Optimization of Bi$_2$O$_3$, TiO$_2$, and Sb$_2$O$_3$ Doped ZnO-Based Low-Voltage Varistor Ceramic to Maximize Nonlinear Electrical Properties

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In ZnO-based low voltage varistor, the two essential features of microstructure determining its nonlinear response are the formation Bi-enriched active grain boundaries as well as a controlled ZnO grain size by secondary spinel-type phases. Besides, the microstructure and phase composition are strongly affected by the dopant concentration during sintering process. In this study, the optimal dopant levels of Bi$_2$O$_3$, TiO$_2$, and Sb$_2$O$_3$ to achieve maximized nonlinear electrical property (alpha) were quantified by the response surface methodology (RSM). RSM was also used to understand the significance and interaction of the factors affecting the response. Variables were determined as the molar ratio of Bi$_2$O$_3$, TiO$_2$, and Sb$_2$O$_3$. The alpha was chosen as response in the study. The 5-level-3-factor central composite design, with 20 runs, was used to conduct the experiments by ball milling method. A quadratic model was established as a functional relationship between three independent variables and alpha. According to the results, the optimum values of Bi$_2$O$_3$, TiO$_2$, and Sb$_2$O$_3$ were obtained 0.52, 0.50, and 0.30, respectively. Under optimal conditions the predicted alpha (9.47) was calculated using optimal coded values from the model and the theoretical value is in good agreement with the value (9.43) obtained by confirmation experiment.

1. Introduction

Multicomponent semiconducting ceramics that are based on ZnO and other cationic oxides exhibit highly nonlinear current-voltage characteristics [1]. ZnO varistor materials are mainly composed of ZnO as well as a balancing mix of other oxides (e.g., oxides from cobalt, nickel, bismuth, titanium, and antimony). These materials combined produce highly non-Ohmic properties and, therefore, are widely used as surge protecting elements in the electrical transmissions and circuits against lightning or temporary overvoltages [2]. With the popularity of miniaturization and integration of electronic devices, low-voltage varistors are in rapidly growing need [3]. The symmetric nonlinear current-voltage (I-V) response of varistor ceramics is closely related to thin insulating layers around the successive ZnO grains [4]. Microstructural studies show that the thin insulating layers around the successive ZnO grains consist of bismuth-rich second phases which promote the formation of potential barriers to electrical conduction at the ZnO homojunctions [5, 6]. Besides, secondary spinel-type phases are also located at these layers which can control the microstructure development during sintering process [7]. TiO$_2$ is a spinel-forming dopant which is commonly used as a grain growth enhancing additive in the production of low-voltage ZnO-based varistor ceramics. ZnO grains’ growth in the
presence of TiO$_2$ was explained by reduction insulating layer mobility due to the presence of spinel particles. As another spinel-forming dopant, Sb$_2$O$_3$ is typically added to produce fine-grained high-voltage varistor by the presence of a spinel phase at the insulating layer [8]. Each of the dopants plays a distinctive role in the subtle tuning of the final nonlinear characteristics of the varistor ceramics and cannot be omitted. Also, the proper ratios among the dopants have to be set in order to obtain the required electrical performance of the varistor ceramics through the process of microstructure development [9]. As a multivariate case, a pervious study has demonstrated that the phase equilibrium formation and as a consequence the microstructure development are strongly influenced by the TiO$_2$/Bi$_2$O$_3$ in low-voltage varistor ceramics and the Sb$_2$O$_3$/Bi$_4$O$_3$ ratio in high-voltage varistor ceramics. This report reveals the synergistic interaction of dopant in order to achieve a desired microstructure with specific electrical properties [10]. The traditional one-factor-at-a-time approach to optimization is time-consuming and incapable of reaching a true optimum because of taking no account of comprehensive effect of factors [11]. On the contrary, the statistical experimental design allows simultaneous investigation of the effects of several process variables, as well as their actual significance on the considered response and possible interrelationship among them, giving maximum information with the fewest number of trial experiments [12–14]. For these purposes, optimization by RSM utilizing polynomial equations has been widely used [15]. This methodology requires minimum experimentation and time, thus providing to be far more effective and also cost-effective than the traditional methods of optimization [16]. In this study, RSM was used for modeling and optimizing of molar ratio of Bi$_2$O$_3$, TiO$_2$, and Sb$_2$O$_3$ as additives to achieve the maximum value of the alpha for low-voltage varistor.

