Stochastic Model of Breakdown Nucleation under Intense Electric Fields

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A plastic response due to dislocation activity under intense electric fields is proposed as a source of breakdown. A model is formulated based on stochastic multiplication and arrest under the stress generated by the field. A critical transition in the dislocation population is suggested as the cause of protrusion formation leading to subsequent arcing. The model is studied using Monte Carlo simulations and theoretical analysis, yielding a simplified dependence of the breakdown rates on the electric field. These agree with experimental observations of field and temperature breakdown dependencies.

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Various modern applications rely on maintaining high electric fields in a vacuum between metallic electrodes [1–4]. In such systems, arcing of current through the vacuum, which leads to a field breakdown (BD), is a major failure mechanism. Even when plasma formation is required [5], arcing nucleation and the mechanism leading up to it play a critical role in system design. Therefore, arcing nucleation and vacuum BD are subjects of interest in application and theory.

In most cases, plasma, created by particles emitted from the cathode, leads to arcing [2,6,7]. While the properties of arcing in relation to the interelectrode environment, as well as the development of the cathode and anode surface states, have been previously studied and remain an active and important area of research, the processes in the cathode leading up to BD nucleation are not yet understood [2,4]. It has been postulated that plastic damage to the cathode surface plays a critical role in the nucleation process. Based on this assumption, it was suggested that a BD can be related to the mobility of defects within the solid [8], and thus the mean time to BD τ would show an exponential dependence on the stress σ . Assuming that the leading contribution to σ is due to Maxwell stress $\sigma = \epsilon_0 E^2/2$, where E is the applied field, led to a $\ln \tau \sim E^2$ relation. This showed a good fit to a compilation of experimental data on the BD rate (BDR ~ $1/\tau$) versus E in metals at room temperature [10].

Understanding the observed limit is of general interest and is important for the design of high gradient applications, specifically in the proposed new CLIC project in CERN [1,4,11]. This led to a concentrated effort to identify the mechanism by which BDs are driven [12,13]. The basic assumption was that an applied field may cause a yield at the surface, which would lead to the formation of a localized protrusion. This protrusion would then enhance the electric current on the surface, leading to heating and thus to plasma formation and arc nucleation. However, molecular dynamics and finite element simulations showed these processes occurring only at $E \gtrsim 1$ GV/m [12,13], significantly more than the observed BD fields, in the range of 100–200 MV/m [4,11].

It is well established that plasticity in metals close to the yield is controlled by stochastic dislocation reactions [14]. Individual crystals, too, deform via a sequence of discrete slip events, as was demonstrated in the compression of micropillars [15]. The probability distribution of these events was measured [16] and reproduced by discrete dislocation dynamics simulations [17], as well as meanfield models [18]. Such systems demonstrate universal critical behavior characteristics of a self-organized critical state controlled by a minimally stable cluster, where in this case the cluster is a pinned dislocation arrangement [19,20]. A model reproducing this type of critical behavior utilizes terms describing the kinetics of the mobile dislocation density ρ , with nucleation at stress concentration sites on free surfaces, $\dot{\rho}_{+}$, as well as their depletion, $\dot{\rho}_{-}$ [21]. While not fully descriptive of the complex dislocation system, this model successfully describes the nature and size dependence of the observed stress-strain curves [22,23].

In a similar fashion, it has been shown, using a stochastic model, that the correlated motion of dislocations can lead to micron-sized surface protrusions, when persistent slip bands, caused by cyclic stress, break through the surface [24,25]. This was observed using SEM in fatigued samples exposed to high-cyclic stresses [26,27]. Samples exposed to strong electric fields, however, do not show such prominent features [4].

Here we explore the possibility that the mechanism leading to arc nucleation is a critical transition in the mobile dislocation population density close to the surface. We propose that, in a cathode subjected to an external electric field, the dislocation density typically fluctuates around a stable level, which depends on *E*. However, at any point in time, there is a finite probability that the density will reach a

critical point, beyond which it will increase deterministically. Arc nucleation will then follow from the surface response to the sudden localized increase in the surface dislocation density, through a mechanism which we do not attempt to address at this stage.

