Fuzzy Logic Implementation of Photo Catalytic Sensor

Abstract
This paper presents novel method to control output error of the Photo Catalytic Sensor (PCS) using fuzzy interface. Whig and Ahmad in 2014, proposed the spice model of PCS which is a type of ion sensitive FET and very useful to estimate the parameter BOD (Biological Oxygen Demand) which is generally used to estimate quality of water. PCS can also be used to calculate the amount of concentration level of oxygen required to purify the polluted air and sanitization of surface. The purpose to control the output error due to variation in temperature is due to the fact that the fluctuation of O2 influence the threshold voltage, which is internal parameter of FET can manifest itself as a voltage signal at output but as a function of trans conductance gain. Hence, a system controlled by fuzzy logic which controls the output error is proposed. This novel method is very useful because of their simplicity, robustness and successful practical applications. The proposed system is relies on the fast operation of PCS which operates under sub-threshold conditions and reduced computation time. The system is more users friendly and the outcomes of simulation are fairly in agreement with the theoretical estimation.

Keywords: Photo catalytic sensor; Field effect transistor; Fuzzy rules; Fuzzy interface

Abbreviations: PCS: Photo Catalytic Sensor; FET: Field Effect Transistor; TiO2: Titanium Oxide; FIE: Fuzzy Inference Engine

Introduction

Environment plays an important role on the outputs of electronic sensors and systems. Electronic sensors used for very precise measurements perform very well under lab environment [1-2]. Semiconductor technology is very popular and widely used for sensor development as it provides an advantage of low power, high speed, small size integration and their signal processing capability. More often, CAD tools are used (Sze, 1994) which provide simulation and synthesis of semiconductor sensors. The SPICE has built in models for most semiconductor devices but there is scarcity of appropriate models for semiconductor sensors. Modelling of MOSFETS [3-4] requires a deep knowledge of the code structure, subroutines and it is strictly linked to a particular version of SPICE. The simplest PCS is O2 sensitive where the sensitive surface is made up of insulator layer like titanium oxide (TiO2) exposed to an electrolyte solution [5].

In these ion sensitive types of sensors the output change some times is deterministic in nature and sometimes random.

The sensor output recorded is not consistent with the external parameter variations and it changes randomly [7]. If the sensor output variation is linear or it following some pattern then we can use the linear model which is able to fit to get the desired output but this does not happen most of the time, this will not work as the output behaviour is nonlinear and non-deterministic [8]. Therefore it is desirable to devise a model which can improve the performance based on mathematical or artificial intelligence techniques. The scope of this paper is to study the behaviour of sensor output signal variations to minimize the output error generated due to temperature variation [9]. In this paper, fuzzy rule base approach has been implemented and it is compared with mathematical curve fitting technique. Fuzzy technique incorporated to solve this kind of engineering problem has produced good results. This approach for modelling random error drift added a new dimension to this kind of engineering problem.

Photo Catalysis Process

The process of photo catalysis is a proficient method for degrading organic compounds. Various literatures are available on the different mechanisms and equations involved in the process for gaining a better knowledge [10]. The semiconductor material consists of two bands which are valence band and conduction band. The energy gap between these two bands is known as band gap given by Eg. The electrons from the valence band jump to conduction band which may be empty when a light of energy higher than band gap energy falls on the semiconductor material. Holes are left behind in the valence band due to excitation of electrons to higher energy band. These holes on reaching the surface of the organic molecule react with water to give OH-radicals for oxidizing the organic pollutants. The dissolved oxygen
in the molecular form acts as a scavenger of the photo generated electrons and forms a superoxide radical ion. Titanium oxide has the ability to cause photo-oxidative destruction of the organic pollutants and is non-corrosive in nature due to which it is used as a catalyst in the process [11]. The oxygen content in any given sample can be determined by observing the change in dissolved oxygen concentration during the process of photo catalysis. In Photo catalysis process a floating gate electrode is used. The sunlight or UV radiations fall on TiO$_2$ which further act as a catalyst to speed up the photo catalysis process. PCS senses the changes in the oxygen concentration and its voltage levels change as an indication. The complete Photo catalysis process is shown in Figure 1.

**Photo Catalysis Sensor**

The SPICE model for PCS is given in [12]. It is basically a MOSFET having structural difference in which the gate terminal is kept inside the solution and diffusion and quantum capacitances are added to overcome the effect of Helmholtz and diffusion layer [13-15]. The cross section of PCS is shown in Figure 2. The threshold voltage equation for the PCS model is given as:

$$V_{th}(PCS) = E_{ref} - \Psi_{sol} + X_{oxd} + \frac{\Phi_s + Q_{ox} + Q_B}{C_{ox}} + 2\Phi f$$  

$\Psi_{sol}$ is an input parameter of the equation which is dependent on the concentration of $O_2$ in the solution and surface dipole potential $\chi^*$. Here $E_{ref}$ is the constant reference electrode potential. For different concentrations of $O_2$ different V-I curves for PCS can be plotted. $\Psi_{sol}$ is a function of $O_2$, and as the saturation cut-off current $I_{ds}$ increases the value of the oxygen concentration level decreases. The circuit for PCS as given in [14] is shown in Figure 3.