2. Experimental Procedures

2.1. Sample Preparation. The samples were fabricated using a conventional ceramic process with nominal composition of ZnO (99.99%), Bi$_2$O$_3$ (99.975%), Sb$_2$O$_3$ (99.6%), and TiO$_2$ (99.9%) as raw powders. The raw chemicals of analytical grade produced by Alfa Aesar were weighed according to the molar ratio [mol%]. The molar ratio was converted into weight, and electronic scales were used for precise measurement up to $10^{-4}$ g. The weighed powders were mixed and milled with high-resistance zirconium balls and acetone in a polypropylene jar for 24 h. The resultant slurries were dried in oven at 100°C for 8 h. The dried pastes were crushed into powders, sieved, and pressed into pellet forms of 5.0 mm radius and 0.7 thickness at 20 MPa by Specac Hydraulic Presser machine. The green compacts were sintered at 1260°C for 2 h, with heating and cooling rates of 5°C min$^{-1}$ [17, 18] by a box furnace (CMTS Model HTS 1400) to prepare the ceramic. To determine the electrical properties, both faces of the ceramic pellets were painted by conductive silver electrode which is called varistor.

2.2. Materials Characterization. The electric field-current density (E-J) characteristics of the varistors were recorded with Keithley 236 source meter at room temperature. The varistors were scanned by DC voltage from 0 to 100 V in step size of 2.5 V. The I-V data was used to calculate the alpha of the varistor by

$$\alpha = \frac{\log J_2 - \log J_1}{\log E_2 - \log E_1},$$

where $E_1$ and $E_2$ are the electrical fields corresponding to the current densities $J_1$ and $J_2$ at 0.1 and 1 mA/cm$^2$, respectively. The breakdown field ($E_B$) was determined at 1 mA/cm$^2$ while the leakage current density ($J_L$) was determined at 80% of breakdown field.

The crystalline phase compositions of the sintered sample were identified by X-ray diffractometer (XRD, PANalytical (Philips) X’Pert Pro PW3040/60, CuKα, and λ = 1.5418 Å) and the data were analyzed by using X’Pert High Score software. In order to make microstructural investigations, one of the surfaces of samples was ground with silicon carbide paper and polished with 1μm diamond suspension, which led to a mirror-like surface. Then, they were thermally etched at 1100°C for 10 min to reveal the microstructural details. To reduce charging effects and to improve the resolution of the image, the etched samples were metalized with a thin coating of gold mounted on Al stub. The surface microstructure and compositional analyses of the sintered samples were examined by VPSEM (LEO 1455) which is attached to EDX.

2.3. Experimental Design. The most popular RSM design is the central composite design (CCD) [19, 20]. A CCD has three groups of design points: (a) two-level factorial or fractional factorial design points, (b) axial points (sometimes called “star” points), and (c) center points. CCDs are designed to estimate the coefficients of a quadratic model. The factorial points, which are one unit distance away from the center of the design space, are used to fit the linear and interaction terms. The star points can provide additional levels of the factor for prediction of the quadratic terms, and their distances from the center are α unit [19]. The α value is equal to $(2^n)^{1/4}$, where $n$ is the number of factors. In this study, $n$ is equal to three effective variables, namely, the molar ratio of Bi$_2$O$_3$ (A), TiO$_2$ (B), and Sb$_2$O$_3$ (C), so that the α value is 1.682. Table 1 summarizes the ranges and levels of the effective variables in five different levels ($-1.682$, $-1$, 0, +1, and $+1.682$) involved in the design strategy.

Accordingly, 20 experiments which determined by consisting 8$(2^n)$ full factorial points, 6 (2n) axial points and 6 center points designed as replications to get a good estimate of experimental error (pure error). Then the design (Table 2) was performed according to the procedure in Section 2.1 and the calculated alpha (Section 2.2) is presented in column of actual value (Table 2). The completed design matrix was used for regression process by Design Expert software version 8.0.7.1 (Stat-Ease Inc., USA).
Table 2: Experimental ranges and levels of the effective variables in initial ceramic powders.

| Symbol | Variable | Unit (%) | Level of variables |
|--------|----------|----------|--------------------|
| A      | Bi₂O₃    | mol      | 0.16 -1.682       |
| B      | TiO₂     | mol      | 0.16 -1.0       |
| C      | Sb₂O₃    | mol      | 0.13 -1.0       |

