Mechanisms of Gamma Oscillations

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Abstract

Gamma rhythms are commonly observed in many brain regions during both waking and sleep states, yet their functions and mechanisms remain a matter of debate. Here we review the cellular and synaptic mechanisms underlying gamma oscillations and outline empirical questions and controversial conceptual issues. Our main points are as follows: First, gamma-band rhythmogenesis is inextricably tied to perisomatic inhibition. Second, gamma oscillations are short-lived and typically emerge from the coordinated interaction of excitation and inhibition, which can be detected as local field potentials. Third, gamma rhythm typically concurs with irregular firing of single neurons, and the network frequency of gamma oscillations varies extensively depending on the underlying mechanism. To document gamma oscillations, efforts should be made to distinguish them from mere increases of gamma-band power and/or increased spiking activity. Fourth, the magnitude of gamma oscillation is modulated by slower rhythms. Such cross-frequency coupling may serve to couple active patches of cortical circuits. Because of their ubiquitous nature and strong correlation with the “operational modes” of local circuits, gamma oscillations continue to provide important clues about neuronal population dynamics in health and disease.

Keywords

inhibitory interneurons, interneuronal network, excitatory-inhibitory loop, spike timing, dynamical cell assembly, irregular spiking, cross-frequency coupling, long-distance communication
INTRODUCTION

The precise timing of neuronal-spike discharges is believed to be important for coding of information (O’Keefe & Recce 1993, Buzsáki & Chrobak 1995, Singer & Gray 1995, Singer 1999). The ability of various neuron types to time their action potentials with millisecond precision depends largely on the presence of fast membrane potential fluctuations (Mainen & Sejnowski 1995, Haider & McCormick 2009). In the intact brain, such high-frequency patterns are often brought about by various endogenous oscillations, the most ubiquitous of which are rhythms in the gamma-frequency range (30–90 Hz) (see Origin and Definition of Gamma Oscillation, sidebar below).


ARE CELL ASSEMBLIES DYNAMICALLY ORGANIZED IN GAMMA CYCLES?

To appreciate the physiological function of the gamma cycle in neural networks, we need to examine the spiking patterns of neurons at this timescale. The exact timing of neuronal spikes can be related to environmental stimuli, overt behavior, local field potential (LFP), or spiking activity of other neurons. Each of these comparisons provides a different “optimum” time window. The best prediction is obtained when information about the spike times of partner neurons are available in the 10- and 30-ms window (Figure 1) (Jensen & Lisman 1996, Borgers & Kopell 2003, Harris et al. 2003, Lisman 2005), i.e., the time window corresponding approximately to a gamma cycle. Neuronal assemblies, i.e., transient neuronal partnerships, can be active repeatedly in successive gamma cycles, or different assemblies can alternate in a rapid sequence.

The gamma-cycle-related lifetime of the cell assembly is closely related to several biophysical properties of neurons, including the time constant of gamma-aminobutyric acid (GABA)\(\alpha\) and \(\alpha\)-amino-3-hydroxy-5-methyl-4-isoxazolepropionic acid (AMPA) receptors (Johnston & Wu 1994), the membrane time constant of cortical pyramidal cells (Destexhe & Paré 1999, Leger et al. 2005), and the critical time window of spike-timing-dependent plasticity (Magee & Johnston 1997, Markram...
et al. 1997). Because these parameters determine the neuron’s ability to integrate inputs from multiple upstream sources, a hypothesized functional role of the cell assembly is to bring together sufficient numbers of peer neurons so that their collective spiking can discharge the postsynaptic neuron (Harris et al. 2003). Consequently, from the point of view of the downstream (“reader” or “integrator”) cell, ensemble activity of upstream neurons whose spikes occur within the gamma-cycle window is classified as a single event (Buzsáki 2010). Upstream neurons whose spikes fall outside this time window become part of another transient assembly.

MODELS OF GAMMA OSCILLATIONS

The similar kinetics of gamma-frequency oscillations in a variety of different brain regions and species have provided clues and constraints about the requirements of their supporting mechanisms. Gamma oscillations have been described in several areas of the neocortex (Gray et al. 1989, Murthy & Fetz 1992, Fries et al. 2001, Sirotà et al. 2008), entorhinal cortex (Chrobak & Buzsáki 1998), amygdala (Halgren et al. 1977, Popescu et al. 2009), hippocampus (Buzsáki et al. 1983, Bragin et al. 1995, Whittington et al. 1995, Mann et al. 2005), striatum (Berke et al. 2004, Tort et al. 2008), olfactory bulb (Adrian 1942, Freeman 1975), and thalamus (Pinault & Deschênes 1992) as well as other areas. Common denominators of these brain regions are the presence of inhibitory interneurons and their actions through GABAA synapses. Synchronization of neurons is substantially more effective by perisomatic inhibitory postsynaptic potentials (IPSPs) than dendritic excitatory (E)PSPs (Lytton & Sejnowski 1991). From these considerations, it is reasonable to assume that a key ingredient of gamma oscillations is GABAA receptor-mediated inhibition.