To explore this option, we employ a zero-dimensional mean-field model to describe the kinetics of the creation and depletion of mobile dislocations in a single slip plane, neglecting interactions between slip planes and the spatial variation of the mobile dislocation density within one plane. In this model, mobile dislocations nucleate at existing sources, and their depletion is due to collisions with obstacles. We formulate the problem in terms of a birth-death master equation [28] for the mobile dislocation population. This formalism is used to calculate an explicit analytical expression for the BDR in one slip plane as a function of E, which agrees well with kinetic Monte Carlo (KMC) simulations. The resulting model is unique in that, for the first time, it treats a BD in metals as a critical transition, due to the stochastic evolution of dislocations under an external field. In contrast with linear evolution models [10,12,13], our model predicts an eventual BD without requiring observable pre-BD surface features, whose absence in the microscopy of post-BD samples has posed a long-standing problem [4].

We calibrate the model for oxygen-free high-conductivity Cu, due to the availability of experimental data, as it is used in the CERN CLIC Collaboration, developing the next-generation linear collider accelerators. The physical parameters that are unknown are found by fitting the results generated by the model to experimental observations, including the temperature and E dependence of the BDR. Following this calibration, the model yields a quantitative agreement with the observed experimental BDRs, without making additional assumptions about the physical characteristics of the system, such as postulating the existence of specific surface or subsurface features.

Deterministic description.—A simplified kinetic model is based on the average creation and depletion rates $\dot{\rho}_+$ and $\dot{\rho}_-$, respectively [21]. For these, we assume that the kinetics are described within slip planes limited by dislocation cells on the order of 10 μ m. Dislocations nucleate at sources, whose density depends on the number of mobile dislocations, at a rate depending on *E*. They are depleted by interactions with other mobile dislocations and existing defects. Thus, the deterministic dynamics of the mobile dislocation density are given by

$$\dot{\rho} = \dot{\rho}_{+} - \dot{\rho}_{-}, \quad \dot{\rho}_{+} = B_{1}(\rho + c)\sigma^{2}e^{\alpha\sigma}; \quad \dot{\rho}_{-} = b_{2}\sigma\rho(\rho + c),$$
(1)

with $\sigma = A_1 + a_2\rho$. The constants A_1 , a_2 , B_1 , b_2 , c, and α depend on the system parameters and are independent of E, except for $A_1 \sim E^2$. The derivation of these functional forms and the relation of the effective constants to the appropriate physical parameters is described in the



FIG. 1. Fixed points of the dynamical equations for ρ , attracting (ρ_*) and repelling (ρ_c) , as functions of *E*, demonstrating the existence of a bifurcation point.

Supplemental Material; see also Fig. S1 [29]. Henceforth, all analytical and numerical results are presented for fitted parameters (see below).

Below a critical field E_c (see below), the equation $\dot{\rho} = 0$ yields two solutions: ρ_* and ρ_c , where $\rho_* < \rho_c$. For $\rho_* < \rho < \rho_c$, we have $\dot{\rho}_- > \dot{\rho}_+$, and the dislocation density deterministically decreases back to ρ_* . Therefore, ρ_* is an attracting fixed point of Eq. (1), while ρ_c is a repelling point. That is, if ρ is larger than ρ_c , it will increase indefinitely, leading to a subsequent BD.

Sufficiently below E_c , where $\rho_* \ll A_1/a_2 \ll \rho_c$, we can find analytical expressions for ρ_* and ρ_c , which yield $\rho_* = (B_1A_1/b_2)e^{\alpha A_1}$ and $\rho_c = (\alpha a_2)^{-1}\ln[b_2/(B_1a_2)]$.

