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# Prediction of Energy Losses in Soft Magnetic Materials Under Arbitrary Induction Waveforms and DC Bias

Olivier de la Barrière, Carlo Ragusa, Member, IEEE, Carlo Appino, and Fausto Fiorillo

Abstract-The Statistical Theory of Losses (STL) provides a simple and general method for the interpretation and prediction of the energy losses in soft magnetic materials. One basic application consists, for example, in the prediction of the loss under arbitrary induction waveform, starting from data available from conventional measurements performed under sinusoidal flux. There are, however, persisting difficulties in assessing the loss when the induction waveform is affected by a DC-bias, because this would require additional experimental data, seldom available to machine designers. In this paper we overcome this problem applying, with suitable simplifications, the dynamic Preisach model. Here the parameters of the STL model obtained exploiting preemptive conventional are measurements only. By this new simplified method, analytical expressions for the loss components are obtained under general supply conditions, including DC-biased induction waveforms.

*Index Terms*-Magnetic hysteresis, magnetic loss, magnetic materials.

## I. INTRODUCTION

THE optimal design of magnetic components like electrical machines or power electronics transformers in modern applications, e.g. renewable energy production [1] or hybrid traction [2], can be quite a complex task. It often requires the optimization of the electrical machine under complex constraints, with a significant number of working points to be satisfied in the torque vs. speed plan [3]. In embedded applications the torque density [4] is a crucial objective, while it is often necessary to simultaneously maximize the efficiency

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[5] [6]. These multi-objective design problems often require the use of stochastic algorithms, such as evolutionary [7] or particles swarm algorithms [8] [9], to overcome the problem of local extrema of the global objective function [10]. They generally call for a huge number of model evaluations for converging to the optimal solution and the electromagnetic model implemented in the optimization process must offer the best compromise between accuracy and computation time.

In this context, iron loss modeling plays a crucial role, because it has a direct impact on the machine efficiency [11] [12]. The loss calculation is a complex problem, often made more demanding by specific working regimes, where, for example, the induction waveform is distorted or affected by a DC-bias, an ubiquitous circumstance in electrical engineering [13][14]. The relevance of DC-biased and asymmetric inductions in power electronics has also been pointed out [15] and the important case of Pulse Width Modulation (PWM) [16] can also be seen as a problem of polarized induction waveform. Indeed, the main difficulty arising with PWM waveforms is the prediction of the loss associated with minor loops. Each of these minor loops can be considered as a DC-biased cycle of small peak-to-peak amplitude [17]. For electrical machine designers, DC-biased inductions waveforms are mostly found in the rotor of electrical machines. To illustrate this point, let us consider a synchronous permanent magnet machine, which is interesting in hybrid vehicles due to its high torque density [18]. The machine has two pole pairs and non-salient rotor (Fig.1). We are interested in the no-load magnetic flux density in the ferromagnetic rotor (which is in many cases made of laminated material). Fig. 1. shows that the radial induction waveform at a given point of the rotor yoke (obtained, for example, with a finite-element computation) is affected by a DC-bias, with superposed undulation. The bias is generated by the permanent magnets and the oscillation around it is the effect of machine slotting. The space and time harmonics of the stator magnetomotive force can also be responsible for variations of the rotor induction around the DC-bias value [19]. In low cost machines with concentrated windings, it may happen that the rotor loss provides the larger contribution to the iron loss [20], while becoming crucial from the thermal viewpoint, because of the difficulty to cool the rotor.

In the literature, three classes of loss models can be found, each of them offering a different compromise between accuracy and computational complexity [21].



Fig. 1. Radial induction waveform in the rotor yoke of a permanent magnet machine.

The phenomenological models derived from the old Steinmetz formula are often used in electrical engineering [22]. Mostly limited to conventional supply conditions, they consist of fast and simple analytical expressions giving the specific power loss as a function of frequency and peak induction. Authors in [23] have proposed suitable modifications of the Steinmetz equation to account for the induction distortion and the DCbias. Experimental aspects of the loss behavior of DC-biased loops have been discussed in [24] [25]. A two-term loss formulation was proposed in [26], where the first term represents the classical contribution, proportional (in terms of power) to  $(B_p; f)^2$ , with  $B_p$  the peak induction, while the second term, proportional to  $B_p^2 f$ , takes into account the DC bias contribution.