I-I Model

Only three requirements are needed for gamma oscillations to emerge, as illustrated by a “stripped-down” network model consisting of only inhibitory interneurons (Figure 2a) (Wang & Rinzel 1992, Whittington et al. 1995, Wang & Buzsáki 1996, Traub et al. 1996b): mutually connected inhibitory interneurons, a time constant provided by GABAA receptors, and sufficient drive to induce spiking in the interneurons. Gamma oscillations in inhibitory-inhibitory (I-I) neuron models can emerge in two different ways (see Irregular Activity of Single Neurons and Gamma Oscillations of Neuron Groups, sidebar below). When the input drive is relatively tonic, neurons can fire spikes with a well-defined periodicity (Figure 2a) (Kopell & Ermentrout 2002). By contrast, when neurons receive stochastic inputs and fire spikes irregularly, sufficiently strong recurrent synaptic interactions will make the asynchronous state unstable against random fluctuations, and oscillations emerge (Figure 2b) (Brunel & Hakim 1999, Brunel 2000, Brunel & Wang 2003, Geisler et al. 2005, Ardid et al. 2010, Economo & White 2012). In both cases, the emerging synchrony
Figure 1
Dynamical cell assemblies are organized in gamma waves. (a) Raster plot of a subset of hippocampal pyramidal cells that were active during a 1-s period of spatial exploration on an open field out of a larger set of simultaneously recorded neurons, ordered by stochastic search over all possible orderings to highlight the temporal relationship between anatomically distributed neurons. Color-coded ticks (spikes) refer to recording locations shown in panel b. Vertical lines indicate troughs of theta waves (bottom trace). Cell-assembly organization is visible, with repeatedly synchronous firing of some subpopulations (circled). (c) Spike timing is predictable from peer activity. Distribution of timescales at which peer activity optimally improved spike-time prediction of a given cell, shown for all cells. The median optimal timescale is 23 ms (red line). Based on Harris et al. (2003).

is caused when a subset of the interneurons begins to discharge together and generates synchronous IPSPs in the partner neurons. In turn, the inhibited neurons will spike again with increased probability when GABA\textsubscript{A} receptor–mediated hyperpolarization has decayed, and the cycle repeats (Figure 2a,b). Because the duration of IPSCs (inhibitory postsynaptic current) is determined by the subunit composition of the GABA\textsubscript{A} receptor (cf. Farrant & Nusser 2005), the frequency of gamma oscillations in the I-I model is determined mainly by the kinetics of the IPSPs and the net excitation of interneurons (Whittington et al. 1995, Wang & Buzsáki 1996).

In vitro experiments provided support for the sufficient role of mutual inhibition among interneurons for the generation of gamma rhythm, for instance sustained by activation of metabotropic glutamate receptors (Whittington et al. 1995). Gamma oscillations can be induced by other means as well, such as activation of muscarinic-cholinergic receptors (Fisahn et al. 1998) or kainate receptors (Fisahn et al. 2004, Hájos & Paulsen 2009). Common to all these conditions is the increased firing of synaptically coupled interneurons. When pyramidal cells and other interneuron types are added to the I-I model network, the entire network can become phase-locked to the gamma oscillations.
E-I Model

