Indeterminacy of spatiotemporal cardiac alternans

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Cardiac alternans, a beat-to-beat alternation in action potential duration (at the cellular level) or in electrocardiogram morphology (at the whole heart level), is a marker of ventricular fibrillation, a fatal heart rhythm that kills hundreds of thousands of people in the United States each year. Investigating cardiac alternans may lead to a better understanding of the mechanisms of cardiac arrhythmias and eventually better algorithms for the prediction and prevention of such dreadful diseases. In paced cardiac tissue, alternans develops under increasingly shorter pacing period. Existing experimental and theoretical studies adopt the assumption that alternans in homogeneous cardiac tissue is exclusively determined by the pacing period. In contrast, we find that, when calcium-driven alternans develops in cardiac fibers, it may take different spatiotemporal patterns depending on the pacing history. Because there coexist multiple alternans solutions for a given pacing period, the alternans pattern on a fiber becomes unpredictable. Using numerical simulation and theoretical analysis, we show that the coexistence of multiple alternans patterns is induced by the interaction between electrotonic coupling and an instability in calcium cycling.

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I. INTRODUCTION

Sudden cardiac death, attributable to unexpected ventricular arrhythmias, is one of the leading causes of death in the United States and kills over 300 000 Americans each year [1]. The induction and maintenance of ventricular arrhythmias has been linked to single-cell dynamics [2,3]. In response to an electrical stimulus, cardiac cells fire an action potential [4], which consists of a rapid depolarization of the transmembrane voltage (V_m) followed by a much slower repolarization process before returning to the resting value (Fig. 1). The time interval during which the voltage is elevated is called the action potential duration (APD). The time between the end of an APD and the beginning of the next one is called the diastolic interval (DI). The time interval between two consecutive stimuli is called the basic cycle length (BCL). When the pacing rate is slow, a periodic train of electrical stimuli produces a phase-locked steady-state response, where each stimulus gives rise to an identical action potential (1:1 pattern). When the pacing rate becomes sufficiently fast, the 1:1 pattern may be replaced by a 2:2 pattern, so-called electrical alternans [5.6], where the APD alternates between short and long values. Recent experiments have established a causal link between alternans and the risk for ventricular arrhythmias [7–9]. Therefore, understanding of the mechanism of alternans is a crucial step in detection and prevention of fatal arrhythmias.

Cellular mechanisms of alternans have been much studied. Summaries on this topic can be found in recent review articles by Shiferaw *et al.* [10] and Weiss *et al.* [11]. At the cellular level, cardiac dynamics involves bidirectional coupling between membrane voltage (V_m) dynamics and intracellular calcium (Ca_{in}) cycling. During an action potential, the elevation of V_m activates L-type Ca currents I_{Ca} to invoke the elevation of [Ca_{in}], which in turn triggers Ca release from the

sarcoplasmic reticulum (SR), a procedure known as calciuminduced calcium release [12]. The $V_m \rightarrow Ca_{in}$ coupling satisfies graded release, where a larger DI leads to an increase in the Ca release at the following beat since it allows more time for L-type Ca channels to recover. On the other hand, Ca release from the SR affects the APD in two ways: it curtails the APD by enhancing the inactivation of L-type Ca currents I_{Ca} ; and it prolongs the APD by intensifying Na⁺-Ca²⁺ exchange currents I_{NaCa} . Therefore, depending on the relative contributions of I_{Ca} and I_{NaCa} , an increase in Ca release may either shorten (negative $Ca_{in} \rightarrow V_m$ coupling) or lengthen the APD (positive $Ca_{in} \rightarrow V_m$ coupling) [13–15]. There exist two main cellular mechanisms of alternans. First, alternans may be attributed to steep APD restitution [5], which is due to a period-doubling instability in the V_m dynamics. In this case, Cain transient alternans, as a slave variable, is induced because V_m regulates [Ca_{in}] via the L-type Ca currents and the sodium-calcium exchange currents. Second, alternans may be caused by a period-doubling instability in Cain cycling, which is associated with a steep relationship between the SR release and SR load [16,17]. In this case, APD alternans is a secondary effect via $Ca_{in} \rightarrow V_m$ coupling. For ease of reference, we call the first mechanism APD-driven alternans and the second *Ca-driven alternans*. Interestingly, APD and Cain transient alternans can be electromechanically (EM) in phase

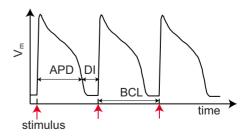


FIG. 1. (Color online) Schematic action potential showing the response of the transmembrane voltage to periodic electrical stimuli.

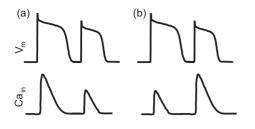


FIG. 2. Schematic illustration of electromechanically in-phase (a) and out-of-phase (b) alternans.

or out of phase [18,19]. In EM in-phase alternans, a longshort-long APD pattern corresponds to a large-small-large [Ca_{in}] pattern; see Fig. 2(a). In contrast, in EM out-of-phase alternans, a long-short-long APD pattern corresponds to a small-large-small [Ca_{in}] pattern; see Fig. 2(b). When alternans happens in isolated cells, the bidirectional coupling between APD and Ca_{in} transient determines the relative phase of APD and Ca_{in} transient alternans. In particular, APDdriven alternans always leads to EM in-phase alternans whereas Ca-driven alternans is EM in phase for positive Ca_{in} $\rightarrow V_m$ coupling and out of phase for negative Ca_{in} $\rightarrow V_m$ coupling [13–15].

The mechanism of alternans in multicellular tissue is more complicated since it involves electrotonic coupling and conduction velocity restitution. Of particular interest is a phenomenon called spatially discordant alternans, in which different regions of the tissue alternate out of phase. Discordant alternans is arrhythmogenic because it forms a dynamically heterogeneous substrate that may promote wave break and reentry [2,3]. To study the spatiotemporal patterns of alternans, Echebarria and Karma derived amplitude equations that are based on APD-driven alternans [20,21]. These amplitude equations not only are capable of quantitative predictions but also provide insightful understanding of the arrhythmogenic patterns. In a recent paper, Dai and Schaeffer [22] analytically computed the linear spectrum of Echebarria and Karma's amplitude equations for the cases of small dispersion and long fibers.

