IDENTIFICATION OF DIFFUSION COEFFICIENTS DURING POST-DISCHARGE NITRIDING

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Abstract- This work presents a model for computing the diffusion coefficients during post-discharge nitriding through an inverse problem of coefficient identification in a diffusion model, which considers several layers and Stefan type conditions. An approximate solution corresponding to a quasi-stationary state is obtained for the model, using qualitative and quantitative information from experimental results. To study the inverse problem we assume that we have information about the nitrogen concentration at different depths in the solid after the stabilization of the layers in the post-discharge nitriding process. To develop a numerical algorithm for the identification of the diffusion coefficients, a functional to be minimized is built. This functional measures the deviation between the data theoretically obtained from the approximate solution of the model and the experimental data.

1. INTRODUCTION

Thermo-chemical nitriding treatment produces an important improvement in the mechanical, tribological and chemical properties in steel, thus enhancing the resistance to their fatigue, corrosion and wear [1, 7, 8].

In most of the mathematical simulations of nitrogen diffusion in iron or steel reported, nitrogen concentration on the surface is taken to be constant from the beginning of the treatment [4, 5, 9,10, 11, 14, 15, 16, 18] and consequently it is considered that the thickness of the layers is zero at the initial moment. The evolution of the nitrogen concentration on the surface is inherent to the process. In processes where the diffusion takes place through a thermo-chemical balance between a gaseous mixture and the solid, i.e. mixtures with ammonia, nitrogen concentration on the surface depends on the nitrogen potential [4, 14, 16].

During weakly ionized plasma assisted processes, pulverization, adsorption and diffusion events take place on the specimen surface [10, 11]. The nitrogen concentration on the surface evolves quickly and corresponds to a dynamic balance between pulverization towards the atmosphere and the diffusion towards the solid.

Control or automation of nitriding processes depends on the means requires to identify the nitrogen concentration on the surface as well as on a mathematical model adapted to estimate the growth kinetics of compact concomitant nitride layers. On the other hand, the understanding and interpretation of the mechanisms of mass transport in solid requires experimental validation of the diffusion coefficients.

In computing diffusion coefficients, using experimental results, it has been assumed that the nitrogen concentration has a parabolic growth profile in each layer from the beginning of the process, nevertheless the nitrogen assumes different forms in each process and depends on each one. If the nitrogen concentration on the surface evolves slowly, the parabolic regime will not be observed during the initial stages, moreover the accuracy in computing diffusion coefficients will be limited [4, 5, 10, 11, 14, 16, 18].

Nitriding post-discharge processes generate an atmosphere of excited neutral or dissociated species [2, 3, 5, 12, 13]. These species produce a fast evolution of the nitrogen concentration on the surface [3].

It has been observed that the presence of atomic nitrogen in post-discharge processes significantly increases the mass transfer to a solid compared to other processes.

Based on experimental results obtained from iron nitriding, in an atmosphere produced by post-discharge microwaves, the present work considers a one-dimensional model of layer kinetic growth, and studies the inverse problem of diffusion coefficients identification in each phase from the consideration of the stabilization of the layers after a certain period of treatment. Such consideration was justified from an experimental point of view and also by taking into account the analytical expression of the layer growth kinetics.

The experimental results obtained by means of treatments assisted during post-discharge generated by a microwave source allow the accurate determination of the values of the diffusion coefficients on the basis of the solution of an inverse problem which identifies the coefficients in a one-dimensional model of kinetic growth of concomitant nitride layers.

2. EXPERIMENTAL PROCEDURE

Samples were obtained from an ARMCO iron bar (25.4 mm in diameter and 7 mm thick, Mn 880 ppm; C and P, 200 ppm; and S 150 ppm). Nitriding was carried out in a post-discharge microwave-generated plasma described elsewhere [3]. The general sequence of the nitriding experiments started with heating up the sample to 893 K in a
tubular resistance furnace in a non-oxidized and non-nitriding atmosphere composed of 26Ar-80H2 sccm at a total pressure of 900 Pa. The applied and reflected were 200 W and 65 W respectively, and the distance from the discharge point was 7 cm. Upon reaching the prescribed temperature, the temperature, the atmosphere was switched to a mixture of 300N2-26Ar-80H2 at 1200 Pa and recording of the nitriding time started. After the nitriding time was completed, the atmosphere was switched back to the initial non-oxidizing, non-nitriding atmosphere.