2.4. Statistical Analysis. The multiple regression equation was used to fit the second-order polynomial equation based on the experimental data (Table 2) as follows:

\[ Y = \beta_0 + \beta_1 X_A + \beta_2 X_B + \beta_3 X_C + \beta_{AA} X_A^2 + \beta_{BB} X_B^2 + \beta_{CC} X_C^2 + \beta_{AB} X_A X_B + \beta_{AC} X_A X_C + \beta_{BC} X_B X_C, \]  

where \( Y \) represents the predicted response, \( \beta_0 \) is the model intercept, \( \beta_1, \beta_2, \beta_3, \beta_{AA}, \beta_{BB}, \) and \( \beta_{CC} \) are linear coefficients, \( \beta_{AB}, \beta_{AC} \) and \( \beta_{BC} \) are quadratic coefficients, \( X_A, X_B, \) and \( X_C \) are the independent variables influencing the response. Fitting (2) to the experimental data by the method of least squares (MLS) allowed estimation of all the coefficients [21]. The MLS is a multiple regression technique used to enquire the relationship between the independent and dependent variables and the estimator can be written as follows [22, 23]:

\[ \beta = (X^T X)^{-1} X^T Y, \]  

where \( \beta \) is a vector of regression coefficients; \( X \) is an extended designed matrix of the coded levels of the input variables; and \( Y \) is a column vector of response determined according to the arrangements points into the experimental design.

Once \( \beta \) has been determined, it is then possible to predict \( Y \) and so calculate a few numbers of statistical lines of evidence whose results appear in analysis of variance (ANOVA): \( F \)-value, probability value (P value), lack of fit, coefficient of determination \( R^2 \)-squared \((R^2)\), adjusted \( R^2 \)-squared \((R_{adj})\), and predicted \( R^2 \)-squared \((R_{pred})\). These lines of evidence confirm the quality of the fitted model by conducting Fisher’s \( F \)-test [24, 25].

3. Results and Discussion

3.1. Model Fitting and Statistical Analysis. In the regression process, the software fitted the actual values with the polynomial equation (2) to obtain the predicted values by using RSM (Table 2). The residuals of the actual and predicted values were based on statistical analysis and model suggestion. The suggested model for the initial additives is a function of the molar ratio of Bi₂O₃ (A), TiO₂ (B), and Sb₂O₃ (C) (4). Consider

\[ Y = -15.56 + 16.25A + 25.25B + 111.24C \]  

\[ + 1.18AB + 6.64AC + 0.4BC \]  

\[ - 21.13A^2 - 32.06B^2 - 194.61C^2. \]  

The results of ANOVA were used to evaluate the statistical significance of the quadratic model in general and its terms in detail (Table 3) [26] were listed in Table 3. According to the results, the model \( F \)-value for the model was 28.79, implying the significance of the suggested model. The regression model can explain most of the variation in the response with the large value of \( F \). In addition, the associated \( P \) value was less than 0.05, which confirmed that this model was very significant. On the base of the results of analysis of error, the lack of fit can be achieved simply by differences between residual errors and replicated error which can provide an estimate of the pure error at the center point. A \( P \) value (0.1846) greater than 0.05 implies that the lack of fit is much less than the pure error which is not significant. Nonsignificant lack of fit is good and displays that the model is suitable to accurately predict the response (alpha).

The \( R_d \) is calculated on the basis of the change in the response relative to the total variation of the response over the range of the independent factors. As shown in Figure 1, the good correlation between the observed and the predicted values \( (R_d = 0.9628) \) for this model indicated that this model could well explain the 96.28% of the variability in the responses. The \( R_{pred} \) and \( R_{adj} \) values should be within 0.2 of each other. The \( R_{pred} \) of 0.779 was in good agreement with the \( R_{adj} \) of 0.929 for the model. Adequate precision is a signal-to-noise ratio. Ratios greater than 4 are desirable. In the present study, a ratio of 13.279 shows that the studied model is an efficient application in the design space according to standard error.

