Synchronization and entrainment of coupled circadian oscillators

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Circadian rhythms in mammals are controlled by the neurons located in the suprachiasmatic nucleus of the hypothalamus. In physiological conditions, the system of neurons is very efficiently entrained by the 24-hour light-dark cycle. Most of the studies carried out so far emphasize the crucial role of the periodicity imposed by the light dark cycle in neuronal synchronization. Nevertheless, heterogeneity as a natural and permanent ingredient of these cellular interactions is seemingly to play a major role in these biochemical processes. In this paper we use a model that considers the neurons of the suprachiasmatic nucleus as chemically-coupled modified Goodwin oscillators, and introduce non-negligible heterogeneity in the periods of all neurons in the form of quenched noise. The system response to the light-dark cycle periodicity is studied as a function of the interneuronal coupling strength, external forcing amplitude and neuronal heterogeneity. Our results indicate that the right amount of heterogeneity helps the extended system to respond globally in a more coherent way to the external forcing. Our proposed mechanism for neuronal synchronization under external periodic forcing is based on heterogeneity-induced oscillators death, damped oscillators being more entrainable by the external forcing than the self-oscillating neurons with different periods.

Keywords: Circadian oscillations; quenched noise; noise-induced oscillators death; modified Goodwin model; noise-induced synchronization.

1. Introduction

Circadian rhythms are light-dark dependent cycles of roughly 24 hours present in the biochemical and physiological processes of many living entities (Reppert & Weaver 2002). In mammals the main mediator between the light-dark periodicity and the biological rhythms is formed by two interconnected suprachiasmatic nuclei (SCN), located in the hypothalamus. These nuclei form the so called “circadian pacemaker” and contain about 10,000 neurons each (Reppert & Weaver 2002; Moore et al. 2002).

The main property of the SCN is that their activity displays self-sustained oscillations in synchrony with the external forcing imposed by the light-dark cycle. The exact mechanism leading to this behavior has been the subject of intense research. It has been shown that, when taken individually, neurons produce oscillations with a constant period ranging from 20 to 28 hours (Honma et al. 2004; Welsh et al. 1995). The oscillatory behavior originates in a regulatory circuit with a negative feedback loop. The relevant question is how this individual oscillatory behavior
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translates into common, global, oscillations of the SCN activity synchronized with the external light stimulus.

It has been shown that the origin of the oscillatory activity of the circadian pacemaker at the global level resides on the interaction between the SCN neurons. Coupling between cells in the SCN is achieved partly by neurotransmitters (Honma et al. 2004; Hastings & Herzog 2004) and it is by means of those neurotransmitters that external forcing by light influences the neuronal synchronization. For example the vasoactive intestinal polypeptide (VIP) has been shown to be necessary in mediating both the periodicity and the internal synchrony of mammalian clock neurons (Shen et al. 2000; Aton et al. 2005; Maywood et al. 2006). Therefore, a model of coupled and forced neurons appears quite naturally as responsible for the circadian rhythms. Along these lines, an interesting mechanism has been put forward recently by Gonze et al. (2005) and by Bernard et al. (2007). They proposed that synchronization to the external forcing is facilitated by the fact that interneuronal coupling transforms SCN into damped oscillators which can then be easily entrained.

In this paper we show that the presence of some level of heterogeneity or dispersion in the intrinsic periods of the oscillators (Schaap et al. 2003; Herzog et al. 2004) can improve the response of the coupled neuronal system to the external light-dark forcing. The proposed mechanism for the improvement of the neuronal synchronization under external periodic forcing bears some similarities with the one proposed in (Gonze et al. 2005; Bernard et al. 2007) in the sense that the oscillators are brought to a regime of oscillator death (Ermentrout 1990; Mirollo & Strogatz 1990), but in our case this regime is induced by the presence of heterogeneity. Once this regime has been reached, the damped oscillators are more entrainable by the external forcing than the self-oscillating neurons with different periods, or the synchronized oscillatory state which appears in the strong coupling regime but with a period larger than the individual neuronal periods.

