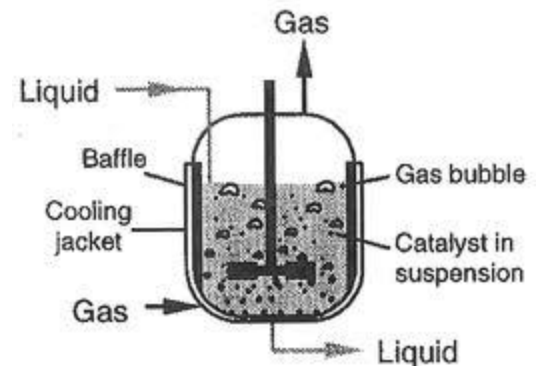


ENGR691X: Fault Diagnosis and Fault Tolerant Control Systems  
Fall 2010

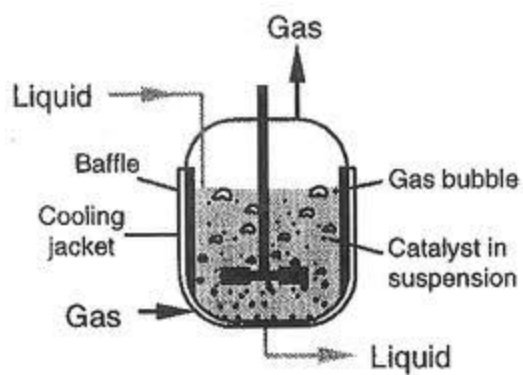
## **Adaptive Fault Tolerant Control of an unstable Continuous Stirred Tank Reactor (CSTR)**

*Group Members:* Maryam Gholamhossein  
Ameneh Vatani

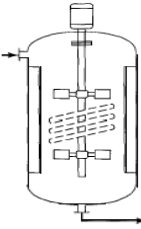




# Why CSTR?

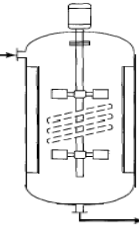


# Outlines



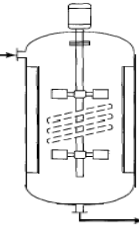
- Continuous Stirred Tank Reactor (CSTR) Model
- Controller
  - Conventional controller
  - Fuzzy logic
  - Adaptive Fault tolerant fuzzy PID controller
- Simulation results & analysis under different fault scenarios
- Conclusion and suggestions

# Reactors



- Chemical reactors are one of the most important part of chemical, biochemical and petroleum processes since they transform raw materials into valuable chemical materials.
- Three classical chemical reactors
  - Batch reactor
  - Continuous stirred-tank reactor (CSTR)
  - Plug flow tubular reactor (PFR)

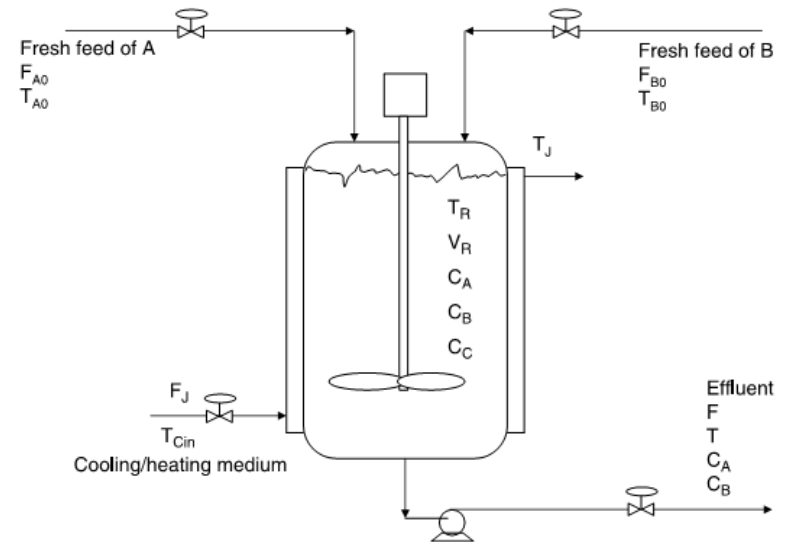
# CSTR Model



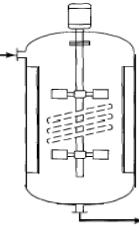
- The CSTR reactor is usually used for liquid-phase or multiphase reactions that have high reaction rates. Reactant streams are continuously fed into the vessel.
- Perfect mixing of the liquid in the reactor is usually assumed, so the modeling of a CSTR involves ordinary differential equations.

# CSTR Model

- Main characteristics of a CSTR
  - Constant temperature
  - Constant concentration
- Reaction types:
  - Exothermic (releasing energy)
  - Endothermic (requiring energy)
  - Reversible (balance of reactants and products)  $A + B \leftrightarrow C + D$
  - Irreversible (proceeding completely to products)  $A + B \rightarrow C$
  - Homogeneous (single-phase)
  - Heterogeneous (multiphase)



# CSTR Model

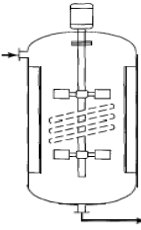


Exothermic and irreversible reactions



Temperature control problems

- Maintaining stable and safe temperature control
- Heat removal methods
  - Jacket cooling
  - Cooling coil



# CSTR Model

- Three-state CSTR model, exothermic-irreversible first-order reaction ( $A \rightarrow B$ )

$$\frac{dC_a}{dt} = \frac{Q}{V}(C_{af} - C_a) - k_0 \exp\left(\frac{-E_a}{RT}\right) C_a$$

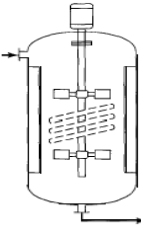
$$\frac{dT}{dt} = \frac{Q}{V}(T_f - T) + \left(\frac{-\Delta H}{\rho C_p}\right) k_0 \exp\left(\frac{-E_a}{RT}\right) C_a - \frac{UA}{V\rho C_p}(T - T_c)$$

$$\frac{dT_c}{dt} = \frac{Q_c}{V_c}(T_{cf} - T_c) + \frac{UA}{V_c \rho_c C_{pc}}(T - T_c)$$

- Dimensionless ...



