

SOLUTION OF NONLINEAR SISO SYSTEM THROUGH DIFFERENT PARADIGMS

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Abstract – This paper shows different methods for solving a nonlinear regression problem. The same dataset with 250 observations of a nonlinear SISO (single input, single output) system is used for the following methods: least-squares regressor by parts, k th order polynomial regressor, Mamdani fuzzy system, and 0-order Takagi-Sugeno fuzzy system. All models are analyzed via R^2 for different hyperparameters (polynomial order, number of intervals, etc...). The best solutions' residues and scatter plots are analyzed and discussed.

Keywords – Nonlinear regression problem, Mamdani fuzzy system, Takagi-Sugeno fuzzy system, polynomial regressor, regression by parts.

I. Least-Squares by parts

The Least-Squares (LS) method is an approximation technique for overdetermined systems that aims to minimize the squared value of the residuals. Such systems are characterized by having more equations than variables and are easily found in practice [1].

Consider the example of a discrete-time SISO system, where $x_n, y_n \in \mathbb{R}$ are, respectively, their input and output values at the instant $n \in \{1, 2, \dots, N\}$. At each instant, one has an equation where the input and output are related through a set of $K + 1$ parameters, $\theta \in \mathbb{R}^{K+1}$. Mathematically, one can define the output variable as

$$y_n = f(x_n; \theta) \quad (1)$$

When $f(\cdot)$ is a linear function, the LS problem is commonly called Ordinary Least-Square (OLS). The least-squares method aims to minimize the following cost function

$$J(\hat{\theta}) = \sum_{n=1}^N e_n^2, \quad (2)$$

where \hat{y}_n and $\hat{\theta}$ are the estimates of y_n and θ , respectively, and $e_n = y_n - \hat{y}_n$ is the residual signal. Although the OLS method has no optimality associated with it, various practical problems, such as regression analysis, can be solved via OLS since no probabilistic assumptions need to be made about the data [1].

The linear regressor output of the SISO system can be expressed as

$$\hat{y}_n = f(x_n; \hat{\theta}) = \hat{a}x_n + \hat{b}, \quad (3)$$

where $\hat{\theta} = [\hat{a} \ \hat{b}]^T$. By using the Equation (3), we can

rewrite the cost function as

$$J(\hat{\theta}) = \sum_{n=1}^N (y_n - \hat{a}x_n - \hat{b})^2. \quad (4)$$

The Equation (4) describes a convex function whose surface is a hyperparaboloid. The minimum value of the cost function corresponds to the set of coefficients sought [2]. By calculating the derivative of $J(\hat{\theta})$ with respect to \hat{a} and \hat{b} , we get

$$\frac{\partial J(\hat{\theta})}{\partial \hat{a}} = -2 \sum_{n=1}^N x_n (\hat{y}_n - \hat{a}x_n - \hat{b}) = 0 \quad (5)$$

and

$$\frac{\partial J(\hat{\theta})}{\partial \hat{b}} = -2 \sum_{n=1}^N (\hat{y}_n - \hat{a}x_n - \hat{b}) = 0, \quad (6)$$

respectively. The solution of this system of equations is given by

$$\hat{a} = \frac{\hat{\sigma}_{xy}}{\hat{\sigma}_x^2} \quad (7)$$

and

$$\hat{b} = \hat{\mu}_y - \hat{a}\hat{\mu}_x, \quad (8)$$

where $\hat{\mu}$ and $\hat{\sigma}^2$ are the sample mean and sample variance with respect to its subscript, respectively, and $\hat{\sigma}_{xy}$ is the estimate of the covariance of x_n and y_n .

Although the solution of the SISO OLS method is rather straightforward, many input-output relationships found in practice have nonlinearities. In these cases, one can resort to applying a transformation to the data in order to linearize the problem. Another approach is to utilize the OLS method in intervals where the scatter plot behaves approximately linear, yielding a set of linear curves with their respective parameters for each path. It is also possible to exploit other nonlinear regression methods, such as polynomial regression. The scatter plot of the dataset, shown in Figure 1, suggests that the input-output relationship is severely nonlinear. Nonetheless, there are intervals where the function can be approximated to a linear curve. A natural hyperparameter arises in this approach: the number of intervals considered. The main trade-off is that the more intervals considered, the better the performance of the coefficient of determination tends to be. However, more parameters are needed to characterize the curve. The

best solution is to solve the nonlinear problem with as few parameters as possible.

