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# Selection of Best Materials and Parametric Optimization of Solar Parabolic Collector Using Fuzzy Logic

Sri P. Mohana Reddy\*, Pathi Venkataramaiah, Devuru Vishnu Vardhan Reddy

Department of Mechanical Engineering, SV University, Tirupati, India Email: \*mohanareddypichela@gmail.com

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# **Abstract**

This paper focused on selection of best materials for absorber tube and reflective surfaces of Solar Parabolic Collector (SPC) using fuzzy logic, after analysing the material data. The glass mirror and Aluminium absorber have been identified as best materials. These selected materials are replaced in existing experimental setup. An experimental design is prepared based on the considered parabolic collector parameters: Absorptivity, Reflectivity and Period of Sun Incidence. During experiments, outlet temperature of water and discharge is recorded for each experimental run. These data are analyzed using fuzzy Logic integrated with the Taguchi method and optimal parameter combination has been found.

# **Keywords**

Materials, Fuzzy Logic, Taguchi Method

#### 1. Introduction

Among the renewable sources of energy, solar energy offers a practical solution for the energy problem which is clouding the prospects of mankind. The use of fossil fuel causes global warming, air pollution and acid rain. One to this harnessing of non-conventional, renewable and environment-friendly energy resources are vital for steering the global energy supplies towards a sustainable path (Silio Salcines and Renedo Estebanez 2009) [1]. Solar thermal systems play an important role in providing non-polluting energy for domestic and industrial applications. Concentrating solar technologies (Kaplanis and Barakos 2002) [2], such as the parabolic collector, compound parabolic collector and parabolic trough can operate at high temperatures and are used to supply industrial

\*Corresponding author.

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P. Venkataramaiah (2013) [5], et al. presented the Simulation and optimisation studies on a solar parabolic collector, and focused on parametric optimisation of a Solar Parabolic Collector done Desirability Functional Analysis approach. Priyabrata (2013) [6], et al. applied fuzzy logic for selection of hydro turbine blade materials. For selection of best materials, various techniques such as Analytical Hierarchical Process (AHP), Fuzzy AHP, etc. are used by many researchers.

Mohana Reddy, *et al.* (2012) [7] proposed a methodology to find best materials for a solar flat-plate collector using the AHP technique. Oguzhan Yilmaz *et al.* (2006) [8] introduced a user-friendly intelligent system for the selection of electro discharge machining (EDM) parameters using MATLAB Fuzzy Logic Toolbox.

Tzeng and Chen (2007) [9] described the application of the fuzzy logic analysis coupled with Taguchi method to optimise the precision and accuracy of the high-speed electrical discharge machining (EDM) process. Rajyalakshmi *et al.* (2012) [10] used fuzzy logic integrated with Taguchi method for optimization of process parameters of wire electrical discharging machine. Sarfaraz Khabbaz *et al.* (2009) [11] applied a simplified fyzzy logic approach for material selection in mechanical engineering design. Vimal Sam Singh, *et al.* (2009) [12] worked on modelling and analysis of thrust force and torque in drilling GFRP composites by multi-facet drill using fuzzy logic. Zadeh, L. (1965) [13] explained the fuzzy sets, information and control. Kao C.C. *et al.*, (2008) [14] introduced *Fuzzy Logic Control of Micro-hole Electrical Discharge Machining*, Journal of Manufacturing Science and Engineering. Lin J. L. *et al.*, (2005) [15] used grey-fuzzy logic for the optimization of the manufacturing process.

Literature review reveals that no researcher uses the fuzzy logic for material selection and parametric studies in the area of solar energy systems. In the present study, fuzzy logic is used for selection of materials and optimisation of process parameters of a solar parabolic collector

# 2. Selection of Material for SPC Using Fuzzy Logic

# 2.1. Different Types of Materials and Its Properties

The different types of materials for absorber tubes and reflective surfaces are considered in this work and shown in **Table 1** and **Table 2**. These materials are chosen for component of SPC to meet all performance requirements, ensure a long working with minimal or no required maintenance, and be cost effective. The properties of materials: density, thermal conductivity, cost, coefficient of thermal expansion, temperature, Corrosion resistantence which are input to fuzzy logic.

