Robustness of Synchrony in Complex Networks and Generalized Kirchhoff Indices

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In network theory, a question of prime importance is how to assess network vulnerability in a fast and reliable manner. With this issue in mind, we investigate the response to external perturbations of coupled dynamical systems on complex networks. We find that for specific, non averaged perturbations, the response of synchronous states depends on the eigenvalues of the stability matrix of the unperturbed dynamics, as well as on its eigenmodes via their overlap with the perturbation vector. Once averaged over properly defined ensembles of perturbations, the response is given by new graph topological indices, which we introduce as \textit{generalized Kirchhoff indices}. These findings allow for a fast and reliable method for assessing the specific or average vulnerability of a network against changing operational conditions, faults or external attacks.

\textbf{Introduction.} Graph theory profoundly impacts many fields of human knowledge, including social and natural sciences, communication technology and electrical engineering, and information sciences and cybernetics \cite{1}. Graphs allow for a convenient modelingization of complex systems where their structure defines the couplings between the system’s individual components, each of them with its own internal dynamics. The resulting coupled differential equations determine the system dynamics and its steady state solutions. Of particular interest is to predict the behavior of the system when it is perturbed away from steady-state, for instance when an electric power plant goes offline in an operating power grid or when a line is cut and information has to be redirected in a communication network. An issue of key importance for network security is how to fast and reliably assess a network’s vulnerability. This is not an easy task: network vulnerability depends on both the system dynamics and the network topology and geometry. It is highly desirable to identify a set of easily computed descriptors that characterize network vulnerability \cite{2}. In this Letter we propose a new family of network descriptors in a two-step approach. We investigate the sensitivity against external perturbations of synchronous states of coupled dynamical systems on complex networks. First, we quantify this sensitivity using performance measures recently introduced in the context of electric power grids \cite{3−5}. Second, by direct calculation of these performance measures, we identify a new class of easily computed topological indices that generally characterize synchrony robustness or fragility under ensemble-averaged perturbations.

Synchronization is ubiquitous \cite{6} in systems of coupled dynamical systems. It follows from the interplay between the internal dynamics of the individual systems and the coupling between them \cite{7−10}. Optimization of synchronization has been investigated from various angles. The synchronous state can be optimal from the point of view of linear stability \cite{11}, the range of parameters that allow synchronization \cite{12−14}, the value that an order parameter takes at synchrony \cite{15} or the volume of the basin of attraction around a stable synchronous fixed point \cite{16−18}. Here we extend these investigations by asking what makes synchronous states more or less fragile against external perturbations. For ensemble-averaged perturbations, the answer is surprisingly simple: synchrony fragility depends on a family of topological indices, which generalize the Kirchhoff index introduced in Ref. \cite{1}. This result is rather general and remains valid for a large class of fragility performance measures quantifying the excursion away from the stable synchronous state, and for rather general synchronizing coupled dynamical systems. Its main restriction is that it applies to not-too-large perturbations, which leave the system inside its original basin of stability.

\textbf{Model and method.} Our analysis focuses on the Kuramoto model \cite{7}

\[ \dot{\theta}_i = P_i - \sum_j b_{ij} \sin(\theta_i - \theta_j), \quad i = 1, \ldots, n, \quad (1) \]

though our results are more general and apply to a wider class of coupled dynamical systems (see Supplemental Material). Eq. (1) models the behavior of a set of \( n \) harmonic oscillators, each with its angle coordinate \( \theta_i \) and its natural frequency \( P_i \), coupled to one another with couplings defined by the weighted adjacency matrix \( b_{ij} \geq 0 \). Kuramoto originally considered identical all-to-all coupling, \( b_{ij} \equiv K/n \) \cite{7}. It was found that for \( K > K_c \), a finite number of oscillators synchronize, with \( \dot{\theta}_i - \dot{\theta}_j = 0 \). This type of frequency synchronization also occurs for nonhomogeneous couplings \( b_{ij} \) defined on a complex network \cite{20}, the case of interest here. Without loss of generality we set \( \sum_i P_i = 0 \), for which the frequency synchronous state has \( \dot{\theta}_i \equiv 0, \forall i \) \cite{21}.

We consider a stable fixed-point solution \( \theta^{(0)} = (\theta_1^{(0)}, \ldots, \theta_n^{(0)}) \) to Eq. (1) with unperturbed natural frequencies \( P^{(0)} \). We then subject this state to a time-dependent perturbation \( P(t) = P^{(0)} + \delta P(t) \), so that angles become time dependent, \( \theta(t) = \theta^{(0)} + \delta \theta(t) \). Lin-
earizing the dynamics of Eq. (1) about \( \theta^{(0)} \), one obtains
\[
\delta \dot{\theta} = \delta P - L(\theta^{(0)}) \delta \theta, \tag{2}
\]
where we introduced the weighted Laplacian matrix
\( L(\theta^{(0)}) \) with matrix elements
\[
\mathbb{L}_{ij} = \begin{cases} 
-b_{ij} \cos(\theta_i^{(0)} - \theta_j^{(0)}), & i \neq j, \\
\sum_k b_{ik} \cos(W_{ij}^{(0)}), & i = j.
\end{cases}
\tag{3}
\]
This Laplacian is minus the stability matrix of the linearized dynamics, and since we consider a stable synchronous state, it is positive semidefinite, with a single eigenvalue \( \lambda_1 = 0 \) with eigenvector \( \mathbf{u}_1 = (1, 1, 1, \ldots, 1)/\sqrt{n} \), and \( \lambda_i > 0, i = 2, 3, \ldots, n \).

The first term on the right-hand side of Eq. (2) perturbs angles away from the synchronous state. To assess the magnitude of this excursion in the spirit of Refs. [3–5] we consider two fragility performance measures
\[
\mathcal{C}_1(T) = \sum_{i} \int_0^T |\delta \theta_i(t)|^2 dt, \tag{4a}
\]
\[
\mathcal{C}_2(T) = \sum_{i} \int_0^T |\dot{\delta \theta}_i(t)|^2 dt. \tag{4b}
\]
Because synchronous states are defined modulo any homogeneous angle shift, the transformation \( \theta_i^{(0)} \rightarrow \theta_i^{(0)} + C \) does not change the synchronous state. Accordingly, only angle shifts with \( \delta \theta_i(t) = 0 \) matter, which is incorporated in the definitions of \( \mathcal{C}_{1,2} \) by subtracting averages
\[
\Delta(t) = n^{-1} \sum_j \delta \theta_j(t) \quad \text{and} \quad \dot{\Delta}(t) = n^{-1} \sum_j \dot{\delta \theta}_j(t).
\]
An alternative procedure is to restrict oneself to perturbations orthogonal to \( \mathbf{u}_1 \) [3–5]. Either procedure ensures, together with the non-negativity of \( \mathbb{L} \), that \( \mathcal{C}_{1,2} \leq \infty \), even when \( T \rightarrow \infty \), if the perturbation is short and weak enough that it leaves the dynamics inside the basin of attraction of \( \theta^{(0)} \). Low values for \( \mathcal{C}_{1,2} \) indicate then that the system absorbs the perturbation with little fluctuations, while large values indicate a temporary fragmentation of the system into independent pieces – \( \mathcal{C}_{1,2} \) measures the coherence of the synchronous state.

