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Robust compact model for bipolar oxide-based resistive switching memories

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Abstract—Emerging non-volatile memories based on resistive switching mechanisms pull intense R&D efforts from both academia and industry. Oxide-based Resistive Random Access Memories (namely OxRAM) gather noteworthy performances, such as fast write/read speed, low power, high endurance and large integration density that outperform conventional Flash memories. To fully explore new design concepts such as distributed memory in logic or biomimetic architectures, robust OxRAM compact models must be developed and implemented into electrical simulators to assess performances at a circuit level. In this paper, we propose a physics-based compact model used in electrical simulator for bipolar OxRAM memories. After uncovering the theoretical background and the set of relevant physical parameters, this model is confronted to experimental electrical data. The excellent agreement with these data suggests that this model can be confidently implemented into circuit simulators for design purpose.

INTRODUCTION

Memory devices based on resistive switching materials are currently pointed out as promising candidates to replace conventional non-volatile memory (NVM) devices based on charge-storage beyond 2x nm technological nodes [1]. Indeed, as compared to conventional floating gate technologies, Resistive RAMs (so-called RRAM) gather fast write/read operations, low power consumption, CMOS voltage compatibility and high endurance [2]. Moreover, the resistive memory element generally consists in simple Metal/Insulator/Metal (MIM) structure. Hence, whatever the underlying physics, the resistive switching memory elements may be advantageously integrated into back-end-of-line (BEOL) enabling NVM solutions to be distributed over CMOS logic. Depending on fundamental physical mechanisms responsible for resistance switching, various RRAM technologies are now categorized by the ITRS. The Redox Memory category, covered in this study, includes Conductive Bridge RAM (CBRAM) [3] and Oxide Resistive RAM (OxRAM) [4] that exhibit voltage polarity-dependent bipolar switching. Besides, RRAM technologies referred as Thermo-Chemical Memories (TCM), or fuse-antifuse memories, are mostly based on materials such as nickel oxide (NiO) that exhibit unipolar switching.

In the OxRAM memory elements addressed in this paper, a MIM structure is generally composed of two passive metallic electrodes sandwiching an active layer, usually an oxygen-deficient oxide. A large number of resistive switching oxides, like HfO2 or Ta2O5, are reported in the literature [5], [6]. Even if OxRAM technology is still in its infancy, it is commonly accepted that the Valency Change Mechanism (VCM) occurs in specific transition metal oxides and the field-assisted motion of anions, such as oxygen ions $O^{2-}$, governs the bipolar resistance switching [7].

After an initial Electroforming step (cf Fig. 1b), the memory element may be reversibly switched between a High Resistance State (Fig. 1d-HRS) and a Low Resistance State (Fig. 1c-LRS). The Electroforming stage corresponds to a voltage-induced resistance switching from an initial very high resistance state (pristine state) to a conductive state. Resistive switching in OxRAM elements corresponds to an abrupt change between HRS ($R_{HRS}$) and LRS ($R_{LRS}$) resistances. This resistance change is achieved by sweeping a voltage across the MIM structure; set operation corresponds to a HRS-to-LRS transition at $V_{Set}$ while reset operation enables turning back the structure into HRS state by applying $V_{Reset}$ of opposite polarity. It has to be mentioned that the Electroforming voltage $V_{forming}$ is generally larger than $V_{Set}$ even if several groups have recently demonstrated forming-free structures by adjusting the oxygen stoichiometry of the active layer [8].

Thanks to their low operating voltage (typically 1 V), their fast read/write access times (tens of nanosecond) [2] and their advantageous integration into BEOL, OxRAM memories pave the way to new design solutions such as distributed memory in logic or biomimetic architectures. Targeting efficient design solutions, the compatibility between memory elements and logic blocks must be beforehand evaluated. Hence, a robust OxRAM compact model is required to assess and validate new concepts before fabrication. This compact model must (i) rely on realistic physical mechanisms described by a set of relevant parameters; (ii) match actual experimental data (quasi-static measurements, temperature dependence, timing...); (iii) guarantee a good predictability up to the system level; (iv) fulfill a computational-efficient implementation.

