Self Adaptive Computation of the Breakdown Voltage of Planar pn–Junctions with Multistep Field Plates

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Abstract

The breakdown voltage highly depends on the electric field in the depletion area whose computation is the most time consuming part of the simulation. We present a self adaptive Finite Element Method which reduces dramatically the required computation time compared to usual Finite Difference Methods. A numerical example illustrates the efficiency and reliability of the algorithm.

It is a well accepted criterion for the breakdown of pn–junctions that the ionization integrals

\[ I_n = \int_0^x \alpha_n \exp\left(-\int_0^\xi (\alpha_n - \alpha_p) \, d\eta\right) \, d\xi \]
\[ I_p = \int_0^x \alpha_p \exp\left(-\int_0^\xi (\alpha_p - \alpha_n) \, d\eta\right) \, d\xi \]

for electrons and holes reach the value one

\[ I_n = 1 \text{ , } I_p = 1 \]  \hspace{1cm} (1)

As impact ionization is likely to happen in regions of large field strength, the condition (2) is tested only for field lines crossing the relative maxima of \(|E|\) (see [1] for details).

As the ionization rates \(\alpha_n, \alpha_p\) depend on \(E\) according to the law of Chynoweth [2]

\[ \alpha_i = \alpha_i \exp\left(-\left(b_i / |E|\right)^m_i\right), \text{ } i = n, p \]

with parameters \(\alpha_i, b_i, m_i, i = n, p\), given in [3], the crucial part of the evaluation of \(I_n, I_p\) is the computation of the electric field. In the case of strongly reverse biased pn–junctions which are considered here the full depletion assumption

\[ n = qD\delta_{\psi_0}, \text{ } p = -qD\delta_{\psi_0} \]

is justified. Here \(\delta_{x,y}\) is defined by \(\delta_{x,y} = 1\) if \(x = y\) and \(\delta_{x,y} = 0\) if \(x \neq y\) (Kronecker symbol), \(\psi\) denotes the potential distribution, \(D\) the doping concentration and \(-\psi_0, 0\) are the voltages applied to the anode and cathode respectively. Substituting (4) in the well–known drift–diffusion equations [4] we are left with the single nonlinear equation [5]

\[ -\text{div}(\varepsilon \nabla \psi) = qD(1 - \delta_{\psi_0} - \delta_{\psi_0}) \]

on the computational domain \(\Omega\) together with transition and boundary conditions. It is well–known (see Hunt/Nassif [6]) that a weak solution of (5) is obtained as the minimum of the energy functional

\[ F(\varphi) = \frac{1}{2}a(\varphi, \varphi) - l(\varphi) \]  \hspace{1cm} (6)
with $a(\cdot, \cdot)$ and $l(\cdot)$ given by
\[
 a(\varphi_1, \varphi_2) = \int_{\Omega} \nabla \varphi_1 \cdot \nabla \varphi_2 d(x, y),
\]
\[
 l(\varphi) = q \int_{\Omega} D\varphi d(x, y),
\]  
(7)
on the convex set $K$,
\[
 K = \{ \varphi \in H | \psi_0 \leq \varphi \leq 0 \text{ almost everywhere on } \Omega \}
\]  
(8)where $H$ consists of all functions $\varphi$ with finite energy
\[
 ||\varphi|| := a(\varphi, \varphi)^{\frac{1}{2}} < \infty,
\]  
(9)satisfying the Dirichlet boundary conditions. Hence we mainly concentrate on the self adaptive Finite Element approximation of the minimization problem:
\[
 \text{Find } \psi \in K \text{ such that } F(\psi) \leq F(\varphi), \ \varphi \in K.
\]  
(10)Once an approximate solution $\Psi$ is computed the ionization integrals $I_n, I_p$ are evaluated as described in Section 5.