We expand angle deviations over the eigenstates \( \mathbf{u}_\alpha \) of \( \mathbb{L} \), \( \delta \theta(t) = \sum \alpha c_\alpha(t) \mathbf{u}_\alpha \), and rewrite Eq. (2) as
\[
\dot{c}_\alpha(t) = \delta P(t) \cdot \mathbf{u}_\alpha - \lambda_\alpha c_\alpha(t), \tag{5}
\]
whose general solution reads
\[
c_\alpha(t) = e^{-\lambda_\alpha t} c_\alpha(0) + e^{-\lambda_\alpha t} \int_0^t dt' e^{\lambda_\alpha t'} \delta P(t') \cdot \mathbf{u}_\alpha. \tag{6}
\]
Being interested in perturbations \( \delta P \) that start at \( t = 0 \), when the system is in the synchronous state with \( \delta \theta(0) = 0 \), we set \( c_\alpha(0) \equiv 0 \). The performance measures of Eqs. (5) are given by
\[
\mathcal{C}_1(T) = \sum_{\alpha \geq 2} \int_0^T c_\alpha^2(t) dt \quad \text{and} \quad \mathcal{C}_2(T) = \sum_{\alpha \geq 2} \int_0^T \dot{c}_\alpha^2(t) dt,
\]
as long as the perturbation is not too large so that the system eventually returns to its initial state. We next introduce generalized Kirchhoff indices in terms of which we express \( \mathcal{C}_{1,2} \) for three different classes of perturbations \( \delta P(t) \).

**Generalized Kirchhoff indices.** The Kirchhoff index originally followed from the definition of the resistance distance in a graph [1]. To a connected graph, one associates an electrical network where each edge is a resistor given by the inverse edge weight in the original graph. The resistance distance is the resistance \( \Omega_{ij} \) between any two nodes \( i \) and \( j \) on the electrical network. The Kirchhoff index is then defined as
\[
K_f = \sum_{i < j} \Omega_{ij}, \tag{7}
\]
where the sum runs over all pairs of nodes in the graph. For a graph with Laplacian \( \mathbb{L} \), it has been shown that \( K_f \) is given by the spectrum \( \{\lambda_\alpha\} \) of \( \mathbb{L} \) as [2, 3, 24]
\[
K_f = n \sum_{\alpha \geq 2} \lambda_\alpha^{-1}. \tag{8}
\]

Up to a normalization prefactor, \( K_f \) gives the mean resistance distance \( \Omega \) over the whole graph. Intuitively, one expects the dynamics of a complex system to depend not only on \( \Omega \), but on the full set \( \{\Omega_{ij}\} \). Higher moments of \( \{\Omega_{ij}\} \) are encoded in generalized Kirchhoff indices \( K_{f_m} \) (see Supplemental Material) which we define as
\[
K_{f_m} = n \sum_{\alpha \geq 2} \lambda_\alpha^{-m}, \tag{9}
\]
for integers \( m \). Below we show that \( \mathcal{C}_{1,2} \) can be expressed as linear combinations of the \( K_{f_m} \)’s corresponding to \( \mathbb{L} \) in Eq. (8). We note that, continued to \( m \in \mathbb{C} \), \( K_{f_m} \) is known as the *spectral zeta function* of \( \mathbb{L} \) [25].

**Dirac delta perturbation.** We first consider\( \delta P(t) = \delta P_0 \tau_0 \delta(t) \) with the Dirac delta-function \( \delta(t) \). Because the perturbation is limited in time, the limit \( T \rightarrow \infty \) can be taken in Eqs. (5)\(^{10}\). One obtains (see Supplemental Material)
\[
\mathcal{C}_1^{\infty} = \sum_{\alpha} \left( \frac{\langle \delta P_0 \cdot \mathbf{u}_\alpha \rangle^2}{2} \right)^2 \lambda_\alpha^{-1}, \tag{10a}
\]
\[
\mathcal{C}_2^{\infty} = \sum_{\alpha} \left( \frac{\langle \delta P_0 \cdot \mathbf{u}_\alpha \rangle^2}{2} \right)^2 \lambda_\alpha. \tag{10b}
\]
Both performance measures depend on the scalar product of the perturbation \( \delta P_0 \) with the eigenmodes \( \mathbf{u}_\alpha \) of \( \mathbb{L} \). Such scalar products occur also when analyzing propagation of disturbances on networks [26]. To get more insight on the typical network response, we define an ensemble of perturbation vectors with \( \langle \delta P_0 \delta P_0 \rangle = \delta_{ij} \langle \delta P_0 \rangle^2 \). Averaging over that ensemble gives
\[
\langle \mathcal{C}_1^{\infty} \rangle = \frac{\langle \delta P_0^2 \rangle}{2n} \tau_0^2 K_f, \tag{11a}
\]
\[
\langle \mathcal{C}_2^{\infty} \rangle = \frac{\langle \delta P_0^2 \rangle}{2n} \tau_0^2 K_{f-1}. \tag{11b}
\]
The network structure determines the performance measures via the spectrum of the weighted Laplacian of Eq. (3). The latter depends on the network structure – its topology and edge weights, as well as the internal dynamics of the oscillators, which modifies the edge weights via angle differences $\theta_i^{(0)} - \theta_j^{(0)}$ determined by $P^{(0)}$. The way all these ingredients determine average network fragility is however simply encoded in $K_{f_{-1}}$ and $K_{f_1}$. We note that Eq. (11a) appeared in slightly different, but equivalent form in Ref. 3.

**Box perturbation.** Next, we go beyond the $\delta$ perturbations discussed so far and consider a perturbation that is extended, but still limited in time, $\delta P(t) = \delta P_0 \Theta(t) \Theta(t_0 - t)$, with the Heaviside function $\Theta(t) = 0$, $t < 0$ and $\Theta(t) = 1$, $t > 0$. Here also, the limit $T \rightarrow \infty$ can be taken in Eqs. (S10). One obtains (see Supplemental Material)

$$\langle \epsilon_1^\infty \rangle = \frac{1}{\lambda_2} \sum_{\alpha \geq 2} \frac{(\delta P_0 \cdot u_\alpha)^2}{\lambda^3}(\lambda_\alpha \tau_0 - 1 + e^{-\lambda_\alpha \tau_0}), \quad (12a)$$

$$\langle \epsilon_2^\infty \rangle = \frac{1}{\lambda_2} \sum_{\alpha \geq 2} \frac{(\delta P_0 \cdot u_\alpha)^2}{\lambda^3}(1 - e^{-\lambda_\alpha \tau_0}). \quad (12b)$$

Compared to Dirac delta perturbations, $\langle \epsilon_1^\infty \rangle$ now depends on $K_{f_2}$ when $\tau_0$ is the longest time scale. This is so, because time-extended perturbations scatter through the network before they are damped by $L$. Accordingly, they depend on details of the network contained in higher moments of the distribution of resistance distances, hence on a generalized Kirchhoff index of higher order.