To be more specific, we will assume that the periods of the individual neurons are random variables drawn from a normal distribution. We will then analyze the global response of the system to the light-dark cycle periodicity as a function of the interneuronal coupling strength, external forcing amplitude and neuronal heterogeneity. We show that the presence of the right amount of dispersion in the periods of the neurons can indeed enhance the synchronization to the external forcing.

Period dispersion arises as a consequence of the cellular heterogeneity at the biochemical level, which is an experimentally well observed fact (Aton & Herzog 2005; Honma et al. 2004). It can act in either physiological or pathological conditions. An example of the latter is the diversification of antigenic baggage present in tumor cells that makes them more difficult to be recognized and captured by the defense mechanisms and therefore more prone to migrate and develop metastasis (González-García et al. 2002). Our results show that some level of disorder can be of help when synchronizing neuronal activity to the external forcing. Although counterintuitive, it has been unambiguously shown that the addition of various forms of disorder can improve the order in the output of a large variety of nonlinear systems. For example, the mechanism of stochastic resonance (Gammaitoni et al. 1998; Hänggi & Marchesoni 2009) shows that the response of a bistable system to a weak signal can be optimally amplified by the presence of an intermediate level of dynamical noise. Stochastic resonance is not a rare phenomenon; it has been repeatedly shown to be relevant in physical and biological systems described by non-
linear dynamical equations (Gammaitoni et al. 1998; Hänggi & Marchesoni 2009). In large systems with many coupled elements, noise is responsible for a large variety of ordering effects, such as pattern formation, phase transitions, phase separation, spatiotemporal stochastic resonance, noise-sustained structures, doubly stochastic resonance, amongst many others (García-Ojalvo & Sancho 1999). All these examples have in common that some sort of order at the macroscopic level appears only in the presence of the right amount of noise or disorder at the microscopic level. Furthermore, it has been proven that noise may play a constructive role in nonlinear systems, by enhancing coherent (periodic) behavior near bifurcations and phase transitions (Neiman et al. 1997; Pikovsky & Kurths 1997). In this paper we introduce non-negligible random heterogeneity into the periods of all neurons, so-called quenched noise. Numerical simulations suggest (data not shown) that the results are valid as well when the quenched noise is introduced into the model parameters. A different approach is the consideration of intracellular stochastic variability due to low molecule numbers (Forger & Peskin 2005) or both variability and heterogeneity.

Close to our work is the study by Ueda et al. (2002), where the effect of fluctuations in neuron parameter values is assessed and it is shown that the coupled system is relatively robust to noise. Previous theoretical studies have addressed the effect of noise on genetic oscillators (Thattai & van Oudenaarden 2001; Steuer et al. 2003; Becskei et al. 2005), and some have proposed an ordering influence of noise on circadian clocks at the single cell level in cases where neither light intensity nor coupling strength by themselves can synchronize the system. Collective phenomena induced by heterogeneity in autonomous, non-forced systems, has also been discussed in the literature. For example (de Vries & Sherman 2001) and (Cartwright 2000) have shown that collective bursting or firing can appear in excitable systems and a general theory of the role of heterogeneity in those systems has been developed by (Tessone et al. 2007). In this paper, we refer to the collective response in systems of non-linear oscillators subjected to the action of an external forcing representing the day-light cycle.

