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Nonlinear analysis of periodic waves in a neural field model

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Various types of brain activity, including motor, visual, and language, are accompanied by the propagation of periodic waves of electric potential in the cortex, possibly providing the synchronization of the epicenters involved in these activities. One example is cortical electrical activity propagating during sleep and described as traveling waves¹. These waves modulate cortical excitability as they progress. Clinically-related examples include cortical spreading depression in which a wave of depolarization propagates in migraine but also in stroke, hemorrhage or traumatic brain injury². Here we consider the possible role of epicenters and explore a neural field model with two nonlinear integro-differential equations for the distributions of activating and inhibiting signals. It is studied with symmetric connectivity functions characterizing signal exchange between two populations of neurons, excitatory and inhibitory. Bifurcation analysis is used to investigate the emergence of periodic traveling waves and of standing oscillations from the stationary, spatially homogeneous solutions, and the stability of these solutions. Both types of solutions can be started by local oscillations indicating a possible role of epicenters in the initiation of wave propagation.

The concept of periodic traveling waves appeared in the early 1970's and since that time investigators have actively explored the mechanisms, computational principles and functional role of these waveforms^{3,4}. This paper focuses on how modeling information can be used to guide electrical stimulation in the targeted manipulation of brain networks in human patients affected by neurological disorders. We explore a neural field model with two nonlinear integro-differential equations and perform a bifurcation analysis. Results suggest that it is possible to determine the direction of propagation and that epicenters may play a role in the initiation of wave propagating in the cortex. These results will guide the determination of the optimal localization to apply external brain stimulation (forcing term) to restore wave propagation in damaged cortical tissue.

I. INTRODUCTION

All types of brain activity are accompanied by formation of spatiotemporal patterns of electric potential in the cerebral cortex, in particular, wave fronts and periodic waves^{5–7}. They provide subthreshold depolarization to individual neurons and increase their spiking probability. They possibly synchronize different parts of the brain and organize spatial phase distributions in a population of neurons⁸. Traveling waves can

also be detected from the phase relations between oscillations recorded in different cortical regions⁹.

EEG recordings from the cortical surface in awake neurosurgical patients performing a memory task can show focal regions or clusters of 2-15 Hz oscillations ¹⁰. These oscillations formed traveling waves that propagated at about 0.25-0.75 m/s and correlated with the subject's performance. In other words, they guide the spatial propagation of neural activity demonstrating large scale spatially coordinated oscillations.

The relevance of traveling waves is actively explored in patients with stroke, subarachnoid hemorrhage or traumatic brain injury. The occurrence of spreading depression or depolarization in these pathologies is a well documented phenomenon. Spreading depression or depolarization of neurons and glial cells is also reported in patients with migraine². However, the relevance of traveling waves is also explored in healthy subjects. Indeed traveling waves have been observed in processes involving cognition (such as perception, learning, recognition, memory, language) (see for example Ref. 10 for working memory or Ref. 11 for language) but also in motor behaviors^{9,12} or vision¹³.

For each human subject the examination of slow waves of cortical activity during sleep-awake states using simultaneous scalp EEG and intracranial recordings allowed the identification of the set of intracranial contacts involved in a larger percentage of detected events¹⁴. These contacts are called «hubs» because the wave had large probability of passing through the region close to the contact. Using probabilities they were able to reconstruct a preferential propagation network for each subject. Slow waves have been reported to propagate across cortical areas at about 1 m/s with multiple propagation paths and several points of origin¹⁴. These waves appear to shape and

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strengthen neuronal networks. There are numerous other evidences of wave propagation during brain activity^{11,15}.

There are two main classes of cortical activity which are described by neural mass models and neural field models (see, e.g., Ref. 16). The former is based on a discrete representation of brain networks with spatially homogeneous neural elements connected to each other. In neural field models, cerebral cortex is considered as a continuous medium with a nonlocal interaction due to neuron connectivity. Such models were first introduced in Ref. 17 followed by many others for one or two neuron populations ^{18–20}, discrete or distributed time delay^{20,21}, linear adaptation²², refractoriness²³, and other questions ^{22,24–27}.

In this work we consider the model

$$\frac{\partial u}{\partial t} = P_{1,1} * \psi_1(u) - P_{1,2} * \psi_2(v) - \sigma u,
\frac{\partial v}{\partial t} = P_{2,1} * \psi_1(u) - P_{2,2} * \psi_2(v) - \sigma v,$$
(1)

where

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$$P_{i,j}(x) = a_{i,j}e^{-b_{i,j}|x|}, i, j = 1, 2,$$

 $a_{i,j}, b_{i,j}$, and σ are positive constants, $\psi_1(u)$ and $\psi_2(v)$ are bounded smooth growing (sigmoid, i.e. having an S-shaped curve) functions on the whole real axis. * denotes convolution. The variables u and v characterize the activity of the neural population of excitatory and inhibitory neurons, respectively. System (1) represents a particular case without time delay of the model studied in Ref. 28. Considered on the whole real axis, under some natural conditions, it has a stationary, spatially homogeneous solution (u_0, v_0) . This solution can be stable or unstable depending on the values of parameters. In particular, it can lose its stability due to a simple real eigenvalue or a pair of complex conjugate eigenvalues crossing the imaginary axis. Linear stability analysis of the stationary homogeneous solution was carried out in Ref. 28. Here we are interested in nonlinear bifurcation analysis allowing the determination of stability of the bifurcating solutions. We will obtain the conditions providing the existence and stability of standing and periodic traveling waves.

Furthermore, we are interested in the initiation of the periodic waves by space localized oscillations. From the biological point of view, they correspond to the oscillations in the hubs, which process received information and send it to other hubs in the form of traveling waves. We show that wave initiation occurs depending on the properties of the localized oscillations.

II. SPECTRUM OF THE LINEARIZED EQUATIONS

Let $(u_0(\sigma), v_0(\sigma))$ be a pair of constants satisfying

$$\sigma u_0 = 2 \frac{a_{1,1}}{b_{1,1}} \psi_1(u_0) - 2 \frac{a_{1,2}}{b_{1,2}} \psi_2(v_0),$$

$$\sigma v_0 = 2 \frac{a_{2,1}}{b_{2,1}} \psi_1(u_0) - 2 \frac{a_{2,2}}{b_{2,2}} \psi_2(v_0).$$

Then $(u_0(\sigma), v_0(\sigma))$ are solutions to (1). By shifting the functions

$$u(x,t) = u_0(\sigma) + \tilde{u}(x,t), \quad v(x,t) = v_0(\sigma) + \tilde{v}(x,t),$$

linearizing the equations (1), and dropping the tildes, we get

$$\frac{\partial u}{\partial t} = \psi_1'(u_0(\sigma))P_{1,1} * u - \psi_2'(v_0(\sigma))P_{1,2} * v - \sigma u(x,t),$$

$$\frac{\partial v}{\partial t} = \psi_1'(u_0(\sigma))P_{2,1} * u - \psi_2'(v_0(\sigma))P_{2,2} * v - \sigma v(x,t).$$

Looking for solutions in the form of

$$u(x,t) = \exp(\lambda t)u(x), \quad v(x,t) = \exp(\lambda t)v(x)$$

and after applying the Fourier transform, we obtain the characteristic equations

$$\begin{split} & \psi_1'(u_0(\sigma)) \hat{P}_{1,1}(\xi) \hat{u}(\xi) - \psi_2'(v_0(\sigma)) \hat{P}_{1,2}(\xi) \hat{v}(\xi) = (\sigma + \lambda) \hat{u}(\xi), \\ & \psi_1'(u_0(\sigma)) \hat{P}_{2,1}(\xi) \hat{u}(\xi) - \psi_2'(v_0(\sigma)) \hat{P}_{2,2}(\xi) \hat{v}(\xi) = (\sigma + \lambda) \hat{v}(\xi), \\ & \text{where} \end{split}$$

$$\hat{P}_{i,j}(\xi) = \frac{2a_{i,j}b_{i,j}}{b_{i,j}^2 + \xi^2}.$$

Let
$$\mu = \mu(\xi; \sigma) = \sigma + \lambda(\xi; \sigma) \in \mathbb{C}$$
 and

$$\hat{Q}_{i,1}(\xi;\sigma) = \psi_1'(u_0(\sigma))\hat{P}_{i,1}(\xi), \hat{Q}_{i,2}(\xi;\sigma) = \psi_2'(v_0(\sigma))\hat{P}_{i,2}(\xi).$$

Then the characteristic polynomial becomes

$$\begin{split} \mu^2 - \left(\hat{Q}_{1,1}(\xi) - \hat{Q}_{2,2}(\xi)\right)\mu - \\ - \left(\hat{Q}_{1,1}(\xi)\hat{Q}_{2,2}(\xi) - \hat{Q}_{1,2}(\xi)\hat{Q}_{2,1}(\xi)\right) = 0 \end{split}$$

with two roots

$$\mu(\xi) = \frac{1}{2} \left(\hat{Q}_{1,1}(\xi) - \hat{Q}_{2,2}(\xi) \right) \pm$$

$$\pm \frac{1}{2} \sqrt{ \left(\hat{Q}_{1,1}(\xi) + \hat{Q}_{2,2}(\xi) \right)^2 - 4 \hat{Q}_{1,2}(\xi) \hat{Q}_{2,1}(\xi)}.$$