According to the terms of statistical analysis, Table 3 also shows the significance effect of each term in the model equation with respective \( P \) values. A significant \( P \) value (<0.05) for each term indicates an active term and a reasonable estimate of its effects. In this case, the independent variables, including A, B, A², B², and C², are significant model terms. In particular, the linear effect of B-TiO₂ and the quadric effect of B-TiO₂ and C-Sb₂O₃ are highly significant terms with \( P \) values <0.0001. The linear and quadric effects of A-Bi₂O₃ are significant with \( P < 0.05 \). In addition, none of the interaction terms had a significant effect on the response \( Y \) \((P > 0.05)\). In order to determine the influence rank of each term in the model, Pareto analysis was used in the form of Pareto chart according the following equation [27]:

\[ P_i = \left( \frac{\beta_i^2}{\sum \beta_i^2} \right) \times 100 \quad (i \neq 0), \]  

where \( P_i \) represents the percentage effect of each factor and \( \beta_i \) represents the coefficient of each term in the polynomial model. As illustrated in Figure 2, the most statistically significant variable in the model was quadratic effect of C-Sb₂O₃ (72%), followed by linear effect of C-Sb₂O₃ (23.5%). However, the antagonistic effect and synergistic effect of each variable on the response (alpha) can be determined by the negative and positive signs of regression coefficients in the model equation.
Table 2: The five-level experimental design of the three additives, Bi$_2$O$_3$, TiO$_2$, and Sb$_2$O$_3$, as ceramic initial powders used in ZnO-based low-voltage varistor. The data was presented by coded and actual values.

| Run | Type    | Coded variable | Actual variable | Y (alpha) |
|-----|---------|----------------|-----------------|-----------|
|     |         | $A$            | $B$             | $C$       | Bi$_2$O$_3$ | TiO$_2$ | Sb$_2$O$_3$ | Actual value | Predicted value |
| 1   | Factorial | $-1$           | $-1$            | $-1$      | 0.3       | 0.3     | 0.2          | 7.19         | 7.10          |
| 2   | Factorial | $+1$           | $-1$            | $-1$      | 0.7       | 0.3     | 0.2          | 6.62         | 5.82          |
| 3   | Factorial | $-1$           | $+1$            | $-1$      | 0.3       | 0.7     | 0.2          | 4.82         | 4.55          |
| 4   | Factorial | $+1$           | $+1$            | $-1$      | 0.7       | 0.7     | 0.2          | 3.45         | 3.46          |
| 5   | Factorial | $-1$           | $-1$            | $+1$      | 0.3       | 0.3     | 0.4          | 6.97         | 6.42          |
| 6   | Factorial | $+1$           | $-1$            | $+1$      | 0.7       | 0.3     | 0.4          | 5.95         | 5.67          |
| 7   | Factorial | $-1$           | $+1$            | $+1$      | 0.3       | 0.7     | 0.4          | 3.65         | 3.90          |
| 8   | Factorial | $+1$           | $+1$            | $+1$      | 0.7       | 0.7     | 0.4          | 3.80         | 3.34          |
| 9   | Axial    | $-1.682$       | 0               | 0         | 0.2       | 0.5     | 0.3          | 7.36         | 7.49          |
| 10  | Axial    | $+1.682$       | 0               | 0         | 0.9       | 0.5     | 0.3          | 5.29         | 5.94          |
| 11  | Axial    | 0              | $-1.682$        | 0         | 0.5       | 0.2     | 0.3          | 6.77         | 7.53          |
| 12  | Axial    | 0              | $+1.682$        | 0         | 0.5       | 0.9     | 0.3          | 3.41         | 3.43          |
| 13  | Axial    | 0              | 0               | $-1.682$  | 0.5       | 0.5     | 0.1          | 3.51         | 3.94          |
| 14  | Axial    | 0              | 0               | $+1.682$  | 0.5       | 0.5     | 0.5          | 2.91         | 3.26          |
| 15  | Central  | 0              | 0               | 0         | 0.5       | 0.5     | 0.3          | 8.93         | 9.10          |
| 16  | Central  | 0              | 0               | 0         | 0.5       | 0.5     | 0.3          | 9.30         | 9.10          |
| 17  | Central  | 0              | 0               | 0         | 0.5       | 0.5     | 0.3          | 9.39         | 9.10          |
| 18  | Central  | 0              | 0               | 0         | 0.5       | 0.5     | 0.3          | 9.83         | 9.10          |
| 19  | Central  | 0              | 0               | 0         | 0.5       | 0.5     | 0.3          | 8.85         | 9.10          |
| 20  | Central  | 0              | 0               | 0         | 0.5       | 0.5     | 0.3          | 8.46         | 9.10          |

Table 3: ANOVA of quadratic model.