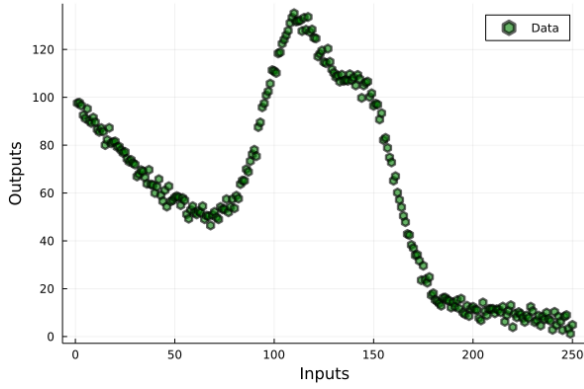


Fig. 1. Scatter plot of the dataset.

In this paper, the OLS algorithm by parts is implemented for different sets of curve intervals, $\{R_i\}_{i=1}^I$, where $I \in \{4, 5, 6\}$ and R_i is the i th interval. Since it is obtained a coefficient of determination, R^2 , for each interval, the mean, μ_{R^2} , and the variance, $\sigma_{R^2}^2$, is analyzed for each configuration. The Figure 2 shows where the delimiters have been placed, in addition to the linear curves obtained by the OLS algorithm. The Algorithm 1 summarizes the behavior of the OLS algorithm by parts, and the Table I shows the performance for each value of I .

