

## ETCE2002/OT-29151

### A Fuzzy Logic Control Model for the Curing Process of Heat-Activated Coupling of FRP Composites to Alloy Pipes

Omer M. Soysal  
Department of Computer Science,  
Southern University,  
Baton Rouge, LA 70813

Patrick F. Mensah  
Department of Mechanical Engineering,  
Southern University,  
Baton Rouge, LA 70813

Amitava Jana  
Department of Mechanical  
Engineering,  
Southern University,  
Baton Rouge, LA 70813

#### ABSTRACT

A fuzzy control (FLC) approach has been developed to control the curing process of a polymeric laminate used as the bonding material for the heat-activated coupling (HAC) of a fiberglass composite to Cu-Ni alloy pipe. Controlling the temperature of the curing environment is required in order to achieve uniform cure within the bonding prepreg. Without temperature control, the alloy side of the coupling is essentially a heat sink. Therefore, a controlled heat source on the alloy pipe is needed to compensate for the heat sink effect that tends to decrease curing temperatures within the laminate. The simulation results obtained show that the curing process of HAC of a composite-to-alloy pipe is significantly improved with required uniform cure characteristics being obtained through the thickness of the laminate.

#### INTRODUCTION

Fiber reinforced plastics (FRP) are materials that have gained increasing usage in engineering applications in the past decade. The curing process of these materials to convert them into structural parts for engineering systems involves heating. The prescribed heating process cures the uncured FRP normally referred to as pre-impregnated (prepreg) laminate. Controlling the temperature of the curing environment is desirable in order to achieve uniform cure within the prepreg laminate with optimum mechanical properties.

For such type of systems a fuzzy logic control approach can be employed. Since the experience of the experts directs the control structure, the control system does not need analytical model. Fuzzy logic models, called fuzzy inference

systems, consist of a number of conditional "if-then" rules. For the designer who understands the system, these rules are easy to write, and as many rules as necessary can be applied to describe the system adequately (although typically only a moderate number of rules are needed).

In fuzzy logic, unlike standard conditional logic, the truth of any statement is a matter of degree; e.g. "how cold is it," "how high should we set the heat?" Fuzzy inference systems rely on membership functions to send input data to the computer so that a correct value between 0 and 1 can be calculated. The degree to which any fuzzy statement is true is denoted by a value between 0 and 1.

Not only do the rule-based approach and flexible membership function scheme make fuzzy systems straightforward to create, but they also simplify the design of systems and ensure that the system can easily be updated and maintained over time. [1]

Fuzzy logic may also be considered as an assortment of decision-making techniques. In many applications like process control, the algorithm's outcome is ruled by a number of key decisions, which are made in the algorithm. Defining the best decision requires extensive knowledge of the system. When experience or understanding of the problem is not available, optimizing the algorithm becomes very difficult. This is the reason why fuzzy logic is useful [2].

Furthermore, fuzzy logic starts with and builds on a set of user-supplied human language rules. Then, the fuzzy systems convert these rules to their mathematical equivalents. This simplifies the job of the system designer and the computer, and

results in much more accurate representations of the way systems behave in the real world.

Additional benefits of fuzzy logic include its simplicity and its flexibility. Fuzzy logic can handle problems with imprecise and incomplete data, and it can model nonlinear functions of arbitrary complexity. "If you don't have a good plant model, or if the system is changing, then fuzzy will produce a better solution than conventional control techniques," Bob Varley [1].

Ever since Mamdani applied fuzzy set theory to control a steam engine and boiler combination by a set of rules, there have been several fuzzy inferencing systems proposed by various researchers reported in literature [3]. In Mamdani type fuzzy inference system, the resultant control action of two rules is shown in Figure. 2. In this case, the resulting action is based on *min max* composition. The final crisp value is obtained by calculating the centroid of area. This process of defuzzification is known as center of area (COA) defuzzification method.