| Source         | Sum of squares | Degree of freedom | Mean square | $F$-value | $P$ value | Suggestion |
|----------------|----------------|-------------------|-------------|-----------|-----------|------------|
| Model          | 100.24         | 9                 | 11.14       | 28.79     | <0.0001   | Significant|
| $A$            | 2.89           | 1                 | 2.89        | 7.47      | 0.0211    | —          |
| $B$            | 20.30          | 1                 | 20.30       | 52.46     | <0.0001   | Significant|
| $C$            | 0.55           | 1                 | 0.55        | 1.41      | 0.2623    | —          |
| $AB$           | 0.018          | 1                 | 0.018       | 0.046     | 0.8346    | —          |
| $AC$           | 0.14           | 1                 | 0.14        | 0.37      | 0.5590    | —          |
| $BC$           | $5.2E-004$     | 1                 | $5.2E-004$  | 1.3$E-003$| 0.9715    | —          |
| $A^2$          | 10.30          | 1                 | 10.30       | 26.62     | 0.0004    | Significant|
| $B^2$          | 23.70          | 1                 | 23.70       | 61.24     | <0.0001   | Significant|
| $C^2$          | 54.58          | 1                 | 54.58       | 141.06    | <0.0001   | Significant|
| Residual       | 3.87           | 10                | 0.39        |           |           |            |
| Lack of fit    | 2.72           | 5                 | 0.54        | 2.35      | 0.1846    | Not significant|
| Pure error     | 1.15           | 5                 | 0.23        |           |           |            |
| Corrected total| **104.11**     | **19**            |             |           |           |            |

$R$-squared: 0.9628  
Standard deviation: 0.62

Adjusted $R^2$: 0.9294  
Coefficient of variation %: 9.84

Predicted $R^2$: 0.7793  
PRESS: 22.98

Adequate precision: 13.279
3.2. RSM Analysis. In order to better understand the relationship between the response (alpha) and the independent variables $A$, $B$, and $C$, three-dimensional surfaces plots were formed based on the model polynomial function (Figure 3). Meanwhile, the maximum achievable alpha point was determined by the partial derivative of model (4) with respect to its variables $A$, $B$, and $C$ as presented in following equations:

\[
\begin{align*}
\frac{\partial Y}{\partial A} \bigg|_{BC} &= 0, \\
\frac{\partial Y}{\partial B} \bigg|_{AC} &= 0, \\
\frac{\partial Y}{\partial C} \bigg|_{AB} &= 0.
\end{align*}
\]

By solving the system of (6), the maximum achievable alpha point was found to be 9.47. The corresponding parameters that yielded this maximum value are $A$-Bi$_2$O$_3$ (0.44 mol%), $B$-TiO$_2$ (0.4 mol%), and $C$-Sb$_2$O$_3$ (0.29 mol%).

Figure 3(a) shows the combined effect of variables $A$ and $B$ on alpha at constant molar ratio of Sb$_2$O$_3$ (0.29). As shown, the alpha increased with increasing molar ratio of Bi$_2$O$_3$ between 0.3 and 0.44 and TiO$_2$ between 0.3 and 0.4; however, with further increase more than the optimum (Bi$_3$O$_5$: 0.44 mol% and TiO$_2$: 0.4 mol%), the alpha decreased.

Figure 3(b) signifies the impact of changing amount of Bi$_2$O$_3$ and Sb$_2$O$_3$ on the alpha while the amount of TiO$_2$ was fixed at 0.4. As shown in the figure, it is clear that when the molar ratio of Bi$_3$O$_5$ and Sb$_2$O$_3$ increases, the alpha increases firstly and then decreases after the maximum point (9.47).

The combined effect of the amount of TiO$_2$ and Sb$_2$O$_3$ has been presented in Figure 3(c) while the amount of Bi$_2$O$_3$ was kept constant at 0.44 mol%. The results show that the maximum alpha (9.47) was recorded at 0.4 and 0.29 mol% of TiO$_2$ and Sb$_2$O$_3$, respectively.

In all plots, addition of Bi$_2$O$_3$ up to the optimum points increases the non-linear coefficient (alpha) of the varistor ceramics. This shows that the bismuth content increases in the ZnO samples, someway advocating the cationic inter-diffusion and the mass transport processes throughout the sintering process. It is well identified that the majority of bismuth-doped ZnO varistors have been made ready through liquid phase sintering process. On the other hand, the growth of the Bi$_2$O$_3$ content leads to a growth of the potential barrier height, causing an increase in the value of alpha [28]. However, further increase beyond optimum points might cause the homogeneous segregation of the additives, causing a decline in the level of alpha [29].