Algorithm 1: OLS algorithm by parts

Input: $\{(x_n, y_n)\}_{n=1}^N$

```

1 for  $I \in \{4, 5, 6\}$  do
2   for  $i \in \{1, 2, \dots, I\}$  do
3     for  $(x_n, y_n) \in R_i$  do
4        $N_i \leftarrow$  number of samples that belongs to  $R_i$ 
5        $\hat{\mu}_x \leftarrow \frac{1}{N_i} \sum x_n$ 
6        $\hat{\mu}_y \leftarrow \frac{1}{N_i} \sum y_n$ 
7        $\hat{\sigma}_{xy} \leftarrow \frac{1}{N_i} \sum x_n y_n - \hat{\mu}_x \hat{\mu}_y$ 
8        $\hat{\sigma}_x^2 \leftarrow \frac{1}{N_i} \sum x_n^2 - \hat{\mu}_x^2$ 
9        $\hat{a} \leftarrow \frac{\hat{\sigma}_{xy}}{\hat{\sigma}_x^2}$ 
10       $\hat{b} \leftarrow \hat{\mu}_y - \hat{a} \hat{\mu}_x$ 

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TABLE I
OLS by parts performance - R^2

I	$i=1$	$i=2$	$i=3$	$i=4$	$i=5$	$i=6$	μ_{R^2}	$\sigma_{R^2}^2$
4	0.962	0.952	0.911	0.627	NaN	NaN	0.863	0.018
5	0.962	0.952	0.895	0.985	0.627	NaN	0.884	0.017
6	0.962	0.952	0.876	0.314	0.985	0.627	0.786	0.059

The best solution is found for $I = 5$, highlighted in the Table I. Although it is expected to achieve better performance as it is increased the number of intervals, it is possible to notice by the Figure 2c that the placement of the delimiters does not contribute for a good linear approximation (there are outliers for $i = 4$). This fact decreases R^2 and makes it worse when compared with the case where $I = 5$.

In addition to the regressor performance, one can analyze

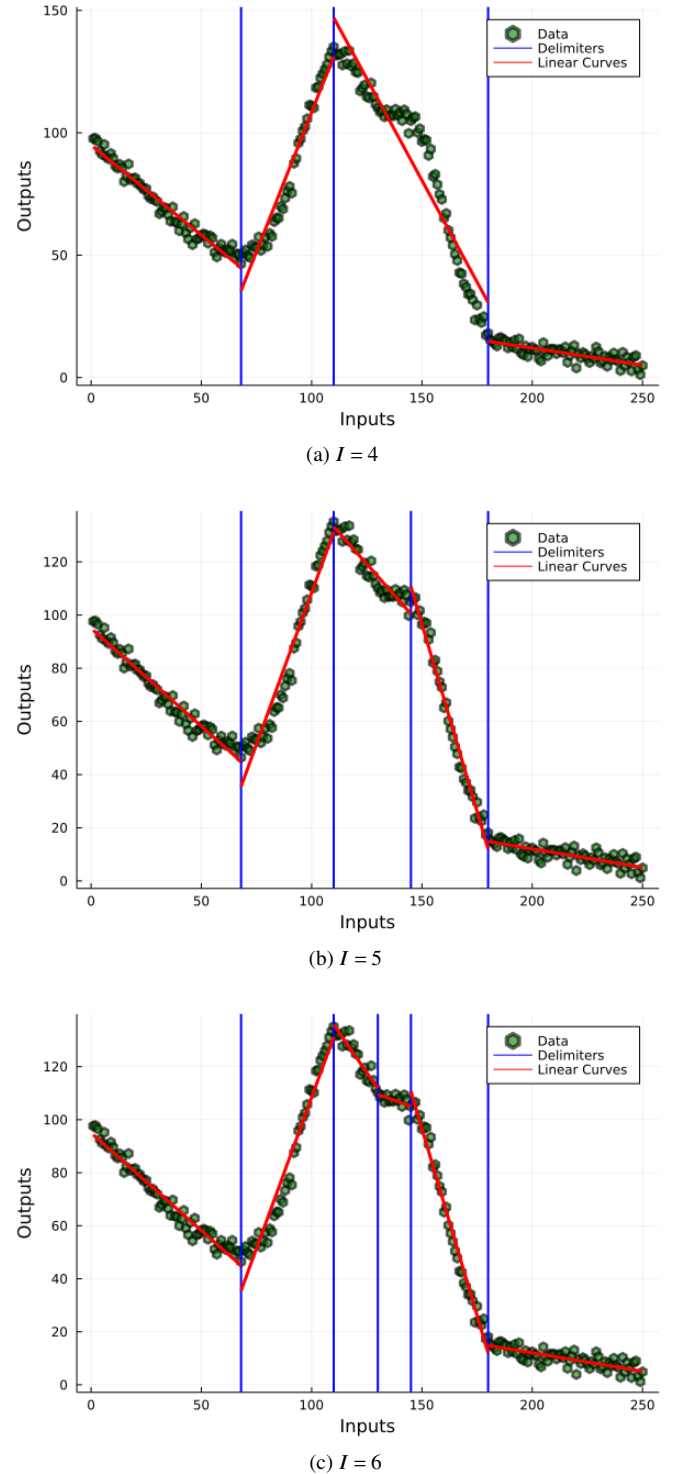


Fig. 2. The OLS method by parts.

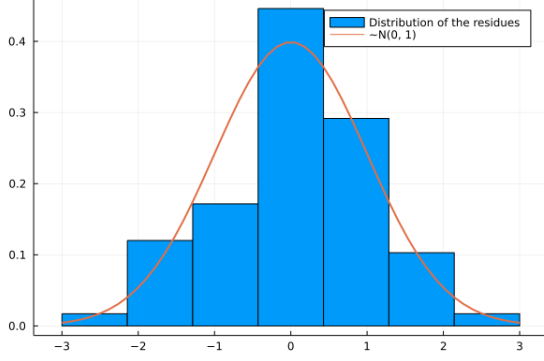


Fig. 3. Distribution of the residuals.

the distribution of the residuals. Let ξ_n be the normalized residual value, that is, $\xi_n = e_n / \sigma_e$. For a good placement of the intervals, the normalized residual distribution approximates to a zero-mean Gaussian distribution with unitary variance, i.e., $\xi_n \sim N(0, 1)$. The Figure 3 shows the distribution (for $i = 1$) of the residuals along with the Gaussian distribution.

II. Polynomial Regression

The Polynomial regression is a nonlinear method which extends the concept of the SISO OLS algorithm. At the instant n , the input x_n is utilized to generate the tuple $(h_{n,1}, h_{n,2}, \dots, h_{n,K})$, where $h_{n,k} = x_n^k$, $\forall k \in \{1, 2, \dots, K\}$. Therefore, the SISO model is transformed into a MISO (Multiple Input, Single Output) model, where the k th input of the n th instant is x_n to the power k . Note that, albeit this model is not linear with relation to x_n , it is with relation to $h_{n,k}$, that is,

$$\begin{aligned} y_n &= f(x_n; \hat{\boldsymbol{\theta}}) = \mathbf{h}_n^T \hat{\boldsymbol{\theta}} \\ &= \hat{\theta}_0 + \hat{\theta}_1 h_{n,1} + \hat{\theta}_2 h_{n,2} + \dots + \hat{\theta}_K h_{n,K}, \end{aligned} \quad (9)$$

where $\hat{\boldsymbol{\theta}} = [\hat{\theta}_0 \ \hat{\theta}_1 \ \dots \ \hat{\theta}_K]^T \in \mathbb{R}^{K+1}$ and $\mathbf{h}_n = [1 \ h_{n,1} \ h_{n,2} \ \dots \ h_{n,K}]^T \in \mathbb{R}^{K+1}$. The matricial notation for all N observations is given by