# CSTR Model



- System dimensionless equations\*:

$x_1$ : dimensionless concentration       $x_2$ : dimensionless reactor temperature

$q_c$ : dimensionless cooling jacket temperature

$x_3$ : dimensionless cooling jacket flow rate       $x_{2f}$ : dimensionless feed flow

$$\frac{dx_1}{dt} = q(x_{1f} - x_1) - x_1k(x_2)$$

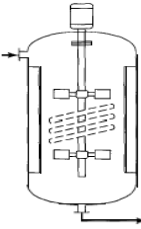
$$\frac{dx_2}{dt} = q(x_{2f} - x_2) - \delta(x_2 - x_3) + \beta x_1k(x_2)$$

$$\frac{dx_3}{dt} = \delta_1[q_c(x_{3f} - x_3) + \delta\delta_2(x_2 - x_3)]$$

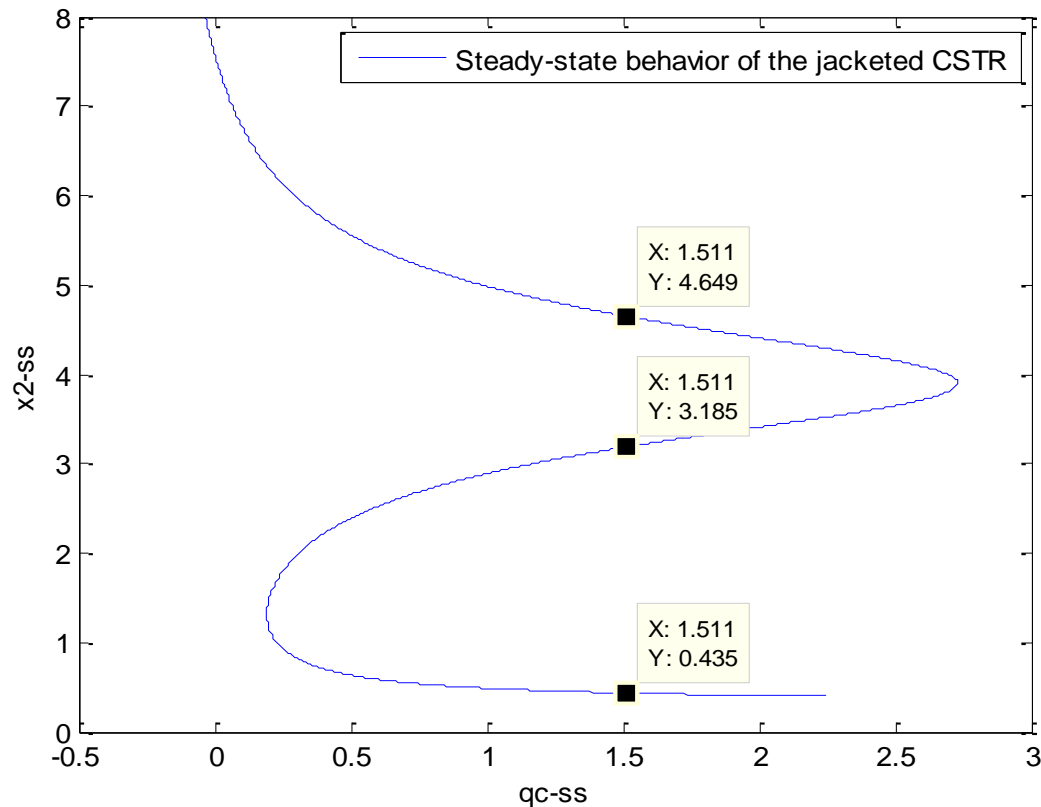
$$k(x_2) = \exp\left(\frac{x_2}{1 + x_2/\gamma}\right)$$

\* Russo L. P., Bequette B. W., "Impact of process design on the multiplicity behavior of a jacketed exothermic CSTR", *AIChE Journal*, 41(1)135