1 Self Adaptive Finite Element Discretization

Let $T$ denote a triangulation of $\Omega \subset \mathbb{R}^2$ with vertices $\mathcal{V}$ and edges $\mathcal{E}$ respectively. Using the standard basis functions $\lambda_p, p \in \mathcal{V}$ with the property $\lambda_p(x) = \delta_{p,q}$ the desired Finite Element approximation $\Psi$ is written as
\[
 \Psi = \Psi^C + \sum_{p \in \mathcal{P}^0} \Psi_p \lambda_p
\]  
(11)with $\Psi^C$ representing the Dirichlet boundary conditions and $\mathcal{P}^0 \subset \mathcal{P}$ denoting the remaining vertices on which the value $\Psi_p$ of $\Psi$ is unknown. Substituting (11) in (10) we obtain the Finite Element discretization
\[
 \text{Find the vector } \Psi = (\Psi_p)_{p \in \mathcal{P}^0} \in K^L \text{ with } K^L = \{ \varphi = (\varphi_p)_{p \in \mathcal{P}^0} | -\psi_0 \leq \varphi_p \leq 0 \}
\]
\[
 \text{such that } \Psi^T(A\Psi - b) = \min_{\Psi^C}.
\]  
(12)The stiffness matrix $A$ and the right-hand side $b$ have the entries
\[
 a_{pq} = \frac{1}{2} a(\lambda_q, \lambda_p), \quad b_p = l(\lambda_p) - a(\Psi^C, \lambda_p), \quad p, q \in \mathcal{P}^0.
\]  
(13)The important question how to chose the triangulation $T$ is left open in classical Finite Element Methods. We propose a self adaptive method which is roughly described as follows:

**Algorithm 1. Self Adaptive Finite Element Algorithm**

**Step 0:** Start with an intentionally coarse triangulation $T_0$ on level 0.

**Step 1:** Compute an iterate $\Psi_k$ sufficiently close to the Finite Element approximation $\Psi_k$ with respect to the triangulation $T_k$ on level $k \geq 0$.

**Step 2:** Compute a local estimate of the total error $||\Psi_k - \psi||$.

**Step 3:** If $\Psi_k$ is accurate enough, then stop. Else refine all edges on which the local errors are too large resulting in a finer triangulation $T_{k+1}$ and go to Step 1.

This approach is widely used in self adaptive methods [7, 8]. For Step 0 we refer to [9]. The remaining steps are described in the subsequent sections.
2 Iterative Solution

A convergent sequence of iterate $\Psi_k^\nu$, $\nu = 0, \ldots$, can be computed by well-known relaxation methods as described for instance in [10]. See also [1, 11, 12]. Assuming that the discretization error behaves like

$$\|\Psi_k - \psi\| \leq cN_k^{-\frac{1}{2}}, \quad k = 0, 1, \ldots$$

(14)

with $N_k = |\mathcal{P}_k|$ denoting the number of unknowns, the iteration should be stopped if

$$\|\Psi_k^\nu - \psi\| \leq \tilde{c}N_k^{-\frac{1}{2}}, \quad \tilde{c} \approx c.$$  

(15)

It can be shown along the lines of [8] that (15) holds with $\tilde{c} \leq 3c$ provided that

$$\|\Psi_0^\nu - \Psi_0\| \leq \frac{1}{2}\|\Psi_0^\nu - \psi\|, \quad k = 0,$$

$$\|\Psi_k^\nu - \Psi_k\| \leq \frac{1}{2}\tilde{c}_{k-1}\|\Psi_{k-1} - \psi\|, \quad k > 0,$$

(16), (17)

with $\tilde{c}_{k-1} = (N_{k-1}/N_k)^{1/2}$, $k > 0$. Hence (16) and (17) are used as stopping criterion for the iterative solution. We write $\Psi_k^\nu := \Psi_k$ as soon as (16), (17) are fulfilled. Note that for $k > 0$ the total error appearing on the right-hand side of (17) is available from Step 2 of Algorithm 1 on level $k - 1$. To estimate the iteration error $\delta_k = \|\Psi_k - \Psi_k^\nu\|$ we use the well-known error estimate for fixed point iterations. Here the contraction number is replaced by the quotient of subsequent iterates. The iterative solution can be accelerated by multi level methods as proposed in [13]. These concepts will enter future revisions of the program.

3 Local Error Estimation

To obtain an estimate $\varepsilon_k$ of the total error $\varepsilon_k = \|\Psi_k - \psi\|$ we proceed in two main steps.