As can be seen in all 3D plots with increasing amount of TiO$_2$ up the optimum points, the nonlinear exponents have been enhanced. According to a previous report, TiO$_2$ increases reactivity of the Bi$_2$O$_3$-rich liquid phase with the solid ZnO throughout the sintering process which avoids Bi$_2$O$_3$ vaporization, probably leading to increase in the nonlinear coefficient [18]. However, with more TiO$_2$ doped (more than optimum points), the dopants will worsen the nonlinear electrical properties of the samples. This might be linked to the sudden reduction of the quantity of Bi$_2$O$_3$ in the varistor ceramics owing to the reaction between Bi$_2$O$_3$ and TiO$_2$ into secondary phases, consuming out the varistor-forming oxide. It was stated that, throughout sintering, Bi$_2$O$_3$ provides for the development of insulating boundary layers that regulate the varistors operation [30].

As shown in all plots, the alpha increased by increasing the amount of Sb$_2$O$_3$ below the optimum points. The increasing of the amount of antimony oxide favors the densification of the ceramic matrix during the sintering process of ZnO-based varistors. Moreover, the higher concentration of antimony also shows a significant role in increasing the grain boundary resistivity, as it heightens the probability of segregation forming effective potential barriers and consequently increasing the level of alpha [31]. However, with more increasing beyond optimum levels the alpha values decreased. This reduction may be attributed to the decrease in the amount of Bi$_2$O$_3$ since Bi is incorporated into spinel particles till Bi$_2$O$_3$ ultimately fades as the amount of Sb$_2$O$_3$ enhanced [32].
3.3. Confirmation Experiment. On the basis of RSM, the optimum values of the test variables were Bi$_2$O$_3$, 0.44 mol%; TiO$_2$, 0.4 mol%; and Sb$_2$O$_3$, 0.29 mol%. Under these conditions, the maximum predicted alpha was 9.47. The verification experiment was carried out by studying under optimal conditions. An actual value of 9.43 ± 0.42 was observed from real experiments, which are in close agreement with the model predicted values (9.47). The results (Table 4) indicated that verification study confirmed the predictivity of the model.

3.4. Structural and Morphological Properties of Validated Varistor. Figure 4 represents the XRD spectrum of the validated polycrystalline ceramic used as core in ZnO-based low-voltage varistor. The patterns confirmed the presence of dominant ZnO phase (ICSD code: 00-005-0664) with hexagonal wurtzite structure and secondary phases. Many secondary phases with small peaks were detected in the ceramics at all sintering temperatures, namely, Bi$_4$Ti$_3$O$_{12}$ (ICSD code: 00-008-0258), Bi$_{12}$TiO$_{20}$ (ICSD code: 00-034-0097), Zn$_2$TiO$_4$.
Table 4: The model experimental predicted and validated values of alpha at the prediction conditions.

|                | Bi₂O₃ (% mol) | TiO₂ (% mol) | Sb₂O₃ (% mol) | Alpha       |
|----------------|---------------|---------------|---------------|-------------|
| Predicted condition | 0.44          | 0.4           | 0.29          | 9.47 (predicted) |
| Validated condition | 0.44          | 0.4           | 0.29          | 9.43 ± 0.42 (actual) |

Figure 4: XRD patterns of the ceramic core of ZnO validated varistor made at the predicted conditions.

The morphology and microstructure of the sample fabricated under the optimum conditions were studied by VPSEM (Figure 5). As shown, the grains were uniformly distributed throughout the microstructure of sample with nearly no pore which resulted the high relative density and consequently the reliable non-linear electrical property. From EDAX analysis, the Bi, Ti, and Sb were found at the grains boundaries (Figure 6).

The electrical properties of the varistor were basis of I-V characteristic measurement that shows breakdown voltage was 120 V/mm with alpha 9.43. The leakage current was 0.013 mA/cm².

4. Conclusion

In this study, low-voltage Bi₂O₃-TiO₂-Sb₂O₃ doped ZnO varistor ceramics have been successfully prepared by conventional fabrication technique. The RSM was used to optimize the molar ratio of Bi₂O₃, TiO₂, and Sb₂O₃ as additives in starting powder and a second-order polynomial equation was developed for describing the influence of key variables on response (alpha). The results of model fitting and statistical analysis demonstrated that variables A, B, A², B², and C² played a key role in the response, while the linear effect of C and interactions of AB, AC, and BC demonstrated a negligible effect on the response. The optimal conditions for the response (alpha) were found as Bi₂O₃ of 0.44 mol%, TiO₂ of 0.40 mol%, and Sb₂O₃ of 0.29 mol%. The predicted alpha value determined as 9.47 under the optimal conditions and the verification study (9.43) confirmed the suitability of the predicted model. As a conclusion, RSM through the CCD model is appropriate for determining the optimal conditions for alpha, understanding the relationships among the independent and response variables, and maximizing the alpha.