The paper is organized as follows. In section 2 we will describe in detail the model of circadian oscillators and the methods we use. It is a coupled extension of the original Goodwin oscillator (Goodwin 1965) as developed by Gonze et al. (2005). In section 3 we analyze the system response to the periodic external forcing, as a function of the external forcing amplitude, coupling strength and neuronal diversity or heterogeneity. By simulating numerically the governing differential equations we identify the range of these parameters for which the extended system oscillates in synchrony and entrained to the external light period. Section a describes the mechanism through which the neuronal heterogeneity favors the synchronization with the external forcing and analyzes the combined influence of the coupling strength, neuronal heterogeneity and light amplitude on the stability of the linearized system of coupled oscillators. We show that a mean variable in this model exhibits a transition from a rhythmic to an arrhythmic dynamics (the so-called oscillator death (Ermentrout 1990; Mirollo & Strogatz 1990)). Concluding remarks are found in section 4.
2. Model and methods

(a) The circadian pacemaker

As stated in the introduction, our aim is to consider the role that the heterogeneity in the population of neurons plays in the global response of the SCN to an external oscillating stimulus. To this end, we consider an ensemble of coupled neurons subject to a periodic forcing. Each of the neurons, when uncoupled from the others and from the external stimulus, acts as an oscillator with an intrinsic period. Heterogeneity is considered insofar the individual periods are not identical, but show some degree of dispersion around a mean value. For each one of the neurons in the SCN we use a four-variable model proposed by Gonze et al. (Gonze et al. 2005), which is based originally on the Goodwin oscillator (Goodwin 1965), to describe circadian oscillations in single cells. The variables of the model are as follows: The clock gene mRNA (X) produces a clock protein (Y), which activates a transcriptional inhibitor (Z) and this in turn inhibits the transcription of the clock gene, closing a negative feedback loop. The mRNA X also excites the production of neurotransmitter V, which in the coupled system will be then the responsible of an additional positive feedback loop. In order to overcome the high Hill coefficients required for self-oscillations, Gonze et al. replaced the linear degradation by nonlinear Michaelis-Menten terms. This leads to the system of equations:

\[
\begin{align*}
\frac{dX}{dt} &= \nu_1 \frac{K_1^4}{K_1^4 + Z^4} - \nu_2 \frac{X}{K_2 + X}, \\
\frac{dY}{dt} &= k_3X - \nu_4 \frac{Y}{K_4 + Y}, \\
\frac{dZ}{dt} &= k_5Y - \nu_6 \frac{Z}{K_6 + Z}, \\
\frac{dV}{dt} &= k_7X - \nu_8 \frac{V}{K_8 + V},
\end{align*}
\]

which, depending on parameters, might produce oscillations in a stable limit cycle. Using the values \( \nu_1 = 0.7 \ \text{nM/h}, \nu_2 = \nu_4 = \nu_6 = 0.35 \ \text{nM/h}, \nu_8 = 1 \ \text{nM/h}, \)
\( K_1 = K_2 = K_4 = K_6 = K_8 = 1 \ \text{nM}, k_3 = k_5 = 0.7/h, k_7 = 0.35/h, \) the period of the limit cycle oscillations is \( T = 23.5 \ \text{h}. \)

For the complete model, we take \( N \) neuronal oscillators, each one of them described by four variables \( (X_i, Y_i, Z_i, V_i), i = 1, \ldots, N, \) satisfying the above evolution equations. Heterogeneity in the intrinsic periods is introduced by multiplying the left-hand-side of each one of the equations (2.1–2.4) by a scale factor \( \tau_i. \) Hence, the intrinsic period \( T_i \) of the isolated neuron \( i \) is \( \tau_i T. \) The numbers \( \tau_i \) are independently taken from a normal random distribution of mean 1 and standard deviation \( \sigma. \) Since the periods must be positive, in the numerical simulations we have explicitly checked that, for the values of \( \sigma \) considered later, \( \tau_i \) never takes a negative value, which would be unacceptable. The standard deviation \( \sigma \) will be taken as a measure of the diversity. A value of \( \sigma = 0.1 \) for example corresponds to a standard deviation of 10% in the individual periods of the uncoupled neurons, close to the observed variation of periods between 20 and 28 hours.