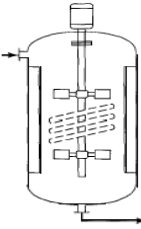
# CSTR Model



- System non-linearity
- Steady-State design and Multiplicity of CSTR



# Controller Design

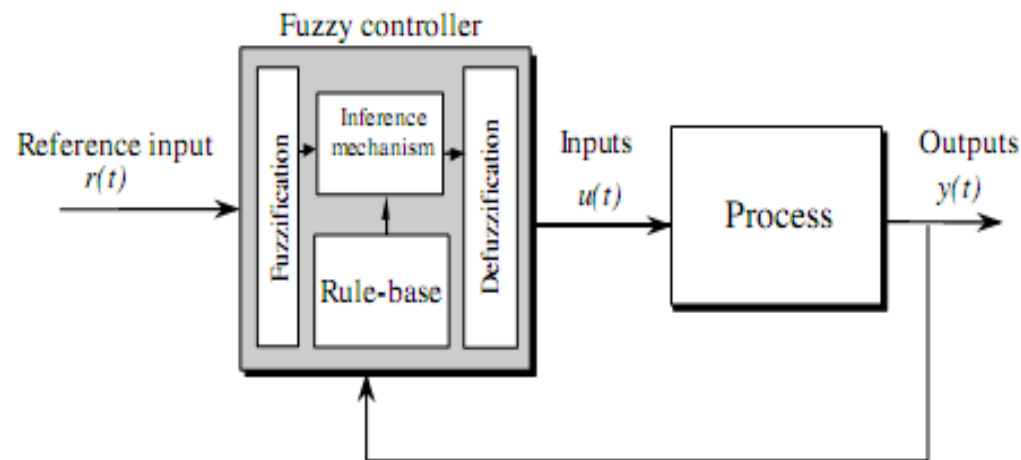


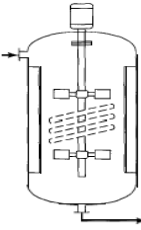
- Conventional controllers

- (PID control, state-space methods, optimal control, robust control,...)
- Designing based on the Mathematical models
- Ignoring heuristic information, as they do not fit into proper mathematical form

## Fuzzy controller

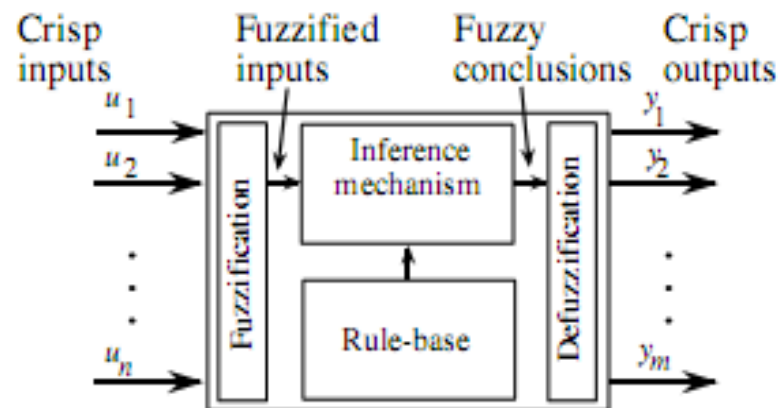
- An artificial decision maker that can operate in a closed-loop control system



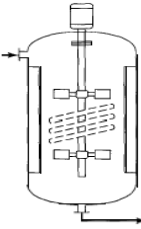


# Controller Design

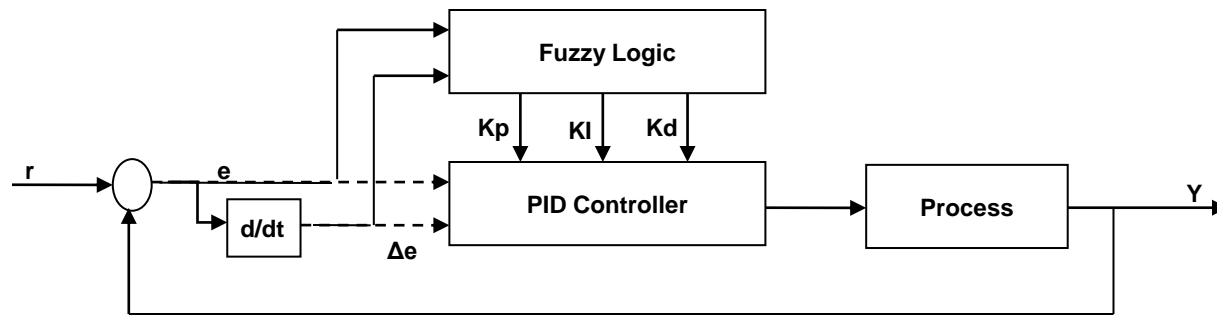
- *Rule-base*, holds the knowledge in the form of a set of rules of how best to control the system (a set of *If-Then* rules)
- *Inference mechanism* (inference engine) evaluates which control rules are relevant at the current time and deciding what the input to the plant should be
- *Fuzzification*, modifying the inputs so that they can be interpreted to the rules in the rule-base
- *Defuzzification*, converting the conclusions of inference mechanism into the plant inputs.



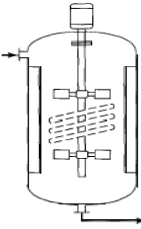
# Controller Design



- Adaptive fuzzy controller scheme  
(Fuzzy controller and conventional controller combination)



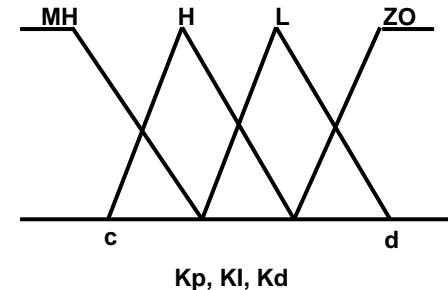
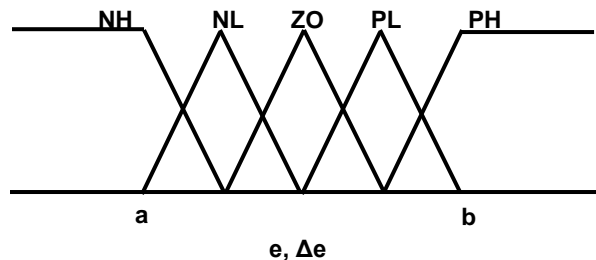
- Tracking and regulatory problem
  - Some continuous process produce different grades of products at different times