The values of $\dot{\rho}_+$ and $\dot{\rho}_-$ approach each other as E increases, as shown in Fig. 1. At $E = E_c$, $\rho_* = \rho_c$, and thus for $E \ge E_c$ the system progresses deterministically to a BD. Notably, for $E < E_c$, reaching ρ_c is a fluctuation-driven stochastic event, leading to an *E*-dependent BDR.

Stochastic model.—To incorporate fluctuations, we model the evolution of the mobile dislocation density ρ as a birth-death Markov process [28]. The rates $\dot{\rho}_+$ and $\dot{\rho}_$ represent the probability per unit time that ρ will increase or decrease, respectively, by $\Delta \rho = 0.1 \ \mu m^{-1}$, corresponding to one dislocation per cell. For one single slip plane, we define $n = \rho / \Delta \rho$ as the instantaneous number of mobile dislocations per cell. By defining $A_2 = a_2 n_c \Delta \rho$, $B_2 = b_2 n_c \Delta \rho$, and $C = c / (n_c \Delta \rho)$, where $n_c = \rho_c / \Delta \rho$, we find that the microscopic birth and death rates as a function of *n* are, respectively,

$$\lambda_n = B_1(n + n_c C)\sigma^2 e^{\alpha\sigma}; \quad \mu_n = (B_2 n / n_c)(n + n_c C)\sigma, \quad (2)$$

with $\sigma(n) = A_1 + A_2 n/n_c$ [40]. The stochastic dynamics are governed by the master equation

$$\frac{\partial P_n(t)}{\partial t} = \lambda_{n-1} P_{n-1}(t) + \mu_{n+1} P_{n+1}(t) - (\lambda_n + \mu_n) P_n(t),$$
(3)

describing the evolution of the probability $P_n(t)$ of finding *n* mobile dislocations per cell at time *t* [28].

In order to find the BDR, we write a recursive equation for T_n , the mean time it takes to reach BD starting from *n* mobile dislocations [28,42]:

$$T_n = \frac{\lambda_n}{\lambda_n + \mu_n} T_{n+1} + \frac{\mu_n}{\lambda_n + \mu_n} T_{n-1} + \frac{1}{\lambda_n + \mu_n}, \quad (4)$$

where $(\lambda_n + \mu_n)^{-1}$ is the average time it takes to jump from n to $n \pm 1$. We solve this equation with an absorbing boundary at n_c and a reflecting wall at n = 0, such that $T_{n_c} = 0$ and $T'_n(n = 0) = 0$. Starting from the vicinity of $n = 0, \tau$ is given by [43]

$$\tau = T_n = \sum_{i=n}^{n_c} \phi_i \left(\sum_{j=0}^i \frac{1}{\lambda_j \phi_j} \right), \tag{5}$$

with $\phi_n = \prod_{m=1}^n \mu_m / \lambda_m$ [28].

In the rest of the Letter, we focus on the regime where the critical number of mobile dislocations, needed for a BD, satisfies $n_c \gg 1$. In this regime, the expression for τ can be simplified. Assuming *a priori* that τ is exponentially large in n_c , we can employ the metastability ansatz $P_n(t) \simeq \pi_n e^{-t/\tau}$ [44], where π_n is the *quasistationary distribution* (QSD) [41,45–47]. Substituting this into Eq. (3) and neglecting the exponentially small term proportional to $1/\tau$, we have $\lambda_{n-1}\pi_{n-1} + \mu_{n+1}\pi_{n+1} - (\lambda_n + \mu_n)\pi_n = 0$, whose solution, for $n \leq n_c$ [48], satisfies [28]

$$\pi_n = \pi_0 \prod_{m=1}^n \frac{\lambda_{m-1}}{\mu_m}.$$
 (6)