Figure 1 presents a cross-sectional view of a sample nitrided for 120 min under the conditions described previously. The nitrides below the compact nitride layer precipitated during the cooling of the sample due to the desaturation of the ferrite; the width of this zone is between 40 to 50 \( \mu \)m. However, the diffusion zone of the nitrogen in the ferrite is about 3 mm.

3. MATHEMATICAL MODEL

Let us propose a mathematical model which describes the growth of the layers during the post-discharge nitriding process. We suppose that the diffusion is one-dimensional and planar and the temperature at every point in the specimen is identical during the whole process. This model presents five steps. The first step ends when the surface reaches the equilibrium concentration \( C_S \) at a time \( t_0 \). This step is modeled by:

\[
\frac{\partial C}{\partial t} = D \frac{\partial^2 C}{\partial x^2}, \quad 0 < x < +\infty, \quad t > 0
\]

(1)

\[
C(x,0) = C_0, \quad 0 < x < +\infty
\]

(2)

\[
\frac{\partial C}{\partial x}(0,t) = \frac{\lambda}{D}(C - C_{eq}) \bigg|_{x=0}, \quad t > 0
\]

(3)

\[
\lim_{x \to +\infty} C(x,t) = C_0 \quad t > 0
\]

(4)

where \( \lambda \) is the kinetic reaction coefficient, \( D \) is the diffusion coefficient, \( C_{eq} \) is the equilibrium nitrogen concentration in the atmosphere remote from the surface and \( C(x,t) \) represents the nitrogen concentration for time \( t \) and depth \( x \). The solution of (1)-(4) is given by:

\[
\frac{C - C_{eq}}{C_0 - C_{eq}} = \text{erf} \left( \frac{x}{2 (Dt_0)^{1/2}} \right) + \exp \left( \frac{\lambda}{D} x + \frac{\lambda^2}{D} t \right) \text{erfc} \left( \frac{x}{2 (Dt_0)^{1/2}} + \frac{\lambda}{D^{1/2}} t^{1/2} \right)
\]

where

\[
\text{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-\xi^2} d\xi, \quad \text{erfc}(x) = 1 - \text{erf}(x) = \frac{2}{\sqrt{\pi}} \int_x^{+\infty} e^{-\xi^2} d\xi
\]

A certain \( t_0 \) should pass for the concentration to reach the threshold value \( C_S \), which follows from
Let us put
\[ f(x) = C(x,t_0) = C_{eq} + \]
\[ (C_0 - C_{eq}) \left\{ \text{erf} \left( \frac{|x|}{2(Dt_0)^{1/2}} \right) + \exp \left( \frac{\lambda^2}{D} x + \frac{\lambda^2}{D} t_0 \right) \text{erfc} \left( \frac{|x|}{2(Dt_0)^{1/2}} + \frac{\lambda}{D^{1/2}} (t_0)^{1/2} \right) \right\} \]

where \( f(x) \) denotes the initial concentration profile when the layer formation begins.

From the starting profile two layers of nitrogen and a diffusion zone begin to be formed and they slowly displace into the interior of the metal. In each layer and diffusion zone, diffusion coefficient takes a constant value \( D_i \), \( i = 1, 2, 3 \). Between adjacent layers and diffusion zone there is a jump in concentration. In each of the first two layers a minimum constant value \( C_{\text{min}}^i \), \( i = 1, 2 \) is reached. For every \( t > t_0 \) let us define \( \xi_i(t) \), \( i = 1, 2 \) as the depth of the corresponding layer. Since diffusion coefficients are constants, it follows that the concentration in each layer is a decreasing function of the depth, then \( \xi_i(t) \) is the depth at which the concentration reaches the minimum value \( C_{\text{min}}^i \), \( i = 1, 2 \) (see Figures 2a-d). As a consequence of the process, there is an experimental value of the concentration \( C_{\text{max}}^i < C_{\text{min}}^i \) and a time \( t = t_i \) such that \( f(\xi_i(t_i)) = C_{\text{max}}^{i+1} \), \( i = 1, 2 \). In this situation we say that the layer was completely formed at time \( t_i \), as shown schematically in Fig. 2 b-d. We shall assume \( t_1 < t_2 \). For \( t \in [t_0, t_1) \) the concentration has a jump in length \( C_{\text{min}}^i - f(\xi_i(t)) \), \( i = 1, 2 \). Moreover, for \( t > t_i \) the length of the jump in concentration in \( \xi_i(t) \) is constant and equal to \( C_{\text{min}}^i - C_{\text{max}}^{i+1} \), \( i = 1, 2 \). In Figures 2a-d we denote \( x_i = \xi_i(t_i) \), \( i = 1, 2 \).