A second type of approach is based on the Dynamic Preisach Model (DPM) [27], which can predict the loss under generic induction waveform, including distorted and DC-biased waveform with and without minor loops. With suitable generalization, the DPM can take into account the skin effect [28], but, as shown in [28] [29], this fully numerical modeling requires heavy calculations, because it is necessary to compute at each instant of time the state of each switching Preisach unit. Recently, however, a simple differential equation describing the dynamic effects of the hysteresis through suitable simplification of the DPM has been proposed [30] [31]. This approach has the advantage of predicting the hysteresis loop and the related loss with good accuracy, close to the one provided by the DPM, but with a largely reduced computational effort.

The third class of models is based on the Statistical Theory of Losses (STL) [32] [33]. By this theory, the loss per cycle W under whatever induction waveform is decomposed at any frequency as:

$$W = W_{\rm h} + W_{\rm cl} + W_{\rm exc} \,, \tag{1}$$

where  $W_{\rm h}$  is the hysteresis (quasi-static) loss,  $W_{\rm cl}$  is the classical loss and  $W_{\rm exc}$  is the excess loss. Following this approach, the classical energy loss component can be computed by the standard equation (valid under negligible skin effect) [33]:

$$W_{\text{class}} = \sigma \frac{d^2}{12} \int_0^{1/f} \left(\frac{dB}{dt}\right)^2 dt , \qquad (2)$$

where  $\sigma$  is the material electrical conductivity, *d* is the sheet thickness, *f* the magnetizing frequency, and *B*(*t*) the instantaneous flux density. Since only the time derivative dB/dt is involved in (2),  $W_{class}$  does not depend on the DC-bias, which influences instead the hysteresis and excess components. The latter can be written as:

$$W_{\rm exc} = \sqrt{\sigma GSV_0} \cdot \int_0^{V_f} \left| \frac{\mathrm{d}J}{\mathrm{d}t} \right|^{3/2} \mathrm{d}t \;, \tag{3}$$

where J is the magnetic polarization ( $J = B - \mu_0 H$ ), S is the cross-sectional area of the lamination, G = 0.1356 is a dimensionless coefficient, and  $V_0$  is a statistical parameter, related to the distribution of the local coercive fields, which depends on the peak induction  $B_p$  and the bias induction  $B_b$ . The authors of [34] [35] have applied the Static Preisach Model (SPM) to compute  $W_{hyst}$ , but their simplified model neglects the  $W_{exc}$  dependence on the DC-bias. The structure dependent  $V_0$  parameter, in particular, is assumed to be equal to the one belonging to the centered hysteresis loop of identical peak-to-peak amplitude, an hypothesis also exploited for the minor loops associated with PWM waveforms [17]. It has been shown that this approximation does not introduce in general important errors, because the excess loss is only a part of the total loss, but it is not physically justified.

This limitation is overcome in this paper, where, by suitable implementation of the simplified DPM [30] [31], it is possible to predict in fast and simple way the loss evolution with the DC bias by predicting the corresponding evolution of the parameters of the STL. It offers an excellent tradeoff between the computation time and the accuracy.

The effect of a DC-bias induction is addressed experimentally, and the theory here outlined is applied to several experimental results obtained on non-oriented and grain-oriented ironsilicon laminations.

## II. THEORETICAL MODEL

We start from a simple dynamical model of hysteresis, introduced in [31] where the relation between the dynamic field H(t) and the magnetic polarization J(t) is decomposed into a dynamic part and a static part. This model is sketched in Fig. 2, where we introduce the *static field*  $H_{\text{stat}}(t)$ , which is defined as the field providing a given magnetic polarization



Fig. 2. Block diagram of the dynamic model.