Conflict of Interests

The authors declare that there is no conflict of interests regarding the publication of this paper.

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References

[1] M. Matsuoka, “Non-ohmic properties of zinc oxide ceramics,” *Japanese Journal of Applied Physics*, vol. 10, no. 6, pp. 736–746, 1971.

[2] L. M. Levinson and H. R. Philipp, “Zinc oxide varistor—A review,” *The American Ceramic Society Bulletin*, vol. 65, no. 4, pp. 639–646, 1986.

[3] W. R. W. Abdullah, A. Zakaria, and M. S. M. Ghazali, “Synthesis mechanism of low-voltage praseodymium oxide doped zinc oxide varistor ceramics prepared through modified citrate gel coating,” *International Journal of Molecular Sciences*, vol. 13, no. 4, pp. 5278–5289, 2012.

[4] M. A. Alim, M. A. Seitz, and R. W. Hirth, “Complex plane analysis of trapping phenomena in zinc oxide based varistor grain boundaries,” *Journal of Applied Physics*, vol. 63, no. 7, pp. 2337–2345, 1988.

[5] S. Anas, R. V. Mangalaraja, M. Poothayal, S. K. Shukla, and S. Ananthakumar, “Direct synthesis of varistor-grade doped nanocrystalline ZnO and its densification through a step-sintering technique,” *Acta Materialia*, vol. 55, no. 17, pp. 5792–5801, 2007.

[6] M. Peiteado, J. F. Fernández, and A. C. Caballero, “Processing strategies to control grain growth in ZnO based varistors,” *Journal of the European Ceramic Society*, vol. 25, no. 12, pp. 2999–3003, 2005.

[7] M. Peiteado, J. F. Fernández, and A. C. Caballero, “Varistors based in the ZnO-Bi₂O₃ system: microstructure control and properties,” *Journal of the European Ceramic Society*, vol. 27, no. 13–15, pp. 3867–3872, 2007.

[8] N. Daneu, N. Novak Gramc, A. Rečnik, M. Maček Kržmanc, and S. Bernik, “Shock-sintering of low-voltage ZnO-based varistor ceramics with Bi₃Ti₅O₁₂ additions,” *Journal of the European Ceramic Society*, vol. 33, no. 2, pp. 335–344, 2013.

[9] S. Bernik and N. Daneu, “Characteristics of ZnO-based varistor ceramics doped with Al₂O₃,” *Journal of the European Ceramic Society*, vol. 27, no. 10, pp. 3161–3170, 2007.

[10] S. Bernik, N. Daneu, and A. Rečnik, “Inversion boundary induced grain growth in TiO₂ or Sb₂O₃ doped ZnO-based varistor ceramics,” *Journal of the European Ceramic Society*, vol. 24, no. 15-16, pp. 3703–3708, 2004.

[11] J. Li, J. Peng, S. Guo, and L. Zhang, “Application of response surface methodology (RSM) for optimization of the sintering process of preparation calcia partially stabilized zirconia (CaO-PSZ) using natural baddeleyite,” *Journal of Alloys and Compounds*, vol. 574, pp. 504–511, 2013.

[12] M. Shah and K. Pathak, “Development and statistical optimization of solid lipid nanoparticles of simvastatin by using 2⁴ factorial design,” *AAPS PharmSciTech*, vol. 11, no. 2, pp. 489–496, 2010.

[13] J. Malakar and A. K. Nayak, “Formulation and statistical optimization of multiple-unit ibuprofen-loaded buoysant system using 2⁴ factorial design,” *Chemical Engineering Research and Design*, vol. 90, no. 11, pp. 1834–1846, 2012.

[14] P. R. Gurus, A. K. Nayak, and R. K. Sahu, “Oil-entrapped sterculia gum-alginete buoysant systems of aceclofenac: development and in vitro evaluation,” *Colloids and Surfaces B: Biointerfaces*, vol. 104, pp. 268–275, 2013.

[15] A. K. Nayak and D. Pal, “Development of pH-sensitive tamarind seed polysaccharide-alginete composite beads for controlled diclofenac sodium delivery using response surface methodology,” *International Journal of Biological Macromolecules*, vol. 49, no. 4, pp. 784–793, 2011.