- Replace the exact solution $\psi$ by the piecewise quadratic approximation $\Phi_k$.
- Localize the computation of $\Phi_k$ by suitable simplifications.

Along the lines of Section 1 we use the representation

$$\Phi_k = \Psi_k^C + \Psi_k^L + \Psi_k^Q$$

(18)

with $\Psi_k^C$ taken from (11) and

$$\Psi_k^L = \sum_{p \in \mathcal{P}_k^L} \Psi_p^L \lambda_p, \quad \Psi_k^Q = \sum_{e \in \mathcal{E}_k^Q} \Psi_e^Q \mu_e$$

(19)

where $\mathcal{E}_k^Q \subset \mathcal{E}_k$ contains all edges which are not part of the Dirichlet boundary and $\mu_e$ is piecewise quadratic satisfying $\mu_e(q) = \delta_{pq}$ for $p = \text{midpoint of } e$ and $q \in \mathcal{P}_k^p \cup \{\text{midpoints of edges in } \mathcal{E}_k^0\}$. Substituting (18) in (10) we obtain the following discretization of higher order.

Find $\Psi_k^L = (\Psi_p^L)_{p \in \mathcal{P}_k^L}$ and $\Psi_k^Q = (\Psi_e^Q)_{e \in \mathcal{E}_k^Q}$ with $(\Psi_k^L, \Psi_k^Q) \in K_{k^L, k^Q}$ and

$$K_{k^L, k^Q} = \{(\varphi^L, \varphi^Q) | -\psi_0 \leq \varphi_p^L \leq 0, \; p \in \mathcal{P}_k^L, \; -\psi_0 \leq \varphi_e^Q + (\varphi_{p_1}^Q + \varphi_{p_2})/2 \leq 0, \; e = (p_1, p_2) \in \mathcal{E}_k^Q\}$$

(20)
such that \( \left( \frac{\psi^L}{\psi^Q} \right)^T \left( \begin{array}{cc} A_{LL} & A_{LQ} \\ A_{QL} & A_{QQ} \end{array} \right) \left( \begin{array}{c} \psi^L \\ \psi^Q \end{array} \right) - \left( \begin{array}{c} b^L \\ b^Q \end{array} \right) \) = \min_{(\varphi^L, \varphi^Q) \in K^{L,Q}}.

Note that the splitting (18) results in a corresponding block structure of the stiffness matrix. Of course (20) has to be simplified for complexity reasons. This is done in two steps.

- Replace the quadratic part \( A_{QQ} \) by its diagonal \( D_{QQ} \).
- Apply one step of a Block-Jacobi Method to the start iterate \((\tilde{\psi}_k, 0)\).

In this way our initial problem (12) is recovered for the approximate linear part \( \tilde{\psi}_k^L \). Note that (12) is already solved up to an approximate iteration error. For the quadratic part we obtain the problem:

\[
\text{Find } \tilde{\psi}_k^Q = (\tilde{\psi}_e^Q)_{e \in \mathcal{E}_k^0} \text{ with } (\tilde{\psi}_k^Q, \tilde{\psi}_k^Q) \in K^{L,Q}
\text{ such that } (\tilde{\psi}_k^Q)^T \left[ D_{QQ} \tilde{\psi}_k^Q - (b^Q - 2A_{QL} \tilde{\psi}_k^L) \right] = \min_{(\tilde{\psi}_k^Q, \tilde{\psi}_k^Q) \in K^{L,Q}}.
\]

Obviously (21) reduces to a one dimensional obstacle problem for each edge \( e \in \mathcal{E}_k^0 \) which is easily solved exactly. Using the energy norm \( |\varphi|_{\mathcal{M}} = (\varphi^T, M \varphi)^{1/2} \) we obtain the total error estimate

\[
(\varepsilon_k) = \left( |\tilde{\psi}_k - \tilde{\psi}_k|^2_{A_{LL}} + |\tilde{\psi}_k^Q|^2_{D_{LL}} \right)^{1/2}.
\]

4 Refinement Techniques

The refinement is based on the local errors

\[
\eta(e) = a_{ee} (\tilde{\psi}_e^Q)^2, \quad e \in \mathcal{E}_k^0,
\]

with \( a_{ee} \) denoting the appropriate diagonal element of \( A_{QQ} \) and \( \tilde{\psi}_k^Q = (\tilde{\psi}_e^Q)_{e \in \mathcal{E}_k^0} \) computed above.