In what follows, we propose a model for the process of post-discharge nitriding from the starting profile \( f(x) \). This model describes the beginning of the layers and interfaces formation. We consider three stages:
and denote the nitrogen concentration at the \( i \)-th layer or diffusion zone for time \( t \) and at depth \( x \), by \( C_i(x,t) \).

Let us define the value of the depth where \( f(x_i^0) = C_{\min}^i \), \( F_i(t) = \max\{C_{\max}^{i+1}, f(\xi_i(t))\} \), \( i = 1, 2 \) and \( x_0^0 = 0, \xi_0(t) = 0 \).

Then, the model has the following form:

\[
\frac{\partial C_i}{\partial t} = D_i \frac{\partial^2 C_i}{\partial x^2}, \quad t > t_0, \quad \xi_{i-1}(t) < x < \xi_i(t), \quad i = 1, 2
\]

\[
t > t_0, \quad \xi_2(t) < x, \quad i = 3.
\]

\[
C_i(x,t_0) = f(x), \quad x_{i-1}^0 < x_i^0, \quad i = 1, 2, \quad x_2^0 < x, \quad i = 3.
\]

\[
C_i(\xi_{i-0}^0(t), t) = C_{\min}^i, \quad t > t_0, \quad i = 1, 2,
\]

\[
C_i(0, t) = C_S, \quad t > t_0,
\]

\[
C_{i+1}(\xi_{i+0}^0(t), t) = C_{\max}^{i+1}, \quad t > t_i, \quad i = 1, 2,
\]

\[
(C_{\min} - F_i(t)) \frac{d\xi_i}{dt} = -D_i \frac{\partial C_i}{\partial x} \bigg|_{x=\xi_{i-0}^0} + \frac{\partial C_{i+1}}{\partial x} \bigg|_{x=\xi_{i+0}^0} + i = 1, 2,
\]

\[
\lim_{x \to +\infty} C_3(x, t) = C_0, \quad t > t_0,
\]

\[
\xi_i(t_0) = x_i^0, \quad i = 1, 2.
\]

Note that this model describes the three different stages of the process, characterized by the following facts:

When \( t \in [t_0, t_1) \) we have \( F_i(t) = f(\xi_i(t)) \), \( i = 1, 2 \). For \( t \in [t_1, t_2) \) we have \( F_i(t) = C_{\max}^2 \), \( F_2(t) = f(\xi_2(t)) \), and finally, for \( t \geq t_2 \) we have \( F_i(t) = C_{\max}^{i+1} \), \( i = 1, 2 \).

To be consistent with the common notation we define \( \xi^3(t) \), \( t > t_0 \) as the depth at which:

\[
C_3(\xi^3(t), t) = 0.1 C_{\max}^3
\]

At the last mentioned stage the three layers are completely formed as shown in Figure 2d. In this case the model corresponds to the growth of layers and movement of interfaces. This last stage corresponds to a period, previous to the layer growth “stabilization”, where layers and interfaces exactly follow the portrait of a moving boundary behavior of the Stefan type.

For large values of time the layer growth becomes negligible, which is justified from experimental essays and also analytically since the interfaces move with velocities proportional to \( \frac{1}{\sqrt{t}} \) and become small for large values of time. So the process arrives to a quasi-steady state, where the variation of \( \frac{\partial C}{\partial t} \) is very small.

### 3.1 Simulation of the concentration profile in the quasi-steady state

We consider that the concentration profile in the quasi-steady state has the form:

\[
C_i(x) = a_i (x - x_i^\ast)^m_i + b_i (x - x_i) + c_i, \quad i = 1, 2, 3
\]

where \( x_i \) is the depth at which each layer reaches the quasi-steady state (experimental observation), \( m_i \) (integers greater than or equal two), \( a_i, b_i, c_i \), \( i = 1, 2, 3 \) are constants to be determined. Following the usual notation for layers, we define:

\[
\tilde{x}_1 = x_1^\ast, \quad \tilde{x}_2 = x_2^\ast, \quad \tilde{x}_3 = x_3^\ast
\]

\[
(5)
\]
Expression (5) guarantees that the last two terms also satisfy the steady equation \( \frac{d^2 C}{dx^2} = 0 \), while the first term should be such that \( m_i (m_i - 1) a_i (x - x_i)^{m_i-2} \) is very small for values of \( x \) in the \( i \)-layer. The form of expression (5) is inspired by Goodman’s method [17, 18].