If the cross-connectivity functions $P_{1,2}(x)$ and $P_{2,1}(x)$ are strong enough (their amplitudes $a_{1,2}$ and $a_{2,1}$ are large enough), we can ensure that the expression under the square root is always negative

$$\left(\hat{Q}_{1,1}(\xi) + \hat{Q}_{2,2}(\xi) \right)^2 - 4 \hat{Q}_{1,2}(\xi) \hat{Q}_{2,1}(\xi) < 0, \quad -\infty < \xi < \infty,$$

which leads to complex values of $\mu(\xi)$ such that

$$\begin{split} \text{Re}\mu(\xi) &= \frac{1}{2} \left(\hat{\mathcal{Q}}_{1,1}(\xi) - \hat{\mathcal{Q}}_{2,2}(\xi) \right), \\ \text{Im}\mu(\xi) &= \pm \frac{1}{2} \sqrt{4 \hat{\mathcal{Q}}_{1,2}(\xi) \hat{\mathcal{Q}}_{2,1}(\xi) - \left(\hat{\mathcal{Q}}_{1,1}(\xi) + \hat{\mathcal{Q}}_{2,2}(\xi) \right)^2}. \end{split}$$

Consider a function $f(\xi) : \mathbb{R} \to \mathbb{R}$ that is related to the real part of the characteristic values $\mu(\xi)$:

$$f(\xi) = \frac{\alpha_1}{\beta_1 + \xi^2} - \frac{\alpha_2}{\beta_2 + \xi^2}, \quad \alpha_i > 0, \quad \beta_i > 0.$$

We are interested in its extreme values and the corresponding frequency ξ where they are attained.

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Lemma 1. If $\alpha_1 = \alpha_2$ or $\beta_1 = \beta_2$ then $|f(\xi)| \le |f(0)|$ for all ξ .

Proof. Let $\alpha_1 = \alpha_2 = \alpha$. Then

$$f(\xi) = \frac{\alpha(\beta_2 - \beta_1)}{(\beta_1 + \xi^2)(\beta_2 + \xi^2)}.$$

Unless $\beta_1 = \beta_2$ and $f(\xi) \equiv 0$, $|f(\xi)|$ is monotonically decreasing on $\xi > 0$ and increasing on $\xi < 0$.

Now let $\beta_1 = \beta_2 = \beta$. Then

$$f(\xi) = \frac{\alpha_1 - \alpha_2}{\beta + \xi^2}.$$

Unless $\alpha_1 = \alpha_2$ and $f(\xi) \equiv 0$, $|f(\xi)|$ is monotonically decreasing on $\xi > 0$ and increasing on $\xi < 0$.

Let $\alpha_1 \neq \alpha_2$ and $\beta_1 \neq \beta_2$; we introduce

$$\beta_3 = \frac{\alpha_1 \beta_2 - \alpha_2 \beta_1}{\alpha_1 - \alpha_2}$$

and note that it can take any real value, not necessarily positive ones. Then the derivative of $f(\xi)$ is

$$\begin{split} f'(\xi) &= 2\xi(\alpha_1 - \alpha_2) \times \\ &\times \frac{(\beta_1 + \xi^2)(\beta_2 + \xi^2) - (\beta_3 + \xi^2)(\beta_1 + \beta_2 + 2\xi^2)}{(\beta_1 + \xi^2)^2(\beta_2 + \xi^2)^2}. \end{split}$$

It vanishes at $\xi = 0$ and, possibly, at

$$\xi^2 = -\beta_3 \pm \sqrt{(\beta_3 - \beta_1)(\beta_3 - \beta_2)} \equiv \eta_\pm$$

if the right hand side is real and positive.

Lemma 2. Both η_+ and η_- are real whenever

$$\beta_3 \leq \min(\beta_1, \beta_2) \text{ or } \beta_3 \geq \max(\beta_1, \beta_2).$$

The value η_{-} is negative when it is real, i.e. whenever

$$\beta_3 \leq \min(\beta_1, \beta_2) \text{ or } \beta_3 \geq \max(\beta_1, \beta_2).$$

The value η_+ is nonpositive whenever

$$\frac{\beta_1\beta_2}{\beta_1+\beta_2} \leq \beta_3 \leq \min(\beta_1,\beta_2) \text{ or } \beta_3 \geq \max(\beta_1,\beta_2).$$

Proof. The first statement is obvious. Let us assume that $\eta_+, \eta_- \in \mathbb{R}$. For the second statement,

1. if $\beta_3 \ge 0$ then $\eta_- < 0$ if and only if

$$\beta_3 > -\sqrt{(\beta_3 - \beta_1)(\beta_3 - \beta_2)},$$

so $0 \le \beta_3 \le \min(\beta_1, \beta_2)$ or $\beta_3 \ge \max(\beta_1, \beta_2)$;

2. if $\beta_3 < 0$ then $\eta_- < 0$ if and only if

$$\beta_1\beta_2 - \beta_3(\beta_1 + \beta_2) > 0$$
,

so simply $\beta_3 < 0$.

For the third statement, $\eta_{+} \leq 0$ if and only if

$$\beta_1\beta_2 - \beta_3(\beta_1 + \beta_2) \le 0$$

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whence the result follows.

Lemma 3. Let $\alpha_1 \neq \alpha_2$, $\beta_1 \neq \beta_2$, and $\beta_3 < \beta_1\beta_2/(\beta_1 + \beta_2)$. The derivative $f'(\xi)$ vanishes at only 3 points:

$$\xi = 0 \text{ and } \xi = \pm \sqrt{-\beta_3 + \sqrt{(\beta_3 - \beta_1)(\beta_3 - \beta_2)}}.$$

Proof. Since $\alpha_1 \neq \alpha_2$ and $\beta_1 \neq \beta_2$, $f'(\xi)$ turns to zero if and only if $\xi = 0$ or $\xi^2 = \eta_{\pm}$. Lemma 2 states that $\eta_- < 0$ and $\eta_+ > 0$ when $\beta_3 < \beta_1 \beta_2 / (\beta_1 + \beta_2)$. So $f'(\xi) = 0$ at $\xi = 0$ and $\xi = \pm \sqrt{\eta_+}$.

Since

$$f'(\xi) = 2(\alpha_1 - \alpha_2) \frac{\beta_1 \beta_2 - \beta_3 (\beta_1 + \beta_2)}{\beta_1^2 \beta_2^2} \xi + O(\xi^2)$$

in the vicinity of the origin, the function $f(\boldsymbol{\xi})$ has global maxima at

$$\xi^2 = -\beta_3 + \sqrt{(\beta_3 - \beta_1)(\beta_3 - \beta_2)}$$

when $\alpha_1 > \alpha_2$, and *global* minima when $\alpha_1 < \alpha_2$. Let

$$\alpha_1 = \psi_1'(u_0(\sigma))a_{1,1}b_{1,1}, \quad \beta_1 = b_{1,1}^2$$

and

$$\alpha_2 = \psi_2'(v_0(\sigma))a_{2,2}b_{2,2}, \quad \beta_2 = b_{2,2}^2,$$

then $\operatorname{Re}\mu(\xi) = f(\xi)$ and the largest real part $\sup_{\xi} \operatorname{Re}\mu(\xi)$ is attained at $\pm \xi_*(\sigma)$:

$$\xi_*(\sigma) = \sqrt{-\beta_3(\sigma) + \sqrt{(\beta_3(\sigma) - b_{1,1}^2)(\beta_3(\sigma) - b_{2,2}^2)}},$$

$$\beta_3(\sigma) = \frac{\psi_1'(u_0(\sigma))a_{1,1}b_{1,1}b_{2,2}^2 - \psi_2'(v_0(\sigma))a_{2,2}b_{2,2}b_{1,1}^2}{\psi_1'(u_0(\sigma))a_{1,1}b_{1,1} - \psi_2'(v_0(\sigma))a_{2,2}b_{2,2}}.$$

Thus $\sup_{\xi} \operatorname{Re} \lambda(\xi; \sigma) = \operatorname{Re} \lambda(\xi_*(\sigma)) = f(\xi_*(\sigma)) - \sigma$.

Theorem 1. *Let* $a_{12}, a_{21} \gg 1$ *and*

$$\frac{b_{22}}{b_{11}} < \frac{\psi_1'(u_0(\sigma))a_{11}}{\psi_2'(v_0(\sigma))a_{22}} < \left(\frac{b_{11}}{b_{22}}\right)^3.$$

Then the spectrum is complex $\operatorname{Im}\lambda(\xi;\sigma)\neq 0$ and $\operatorname{Re}\lambda(\xi;\sigma)<\operatorname{Re}\lambda(\xi_*(\sigma);\sigma)$ for all $\xi\neq\pm\xi_*(\sigma)$.

Proof. Follows from Lemmas 1-3 once we rewrite the constraint $\beta_3 < \beta_1 \beta_2 / (\beta_1 + \beta_2)$.