**Noisy perturbation.** We finally consider fluctuating perturbations characterized by zero average and second moment $\delta P_1(t_1) \delta P_2(t_2) = \delta_0 \delta_0 \exp[-|t_1 - t_2|/\tau_0]$ correlated over a typical time scale $\tau_0$. Because this perturbation is not limited in time, we consider $\langle \epsilon_1,2(T) \rangle$ at finite but large $T$. Keeping only the leading-order term in $T$, we have (see Supplemental Material)

$$\langle \epsilon_1^\infty \rangle = \frac{1}{\lambda_2} \sum_{\alpha \geq 2} \frac{(\delta P_0 \cdot u_\alpha)^2}{\lambda_\alpha(\lambda_\alpha + \tau_0^{-1})} + O(T^0), \quad (14a)$$

$$\langle \epsilon_2^\infty \rangle = \frac{1}{\lambda_2} \sum_{\alpha \geq 2} \frac{(\delta P_0 \cdot u_\alpha)^2}{\lambda_\alpha + \tau_0^{-1}} + O(T^0). \quad (14b)$$

The response is determined by the overlap of the perturbation vector with the eigenmodes of $L$. The noise amplitude $\delta P_{0i}$ is localized on the set $N_{0i}$ of noisy nodes. Averaging over an ensemble of perturbations defined by all permutations of the noisy nodes over all nodes (see Supplemental Material), $(\langle \epsilon_1,2 \rangle)$ is given by Eqs. (14) with $\langle \sum_i \delta P_{0i}^2 u_{0i}^2 \rangle \rightarrow \langle \delta P_{0i}^2 \rangle$. If $\tau_0^{-1}$ lies inside the spectrum of $L$, $\epsilon_1,2$ are functions of the spectrum of $L$ and the inverse correlation time $\tau_0^{-1}$. If, on the other hand, $\tau_0^{-1}$ lies outside the spectrum of $L$, averaged measures are directly expressable as infinite sums over generalized Kirchhoff

### Numerical Simulations

To confirm our results numerically, we focus on $C_1$ for both box and noisy perturbations, varying their time scale $\tau_0$. We consider Eq. (1) with two types of networks, (i) small-world networks, where a cycle graph with constant coupling $b_{ij} = b_0$ for any node $i$ to its 4 nearest neighbors undergoes random rewiring with probability $p \in [0,1]$, and (ii) simple cyclic networks where each node is coupled to its nearest- and $q$-th-neighbors with a constant coupling $b_{i,i \pm q} = b_0$ (see inset in Fig. S1). In both cases, we fix the number of nodes to $n = 50$. In all cases, the unperturbed natural frequencies vanish, $P_{0i}^{(0)} = 0$. The box perturbation has $\delta P_0 = (0,0,...,\delta P_{0i},0,...,\delta P_{0i},0,...)$ with $\delta P_{0i} = -|\delta P_{0i}| = 0.01 b_0$, and averaging is performed over all pairs of nodes $(i_1,i_2)$. The noisy perturbation acts on all nodes, and we construct noise sequences $P_i(t)$ satisfying $\delta P_{0i}(t_1) \delta P_{0i}(t_2) = \delta_0 \delta_0 \exp[-|t_1 - t_2|/\tau_0]$ using the method described in Ref. 30, with $\delta P_{0i} = 0.01 b_0$.

The theory is numerically confirmed for small-world networks in Fig. S2, where $C_1$ decreases monotonically as the rewiring probability $p$ increases, in complete agreement with Eqs. (13a) and (14a) (colored solid lines). This is qualitatively understood as follows. As $p$ increases and more network edges are rewired, more cou-
between oscillators, but is entirely determined by gener-
eralized Kirchhoff indices, which we introduce in Eq. [10]. The latter are both spectral and
analytical systems, both beyond the Kuramoto model con-
tained in Eq. (9). The latter are both spectral and
to determine few of the smallest eigenvalues of
the virtual network defined
only through generalized Kirchhoff indices, which we
considered performance measures averaged over ergodic en-
sambles of perturbations. In this case, they depend on
L only through generalized Kirchhoff indices, which we
introduced in Eq. [9]. The latter are both spectral and
topological in nature, as they can be reexpressed in terms
of the resistance distances in the virtual network defined
by L (see Supplemental Material). A network’s average
or global fragility can therefore be easily quantified by a
direct calculation of generalized Kirchhoff indices. This is
a computationally easy task, requiring in most instances
to determine few of the smallest eigenvalues of L, and
that, for a given system, can be done for few typical fixed
points once and for all. Our findings are rather general
and generalized Kirchhoff indices naturally characterize the
fragility of synchronous states for many coupled dy-
namical systems, both beyond the Kuramoto model con-
sidered here as well as for other types of perturbation not
discussed here (see Supplemental Material).

Two extensions of this work should be considered. First, our approach has been based on the implicit as-
sumption that the perturbation is sufficiently weak, such
that the system stays close to its initial state. Criteria
for acute vulnerability should account for the breakdown of
this assumption and quantify the perturbation thresh-
old above which networks either lose synchrony or change

Figure 1. (Color online) Performance measure $\langle C_1^\infty \rangle$ for box
perturbation, left panel and $\overline{C}_f/T$ for noisy perturbation,
right panel) for the small-world model with $n = 50$ nodes as a
function of the rewiring probability $p$ (for box
perturbation, left panel) and with $\tau_0 = 0.1/b_0$
(black), $0.5/b_0$ (blue), $1/b_0$ (red), $10/b_0$ (green) and $50/b_0$
(violet). Solid lines give Eqs. (13a) (left) and (14a) (right)
calculated numerically over an ensemble of networks obtained
from 20 different rewirings. The dotted-dashed lines give
$K_f$ and the dashed lines $K_f$, both vertically shifted. In the right
panel, $\overline{C}_f(T)$ is averaged over $T' \in [T - 200/b_0, T + 200/b_0]$ with
$T = 800/b_0$, and error bars give the standard deviation of
numerically obtained values with 10 different noise sequences.

Further insight into synchrony fragility is obtained
when considering our cyclic graph model with nearest-
and $q$th-neighbor coupling. If the range of the coupling
were the only ingredient determining the fragility of the
synchronous state, then one would observe a monotonic
decrease of $C_1$ as a function of $q$. Fig. S2 shows numerical
results for the cyclic graphs and five values of $\tau_0$
ranging from $\lambda_0, \tau_0 \lesssim 1$ to $\lambda_0, \tau_0 \gtrsim 1$, $\forall \alpha$. Analytical
results of Eqs. (13a) and (14a), in particular, the crossover
from $\langle C_1^\infty \rangle \sim K_f$ to $\langle C_1^\infty \rangle \sim K_{f2}$ predicted in Eq. (13a)
when $\tau_0$ increases, are clearly confirmed. Particularly re-
markable is that $K_f$ and $K_{f2}$ are not monotonous in the
coupling range $q$ (see Supplemental Material), which is
clearly reflected in the behavior of $\langle C_1^\infty \rangle$. This unambigu-
ously demonstrates that average fragility of synchrony
does not depend trivially on the range of the couplings
between oscillators, but is entirely determined by gen-
eralized Kirchhoff indices.