In the literature, many groups have proposed physical models for Set/Reset mechanisms [8]–[14], but their intrinsic complexity excludes any implementation into electrical simulators. In contrast, other models are fully compatible with simula-
Nevertheless, recent results showed that setting effect by drift/diffusion of oxygen vacancies [9]–[12]. In transient simulations, temperature dependence, dynamical characteristics...) based on the ground and the set of relevant physical parameters, this model is confronted to experimental electrical data (DC behavior, voltage or time dependencies [15]–[17]. In this context, we propose a robust physics-based compact model for bipolar OxRAM memories that supports efficient implementation into electrical simulators. After uncovering the theoretical background, and the conduction (Poole-Frenkel, Schottky emission, space charge limited current...), but an ohmic behavior is considered for the sake of simplicity.

A. Set/Reset operation

Set operation relies on an electrochemical reaction whose charge transfer rate can be described by the Butler-Volmer equation [19]. From this equation the electrochemical reduction rate $\tau_{Red}$ (Eq. 1) and oxidation rate $\tau_{Ox}$ (Eq. 2) can be derived, here $k_b$ denotes the Boltzmann constant, $E_a$ an activation energy, $\alpha$ the charge transfer coefficient (ranging between 0 and 1) and $\tau_{RedOx}$ the nominal redox rate. The growth/destruction of the filament then results from the interplay between both redox reaction velocities through the master Eq. 3, with the CF radius $r_{CF}$ ranging from 0 to $r_{CFmax}$.

$$\tau_{Red} = \tau_{RedOx} \cdot e^{\frac{E_a - q \cdot (1 - \alpha) \cdot V_{Cell}}{k_b \cdot T}}$$  \hspace{1cm} (1)

$$\tau_{Ox} = \tau_{RedOx} \cdot e^{\frac{E_a - q \cdot \alpha \cdot V_{Cell}}{k_b \cdot T}}$$  \hspace{1cm} (2)

$$\frac{dr_{CF}}{dt} = \frac{r_{CFmax} - r_{CF}}{\tau_{Red} + \tau_{Ox}}$$  \hspace{1cm} (3)

where $V_{Cell}$ is the voltage applied between the top and the bottom electrodes, $q$ is the elementary charge of electron, $k_b$ is the Boltzmann constant, $T$ is the temperature in the structure.

B. Electroforming stage

In addition to the Set operation, Electroforming converts a highly resistive pristine oxide into a switchable sub-oxide region. After this step, standard Set/Reset operation may then occur. Due to the higher voltage bias required during Forming, with respect to Set operation, a CF is generally formed concomittantly to the sub-oxide region after Forming (Fig. 1c). The Electroforming rate $\tau_{Form}$ is given in Eq. 4. The growth of the sub-oxide region is controlled by Eq. 4, where $E_{aForm}$ is the activation energy for Electroforming and $\tau_{Form0}$ the nominal forming rate.

$$\tau_{Form} = \tau_{Form0} \cdot e^{\frac{E_{aForm} - q \cdot \alpha \cdot V_{Cell}}{k_b \cdot T}}$$  \hspace{1cm} (4)

$$\frac{dr_{CFmax}}{dt} = \frac{r_{work} - r_{CFmax}}{\tau_{Form}}$$  \hspace{1cm} (5)
C. Temperature dependence

As shown by Govoreanu et al. [22], temperature plays crucial role on the reaction rates. In our model, the local temperature of the filament is computed from the heat equation given in Eq. 6. Considering a cylinder-shaped filament, the temperature is given by Eq. 7. For instance, Eq. 8 gives maximum temperature reached into the CF at \( x = 0 \), the middle of filament.

\[
\sigma(x) \cdot F(x)^2 = -K_th \frac{\partial^2 T(x)}{\partial x^2} \tag{6}
\]

\[
T(x) = T_amb + \frac{V_{\text{Cell}}^2}{2 \cdot L_x \cdot K_th} \left( \frac{L_x^2}{4} - x^2 \right) \sigma_{\text{eq}} \tag{7}
\]

\[
T = T_amb + \frac{V_{\text{Cell}}^2}{8 \cdot K_th} \cdot \sigma_{\text{eq}} \tag{8}
\]

\[
\sigma_{\text{eq}} = \sigma_{CF} \cdot \frac{r_{CF}^2}{r_{\text{work}}^2} - \sigma_{OX} \cdot \frac{r_{CF_{\max}}^2 - r_{CF}^2}{r_{\text{work}}^2} \tag{9}
\]

where \( \sigma(x) \) is the local electrical conductivity, \( F(x) \) is the local electric field, \( T_amb \) is the ambient temperature and \( \sigma_{CF} \) (resp. \( \sigma_{OX} \)) the electrical conductivity of the conductive filament (resp. switchable sub-oxide) and \( K_th \) is the thermal conductivity. \( \sigma_{eq} \) is the equivalent electrical conductivity in the work area.