Fukuhara, Ishii and Agusa studied on an internal welding robot system to promote pipeline-joining efficiency in inside/outside simultaneous welding. They used fuzzy control system to stabilize the welding arc. The system consists of six major units: a welding head, a hydraulically-operated line-up clamp, a self-drive mechanism, an imager controller, a system controller and a welding power supply unit. The welding head has a welding torch, a weaving mechanism, charge couple device cameras and a laser beam projector, and welding operation can be remotely controlled with an arc monitoring system. An image sensor/processor system is used to locate the center of the weld seam and the power supply unit is characterized by a fuzzy control system that stabilizes the welding arc [4].

Even though FLC has been applied widely in various classical as well as complicated engineering problems not much work has been done in the application of FLC to curing processes of FRP composite. In this study an attempt has been made to apply the FLC to solve the critical problem of nonlinear heat diffusion with chemical kinetics in the curing of FRP composite systems. This approach is required in order to control temperature associated with the curing process.

## NOMENCLATURE

K: Thermal diffusivity, m<sup>2</sup>/sec  
k: Thermal conductivity, W/m K  
C<sub>p</sub>: The specific heat capacity, J/Kg K  
h: Convective heat transfer coefficient, W/m<sup>2</sup> K  
r: Radial dimension, m  
z: Axial dimension, m  
i: Radial grid dimension  
j: Axial grid dimension  
t: Time, sec

N: Iteration number  
T: Temperature, °C  
 $\dot{g}$ : Thermal energy generation term, W/m<sup>3</sup>  
E: Error  
Err: Error change  
T<sub>o</sub>: Observed temperature, °C  
T<sub>at</sub>: Target temperature, °C  
ΔT<sub>c</sub>: Control temperature, °C  
N: Negative  
NB: Negative Big  
NVB: Negative Very Big  
Z: Zero  
P: Positive  
PB: Positive big  
PVB: Positive very big  
c<sup>r</sup>: Center of gravity value for rule r  
x<sub>i</sub>: The crisp value for input i to the fuzzy logic controller  
ρ: Density, Kg/m<sup>3</sup>  
α: Degree of curing  
β: Thermal conductivity constant  
η: Specific heat constant  
μ: Membership value

## PLANT MODEL

Figure 1 shows a schematic representation of HAC pipes. The pre-impregnated fiber and organic matrix material is selected to have the same properties as the FRP composite pipe being joined after it has cured. It is assumed that both the pipe and joint section have uniform, isotropic and temperature dependent thermo-physical properties. Therefore, under these assumptions the differential equation that governs the transport of thermal energy through the composite system wall with internal heating is the heat diffusion equation in two dimensions with energy generation is nonlinear and nonhomogenous as in [5]:

$$\rho C_p(T) \frac{\partial T}{\partial t} = \frac{1}{r} \frac{\partial}{\partial r} \left( k(T) r \frac{\partial T}{\partial r} \right) + \frac{\partial}{\partial z} \left( k(T) \frac{\partial T}{\partial z} \right) + \dot{g}(T) \quad (1)$$

The thermal properties C<sub>p</sub>(T) and k(T) are linear function of temperature as follows [6]:

$$c_p = c_p + c_{po} \eta T \quad (2)$$

$$k = k + k_{po} \beta T \quad (3)$$

Thermal energy generation due to chemical kinetics of the curing prepreg was accounted for using an autocatalytic kinetic model as follows [6]

$$\dot{g} = \rho H_r Z e^{-E/RT} \alpha^m (1 - \alpha)^n \quad (4)$$

## Initial conditions

The pipes and the prepreg are initially at room temperature.  
 $(T_1 = T_2 = T_3)_{t=0} = T_0$  (5)

Temperature Boundary Conditions

The boundary conditions used in this study is derived from manufacturers cure cycle required for the prepreg material:

Prepreg Boundary:

$$T = \begin{cases} At & 0 \leq t \leq t_1 \\ At_1 & t_1 \leq t \leq t_2 \\ -\frac{At_1}{t_3 - t_2}(t_3 - t) & t_2 \leq t \leq t_3 \end{cases} \quad (6)$$

Radial Outer Boundaries ( -a < z < a):

$$-k_1 \frac{\partial T_i}{\partial t} + hT_i = hT_\infty \quad (7)$$

where  $T_0$  is the initial temperature of the joint.  $T_1, T_2, T_3, h, k_1, k_2,$  and  $k_3$  are temperature, convective heat transfer coefficient, and thermal conductivity in the prepreg, fiberglass composite pipe, and the alloy pipe, respectively.  $T_\infty$  is the environmental temperature.