# Controller Design

- Fuzzy adaptation module steps:

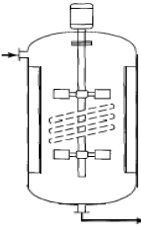
1) Defining the input & output membership functions



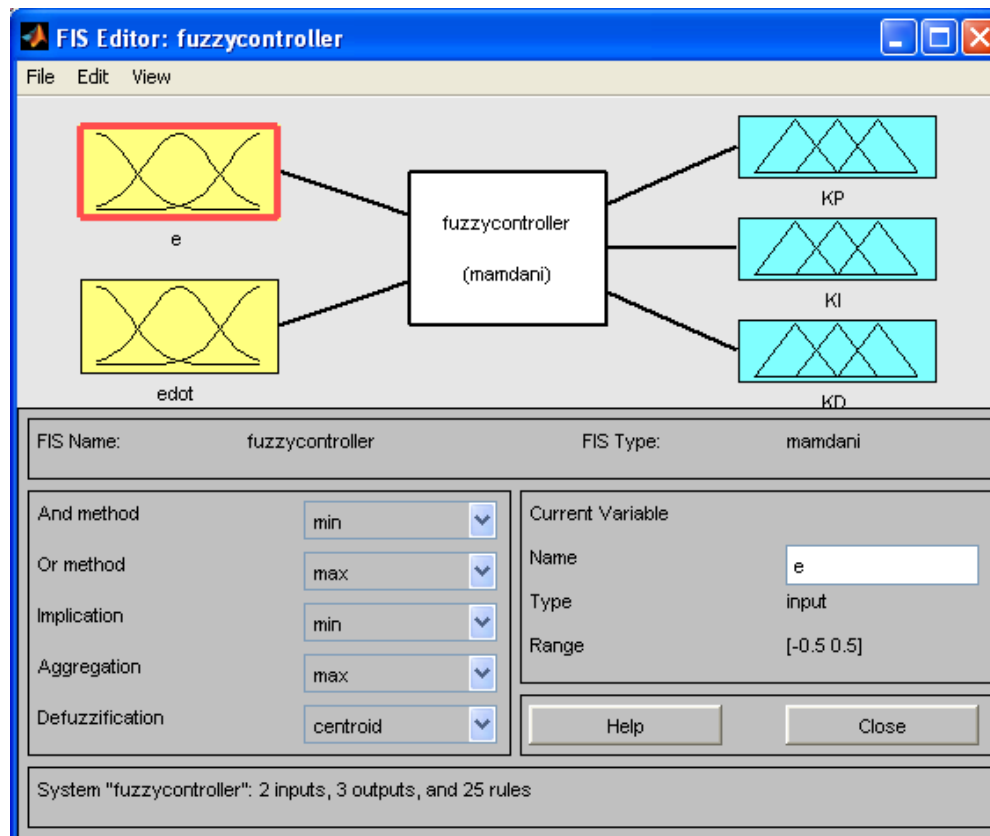
- 2) Defining the fuzzification and defuzzification methods
- 3) Defining Inference mechanism
- 4) Defining the Rules in the form of linguistic structure  
*(one of fuzzy implementation challenges!)*

*If e is X and  $\Delta e$  is Y, then  $KI=U$ ,  $Kp=V$ ,  $Kd=Z$*

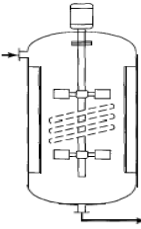
# Controller Design



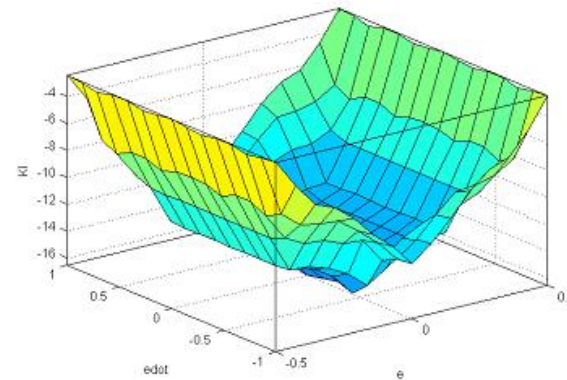
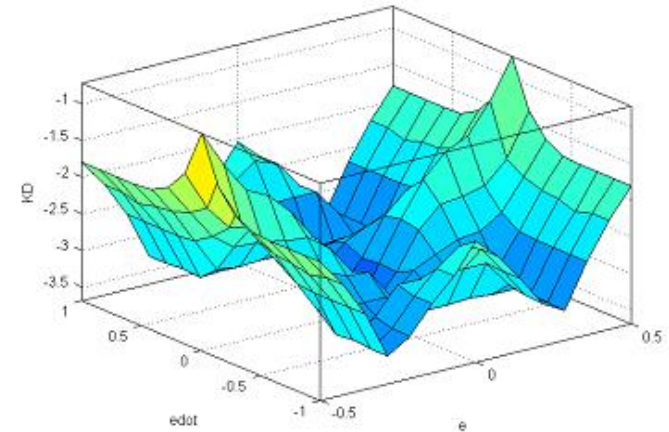
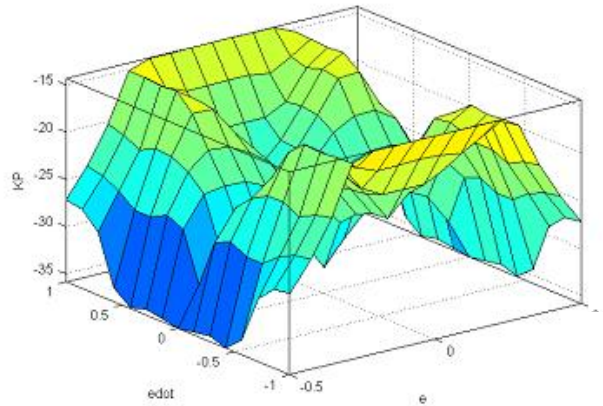
- Fuzzy controller inputs: Error ( $e$ ) and error changes ( $\Delta e$ )
- Fuzzy controller outputs: PID gains ( $K_p, K_c, K_d$ )
- Fuzzy Inference Strategy: Mamdani
- defuzzification method: Centriod