Conclusion. Using both performance measures de-
defined in Eqs. [10], we have expressed synchrony fragility
in terms of the weighted Laplacian matrix $L$ of the sys-
tem’s network. We have first shown that the response
to specific perturbations is determined by both the spec-
trum of $L$ and its eigenmodes $u_\alpha$ through their scalar
product $\delta P \cdot u_\alpha$ with the perturbation vector. Eqs. (10),
(12) and (14) clearly indicate that perturbations overlap-
ning with the eigenmodes with smallest Lyapunov expo-
\[\text{Fig. 2. (Color online) Performance measure } \langle C_1^\infty \rangle \text{ for box}
\text{perturbation, left panel and } \overline{C}_f/T \text{ for noisy perturbation,}
\text{right panel) for the cyclic graph with } n = 50 \text{ nodes with nearest-}
\text{and } q\text{th-neighbor coupling, } b_{i,i+1} = b_{i,i+q} = b_0,
\text{as a function of } q \text{ and with } \tau_0 = 0.1/b_0 \text{ (black), } 0.5/b_0
\text{ (blue), } 1/b_0 \text{ (red), } 10/b_0 \text{ (green) and } 50/b_0 \text{ (violet). Solid}
\text{lines give Eqs. (13a) (left) and (14a) (right). The dotted-dashed}
\text{lines give } K_f \text{ and the dashed lines } K_{f2}, \text{ both vertically shifted. In the right}
\text{panel, } \overline{C}_f(T) \text{ is averaged over } T' \in [T - 200/b_0, T + 200/b_0]
\text{with } T = 800/b_0, \text{ and error bars give the standard deviation of numerically}
\text{obtained values with } 10 \text{ different realizations of noisy per-
turbations. The inset sketches the model for } n = 8 \text{ and } q = 3.\]
their synchronous state. Second, synchrony fragility for second-order systems with inertia should be considered, investigating in particular more closely the case of electric power grids under the influence of fluctuating power injections \cite{31}. Work along those lines is in progress.

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GENERALIZED KIRCHHOFF INDICES

For a complex graph determined by its Laplacian matrix $L$, we introduce a family of graph invariants

$$Kf_m = n \sum_{\alpha \geq 2} \lambda_{\alpha}^{-m}, \quad (S1)$$

where $\lambda_\alpha$ is an eigenvalue of $L$. We call them generalized Kirchhoff indices because the Kirchhoff index introduced in Ref. [S1] can be expressed as $[S2, S3]$:

$$Kf_1 = n \sum_{\alpha \geq 2} \lambda_{\alpha}^{-1}. \quad (S2)$$

We show that, just like the original Kirchhoff index $Kf_1$, generalized Kirchhoff indices can be expressed as functions of the resistance distances between any pair of nodes $(i, j)$ in the network. The network’s Laplacian matrix $L$ has one zero eigenvalue. We therefore define the matrix

$$\Gamma = L + u_1^T u_1, \quad (S3)$$

in terms of which the resistance distance $\Omega_{ij}$ between nodes $i$ and $j$ is defined as $[S1]$:

$$\Omega_{ij} = \Gamma_{ii}^{-1} + \Gamma_{jj}^{-1} - 2\Gamma_{ij}^{-1}. \quad (S4)$$

This can be rewritten in terms of the eigenvectors of $L$ as $[S1]$:

$$\Omega_{ij} = \sum_{\alpha \geq 2} \left(\frac{u_{\alpha,i} - u_{\alpha,j}}{\lambda_{\alpha}}\right)^2, \quad (S5)$$

where the zero mode corresponding to $\lambda_1 = 0$ is omitted in the sum. We show that $Kf_1$ and $Kf_2$ can be rewritten in terms of resistance distances. For $Kf_1$, one has

$$\sum_{i<j} \Omega_{ij} = \frac{1}{2} \sum_{i, j} \sum_{\alpha \geq 2} \frac{(u_{\alpha,i} - u_{\alpha,j})^2}{\lambda_{\alpha}} = Kf_1, \quad (S6)$$

because the eigenvectors $\alpha \geq 2$ of $L$ satisfy $\sum_i u_{\alpha,i} = 0$ and $\sum_i u_{\alpha,i}^2 = 1$. To express $Kf_2$, higher moments of the distribution of resistance distances are needed. One has,

$$\sum_{i,j} \Omega_{ij}^2 = \sum_{i,j; \alpha, \beta \geq 2} \frac{(u_{\alpha,i} - u_{\alpha,j})^2(u_{\beta,i} - u_{\beta,j})^2}{\lambda_{\alpha}\lambda_{\beta}} = 2n \sum_{i; \alpha, \beta \geq 2} \frac{u_{\alpha,i}^2 u_{\beta,i}^2}{\lambda_{\alpha}\lambda_{\beta}} + \frac{2(Kf_1)^2}{n^2} + \frac{4Kf_2}{n}, \quad (S7)$$

$$\sum_{i,j,k} \Omega_{ij}\Omega_{jk} = \sum_{i,j,k; \alpha, \beta \geq 2} \frac{(u_{\alpha,i} - u_{\alpha,j})(u_{\beta,j} - u_{\beta,k})}{\lambda_{\alpha}\lambda_{\beta}} = \frac{3(Kf_1)^2}{n} + n^2 \sum_{i; \alpha, \beta \geq 2} \frac{u_{\alpha,i}^2 u_{\beta,i}^2}{\lambda_{\alpha}\lambda_{\beta}}. \quad (S8)$$

Combining the latter two equations, one has

$$\frac{n}{4} \left( \sum_{i,j} \Omega_{ij}^2 \right) - \frac{1}{2} \left( \sum_{i,j,k} \Omega_{ij}\Omega_{jk} \right) + \frac{(Kf_1)^2}{n} = Kf_2. \quad (S9)$$

For $m \geq 3$, it is possible though algebraically tedious to show that $Kf_m$ can be expressed in a similar way in terms of higher moments of resistance distances.
DIRECT CALCULATION OF FRAGILITY MEASURES

The fragility performance measures introduced in Eqs. (4a,b) of the main text can be rewritten in terms of the coefficients of the expansion $\delta \tilde{\theta}(t) = \sum_\alpha c_\alpha(t) u_\alpha$ of angle displacements over the eigenvectors $u_\alpha$ of $\mathbf{L}$. One has

$$
\mathfrak{C}_1(T) = \sum_{\alpha \geq 2} \int_0^T c^2_\alpha(t) dt, \quad \mathfrak{C}_2(T) = \sum_{\alpha \geq 2} \int_0^T \dot{c}^2_\alpha(t) dt.
$$

(S10)

The coefficients $c_\alpha(t)$ are solutions of

$$
c_\alpha(t) = e^{-\lambda_\alpha t} c_\alpha(0) + e^{-\lambda_\alpha t} \int_0^t dt' e^{\lambda_\alpha t'} \delta \mathbf{P}(t') \cdot u_\alpha,
$$

(S11)

and in our case where the perturbation starts at $t = 0$, $c_\alpha(0) = 0$. We treat sequentially, and in some additional details, the three perturbations considered in the main text.