value under quasi-static conditions. The static field  $H_{\text{stat}}(t)$  is obtained from the dynamic field H(t) (step 1 in Fig. 2) and the polarization J(t) is subsequently computed via the Static Hysteresis Model. As discussed in [31], the following dynamic equation relates H(t) and  $H_{\text{stat}}(t)$ 

$$\dot{H}_{\text{stat}} = \text{sign} \Big[ H(t) - H_{\text{stat}}(t) \Big] \frac{9}{16} k_{\text{d}} \Big[ H(t) - H_{\text{stat}}(t) \Big]^2, \qquad (4)$$

where  $k_d$  is a suitable parameter governing the dynamics of the magnetization reversal [27]. From (4) we obtain the dynamic field H(t) as

$$H(t) = H_{\text{stat}}(t) + sign(\dot{H}_{\text{stat}}) \cdot \frac{4}{3} \sqrt{\frac{|\dot{H}_{\text{stat}}|}{k_{\text{d}}}}.$$
(5)

The second term on the right hand side of (5) is the excess field  $H_{\text{exc}} = H - H_{\text{stat}}$ , which depends on the time derivative of the static field  $H_{\text{stat}}$ . Equation (5) brings then the dynamic field back to the static one, which can in turn be related to the polarization J(t) by means of the Static Preisach Model (SPM). By this model, one finds that for a generic hysteresis cycle taken between  $H_{\text{m}}$  and  $H_{\text{M}}$ , the descending branch  $J_{\text{d}}$  of the given cycle can be computed according to

$$J_{\rm d} = F(H_{\rm M}, -H_{\rm M}) - 2 \cdot F(H_{\rm M}, H_{\rm stat}), \tag{6}$$

where  $F(\alpha,\beta)$  is the Everett's function [36],  $H_m$  and  $H_M$  are the fields required to reach, under static conditions, the lower and upper value of the polarization waveform, and  $H_m \le H_{\text{stat}} \le H_M$ . The condition  $|H_m| \le H_M$  is posed, implying positive DC bias. A similar equation applies for the ascending branch,  $J_a$ :

$$J_{a} = F(H_{M}, -H_{M}) - 2 \cdot F(H_{M}, H_{m}) + 2 \cdot F(H_{\text{stat}}, H_{m}).$$
(7)

In the following we shall apply the method discussed in [36], where the Everett's function is computed on the basis of the minimum preemptive experimental data provided by the *limit* (very high peak induction) experimental cycle. Following [33], the function  $F(\alpha, \beta)$  for generic switching fields  $\alpha$  and  $\beta$  in the Preisach plane is obtained as

$$F(\alpha,\beta) = -\frac{1}{2}J_{d,\text{lim}}(-\alpha) - \frac{1}{2}J_{d,\text{lim}}(\beta) + \Phi(\alpha) \cdot \Phi(-\beta), \qquad (8)$$

where  $J_{d,lim}(\alpha)$  represents the descending branch of the experimental *limit cycle* and  $\Phi$  is the function

$$\Phi(H_{\text{stat}}) = -\sqrt{J_{\text{d,lim}}(0)} \cdot \exp\left[\int_{0}^{H_{\text{stat}}} \frac{-\dot{J}_{\text{d,lim}}(-\alpha) + \mu_{\text{rev}}(\alpha)}{J_{\text{d,lim}}(\alpha) + J_{\text{d,lim}}(-\alpha)} d\alpha\right],$$
(9)

where  $\dot{J}_{\rm d,lim}(\alpha) = dJ_{\rm d,lim}(\alpha)/d\alpha$  is the slope of the descending branch of the experimental *limit cycle*,  $-H_{\rm M,sat} \le H_{\rm stat} \le H_{\rm M,sat}$ , and  $\mu_{\rm rev}$  is the reversible permeability of the material.

