers design

The conventional PI controllers are designed using the method developed by Issaksson and Graeb [4], which gives the parameters depicted in Table 1.

The considered PI fuzzy controller for each loop is shown in Figure 6. Where $G_{ce}, G_{de}$ and $G_u$ are gains scaling factors and heir values are reported in Table 1.

The membership functions of inputs (error $e$ and derivative error $de$) and output of each controller are triangular with five sets.

The fuzzy rules describing the controller structure are shown in Table 2.

The inference engine used for fuzzy rules processing is sum–prod method.

The center–of–gravity method is used to fuzzify the overall subset representing output control variable.

<table>
<thead>
<tr>
<th>Loops</th>
<th>Fuzzy controllers</th>
<th>Conventional controllers</th>
</tr>
</thead>
<tbody>
<tr>
<td>$Ge$</td>
<td>$Gde$</td>
<td>$\Delta u$ $K_c$ $T_i$</td>
</tr>
<tr>
<td>$[f_j^c - x_j]$</td>
<td>0.003</td>
<td>0.0025</td>
</tr>
<tr>
<td>$[f_i^c - x_i]$</td>
<td>0.002</td>
<td>0.0050</td>
</tr>
</tbody>
</table>

Table 2 Rule base

<table>
<thead>
<tr>
<th>$\Delta u$</th>
<th>$e$</th>
<th>$de$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$N$</td>
<td>$B$</td>
<td>$N$</td>
</tr>
<tr>
<td>$N$</td>
<td>$B$</td>
<td>$N$</td>
</tr>
<tr>
<td>$Z$</td>
<td>$B$</td>
<td>$N$</td>
</tr>
<tr>
<td>$P$</td>
<td>$Z$</td>
<td>$N$</td>
</tr>
</tbody>
</table>

Where:

- NB: negative big.
- NS: negative small.
- ZE: zero.
- PS: positive small.
- PB: positive big.

![Fig. 5. The diagonal elements of the distillation column superposed by the Gershgorin circles.](image)

![Fig. 5. PI Fuzzy controller.](image)
4.5 Simulation results
The transient responses for well–tuned feedback control in response to a feed composition upset are given in Figure 6, and the control performances are summarized in the IAE values in Table 3. Note that both fuzzy and conventional control assures the set point tracking. Based on the total IAE values (0.2791 for the conventional controllers and 0.139 for the fuzzy controllers) and the response time, the performances obtained with the fuzzy multiloop control are better than those obtained using the conventional multiloop control for the feed composition disturbance.

The dynamic responses for a set point change in the top composition controller of +0.005 mole fraction, with the other set point and all disturbances constant, are given in Figure 7. The results summarized in Table 3, show that the total IAE values are 0.055 for the conventional control and 0.08783 for the fuzzy control. In this case, the former system control gives the better results, but the controlled variable \( x_b \) is more affected in relation to the fuzzy control case and note that the transient responses for \( x_d \) present the oscillations.

| Loops | IAE
<table>
<thead>
<tr>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Fuzzy controllers</td>
</tr>
<tr>
<td>Set point change</td>
<td>Perturbation rejection</td>
</tr>
<tr>
<td>([\dot{x}_c - x_c] )</td>
<td>0.08700</td>
</tr>
<tr>
<td>([\dot{x}_c - x_c] )</td>
<td>0.00083</td>
</tr>
</tbody>
</table>

5 Conclusion
In this paper, a systematic design procedure to obtain a fuzzy multiloop control for distillation column is presented. The key idea of the design procedure is to introduce a compensator in cascade with the distillation column in order to reduce the interaction that occurs between their strategic variables. The main advantage of the considered multiloop control strategy is the simplicity in determining the control rules and controllers’ parameters since each loop is treated independently both in fuzzy and conventional control.

The validity of the proposed strategy, in which a compensator is introduced and designed using an interaction analysis method, is confirmed through simulations results and demonstrates the superior performance of the fuzzy multiloop control scheme since it permits attaining desired responses to changes in set point and achieving sufficient feedback properties (disturbance rejection).

Appendix
The application of the method to control a highly interactive system proposed by Khelassi, et al. [6] gives:

\[
G(s)K(s) = G^*(s)
\]  

(3)

Where \( G(s) \) is the distillation column model, \( K(s) \) the compensator and \( G^*(s) \) is the cascade model.

The Direct Nyquist Array (DNA) is used as an interaction method to design \( K(s) \), and the considered control configuration is defined by the diagonally elements of \( G^*(s) \).
According to the DNA of the $G(s)$ in Figure 3, the compensator will be designed in order to reduce the effect of $f_v$ on the controlled variable $x_d$, hence, the idea consist to determine, using the DNA plot of $G(s)$, a linear combination between columns or lines of the DNA so that the effect of $f_v$ on $x_d$ will be negligible, this is achieved by the subtraction of line 2 from line 1 which gives the following cascade.

$$G^\ast(s) = \begin{bmatrix}
g_{11}(s) & g_{12}(s) \\
\alpha g_{11}(s) - g_{21}(s) & \alpha g_{12}(s) - g_{22}(s)
\end{bmatrix}$$

(4)

Where $\alpha$ is terminated so that the interaction transmittances are almost reduced and $G^\ast(s)$ will be diagonally dominant. The parameter $\alpha$ is determined as follow.

$$\alpha = \frac{g_{21}(0)}{g_{11}(0)}$$

(5)

and

$$K(s) = G^{-1}(s)G^\ast(s)$$

(6)

In order to simplify the compensator structure and to assure the physical realization of the compensator given in (6), $K$ is taken as:

$$K = G^{-1}(0)G^\ast(0)$$

(5)

References