Interfacial condition

$$\begin{aligned} (k_1 \frac{\partial T_1}{\partial r})_{r=R_1} &= (k_2 \frac{\partial T_2}{\partial r})_{r=R_1} & (0 < z < a) \\ (T_1)_{r=R_1} &= (T_2)_{r=R_1} & (0 < z < a) \\ (T_2)_{r=R_0} &= (T_3)_{r=R_0} & (0 < z < b) \\ (k_2 \frac{\partial T_2}{\partial r})_{r=R_0} &= (k_3 \frac{\partial T_3}{\partial r})_{r=R_0} & (0 < z < b) \end{aligned} \quad (8)$$

Symmetrical condition

$$\begin{aligned} (\frac{\partial T_1}{\partial z})_{z=0} &= (\frac{\partial T_2}{\partial z})_{z=0} = (\frac{\partial T_3}{\partial z})_{z=0} = 0 \\ (\frac{\partial T_3}{\partial r})_{r=0} &= 0 \end{aligned} \quad (9)$$

**FUZZY LOGIC CONTROL SYSTEM**

In this section a fuzzy logic model is constructed for a proper curing process as explained in the introduction of this study. A FLC is composed of three main parts, namely fuzzyfication, inference engine and defuzzyfication [9]. In the fuzzyfication part, inputs are fuzzyfied according to the fuzzyfication method chosen. In this study singleton fuzzyfier is

used. Labels and their corresponding domains are defined based on expert knowledge.

In the inference engine part, rules that will be used to calculate the fuzzyfied control value are embedded into the F.L.C. model. These rules are determined based on the expert knowledge. In this study minimum inference engine is used.

In the defuzzyfication part, the fuzzyfied output obtained from the inference engine is converted to the crisp value according to the defuzzyfication method. In this study the center of gravity defuzzyfier method is used.

The block diagrams of the system are shown in Figures 3 and 4.

Rules:

Error is defined as:

$$E = T_i - T_o \quad (10)$$

Error change is defined as:

$$Er(n + 1) = \Delta E(n + 1) = E(n + 1) - E(n) \quad (11)$$

Rule-1: IF error is N AND error change is N THEN control temperature should be NVB

Rule-2: IF error is N AND error change is Z THEN control temperature should be NB

Rule-3: IF error is N AND error change is P THEN control temperature should be N

Rule-4, 5, and 6: IF error is Z THEN control temperature should be Z

Rule-7: IF error is P AND error change is N THEN control temperature should be P

Rule-8: IF error is P AND error change is Z THEN control temperature should be PB

Rule-9: IF error is P AND error change is P THEN control temperature should be PVB

Using singleton fuzzyfier, minimum inference engine and center of gravity defuzzyfier, the defuzzyfier output [10] will be:

$$\Delta T_c = \frac{\sum_{r=1}^9 c^r \min_{i=1}^2 \mu_{A_i^r}(x_i)}{\sum_{r=1}^9 \min_{i=1}^2 \mu_{A_i^r}(x_i)} \quad (12)$$

where

$c^r$ : Center of gravity value for rule r.

$x_i$ : Inputs to the fuzzy logic controller, E and Er.

$\mu_{A_i^r}(x_i)$ : Membership value of  $x_i$  for  $A^k$  label.

**ALGORITHM**

A finite difference program based on Alternating Direct Implicit (ADI) methods was developed in a parallel



**FUTURE STUDY**

For the optimum control parameters, advanced fuzzy logic control method, such as adaptive F.L.C., can be used. Moreover, the computation of required thermal energy to map the output of the controller i.e. the necessary boundary conditions of alloy side are to be evaluated for the smooth running of the proposed fuzzy control system. This aspect is currently under investigation.

**ACKNOWLEDGMENTS**

The authors would like to thank the National Aeronautics Space Administration via the Faculty Awards for Research programs for their support for this project under contract number NAG8-1536.