Spatial patterns of alternans have been investigated in experiments [7,23-25]. Recently, Aistrup et al. [26] used single-photon laser-scanning confocal microscopy to measure Ca signaling in individual myocytes. They found that Ca alternans is spatially synchronized at low pacing rates whereas dyssynchronous patterns, where a number of cells are out of phase with adjoining cells, arise when the pacing rate increases. Aistrup et al. also observed subcellular alternans at fast pacing, where Ca alternans is spatially dyssynchronous within a cell. Using simulations of onedimensional (1D) homogeneous tissue, Sato et al. [15] found that, in cardiac fibers with negative $Ca \rightarrow V_m$ coupling, Ca alternans reverses phase over a length scale of one cell whereas, in fibers with positive $Ca \rightarrow V_m$ coupling, Ca alternans changes phase over a much larger length scale. They interpreted this difference by showing that negative Ca $\rightarrow V_m$ coupling tends to desynchronize two coupled cells while positive $Ca \rightarrow V_m$ coupling tends to synchronize the coupled cells.

Motivated by the aforementioned experimental and theoretical work, this paper aims to explore spatial patterns of cardiac alternans. Through extensive numerical simulations, we find that complex spatial patterns of Ca alternans with phase reversals in adjacent cells can happen in homogeneous fibers with both negative and positive $Ca \rightarrow V_m$ couplings. Most surprisingly, we find that the spatiotemporal pattern of cardiac alternans is not determined by the pacing period alone. Specifically, when calcium-driven alternans develops in multicellular tissue, there coexist multiple spatiotemporal patterns of alternans regardless of the length of the fiber, the junctional diffusion of Ca, and the type of $Ca \rightarrow V_m$ coupling. We further investigate the mechanism that leads to the coexistence of multiple alternans solutions. Our analysis shows that multiple alternans solutions are induced because of the interaction between electrotonic coupling and an instability in Ca_{in} cycling.

II. MODEL DESCRIPTION

A. Membrane dynamics

We adopt a model of membrane dynamics that combines the calcium dynamics model developed by Shiferaw et al. [27] and the canine ionic model by Fox et al. [28]. In the following, we will refer to this model as the Shiferaw-Fox model. Detailed formulations of the model can be found in [27,29]. The Shiferaw-Fox model has adopted two sets of parameters in the calcium dynamics to account for negative and positive $Ca_{in} \rightarrow V_m$ couplings. Besides the phase difference, the two sets of parameters also produce alternans at different values of BCL. Using Shiferaw's default parameters, we find that alternans happens at BCL \approx 401 ms for negative $Ca_{in} \rightarrow V_m$ coupling and BCL ≈ 323 ms for positive $Ca_{in} \rightarrow V_m$ coupling. Figure 3 shows the bifurcation diagrams in APD and peak value of $[Ca_{in}]$ for negative $Ca_{in} \rightarrow V_m$ coupling. The bifurcation diagrams for positive $Ca_{in} \rightarrow V_m$ coupling are similar and thus are not shown here. We note that, in simulations of isolated cells using the Shiferaw-Fox model, alternans solutions do not depend on the initial condition nor on the pacing history. However, as we will show in the following, fibers based on the Shiferaw-Fox model possess multiple alternans solutions, which are sensitive to the initial condition and the pacing protocol.

B. Simulation of fiber models

We study paced, homogeneous fibers, which can be modeled using the cable equation

$$\frac{\partial v}{\partial t} = D \frac{\partial^2 v}{\partial x^2} - \frac{1}{C_m} (I_{\text{ion}} + I_{\text{ext}}), \qquad (1)$$

where v represents V_m , $D=5 \times 10^{-4}$ cm²/ms represents the effective diffusion coefficient of v in the fiber, $C_m = 1 \ \mu F/cm^2$ represents the transmembrane capacitance, I_{ion} is the total ionic current, and I_{ext} represents the external current stimulus. The ionic current I_{ion} is computed using the Shiferaw-Fox model. The current stimulus has duration 1 ms and amplitude 80 $\mu A/\mu F$. This paper studies fibers of various lengths. For two coupled cells, we pace the left cell. For longer fibers, we pace the leftmost few cells to ensure propagation. For example, the leftmost five cells are paced in

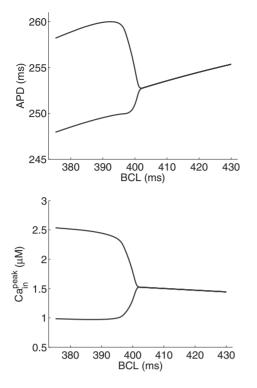


FIG. 3. Bifurcation diagrams of an isolated cell using the Shiferaw-Fox model.

simulating a fiber of 100 cells. The cable equation (1) is solved using the finite difference method with a space step of $\Delta x=0.015$ cm and time step of $\Delta t=0.1$ ms. No-flux boundary conditions are imposed at both ends of the fiber [27,29].

C. Pacing protocols

To study the onset and development of alternans, we pace both single cells and fibers of various lengths with several pacing protocols, which are briefly described below.

(i) In the *downsweep protocol* [30], the cell or fiber is paced periodically with period BCL until it reaches steady state. Then the pacing period is reduced by ΔB and the procedure is repeated many times. Note that this protocol is also known as the *dynamic pacing protocol* [31].

(ii) The *perturbed downsweep protocol*, proposed by Kalb *et al.* [30], can be regarded as a perturbation to the downsweep protocol. At each pacing period BCL, the cell or fiber is first paced N beats to reach steady state. Then, a longer pacing period is applied at the (N+1)st pacing, after which the original pacing period is applied for ten beats to allow the tissue to recover its previous steady state. Next, a shorter pacing period is applied and followed by ten beats of the original pacing period. Finally, the pacing period is reduced by ΔB and the procedure is repeated.

(iii) To explore the possibility for multiple alternans solutions, we set up certain initial conditions and pace the tissue with period BCL to reach steady state, a process we call *direct pacing*.

(iv) To explore the origin of an alternans pattern, we use the *upsweep protocol* [30], which is a reversed downsweep protocol.