Table 1. Material properties for reflecting surfaces.

| C      | Madadal            | Material properties |             |           |           |            |  |
|--------|--------------------|---------------------|-------------|-----------|-----------|------------|--|
| S. no. | Material           | Density (g/cm³)     | T.C (W/m k) | Corrosion | Temp (°C) | Price (Rs) |  |
| 1      | Stainless steel    | 7.93                | 16.2        | 9         | 115       | 65         |  |
| 2      | Glass mirror       | 2500                | 1.05        | 7         | 120       | 720        |  |
| 3      | Polished Aluminium | 2.70                | 215         | 6         | 110       | 129        |  |
| 4      | Sliver mirror film | 2100                | 1.15        | 10.3      | 100       | 60         |  |
| 5      | Acrylic mirror     | 2400                | 1.10        | 7         | 85        | 50         |  |

Table 2. Material Properties for absorber tube.

| S. no. | Material             |                 |             | Material properties |           |            |  |
|--------|----------------------|-----------------|-------------|---------------------|-----------|------------|--|
|        | Material             | Density (g/cm³) | T.C (W/m k) | Corrosion           | Temp (°C) | Price (Rs) |  |
| 1      | Aluminium tube       | 2.65            | 210         | 6                   | 125       | 110        |  |
| 2      | Glass tube           | 2500            | 1.05        | 7                   | 120       | 720        |  |
| 3      | Copper tube          | 8920            | 400         | 7.5                 | 90        | 440        |  |
| 4      | Stainless steel tube | 7.93            | 16.2        | 9                   | 120       | 65         |  |

#### 2.2. Material Selection Using Fuzzy Logic

Fuzzy logic is proved as effective tool for inferring or deducing information from a given set of facts and the reasoning process is simple, compared to computationally precise systems. Fuzzy logic can select the parameters or also materials rationally for optimizing complex combinations of technical and other properties which is hard to achieve manually. Hence, fuzzy logic theory can be used to select the optimum material for a function or optimum process parameters from a preranked groups based on their relevant properties. With these advantages the fuzzy logic is applied in this paper.

The fuzzy logic control allows the existence of uncertainty in handling parameter values. Fuzzy logic system comprises a fuzzifier, membership functions, a fuzzy rule base, an inference engine, and defuzzifier. The fuzzifier uses membership functions to fuzzify input values of each performance characteristic are shown in Figure 1.

The inference engine performs fuzzy reasoning on fuzzy rules to generate a fuzzy value. Finally, the defuzzifier converts fuzzy predicted value into a response value. In the following, the concept of fuzzy reasoning is described based on the five input-one-output fuzzy logic units. The fuzzy rule base consists of a group of if-then control rules with the five inputs, x1, x2, x3, x4 and x5 and one output y, i.e.

In the case of a material selection the total number of rules is equal to the number of fuzzy sets raised to the number of material properties being considered. Performance indices are calculated based on the material attributes. The indices are used to rank the performance of candidate materials from best to worst. In this paper the material properties such as density, thermal conductivity, coefficient of thermal expansion, corrosion resistant and cost are given as the input to fuzzy logic tool box to select best material for solar parabolic collector as shown in **Figure 2**.

Rule 1: if x1 is A1 and x2 is B1 and x3 is C1 and x4 is D1 and x5 is E1 then y is F1 else Rule 2: if x1 is A2 and x2 is B2 and x3 is C2 and x4 is D2 and x5 is E2 then y is F2 else

Rule n: if x1 is An and x2 is Bn and x3 is Cn and x4 is Dn and x5 is En then y is Fn else

Ai, Bi, Ci, Di and Ei are fuzzy subsets defined by the corresponding membership functions, *i.e.*,  $\mu$ Ai,  $\mu$ Bi,  $\mu$ Ci,  $\mu$ Di,  $\mu$ Ei. In this section, three fuzzy subsets are assigned to the five inputs, as shown in Figure 2.