Thus, the problem is to determine the relationships between \( m_i, a_i, b_i, c_i, \quad i = 1, 2, 3 \) such that conditions for concentration decrease at each phase, and a concentration jump and null net flow at the interfaces are fulfilled. More exactly we write the following:

\[

c_i (x) = a_i (x - x_{i,j})^m + b_i (x - x_{i,j}) + C_{m, i}^1
\]

\[

c_2 (x) = a_2 (x - x_{i,j})^n + b_2 (x - x_{i,j}) + C_{m, i}^2
\]

\[

c_3 (x) = a_3 (x - x_{i,j})^k.
\]

because \( C_{m, i}^3 \) may be considered null and from the null net flow at the third interface, then it follows that \( b_3 = 0 \).

Expression (6) guarantees:

\[

c_1 (x_{i,j}) = C_{m, i}^1, \quad c_2 (x_{i,j}) = C_{m, i}^2, \quad c_3 (x_{i,j}) = 0
\]

In a straightforward manner we obtain:

\[
a_3 = (-1)^k \left( \frac{C_m^3}{x_{i,j} - x_{i,j}} \right)^k, \quad b_3 = -k \frac{D_3^3}{D_2^2} \left( \frac{C_m^3}{x_{i,j} - x_{i,j}} \right)
\]

\[
a_2 = \left( \frac{1}{x_{i,j} - x_{i,j}} \right)^n \left( \frac{C_m^2 - C_{m, i}^2}{x_{i,j} - x_{i,j}} - k \frac{(x_{i,j} - x_{i,j})}{D_3^3} \frac{D_2^3}{C_m^2} \right)
\]

\[
b_2 = \frac{D_2^2}{D_1^1} \left[ \frac{(n-1)k}{x_{i,j} - x_{i,j}} \frac{D_3^3}{D_2^2} C_m^3 - \frac{n}{x_{i,j} - x_{i,j}} \left( C_m^2 - C_{m, i}^2 \right) \right]
\]

\[
a_1 = (-1)^m \left( \frac{C_m^3}{x_{i,j}^m} \right) \left( x_{i,j} - x_{i,j} \right)^m + \frac{D_2^3}{D_1^1} \left[ \frac{(n-1)k}{x_{i,j} - x_{i,j}} \frac{D_3^3}{D_2^2} C_m^3 - \frac{n}{x_{i,j} - x_{i,j}} \left( C_m^2 - C_{m, i}^2 \right) \right]
\]

where \( m, n \) and \( k \) are integers greater than or equal two. In the numerical experiments the linear case \( (m = n = k = 0) \) is also considered.

4. INVERSE PROBLEM AND NUMERICAL RESULTS

To find the diffusion coefficients \( D_i, \quad i = 1, 2, 3 \) we use the approximate expression (6) of the quasi-steady problem and a nonlinear optimization algorithm to solve the least square problem.

Experimental measures of the quasi-steady depth of the layers and diffusion zone are taken, as well as the experimental values of \( C_S, C_{m, i}^1, C_{m, i}^2 \) and the diffusion coefficients \( D_i \). All these values were reported in the references or were obtained in the laboratory by the procedure described in the second section above.

\[
x_{i,j} = 7 \mu m, \quad x_j = 17 \mu m, \quad x_{i,j} = 67 \mu m,
\]

\[
C_S = 19.25, \quad C_{m, i}^1 = 19, \quad C_{m, i}^2 = 9.625, \quad C_{m, i}^3 = 6.5, \quad C_m^3 = 0.365, \quad C_{m, i}^3 = 0
\]

\[
D_1 = 3.02204 \times 10^{-13}, \quad 1.0259 \times 10^{-13}, \quad 7.713003 \times 10^{-13} \text{ m}^2 / \text{seg},
\]

\[
D_2 = 1.27401 \times 10^{-14}, \quad 2.186 \times 10^{-13}, \quad 3.6958 \times 10^{-13} \text{ m}^2 / \text{seg},
\]

\[
D_3 = 1.83122 \times 10^{-11} \text{ m}^2 / \text{seg},
\]

From this data, theoretical expressions for concentration profiles at each layer and diffusion zone are computed, using the linear and nonlinear models (5) and relations (7)-(10). Then a simulation process is performed, generating artificial data from the theoretical expressions of the profiles (adding random noise) in order to test the feasibility and sensibility of the solution of the inverse problems.
Due to technical difficulties in obtaining real measurements in the first two layers, the artificial data \( \hat{C}_i(y_j) \) are generated only at depths \( y_j = 2, 4, 6, 8 \mu m \), corresponding to those layers, and at depths \( y_j \mu m, j = 5, \ldots, N \), in the diffusion zone.