Two additional factors influence the dynamics of single cell oscillations: forcing by light and intercellular coupling. Both are assumed to act independently from the
negative feedback loop and are added as independent terms in the transcription rate of $X$ (Gonze et al. 2005). Light is incorporated through a periodic time-dependent function $L(t)$, which can be realized in various forms. In the majority of the presented results we have used a sinusoidal signal, $L(t) = \frac{L_0}{2} (1 + \sin \omega t)$. In some cases, for comparison and to simulate different day lengths, we have used a square wave $L(t) = \begin{cases} L_0, & \text{if } (t \mod 24h) < t_{\text{light}} \\ 0, & \text{otherwise} \end{cases}$. In both ways the signal oscillates between the values $L(t) = 0$ and $L(t) = L_0$ with a period $2\pi/\omega = 24h$.

Coupling between the neurons is assumed to depend on the concentration $F$ of the synchronizing factor (the neurotransmitter) in the extracellular medium, which builds-up by contributions from all neurons. Under a fast transmission hypothesis, the extracellular concentration is assumed to equilibrate to the average, mean-field, cellular neurotransmitter concentration, $F = \frac{1}{N} \sum_{i=1}^{N} V_i$. The resulting model is:

$$\tau_i \frac{dX_i}{dt} = \nu_1 \frac{K_4^4 X_i}{K_2 + X_i} - \nu_2 Y_i \frac{K_4^4 + K_4 Z_i}{K_2 + X_i} + \nu_6 Y_i \frac{K_4^4 + K_4 Z_i}{K_c + K_4 F} + L(t) \quad (2.5)$$

$$\tau_i \frac{dY_i}{dt} = k_3 X_i - \nu_4 Y_i \frac{K_4^4 + K_4 Z_i}{K_2 + X_i} \quad (2.6)$$

$$\tau_i \frac{dZ_i}{dt} = k_5 Y_i - \nu_6 Z_i \frac{K_4^4 + K_4 Z_i}{K_2 + X_i} \quad (2.7)$$

$$\tau_i \frac{dV_i}{dt} = k_7 X_i - \nu_8 V_i \frac{K_4^4 + K_4 Z_i}{K_2 + X_i} \quad (2.8)$$

$$F = \frac{1}{N} \sum_{i=1}^{N} V_i \quad (2.9)$$

with $\nu_c = 0.4 \text{nM/h}$, $K_c = 1 \text{nM}$.

There is experimental evidence supporting the assumption of a chemical (rather than electrical) mechanism of inter-cell communication among SCN neurons as a synchronization factor and, in fact, mechanisms other than neurotransmitters or electrical coupling for the SCN communication have been suggested (e.g. by Pol & Dudek (1993)). Furthermore, more realistic modeling which takes into account all variables known to participate of the negative feedback loop has been introduced. These models may include up to 10 variables and corresponding equations for each single cell (Bernard et al. 2007).

It seems, however, that in order to get understanding of the SCN dynamics, a sufficient tool is the 4 variable model described above. In fact, the synchronization of damped oscillators is independent from the particular intracellular model used and as discussed by (Bernard et al. 2007), this system, the model developed by (Leloup & Goldbeter 2003), and other simple negative feedback oscillators have similar synchronization properties. In this paper we have decided to use the simpler 4-variable model although most of our results are also valid in the more complex 10-variable model.

A model close to (2.5–2.9) has been used by Ullner et al. (2009), where the authors investigate how the interplay between fluctuations of constant light and intercellular coupling affects the dynamics of the collective rhythm in a large ensemble of non-identical, globally coupled oscillators. In their case, however, an inverse de-
dependence of the cell-cell coupling strength on the light intensity was implemented, in such a way that the larger the light intensity the weaker the coupling.