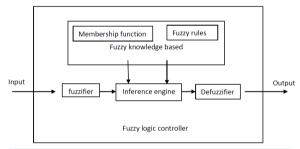


Figure 1. Fuzzy logic system.

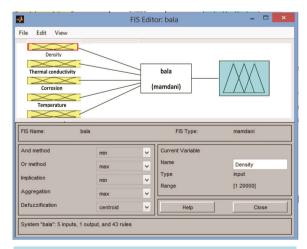


Figure 2. FIS for material selection.

It is observed from **Figure 3** that the fuzzy set for the density values between 5000 gm/cm<sup>3</sup> and 15000 gm/cm<sup>3</sup> are as: below 5000 gm/cm<sup>3</sup> is having excellent fuzzy set. Between 5000 - 15000 gm/cm<sup>3</sup> it is having good fuzzy set.

From **Figure 4**, the fuzzy sets for the corrosion resistance values between PREN (pitting resistance equivalent number) values 32 and 78 are as: above 78 it is having excellent fuzzy set. Between 32 - 78 it is having good fuzzy set.

From **Figure 5**, the fuzzy sets for the thermal conductivity values between 0.25 and 0.75 are as: above 0.75 it is having excellent fuzzy set, between 0.25 - 0.75, it is having good fuzzy set.

From **Figure 6**, the fuzzy sets for the temperature values between 75°C and 170°C are as: above 170°C it is having excellent fuzzy set, between 75°C - 170°C it is having good fuzzy set.

From Figure 7, the fuzzy sets for the cost values between 600 and 1500 are as: below 600 it is having excellent fuzzy set, between 600 - 1500 it is having good fuzzy set.

Five fuzzy subsets are assigned to the output, as shown in **Figure 8**. Forty three rules were developed based on their input values. By taking the max-min compositional operation, the fuzzy reasoning of these rules yields a fuzzy output that can be used to find the best material of SPC by giving rank to the response value.

#### 2.3. Fuzzy Evaluation Results

This section presents the results of the material selection for solar parabolic collector. The results obtained from fuzzy logic are shown in **Table 3** and **Table 4**. From these tables, the best materials Glass mirror for reflective

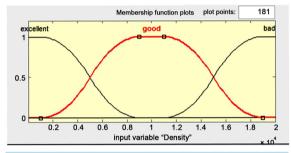


Figure 3. Membership function of density.

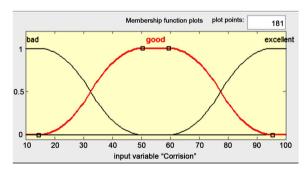


Figure 4. Membership function of corrosion.

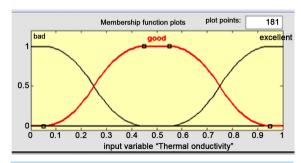


Figure 5. Membership function of thermal conductivity.

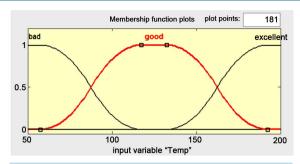


Figure 6. Membership function of temperature.

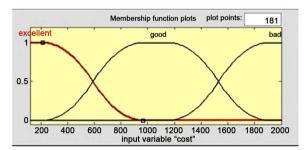


Figure 7. Membership function of cost.

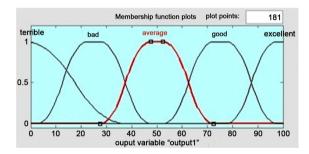


Figure 8. Membership function of output.