# Controller Design

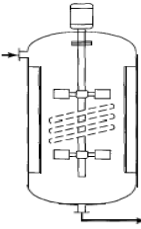


- AFTCS or PFTCS?!



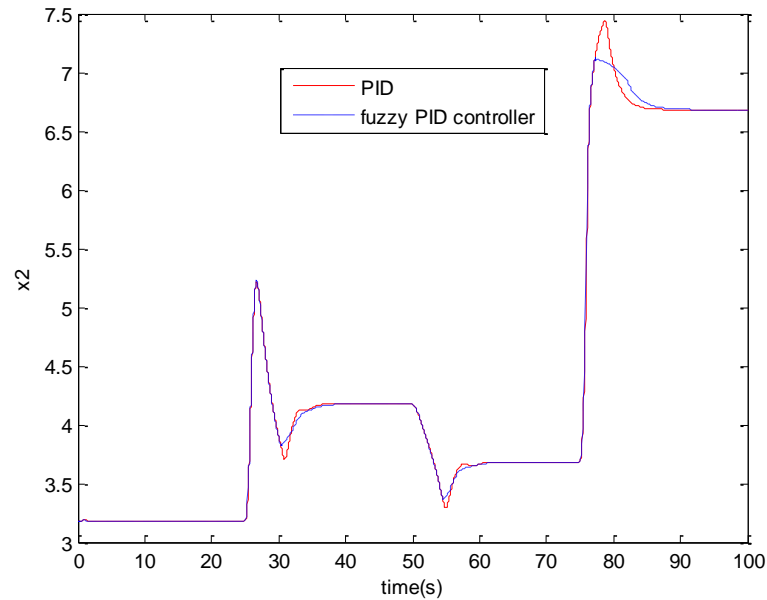
- So where is the FDD part?