Dirac delta perturbation

We first consider $\delta \mathbf{P}(t) = \delta \mathbf{P}_0 \tau_0 \delta(t)$. Inserting it into Eq. (S11) one obtains,

$$
c_\alpha(t) = (\delta \mathbf{P}_0 \cdot u_\alpha) \tau_0 e^{-\lambda_\alpha t}.
$$

(S12)

This directly gives

$$
\mathfrak{C}_1(T) = \frac{\sum_{\alpha \geq 2} (\delta \mathbf{P}_0 \cdot u_\alpha)^2 \tau_0^2}{2 \lambda_\alpha} (1 - e^{-2\lambda_\alpha T}) \cdot \mathfrak{C}_2(T) = \frac{\sum_{\alpha \geq 2} (\delta \mathbf{P}_0 \cdot u_\alpha)^2 \tau_0^2}{2} \lambda_\alpha (1 - e^{-2\lambda_\alpha T}).
$$

(S13)

Taking the limit $\lambda_\alpha T \gg 1$, one obtains Eqs. (10a,b) of the main text,

$$
\mathfrak{C}_1(T \to \infty) = \mathfrak{C}_1^\infty = \frac{\tau_0^2}{2} \sum_{\alpha \geq 2} \frac{(\delta \mathbf{P}_0 \cdot u_\alpha)^2}{\lambda_\alpha},
$$

(S14a)

$$
\mathfrak{C}_2(T \to \infty) = \mathfrak{C}_2^\infty = \frac{\tau_0^2}{2} \sum_{\alpha \geq 2} (\delta \mathbf{P}_0 \cdot u_\alpha)^2 \lambda_\alpha.
$$

(S14b)

Averaging over the ensemble of perturbation vectors defined by $\langle \delta \mathbf{P}_{0i} \rangle = 0$, $\langle \delta \mathbf{P}_{0i} \delta \mathbf{P}_{0j} \rangle = \delta_{ij} \langle \delta \mathbf{P}_{0i}^2 \rangle$, we have $\langle (\delta \mathbf{P}_0 \cdot u_\alpha)^2 \rangle = \frac{1}{2} \sum_{i,j} \langle \delta \mathbf{P}_{0i} \delta \mathbf{P}_{0j} \rangle u_{\alpha,i} u_{\alpha,j} = \langle \delta \mathbf{P}_{0i}^2 \rangle$, $\forall \alpha$. The averaged fragility measures become,

$$
\langle \mathfrak{C}_1^\infty \rangle = \langle \mathfrak{C}_1^\infty \rangle = \frac{\sum_{\alpha \geq 2} 1}{\lambda_\alpha} = \frac{\sum_{\alpha \geq 2} (\delta \mathbf{P}_0 \cdot u_\alpha)^2 \tau_0}{2n} K f_1,
$$

(S15a)

$$
\langle \mathfrak{C}_2^\infty \rangle = \langle \mathfrak{C}_2^\infty \rangle = \frac{\sum_{\alpha \geq 2} \lambda_\alpha}{\lambda_\alpha} = \frac{\sum_{\alpha \geq 2} (\delta \mathbf{P}_0 \cdot u_\alpha)^2 \tau_0}{2n} K f_1,
$$

(S15b)

just as in Eqs. (11a,b) in the main text.

Box perturbation

We next consider $\delta \mathbf{P}(t) = \delta \mathbf{P}_0 \Theta(t)(\tau_0 - t)$. Eq. (S11) with $c_\alpha(0) = 0$ now gives,

$$
c_\alpha(t) = \begin{cases} 
(\delta \mathbf{P}_0 \cdot u_\alpha)(1 - e^{-\lambda_\alpha t}) / \lambda_\alpha, & t \leq \tau_0, \\
(\delta \mathbf{P}_0 \cdot u_\alpha)(e^{\lambda_\alpha(\tau_0 - t)} - e^{-\lambda_\alpha t}) / \lambda_\alpha, & t > \tau_0.
\end{cases}
$$

(S16)

Eqs. (S10) become

$$
\mathfrak{C}_1(T) = \sum_{\alpha \geq 2} \frac{(\delta \mathbf{P}_0 \cdot u_\alpha)^2}{\lambda_\alpha^2} (\lambda_\alpha \tau_0 - 1 + e^{-\lambda_\alpha \tau_0} - e^{2\lambda_\alpha(\tau_0 - T)} / 2 + e^{\lambda_\alpha(\tau_0 - T)} - e^{-2\lambda_\alpha T} / 2),
$$

(S17a)

$$
\mathfrak{C}_2(T) = \sum_{\alpha \geq 2} \frac{(\delta \mathbf{P}_0 \cdot u_\alpha)^2}{\lambda_\alpha} (1 - e^{-\lambda_\alpha \tau_0} - e^{2\lambda_\alpha(\tau_0 - T)} / 2 + e^{\lambda_\alpha(\tau_0 - 2T)} - e^{-2\lambda_\alpha T} / 2).
$$

(S17b)
Taking the limit $\lambda_\alpha T \gg 1$, one recovers Eqs. (12a,b) in the main text,

\begin{align}
\mathcal{E}_1^\infty &= \sum_{\alpha \geq 2} \frac{(\delta P_\alpha \cdot u_{\alpha,i})^2}{\lambda_\alpha^3} (\lambda_\alpha \tau_0 - 1 + e^{-\lambda_\alpha \tau_0}), \\
\mathcal{E}_2^\infty &= \sum_{\alpha \geq 2} \frac{(\delta P_\alpha \cdot u_{\alpha,i})^2}{\lambda_\alpha} (1 - e^{-\lambda_\alpha \tau_0}).
\end{align}

Following the same averaging procedure as for the $\delta$-perturbation, one finally recovers Eqs. (13a,b) in the main text,

\begin{align}
\langle \mathcal{E}_1^\infty \rangle &= \langle \delta P_0 \rangle \sum_{\alpha \geq 2} \frac{1}{\lambda_\alpha^2} (\lambda_\alpha \tau_0 - 1 + e^{-\lambda_\alpha \tau_0}), \\
\langle \mathcal{E}_2^\infty \rangle &= \langle \delta P_0 \rangle \sum_{\alpha \geq 2} \frac{1}{\lambda_\alpha} (1 - e^{-\lambda_\alpha \tau_0}).
\end{align}

The asymptotic behaviors for $\lambda_\alpha \tau_0 \ll 1$ and $\lambda_\alpha \tau_0 \gg 1$ are easily computed via a Taylor-expansion.