This solution is shown in Fig. 2(a), where π_0 is found via normalization [28]. Using Eq. (3) for $n = n_c + 1$ and the metastability ansatz, we thus have $\tau \simeq (\lambda_{n_c} \pi_{n_c})^{-1}$ [41, 45–47]. Using Eqs. (2) and (6), expanding in $n_c \gg 1$, and applying the Stirling approximation up to subleading order, τ gives way to a WKB-like solution [41]

$$\tau = \mathcal{A}e^{n_c\Delta S},\tag{7}$$

where

$$\Delta S = \ln \frac{B_2}{A_1 B_1} - \alpha A_1 \left(1 + \frac{1}{2\eta} \right) - (\eta + 1) \ln \left(1 + \frac{1}{\eta} \right),$$
$$\mathcal{A} = \sqrt{\frac{2\pi}{n_c}} \frac{e^{-\alpha A_1 [1 + (1/2\eta)]}}{A_1^2 B_1 C} \left(1 + \frac{1}{\eta} \right)^{-1/2},$$
(8)

and $\eta = A_1/A_2$. Here $n_c \Delta S$ can be viewed as a barrier to a BD [49]. Note that, in the experimentally relevant electric-field range, we observe that



FIG. 2. Analytical versus simulated values: (a) probability of being in state n, at $t \ll \tau$, as calculated from the analytical expression [Eq. (6)] and numerical simulation. The lines, from bottom to top, are for E = 200, 230, and 260 MV/m. The simulation points include measurements of the probability for $n > n_c$, above the QSD validity regime. (b) τ normalized by $\tau(E_0 = 180 \text{ MV/m})$ as a function of E, calculated using the metastable approximation [Eq. (7), solid line], the exact formula [Eq. (5), dashed line], and the simulation (triangles).

$$\tau \sim \exp[\gamma (1 - E/E_0)] \tag{9}$$

with E_0 a reference field and γ a dimensionless constant independent of *E*, as demonstrated in Fig. 2(b) [50].

The analytical results were compared to a KMC simulation, tracking the time evolution of *n*. In this simulation, $n \rightarrow n \pm 1$ changes randomly using the transition rates in Eq. (2), with the time elapsed between changes determined using an exponential distribution with mean $(\lambda_n + \mu_n)^{-1}$. The numerically estimated QSD and τ agree with the analytical solution; see Figs. 2 and S2 [29].

Parameter range.—The model includes six constants which depend on the material properties and on specific mechanisms that control the reactions of dislocations to the applied field [29]. These constants depend on four unknown parameters: (i) β , relating the stress at the nucleation sites to E; (ii) κ , representing temperatureindependent factors affecting nucleation, such as the nucleation attempt frequency and activation entropy; (iii) the activation energy E_a ; and (iv) the activation volume Ω for releasing new mobile dislocations.



FIG. 3. Experimental BDRs with fitted theoretical lines using Eq. (7): (a) BDR versus *E* for various Cu accelerating structures [11]. (b) BDR variation with *E* at room temperature (two lines on the left) and at 45 K (line on the right) [51]. (c) BDR versus *E* for various Cu accelerating structures [11,52], with *E* rescaled so that all measurements are fitted with $\beta = 4.8$.



FIG. 4. Fit of free model parameters: (a) LSQ parameter as a function of Ω and E_a , with $\beta = 4.8$ and $\kappa = 0.41$. (b) LSQ parameter as a function of β and κ , with $\Omega = 5.4$ eV/GPa and $E_a = 0.08$ eV.

The values of these parameters can be found and validated by a comparison to experimental measurements of the BDR as a function of *E* for different structures [11], shown in Fig. 3(a), and for one structure at both room temperature and 45 K [51], shown in Fig. 3(b). The results include measurements from various geometries, leading to a significant variation in the local field at the BD site [11]. As this translates to a variation in β only, results were scaled using β to those of a reference set [52], so that all sets produce identical BDRs at *E*= 180 MV/m. This led to a single normalized data set shown in Fig. 3(c). As normalization was done for each structure at 300 K, results at 45 K are presented for the rescaled *E*= 250 MV/m rather than the measured 300 MV/m.