## 1) Calculation of the hysteresis loss component

The hysteresis loss component associated with the cycle taken between  $H_{\rm m}$  and  $H_{\rm M}$  is calculated as

$$W_{\text{hyst}}(H_{\text{M}}, H_{\text{m}}) = \oint_{\text{cycle}} H_{\text{stat}} \, \mathrm{d}J = \int_{H_{\text{m}}}^{H_{\text{M}}} H_{\text{stat}} \left(\frac{\mathrm{d}J_{\text{a}}}{\mathrm{d}H_{\text{stat}}} - \frac{\mathrm{d}J_{\text{d}}}{\mathrm{d}H_{\text{stat}}}\right) \mathrm{d}H_{\text{stat}}, \quad (10)$$

where the slopes of the ascending and descending branches are introduced. By further introducing the quantity  $\varphi = d\Phi(H_{stat})/dH_{stat}$ , we obtain from (6)

$$\frac{\mathrm{d}J_{\mathrm{d}}}{\mathrm{d}H_{\mathrm{stat}}} = 2\varphi\left(-H_{\mathrm{stat}}\right)\left[\Phi\left(H_{\mathrm{stat}}\right) - \Phi\left(H_{\mathrm{M}}\right)\right] \tag{11}$$

and

$$\frac{\mathrm{d}J_{\mathrm{a}}}{\mathrm{d}H_{\mathrm{stat}}} = 2\varphi(H_{\mathrm{stat}}) \left[\Phi(H_{\mathrm{stat}}) - \Phi(H_{\mathrm{m}})\right] \tag{12}$$

for the descending and ascending branch, respectively. Eq. (10) can then be written as

$$W_{\text{hyst}} = \int_{H_{\text{m}}}^{H_{\text{M}}} H_{\text{stat}} \cdot \left\{ 2\varphi \left( H_{\text{stat}} \right) \left[ \Phi \left( H_{\text{stat}} \right) - \Phi \left( H_{\text{m}} \right) \right] -2\varphi \left( -H_{\text{stat}} \right) \left[ \Phi \left( H_{\text{stat}} \right) - \Phi \left( H_{\text{M}} \right) \right] \right\} dH_{\text{stat}} , \qquad (13)$$

thereby providing a formal expression for the hysteresis loss component.

2) Calculation of the excess loss component The excess loss component is derived from (5) as

$$W_{\rm exc} = \int_{0}^{1/f} H_{\rm exc} \dot{J} \, dt = \int_{0}^{1/f} \frac{4}{3} \sqrt{\frac{dH_{\rm stat}}{dJ} / k_{\rm d}} \cdot \left| \dot{J} \right|^{3/2} dt \,, \tag{14}$$

where the term  $dH_{\text{stat}}/dJ$  is the inverse of the local slope of the static hysteresis loop. Recalling the expression (3) for  $W_{\text{exc}}$  provided by the STL and comparing it with (14), the statistical parameter  $V_0$ , lumping the effect of the local coercive fields in the STL, is obtained as

$$V_0(J_p, J_b) = \frac{1}{\sigma GS} \left( \frac{1}{T} \int_0^{1/f} \frac{4}{3} \sqrt{\frac{\mathrm{d}H_{\mathrm{stat}}}{\mathrm{d}J} / k_\mathrm{d}}} \,\mathrm{d}t \right)^2. \tag{15}$$

In this derivation we have assumed, without loss of generality, that the statistical parameter  $V_0$  depends only on the bias polarization  $J_b$  and the peak-to-peak swing  $2J_p$ , while being independent on the induction waveform. This permits one to simplify the calculation by assuming a triangular J(t).

In order to perform the integration (15) as a function of the variable  $H_{\text{stat}}$ , we decompose the cycle into the ascending and the descending branches and after some mathematics we obtain

$$V_0(J_p, J_b) = \frac{1}{9\sigma GSk_d J_p^2} \left[ \int_{H_m}^{H_m} \left( \sqrt{\frac{dJ_a}{dH_{stat}}} + \sqrt{\frac{dJ_d}{dH_{stat}}} \right) dH_{stat} \right]^2.$$
(16)