III. SPATIOTEMPORAL PATTERNS OF ALTERNANS: NUMERICAL EXPLORATION

We simulate fibers using the Shiferaw-Fox model with both negative and positive $Ca_{in} \rightarrow V_m$ couplings under various conditions. Default parameters in Shiferaw's code [29] are used unless otherwise specified. Despite quantitatively significant differences, we find both types of coupling lead to the coexistence of multiple alternans solutions. For clarity, we start with the results for negative $Ca_{in} \rightarrow V_m$ coupling and defer the results for positive $Ca_{in} \rightarrow V_m$ coupling to a later section.

A. Coexistence of multiple solutions

We first consider a homogeneous fiber of 100 cells with negative $Ca_{in} \rightarrow V_m$ coupling. To our surprise, numerical simulations show that, when the fiber is in alternans, there coexist multiple solutions for a given pacing period. For example, Fig. 4 shows six selected solutions of alternans for the fiber paced at BCL=375 ms. Here, the steady-state solutions in Figs. 4(a)-4(c) are obtained using the downsweep protocol with step size $\Delta B=1$, 2, and 25 ms, respectively. The pacing protocols are started from BCL=500 ms in Figs. 4(a) and 4(c) and from BCL=499 ms in Fig. 4(b). We note that the solution of a downsweep protocol is not influenced by the initial condition at the starting, long BCL; instead, the solution is sensitive to the step size ΔB . The steady-state solutions in Figs. 4(d)-4(f) are obtained by pacing the fiber at BCL=375 ms with prescribed initial conditions for 200 beats, the so-called direct pacing. The initial condition of the fiber in Fig. 4(d) is uniform, i.e., all cells are assigned the same resting voltage, gating variables, and ionic concentrations. The initial condition in Fig. 4(e) is the same as that in Fig. 4(d) except that [Ca_{in}] is assigned to be 0.54 μM for the first 35 cells and 0.66 μM for the remaining cells. Interestingly, this initial condition leads to a steady-state [Ca_{in}] pattern, which, besides the phase reversal between cells 35 and 36, has another phase reversal between cells 11 and 12. The initial condition in Fig. 4(f) is the same as that in Fig. 4(d)except that [Cain] is randomly assigned for cells on the fiber according to a uniform distribution in the interval of 0.45 $-0.75 \ \mu M$. In all protocols, we pace the fiber for 200 beats at each BCL and plot the last ten beats at BCL=375 ms. The simulation results in Fig. 4 demonstrate that the alternans on a fiber is not solely determined by the pacing period. Instead, the solution is sensitive to the pacing protocol and the initial condition. It is worth noting that, besides the solutions shown in Fig. 4, there exist many other solution patterns. In particular, there exist many complex patterns similar to Fig. 4(f). In the following, we will verify whether the phenomenon is influenced by the length of the fiber, junctional Ca diffusion, or $\operatorname{Ca}_{\operatorname{in}} \to V_m$ coupling.

B. Influence of the length of the fiber

We repeat the numerical simulations for fibers of various lengths and find that the phenomenon persists regardless of the length of the fiber. Probably the most illuminating example is a "fiber" of two coupled cells. Denoting the voltage ິຂິ 260

250

(MM)

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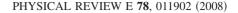
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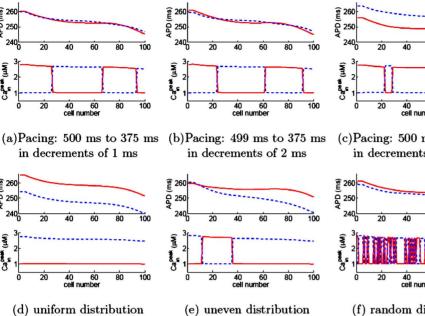
APD 25

(MM)

Cape

240





at cell 1 as V_1 and that at cell 2 as V_2 , we can simulate the two coupled cells by integrating the following equations [32]:

$$\frac{dV_1}{dt} = -\frac{1}{C_m}(I_{\rm ion} + I_{\rm ext}) + D\frac{(V_2 - V_1)}{\Delta x^2},$$
 (2)

$$\frac{dV_2}{dt} = -\frac{1}{C_m} I_{\rm ion} + D \frac{(V_1 - V_2)}{\Delta x^2}.$$
 (3)

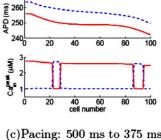
Note that only cell 1 is paced in this case. Using various pacing protocols and initial conditions, we find the two coupled cells can have spatially desynchronized [Fig. 5(a)] and synchronized [Fig. 5(b)] alternans. Note that, in the desynchronized pattern, APD in both cells exhibits a beat-tobeat variation of a few hundredths of milliseconds, which is impossible to observe in experiments.

C. Influence of junctional Ca diffusion

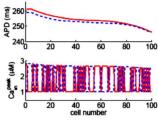
The original Shiferaw-Fox model does not include the diffusion of Ca between neighboring cells [33]. Physiologically, there exists gap junctional Ca diffusion although its magnitude is several orders lower than voltage diffusion [34,35] and thus it is typically neglected in cardiac modeling. One may wonder whether including junctional Ca diffusion will affect the simulation results. To answer this question, we introduce junctional Ca diffusion to the Shiferaw-Fox model by modifying the equation governing [Ca_{in}] as follows [32]:

$$\frac{\partial C_i}{\partial t} = D_c \frac{\partial}{\partial x} \left(\frac{\partial C_i}{\partial x} - \frac{Z_c F}{RT} C_i \frac{\partial v}{\partial x} \right) - I_c, \tag{4}$$

where C_i represents [Ca_{in}], $D_c = 3 \times 10^{-9}$ cm²/ms is the Ca diffusion coefficient [34,35], Z_c is the valence of Ca, F is the Faraday constant, R is the gas constant, T=300 K is the temperature, and I_c represents the Ca_{in} currents in the Shiferaw-Fox model. With this modification, we repeat the



in decrements of 25 ms



(f) random distribution

FIG. 4. (Color online) Six selected alternans solutions for a homogeneous fiber of 100 cells with negative $Ca \rightarrow V_m$ coupling when paced at BCL=375 ms. (a)-(c) are obtained using downsweep protocols from 500 to 375 ms in steps of 1, 2, and 25 ms, respectively. (d)-(f) are obtained by directly pacing at BCL=375 ms with different initial distributions in Cain; see text for details. To verify if a solution reaches steady state, we plot the last ten beats of the simulation results, where odd beats are represented by red solid lines and even beats by blue dashed lines. The coincidence of all odd beats and that of all even beats confirm that the solution is indeed in steady state.

numerical simulations and find there still exist multiple alternans solutions.