$$\mathbf{y} = \mathbf{H} \hat{\boldsymbol{\theta}} \quad (10)$$

where $\mathbf{y} = [y_1 \ y_2 \ \dots \ y_N]^T \in \mathbb{R}^N$ and

$$\mathbf{H} = \begin{bmatrix} 1 & h_{1,1} & h_{1,2} & \dots & h_{1,K} \\ 1 & h_{2,1} & h_{2,2} & \dots & h_{2,K} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & h_{N,1} & h_{N,2} & \dots & h_{N,K} \end{bmatrix} \in \mathbb{R}^{N \times (K+1)} \quad (11)$$

The cost function is given by

$$\begin{aligned} J(\hat{\boldsymbol{\theta}}) &= (\mathbf{y} - \mathbf{H} \hat{\boldsymbol{\theta}})^T (\mathbf{y} - \mathbf{H} \hat{\boldsymbol{\theta}}) \\ &= \mathbf{y}^T \mathbf{y} - 2\mathbf{y}^T \mathbf{H} \hat{\boldsymbol{\theta}} + \hat{\boldsymbol{\theta}}^T \mathbf{H}^T \mathbf{H} \hat{\boldsymbol{\theta}} \end{aligned} \quad (12)$$

By differentiating the Equation (12) with respect to $\hat{\boldsymbol{\theta}}$ and

setting its result to zero, we get

$$\begin{aligned} \frac{\partial J(\hat{\boldsymbol{\theta}})}{\partial \hat{\boldsymbol{\theta}}} &= -2\mathbf{H}^T \mathbf{y} + 2\mathbf{H}^T \mathbf{H} \hat{\boldsymbol{\theta}} = 0 \\ \therefore \hat{\boldsymbol{\theta}} &= \mathbf{H}^\dagger \mathbf{y}, \end{aligned} \quad (13)$$

where $\mathbf{H}^\dagger = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T$ is the pseudoinverse of \mathbf{H} , also called left inverse since \mathbf{H} is a tall matrix ($N > K$). The fact that \mathbf{H} is full rank ($\text{rank}(\mathbf{H}) = K + 1$) guarantees the invertibility of $\mathbf{H}^T \mathbf{H}$ [3]. The procedure of the K th-order polynomial regression is summarized in the Algorithm 2, where the hyperparameter K is analyzed for the set $\{5, 6, 7\}$.

Algorithm 2: K th-order polynomial regressor.

Input: $\{(x_n, y_n)\}_{n=1}^N$
1 for $K \in \{5, 6, 7\}$ **do**
 2 $\mathbf{y} \leftarrow$ generated from $\{y_n\}_{n=1}^N$
 3 $\mathbf{H} \leftarrow$ generated from $\{x_n\}_{n=1}^N$
 4 $\mathbf{H}^\dagger = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T$ /* The pseudoinverse */
 5 $\hat{\boldsymbol{\theta}} = \mathbf{H}^\dagger \mathbf{y}$

The Figure 4 shows the curve fitting for the K th-order polynomial regressor. As shown in the Table II, the 6th-order polynomial curve reached the best result. For this case, the model achieved the Pearson correlation coefficient between \mathbf{y} and $\hat{\mathbf{y}}$ of 97.84%.

TABLE II
OLS by parts performance - R^2

$K = 5$	$K = 6$	$K = 7$
0.85	0.957	0.566

III. ORGANIZATION OF THE PAPER

This section presents the main issues for editing the manuscript.

A. General Organization

The papers that shall be published in the Brazilian Power Electronics Journal must contain the following main sections: 1) Title; 2) Authors and Affiliations; 3) Abstract and Keywords; 4) Introduction; 5) Body Text; 6) Conclusions; 7) References; 8) Biographies. This order must be respected, unless the authors add some items, such as: Nomenclature; Appendices and Acknowledgements.

Some comments regarding the main items of the manuscripts are presented below.