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$$\operatorname{Re}\lambda(\xi;0) < \operatorname{Re}\lambda(\xi_*(0);0), \quad \xi \neq \xi_*(0),$$
 (H1)

and

$$\lambda(\xi_*(0);0) = \pm i\omega_* \neq 0. \tag{H2}$$

Condition (H1) follows from Theorem 1 while the Hopf bifurcation condition (H2) can be satisfied with a careful choice of parameters. We also assume the transversality condition

$$\begin{split} \left. \left(\frac{d}{d\sigma} \text{Re} \lambda \left(\xi_*(0); \sigma \right) \right) \bigg|_{\sigma = 0} &= \\ &= -1 + \frac{1}{2} \psi_1''(u_0(0)) u_0'(0) \hat{P}_{11}(\xi_*(0)) - \\ &\quad - \frac{1}{2} \psi_2''(v_0(0)) v_0'(0) \hat{P}_{22}(\xi_*(0)) \neq 0. \end{split} \tag{H3}$$

That is the pair of imaginary eigenvalues passes the imaginary axis with nonzero speed.

Let $\xi_* = \xi_*(0)$ and $l_* = 2\pi/\xi_*$. Consider the Lebesgue space $L^2(0, l_*)$ with standard inner product and norm:

$$\langle f,g \rangle_{L^2(0,l_*)} = \int_0^{l_*} f \overline{g} dx, \quad \|f\|_{L^2(0,l_*)} = \sqrt{\langle f,f \rangle_{L^2(0,l_*)}}.$$

The complex exponentials $e_n=\frac{1}{\sqrt{l_*}}e^{\mathrm{i}n\xi_*x}$ form an orthonormal basis in $L^2(0,L_*)$. Also, consider the Sobolev space of l_* -periodic twice-differentiable functions $H^2_{per}(0,l_*)$ with inner product

$$\langle f,g\rangle_{H^2_{per}(0,l_*)} = \sum_{n\in\mathbb{Z}} (1+n^4\xi_*^4)\langle f,e_n\rangle_{L^2(0,l_*)}\langle e_n,g\rangle_{L^2(0,l_*)}$$

and Euclidean norm. We consider smooth functions to make the analysis extendable to diffusive models.

Let

$$w = \begin{bmatrix} u \\ v \end{bmatrix} \in X = \left(H_{per}^2(0, l_*)\right)^2$$

with inner product

$$\langle w, \tilde{w} \rangle_X = \langle w_1, \tilde{w}_1 \rangle_{H^2_{per}(0,l_*)} + \langle w_2, \tilde{w}_2 \rangle_{H^2_{per}(0,l_*)}$$

where w_j and \tilde{w}_j are the components of two-dimensional vector functions.

Let $M_k(\sigma) \in \mathbb{R}^{2 \times 2}$ be diagonal matrices

$$M_k(\sigma) = \frac{1}{k!} \begin{bmatrix} \psi_1^{(k)}(u_0(\sigma)) & 0\\ 0 & -\psi_2^{(k)}(v_0(\sigma)) \end{bmatrix}$$

and $\mathscr{P}_n \in \mathbb{R}^{2 \times 2}$ be Fourier filtering matrices

$$\mathscr{P}_n = \begin{bmatrix} \hat{P}_{11}(n\xi_*) & \hat{P}_{12}(n\xi_*) \\ \hat{P}_{21}(n\xi_*) & \hat{P}_{22}(n\xi_*) \end{bmatrix}.$$

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Any function $w \in X \subset (L^2(0, l_*))^2$ can be represented by its Fourier series

$$w = \sum_{n \in \mathbb{Z}} w_n e_n, \quad w_n = \begin{bmatrix} \langle w_1, e_n \rangle_{L^2(0, l_*)} \\ \langle w_2, e_n \rangle_{L^2(0, l_*)} \end{bmatrix} \in \mathbb{C}^2.$$

We can rewrite the initial equations as

$$w_t = L(\sigma)w + F(w, \sigma), \quad w \in X,$$

where

$$L(\sigma)w = \sum_{n \in \mathbb{Z}} (\mathscr{P}_n M_1(\sigma) - \sigma_* - \sigma) w_n e_n$$

and

$$F(w,\sigma) = \sum_{k\geq 2} \sum_{n\in\mathbb{Z}} \mathscr{P}_n M_k(\sigma) (w^{\odot k})_n e_n.$$

We use \odot to denote the Hadamard (element-wise) product of vectors and write $w^{\odot 2} = w \odot w$, $w^{\odot 3} = w \odot w \odot w$, etc., for Hadamard products of w with itself.

The characteristic equation at $\sigma = 0$ decouples into

$$(\mathscr{P}_n M_1(0) - \sigma_* - \lambda) w_n = 0, \quad n \in \mathbb{Z},$$

and $\lambda \in \mathbb{C}$ is a characteristic value if there exist $n \in \mathbb{Z}$ and $w_n \neq 0 \in \mathbb{C}^2$ that satisfy the equation above. Owing to (H2), there is a unique pair of imaginary characteristic values $\pm i\omega_*$ corresponding to $n = \pm 1$. They are degenerate because the connectivity functions are symmetric and the system as a whole is O(2)-equivariant. Let $\zeta_1 \neq 0 \in \mathbb{C}^2$ be an eigenvector:

$$\begin{split} &(\mathscr{P}_1 M_1(0) - \sigma_* - \mathrm{i} \omega_*) \zeta_1 = 0, \\ &\zeta_1 = \begin{bmatrix} \psi_2'(\nu_0(0)) \hat{P}_{12}(\xi_*) & \psi_1'(u_0(0)) \hat{P}_{11}(\xi_*) - \sigma_* - \mathrm{i} \omega_* \end{bmatrix}^T. \end{split}$$

Consider vector functions

$$q_1 = \zeta_1 e_1, \quad q_2 = \zeta_1 e_{-1}, \quad q_3 = \overline{\zeta_1} e_1, \quad q_4 = \overline{\zeta_1} e_{-1}.$$

In pairs, they span the eigenspaces

$$\mathcal{N}(L(0) - i\omega_*) = \operatorname{span}\{q_1, q_2\} \subset X,$$

 $\mathcal{N}(L(0) + i\omega_*) = \operatorname{span}\{q_3, q_4\} \subset X;$

together they span the center subspace

$$\mathcal{Q} = \text{span}\{q_1, q_2, q_3, q_4\} \subset X.$$

The basis $\{q_j\}$ is not orthogonal so we will use a biorthogonal system to get a projection operator onto \mathscr{Q} . Let $\zeta_1^* \neq 0 \in \mathbb{C}^2$ be an eigenvector of the adjoint matrix

$$(\mathscr{P}_1 M_1(0) - \sigma_* - \mathrm{i}\omega_*)^H \zeta_1^* = 0$$

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normalized so that $\langle \zeta_1, \zeta_1^* \rangle = 1$. Further, let

$$\begin{split} q_1^* &= \frac{1}{1 + \xi_*^4} \zeta_1^* e_1, \quad q_2^* = \frac{1}{1 + \xi_*^4} \zeta_1^* e_{-1}, \\ q_3^* &= \frac{1}{1 + \xi_*^4} \overline{\zeta_1^*} e_1, \quad q_4^* = \frac{1}{1 + \xi_*^4} \overline{\zeta_1^*} e_{-1} \end{split}$$

be the basis vectors of the eigenspaces

$$\mathcal{N}(L(0)^* + i\omega_*) = \text{span}\{q_1^*, q_2^*\} \subset X,$$

 $\mathcal{N}(L(0)^* - i\omega_*) = \text{span}\{q_3^*, q_4^*\} \subset X.$

It is easy to show that $\{q_i^*\}$ and $\{q_i\}$ are biorthogonal systems

$$\langle q_i, q_j^* \rangle_X = \begin{cases} 1, & i = j, \\ 0, & i \neq j. \end{cases}$$

We can thus define a projection operator

$$\pi: X \to \mathscr{Q}, \quad \pi w = \sum_{j=1}^{4} \langle w, q_j^* \rangle_X q_j$$

and a complementary subspace

$$\tilde{\mathcal{Q}} = \{ w \in X : \pi w = 0 \}.$$

Then $X = \mathcal{Q} \oplus \tilde{\mathcal{Q}}$.

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For a given $w \in X$, let $z \in \mathbb{C}^4$ (note that $z_4 = \overline{z_1}$ and $z_2 = \overline{z_3}$ as the functions are real-valued) be the vector of coefficients of πw with respect to the basis $\{q_j\}$. Let γ be the velocity of the critical characteristic value:

$$\gamma = \left. \left(\frac{d}{d\sigma} \lambda(\xi_*; \sigma) \right) \right|_{\sigma = 0}.$$

According to the center manifold theorem, the equations for z_1 and z_3 on the manifold have the following normal form²⁹:

$$\dot{z}_1 = (i\omega_* + \gamma\sigma)z_1 + c_1z_1|z_1|^2 + c_2z_1|z_3|^2 + \text{h.o.t.},
\dot{z}_3 = (i\omega_* + \gamma\sigma)z_3 + c_1z_3|z_3|^2 + c_2z_3|z_1|^2 + \text{h.o.t.},$$

where h.o.t. = $\mathcal{O}(\sigma(|z_1|^3 + |z_3|^3) + \sigma^2(|z_1| + |z_3|) + (|z_1| + |z_2|)^5)$. See Appendix A for the computation of the cubic coefficients c_1 and c_2 of the normal form.