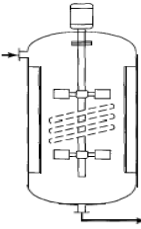




# Simulation results

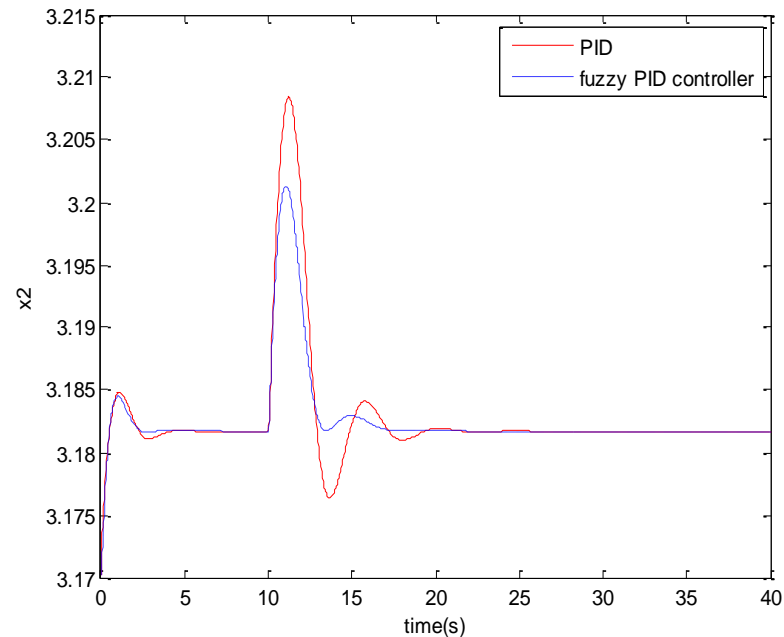
- Fault free tracking response



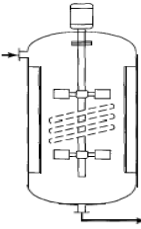


# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios

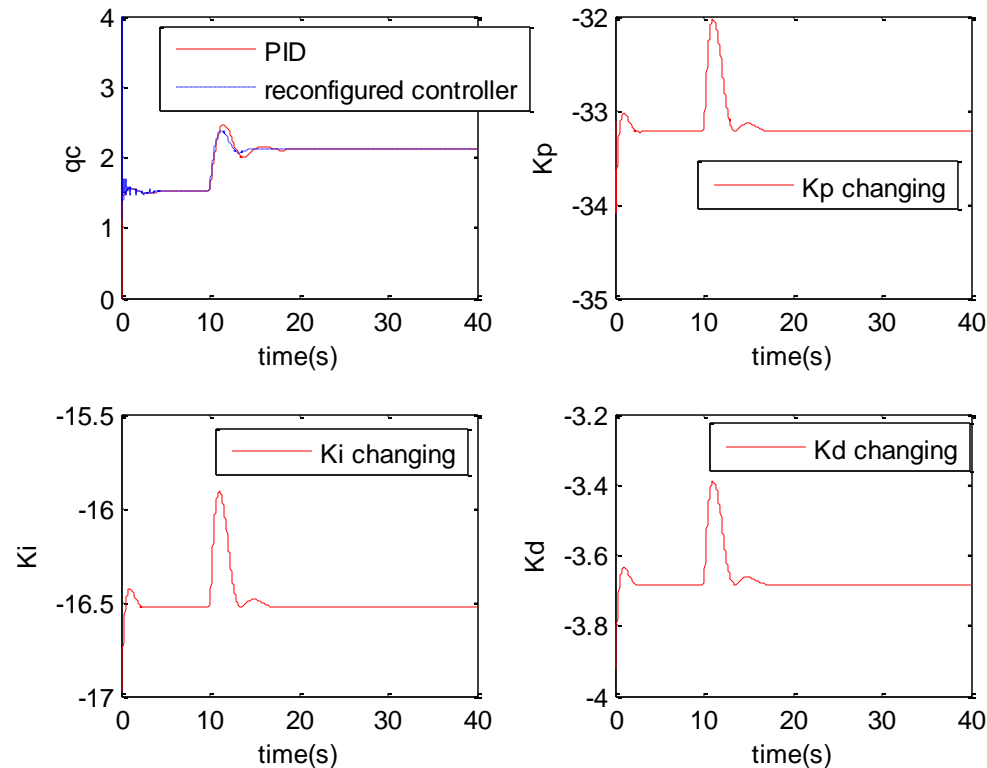


**System output response to 15% actuator failure**

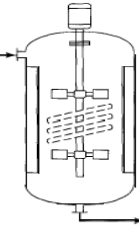


# Simulation results

- Actuator faults scenarios
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- Sensor Faults scenarios