Introducing then (11) and (12) in (16), we get

$$V_{0}(J_{p},J_{b}) = \frac{1}{9\sigma GSk_{d}J_{p}^{2}} \left( \int_{H_{m}}^{H_{M}} \left\{ \sqrt{2\varphi(H_{\text{stat}}) \left[ \Phi(-H_{\text{stat}}) - \Phi(-H_{m}) \right]} + \mu_{\text{rev}}(H_{\text{stat}}) \right\} + \sqrt{2\varphi(-H_{\text{stat}}) \left[ \Phi(H_{\text{stat}}) - \Phi(H_{M}) \right]} + \mu_{\text{rev}}(H_{\text{stat}}) \left\{ \Phi(H_{\text{stat}}) - \Phi(H_{M}) \right\} + \mu_{\text{rev}}(H_{\text{stat}}) \left\{ \Phi(H_{M}) - \Phi(H_{M}) \right\} + \mu_{\text{rev}}(H_{\text{stat}}) \left\{ \Phi(H_{M}) - \Phi(H_{M}) \right\} + \mu_{\text{rev}}(H_{\text{stat}}) \left\{ \Phi(H_{M}) - \Phi(H_{M}) \right\} + \mu_{\text{rev}}(H_{M}) \right\} + \mu_{\text{rev}}(H_{M}) + \mu_$$

To simplify the whole matter, one might emulate the *limit* cycle by an analytical expression and make explicit the function  $\Phi$  and its derivative  $\varphi$ . Typically, a hyperbolic tangent expression does provide a good description of the *limit* cycle [37] and the descending branch of the *limit* cycle is correspondingly obtained as

$$J_{\rm d}(H_{\rm stat}) = A_0 \tanh \frac{H_{\rm stat} + H_{\rm C}}{\varsigma H_{\rm C}} + \mu_{\rm rev} \cdot H_{\rm stat}, \qquad (18)$$

with  $A_0$ ,  $H_C$ ,  $\varsigma$ , and  $\mu_{rev}$  fitting parameters. Combining (18) and (9) we get

$$\Phi(H_{\text{stat}}) = \frac{\sqrt{A_0 \tanh(1/\varsigma) \cdot \cosh(1/\varsigma)}}{\cosh\left(\frac{H_{\text{stat}} - H_{\text{C}}}{\varsigma H_{\text{C}}}\right)} \frac{\exp\left(-\frac{(1/\varsigma)}{\tanh(2/\varsigma)}\right)}{\exp\left(\frac{H_{\text{stat}} - H_{\text{C}}}{\varsigma H_{\text{C}} \tanh(2/\varsigma)}\right)}$$
(19)

and after derivation

$$\varphi(H_{\text{stat}}) = \frac{1}{\varsigma H_{\text{c}} \sinh\left(2/\varsigma\right)} \frac{\cosh\left(\frac{H_{\text{stat}} + H_{\text{c}}}{\varsigma H_{\text{c}}}\right)}{\cosh\left(\frac{H_{\text{stat}} - H_{\text{c}}}{\varsigma H_{\text{c}}}\right)} \Phi(H_{\text{stat}}) \cdot$$
(20)

By introducing the so-obtained  $\Phi(H_{\text{stat}})$  and  $\varphi(H_{\text{stat}})$  in (13) and (17), the hysteresis loss  $W_{\text{hyst}}$  and the excess loss  $W_{\text{exc}}$  (via (3)) can be numerically calculated. To remark that if a fitting formula for the *limit* experimental cycle different from (18) is assumed, the expressions (19) and (20) for the functions  $\Phi$  and  $\varphi$  are modified, but those for  $W_{\text{hyst}}$  and  $V_0$  remain valid.

#### III. EXPERIMENTAL RESULTS

We discuss now two examples of validation of the discussed predictive model based on experiments performed on non-oriented Fe-Si sheets of thickness d = 0.194 mm, and grain-oriented Fe-Si laminations of thickness d = 0.28 mm.