For example, Fig. 6 shows three selected solutions for a homogeneous fiber of 100 cells paced at BCL=375 ms. Figures 6(a) and 6(b) are obtained using downsweep protocols like those in Figs. 4(b) and 4(c), respectively. Figure 6(c) is

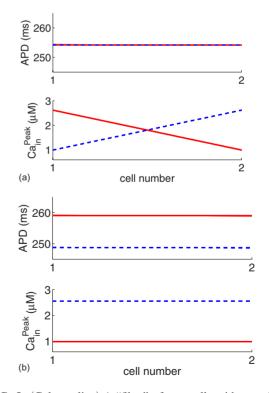


FIG. 5. (Color online) A "fiber" of two cells with negative Ca $\rightarrow V_m$ coupling can have both spatially (a) desynchronized and (b) synchronized alternans solutions. The pacing period is 375 ms. The last ten beats of the steady-state solutions are plotted, where odd beats are represented by red solid lines and even beats by blue dashed lines.

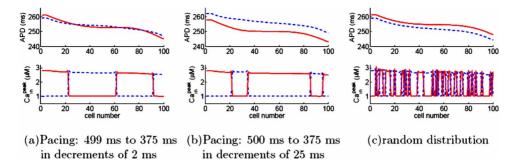


FIG. 6. (Color online) Three selected alternans solutions for a homogeneous fiber model, where junctional Ca diffusion is included. The cells have negative $Ca \rightarrow V_m$ coupling and the pacing period is 375 ms. (a) and (b) are obtained using downsweep protocols and (c) is obtained by direct pacing with random initial distribution of Ca_{in} . The last ten beats of the steady-state solutions are plotted, where odd beats are represented by red solid lines and even beats by blue dashed lines.

obtained using a random initial distribution in $[Ca_{in}]$ [cf. Fig. 4(f)]. Comparing results in Figs. 4 and 6 shows that the coexistence of multiple solutions is not influenced by junctional Ca diffusion.

D. Influence of $Ca_{in} \rightarrow V_m$ coupling

Simulations show that fibers with positive $Ca_{in} \rightarrow V_m$ coupling also possess multiple alternans solutions. For example, Fig. 7 shows three selected solutions for a homogeneous fiber of 100 cells with positive $Ca_{in} \rightarrow V_m$ coupling, paced at BCL=300 ms. Figure 7(a) is obtained via a perturbed down-sweep protocol, where the step size is $\Delta B=25$ ms and a long and a short perturbation of ± 20 ms are applied at each BCL; see Sec. II for details of the perturbed downsweep protocol. Figures 7(b) and 7(c) are obtained by directly pacing the cell at BCL=300 ms using a uniform and a random initial distribution in [Ca_{in}], respectively.

We simulate two coupled cells with positive $Ca_{in} \rightarrow V_m$ coupling and find that alternans can be both spatially synchronized and spatially desynchronized; see Fig. 8. In Figs. 5 and 8, the values of the APD are almost identical in the two cells because the conduction time across a cell's length is negligible. Thus, these examples indicate that the coexistence of multiple solutions does not depend on steep CV restitution. We note that, using a downsweep protocol with step size of 2 ms, Sato *et al.* [15] also studied alternans in two coupled cells. They observed the desynchronized solution for

negative $\operatorname{Ca}_{in} \to V_m$ coupling and the synchronized solution for positive $\operatorname{Ca}_{in} \to V_m$ coupling. Using numerical simulation, Sato *et al.* showed that negative $\operatorname{Ca}_{in} \to V_m$ coupling tends to desynchronize two coupled cells whereas positive $\operatorname{Ca}_{in} \to V_m$ coupling tends to synchronize them. Thus, we hypothesize that the synchronized solution for negative $\operatorname{Ca}_{in} \to V_m$ coupling and the desynchronized one for positive coupling are induced by electrotonic coupling, which will be verified in the next section.

IV. ORIGIN OF THE INDETERMINACY

A. Numerically tracing an alternans solution

The coexistence of multiple alternans solutions in a fiber is surprising because the underlying cell's model possesses a supercritical period-doubling bifurcation, describing a transition from a unique 1:1 solution to a unique alternans solution (Fig. 3). The transition to alternans in a fiber appears to be more complicated. First, when paced at sufficiently large values of BCL, a fiber has a single 1:1 solution, regardless of the initial condition. However, as shown in the previous section, the fiber can develop multiple patterns of alternans. Therefore, it is interesting to ask how different alternans patterns are related to one another and how they are connected to the 1:1 solution. To address these questions, we utilize the upsweep protocol to trace the "origins" of different alternans solutions. Starting at BCL=375 ms from a pattern in Fig. 4,

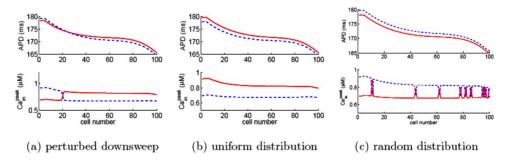


FIG. 7. (Color online) Three selected alternans solution in a homogeneous fiber of 100 cells with positive $Ca \rightarrow V_m$ coupling. The pacing period is 300 ms in all panels. (a) is obtained using a perturbed downsweep protocol. (b) and (c) are obtained by direct pacing with initial uniform distribution and random distribution of Ca_{in} , respectively. (See text for details.) The last ten beats of the steady-state solutions are plotted; odd beats are red solid and even beats are blue dashed.