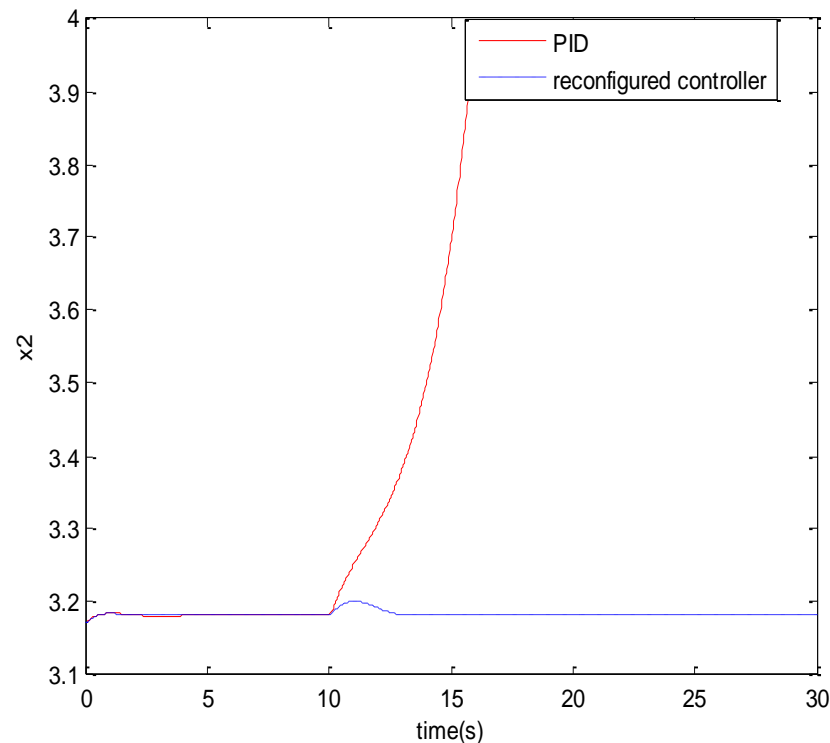


**Control input signals and controller gains under 15% actuator failure**

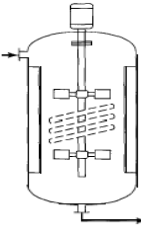


# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios

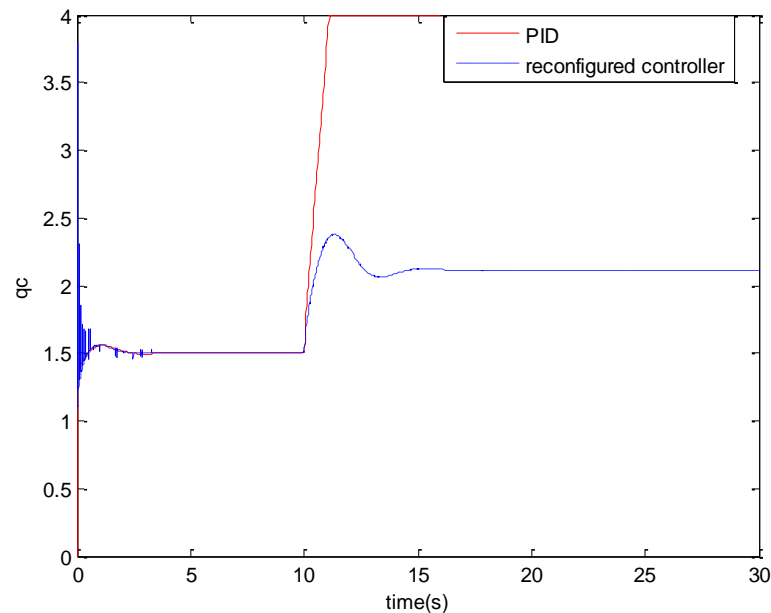


**System output response to 25% actuator failure**

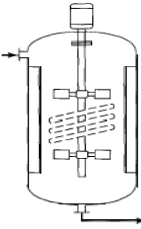


# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios

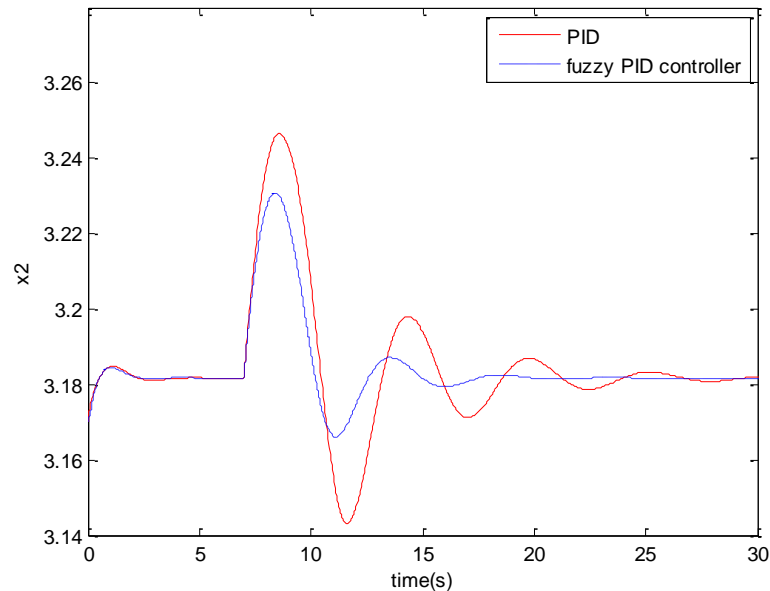


**Control input signals under 25% actuator failure**

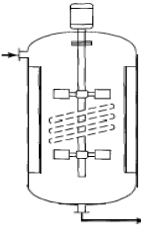


# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios

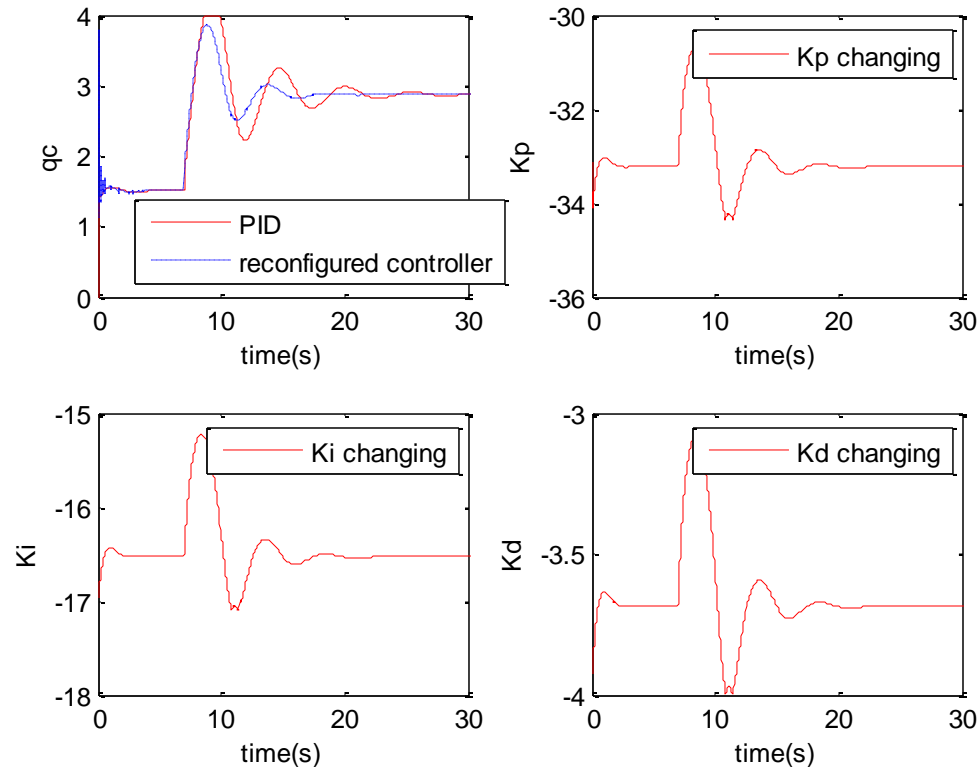


**System output response to changing  $x2f$  from 0 to 0.08**

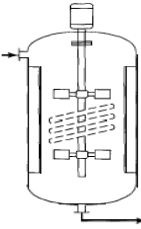


# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
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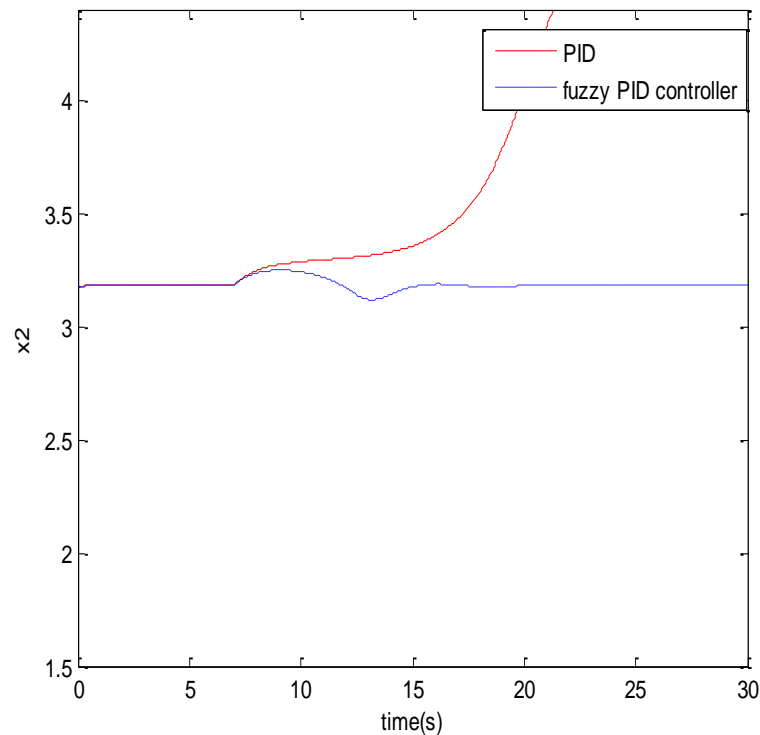


**Control input signals and controller gains changing  $\times 2f$  from 0 to 0.08**



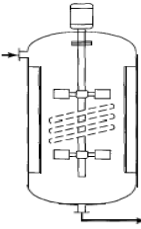
# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios



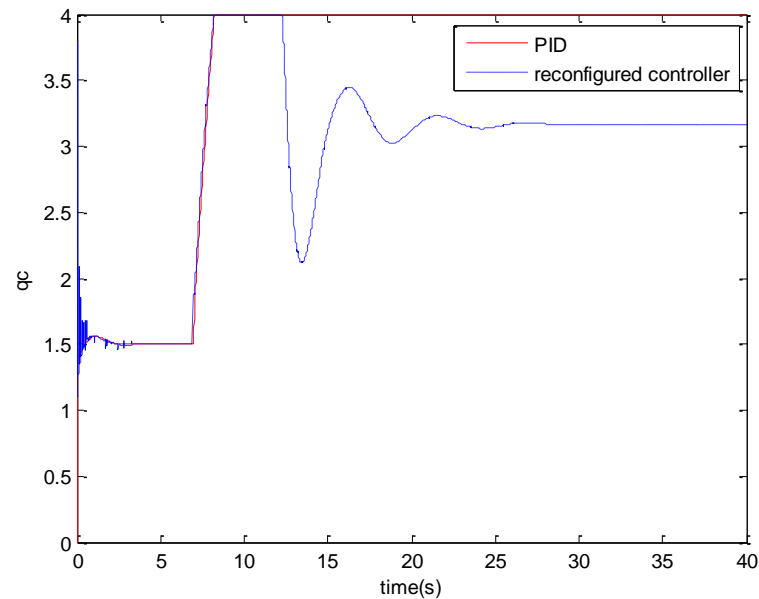