(b) Measures of synchrony and entrainment

Due both to the effect of coupling and of forcing, the neurons might synchronize their oscillations. There are several possible measures of how good this synchronization is. In this paper, the interneuronal synchronization will be quantified by the parameter of synchrony $\rho$, defined as

$$\rho = \sqrt{1 - \frac{\sum_{i=1}^{N} (V_i(t) - F(t))^2}{\sum_{i=1}^{N} V_i(t)^2}} = \sqrt{\frac{\langle F(t)^2 \rangle}{\langle \sum_{i=1}^{N} V_i(t)^2 \rangle}},$$ (2.10)

where $\langle \ldots \rangle$ denotes a time average in the long-time asymptotic state. The parameter $\rho$ varies between a value close to 0 (no synchronization) and 1 (perfect synchronization, with all neurons in phase, $V_i(t) = V_j(t)$, $\forall i, j$). It is important to note that even if the neurons synchronize perfectly their oscillations, the period of those oscillations does not necessarily coincide with the mean period $T$ of the individual oscillators or with the period $2\pi/\omega$ of the external forcing. In fact, in the unforced (no light) case, the period of the common oscillations (for the set of parameters given before and a dispersion of $\sigma = 0.05$ and coupling $K = 0.5$) is approximately equal to 26.5 h whereas the period of the forcing is $2\pi/\omega = 24$ h and the mean period of the individual uncoupled oscillators is $T = 23.5$ h (Gonze et al. 2005).

Besides the previous measure of synchronization amongst the oscillators, we are also concerned about the quality of the global response of the neuronal ensemble to the external forcing $L(t)$. A suitable measure of this response can be defined using the average gene concentration,

$$X(t) = \frac{1}{N} \sum_{i=1}^{N} X_i(t),$$ (2.11)

and computing the so-called spectral amplification factor $R$ (Gammaitoni et al. 1998),

$$R = \frac{4}{L_0^2} \left| \langle e^{-i\omega t} X(t) \rangle \right|^2.$$ (2.12)

$R$ is nothing but the normalized amplitude of the Fourier component at the forcing frequency $\omega$ of the time series $X(t)$. We will show that, under some circumstances, the response $R$ will increase with the intrinsic diversity $\sigma$ and that the period of the oscillations at the global level coincides with that of the external forcing, these being the main results of this paper.

3. Results

The synchronization properties of the set of circadian oscillators is influenced by the amplitude of the external forcing $L_0$, the coupling strength $K$ and the diversity in the individual periods $\sigma$. The role of the first two has been studied in (Bernard
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Figure 1. Synchrony order parameter $\rho$ (see Eq. (2.10)). Values are coded in colour levels, and displayed as a function of $L_0$ and $\sigma$ for several values of $K$. Data from numerical simulations of $N = 1000$ neurons with dynamics ruled by Eqs. (2.5–2.9). Synchrony among the neurons (yellow region) is favored by strong or very weak light intensity $L_0$, low diversity $\sigma$ and large coupling $K$. The thick black line is the linear stability limit discussed in section a (see also Fig. 7).

et al. 2007; Becker-Weimann et al. 2004; Gonze et al. 2005). In this section we focus on the heterogeneity of neuronal periods and analyze the combined influence of $L_0$, $K$ and $\sigma$ on the different parameters quantifying interneuronal synchronization and response to the forcing.

Fig. 1 shows colour plots of the parameter of synchrony $\rho$ as a function of the diversity $\sigma$ and the light intensity $L_0$, for different values of the coupling strength $K$. High values of the light intensity $L_0$ favor interneuronal synchrony. Also in agreement with its intuitive disordering role, high neuronal diversity leads to a low synchrony parameter $\rho$ in several parts of the diagrams. However, there is a region of values of $L_0 \in [0, L_{max}]$ for which there is a non-monotonous dependence of the synchrony order parameter with respect to the diversity. This can be seen more clearly in panel (a) of Fig. 2 where we plot $\rho$ as a function of diversity $\sigma$ for fixed values of $K = 0.6$ and $L_0 = 0.005$. $\rho$ first decreases by increasing $\sigma$ within the interval $0 \leq \sigma \leq 0.05$, but then it develops a maximum. The range of values of $L_0$ for which this non-monotonous behavior is observed depends on the coupling constant $K$: the larger $K$, the larger $L_{max}$.