The parameter evaluation was done in two steps: First, a fitting was performed for the reference set [52] and the data from the temperature-varied structure [51], using a least square (LSQ) fit demanding (i) consistency with the experimental values of $\tau(T = 300 \text{ K}, E = 180 \text{ MV/m})$, $\tau(T = 45 \text{ K}, E = 300 \text{ MV/m})$, and $\gamma(T = 300 \text{ K})$ of Eq. (9), (ii) consistency with theoretical estimates of $E_a \gtrsim 0.1$, and (iii) validity of the approximation $n_c \gg 1$ at the corresponding temperatures and fields as in (i) [53]. Next, the rest of the data sets [11] were used to validate the quality of the fit. Two cross sections of the resulting LSQ fit in the parameter space $(\Omega, E_a, \beta, \kappa)$ are plotted in Fig. 4. We find that the LSQ parameter has a minimum at $\beta = 4.8 \pm 0.1, \ \kappa = 0.41 \pm 0.02, \ \Omega = 5.4 \pm 0.2 \ \text{eV/GPa},$ and $E_a = 0.08 \pm 0.002$ eV, marked on the graphs in Fig. 4.

Our results for E_a are consistent with mobile dislocation nucleation from preexisting sources [54], significantly lower than the activation energy for nucleation in pristine crystalline structures [55–57]. Furthermore, the activation volume $\Omega = 55b^3$, with *b* the Burgers vector, is within the experimental range $10b^3 < \Omega < 124b^3$ [54,56].

Discussion.—The model can be consistently fitted to all available experimental data sets, with a single free parameter β adjusted to account for the geometrical difference between experimental structures. Thus, the model allows us to make predictions for BDRs over a wide range of physical

parameters beyond those of the current measurements, as demonstrated in Fig. 3.

According to the proposed model, BD nucleation is preceded by a critical increase in the number of mobile dislocations. This can create an early-warning signal for imminent BD through the monitoring of characteristic fluctuations [58], which includes indirect measurements such as thermionic current emissions or direct measurements of acoustic signals from increased fluctuations in the mobile dislocation populations [59]. Furthermore, it is expected that the standard deviation of the QSD, representing the typical fluctuations of the pre-BD mobile dislocation population, will increase significantly as the BD is approached [29].

Conversely, our model does not depend upon the appearance of an observable surface protrusion before a BD. This is in line with the fact that no observable sub-BD features have been observed in metallic electrodes exposed to a strong electric field [4]. BD sites are characterized by a large crater created by the arc [2], obliterating any remains of possible pre-BD features. Such features, however, should have been found further away from the BD site if they existed.

An understanding of the mechanism which leads to BD nucleation can facilitate the development of a better design of electrodes, focusing on limiting the nucleation and mobility of dislocations, in order to lower the BDR. It is well established that, in order to stabilize a significant field, an electrode has to undergo *conditioning* via a series of field exposures and BDs at lower fields [60]. Conditioning includes both an initial extrinsic process (resulting from the removal of contaminates) and a long-term intrinsic process resulting from modifications of the electrode. In line with our model, this comes about as a result of surface hardening [9]. In addition, by controlling the dislocation mobility, our model offers a direction for improving the electrode performance.

In conclusion, we presented a model describing BD nucleation as a stochastic process driven by the creation and depletion of dislocations within the electrode. BD nucleation in this case is a result of a critical transition in the mobile dislocation population density. The model was formulated using a set of parameters describing known material properties and unknown parameters describing interactions specific to the response of the dislocation population to the applied field. Measurements in various fields and temperatures were used to fit the parameters and validate the model. This model is unique, as it does not require pre-BD features, and offers a simple intrinsic mechanism for a BD at fields lower than the deterministic limit. Establishing such a model may provide opportunities for improving the design of future electrodes, aiming to limit the dislocation mobility, as well as offer ways to identify pre-BD early-warning signals through the evolution of the dislocation population.

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