Here $C_q$ is the resultant of $C_{ox}$ and $C_q$ which are oxide and quantum capacitances respectively. The equivalent capacitance $C_M$ is given as:

$$\frac{1}{C_M} = \frac{1}{C_q} + \frac{1}{C_{ox}}$$  

The drain current equation in non-saturation mode for PCS is given as:

$$I_{ds} = C_{ox} \mu \frac{W}{L} \left[ (V_{gs} - V_f) V_{ds} - \frac{1}{2} V_{ds}^2 \right]$$  

Where

- $C_{ox}$: Oxide capacitance per unit area,
- $\mu$: Mobility of electrons in the channel,
- $W$: Channel width,
- $L$: Length of the Channel

Various Process Parameters including length of channel and channel width are chosen according to the 120 nm CMOS process model. According to the characteristics of the MOSFET gate to source voltage, $V_{gs}$ known as reference voltage drain current is allowed to vary with drain to source voltage keeping reference voltage constant. Comparing PCS with MOSFET keeping the concentration of $O_2$ = 1mg/l it is found that the curve resembles with the characteristic $V_{ds}/I_{ds}$ curve of MOSFET keeping $V_{gs}$ constant. Now keeping the reference voltage $V_{gs}$ = 0 it is observed that for different concentration levels of $O_2$ different $V_{ds}/I_{ds}$ curves are obtained as shown in Figure 4. From the above it is observed that as the oxygen concentration level decreases saturation cut off current $I_{ds}$ increases hence it is concluded that PCS can be treated as MOSFET on the basis that the chemical input parameter $\Psi_{sol}$ is a function of $O_2$ ($\Psi_{sol} = f(Oxygen)$). For the different values of oxygen content the curves between $I_{ds}$ and $V_{ds}$ is shown in Figure 4.
Fuzzy Logic Implementation of Photo Catalytic Sensor

Fuzzy Implementations

Fuzzy approach is based upon IF-THEN fuzzy rules. Fuzzy systems can have multiple inputs and single output system. Fuzzy system consist of four major components i.e., fusilier, fuzzy rule base, fuzzy inference engine and at output defuzzifier [15]. Each component has its own role. For example, fuzzy rule base contains the fuzzy IF-THEN rules. Fuzzy inference engine (FIS) using inference method maps the fuzzy sets in the input space to the output space using fuzzy logic. Fusilier and defuzzifier transform the input variables to the fuzzy sets and vice versa respectively [16]. Fuzzy Rule Base implementation requires first defining the linguistic variables [9, 10]. Based upon our acquired data of photo catalytic sensor, the declaration of fuzzy variables to the inputs and outputs is defined. It is clear from the Table 1 that the temperature varies between 10 to 45 °C. Hence, we have defined the temperature range into three categories i.e., Low, Medium and High as given in Table 1. The second input to the fuzzy system is sensor output data in microvolt (µV) which act as the input to get the desired output or corrected output. Since the sensor data varies in between 5 to 20 µV range as depicted in the Table 1. The third column in the Table 1 shows the fuzzy label for sensor output data range.

Table 1: Fuzzy Input Labels.

| Range | Temperature Range (°C) | Sensor Data Range (µV) |
|-------|------------------------|------------------------|
| Low   | Low (10-20)            | L-SEN(5-10)            |
| Medium| Medium(20-30)          | M-SEN(7-17)            |
| High  | High(30-45)            | H-SEN(10-20)           |

The output from the fuzzy system will be the desired output or corrected output. Our aim is that sensor output does no change with respect to temperature variation rapidly. It should behave like at room temperature or at Low temperature range. So corrected output should lie in the “L-SEN” range 5 to 10 µV. We have defined the sensor output range in Low (L-SEN) and Medium (M-SEN) for sensor data range as shown in Table 2. Based upon these fuzzy label declarations, triangular membership function has been chosen and defined. The other membership functions like Gaussian and tepozodial can also be defined. The Input membership function of the Sensor input data is shown below in the Figure 6. Similarly output membership function is defined and it is shown in Figure 7. Based upon the input output membership functions, the fuzzy rules [7] which will give the desired output from the fuzzy system. Nine possible combinations have been considered at the input side because there are three membership functions for each input. These are listed below which are capable of producing the desired output. Since the two inputs occur simultaneously, so AND operator has been chosen in IF statements.

R1: If temperature is low [10-20] AND sensor output is low [5-10] THEN final output is low [5-10]
R2: If temperature is medium [20-30] AND sensor output is low [5-10] THEN final output is low [5-10]
R3: If temperature is high [30-45] AND sensor output is medium [7-17] THEN final output is medium [20-30].