The relations between c_1 and c_2 determine the stability of periodic traveling wave and standing wave solutions²⁹. To see this, we make a polar change of coordinates,

$$z_1 = \rho_1 e^{i\theta_1}, \quad z_3 = \rho_3 e^{i\theta_3},$$

and keep two equations out of four:

$$\dot{\rho}_{1} = \rho_{1}(\sigma \text{Re}\gamma + \text{Re}c_{1}\rho_{1}^{2} + \text{Re}c_{2}\rho_{3}^{2}) + \text{h.o.t.},
\dot{\rho}_{3} = \rho_{3}(\sigma \text{Re}\gamma + \text{Re}c_{1}\rho_{3}^{2} + \text{Re}c_{2}\rho_{1}^{2}) + \text{h.o.t.}.$$
(2)

These are the amplitude equations as ρ_1 and ρ_3 represent the amplitudes of counterpropagating periodic traveling waves. The system (2) can have up to four equilibria (see Table I),

TABLE I: Equilibria of the amplitude equations (2).

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Type of solution	$ ho_1$	$ ho_3$
Constant	0	0
Periodic traveling wave	$\sqrt{-rac{\sigma \mathrm{Re}\gamma}{\mathrm{Re}c_1}}$	0
Periodic traveling wave	0	$\sqrt{-\frac{\sigma \text{Re}\gamma}{\text{Re}c_1}}$
Standing wave	$\sqrt{-rac{\sigma \mathrm{Re}\gamma}{\mathrm{Re}c_1+\mathrm{Re}c_2}}$	$\sqrt{-\frac{\sigma \text{Re}\gamma}{\text{Re}c_1 + \text{Re}c_2}}$

and the stability of these equilibria defines the stability of periodic traveling waves and standing waves.

Let $\sigma Re \gamma > 0$ then

$$Rec_1 < 0$$
, $Re(c_1 + c_2) < 0$, $Re(c_1 - c_2) < 0$

leads to asymptotically stable standing waves while

$$Rec_1 < 0$$
, $Re(c_1 + c_2) < 0$, $Re(c_1 - c_2) > 0$

produces asymptotically stable periodic traveling waves.

IV. ONE CLASS OF PARAMETERS

The net result of Sections II-III are the explicit expressions of the normal form coefficients, which can be easily evaluated for a given set of parameters. However, these expressions are too cumbersome to aid in the *search* of suitable parameters. Theorem 1 is also of merely theoretical interest: the condition $a_{1,2}, a_{2,1} \gg 1$ which ensures that the whole spectrum lies in $\mathbb{C} \setminus \mathbb{R}$ is too vague to be verifiable. In this section, we will present and semi-rigorously analyze a particular class of parameters that can be studied in more details.

Let $\psi_1 = \psi_2 = \psi$ be a sigmoid function. In our analysis, we will use a number of its properties such as

$$\psi'(0) > 0$$
, $\psi''(0) = 0$, $\psi'''(0) < 0$, $\psi''''(0) = 0$.

Let R > 0 be the ratio of parameters

$$\frac{a_{1,2}}{b_{1,2}} = \frac{a_{2,1}}{b_{2,1}} = \frac{a_{2,2}}{b_{2,2}} = R$$

and suppose that the ratio $a_{1,1}/b_{1,1}$ is perturbed slightly

$$\frac{a_{1,1}}{b_{1,1}}=R+\varepsilon,\quad \varepsilon\in\mathbb{R}.$$

Next, assume that the parameters $b_{i,j}$ are equal

$$b_{1,1} = b_{1,2} = b_{2,1} = b, \quad b > 0,$$

except for $b_{2,2}$ that is smaller:

$$b_{2,2} = vb$$
, $0 < v < 1$.

Lemma 4. For all small values of ε , system (1) has constant solutions

$$u_0 = 2\frac{\psi(0)}{\sigma_*} \left(1 + 2R \frac{\psi'(0)}{\sigma_*} \right) \varepsilon + \mathcal{O}(\varepsilon^2),$$

$$4R \frac{\psi(0)}{\sigma_*} \frac{\psi'(0)}{\sigma_*} = 2 \frac{2}{\sigma_*} \frac{2}{\sigma$$

$$v_0 = 4R \frac{\psi(0)}{\sigma_*} \frac{\psi'(0)}{\sigma_*} \varepsilon + \mathscr{O}(\varepsilon^2).$$

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Proof. At $\varepsilon = 0$, we have

$$\sigma_* u_0 = \sigma_* v_0 = 2R(\psi(u_0) - \psi(v_0))$$

and $u_0 = v_0 = 0$. By the inverse function theorem there is a smooth branch of solutions $(u_0(\varepsilon), v_0(\varepsilon))$ whose first derivatives satisfy the equation

$$\begin{bmatrix} \sigma_* - 2R\psi'(0) & 2R\psi'(0) \\ -2R\psi'(0) & \sigma_* + 2R\psi'(0) \end{bmatrix} \begin{bmatrix} \frac{du_0}{d\varepsilon}(0) \\ \frac{dv_0}{d\varepsilon}(0) \end{bmatrix} = \begin{bmatrix} 2\psi(0) \\ 0 \end{bmatrix}.$$

Note that the Jacobian matrix is full-rank for all admissible values of R, σ_* , and $\psi'(0)$, being a sum of a scalar matrix and a rank-1 matrix. Its inverse can be computed, for example, with the Sherman–Morrison formula.

A. Spectrum

By the properties of sigmoid functions,

$$\psi'(u_0) = \psi'(0) + \mathscr{O}(\varepsilon^2). \quad \psi'(v_0) = \psi'(0) + \mathscr{O}(\varepsilon^2).$$

Then

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$$\begin{split} \hat{Q}_{1,1}(\xi) &= 2\psi'(0)(R+\varepsilon)\frac{b^2}{b^2+\xi^2} + \mathcal{O}(\varepsilon^2), \\ \hat{Q}_{1,2}(\xi) &= \hat{Q}_{2,1}(\xi) = 2\psi'(0)R\frac{b^2}{b^2+\xi^2} + \mathcal{O}(\varepsilon^2), \\ \hat{Q}_{2,2}(\xi) &= 2\psi'(0)R\frac{v^2b^2}{v^2b^2+\xi^2} + \mathcal{O}(\varepsilon^2). \end{split}$$

Recall that the characteristic values are

$$\mu(\xi) = \frac{1}{2} \left(\hat{Q}_{1,1}(\xi) - \hat{Q}_{2,2}(\xi) \right) \pm \pm \frac{1}{2} \sqrt{\left(\hat{Q}_{1,1}(\xi) + \hat{Q}_{2,2}(\xi) \right)^2 - 4 \hat{Q}_{1,2}(\xi) \hat{Q}_{2,1}(\xi)}.$$

Consider the expression under the square root

$$\begin{split} \left(\hat{Q}_{1,1}(\xi) + \hat{Q}_{2,2}(\xi)\right)^2 - 4\hat{Q}_{1,2}(\xi)\hat{Q}_{2,1}(\xi) = \\ &= \left(\frac{2\psi'(0)Rb^2}{(b^2 + \xi^2)}\right)^2 \left[\left(1 + \nu^2 \frac{b^2 + \xi^2}{\nu^2b^2 + \xi^2}\right)^2 + \\ &\quad + 2\left(1 + \nu^2 \frac{b^2 + \xi^2}{\nu^2b^2 + \xi^2}\right)\varepsilon - 4R^2\right] + \mathcal{O}(\varepsilon^2). \end{split}$$

It has the same sign as

$$\left(1 + v^2 \frac{b^2 + \xi^2}{v^2 b^2 + \xi^2}\right) - \left(2 - \frac{\varepsilon}{R} + \mathscr{O}(\varepsilon^2)\right) =
= \frac{\varepsilon}{R} - (1 - v^2) \frac{\xi^2}{v^2 b^2 + \xi^2} + \mathscr{O}(\varepsilon^2).$$

Since $\frac{\xi^2}{v^2b^2+\xi^2}$ is monotonically increasing as a function of ξ^2 , there is a value

$$\xi_0^2 = \frac{v^2 b^2}{1 - v^2} \frac{\varepsilon}{R} + \mathscr{O}(\varepsilon^2),$$

if $0<\varepsilon/R<1-\nu^2$, such that the expression under the square root is positive for $0\le\xi^2<\xi_0^2$ and negative for $\xi^2>\xi_0^2$. It follows that

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$$\mu(\xi) \in egin{cases} \mathbb{R}, & 0 \leq \xi^2 < \xi_0^2, \ \mathbb{C} \setminus \mathbb{R}, & \xi^2 > \xi_0^2. \end{cases}$$

If ε < 0, the whole spectrum lies in $\mathbb{C} \setminus \mathbb{R}$ and we can apply Theorem 1 to show that

$$\operatorname{Re}\mu(\xi) < \operatorname{Re}\mu(\xi_*), \quad \xi \neq \pm \xi_*$$

when $\varepsilon + \mathcal{O}(\varepsilon^2) > R(v^2 - 1)$.