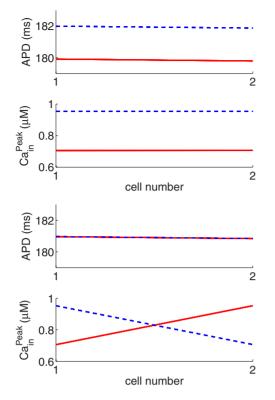


FIG. 8. (Color online) A "fiber" of two cells with positive Ca $\rightarrow V_m$ coupling can have both spatially synchronized and desynchronized alternans solutions. The pacing period is BCL=300 ms. The last ten beats of the steady-state solutions are plotted; odd beats are red solid and even beats are blue dashed.

we increase BCL by 1 ms every 200 beats until the alternans solution either changes to a different alternans pattern or becomes a 1:1 pattern. Then, we mark that value of BCL as the "starting" point for the studied alternans pattern. Under this protocol, each solution in Fig. 4 transforms into another alternans solution at a starting BCL shown in Table I. Specifically, pattern (a) "starts" at BCL=416 ms and patterns (b)-(f) start at BCL values between 402 and 405 ms. Further numerical simulations show that the fiber first undergoes alternans at BCL=425 ms. This is counterintuitive because, in the underlying cell's model, the onset of alternans occurs at 401 ms (cf. Fig. 3). Since alternans solutions in the fiber can occur earlier than alternans in the single-cell model, we hypothesize that these solutions are induced from the interaction between cellular dynamics and electrotonic coupling (*in*duced alternans). We further hypothesize that alternans may also result from the intrinsic instability mechanism that leads to alternans in the underlying cellular model (intrinsic alternans). Noticing that the spatially concordant alternans in pattern (d) starts at BCL=402 ms, near the onset of alternans in the single-cell model, we hypothesize that concordant alternans is a result of intrinsic alternans.

TABLE I. Origins of the alternans patterns in Fig. 4.

Pattern	(a)	(b)	(c)	(d)	(e)	(f)
Starting BCL (ms)	416	405	404	402	403	404

Using upsweep protocols, we also trace the origins of the alternans in Fig. 5 for the two coupled cells. Numerical simulations show that the desynchronized solution in Fig. 5(a) stops at BCL=405 ms and the synchronized solution in Fig. 5(b) stops at BCL=401 ms. Moreover, we also find that different alternans solutions for fibers with positive $Ca_{in} \rightarrow V_m$ coupling stop at different values of BCL. Thus, the bifurcation structure for a fiber is much more complicated than that for a single cell. Moreover, the numerical results suggest that the multiple alternans patterns created from different bifurcations are due to the interaction between electrotonic coupling and instability in calcium cycling.

B. Bifurcation mechanism

Shiferaw *et al.* showed that the bidirectional V_m -Ca coupling in the Shiferaw-Fox model can be captured by a 2D mapping model [13]. The work of Shiferaw *et al.* provides a theoretical framework for understanding various experimental observations including electromechanically in-phase and out-of-phase alternans and quasiperiodic oscillations of voltage and calcium. Here, we follow their approach and take a mapping model in the following form [36]:

$$A_{n+1} = f(D_n) + \gamma (C_{n+1} - C_{\text{crit}}),$$
 (5)

$$C_{n+1} = \mu D_n - g(C_n), \tag{6}$$

where A_n , D_n , and C_n represent the APD, DI, and peak [Ca_{in}] concentration of the *n*th beat, respectively. By definition, it follows that D_n =BCL- A_n ; see Fig. 1. The function *f* stands for the APD restitution. The second term in Eq. (5) accounts for the influence of [Ca_{in}] on APD. Negative Ca_{in} $\rightarrow V_m$ coupling corresponds to $\gamma < 0$ and positive coupling to $\gamma > 0$. Due to graded release, μ has to be positive [15]. The function *g* determines the relation between Ca concentrations in two consecutive beats.

Using the map (5) and (6), we will show how the interaction between electrotonic coupling and instability in calcium cycling leads to multiple alternans patterns in fibers. Since the analysis below does not depend on the forms of fand g, we do not specify their forms here. Instead we present numerical examples based on concrete forms of f and g in the Appendix. The numerical examples there show that fiber models based on the map (5) and (6) are able to reproduce the phenomena observed in simulations of the Shiferaw-Fox model.

1. Bifurcation for single cells

Using a mapping model similar to (5) and (6), Shiferaw *et al.* [13] carried out a bifurcation analysis to show how different alternans solutions in single cells arise as a result of V_m -Ca coupling. Here, we briefly review that analysis, which serves as a starting point to understand bifurcation for fibers. To this end, consider a paced cell described by map (5) and (6). A 1:1 solution is a fixed point of the map and its stability is determined by the Jacobian matrix

$$J = \begin{pmatrix} -f' - \gamma \mu & -\gamma g' \\ -\mu & -g' \end{pmatrix}, \tag{7}$$

where f' measures the slope of the APD restitution and g' measures the slope of the [Ca_{in}] relation, and all derivatives are evaluated at the fixed point.

The strength of V_m -Ca coupling varies among species [13]. In this paper, we assume a weak coupling between V_m dynamics and Ca cycling, which allows easy analysis and yields insight on the instability mechanisms in cardiac cells as well as in fibers. To this end, we assume $\gamma \mu \ll f', g'$ in the following. Using a perturbation technique [37], we find that the eigenvalues of J to first order in $\gamma \mu$ are

$$-f'\left(1+\frac{\gamma\mu}{f'-g'}\right)$$
 and $-g'\left(1+\frac{\gamma\mu}{g'-f'}\right)$. (8)

Since $\gamma \mu$ is small, one can see that a period-doubling bifurcation occurs if one of the slopes, f' or g', becomes sufficient large (compared to 1). Thus, if the bifurcation is caused by increase in f', we say the alternans is APD driven. On the other hand, it is called a Ca-driven alternans if the bifurcation is due to increase in g'. For simplicity, if alternans is APD driven, we assume g' remains small and will not induce dynamic instability for all physiologically interesting parameter regimes, and vice versa.

APD-driven alternans. In case of APD-driven alternans, it follows that f' > g'. To first order in $\gamma \mu$, the unstable eigenvector is

$$\left(1, \quad \frac{1}{f'-g'} - \frac{\gamma \mu f'}{(f'-g')^3}\right)^T, \tag{9}$$

which is in phase in APD and [Ca_{in}]. Therefore, APD-driven alternans is electromechanically in phase.