Afterwards, we use the generated data to solve the least-square problem of minimizing the function:

\[
J\left( \frac{D_2}{D_1}, \frac{D_3}{D_2} \right) = \alpha_1 \left[ C_1(2) - \hat{C}_1(2) \right]^2 + \alpha_2 \left[ C_1(4) - \hat{C}_1(4) \right]^2 + \alpha_3 \left[ C_1(6) - \hat{C}_1(6) \right]^2 + \alpha_4 \left[ C_2(8) - \hat{C}_2(8) \right]^2 + \sum_{j=5}^{N} \left[ C_3(y_j) - \hat{C}_3(y_j) \right]^2
\]

where the function \( J \) represents the error between the analytical and experimental concentration data, meanwhile \( \alpha_i, i = 1, \ldots, 4 \), \( \alpha_j, j = 5, \ldots, N \) represent different weights in \( J \) which may improve the estimates of \( \frac{D_2}{D_1} \) and \( \frac{D_3}{D_2} \).

Computing was performed on a PC at 900 Mhz, using MATLABregression programs version 6.5. The results of preliminary numerical experiments for the linear model \( (m = n = k = 0) \) yield:

- Estimates of \( \frac{D_2}{D_1} \) and \( \frac{D_3}{D_2} \) are not sensitive to the increase of the number of measurements in the diffusion zone when the artificial data do not have random errors.
- Estimates are very sensitive to the initial vector in the optimization process. It seems that an initial of \( \frac{D_3}{D_2} \) close to the order of the actual values is the best option, while for \( \frac{D_3}{D_2} \) the initial value does not influence the result when random errors are not considered.
- Estimates are also sensitive to measurements errors.

The results of the numerical experiments for a nonlinear model \( (m = n = 2, k = 3) \), using a more robust optimization algorithm yield:

- Estimates are exacts when the artificial data do not have random errors.
- Estimates are sensitive to measurement errors, specially the quotient \( \frac{D_3}{D_1} \). A regularization process is required.
- Under the presence of error measurements, the estimates are not significantly different when the data increase in the diffusion zone and large weight coefficients are included in the sum of squares \( J \).

In the nonlinear case, several thousands of problems were solved, generating concentration data \( Z_{data} \) with random perturbation of the values of \( D_1, D_2, D_3 \) given in (11). The concentration data were also perturbed with random errors of 3 %, while random generated quotients \( \frac{D_2}{D_1} \) and \( \frac{D_3}{D_2} \) were used as initial solutions for the optimization algorithm. The following graphics, corresponding to MATLABfigures of 240 problems runs, illustrate the numerical results.

4.1 Some illustrative graphics

Figure 3 shows values of the maximal residual errors between the artificial data and the model data, given by the diffusion coefficients estimated only from four measurements in the first two layers and using weight coefficients equal to one (diamonds). These values are compared to the same difference related to diffusion coefficients estimated from twelve measurements and weight coefficients equal to \( 10^6 \) (asterisks). The residuals values are always small and there are no significant residual differences between both estimates.
Figure 4 shows the ANOVA box picture, which compares residual mean differences and exhibits the distribution of the corresponding data around the mean value.
Figure 5 shows the ANOVA comparison interval, which exhibits no significant difference between the residual means. Only in very few cases (1% of all the numerical experiments) significant differences between those mean values were noticed.

5. CONCLUSIONS

A new approach to a free boundary model with Stefan type conditions which describes the nitriding post-discharge process is presented. Under the assumption of reaching a quasi-steady state for large times, an approximate analytical solution is proposed, which is used to recover diffusion coefficients in each layer through an optimization algorithm. Numerical experiments show a good fitness for the data and quotient $D_1 / D_2$. However, a regularization process should be considered because the quotient $D_1 / D_2$ is always badly estimated when measurement errors are taken into account.

REFERENCES