**Noisy perturbation**

We finally consider fluctuating perturbations characterized by zero average $\overline{\delta P_{0i}} = 0$, and second moments $\overline{\delta P_1(t_1) \delta P_2(t_2)} = \delta_{ij} \delta P_0^2 \exp[-\tau_0^{-1}|t_1 - t_2|]$ correlated over a typical time scale $\tau_0$. With this ensemble average, one obtains,

\begin{align}
\overline{\mathcal{E}_1}(T) &= \sum_{\alpha \geq 2} \int_0^T \overline{\mathcal{E}_1^\alpha(t)} dt \\
&= \sum_{\alpha \geq 2} \int_0^T e^{-2\lambda_\alpha t} \int_0^t \int_0^{t_1} e^{\lambda_\alpha (t_1 + t_2)} \overline{\delta \mathbf{P}(t_1) \cdot u_{\alpha,i} \delta \mathbf{P}(t_2) \cdot u_{\alpha,i}} dt_1 dt_2 dt \\
&= \sum_{\alpha \geq 2} \sum_i (\delta P_{0i} u_{\alpha,i})^2 \int_0^T e^{-2\lambda_\alpha t} \int_0^t \int_0^{t_1} e^{\lambda_\alpha (t_1 + t_2)} e^{-|t_1 - t_2|/\tau_0} dt_1 dt_2 dt \\
&= \sum_{\alpha \geq 2} \sum_i (\delta P_{0i} u_{\alpha,i})^2 \left[ \frac{T}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1}) + 1 - e^{-2\lambda_\alpha T}} \right] + \frac{2\lambda_\alpha^2 (\lambda_\alpha - \tau_0^{-1}) + 2(e^{-\lambda_\alpha + \tau_0^{-1}} - 1)}{\lambda_\alpha + \tau_0^{-1} \lambda_\alpha (\lambda_\alpha - \tau_0^{-1})^2}.
\end{align}

To calculate $\overline{\mathcal{E}_2}(t)$ we note that,

\begin{equation}
\overline{\mathcal{E}_2}(t) = \lambda_\alpha^2 \overline{\mathcal{E}_1}(t) + (\overline{\delta \mathbf{P}(t) \cdot u_{\alpha,i}}^2 - 2\lambda_\alpha e^{-\lambda_\alpha t}) \int_0^t e^{\lambda_\alpha t} \overline{\delta \mathbf{P}(t') \cdot u_{\alpha,i} \delta \mathbf{P}(t) \cdot u_{\alpha,i}} dt'.
\end{equation}

One obtains, after some algebra

\begin{align}
\overline{\mathcal{E}_2}(T) &= \sum_{\alpha \geq 2} \sum_i (\delta P_{0i} u_{\alpha,i})^2 \left[ \frac{T}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1}) + 1 - e^{-2\lambda_\alpha T}} \right] + \frac{2\lambda_\alpha^2 (\lambda_\alpha - \tau_0^{-1}) + 2(e^{-\lambda_\alpha + \tau_0^{-1}} - 1)}{\lambda_\alpha + \tau_0^{-1} \lambda_\alpha (\lambda_\alpha - \tau_0^{-1})^2}.
\end{align}

For $\lambda_\alpha T \gg 1$, one recovers Eqs. (14a,b) in the main text,

\begin{align}
\overline{\mathcal{E}_1}(T) &= \sum_{\alpha \geq 2} \sum_i (\delta P_{0i} u_{\alpha,i})^2 \frac{T}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1})} + \mathcal{O}(T^0), \\
\overline{\mathcal{E}_2}(T) &= \sum_{\alpha \geq 2} \sum_i (\delta P_{0i} u_{\alpha,i})^2 \frac{\tau_0^{-1} T}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1})} + \mathcal{O}(T^0).
\end{align}

Averaging over all permutations, $\sigma$, of the components of $\delta \mathbf{P}_0 = (\delta P_{01}, \ldots, \delta P_{0n})$, one has the following identity,

\begin{equation}
\frac{1}{n!} \sum_{\sigma} \sum_i (\delta P_{0\sigma(i)} u_{\alpha,i})^2 = \frac{(n - 1)!}{n!} \left( \sum_i \delta P_{0i}^2 \right) \left( \sum_j u_{\alpha,j}^2 \right) = \frac{(\delta P_0^2)^2}{n} \equiv \langle (\delta P_0^2)^2 \rangle.
\end{equation}
We finally obtain the leading-order contribution in $T$,

\begin{equation}
(\mathcal{E}_1)(T) = \langle \delta P_0^2 \rangle \sum_{\alpha \geq 2} \frac{T}{\lambda_\alpha (\lambda_\alpha + \tau_0^{-1})} + \mathcal{O}(T^0),\end{equation}

\begin{equation}
(\mathcal{E}_2)(T) = \langle \delta P_0^2 \rangle \sum_{\alpha \geq 2} \frac{\tau_0^{-1} T}{(\lambda_\alpha + \tau_0^{-1})} + \mathcal{O}(T^0),\end{equation}

which can be Taylor-expanded in geometric series when either $\lambda_\alpha \tau_0 > 1$ or $\lambda_\alpha \tau_0 < 1$, $\forall \alpha$ to obtain \(\langle \mathcal{E}_{1,2} \rangle = n^{-1} \langle \delta P_0^2 \rangle / T \sum_{m=0}^{\infty} C_{1,2}^{(m)}\). One easily recovers the coefficients \(C_{1,2}^{(m)}\) given in Eqs. (15a,b) in the main text. Note that, when $\lambda_\alpha \tau_0 \ll 1$,

\begin{equation}
(\mathcal{E}_1)(T) \sim \frac{T \langle \delta P_0^2 \rangle}{n} C_{1}^{(0)} = \frac{T \langle \delta P_0^2 \rangle}{n} \tau_0 K_{f1},
\end{equation}

\begin{equation}
(\mathcal{E}_2)(T) \sim \frac{T \langle \delta P_0^2 \rangle}{n} C_{2}^{(0)} = \frac{T \langle \delta P_0^2 \rangle}{n} \tau_0 K_{f0}.
\end{equation}

For $\lambda_\alpha \tau_0 \gg 1$, one obtains,

\begin{equation}
(\mathcal{E}_1)(T) \sim \frac{T \langle \delta P_0^2 \rangle}{n} C_{1}^{(0)} = \frac{T \langle \delta P_0^2 \rangle}{n} \tau_0 K_{f2},
\end{equation}

\begin{equation}
(\mathcal{E}_2)(T) \sim \frac{T \langle \delta P_0^2 \rangle}{n} C_{2}^{(0)} = \frac{T \langle \delta P_0^2 \rangle}{n \tau_0} K_{f1}.
\end{equation}

**Perturbations as Fourier series**

From Eq. (6) in the main text, it is clear that, with $c_\alpha(0) = 0$, $\mathcal{E}_{1,2}$ will always be a sum over eigenmodes of $L$ labeled $\alpha$, and that each term in that sum contains a factor $(\delta P \cdot u_\alpha)^2$, regardless of the choice of perturbation. Our first main conclusion, that the response of the synchronous state under specific, nonaveraged perturbation depends on the overlap of its eigenmodes with the perturbation vector is therefore rather general.