Table 3. Performance indices of reflective surfaces.

| S. no. | Material           | Performance indices | Rank |
|--------|--------------------|---------------------|------|
| 1      | Stainless steel    | 50                  | 2    |
| 2      | Glass mirror       | 89.5183             | 1    |
| 3      | Polished Aluminium | 50                  | 2    |
| 4      | Sliver mirror film | 21.5735             | 4    |
| 5      | Acrylic mirror     | 19.9873             | 5    |

Table 4. Performance indices of absorber tube.

| S. no | Material             | Performance indices | Rank |
|-------|----------------------|---------------------|------|
| 1     | Aluminium tube       | 90.8965             | 1    |
| 2     | Glass tube           | 89.5183             | 2    |
| 3     | Copper tube          | 18.5                | 4    |
| 4     | Stainless steel tube | 50                  | 3    |

surface and Aluminium for absorber tube are selected based on their ranks.

# 3. Experimental Work

## 3.1. Description of Experimental Setup

Experimental setup consists of

- Parabolic Shaped Structure (made of MS Flat).
- Two legs to support the structure.
- Reflective Surface.
- Absorber tube.
- Auto Tracking System.
- Storage tank (Inlet & Outlet).
- Piping system (**Figure 9**).

Based on the reference [5] experimental setup has been prepared for conducting experiments, using the best materials obtained fuzzy logic. The entire set is placed in the N-S direction to face the axis of the parabolic trough towards east. The parabolic structure has a flexibility to change the reflective surface and Absorber tube materials. A provision has been made to change the position of absorber tube away from the focal point and also a mechanism is provided to change the angle of the absorber. A hydraulic actuator coupled with Light sensor is attached to the system for Auto-tracking. Light sensor senses the intensity of light and give signal to servo switch intern the signal is given to the hydraulic actuator.

#### 3.2. Process Parameters and Their Levels

The process parameters considered in this work are: Absorptivity, Reflectivity and Period of Sun Incidence (POI). The Absorptivity is considered at three levels by coating with three coats; three Coating, are prepared by mixing silicon, black nickel and chromium with black enamel paint one at a time respectively and used for coating of best absorber tube which is selected through fuzzy logic. The reflective surface is considered at two levels, one is Glass mirror and other is silver mirror film. The process parameters and their levels are shown in **Table 5**.

# 3.3. Design of Experiment and Experimental Data

A Taguchi Experimental  $L_{18}$  design has been prepared based on the data of **Table 5**. Experiments are conducted according to the Taugchi design and the experimental results temperature and discharge are recorded as shown in **Table 6**.



Figure 9. Experimental setup of solar parabolic collector.

Table 5. Process parameters and their levels.

| S. no. | <b>Process Parameters</b>   | Level 1                                   | Level 2                                      | Level 3                                   |
|--------|-----------------------------|---|--|---|
| 1      | Reflectivity (A)            | Silver Mirror Film (SMF) (0.95)           | Glass Mirror (GM) (0.97)                     | -   |
| 2      | Absorptivity (B)            | U-Tube Aluminium<br>Coated with Si (0.83) | U-Tube Aluminium Coated with Black Ni (0.93) | U-Tube Aluminium<br>Coated with Cr (0.97) |
| 3      | Period of Sun Incidence (C) | 9.00 AM - 11.00 Am                        | 11.00 AM - 01.00 PM                          | 01.00 PM - 04.00 PM                       |

Table 6. Experimental results.

|          |      | Parameters |   | Experime         | ntal results       |
|----------|------|------------|---|------------------|--------------------|
| Exp. run | A    | В          | С | Temperature (°C) | Discharge (lit/hr) |
| 1        | 0.95 | 0.83       | 1 | 60               | 10                 |
| 2        | 0.95 | 0.83       | 2 | 70               | 7                  |
| 3        | 0.95 | 0.83       | 3 | 45               | 22                 |
| 4        | 0.95 | 0.93       | 1 | 60               | 12                 |
| 5        | 0.95 | 0.93       | 2 | 55               | 14                 |
| 6        | 0.95 | 0.93       | 3 | 45               | 20                 |
| 7        | 0.95 | 0.97       | 1 | 50               | 16                 |
| 8        | 0.95 | 0.97       | 2 | 60               | 12                 |
| 9        | 0.95 | 0.97       | 3 | 45               | 20                 |
| 10       | 0.97 | 0.83       | 1 | 50               | 18                 |
| 11       | 0.97 | 0.83       | 2 | 70               | 9                  |
| 12       | 0.97 | 0.83       | 3 | 55               | 15                 |
| 13       | 0.97 | 0.93       | 1 | 60               | 13                 |
| 14       | 0.97 | 0.93       | 2 | 65               | 10                 |
| 15       | 0.97 | 0.93       | 3 | 63               | 11                 |
| 16       | 0.97 | 0.97       | 1 | 62               | 10                 |
| 17       | 0.97 | 0.97       | 2 | 50               | 14                 |
| 18       | 0.97 | 0.97       | 3 | 45               | 18                 |