**REFERENCES**

[1] [http://www-dse.doc.ic.ac.uk/~nd/surprise\\_96/journal/vol2/jp6/article2.html](http://www-dse.doc.ic.ac.uk/~nd/surprise_96/journal/vol2/jp6/article2.html)  
 [2] <http://www.st.com/stonline/books/ascii/docs/2498.htm>  
 [3] Jang, J-S., Sun, C-T., "Neuro-Fuzzy Modeling and Control," *Proceedings of the IEEE*, March 1995, pp. 378-406  
 [4] N. Fukuhara, H. Ishii, K. Agusa, *Development of an inside/outside simultaneous welding robot for pipeline*,

Robot, n110, May, 1996 JIRA, Tokyo, Japan, pp.73-80, ISSN: 0387-1940.  
 [5] P.F., Mensah, O. Soysal, G., Li, *Modeling of heat-activated coupled composite pipes in a cryogenic thermal environment*, Proceeding of ETCE 2001: 23rd ASME Energy Sources Technology Conference and Exposition February 5-7, 2001, Houston, Texas, p.2  
 [6] P.F., Mensah, M., Stubblefield, S.S. Pang and D. Wingard, *Thermal Analysis Characterization of Fiberglass Epoxy Prepreg Used to Join Composite Pipes*, Polymer Engineering and Science, April 1999, Vol. 39, No. 4.  
 [7] P.F., Mensah, O. Soysal, A. Jana, M., Stubblefield, *Transient two-dimensional numerical modeling of asymmetric curing process*, Proceeding of ASME ETCE2002, ASME Engineering Technology Conference on Energy, February 4-6, 2002, Houston, TX  
 [8] L.X., Wang, *A course in fuzzy systems and control*, Prentice-Hall, 1997, p.7  
 [9] L.X., Wang, *A course in fuzzy systems and control*, Prentice-Hall, 1997, p.120