## A. Practical implementation of the model

Let us take the case where a hysteresis loop of peak-to-peak amplitude  $2J_p$  is measured under bias polarization of value  $J_b$ . In order to apply (13) and (17) and arrive at  $W_{hyst}$  and  $V_0$ , we need to determine the static fields  $H_{\rm m}$  and  $H_{\rm M}$  permitting one to reach the polarization levels  $J_{\rm b}$ - $J_{\rm p}$  and  $J_{\rm b}$ + $J_{\rm p}$ , respectively. This might be quite a complex problem, because one should take into account the magnetic history of the material prior to the moment where the biased cycle is executed. In practical applications, no information is typically available regarding such a history and a numerical inversion of the static hysteresis model is required. A simpler strategy is therefore adopted here, in order to approximately determine  $H_{\rm m}$  and  $H_{\rm M}$ , which consists in determining the anhysteretic magnetization curve, identified with the single valued J(H) curve intermediate between the ascending and descending branches of a major hysteresis loop. This J(H) can be inverted quite easily, in order to derive  $H_{\rm m}$  and  $H_{\rm M}$  from the peak polarization values  $J_b$ - $J_p$  and  $J_b$ + $J_p$ , as illustrated in Fig. 3. In conclusion, only the major hysteresis loop of the material is required by the model, in order to proceed with the identification process and for achieving the anhysteretic J(H)curve. A flow chart on the model implementation is given in Fig. 4.



Fig. 3. Determination of the fields  $H_m$  and  $H_M$  corresponding to the lower  $J_b$ - $J_p$  and upper  $J_b$ + $J_p$  polarization values, respectively.

#### B. Model identification

#### 1) "Static" parameters

According to (19) and (20), the previously introduced parameters  $A_0$ ,  $H_C$ ,  $\varsigma$ , and  $\mu_{rev}$  need to be retrieved in order to calculate the  $\Phi$  and  $\varphi$  functions. Their values are obtained by fitting the descending branch of the major static hysteresis loop by (18). This operation provides for the NO sheets the values:  $A_0 = 1.33$  T,  $H_C = 39.4$  A/m,  $\varsigma = 1.1$ , and  $\mu_{rev} = 150 \cdot \mu_0$ . The measured conductivity of this Fe-Si alloy is  $\sigma = 1.99 \cdot 10^6$  S·m<sup>-1</sup>.



Fig. 4. Flow chart of the model implementation.

## 2) Derivation of the dynamic constant k<sub>d</sub>

The constant  $k_d$  is related to the function  $V_0(J_p, J_b = 0)$ , which can be determined by the conventional loss separation (3), according to the STL, on symmetric loops  $(J_b = 0)$  [17]. Once the curve  $V_0(J_p, J_b = 0)^{1/2} \cdot J_p^{3/2}$  is obtained in this way, calculation of  $V_0(J_p, J_b = 0)$  is made by (17) and the value of the constant  $k_d$  providing best fitting of the curve is retained. Fig. 5. illustrates such a fitting (least squares algorithm), providing the value of the dynamic constant  $k_d = 1200 \text{ mA}^{-1}\text{s}^{-1}$ . To sum up the identification procedure, a flow chart is given in Fig. 6.

## C. Experiments

## 1) Experimental procedure

Measurements on NO sheets have been performed with and without DC bias up to 1000 Hz, below the frequency range where the skin effect is likely to appear [31]. A 700-turn Epstein frame has been employed up to 400 Hz, substituted by a 200-turn frame at higher frequencies. Unambiguous measurements can be performed without DC bias and do not need detailed discussion here. Specifications are instead required for the DC-biased measurements, because several options regarding the way the experiments are performed are possible, according to the magnetic history of the material. In [34] the sample is, for example, brought to the saturated state and the minor loop is run along the descending branch of the major cycle. In the present experiments, the sample has been



Fig. 5. Behavior of the quantity  $(V_0)^{1/2} \cdot J_p^{3/2}$  versus  $J_p$  with centered hysteresis loops in the NO 0.194 mm thick Fe-Si sheet. The results obtained by conventional loss separation procedure (experimental) are compared with the prediction of (17) using best fitting value of the dynamic constant  $k_d$ .

carefully demagnetized at a low frequency before applying the DC bias. As shown in Fig. 7, the DC-biased cycle, of frequency f = 1/T, is nested inside a major symmetric loop of period equal to 10*T*. The acquisition of the specific DC-biased cycle is carried out at the end of the positive half-period of the major cycle. Before acquisition, a number of identically DC-biased cycles are made in order to reach steady state (it was shown in [34] that a few cycles are required before reaching a stable periodic behavior of the minor loop).