1) Title: The title of the paper should be as succinct as possible, stating the subject of the paper in a very clear manner. It should be centered at the top of the first page, in bold, type size 14 points, with the whole title in capital letters.

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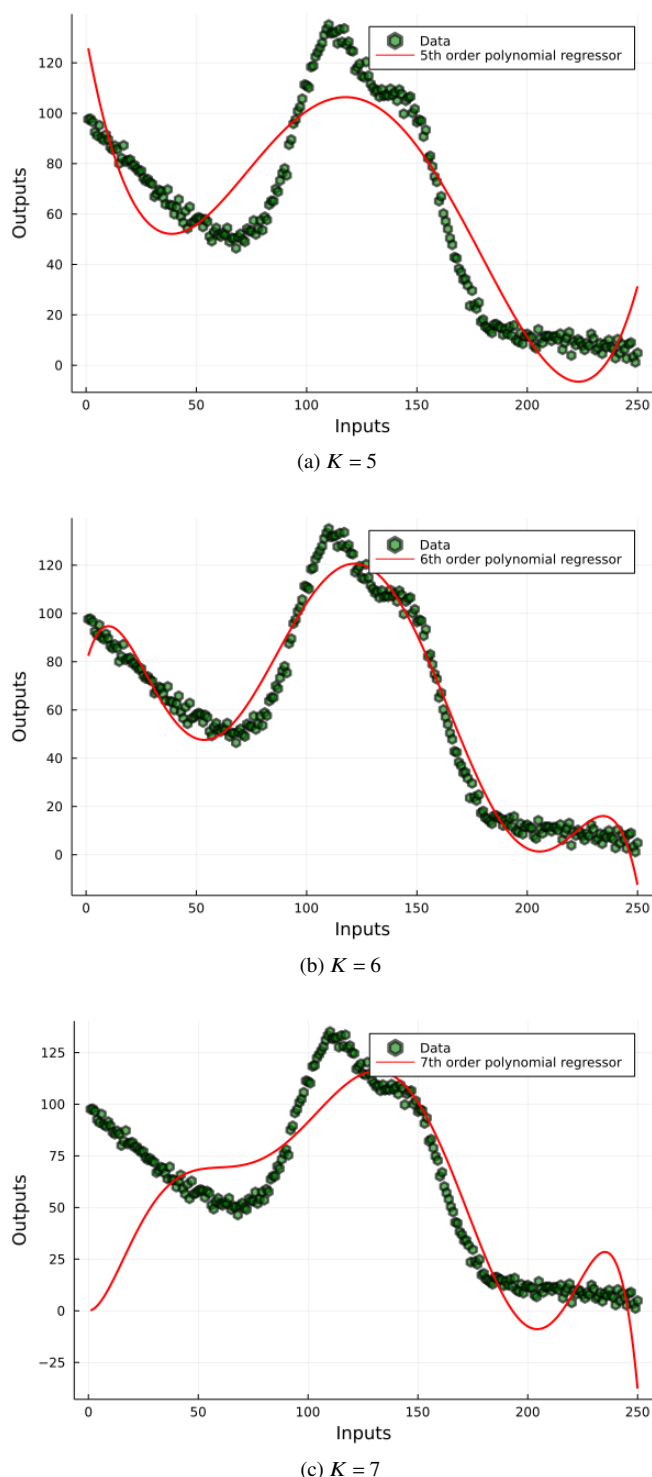


Fig. 4. The K th-order polynomial regressor.

forms (type size 12 points). Immediately below the authors' names, their affiliations, with city, state and country, must be informed (type size 10 points). The electronic addresses must be informed just below the affiliations (type size 10 points).

3) *Abstract and keywords*: This part is considered one of the most important in the whole paper. It is based on information in Abstract and Keywords that technical papers are indexed and stored in databases.

The Abstract should have no more than 200 words, indicating the main ideas contained in the paper, as well as procedures and obtained results. The Abstract should not be confused with the Introduction and should not have any abbreviations, references, figures, etc. For writing the Abstract, as well as the whole manuscript, you should use passive voice, e. g., "... the experimental results show that..." instead of "... the results we obtained show that...". The word Abstract must be written both in italic and in bold. The Abstract text should be in bold.

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7) *References*: The citation of references throughout the text should appear between square brackets, just before the punctuation mark at the end of the sentence in which the reference is inserted. Only the number of the references should be used, avoiding citations such as "... according to the reference [2]...".

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Regarding books, following the name of the authors (initials followed by last name), the title should be in italic, and then should come the publisher, number of edition and place and year of publication.

At the end of these guidelines, there is an example of how the references should be inserted ?, ?, ?, ?, ?, ?, ?.

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- Full name (in bold and underlined);
- Place and year of undergraduation and graduation conclusions;
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On the last page of the paper, the authors should distribute the contents evenly, using both columns, in a way that both end in a parallel manner.

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The organization of the manuscript in titles and subtitles is important to divide it in sections, which help the reader to find subjects of interest in the paper. They also help the authors to develop their paper in an orderly form. The paper can be organized in primary, secondary and tertiary sections.

The primary sections are the titles of the actual sections. They are written in capital letters in the center of the column separated by a blank line above and another one below them, and sequential Roman numerals should be used.