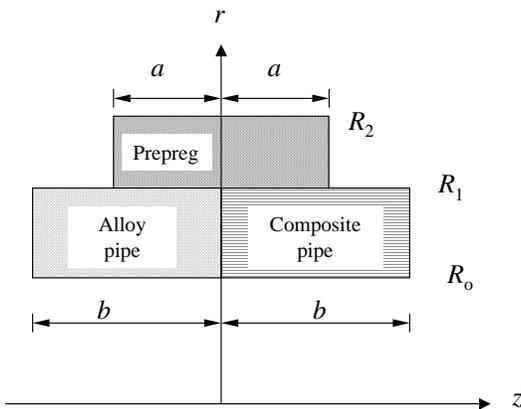


Figure 1 The Pipe Structure

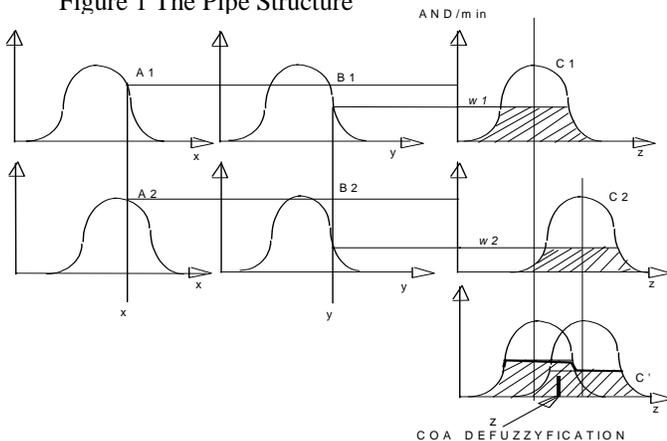


Figure 2 Fuzzy Inferencing; Mamdani Type

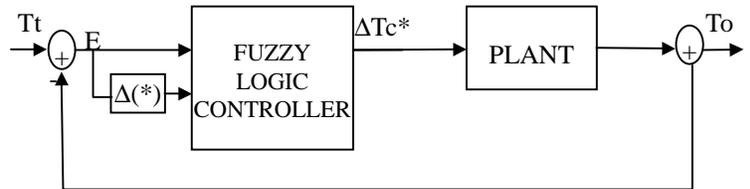


Figure 3 System Block Diagram

Table 1 Look Up Table

		ERROR CHANGE		
		N	Z	P
ERROR	N	$\Delta T_c$ : NVB	$\Delta T_c$ : NB	$\Delta T_c$ : N
		RULE-1	RULE-2	RULE-3
	Z	$\Delta T_c$ : Z	$\Delta T_c$ : Z	$\Delta T_c$ : Z
		RULE-4	RULE-5	RULE-6
	P	$\Delta T_c$ : P	$\Delta T_c$ : PB	$\Delta T_c$ : PVB
		RULE-7	RULE-8	RULE-9

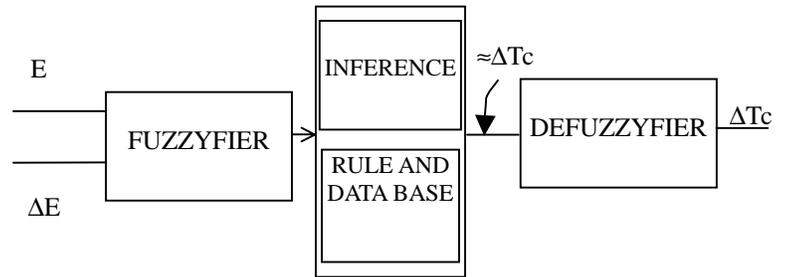


Figure 4 F.L.C. Block Diagram

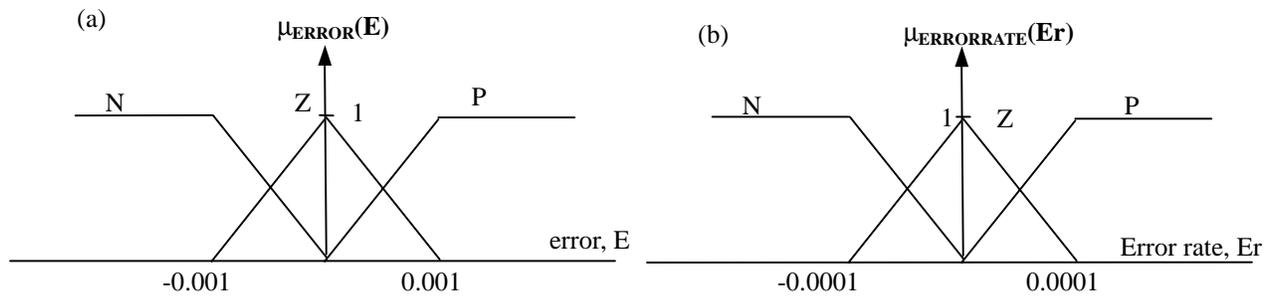


Figure 5 Input Membership Function: a) Error b) Error change

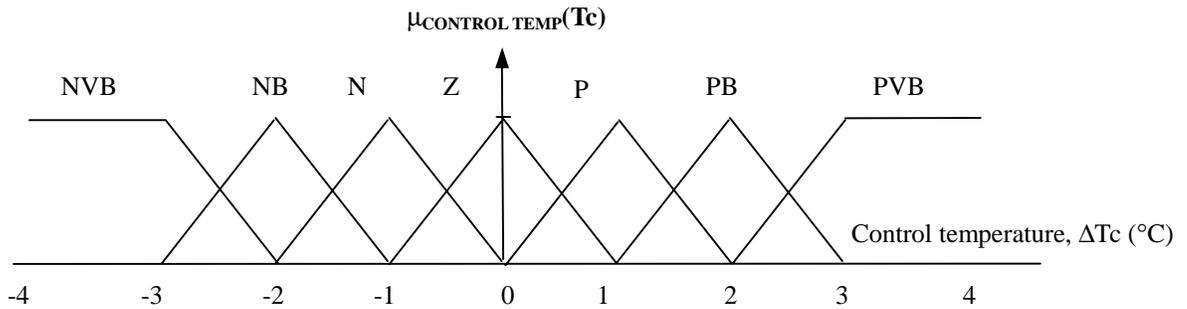


Figure 6 Output Membership Function (normalized by 4/0.0047)