Suppose $\varepsilon \geq 0$. Note that even though the spectrum is not completely complex, we can still apply Theorem 1. Indeed, the frequency ξ_* equals

$$\xi_*^2 = vb^2 \left(1 - \frac{1+v}{2(1-v)} \frac{\varepsilon}{R} \right) + \mathcal{O}(\varepsilon^2)$$

and clearly $\xi_*^2 > \xi_0^2$. Then $\text{Re}\mu(\xi) < \text{Re}\mu(\xi_*)$ for all $\xi^2 > \xi_0^2$, $\xi \neq \pm \xi_*$, provided that

$$v^2 < 1 + \frac{\varepsilon}{R} + \mathcal{O}(\varepsilon^2) < 2 - v^2 < \frac{1}{v^2}.$$

Under the same conditions, $\hat{Q}_{1,1}(\xi) - \hat{Q}_{2,2}(\xi)$ is growing on $(0,\xi_0)$ and

$$\hat{Q}_{1,1}(\xi) - \hat{Q}_{2,2}(\xi) \le 4\psi'(0)\varepsilon + \mathcal{O}(\varepsilon^2), \quad 0 \le \xi < \xi_0.$$

Using monotonicity properties we can also derive that

$$\begin{split} \left(\hat{Q}_{1,1}(\xi) + \hat{Q}_{2,2}(\xi)\right)^2 - 4\hat{Q}_{1,2}(\xi)\hat{Q}_{2,1}(\xi) < \\ < \left(\hat{Q}_{1,1}(0) + \hat{Q}_{2,2}(0)\right)^2 - 4\hat{Q}_{1,2}(\xi_0)\hat{Q}_{2,1}(\xi_0) = \\ = \left(4\psi'(0)R\right)^2 \frac{1 + v^2}{1 - v^2} \frac{\varepsilon}{R} + \mathscr{O}(\varepsilon^2) \end{split}$$

on $(0, \xi_0)$. We thus get an upper bound

$$\mu(\xi) < 2\psi'(0)R\sqrt{\frac{1+\nu^2}{1-\nu^2}\frac{\varepsilon}{R}} + \mathscr{O}(\varepsilon), \quad 0 \leq \xi < \xi_0,$$

which shows that $\mu(\xi)={\rm Re}\mu(\xi)<{\rm Re}\mu(\xi_*)$ for $\xi^2<\xi_0^2$. As a result,

$$\operatorname{Re}\mu(\xi) < \operatorname{Re}\mu(\xi_*), \quad \xi \neq \pm \xi_*.$$

B. Hopf bifurcation

For the Hopf bifurcation to occur, we must have

$$\mu(\xi_*) = \sigma_* + i\omega_*, \quad \omega_* \neq 0.$$

Straightforward calculations show that

$$\begin{split} \operatorname{Re}\mu(\xi_*) &= \frac{\psi'(0)R}{1+\nu} \left(1-\nu+\frac{\varepsilon}{R}\right) + \mathscr{O}(\varepsilon^2), \\ \operatorname{Im}\mu(\xi_*) &= \frac{\psi'(0)R}{1+\nu} \left(\sqrt{(1-\nu)(3+\nu)} - \frac{\varepsilon}{R}\right) + \mathscr{O}(\varepsilon^2), \end{split}$$

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whence we get the bifurcation condition

$$\sigma_* = \frac{\psi'(0)R}{1+\nu} \left(1-\nu + \frac{\varepsilon}{R}\right) + \mathcal{O}(\varepsilon^2).$$

In Figure 1, we show Hopf bifurcation curves in the (R, v)parameter space for different values of $\sigma_*/\psi'(0)$.

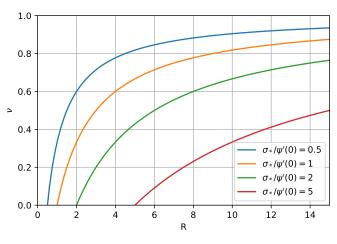


FIG. 1: Curves in the (R, ν) parameter space where Hopf bifurcation occurs for different values of $\sigma_*/\psi'(0)$ and $\varepsilon = 0$.

C. Normal form

Let σ be a small perturbation of the critical parameter value σ_* . Then the real part of the velocity of the critical characteristic value is negative

$$\operatorname{Re} \gamma = \left. \left(\frac{d}{d\sigma} \operatorname{Re} \lambda(\xi_*(0); \sigma) \right) \right|_{\sigma=0} = -1 + \mathscr{O}(\varepsilon^2) < 0$$

and the condition $\sigma \text{Re} \gamma > 0$ transforms into $\sigma < 0$.

Consider matrices $M_k(0)$. It follows from our derivations

$$\begin{split} M_1(0) &= \begin{bmatrix} \psi'(0) + \mathscr{O}(\varepsilon^2) & 0 \\ 0 & -\psi'(0) + \mathscr{O}(\varepsilon^2) \end{bmatrix}, \\ M_2(0) &= \frac{1}{2} \begin{bmatrix} \psi'''(0)u_0 + \mathscr{O}(\varepsilon^2) & 0 \\ 0 & -\psi'''(0)v_0 + \mathscr{O}(\varepsilon^2) \end{bmatrix}, \\ M_3(0) &= \frac{1}{6} \begin{bmatrix} \psi'''(0) + \mathscr{O}(\varepsilon^2) & 0 \\ 0 & -\psi'''(0) + \mathscr{O}(\varepsilon^2) \end{bmatrix}. \end{split}$$

In particular, $||M_2(0)|| = \mathcal{O}(\varepsilon)$. Note also that

$$\mathscr{P}_1 = \begin{bmatrix} \frac{2R}{1+\nu} \left(1 + \frac{2-\nu}{2-2\nu} \frac{\varepsilon}{R} \right) & \frac{2R}{1+\nu} \left(1 + \frac{\nu}{2-2\nu} \frac{\varepsilon}{R} \right) \\ \frac{2R}{1+\nu} \left(1 + \frac{\nu}{2-2\nu} \frac{\varepsilon}{R} \right) & \frac{2R\nu}{1+\nu} \left(1 + \frac{1}{2-2\nu} \frac{\varepsilon}{R} \right) \end{bmatrix} + \mathscr{O}(\varepsilon^2).$$

Then the eigenvector ζ_1 equals

$$\zeta_1 = \frac{\psi'(0)R}{1+\nu} \left[\frac{2+\frac{\nu}{1-\nu}\frac{\epsilon}{R}}{1+\nu-i\sqrt{(1-\nu)(3+\nu)}+(\frac{1}{1-\nu}-i)\frac{\epsilon}{R}} \right] + \mathscr{O}(\epsilon^2). \text{ and periodic traveling waves are asymptotically stable for all parameters of this class as long as } |\epsilon| \text{ is not too large.}$$

For simplicity, we will only write the $\mathcal{O}(1)$ terms in what follows. In that case, we have

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$$\zeta_1^* = \frac{1+\nu}{\psi'(0)R} \begin{bmatrix} \frac{1}{4} + \frac{\mathrm{i}}{4} \frac{1+\nu}{\sqrt{(1-\nu)(3+\nu)}} \\ -\frac{\mathrm{i}}{2} \frac{1}{\sqrt{(1-\nu)(3+\nu)}} \end{bmatrix} + \mathscr{O}(\varepsilon).$$

As $\|\zeta_1\| = \mathcal{O}(1)$ and $\|M_2(0)\| = \mathcal{O}(\varepsilon)$, the vectors h_i , j =1,..., 5 from Appendix A are all small, $||h_i|| = \mathcal{O}(\varepsilon)$. Based on the property of the Hadamard product,

$$||x\odot y|| \le ||x|| ||y||,$$

we can conclude that

$$\begin{split} \|M_2(0)(\overline{\zeta_1}\odot h_1)\| &= \mathscr{O}(\varepsilon^2), \quad \|M_2(0)(\zeta_1\odot h_2)\| = \mathscr{O}(\varepsilon^2), \\ \|M_2(0)(\zeta_1\odot h_3)\| &= \mathscr{O}(\varepsilon^2), \quad \|M_2(0)(\overline{\zeta_1}\odot h_4)\| = \mathscr{O}(\varepsilon^2), \\ \|M_2(0)(\zeta_1\odot h_5)\| &= \mathscr{O}(\varepsilon^2). \end{split}$$

As a consequence, the normal form coefficients take a much simpler form:

$$egin{aligned} c_1 &= rac{1}{2l_*} \left\langle M_3(0)(\zeta_1 \odot \zeta_1 \odot \overline{\zeta_1}), \mathscr{P}_1^T \zeta_1^*
ight
angle + \mathscr{O}(arepsilon^2), \ c_2 &= rac{1}{l_*} \left\langle M_3(0)(\zeta_1 \odot \overline{\zeta_1} \odot \zeta_1), \mathscr{P}_1^T \zeta_1^*
ight
angle + \mathscr{O}(arepsilon^2). \end{aligned}$$

Recall that for periodic traveling waves to be asymptotically stable, three conditions need to be met:

$$Rec_1 < 0$$
, $Re(c_1 + c_2) < 0$, $Re(c_1 - c_2) > 0$.

In our case we have

$$Re(c_1 + c_2) = 3Rec_1 + \mathcal{O}(\varepsilon^2),$$

$$Re(c_1 - c_2) = -Rec_1 + \mathcal{O}(\varepsilon^2).$$

This means that the stability of periodic traveling waves depends on a single condition

$$Rec_1 < 0$$
.