As stated before, the fact that neurons synchronize amongst themselves does not mean that they synchronize to the forcing by light. To study this point, we have computed the individual periods $T_i$, $i = 1, \ldots, N$, of the oscillators in the ensemble. In those cases in which the concentrations do not oscillate with exact periodicity, we still define the period as the average time between maxima of the dynamical variables. In Fig. 4 we plot the mean value $\bar{T} = \frac{1}{N} \sum_{i=1}^{N} T_i$ as a function of $\sigma$ and

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Figure 2. Main parameters used for characterizing the synchronization of circadian oscillators as a function of the variance $\sigma$. (a) the synchrony parameter $\rho$; (b) the mean $\bar{T}$ of the individual periods $T_i$; (c) the response order parameter $R$; (d) the maximum real part of the eigenvalues of the linearized system.

$L_0$ for different values of $K$. As the dispersion in $T_i$ is small, it turns out that $\bar{T}$ is close to the period of the average variable $X(t)$.

Although, by construction, individual neurons have periods that fluctuate around $T = 23.5$ h, it turns out that the period of the resulting synchronized oscillations that occur in the unforced but coupled ($L_0 = 0, K > 0$) case, increases with increasing coupling $K$. For example, $\bar{T} \approx 30$ h for $K = 0.6$, mostly independent of the value of $\sigma$. As the forcing sets in, at low values of the coupling strength, the mean period is now $\bar{T} = 24$ h for all values of $L_0$ and $\sigma$. As the coupling between neurons increases, larger values of $L_0$ and/or $\sigma$ are needed in order for the mean period to coincide with that of the external forcing. An important feature that emerges from these plots is that for low light intensity it is possible to achieve a mean period of 24 h by increasing the neuronal diversity. For example, in the areas at the left of the different panels of Fig.4, or in panel (b) of Fig.2 corresponding to the case...
Figure 3. Characterization of synchrony for a light signal of square wave form. The synchrony parameter $\rho$, the mean $\bar{T}$ of the individual periods $T_i$ and the response order parameter $R$ (from top to bottom) measured for square wave stimuli of various day lengths (8h, 12h and 16h, from left to right).

$K = 0.6$, while identical neurons have periods of $\approx 30$ h, increasing $\sigma$ induces an adjustment of the period to 24 h. The transition between $\bar{T} = 24$ h and $\bar{T} \neq 24$ h is rather sharp, specially for large $K$. This is a clear manifestation that diversity indeed is able to improve the response to the external forcing. The same conclusion about the constructive role of diversity can be reached by looking at the measure of response $R$ (see Figs. 5 and 2(c)). These figures show that there is a region in parameter space in which the system response to the periodic light forcing displays a maximum value as a function of diversity $\sigma$. This indicates that it is possible to improve neuronal synchronization to the daily-varying light input by taking $\sigma$ close to an optimal value. Too small or too large diversity will not yield an optimal response. This is a clear manifestation that diversity indeed is able to improve the response to the external forcing.

A complementary perspective on this constructive role of diversity is attained looking at spectral amplification factor, $R$, from Eq. (2.12). This is a normalized measure of the amplitude of the oscillation of the neuronal system at the frequency of the daily forcing. Figures 5 and 2(c) show that there is a region in parameter space in which the system response to the periodic light forcing displays a maximum value as a function of diversity $\sigma$. In fact this maximum is very large as compared with the $R$ value at zero diversity, so that one can say that one of the most noticeable effects of a non-vanishing neuronal diversity is to give the system the capacity to respond efficiently to the 24h forcing in situations of small or no response at this frequency in the absence of diversity (the non-diverse neuronal ensemble could be oscillating at a different frequency, as revealed by high values of $\rho$). In summary,
it is possible to largely improve neuronal synchronization to the daily-varying light input by taking $\sigma$ close to an optimal value. Too small or too large diversity will not yield an optimal response at this frequency, although the response is generally larger than for zero diversity.