Table 2: Fuzzy Output Labels.

| Range | Sensor Data Range (µV) |
|-------|------------------------|
| Low   | Low Output(10-20)      |
| Medium| Medium Output(20-30)    |

Figure 4: $I_{ds}$ Vs $V_{ds}$ curve for different value of oxygen content.

Figure 5: Block diagram of Fuzzy Logic Implementation.

Figure 6: Input Membership function.
Principle of fuzzy based PCS System

The performance specifications of the system can be improved by tuning value of various parameters. By self-tuning it mean the characteristics of the controller to change or adjust its controlling parameters on-line automatically so as to have the most appropriate values of these parameters, which help system to get desired value. Fuzzy self-tuning works on the basis of control rules, which can be obtained by theoretical and experimental analysis of any system of expert. Thus, fuzzy logic can tune the internal parameters with the help of rule base on–line. This provides better performance than the conventional or simple fuzzy controller.

Design of self-tuning Fuzzy PID Controller

The self-tuning Fuzzy controller takes “error” \( e \) and “rate of change of error” \( ec \) as input to fuzzy logic controller and modify value of three parameters “L-SEN”, “M-SEN” and “H-SEN” online as output. Thus we have total of five linguistic variables \( e, ec, L-SEN, M-SEN, H-SEN \). Total of seven fuzzy value NB, NM, NS, ZO, PS, PM, PB are chosen for each of the linguistic variable. The region \( e \) and \( ec \) are between -3 to 3 whereas by doing some interpolation region of L-SEN, M-SEN and H-SEN are kept between 0 and 1. Interpolation tries to keep the value of the variable within specified region. The linguistic rules are important part of FIS. These rules are called the rule base. These rules are created with the help of human knowledge and expertise upon behaviour of the system under different condition. Any number of such rules can be formed to give the controller direction for action.

Results and Discussion

Rules for fuzzy interface system design for different values of L-SEN, M-SEN and H-SEN are shown in Table 3-5. Thus in a fuzzy interface system design there are total of two input as “e” and “ec” and three output as L-SEN, M-SEN and H-SEN as shown in Figure 8. Hence by these three tables total 49 rules are formed in FIS using MATLAB. Two input three outputs Fuzzy inference system has been designed. Inputs are “error” and “rate of change of error” \( ec \) and outputs are three parameter L-SEN, M-SEN and H-SEN of controller. The transient analysis of Fuzzy system design is shown in Figure 9. The output of fuzzy interface is linear shows that the sensor system is stable and the system is independent from the impact of external parameters. The surface plots corresponding to L, M and H-SEN is plotted with respect to two parameters \( e \) and \( ec \) as shown in the Figure 10-12. It is observed for the figures that the L-SEN (sensor output range at low temperature) has a little effect of change in \( e \) and \( ec \) value.

Table 3: Rule base for L-SEN.

| e/ec | NB | NM | NS | ZO | PS | PM | PB |
|------|----|----|----|----|----|----|----|
| NB   | PB | PB | PM | PM | PS | ZO | ZO |
| NM   | PB | PB | PM | PS | ZO | NS | NS |
| NS   | PM | PM | PS | ZO | NS | NM | NM |
| ZO   | PM | PM | PS | ZO | NS | NM | NM |
| PS   | PS | PS | ZO | NS | NM | NM | NB |
| PM   | PS | ZO | NS | NM | NM | NB | PB |
| PB   | ZO | ZO | NM | NM | NB | PB | PB |

Table 4: Rule base for M-SEN.

| e/ec | NB | NM | NS | ZO | PS | PM | PB |
|------|----|----|----|----|----|----|----|
| NB   | NB | NB | NM | NM | NS | ZO | ZO |
| NM   | NB | NB | NM | NS | NS | ZO | ZO |
| NS   | NB | NM | NS | NS | ZO | PS | PS |
| ZO   | NM | NM | NS | ZO | PS | PM | PM |
| PS   | NM | NS | ZO | PS | PS | PM | PB |
| PM   | ZO | ZO | PS | PS | PM | PB | PB |
| PB   | ZO | ZO | PS | PM | PM | PB | PB |
Conclusion

The error control using fuzzy logic interface is presented in this paper. Temperature variation in the environment where these sensors installed degrades the performance due to change in the properties of material. Data has been acquired from sensor under temperature varying environment. The output of the sensor has shown non-deterministic and non-linear behaviour with respect to temperature change. We have developed the fuzzy based technique to model and minimize the output variation. Fuzzy approach has shown better results with Sugeno model as compared to the Mamdani model. It means that fuzzy approach has ability to produce much better results as compared to other mathematical techniques. In future, Genetic Algorithm can also be implemented to check the results with fuzzy system as a comparison. Also, this study may be extended for further improvements in terms of power and size, besides the wiring and layout characteristics level.

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