Direct calculations give

$$c_1 = \frac{1}{3l_*} \left(\frac{\psi'(0)R}{1+\nu} \right)^3 \frac{\psi'''(0)}{\psi'(0)} \left(1 - \nu + i\sqrt{(1-\nu)(3+\nu)} \right) + \mathcal{O}(\varepsilon).$$

Therefore, since v < 1, $\psi'(0) > 0$, and $\psi'''(0) < 0$, the real part is negative,

$$\operatorname{Re} c_1 = \frac{1}{3l_*} \left(\frac{\psi'(0)R}{1+\nu} \right)^3 \frac{\psi'''(0)}{\psi'(0)} (1-\nu) + \mathcal{O}(\varepsilon) < 0,$$

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TABLE II: Example of parameters.

<i>a</i> ₁₁	a_{12}	a ₂₁	a_{22}	b ₁₁	b ₁₂	b ₂₁	b ₂₂	σ_*
3.05	3.00	3.00	0.30	1.00	1.00	1.00	0.10	1.00

V. NUMERICAL EXAMPLES

For the numerical examples, we choose $\psi_1 = \psi_2 = \psi$ with

$$\psi(u) = \frac{2}{\pi} \arctan(0.6782u) + 1.$$

Consider the parameters from Table II. The constant solution is $u_0 = 0.404$, $v_0 = 0.287$. Upon linearization about (u_0, v_0) we can find that the critical spatial frequency is $\xi_* = 0.318$ and that $\lambda(\xi_*) = \pm 1.86$ i (see Figure 2). The computation of

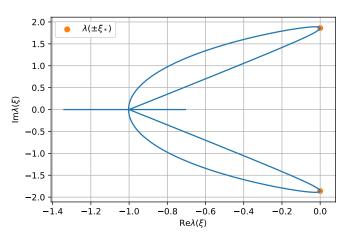


FIG. 2: Distribution of characteristic values $\lambda(\xi)$ on the complex plain for the parameters from Table II.

the normal form gives

$$c_1 \approx -0.0182 - 0.0386i$$
, $c_2 \approx -0.0365 - 0.0773i$,

which means that the periodic traveling waves are asymptotically stable.

For numerical verification, we ran numerical simulations on the interval $(0,2\pi/\xi_*)$ with periodic boundary conditions. We computed the convolutions using the fast Fourier transform and the convolution theorem.

In Figure 3, we present numerical simulations for box-like initial states:

$$u_0(x) = 2\chi\{x < \pi/\xi_*\} - 1, \quad v_0(x) = 2\chi\{x < \pi/\xi_*\} - 1.$$

Here, $\chi\{\Omega\}$ is the indicator function of the set Ω such that $\chi\{\Omega\}(x) = 1$ if $x \in \Omega$ and $\chi\{\Omega\}(x) = 0$ if $x \notin \Omega$. In the formulas above, both u_0 and v_0 are equal to 1 on $(0, \pi/\xi_*)$ and to -1 on $(\pi/\xi_*, 2\pi/\xi_*)$. As the simulations show, the solution retains the symmetry of the initial state and becomes a standing wave. However, if we shrink or enlarge the positive half to $(0, \pi/\xi_* \pm \varepsilon)$ for one of the components (let it be v_0), the

solution turns into a periodic traveling wave since the initial state breaks the reflection symmetry of the system.

It is also important to understand how the system evolves from zero initial state in the presence of a forcing term. Namely, we are interested in time-oscillating point sources like

$$f(x,t) = \delta(x - x_0)\sin(t),$$

where $\delta(x)$ is the δ -function. So we want to solve the following initial value problem:

$$\frac{\partial u}{\partial t} = P_{1,1} * \psi_1(u) - P_{1,2} * \psi_2(v) - \sigma u + f_1(x,t),
\frac{\partial v}{\partial t} = P_{2,1} * \psi_1(u) - P_{2,2} * \psi_2(v) - \sigma v + f_2(x,t),
u(x,0) = 0, \quad v(x,0) = 0.$$

In Figure 4, we use

$$f_i(x,t) = 0.1\delta(x-x_i)\sin(\omega t), \quad j = 1, 2,$$

as the forcing terms. If $x_1 = x_2$, that is the point sources are placed at the exactly same positions, the zero initial state evolves into a standing wave. Once again, this is because the reflection symmetry is not broken. Meanwhile, should the point sources be placed at different (possibly very close) positions, $x_1 \neq x_2$, the solution is a periodic traveling wave, which propagates in the direction determined by the relative position of the two sources.

In Figure 5, we explore the case where the forcing terms have different oscillation frequencies:

$$f_1(x,t) = 0.1\delta(x - \pi/2\xi_*)\sin(At),$$

 $f_2(x,t) = 0.1\delta(x)\sin(t).$

We observed that for a fixed distance between the forcing terms, the propagation direction of the traveling waves depends on the ratio of the forcing term frequencies A>0. Namely, there is a threshold value A_* such that waves propagate in different directions for $A< A_*$ and $A>A_*$. Moreover, as $A\to A_*$, the solution spends more time in the form of a standing wave, which suggests that at $A=A_*$ the solution is a standing wave.

VI. DISCUSSION

With the development of modern technologies it has been possible to demonstrate the existence of periodic traveling waves in cortex during many human activities⁷. These propagating waves help explain brain dynamics observed at the population level³⁰. For example, Huang et al.^{31,32} studied brain wave dynamics for years using voltage-sensitive dye imaging and demonstrated the presence of spiral dynamics in the intact mammalian brain during pharmacologically induced oscillations as well as during sleep-like states. These authors showed that traveling waves are emergent properties which can influence not only brain oscillation frequencies and spatial coherence but they can also modulate their amplitude. As a result

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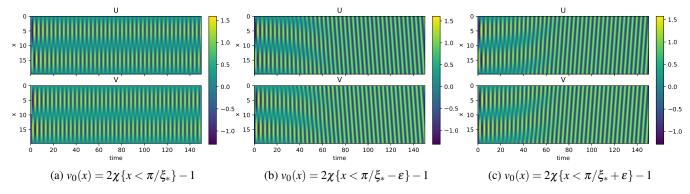


FIG. 3: Numerical simulation with the initial state $u_0(x) = 2\chi\{x < \pi/\xi_*\} - 1$ and $\sigma = 0.95$. Level lines of the solution u(x,t) (upper row) and v(x,t) (lower row) are shown for three different initial states $v_0(x)$. Standing waves are observed for the symmetric initial state (a), periodic traveling waves for the asymmetric initial states (b,c).

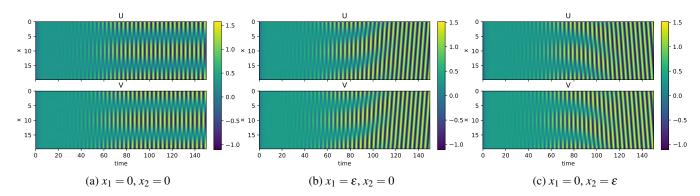


FIG. 4: Numerical simulation with the forcing terms $f_j(x,t) = 0.1\delta(x-x_j)\sin(\omega t)$, j=1,2, and $\sigma=0.92$. Level lines of the solution u(x,t) (upper row) and v(x,t) (lower row) are shown for three different forcing terms. Standing waves are observed for the symmetric forcing term (a), periodic traveling waves for the asymmetric forcing terms (b,c).

these waves «may contribute to both normal cortical processing and to pathological patterns of activity such as those found in epilepsy» (Ref. 30, P. 1591). Interestingly Stead et al.³³ observed «microseizures» (spiral waves) on isolated research microelectrodes in both epileptic patients (in regions prone to generate epileptic seizures) but also in control subjects. They conclude that only «the density of these phenomena distinguishes normal from epileptic brain» (Ref. 33, P. 2789).

Although the functional role of traveling waves is not completely elucidated yet, new techniques based on targeted neuromodulation of traveling waves using electric fields are already being explored as potential therapeutic strategies. For example, Whalen et al.² were able to control the modulation, suppression and prevention or to increase propagation velocity of spreading depression with DC electrical fields in brain slices thus demonstrating «the potential feasibility of electrical control and prevention of spreading depression» (Ref. 2, P. 8769). Using coupled oscillators Ermentrout and Kleinfeld³⁴ also studied electrical waves in the cortex and speculated as to their computational role.

Neural field models provide an appropriate tool to study these waves but the results differ for different models. A single equation with symmetric connectivity functions for activating and inhibitory signals does not have solutions in the form of periodic traveling waves. In the case of asymmetric functions, such solutions exist, they are stable, and solutions with different frequencies (speeds) are observed for the same values of parameters³⁵. The last property is important for the understanding of the existence of the waves with different frequencies during brain activity.

A single equation with symmetric connectivity functions can also manifest periodic waves in the presence of time delay³⁶. The periodic waves bifurcate from the stationary, spatially homogeneous solution being unstable, and they become stable if time delay exceeds some critical value. As before, the coexistence of waves with different frequencies is observed. Another important characterization of periodic waves related to the choice of initial conditions should also be taken into account. We will return to this question below.