#### 4. Determination of Optimum Process Parameters Using Fuzzy Logic

A fuzzy logic unit comprises a fuzzifier, membership functions, a fuzzy rule base, an inference engine, and a defuzzifier. First, the fuzzifier uses membership functions to fuzzify the S/N ratios. Next, the inference engine performs a fuzzy reasoning on fuzzy rules to generate a fuzzy value. Finally, the defuzzifier converts the fuzzy value into a MRPI. In the following, the concept of fuzzy reasoning is described briefly based on the two input-one-output fuzzy logic units. The fuzzy rule base consists of a group of if-then control rules with the two inputs, x1, x2 and one output y, i.e.

Rule 1: if x1 is A1 and x2 is B1 then y is C1 else

Rule 2: if x1 is A2 and x2 is B2 then y is C 2 else

Rule n: if x1 is An and x2 is Bn then y is C n.

Ai, Bi and Ci are fuzzy subsets defined by the corresponding membership functions, *i.e.*,  $\mu$ Ai,  $\mu$ Bi and  $\mu$ Ci. In the present work, three fuzzy subsets are assigned to the two inputs, as shown in **Figure 10** and **Figure 11**.

Five fuzzy subsets are assigned to the output, as shown in **Figure 12**. Nine fuzzy rules were developed based on the fact that a higher S/N ratio gives better performance. By taking the max-min compositional operation, the fuzzy reasoning of these rules yields a fuzzy output. Finally, a defuzzification method, called the centre of gravity method. Is adopted to transform the fuzzy inference output  $\mu_D$   $(Y_1, Y_2)$  into a non-fuzzy value  $Y_0$  [3], which is known as the crisp output, that is calculated with the help

$$MRPI(Y_0) = \frac{\sum Y \mu D(Y1)}{\sum \mu D(Y2)}$$

The non-fuzzy value  $Y_0$  gives MRPI. Invariably, a larger MRPI is preferred, which gives a better performance characteristic. **Table 7** shows the results of the S/N ratios and MRPI for different experiment.

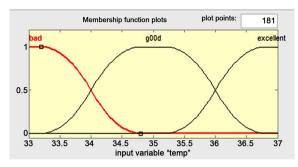


Figure 10. Membership function for S/N ratios of temp.

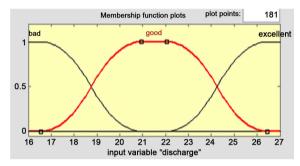


Figure 11. Membership function for S/N ratios of discharge.

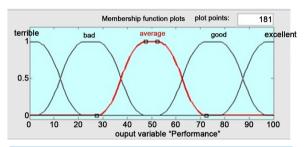


Figure 12. Membership function for S/N ratios of MRPI.

#### 4.1. Analysis and Discussion on Experimental Results Using MRPI

The experimental scheme used in this work is based on Taguchi's orthogonal array, by which it is possible to separate the effect of each process parameter on the MRPI at different levels. The influence of each process parameter can be more clearly presented in means of the data mean response graph. The data mean graph shows the change in the response when a given factor goes from level 1 to level 3.

The optimum process parameters of the SPC are identified as shown in **Figure 13** and **Table 8**. The absorptivity influences more and the Period of Sun Incidence has less influence on the responses.