Our approach can moreover be extended to any perturbation that can be expanded in a Fourier series,

\begin{equation}
\delta P(t) = \sum_p \left( \delta P_p^+ \exp[2\pi ipt/\tau_0] + \delta P_p^- \exp[-2\pi ipt/\tau_0] \right).
\end{equation}

The condition that the components of the perturbation vector are real, $\delta P_i \in \mathbb{R}$, gives either $\delta P_{p,i}^+ = \delta P_{p,i}^- \in \mathbb{R}$ or $\delta P_{p,i}^+ = -\delta P_{p,i}^- \in i\mathbb{R}$. Because $\delta P(t = 0) = 0$, we consider only the latter case in what follows. Eq. (6) in the main text gives

\begin{equation}
c_\alpha(t) = \exp[-\lambda_\alpha t] \sum_{p,i} u_{\alpha,i} \delta P_{p,i}(t) \left( \frac{e^{(\lambda_\alpha + 2\pi ip/\tau_0)t} - 1}{\lambda_\alpha + 2\pi ip/\tau_0} - \frac{e^{(\lambda_\alpha - 2\pi ip/\tau_0)t} - 1}{\lambda_\alpha - 2\pi ip/\tau_0} \right).
\end{equation}

In the long time limit we obtain

\begin{equation}
c_\alpha(t \to \infty) = \sum_{p,i} u_{\alpha,i} |\delta P_{p,i}(t)| \frac{(4\pi p/\tau_0) \cos(2\pi pt/\tau_0) - 2\lambda_\alpha \sin(2\pi pt/\tau_0)}{\lambda_\alpha^2 + 4\pi^2 p^2/\tau_0^2}.
\end{equation}

To get the average of the performance measure $\mathcal{E}_1$, we square this expressions, average it over an homogeneous ensemble of perturbation as in the main text and sum over $\alpha$. For a sufficiently long duration of perturbation, integrating over time gives the dominant contribution to the fragility performance measures (under the assumption that $T$ is large, but shorter than the duration of the perturbation)

\begin{equation}
\mathcal{E}_1(T) \sim \frac{T \tau_0}{2} \sum_{\alpha,p} \left( \langle \delta P_{p,0}^2 \rangle \frac{4\lambda_\alpha^2 \tau_0^2 + 16\pi^2 p^2}{(\lambda_\alpha^2 \tau_0^2 + 4\pi^2 p^2)^2} \right).
\end{equation}

For each Fourier harmonics, the denominator can be Taylor-expanded, depending on whether $\lambda_\alpha \tau_0 > 2\pi p$ of $\lambda_\alpha \tau_0 < 2\pi p$. When $2\pi p/\tau_0$ lies outside of the spectrum of the Laplacian, Eq. (S33) allows to express $\mathcal{E}_1(T)$ as a sum over even-order generalized Kirchhoff indices.
KIRCHHOFF INDICES AND PHASE DYNAMICS

Kirchhoff indices in the cycle model with nearest and $q^{th}$-neighbor coupling

The eigenvalues $\lambda_{\alpha}$ of the Laplacian of our model with uniform nearest and $q^{th}$-neighbor coupling are obtained by a Fourier transformation and are given by,

$$\lambda_{\alpha} = 4 - 2 \cos(k_{\alpha}) - 2 \cos(k_{\alpha}q) , \quad \alpha = 1, \ldots, n ,$$  \hspace{1cm} (S34)

where $k_{\alpha} = \frac{2\pi(\alpha-1)}{n}$. Then one obtains,

$$K_{f1} = n \sum_{\alpha \geq 2} \frac{1}{4 - 2 \cos(k_{\alpha}) - 2 \cos(k_{\alpha}q)} .$$  \hspace{1cm} (S35)

The dependence of the denominator of Eq. (S35) with $q$ makes it is clear that $K_{f1}$ is a non monotonous function of $q$. This is shown in Fig. S1 for $n = 50$. As mentioned in the main text, two seemingly similar choices of long range interactions may lead to large differences of the Kirchhoff index. This translates into large variations of the network’s average resistance distance, which can be understood topologically in terms of the commensurability of the $q^{th}$-neighbor coupling with the number $n$ of nodes in the network. Since the resistance distance between any two nodes accounts for all paths between them, one may expect that $q^{th}$-neighbor couplings provide short, alternative paths between nodes, effectively bringing them closer to each other. This is however not always the case. In fact, if $n$ is a small integer multiple of $q$, or nearly so, paths involving multiple $q$-range hops, starting from a given node, come back to the initial node or close to it. Such paths only allow to reach nodes in the close vicinity of the starting node and do not reduce significantly the resistance distance between many nodes close to the original one. In contrast, without commensurability between $q$ and $n$, the resistance distance between most of the nodes to the original one is reduced.

This is illustrated in Fig. S1 which plots $K_{f1}$ and $K_{f2}$ as a function of $q$ for $n = 50$. As expected peaks are present for $q = 10, 17 $ and $24$ such that $n/q$ is a small integer or close to a small integer. The insets of Fig. S1 sketch how despite these long range interactions some portions of the network keep the same geodesic distance to the red node 1.

Influence on the phase dynamics

The two seemingly similar cycle models with $n = 50$, nearest- and $q^{th}$-neighbor coupling with $q = 17$ and $q = 19$ considered in Fig. 2 in the main text nevertheless behave strikingly differently under an external perturbation. This is so, because they have very different generalized Kirchhoff indices $K_{f1}$ and $K_{f2}$. To illustrate this behavioral discrepancy, Fig. S2 compares the phase dynamics for the box perturbation for these two graphs. The left panel is for the cyclic graph with $q = 17$, which has a fragility performance measure $\mathcal{C}_1$ bigger than for the cyclic graph with $q = 19$ in the right panel. Clearly the angle deviations on the left panel spread more and take more time to return to the initial fixed point after the perturbation than on the right panel. From Eq. (13b) in the main text, $\langle \tau_0 \rangle$ is proportional to $K_{f2}$ in the corresponding limit $\lambda_{\alpha} \tau_0 \gg 1 \forall \alpha$. The numerically obtained values (indicated in Fig. S2) of $\mathcal{C}_1$ follow that trend, though not exactly, as expected for this single realization of perturbation.