Calcium-driven alternans. In the case of Ca-driven alternans, it follows that g' > f'. To first order in $\gamma\mu$, the unstable eigenvector is

$$\left(\frac{\gamma\mu g'}{g'-f'}, \quad 1\right)^T.$$
 (10)

For positive $Ca_{in} \rightarrow V_m$ coupling, it follows that $\gamma > 0$ and thus alternans is electromechanically in phase. On the other hand, for negative $Ca_{in} \rightarrow V_m$ coupling, it follows that $\gamma < 0$ and thus alternans is electromechanically out of phase.

2. Bifurcation for two coupled cells

To understand the bifurcation mechanism for fibers, we start with the extreme case of two coupled cells. Denoting the locations of the cells by x_1 and x_2 , we represent the APD, DI, and $[Ca_{in}]$ of cell *i* for the *n*th beat as $A_n(x_i)$, $D_n(x_i)$, and $C_n(x_i)$, respectively. Following Fox *et al.* [38], we account for electrotonic coupling between the cells using weighted averaging:

$$A_{n+1}(x_1) = \rho(0)F(x_1) + \rho(1)F(x_2), \tag{11}$$

$$A_{n+1}(x_2) = \rho(1)F(x_1) + \rho(0)F(x_2), \qquad (12)$$

$$C_{n+1}(x_1) = \mu D_n(x_1) - g(C_n(x_1)), \qquad (13)$$

$$C_{n+1}(x_2) = \mu D_n(x_2) - g(C_n(x_2)), \qquad (14)$$

where

$$F(x_i) = f(D_n(x_i)) + \gamma [C_{n+1}(x_i) - C_{crit}], \quad (15)$$

and $\rho(0)$ and $\rho(1)$ are the weighting functions. Here, we have neglected the effect of dispersion due to conduction velocity restitution since the fiber is short and thus the influence of conduction restitution on APD is insignificant. On the time scale of one APD, V_m diffuses on a spatial scale of a few tens of cells [20]. Thus, the coupling in APD between two neighboring cells is strong. As a result, $\rho(0) \approx \rho(1)$ and $\rho(0)$ $+\rho(1)=1$. Therefore, we let $\rho(0)=1/2+\varepsilon$ and $\rho(1)=1/2-\varepsilon$, where $0 < \varepsilon < 1$. For a 1:1 solution, the two cells have the same values of APD and [Ca_{in}]. Thus, the Jacobian matrix can be written as

$$J = \begin{pmatrix} \rho(0)(-f' - \gamma\mu) & \rho(1)(-f' - \gamma\mu) & -\rho(0)\gamma g' & -\rho(1)\gamma g' \\ \rho(1)(-f' - \gamma\mu) & \rho(0)(-f' - \gamma\mu) & -\rho(1)\gamma g' & -\rho(0)\gamma g' \\ -\mu & 0 & -g' & 0 \\ 0 & -\mu & 0 & -g' \end{pmatrix},$$
(16)

where all derivatives are evaluated at the 1:1 solution.

Using perturbation theory [37], we find that the eigenvalues to first order in $\gamma\mu$ and ε are

$$-f'\left(1+\frac{\gamma\mu}{f'-g'}
ight), \quad -g'\left(1+\frac{\gamma\mu}{g'-f'}
ight), \quad -2\varepsilon f',$$

$$-g' - 2\varepsilon \gamma \mu. \tag{17}$$

Note that the first two eigenvalues are the same as those for the case of single cells [cf. Eq. (8)]. Thus, they are due to intrinsic membrane dynamics. The last two eigenvalues are induced by the electrotonic coupling as manifested by the small coupling parameter ε .

APD-driven alternans. For APD-driven alternans, the intrinsic bifurcation is the only mechanism to produce alternans. The corresponding eigenvector to first order in $\gamma\mu$ and ε is

$$\left(1, \quad 1, \quad \frac{1}{f'-g'} - \frac{\gamma\mu f'}{(f'-g')^3}, \quad \frac{1}{f'-g'} - \frac{\gamma\mu f'}{(f'-g')^3}\right)^T.$$
(18)

Thus, APD-driven alternans tends to be electromechanically in phase and spatially synchronized.

Calcium-driven alternans. For Ca-driven alternans, bifurcation can happen intrinsically when $g' + \gamma \mu g' / (g' - f') = 1$ (*intrinsic bifurcation*). Bifurcation can also be induced by electrotonic coupling when $g' + 2\varepsilon \gamma \mu = 1$ (*induced bifurcation*). To first order in $\gamma \mu$ and ε , the eigenvector corresponding to the intrinsic bifurcation is

$$\left(\frac{\gamma\mu g'}{g'-f'}, \frac{\gamma\mu g'}{g'-f'}, 1, 1\right)^T.$$
 (19)

To first order in $\gamma\mu$ and ε , the eigenvector corresponding to the induced bifurcation is

$$(2\varepsilon\gamma\mu, -2\varepsilon\gamma\mu, 1, -1)^T.$$
 (20)

Therefore, while the intrinsic bifurcation gives a spatially synchronized solution, the induced bifurcation leads to a spatially desynchronized pattern, and the electromechanical phase is determined by the sign of γ .

Subtracting the intrinsic eigenvalue from the induced one yields $\gamma \mu [g'(g'-f')-2\varepsilon]$, which has the same sign as γ since g' > f' (Ca-driven alternans) and $\mu > 0$ (graded release). If $\gamma < 0$, the induced eigenvalue is more negative; therefore, the spatially desynchronized pattern, arising from the induced bifurcation, occurs before the spatially synchronized pattern, arising from the intrinsic bifurcation. On the other hand, if $\gamma > 0$, the intrinsic eigenvalue is more negative; therefore, the spatially desynchronized pattern occurs after the spatially synchronized pattern. The analysis is in agreement with numerical simulations of the Shiferaw-Fox model using upsweep protocols. Therefore, when the BCL is continuously decreased from a 1:1 solution, it is more likely that a spatially synchronized solution for the positive Ca_{in} $\rightarrow V_m$ coupling case will be observed, whereas it is more likely that a spatially desynchronized pattern will be observed in the negative $Ca_{in} \rightarrow V_m$ coupling case. This is probably why Sato et al. observed different solution patterns for negative and positive $Ca_{in} \rightarrow V_m$ couplings in their downsweep simulations.