Let’s us mention that during Set operation, the temperature increases due to the increase of CF radius; a positive feedback loop thus occurs leading to a self-accelerated reaction. Conversely, during Reset operation, both the radius and the temperature decrease in the CF, this feature lead to a self-limited reaction also referred as a soft Reset [23].

D. Current through the MIM structure

The total current flowing through the OxRAM memory element is the sum of three different contributions (Eq. 10): the first one is related to the conductive area (\( I_{CF} \)); the second one that describes the conduction through the switchable sub-oxide (\( I_{\text{Sub-oxide}} \)); the last contribution arises from conduction through the unswitched pristine oxide (\( I_{\text{Pristine}} \)). \( I_{CF} \) and \( I_{\text{Sub-oxide}} \) (Eqs. 11 & 12 respectively) are described as ohmic contributions; this assumption as already been applied efficiently for TCM [24] and has proven to be accurate without sacrificing the ease of numerical implementation. Conduction in the pristine oxide region is described by means of a tunneling current given in Eq. 13 [25].

\[
I_{\text{Cell}} = I_{\text{Sub-oxide}} + I_{CF} + I_{\text{Pristine}} \tag{10}
\]

\[
I_{CF} = F \cdot \pi \cdot \sigma_{CF} \cdot r_{CF}^2 \tag{11}
\]

\[
I_{\text{Sub-oxide}} = F \cdot \pi \cdot \sigma_{OX} \cdot (r_{CF_{\max}}^2 - r_{CF}^2) \tag{12}
\]

\[
I_{\text{Pristine}} = S_{\text{Cell}} \cdot A \cdot F^2 \cdot \exp \left( -\frac{B}{F} \right) \tag{13}
\]

\[
A = \frac{m_e \cdot q^3}{8 \pi \cdot h \cdot m_e \cdot \phi_b} \tag{14}
\]

where \( B = \sqrt{2 m_e q L_x F} \) and \( m_e \) and \( m_{ox} \) are the effective electron masses into the cathode and oxide respectively; \( h \) is the Planck constant; \( \phi_b \) the metal-oxide barrier height; \( S_{\text{Cell}} \) the section of the device.

E. Numerical implementation

The implementation of a compact model into electrical simulation tools requires a discrete resolution of a set of differential equations. If the time step is sufficiently small, \( \tau_{Red}, \tau_{Ox} \) and \( \tau_{Form} \) be assumed constant and the discrete forms of Eqs. 3&5 are given in Eqs. 14&15. Solving these differential equations step by step ensures a better convergence of the simulation.

\[
r_{CF_{max},i+1} = (r_{CF_{max}} - r_{\text{work}}) \cdot e^{\frac{-\Delta t}{\tau_{Form}}} + r_{\text{work}} \tag{14}
\]

\[
r_{CF_{i+1}} = \left( r_{CF_i} - r_{CF_{max}} \cdot \frac{\tau_{eq}}{\tau_{Red}} \right) e^{\frac{\Delta t}{\tau_{eq}}} + r_{CF_{max}} \cdot \frac{\tau_{eq}}{\tau_{Red}} \tag{15}
\]

where \( \tau_{eq} = \frac{\tau_{Red} \cdot \tau_{Ox}}{\tau_{Red} + \tau_{Ox}} \).