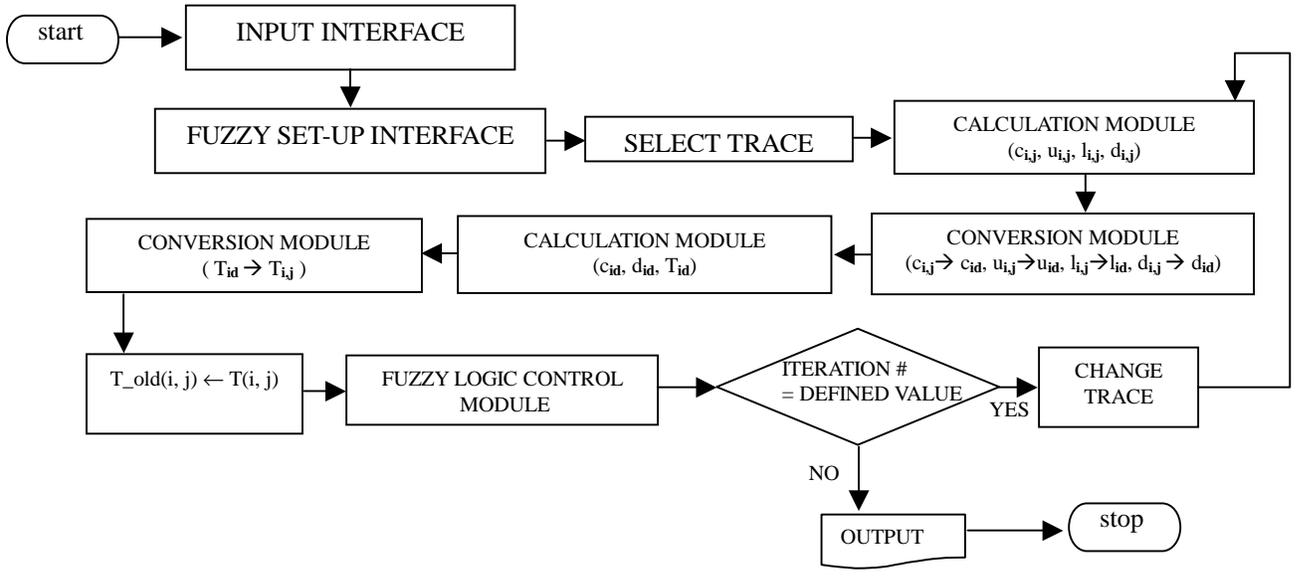
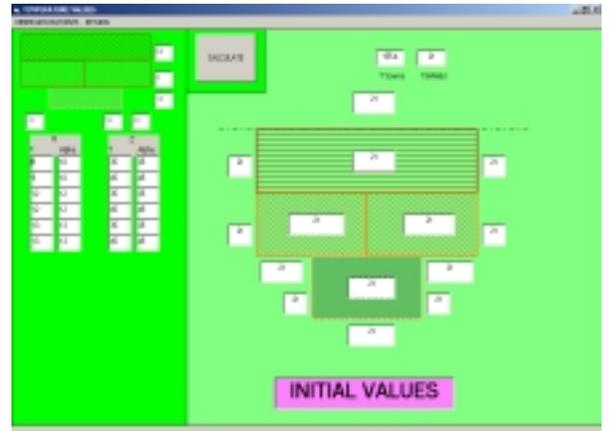


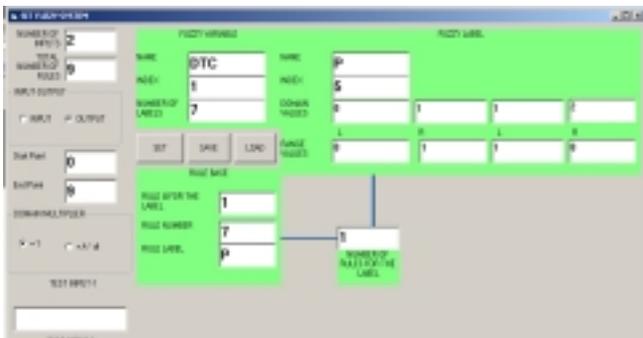
Figure 7 Flowchart of the Algorithm



(a)