The conclusion based on two coupled cells may be extended to short fibers; however, we expect long fibers to have more complicated phenomena than presented here; in particular, because the dispersion effect may play an important role in long fibers as shown by Echebarria and Karma [20,21] and by Dai and Schaeffer [22]. Nevertheless, the simple case sheds light on the effect of electrotonic coupling, the difference between APD- and Ca-driven alternans, and the difference between negative and positive $Ca_{in} \rightarrow V_m$ couplings.

3. Understanding bifurcation for fibers

For Ca-driven alternans in fibers, we start with the extreme case of $\mu=0$. As can be seen from the cellular model (5) and (6), in this case, Ca_{in} dynamics becomes independent of V_m dynamics. Because Ca_{in} dynamics is not influenced by electrotonic coupling, when alternans develops in a fiber, cells can arbitrarily choose their phases in $[Ca_{in}]$. Depending on the initial conditions, Ca alternans on a fiber can take different spatial patterns. When μ is small but nonzero, [Ca_{in}] values in different cells are weakly coupled due to feedback of V_m dynamics and electrotonic coupling. If this weak coupling does not suppress the coexistence of multiple solutions, the alternans pattern will become sensitive to initial conditions and pacing protocols. Under certain pacing protocols, Cain alternans on a fiber may possess a complex pattern with multiple phase reversals, as shown in Figs. 4(f), 6(c), and 7(c). In these examples, the spatial patterns of APD are much less complex, which is because APD on the fiber is an averaging effect over many cells due to the fast diffusion of V_m [20,21]. As a result of electrotonic coupling, different spatial patterns of Cain alternans may correspond to similar spatial patterns of APD. Moreover, as shown in simulations of the Shiferaw-Fox model (see Figs. 4, 6, and 7) and of the mapping model (see Fig. 10 in the Appendix), spatial dyssynchrony in Cain alternans may induce spatially discordant APD alternans, which verifies the hypothesis of Aistrup et al. [26].

V. SUMMARY AND DISCUSSION

Using numerical simulation and theoretical analysis, we have investigated spatiotemporal patterns of calcium-driven alternans. The main finding is that, although alternans of an isolated cell is solely determined by the pacing period, the solution in a fiber is sensitive to pacing protocols and initial conditions. To the author's knowledge, this is the first report on the coexistence of multiple alternans solutions for cardiac fibers. We have further verified that the coexistence of multiple alternans solutions is independent of the length of the fiber, junctional Ca diffusion, and the type of $Ca_{in} \rightarrow V_m$ coupling. Since multiple solutions also exist for fibers of as few as a couple of cells, the phenomenon does not require steep conduction velocity restitution. Another interesting observation is that complex patterns of Cain alternans with multiple phase reversals in neighboring cells may arise in a homogeneous fiber with both negative and positive $Ca_{in} \rightarrow V_m$ couplings. The simulation results also verify the hypothesis of Aistrup et al. that spatially desynchronized Ca signaling may lead to spatially discordant APD alternans [26].

We have also explored the bifurcation mechanism for the coexistence of multiple solutions. Numerical simulations and symbolic bifurcation analyses trace the onset of multiple alternans patterns to a number of bifurcations induced by the interaction between electrotonic coupling and an instability in calcium cycling. The bifurcation here bears similiarity to that of a ring model. For example, Courtmanche *et al.* [39] used the Beeler-Reuter model [40] to show that, when the length of the ring is reduced, there is an infinite-dimensional Hopf bifurcation, which creates an infinite number of quasiperiodic solutions. Since the Beeler-Reuter model does not account for intracellular calcium cycling, it will be interesting to investigate how calcium-induced alternans propagates in a ring.

The simulations in this paper lead to another interesting observation-the onset of altrenans in a fiber does not occur at the same value of BCL as that in the underlying single-cell model, a phenomenon against the common wisdom. This counterintuitive observation indicates that one cannot directly relate the behavior of fibers to that of single-cell models. Another intriguing example about the differences between alternans in fibers and alternans in single cells is presented in the work of Cherry and Fenton [41], who analyzed two models of canine ventricular myocytes: the Fox-MchHarg-Gilmour model [28] and the Hund-Rudy (HR) model [42]. Simulations of both models show that the bifurcation structures for fibers are different than those for the corresponding single cells (see Fig. 5 in [41]). Most interestingly, the HR 0D model (a single cell) shows alternans for BCL between 180 and 230 ms whereas the HR 1D model (a 1.25-cm-long fiber) does not undergo alternans for all values of BCL studied. Examples in [41] as well as examples in the current paper show that stability of a fiber may not be inferred from stability of the single cell model, and vice versa.

The results in this paper are based on numerical simulations and hypothesized mapping models. Here, we describe potential experiments to verify the numerical observations. Probably the simplest experiment would measure the spatial patterns of APD or Ca alternans on a fiber under various pacing protocols. It is well know that in vitro experiments suffer nonstationary drifts due to tissue dying. However, this is a slower process compared to typical experimental protocols, which last a few tens of minutes. Moreover, the time constant of cardiac tissue is on the order of a few tens of seconds [30]. Therefore, it is possible to conduct multiple well-designed protocols in a time interval when tissue dynamics remains stationary. Other related experiments include the work of Fenton [43], where the action potential is measured using a microelectrode at a cell in a slice of paced dog epicardium tissue. Fenton showed that the onset and the form of alternans at the measured cell depend on the pacing history. While Fenton's observations may be attributable to factors such as cardiac memory, the coexistence of multiple alternans patterns in cardiac tissue may be another contributing element.