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The tertiary sections are subdivisions of the secondary sections. Only the first letter of the first word of the section should be a capital letter. The designation of the tertiary sections should be done with Arabic numerals, followed by parentheses. They should be in italic.

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B. Figures and Tables

Tables and figures (drawings or pictures) should be inserted in the text right after they are mentioned for the first time, as long as they fit the size of the columns; if necessary, use the whole page. Figures resolution should be at least 300 dpi and vector files should be preferably used for better print quality. Table captions should be above the tables and figure captions should be below the figures. The tables should have titles and they are designated by the word Table, being numbered in sequence by Roman numerals. Table captions must be centered and in bold.

Figures also need captions and they are designated by Figure in the text (Fig. in the caption itself), numbered with Arabic numerals in a sequenced manner, left- and right-justified, as shown in the example. The designation of the parts of a figure is done by adding lowercase letters to the numbers of the figures starting with the letter a, e.g. Figure 1(a).

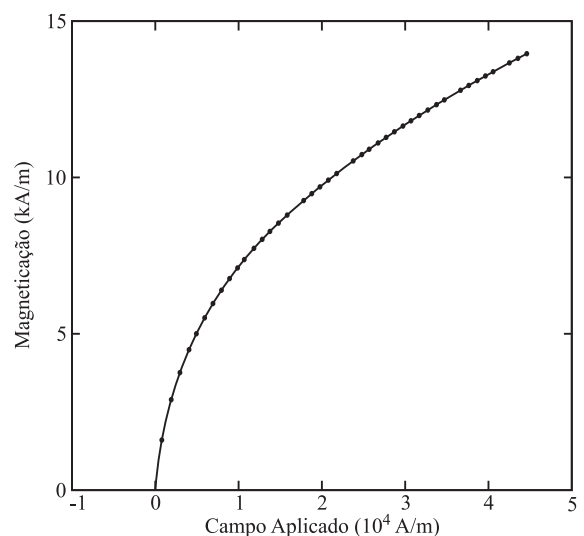


Fig. 5. Magnetization as a function of applied field. (Note that “Fig.” is abbreviated and there is a period after the figure number followed by two spaces.)

To better understand graphs, the definition of their axes

should be done with words not letters, except when referring to waveforms and phase planes. The units should be between parentheses. For example, use the denomination “Magnetization (A/m)”, instead of “M (A/m)”.

Figures and tables should be positioned preferably in the beginning or the end of the column, avoiding putting them in the middle. Avoid tables and figures whose sizes exceed the size of the columns. The figures should preferentially be in black, with a white background, since the printed version of the journal is in black and white. Their lines should be thick, so the impression is readable.

C. Abbreviations and Acronyms

Abbreviations and acronyms must be defined the first time they are used in the text, e.g. “... Pulse-Width Modulation (PWM)...”.

D. Equations

Number equations consecutively with equation numbers in parentheses flush with the right margin, as in (1). The equations should be written in a compact form, centered in the column. If a nomenclature section is not included in the beginning of the text, the quantities should be defined right after the equation, such as:

$$\Delta I_L = I_o + \frac{\sqrt{3}}{2} \frac{V_i}{Z} \quad (14)$$

where:

- ΔI_L - resonant inductor peak current;
- I_o - load current;
- V_i - source voltage;
- Z - characteristic impedance.

V. CONCLUSIONS

This paper was fully written in accordance with the guidelines for submissions of papers in English.

ACKNOWLEDGEMENTS

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