An external signal of square wave form and with different day lengths lead to similar results. As can be seen in figure 3 the response $R$ to the external signal passes through a maximum at an intermediate value of diversity. The mean period and the synchrony parameter behave as in the case with a pure sinusoidal as the driving force. Furthermore, the qualitative result is independent of the chosen day length.

(a) Diversity-induced oscillator death

Why does an increase in the diversity of the oscillators lead to an improved response to the external forcing? We argue that the main effect of the increase of the diversity is to take the oscillators into a regime of oscillator death (Ermentrout 1990; Mirollo & Strogatz 1990) in which they can be easily entrained by the varying part of the forcing. To understand this mechanism we first split the forcing into a constant (the mean) and a time varying part: $L(t) = L_0 + \frac{L_0}{2}\sin(\omega t)$. Taking only the constant part, $L(t) = \frac{L_0}{2}$, Figs. 6(a)–(c) show that the oscillators go from self-sustained oscillations to oscillator death, i.e. the amplitude of the self-sustained oscillations decreases, as $\sigma$ increases. Once oscillators are damped, they would re-
Figure 5. Colour plots of the spectral amplification factor $R$ as defined in Eq. (2.12) in logarithmic scale. Too high light intensity $L_0$ and too much diversity $\sigma$ lower the response of $X$ to the external frequency. This also happens for low light and low diversity (where the neurons oscillate at a frequency $\neq 24\,\text{h}$, see Fig. 4). Between both limits $R$ passes a local maximum.

spond quasi-linearly to periodic forcing, at least if this forcing is not too large, and linear oscillators always become synchronized to the external forcing, independently of their internal frequency. This is consistent with what is seen in figures 6(d)–(f), where the neurons in the case of low heterogeneity oscillate synchronously with each other, but their common period is larger than the one of the light forcing. Only when diversity brings the neurons to oscillator death can all of them be entrained to the period of the forcing signal. The mechanism is related to the one discussed by Gonze et al. (2005) and Bernard et al. (2007), but here we stress that neuron heterogeneity, as opposed to internal neuron parameters and couplings, is enough to damp the collective neuron oscillations and bring the system to a non-oscillating state where it can be more easily entrained. It is interesting to note that it has been shown experimentally for fruitflies that only a subset of the pacemaker neurons sustain cyclic gene expression after changing the laboratory light conditions to constant darkness, whereas the oscillations of the other pacemaker neurons are damped out (Veleri et al. 2003). Although this does not reveal the mechanism by which the oscillations die out it suggests that some of the circadian oscillators do indeed work in the damped regime, at least in Drosophila.

An alternative way of checking this mechanism based on diversity-induced oscillator death is by analyzing the stability of the steady state of the system of Eqs. (2.5–2.9) when considering a constant forcing $L(t) = \frac{L_0}{T}$. The numerical calculation of the fixed point of the dynamics is greatly simplified by the fact that the concentrations of the biochemical variables are the same for each one of the $N$ neu-
Figure 6. Figures (a), (b) and (c) represent the time-dependent amplitude of the $V_i$ variable for a few selected neurons in the presence of constant light and increasing $\sigma$, while Figures (d), (e) and (f) represent the amplitude of the same neurons with sinusoidal light and increasing $\sigma$. The thin line on the bottom of the graphs is the external light signal. $K = 0.6$.

In summary, increasing the diversity or the (constant) forcing term decreases (on average) the maximum eigenvalue of the coupled system and thus a Hopf bifurcation can be crossed backwards, so that self-oscillations disappear. When applying the periodic external forcing on the system formed by self-sustained neurons, coherence with the external frequency is difficult to achieve because there is the competing effect of mutual neuron synchronization to a different frequency. However, when the periodic external forcing is applied on the system of damped neurons, they all synchronize to the external forcing, and thus with each other since this is the only dynamical regime available to forced damped oscillators (if forcing is not too...
strong to excite further resonances). Increased coupling strength increases the range of unentrained self-oscillations.