These equations were then implemented within an ELDO compact model following the flowchart given in Fig. 2. At each call of the OxRAM instance during a transient simulation, the previous state of the filament as well as the applied voltage are provided to the model in order to take into account for the memory effect. The new filament state and the current are then computed as function of these inputs and the given time step.
II. MODEL VALIDATION

To validate the proposed theoretical approach, the model was confronted to quasi-static and dynamic experimental data extracted. First, the compact model was calibrated on recent electrical data measured on HfO$_2$-based OxRAM devices [18]. In this study, the memory elements consisted in a Ti/HfO$_2$/TiN stack with a 5 nm tick hafnium oxide. The set of physical parameters given in Table I, together with the actual hafnium oxide film thickness, the present model shows an excellent agreement with the experimental data for both HfO$_2$-based memory elements [18]. Using the set of physical parameters given in Table I, together with the actual hafnium oxide film thickness, the present model shows an excellent agreement with the experimental data for both Set and Reset operations.

A peculiar attention must be paid to the dependence of the resistance in LRS state against the maximum current allowed during Set operations. As reported in previous works, the resistance in LRS state (noted $R_{LRS}$) and Reset current strongly depend on the maximum current reached during the preceding Set operation [4], [26]–[30] (referred as $I_{CompSet}$). This feature can be understood in terms of reduction of CF radius that concomitantly increases the resistance of the MIM structure [26]. Fig. 5 shows the experimental evolutions of reset current $I_{reset}$ as a function of maximum set current $I_{CompSet}$ [26]. The proposed model matches well the experimental data obtained by various authors and confirms the scalability trend of Reset current $I_{reset}$ in resistive memories. The reset current may be scaled down by limiting the maximum Set current through an integrated select device (e.g. transistor or load resistor) in series with memory element.

Another marker of OxRAM is the soft-Reset that induces a dependency between resistance in HRS ($R_{HRS}$) and the stop voltage ($V_{stop}$) during the preceding reset operation [23]. This feature can be understood as an incomplete destruction of the CF as shown in Fig. 1d. Fig. 6a shows that the dependence of $R_{HRS}$ versus $V_{stop}$ is well captured by the model thanks to the thermal activation of Set and Reset mechanisms. Indeed, since the filament has a uniform section, no thermal confinement occurs. There is an interplay between temperature, current...
Density and the CF radius. Since both the CF radius and the temperature decrease during Reset, the reaction-rate, which also depends on temperature (Eq. 2), will decrease. This feature means that the reaction-rate is self-limited leading to a soft-Reset.

B. Dynamic characteristic

Another important feature for designers is the dependence of Set and Reset switching times as a function of the applied voltage $V_{Cell}$. Fig. 6b shows switching time for Set operation as a function of voltage presented in [23] and corresponding simulation results.

Fig. 6: a) Experimental $R_{HRS}$ as a function of stop voltage during Reset operation and b) switching time for Set operation as a function of voltage presented in [23] and corresponding simulation results.

As shown in Fig. 7 and as already demonstrated in Refs. [31], [32], the Electroforming voltage $V_{Forming}$ is activated in temperature, typically $-0.005V/°C$, so that $V_{Forming}$ is almost divided by a factor 2 from room temperature (RT) to $473K$. If RRAM switching capabilities are evaluated in temperature, one can see that both set/reset voltages exhibit less than $50mV$ variation in the investigated temperature range. Again, both of these behaviors (i.e., high thermal activation of $V_{Forming}$ and low activation of $V_{Set}$/$V_{Reset}$) are well captured by our model with the set of physical parameters given in Table I.

D. Device-to-device variability

Even if memory devices relying on a resistance change are attracting a lot of R&D effort, their technological deployment is still in its infancy. Therefore, many challenging tasks such as electrical or technological variability are still to be addressed to fulfill the requirements of mass production. Let us mention that there are two relevant types of variability: device-to-device variability that characterizes the uniformity within a memory array; cycle-to-cycle variability that characterizes device stability along with cycling [33], [34]. As a consequence, there
is an increasing demand to implement such variability in the compact model to apprehend their impact at a circuit level.

Fig. 8 depicts $I(V)$ characteristics for Electroforming, Set and Reset operations measured on a large number of memory elements coming from [18], [23]. A standard deviation of about ±5% on selected parameters ($\alpha$ and $L_x$) enables satisfactorily accounting for the device-to-device variability. This dispersion can be explained by local variations of thickness, referring to $L_x$, and composition, affecting $\alpha$. A modulation of $\alpha$ impacts the creation/destruction kinetics of the CF which – in turn – introduces a variability in the programming voltage ($V_{\text{Set}}$ & $V_{\text{Reset}}$) and in the CF width. In the other hand, $L_x$ variation impacts the current though the virgin oxide, thus the current before forming. All of these variations can be interpreted in terms of local structural or chemical variation of the oxide: crystallinity, grain boundaries, and interface roughness.