As pointed out by Qu and Weiss [44], because the Ca_{in} and V_m instabilities are always being regulated by each other, it remains a challenge to identify the sources of instabilities in cardiac experiments. Recent work by Sato *et al.* [45] and by Jordan and Christini [46] has developed theoretical criteria to assess the relative contributions of V_m and Ca_{in} dynamics in inducing cardiac alternans. The coexistence of multiple alternans patterns observed here may provide another possible criterion for this purpose. If the phenomenon is verified in experiments, it will raise questions in detection of alternans. New theories will also be needed to understand the de-

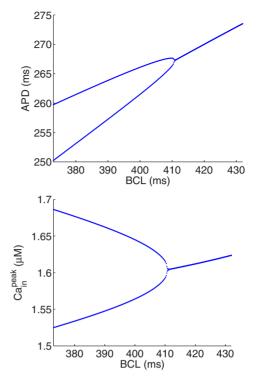


FIG. 9. (Color online) Bifurcation diagrams of the map (5) and (6) with negative V_m -Ca coupling.

velopment of different alternans patterns to ensure better prediction and control of spatiotemporal alternans.

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APPENDIX: SIMULATION OF THE MAPPING MODEL

Here, we present numerical examples for single cells and for fibers using the mapping model (5) and (6). For simulation purposes, we let $f(D_n) = \alpha(1 - \beta e^{-D_n/\tau})$ and $g(C_n) =$ $-s(C_n - C_{crit}) - C_{min}$. Inspired by the results of Shiferaw *et al.* [Fig. 10(d) in [27]], we choose the slope *s* in *g* to be *s* $= \frac{1}{2}(1 - \theta)s_1 + \frac{1}{2}(1 + \theta)s_2$, where $\theta = \tanh[k(C - C_{crit})]$. To produce numerical results, we adopt the following set of parameters: $\alpha = 500$ ms, $\beta = 0.62$, $\tau = 500$ ms, $\gamma = -30$ ms/ μM , μ $= 0.0025 \ \mu M/ms$, $C_{crit} = 1.6 \ \mu M$, $C_{min} = 1.25 \ \mu M$, $s_1 = -2.5$, $s_2 = 0.01$, and $k = 10 \ \mu M^{-1}$. Note that we have arbitrarily chosen a negative value of γ although a positive value of γ will produce similar phenomena. Moreover, we make the

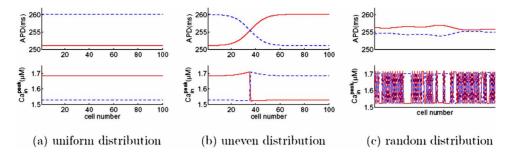


FIG. 10. (Color online) Simulation of the coupled-map model: three selected alternans solutions for a homogeneous fiber of 100 cells with negative $Ca \rightarrow V_m$ coupling. The pacing period is 375 ms in all panels. (a) starts from a uniform initial distribution in Ca_{in} . (b) starts from an initial condition where Ca_{in} is set to be 1.5 μM in the first 35 cells and 1.7 μM in the remaining cells. (c) starts from a random initial distribution in Ca. The last ten beats of the steady-state solutions are plotted, where odd beats are represented by red solid lines and even beats by blue dashed lines.

magnitude of $\gamma\mu$ small to simulate weak V_m -Ca coupling.

1. Alternans for single cells

With the chosen parameters, map (5) and (6) causes alternans through a period-doubling bifurcation at B = 410.958 ms; see the bifurcation diagrams in Fig. 9. For this simple model, it can be analytically shown that, for a given BCL, the cell has either a unique 1:1 solution or a unique alternans solution. However, as we will see below, a fiber has multiple alternans solutions induced from electrotonic coupling.

2. Alternans for fibers

Based on the mapping model (5) and (6) for single cells, we construct a coupled-map model to simulate fibers. Coupled-map models have been used by a few authors to study propagation of the action potential [38,47–49]. Here, we follow that approach. Specifically, we consider a fiber consisting of M identical cells, located at x_i , i=1,2,...,M. The distance between two neighboring cells is denoted by Δx . At cell x_i , the APD and DI are related by the following equation:

$$D_{n+1}(x_i) = T_{n+1}(x_i) - A_{n+1}(x_i), \tag{A1}$$

where $T_{n+1}(x_i)$ is the time interval between two consecutive activations of site x_i . The time $T_{n+1}(x_i)$ is determined by the propagation time from the pacing site to x_i , that is,

$$T_{n+1}(x_i) = B + \sum_{j=1}^{i-1} \frac{\Delta x}{V_{n+1}(x_i)} - \sum_{j=1}^{i-1} \frac{\Delta x}{V_n(x_i)},$$
 (A2)

where $V_n(x_i)$ stands for the conduction velocity. We adopt a conduction velocity from Fox *et al.* [48], $V_n(x_i) = V_{max}(1 - \exp\{-[D_n(x_i) + \beta]/\delta\})$ with parameters $V_{max} = 0.72$ cm/ms, $\beta = 17.408$, and $\delta = 14$. To account for the effect of electrotonic coupling, we modify the weighted averaging formula in Fox *et al.* [38] as follows:

$$A_{n+1}(x_i) = \frac{\sum_{\substack{j=\max(-30,1-i)\\\min(30,M-i)\\\min(30,M-i)}} w(j)\overline{A}_{n+1}(x_j)}{\sum_{\substack{j=\max(-30,1-i)}} w(j)},$$
 (A3)

where $\overline{A}_{n+1}(x_j) = f(D_n(x_j)) + \gamma [C_{n+1}(x_j) - C_{crit}]$ and $w(j) = \exp(-0.0067j^2)$. The equation for [Ca_{in}] reads

$$C_{n+1}(x_i) = \mu D_n(x_i) + g(C_n(x_i)).$$
(A4)

Simulations of the coupled-map model (A1)–(A4) show that the alternans pattern on a fiber depends on the pacing history, a phenomenon consistent with that observed in simulations of the Shiferaw-Fox model. For example, Fig. 10 shows three selected solutions for a homogeneous fiber of 100 cells paced at B=375 ms. The panels in Fig. 10 have different initial conditions in [Ca_{in}]. Figure 10(a) starts from a uniform initial distribution. Figure 10(b) starts from an initial condition where [Ca_{in}] is set to be 1.5 μM in the first 35 cells and 1.7 μM in the remaining cells. Figure 10(c) starts from a random initial distribution. Thus, the coupled-map model is able to reproduce the coexistence of multiple alternans solutions as observed in the Shiferaw-Fox model.

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