## 2) Results

A peak polarization  $J_p = 0.5$  T has been chosen, and three different values of DC bias have been considered:  $J_b = 0$  T (centered case),  $J_b = 0.5$  T, and  $J_b = 0.75$  T. A conventional loss separation procedure is performed, by which the experimental excess loss  $W_{exc}$  is obtained by subtracting the measured hysteresis loss and the classical loss components



Fig. 6. Flow chart of the identification procedure, by which the parameters A,  $H_c$ ,  $\zeta$ ,  $\mu_{rev}$ , and  $k_d$  are obtained from the static and dynamic data.



Fig. 7. Application of the DC-biased polarization waveform.

from the measured energy loss (symbols in Fig. 8a). The value of the parameter  $V_0(J_p, J_b)$  is then obtained by (16) and the excess loss is calculated with (3) (lines in Fig. 8a). It is found that the increase of the slope of the experimental  $W_{\text{exc}}(f)$  versus  $f^{1/2}$  behavior is correctly taken into account by the function  $V_0(J_p, J_b)$ . Fig.8b shows the behavior of the measured loss (symbols) and the predicted one (lines) obtained by adding the classical loss (2), the previously calculated  $W_{\text{exc}}(f)$ , and the hysteresis loss  $W_{\text{hyst}}$  calculated with (13). Neglecting the DC bias can lead to significant errors: for example, for a bias of 0.75 T, the loss at 100 Hz is ~23% greater than the value obtained for unbiased conditions.

## 3) Validation on grain-oriented Fe-Si sheets

An advantage of this material is its relatively important permeability, making it easier to reach higher polarization levels than in NO sheets. The GO sheets are also characterized by wide domain wall spacing, the source of a remarkable excess loss contribution, and provide a significant applicative example of the model. The same peak-to-peak induction of  $2 \cdot J_p = 1$  T has been considered, and DC bias up to  $J_b = 1.2$  T has been reached. The previously described identification procedure provides the following parameters:  $A_0 = 1.87$  T,  $H_C = 6.8$  A/m,  $\varsigma = 0.5$ , and  $\mu_{rev} = 200 \cdot \mu_0$ . The conductivity is  $\sigma = 2.08 \cdot 10^6$  S·m<sup>-1</sup>. The dynamic constant has been found equal to  $k_d = 57$  mA<sup>-1</sup>s<sup>-1</sup>. The results, showing a satisfying agreement with the theory, are given in Fig. 9.



Fig. 8 Non-oriented Fe-Si 0.194 mm thick sheet under sinusoidal induction waveform ( $J_p = 0.5 \text{ T}$ ;  $J_b = 0$ ,  $J_b = 0.5 \text{ T}$ , and  $J_b = 0.75 \text{ T}$ ). Experimental values and outcomes of the proposed theoretical model are compared vs. frequency. a) Excess loss  $W_{exc}$ . b) Total loss W.



Fig. 9. Grain-oriented Fe-Si 0.28 mm thick sheets under sinusoidal induction waveform ( $J_p = 0.5$  T;  $J_b = 0$ ,  $J_b = 0.75$  T, and  $J_b = 1.2$  T). Experimental and theoretical loss vs. frequency: a) Excess loss  $W_{exc}$ ; b) Total loss *W*.

## IV. CONCLUSION

In this article, the importance of DC-biased loops in applications has been highlighted. A simple computational scheme, based on the extension of the Statistical Loss Theory to DC-biased waveforms, has been proposed. Semi-analytical formulae have been derived for the hysteresis contribution, and the  $V_0$  parameter of the Statistical Theory of Losses. The model requires acquisition and fitting of the major symmetric static hysteresis loop, and the identification of a single constant  $k_d$  governing the dynamic magnetization process. This dynamic constant is obtained by fitting the statistical parameter  $V_0(J_p)$  experimentally found with unbiased sinusoidal induction. Two examples of application of the theoretical model, regarding 0.194 mm thick non-oriented and 0.28 mm thick grain-oriented Fe-Si sheets, have been discussed. Measurements performed under DC-bias are shown to be in good agreement with the prediction, both regarding the excess loss and the hysteresis loss components. This model appears then to provide a computationally simple predicting tool for electrical engineers facing the problem of DC-biased regimes, relieving the designer from the quest for cumbersome measurements under polarized induction.

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