Fig. 9 shows the cumulative distributions of switching voltages (Fig. 9a) and resistances in pristine and HRS states (Fig. 9b) extracted from $I(V)$ curves measured on few tens of devices (Fig. 8). The resistance in LRS is controlled by a select transistor. Since the amount of current available during Set controls $R_{\text{LRS}}$, it is likely that $R_{\text{LRS}}$ is partially impacted by the transistor variability which not studied here. Electrically, these variations impact Set and Reset voltages (Fig. 9a), but also the LRS (resp. HRS) resistance. Note that the parameters variation used to plot Fig. 9 simulation data are included in a distribution following a Gaussian law.

Fig. 9b plots the pristine and HRS resistance variations for the OxRAM considered in this study. The model (lines) is consistent with experimental trends (symbols). Since LRS is mostly controlled by external compliance current ($I_{\text{CompSet}}$), pristine and high resistance state are generally more affected than LRS in terms of electrical variation.

In conclusion, this paper deals with a physics-based compact model that is demonstrated robust for simultaneously describing Electroforming, Set and Reset operations in bipolar resistive switching memories based on HfO$_2$ active layer. By gathering local electrochemical reactions and heat equation in a single master equation, the model enables accounting for both creation and destruction of conductive filaments. The simulation results satisfactorily match quasi-static and dynamic experimental data measured on actual resistive switching devices. Beside, the compact model may be used as a suitable tool for predicting the temperature dependence of switching parameters. Finally, the model fulfills the expectations in terms of implementation into circuit simulators and enables forecasting relevant trends required for designing innovative biomimetic architectures or for proposing novel solutions of distributed memory in logic.

**APPENDIX A**

**Set/Reset EQUATION BASED ON REDOX KINETIC**

The Butler-Volmer equation describes the electrical current on an electrode with respect to the electrode potential [19]:

$$j = j_0 \cdot \left[ e^{\frac{\alpha z F}{R T} (E - E_{eq})} - e^{\frac{(1-\alpha) z F}{R T} (E - E_{eq})} \right]$$  \hspace{1cm} (17)

$j$ is the current density, $j_0$ the exchange current density, $E$ is the electrode potential, $E_{eq}$ is the equilibrium potential, $T$ is the temperature, $z$ is the number of electrons involved in the electrode reaction, $F$ is the Faraday constant, $R$ is the universal gas constant, $\alpha$ is the charge transfer coefficient. This expression derives from the electrochemical kinetics through a simple redox reaction, that is to say a single step mechanism:

$$\text{O}^{2+} + z e^{-} \overset{\text{red}}{\underset{\text{ox}}{\rightleftharpoons}} R$$  \hspace{1cm} (18)
with R is the reductant and O is the oxidant. In this case, the reaction rates for both reduction and oxidation processes may be expressed by the classical expression:

\[ \nu_{\text{red}} = k_0 e^{-\frac{\Delta G_{0,\text{red}} + q \cdot F \cdot (E - E_{\text{eq}})}{R \cdot T}} \cdot C_R \]  

\[ \nu_{\text{ox}} = k_0 e^{-\frac{\Delta G_{0,\text{ox}} - (1 - \gamma) \cdot q \cdot F \cdot (E - E_{\text{eq}})}{R \cdot T}} \cdot C_O \]  

\[ \frac{dC_A}{dt} = \nu_{\text{red}} - \nu_{\text{ox}} \]  

\[ \Delta G_0 = \ln \left( \frac{\text{SetR}}{\text{SetO}} \right) \]

Equation (21) gives the concentration of metallic species, and the description of metallic and ionic species is given by Eq. (22). The redox potential \( E \) is assumed to be equal to \( -|V_{\text{Cell}}| \). Regarding Eq. (1), it is assumed that electron density available during \( \text{Set} \) could not limit the reduction process. Indeed, during \( \text{Set} \) the growth of the CF is initiated at the cathode side. As the CF grows toward the anode, it is assumed to have isotropic and the redox potential \( E \) is assumed to be equal to \( -|V_{\text{Cell}}| \). Regarding Eq. (18), we assumed that electron density available during \( \text{Set} \) could not limit the reduction process. Indeed, during \( \text{Set} \) the growth of the CF is initiated at the cathode side. As the CF grows toward the anode, it is assumed to have isotropic and the redox potential \( E \) is assumed to be equal to \( -|V_{\text{Cell}}| \).