The models with two different neuron populations, excitatory and inhibitory, can also describe the propagation of periodic waves 17,18,21,22,28 . In this work we considered a particular case of the model Ref. 28 without time delay and with symmetric connectivity functions. Nonlinear bifurcation analysis shows two types of solutions: standing oscillating solutions with the behavior qualitatively similar to the product of two periodic functions, $cos(\omega t) cos(kx)$, and periodic traveling waves similar to $cos(\omega t + kx)$. Their stability depends on

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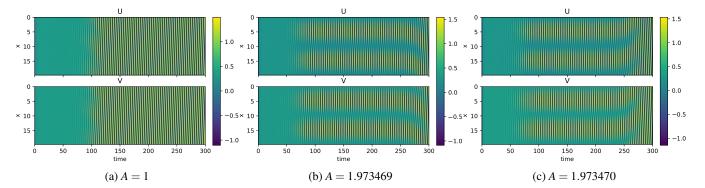


FIG. 5: Numerical simulation with the forcing terms $f_1(x,t) = 0.1\delta(x - \pi/2\xi_*)\sin(At)$, $f_2(x,t) = 0.1\delta(x)\sin(t)$, and $\sigma = 0.92$. Level lines of the solution u(x,t) (upper row) and v(x,t) (lower row) are shown for three different forcing terms. The traveling waves move to the right for A below the threshold value (a,b) and to the left for A above the threshold value (c).

the values of parameters, and they are mutually exclusive.

From the biological point of view we are interested not only in the existence and stability of periodic traveling waves but also in their initiation, that is, under what conditions (initial, boundary, etc), the solution will approach the wave. This question can be potentially important to determine the role of epicenters (hubs) in the functioning of brain connectome. The epicenters receive the information from other epicenters or from the external body organs, they process this information and send it further downstream. Assuming that the synchronization of epicenters is effectuated by cortical waves, we can suppose that the waves are initiated locally by the activated epicenter. In the case of a single equation, such initiation appears to be possible for asymmetric connectivity function but not for symmetric connectivity functions and time delay³⁶. The latter requires the initial conditions close to the wave on the whole space interval. The model considered in this work admits wave initiation by local forcing terms (additional terms in the equations). If the forcing term is symmetric, then standing oscillating solutions are observed. In the case of asymmetric forcing term, under the parameters for which the periodic wave is stable, solution converges to the wave.

Let us note that the direction of propagation is one of the important parameters characterizing cortical waves. We show that this direction is determined by the forcing term. In the 2D setting we expect that we can obtain any direction between 0 and 2π . Hence, the epicenters can exchange directed signals ¹². Wave initiation by local forcing terms is observed in the case of brain stimulation with electrodes ³⁷. External brain stimulation can be used in order to restore wave propagation in a partially damaged tissue ^{35,36}. Recent perspectives propose to control brain traveling waves in human subjects by stimulating preidentified nodes using electrical stimulation. One of these initiatives propose to modulate the activity of cortical epicenters of the language network to treat chronic conditions such as post stroke aphasia ^{38,39}.

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DATA AVAILABILITY

Data sharing is not applicable to this article as no new data were created or analyzed in this study.

Appendix A: Computation of the normal form coefficients

To compute the cubic coefficients $c_1 \in \mathbb{C}$ and $c_2 \in \mathbb{C}$, we introduce a function

$$\tilde{F}(w,\sigma) = F(w,\sigma) + L(\sigma)w - L(0)w - \sigma L'(0)w$$

and use the formulas given in Ref. 29. The computation involves second and third-order Fréchet derivatives of $\tilde{F}(w,\sigma)$ ($\tilde{F}_{ww}(0,0)$ and $\tilde{F}_{www}(0,0)$) at w=0 and $\sigma=0$:

$$\begin{split} \tilde{F}_{2}(w_{1},w_{2}) &= \sum_{n \in \mathbb{Z}} \mathscr{P}_{n} M_{2}(0) (w_{1} \odot w_{2})_{n} e_{n}, \\ \tilde{F}_{3}(w_{1},w_{2},w_{3}) &= \sum_{n \in \mathbb{Z}} \mathscr{P}_{n} M_{3}(0) (w_{1} \odot w_{2} \odot w_{3})_{n} e_{n}. \end{split}$$

Nonlinear analysis of periodic waves in a neural field model

It remains to evaluate the following expressions:

$$\begin{split} g_1 &= \frac{1}{2} (2\mathrm{i}\omega_* - L(0))^{-1} \tilde{F}_2(q_1, q_1), \\ g_2 &= -L(0)^{-1} \tilde{F}_2(\overline{q_1}, q_1), \\ g_3 &= (2\mathrm{i}\omega_* - L(0))^{-1} \tilde{F}_2(q_1, q_3), \\ g_4 &= -L(0)^{-1} \tilde{F}_2(q_1, \overline{q_3}), \\ g_5 &= -L(0)^{-1} \tilde{F}_2(q_3, \overline{q_3}). \end{split}$$

We treat them one by one:

1.

$$\begin{split} \tilde{F}_2(q_1, q_1) &= \frac{1}{\sqrt{l_*}} \mathscr{P}_2 M_2(0) (\zeta_1 \odot \zeta_1) e_2, \\ g_1 &= h_1 e_2, \quad h_1 \in \mathbb{C}^2, \\ (2i\omega_* - \mathscr{P}_2 M_1(0)) h_1 &= \frac{1}{2\sqrt{l_*}} \mathscr{P}_2 M_2(0) (\zeta_1 \odot \zeta_1); \end{split}$$

2.

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$$\begin{split} \tilde{F}_2(\overline{q_1}, q_1) &= \frac{1}{\sqrt{l_*}} \mathscr{P}_0 M_2(0) (\overline{\zeta_1} \odot \zeta_1) e_0, \\ g_2 &= h_2 e_0, \quad h_2 \in \mathbb{C}^2, \\ (\mathscr{P}_0 M_1(0)) h_2 &= -\frac{1}{\sqrt{l_*}} \mathscr{P}_0 M_2(0) (\overline{\zeta_1} \odot \zeta_1); \end{split}$$

minear analysis of periodic waves in a neural net

$$ilde{F}_{2}(q_{1},q_{3}) = rac{1}{\sqrt{l_{*}}}\mathscr{P}_{2}M_{2}(0)(\zeta_{1}\odot\overline{\zeta_{1}})e_{2}, \ g_{3} = h_{3}e_{2}, \quad h_{3} \in \mathbb{C}^{2}, \ (2\mathrm{i}\omega_{*} - \mathscr{P}_{2}M_{1}(0))h_{3} = rac{1}{\sqrt{l_{*}}}\mathscr{P}_{2}M_{2}(0)(\zeta_{1}\odot\overline{\zeta_{1}});$$

11

4.
$$\begin{split} \tilde{F}_{2}(q_{1},\overline{q_{3}}) &= \frac{1}{\sqrt{I_{*}}}\mathscr{P}_{0}M_{2}(0)(\zeta_{1}\odot\zeta_{1})e_{0}, \\ g_{4} &= h_{4}e_{0}, \quad h_{4} \in \mathbb{C}^{2}, \\ (\mathscr{P}_{0}M_{1}(0))h_{4} &= -\frac{1}{\sqrt{I_{*}}}\mathscr{P}_{0}M_{2}(0)(\zeta_{1}\odot\zeta_{1}); \end{split}$$

$$\begin{split} \tilde{F}_2(q_3,\overline{q_3}) &= \frac{1}{\sqrt{l_*}} \mathscr{P}_0 M_2(0) (\overline{\zeta_1} \odot \zeta_1) e_0, \\ g_5 &= h_5 e_0, \quad h_5 \in \mathbb{C}^2, \\ (\mathscr{P}_0 M_1(0)) h_5 &= -\frac{1}{\sqrt{l_*}} \mathscr{P}_0 M_2(0) (\overline{\zeta_1} \odot \zeta_1). \end{split}$$

We also need two terms with third-order derivatives:

$$ilde{F}_3(q_1,q_1,\overline{q_1}) = rac{1}{l_*} \mathscr{P}_1 M_3(0) (\zeta_1 \odot \zeta_1 \odot \overline{\zeta_1}) e_1, \ ilde{F}_3(q_1,q_3,\overline{q_3}) = rac{1}{l_*} \mathscr{P}_1 M_3(0) (\zeta_1 \odot \overline{\zeta_1} \odot \zeta_1) e_1.$$

The normal form coefficients are then given by the following formulas²⁹:

$$c_{1} = \frac{1}{2} \langle \tilde{F}_{3}(q_{1}, q_{1}, \overline{q_{1}}) + 2\tilde{F}_{2}(\overline{q_{1}}, g_{1}) + 2\tilde{F}_{2}(q_{1}, g_{2}), q_{1}^{*} \rangle_{X},$$

$$c_{2} = \langle \tilde{F}_{3}(q_{1}, q_{3}, \overline{q_{3}}) + \tilde{F}_{2}(\overline{q_{3}}, g_{3}) + \tilde{F}_{2}(q_{3}, g_{4}) + \tilde{F}_{2}(q_{1}, g_{5}), q_{1}^{*} \rangle_{X}.$$

It follows that

3.