**System output response to changing  $x_2f$  from 0 to 0.1**



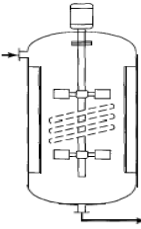


# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios

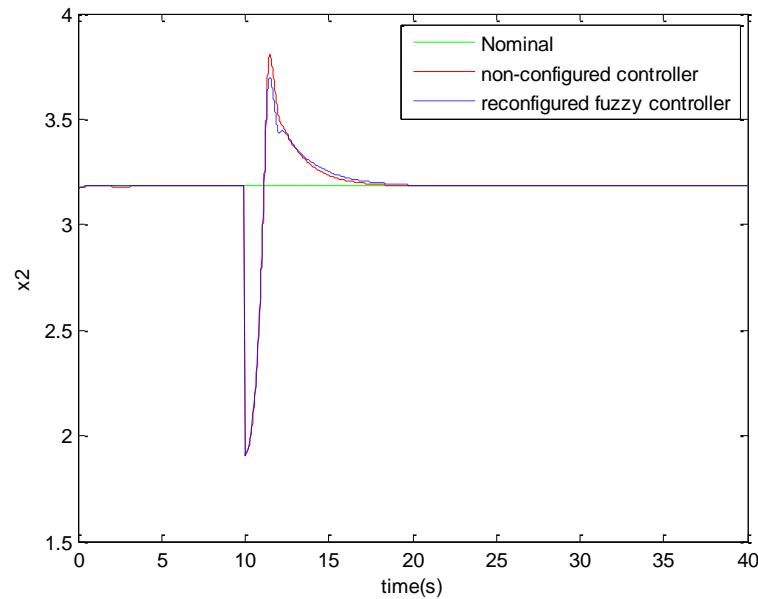


**Control input signals to changing  $x2f$  from 0 to 0.1**



# Simulation results

- Actuator faults scenarios
- System parameter fault scenarios
- Sensor Faults scenarios



**Output responses in the presence of 40% sensor fault**

# Conclusions & Suggestions



- In this project the fault tolerant control of a CSTR model under different faults is accomplished.
- Defining the proper fuzzy rules was a very challenging and time-consuming task!
- In spite of the conventional definition for Active FTCS which obligated the system to have a FDD block; here in this project FDD block is inherent in the fuzzy controller.
- When the fault percentage exceeds specific values, the conventional PID fails to control the CSTR while the fuzzy PID can have the pre-fault performance after a short transient time.

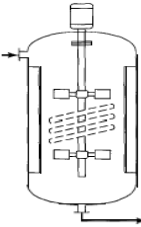
# Conclusions & Suggestions



- Extending this controller to a MIMO system.
- Taking other parameters as input of fuzzy controller.



**Thank you for your attention!**



# Simulation results