#### **Optimum parameters levels:**

It is observed that the optimum combination of influential parameters is A2 B2 C2. Which means:

**A2**: Reflectivity at level 2 (Glass Mirror) (0.97).

**B2**: Absorptivity at level 2 (U-tube Aluminium coated with Black Ni) (0.93).

C2: Period of Sun Incidence at level 2 (11.00 A.M - 1.00 P.M).

#### 4.2. Confirmation Test

The optimum parameters obtained from fuzzy logic are Reflectivity (A2 at level 2 -0.97), Absorptivity (B2-0.93), Period of Sun Incidence (C2-11.00 AM-1.00 PM) Ref section 4.1. An experiment is conducted for this combination of parameters for confirmation. The confirmation test (experiment) results are given in **Table 9** and the conformation test result is satisfactory.

Table 7. S/N ratios and MRPI.

| Exp. run | S/N ratios of temp (Y <sub>1</sub> ) | S/N ratios of discharge (Y <sub>2</sub> ) | MRPI (Y <sub>0</sub> ) |
|----------|--------------------------------------|---|------------------------|
| 1        | 35.5630                              | 20.000                                    | 50.2605                |
| 2        | 36.9020                              | 16.9020                                   | 50.4308                |
| 3        | 33.0643                              | 26.5836                                   | 50                     |
| 4        | 35.5630                              | 21.5836                                   | 53.1153                |
| 5        | 34.8073                              | 22.9226                                   | 52.431                 |
| 6        | 33.0643                              | 26.0206                                   | 49.3673                |
| 7        | 33.9794                              | 24.0824                                   | 48.4023                |
| 8        | 35.5630                              | 21.5836                                   | 53.1153                |
| 9        | 33.0643                              | 26.0206                                   | 49.3673                |
| 10       | 33.9794                              | 25.1055                                   | 55.875                 |
| 11       | 36.9020                              | 19.0849                                   | 65.6235                |
| 12       | 34.8073                              | 23.5218                                   | 56.2179                |
| 13       | 35.5630                              | 22.2789                                   | 53.1504                |
| 14       | 36.2583                              | 20.0000                                   | 64.6842                |
| 15       | 35.9868                              | 20.8279                                   | 62.0754                |
| 16       | 35.8478                              | 20.0000                                   | 55.5266                |
| 17       | 33.9794                              | 22.9226                                   | 39.886                 |
| 18       | 33.0643                              | 25.1055                                   | 44.6953                |

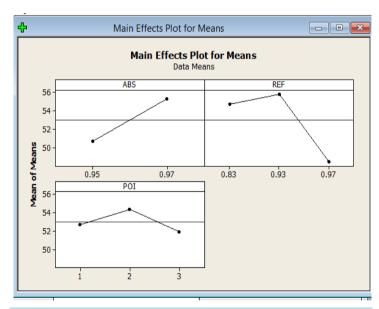


Figure 13. Main effects for data means.

Table 8. Optimum process parameters.

| Level | A     | В     | С     |
|-------|-------|-------|-------|
| 1     | 34.10 | 34.72 | 34.43 |
| 2     | 34.75 | 34.89 | 34.59 |
| 3     |       | 33.66 | 34.26 |
| Delta | 0.65  | 1.23  | 0.32  |
| Rank  | 2     | 1     | 3     |

Table 9. Confirmation test results.

| Exp. Run |      | Parameters |     | Confirmation E   | xperiment results  |
|----------|------|------------|-----|------------------|--------------------|
| Exp. Kun | A    | В          | C   | Temperature (°C) | Discharge (lit/hr) |
| 1        | 0.97 | 0.93       | 2.0 | 65               | 14                 |

#### 5. Conclusion

In this work, fuzzy logic is used for the material selection of Solar Parabolic Collector and for identification of process parameter by analyzing the data related to material properties such as price, thermal conductivity, density, temperature and corrosion. It is identified that Glass mirror and Aluminium tube are the best reflecting material and best absorbing material respectively. Finally, experimental response data are analyzed using fuzzy logic and optimum parameters levels have been identified. And these levels are confirmed experimentally.

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