(b)



(c)

Figure 8 User Interface of The Heat Control Simulator: a and b) Input Screen c) Fuzzy System Set-up Screen

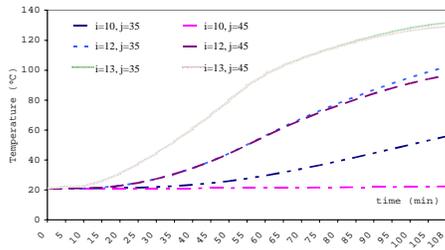


Figure 9 Radial Temperature History of Prepreg Curing Process without Control

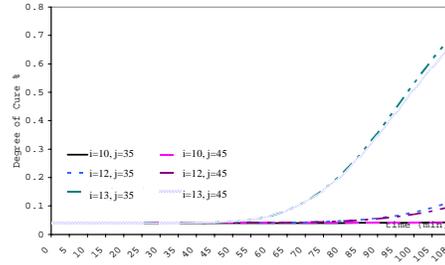


Figure 10 Radial Degree of Cure of Prepreg on Alloy and Composite Pipe without Control

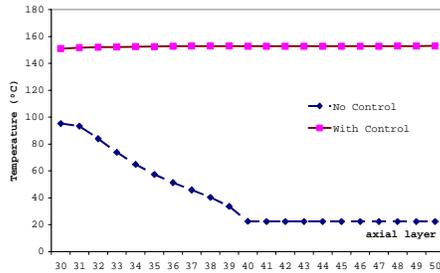


Figure 11 Comparison of Temperature Distribution at the Joint Interface with and without control

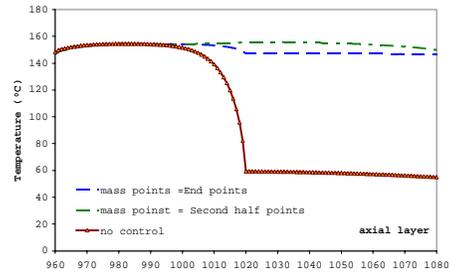


Figure 12 Comparison of Temperature Distribution at the Joint Interface with and without control for a Larger Size

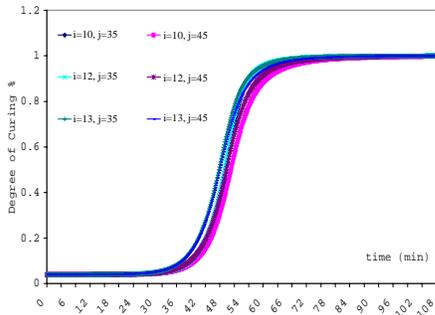


Figure 13 Radial Degree of Cure of Prepreg on Alloy and Composite Pipe For a Large Size Pipe with Control

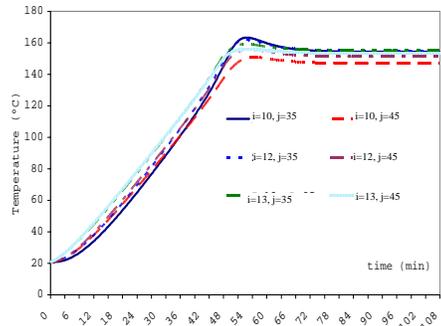


Figure 14 Radial Temperature History of Prepreg Curing Process For a Large Size Pipe with Control

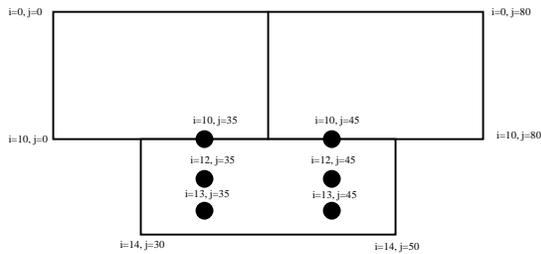


Figure 15 The Points for which the Temperature History and Degree of Curing Data are acquired