Now the subset \( \mathcal{E}_k' \subset \mathcal{E}_k \) which is marked for refinement consists of all edges \( e \) for which \( \eta(e) \) is large, i.e.

\[
\mathcal{E}_k' = \{ e \in \mathcal{E}_k^0 | \eta(e) \geq \eta_k \}
\]

with some suitable threshold value \( \eta_k \). For \( k > 0 \) a guess of the maximal error occurring on the next level in case of uniform refinement is used as threshold value. This guess is made by extrapolation as proposed in [14]. See [10] for details.

Let \( T^i_k \subset T_k \) denote the set of all triangles with at least one edge in \( \mathcal{E}_k' \). Then each \( t \in T^i_k \) is refined using either red or blue refinement as shown in Figures 1 and 2. Note that blue refinement is used to improve the angles of flat triangles which may be used for the efficient resolution of horizontal layer structures typically arising in semiconductor devices. We refer to [10] and [9] for details. After possible refinement of further triangles which may be necessary for structural reasons the process is stopped by green closures as shown in Figure 3.

5 Ionization Integrals

Once an approximation \( \tilde{E}_k = -\nabla \tilde{\psi}_k \) of the electric field is computed on level \( k \), the evaluation of the ionization integrals \( I_n, I_p \) defined in (1) is performed as follows:
Algorithm 2. Ionization Integrals

Step 1: Find the set $\mathcal{T}_k^{\text{max}} \subset \mathcal{T}_k$ of triangles on which $|\hat{E}_k|$ has a relative maximum.

Step 2: For each $t \in \mathcal{T}_k^{\text{max}}$ collect the field line $\gamma_t$ intersecting $t$.

Step 3: For each $\gamma_t, t \in \mathcal{T}_k^{\text{max}}$ evaluate the corresponding integrals $\tilde{I}_n(\gamma_t), \tilde{I}_p(\gamma_t)$. Select the maximal ionization integrals $\tilde{I}_p, \tilde{I}_n$ for all $t \in \mathcal{T}_k^{\text{max}}$.

Recall that $\hat{E}_k$ is constant on each $t \in \mathcal{T}_k$. Hence the evaluation of the ionization integrals with $E$ replaced by $\hat{E}_k$ can be performed exactly. Finally $\tilde{I}_n, \tilde{I}_p$ are used to check the criterion (2) for the breakdown of the considered device.

6 Numerical Results

The method is applied to a planar pn–junction with multistep field plate shown in Figure 4. Note that the representation is not in scale. We refer to [10] for the physical data.

Figure 6 shows an initial triangulation $\mathcal{D}_0$ generated semi–automatically by the program BOXES [9] with the equipotential lines of the corresponding approximation $\Psi_0$ depicted in Figure 7. The Figures 8 and 9 show the final triangulation $\mathcal{D}_6$ generated automatically from $\mathcal{D}_0$ and the corresponding approximation $\Psi_6$ for the applied voltage $-\psi_0 = 800V$. In Figure 9 we have also depicted the boundary of the depletion area $\Omega_D$ together with the field line $\gamma$ along which the maximal ionization integral $\tilde{I}_p$ is obtained. The dependence of $\tilde{I}_p$ on the applied voltage $-\psi_0$ is shown in Figure 5.
Figure 4: Geometry of the computational domain

Figure 5: Dependence of the ionisation integral $I_p$ on the applied voltage $-\psi_0$
Figure 6: Initial triangulation generated by BOXES (371 nodes)

Figure 7: Level curves of initial solution $\psi_0$ corresponding to $\mathcal{T}_0$
Figure 8: Final triangulation $T_6$ obtained by self-adaptive refinement (2042 nodes)

Figure 9: Level curves of final solution $\Psi_6$
References


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