[16] J. Malakar, A. K. Nayak, and D. Pal, “Development of cloxacillin loaded multiple-unit alginete-based floating system by emulsion-gelation method,” *International Journal of Biological Macromolecules*, vol. 50, no. 1, pp. 138–147, 2012.

[17] S. Bernik, “Microstructural and electrical characteristics of ZnO based varistor ceramics with varying TiO₂/Bi₂O₃ ratio,” *Advances in Science and Technology*, pp. 151–158, 1999.

[18] Y. Abdollahi, A. Zakaria, R. S. Aziz et al., “Optimizing Bi₂O₃ and TiO₂ to achieve the maximum non-linear electrical property of ZnO low voltage varistor,” *Chemistry Central Journal*, vol. 7, no. 1, pp. 137, 2013.

[19] L. Sun and C. Zhang, “Evaluation of elliptical finned-tube heat exchanger performance using CFD and response surface methodology,” *International Journal of Thermal Sciences*, vol. 75, pp. 45–53, 2014.

[20] A. Zuorro, M. Fidaleo, and R. Lavecchia, “Response surface methodology (RSM) analysis of photodegradation of sulfonated diazo dye Reactive Green 19 by UV/H₂O₂ process,” *Journal of Environmental Management*, vol. 127, pp. 28–35, 2013.

[21] M. Yolmeh, M. B. Habibi Naja, and R. Farhoosh, “Optimisation of ultrasound-assisted extraction of natural pigment from annatto seeds by response surface methodology (RSM),” *Food Chemistry*, vol. 155, pp. 319–324, 2014.

[22] R. H. Myers and C. M. Anderson-Cook, *Response Surface Methodology: Process and Product Optimization Using Designed Experiments*, John Wiley & Sons, New York, NY, USA, 2009.

[23] M. Khayet, C. Cojocaru, and M. Essalhi, “Artificial neural network modeling and response surface methodology of desalination by reverse osmosis,” *Journal of Membrane Science*, vol. 368, no. 1-2, pp. 202–214, 2011.

[24] D. R. Cox and C. A. Donnelly, *Principles of Applied Statistics*, Cambridge University Press, New York, NY, USA, 2011.

[25] D. Freedman, R. Pisani and R. Purves, *Statistics*, WW Norton, New York, NY, USA, 2007.

[26] Y. Yuan, Y. Gao, L. Mao, and J. Zhao, “Optimisation of conditions for the preparation of β-carotene nanoemulsions using response surface methodology,” *Food Chemistry*, vol. 107, no. 3, pp. 1300–1306, 2008.

[27] M. Zarei, A. Niaei, D. Salari, and A. Khataee, “Application of response surface methodology for optimization of peroxy-coagulation of textile dye solution using carbon nanotube-PTFE cathode,” *Journal of Hazardous Materials*, vol. 173, no. 1–3, pp. 544–551, 2010.

[28] Y. Romaguera, L. Yeyet, F. Guerrero, L. Agullera, J. Pérez, and J. D. L. S. Guerra, “Influence of Bi³⁺ cation on microstructure and electrical properties of the ZnO,” *Revista Cubana de Química*, vol. 21, no. 3, 2010.

[29] L. Cheng, G. Li, K. Yuan, L. Meng, and L. Zhong, “Improvement in nonlinear properties and electrical stability of ZnO varistors with B₂O₃ additives by nano-coating method,” *Journal of the American Ceramic Society*, vol. 95, no. 3, pp. 1004–1010, 2012.

[30] H. Feng, Z. Peng, X. Fu et al., “Effect of TiO₂ doping on microstructural and electrical properties of ZnO-Pr₆O₁₁-based varistor ceramics,” *Journal of Alloys and Compounds*, vol. 497, no. 1-2, pp. 304–307, 2010.
[31] F. L. Souza, J. W. Gomes, P. R. Bueno et al., “Effect of the addition of ZnO seeds on the electrical properties of ZnO-based varistors,” *Materials Chemistry and Physics*, vol. 80, no. 2, pp. 512–516, 2003.

[32] T. Watanabe, Y. Tokoro, Y. Sato, and S. Yoshikado, “Effects of Sb, Zr, and Y addition on the electrical characteristics of Bi-based ZnO varistors,” *Journal of Physics*, vol. 339, no. 1, Article ID 012007, 2012.