$\lambda_2$ vs. Generalized Kirchhoff indices

Here we show that the generalized Kirchhoff indices give more information on the fragility of synchronous states than the smallest Lyapunov exponent, $\lambda_2$. We compare star and cycle graphs with the same number of nodes. In both cases the eigenvalues of the Laplacian matrix can be calculated analytically. The spectrum of the Laplacian of a star graph with $n$ nodes is $\{0, 1, n\}$, with the eigenvalue 1 having multiplicity $n - 2$, thus $\lambda_2 = 1$, $K_{f1} = (n - 1)^2$, and $K_{f2} = (n^2 - 2n - 2)/n$. The spectrum of the Laplacian of a cycle graph with only nearest neighbor coupling is $\lambda_{\alpha} = 2[1 - \cos(2\pi(\alpha - 1)/n)]$, with $\alpha = 1, \ldots, n$. One has, $\lambda_2 = 2[1 - \cos(2\pi/n)]$, $K_{f1} = n(n^2 - 1)/12$, and $K_{f2} = n(n^2 - 1)(n^2 + 1)/720$. The density of eigenvalues approaches zero increases with $n$ in the cycle graph, while in the star graph the Lyapunov exponents accumulate at a finite value as $n$ increases (i.e. $\lambda_2 = \lambda_3 = \ldots = \lambda_{n-1} = 1$).

Thus we expect a crossover in the vulnerability of these two network topologies as $n$ is increased. Fig. S3 shows $\lambda_2$ (left panel), and the performance measure $\mathcal{C}_1$ in both the limits $\lambda_{\alpha} \tau_0 \ll 1$ (center panel) and $\lambda_{\alpha} \tau_0 \gg 1$ (right panel) as a function of the number of nodes. For $n < 6$, the cycle network has a larger $\lambda_2$ and a smaller $\mathcal{C}_1$ compared to the
Figure S1. (Color online) $Kf_1$ (green) and $Kf_2$ (violet) for a cyclic graph with $n = 50$ with nearest and $q^{th}$-neighbor coupling, $b_{i,i\pm 1} = b_{i,i\pm q} = b_0$. The inset sketches the model for $q = 17$, 19, and 24 and illustrates one path involving $q^{th}$-range interactions starting from node 1 (red). The addition of $q^{th}$-neighbor interactions does not reduce the geodesic distance between the reference node (red) and the set of nodes colored in blue.

Figure S2. (Color online) Phases $\delta \theta_i$ for the cyclic graph with $n = 50$ with nearest and $q^{th}$-neighbor coupling, $b_{i,i\pm 1} = b_{i,i\pm q} = b_0$ with $q = 17$ (left panel) and $q = 19$ (right panel), as a function of the normalized time $t/\tau_0$, for a box perturbation with $\tau_0 = 50/b_0$ and perturbation vector with non zero components $\delta P_{0,1} = b_0, \delta P_{0,11} = -b_0$. 

$C_1^\infty = 275.71$

$Kf_2 = 4575.22$

$C_1^\infty = 23.81$

$Kf_2 = 567.17$
star network which means that the largest contribution to $Kf_1$ and $Kf_2$ comes from $\lambda_2^{-1}$. However, for $6 > n > 8$, both $Kf_1$ and $Kf_2$ are not dominated by $\lambda_2^{-1}$, therefore the cycle network is less fragile under an external perturbation than the star network even though it has a smaller $\lambda_2$. For $n = 8$ and $n = 9$, the cycle network is less fragile against short time perturbation but more fragile against long time perturbation compared to the star network. This reflects the fact that for those values, $Kf_1$ is smaller for the cycle than for the star network, while the relation is opposite for $Kf_2$ [see the corresponding relation between $C_1$ and generalized Kirchhoff indices in Eq. (13a) in the main text].

We apply the same analysis to small-world graphs, which are obtained from a $n = 20$ cycle network with first and second nearest neighbor couplings, which are rewired [S6]. Fig. S4 shows three graphs obtained with this procedure, which have different relations between $\lambda_2$ and their generalized Kirchhoff indices $Kf_1$ and $Kf_2$. Graph 1 has a smaller $\lambda_2$ but smaller $Kf_1$ or $Kf_2$ compared to graph 2, while graph 3 has a smaller $\lambda_2$, similar $Kf_1$ and larger $Kf_2$ compared to graph 2. These relations are reflected by the performance measures $C_1$ for box perturbations presented in Table I.

Figure S3. (Color online) $\lambda_2$ (left panel) and performance measure $(\langle C_\infty \rangle_1)$ for a box perturbation with $\tau_0 = 50/b_0$ (center panel) and $\tau_0 = 0.1/b_0$ (right panel) as a function of the number of nodes for the star network (red) and the cyclic network (blue). Solid lines give Eq. (13a) of the main text. The dashed lines give the two limits of Eq. (13a) : $\lambda_\alpha \tau_0 \ll 1$ (center panel) and $\lambda_\alpha \tau_0 \gg 1$ (right panel)

Graph 1

$Kf_1 : 131.68$

$Kf_2 : 70.45$

$\lambda_2 : 0.834$

Graph 2

$Kf_1 : 134.86$

$Kf_2 : 73.83$

$\lambda_2 : 0.954$

Graph 3

$Kf_1 : 134.2$

$Kf_2 : 76.53$

$\lambda_2 : 0.835$

Figure S4. Starting from a homogeneous cyclic network with first and second nearest neighbor couplings and size $n = 20$, graphs 1, 2 and 3 have been obtained after rewiring some randomly chosen edges.
BEYOND KURAMOTO

Instead of the Kuramoto model defined by Eq. (1) in the main text, one may consider other models of coupled dynamical systems. Two extensions have to be differentiated. First, one may consider different coupling than the sine-coupling in the Kuramoto model. This is straightforwardly included in our approach and leads only to a differently weighted Laplacian matrix, giving, for a fixed network, a different Lyapunov spectrum and different eigenmodes \( u_\alpha \), but leaving all expressions for the performance measures unchanged. Second, one may consider dynamical systems with more internal degrees of freedom, such as the one considered by Pecora and Carroll \([S5]\)

\[
\dot{x} = P(x) + B \otimes H(x), \quad (S36)
\]

with \( x = (x_1, x_2, ... x_n) \), \( P(x) = (f(x_1), f(x_2), ... f(x_n)) \), \( B \) the Laplacian matrix of the graph considered and \( H \) a function defining the coupling between adjacent dynamical systems with coordinates \( x_i \in \mathbb{R}^d \). Our approach assumes the existence of a synchronous state with \( x_1^{(0)} = x_2^{(0)} = ... = x_n^{(0)} \). Linearizing about the synchronous state with \( x = x^{(0)} + \delta x \) and considering a perturbation gives, instead of Eq. (2) in the main text,

\[
\delta \dot{x} = \delta P + [I \otimes DP(x^{(0)}) + B \otimes DH(x^{(0)})] \delta x, \quad (S37)
\]

where \( DP \) and \( DH \) are Jacobian matrices. The first term on the right-hand side is similar to the perturbation considered in the main text, and the third one is a generalization of the Laplacian term in Eq. (2) in the main text, where the network Laplacian is extended to take account of additional nodal degrees of freedom. The new second term occurs because \( P \) now depends on internal degrees of freedom \( x \) (it does not in the Kuramoto model). The formula given in the main text for the performance need now to be evaluated with the eigenvalues \( \Lambda_{\alpha,l} \) and eigenmodes \( U_{\alpha,l} \) of \( I \otimes DP(x^{(0)}) + B \otimes DH(x^{(0)}) \), \( l = 1, 2, ... d \).

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