Finally using \( E_a[\beta] \) rather than \( \Delta G_0[\gamma \text{ mol}^{-1}] \), CF radius \( r_{\text{CF}} = r_{\text{CFmax}} \times C_R \) is calculated from the concentration of metallic species (Eq. (21)) and the description of \( \text{Set} \) and \( \text{Reset} \) mechanisms enables proposing a self-consistent master equation relying on a chemical kinetics Eq. (3).

\[ E_a - q \cdot \alpha \cdot V_{\text{Cell}} \]

\[ \tau_{\text{Red}} = \tau_{\text{RedOx}} \cdot e^{\frac{k_0 \cdot T}{r_{\text{CF}}}} \]

\[ \tau_{\text{Ox}} = \tau_{\text{RedOx}} \cdot e^{\frac{k_0 \cdot T}{r_{\text{CF}}}} \]

\[ \frac{dr_{\text{CF}}}{dt} = \frac{r_{\text{CFmax}} - r_{\text{CF}}}{\tau_{\text{Red}}} - \frac{\alpha \cdot E_a}{r_{\text{CF}} \cdot \tau_{\text{Ox}}} \]

\[ k_0 e^{-\frac{\Delta G_0}{R \cdot T}} = \frac{1}{\tau_{\text{RedOx}} \cdot e^{\frac{k_0 \cdot T}{r_{\text{CF}}}}} \]

**APPENDIX B**

**EXPRESSION ANALITIQUE**

It is interesting to note that an analytical expression of \( R_{\text{HRS}} \) as a function of stop voltage during \( \text{Reset} \) operation can be expressed simply if \( r_{\text{CF}} \neq 0 \):

\[ R_{\text{HRS}} \sim \frac{L_x \cdot V_{\text{Stop}}}{8\pi K_{\text{th}} \cdot r_{\text{work}}^2 \cdot \left( \frac{E_a + q \cdot (1 - \gamma) \cdot V_{\text{Stop}} - T_{\text{amb}}}{k_b \cdot \ln \left( \frac{t}{\tau_{\text{RedOx}}} \right)} \right)} \]

It is interesting to observe that the switching time during the \( \text{Set} \) is directly proportional to \( \tau_{\text{Red}} \). In this way the slope of \( \ln (t_{\text{switch}}) \) can be explained:

\[ \ln (t_{\text{Set}}) = \frac{q \cdot \alpha}{k_b \cdot T_{\text{amb}}} \left( \frac{E_a}{q \cdot \alpha} - V_{\text{Cell}} \right) + \text{cst} \]

\[ \ln (t_{\text{Forming}}) = \frac{q \cdot \alpha}{k_b \cdot T_{\text{amb}}} \left( \frac{E_{\text{form}}}{q \cdot \alpha} - V_{\text{Cell}} \right) + \text{cst} \]

The analytic expressions of \( V_{\text{Set}} \) and \( V_{\text{Form}} \) can be obtained from the expressions of the kinetic reaction \( \text{Set} \) (Eq. (1)) and \( \text{Electroforming} \) (Eq. (4)):

\[ V_{\text{Set}} \approx \frac{E_a}{q \cdot \alpha} - k_b \cdot T_{\text{amb}} \cdot \ln \left( \frac{E_a}{q \cdot \alpha} \cdot \beta \cdot \frac{1}{\tau_{\text{RedOx}}} \right) \]

\[ V_{\text{Form}} \approx \frac{E_{\text{form}}}{q \cdot \alpha} - k_b \cdot T_{\text{amb}} \cdot \ln \left( \frac{E_{\text{form}}}{q \cdot \alpha} \cdot \beta \cdot \frac{1}{\tau_{\text{Form}}} \right) \]

These expressions can help to extract \( E_a, E_{\text{form}}, \tau_{\text{RedOx}} \) and \( \tau_{\text{Form}} \).

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