Oscillator death by diversity is not particular to this system. In (Mirolo & Strogatz 1990) the authors analyze a large system of limit cycle-oscillators with mean field coupling and randomly distributed frequencies. They proved that when the coupling is sufficiently strong and the distribution of frequencies has a sufficiently large variance, the system undergoes “amplitude death”. In their approach the oscillators pull each other off their limit cycles, which is translated into a stable equilibrium point for the coupled system. Thus, this mechanism suggests that the quenched noise we introduced in the system “pushes apart” the limit cycles of the different neurons, so that their competition enlarges the range of parameters where fixed point behavior is stable.

A qualitative argument explaining the diversity-induced oscillator death in our system of coupled neurons goes as follows: We know from Gonze et al. (2005) that a single oscillator can switch from a limit cycle to a stable steady state by adding a constant mean field (the term containing $F$ in (2.5) but with time-independent $F$) of sufficient strength to Eq. (2.1). A constant light forcing term has the same effect (see the zero coupling case in fig. 7). Furthermore we have observed that the amplitude of the oscillations decreases with rising diversity (compare figs. 6), but the mean does not change. In a system with low diversity we have large oscillations of $F$ around that mean value. If this value, taken as a constant, determines a stable steady state, then we argue that the large oscillations lead the system into unstable regions, whereas, by increasing $\sigma$ the amplitude is decreased and the concentrations do not leave neighbourhood of the stable fixed point, thus finding themselves damped all the time. This is a possible mechanism for the diversity-induced oscillator death phenomenon.

4. Concluding Remarks

In this work we have analyzed the role of diversity in favoring the entrainment of a system of coupled circadian oscillators. We introduce non-negligible heterogeneity in the periods of all neurons in the form of quenched noise. This is achieved by rescaling the individual neuronal periods by a scaling factor drawn from a normal distribution. The system response to the light-dark cycle periodicity is studied as a function of the interneuronal coupling strength, external forcing amplitude and neuronal heterogeneity.

Most of the cases of order induced by heterogeneity or noise carried out so far (Gammaitoni et al. 1998; Hänggi & Marchesoni 2009; Tessone et al. 2006, 2007; Toral et al. 2009; Pikovsky & Kurths 1997; Ulher et al. 2009), emphasize the fact the diversity directly improves oscillator synchronization. In our case the mechanism is rather different. Diversity does not improve system synchronization directly. This is achieved indirectly, by a leading first to a diversity-induced stabilization of the fixed points of the neurons forming the system. Once steady concentrations are asymptotically stable, it is much better entrainable by the external forcing, so that the damped neurons adapt easily to the external forcing (and then, in addition, they appear as synchronized between them). The synchronization arises therefore not as a result of a direct diversity-induced neuronal synchronization but indirectly, as a result of the diversity-induced oscillator death. Our results indicate therefore that
the right amount of heterogeneity helps the extended system to respond globally in a more coherent way to the external forcing. In addition to the robustness of the results against use of non-sinusoidal forcing we have checked that resonances in the responses to the external forcing and matching of the circadian period to the external forcing appear in more complex models, such as the 10-variable model of (Bernard et al. 2007) with diversity in the time scales $\tau_i$, or the 4-variable model of (Gonze et al. 2005) with heterogeneity in all the reaction rate parameters $\nu_i$.

We expect that a similar behavior will be found in models of non-mammalian clocks like those of Drosophila (Smolen et al. 2004), Arabidopsis (Locke et al. 2005), Neurospora (Heintzen & Liu 2007) or Cyanobacteria (Dong & Golden 2008).

Of course, it is an open question whether the observed diversity in the periods of the neurons of the SCN has been tuned by evolution in order to display a maximum response to the 24 h dark-light natural cycle. A detailed experimental check of our predictions would require to be able to vary the amount of diversity. In this sense we suggest the possibility of using chimeric organisms (Low-Zeddies & Takahashi 2001) to introduce diversity in a controlled way.

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