5.

$$\begin{split} c_1 &= \left\langle \frac{1}{2l_*} \mathscr{P}_1 M_3(0) (\zeta_1 \odot \zeta_1 \odot \overline{\zeta_1}) + \frac{1}{\sqrt{l_*}} \mathscr{P}_1 M_2(0) (\overline{\zeta_1} \odot h_1) + \frac{1}{\sqrt{l_*}} \mathscr{P}_1 M_2(0) (\zeta_1 \odot h_2), \zeta_1^* \right\rangle \\ &= \frac{1}{\sqrt{l_*}} \left\langle \frac{1}{2\sqrt{l_*}} M_3(0) (\zeta_1 \odot \zeta_1 \odot \overline{\zeta_1}) + M_2(0) (\overline{\zeta_1} \odot h_1) + M_2(0) (\zeta_1 \odot h_2), \mathscr{P}_1^T \zeta_1^* \right\rangle, \\ c_2 &= \left\langle \frac{1}{l_*} \mathscr{P}_1 M_3(0) (\zeta_1 \odot \overline{\zeta_1} \odot \zeta_1) + \frac{1}{\sqrt{l_*}} \mathscr{P}_1 M_2(0) (\zeta_1 \odot h_3) + \frac{1}{\sqrt{l_*}} \mathscr{P}_1 M_2(0) (\overline{\zeta_1} \odot h_4) + \frac{1}{\sqrt{l_*}} \mathscr{P}_1 M_2(0) (\zeta_1 \odot h_5), \zeta_1^* \right\rangle \\ &= \frac{1}{\sqrt{l_*}} \left\langle \frac{1}{\sqrt{l_*}} M_3(0) (\zeta_1 \odot \overline{\zeta_1} \odot \zeta_1) + M_2(0) (\zeta_1 \odot h_3) + M_2(0) (\overline{\zeta_1} \odot h_4) + M_2(0) (\zeta_1 \odot h_5), \mathscr{P}_1^T \zeta_1^* \right\rangle. \end{split}$$

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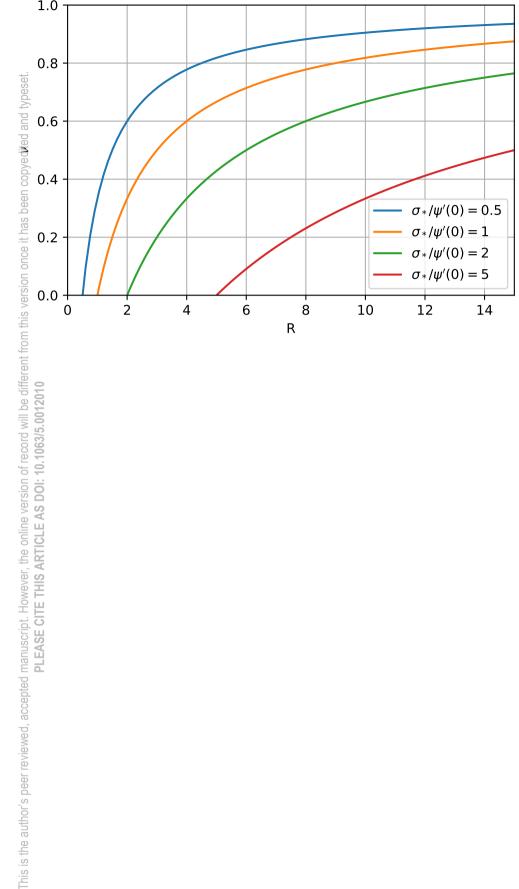
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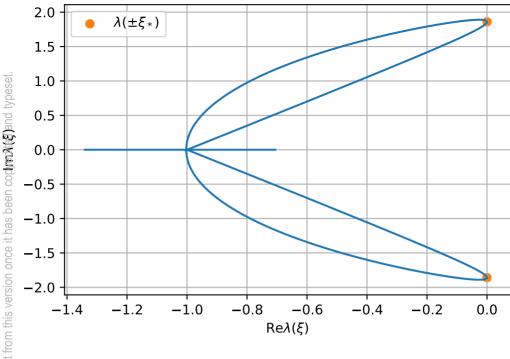


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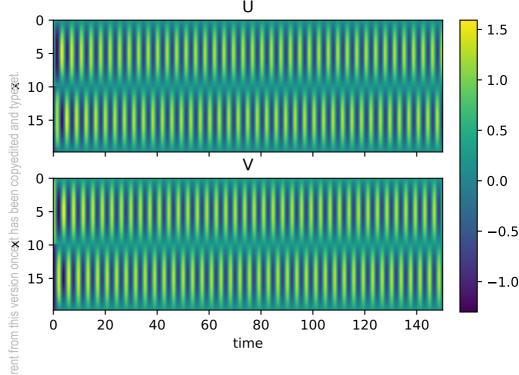
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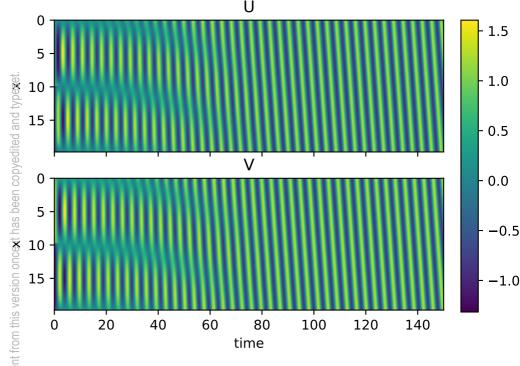
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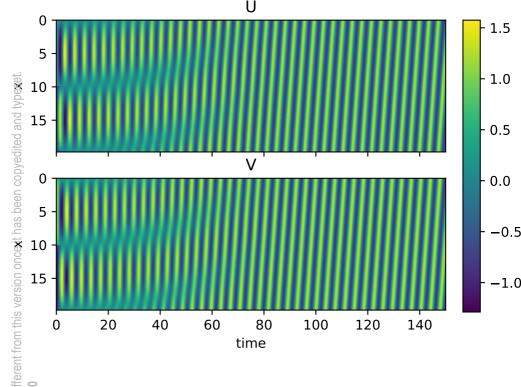




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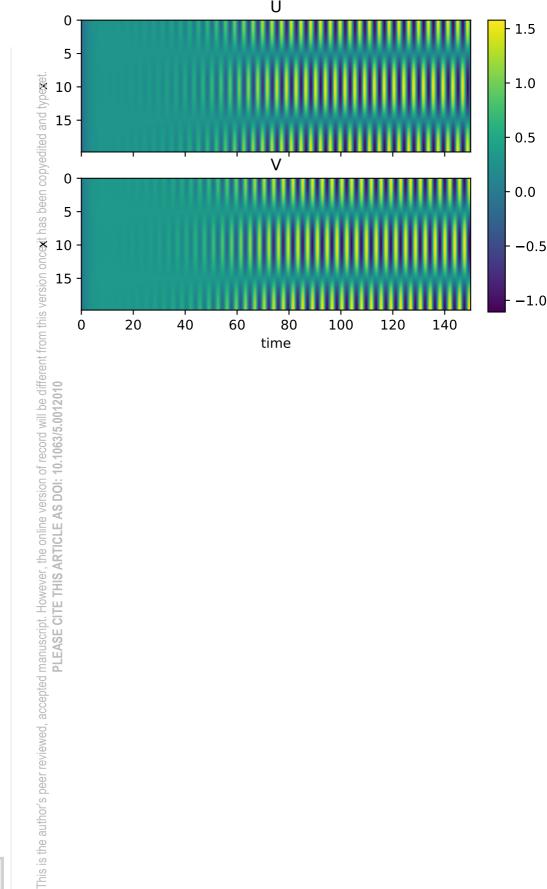




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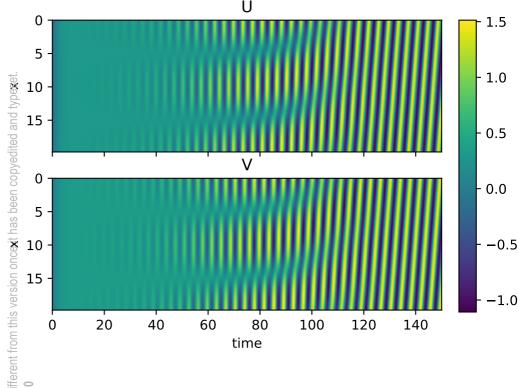
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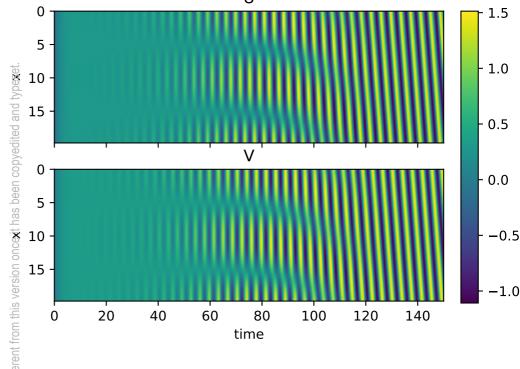




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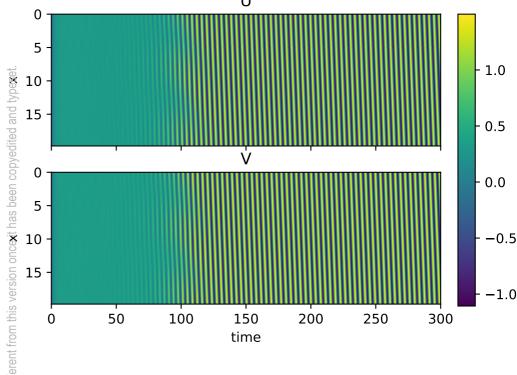




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