The earliest model of gamma oscillations is based on the reciprocal connections between pools of excitatory pyramidal (E) and inhibitory (I) neurons (Wilson & Cowan 1972, Freeman 1975, Leung 1982, Ermentrout & Kopell 1998, Borgers & Kopell 2003, Brunel & Wang 2003, Geisler et al. 2005). In such two-neuron pool models (Figure 2c), fast excitation and the delayed feedback inhibition alternate, and with appropriate strength of excitation and inhibition, cyclic behavior may persist for a while. E-I models can also exhibit two distinct regimes, depending on whether single neurons behave periodically or highly stochastically. In the model, axon conduction and synaptic delays lead to a phase shift (∼5 ms or up to 90°) between the pyramidal and interneuron spikes, and these delays determine the frequency of the gamma rhythm (Freeman 1975, Leung 1982). An appeal of the E-I model is that the delay between the timing of pyramidal cell and interneuron spikes is a prominent feature of gamma oscillations both in vivo and in vitro (Figure 3) (Bragin et al. 1995, Csicsvari et al. 2003, Mann et al. 2005, Mann et al. 2005, Hajos & Paulsen 2009). In further support of the model, weakening the E-I connection by genetic knock down of AMPA receptors on fast spiking interneurons reduces the amplitude of gamma oscillations (Fuchs et al. 2007). The mainstream I-I and E-I models have been developed to explain gamma oscillations in the cortex but other gamma frequency oscillations may possibly arise from other mechanisms as well (Wang 1993, Gray & McCormick 1996, Wang 1999, Minlebaev et al. 2011).

IRREGULAR ACTIVITY OF SINGLE NEURONS AND GAMMA OSCILLATIONS OF NEURON GROUPS

A fruitful debate persists between researchers who study population gamma oscillations and ponder their functions, and researchers who study single-neuron data and observe that neuronal-spike trains are often irregular and by some measures approximate a Poisson process (Softky & Koch 1993). Recent work has offered a novel theoretical framework in which population rhythms can arise from irregularly firing neurons, thereby bridging these contrasting dynamical aspects of cortical dynamics (c.f., Wang 2010).

CELLULAR-NETWORK MECHANISMS OF GAMMA OSCILLATIONS

Perisomatic Inhibition Is Critical for Gamma Oscillations

The first support for the involvement of fast-spiking interneurons in gamma oscillations came from the correlation (spike-field coherence) between their spikes and locally recorded LFP gamma oscillations in the hippocampus of behaving rats (Figure 3) (Buzsáki et al. 1983). Putative fast-spiking interneurons and histologically verified parvalbumin (PV)-immunoreactive basket cells often show a broad peak in their autocorrelograms and spectrograms at gamma frequency (Figure 3b), and the occurrence of their spikes follows those of the surrounding pyramidal neurons by a few milliseconds (Figure 3d,f) (Bragin et al. 1995, Csicsvari et al. 2003, Mann et al. 2005, Hajos & Paulsen 2009), as in E-I models. As expected from the spike-LFP relationship (Figure 3a,c) postsynaptic potentials phase-locked to the LFP gamma rhythm are present in pyramidal neurons. These gamma-correlated postsynaptic potentials in pyramidal cells reverse their polarity close to the equilibrium potential of Cl− (Figure 3e,g), indicating that the gamma-rhythm-related inhibition is mediated by GABA_A receptors (Soltesz & Deschênes 1993, Whittington et al. 1995, Penttonen et al. 1998, Hasenstaub et al. 2005, Mann et al. 2005). The IPSPs paced by the PV basket cells produce coherent transmembrane fluctuations in the target pyramidal cell population (Penttonen et al. 1998, Gloveli et al. 2005, Hasenstaub et al. 2005, Mann et al. 2005, Quilichini et al. 2010) and can be detected as a strong current source in the cell-body layer (Figure 3d) (Csicsvari et al. 2003).
Figure 2
I-I and E-I models of gamma oscillations. (a) Clock-like rhythm of coupled oscillators in an interneuronal (I-I) population. (Upper panel) Single interneurons fire spikes periodically at ~40 Hz. Mutual inhibition via GABA_A receptors quickly brings them to zero-phase synchrony; (lower panel) two example neurons. Adapted from Wang & Buzsáki (1996). (b,c) Sparsely synchronous oscillations in a neural circuit where single neuronal spiking is stochastic. Adapted from Geisler et al. (2005). (b) Interneuronal population in noise-dominated regime typically exhibits gamma power in the higher frequency range, in contrast to (a) the clock-like rhythmic case. (c) Reciprocally connected E-I network where pyramidal cells send fast excitation via AMPA receptors to interneurons, which in turn provide inhibition via GABA_A receptors, leading to coherent oscillations in the gamma-frequency range.

Resonance:
phenomenon describing a neuron or a neural circuit that is maximally responsive to an oscillatory input at a preferred frequency.

et al. 2003, Mann et al. 2005). The interconnected PV-basket interneuron network with its divergent output to pyramidal cells provides an anatomical substrate for coherent timing of the pyramidal cells (Figure 3d) (Kisvárday et al. 1993, Buhl et al. 1994, Sik et al. 1995). Altogether, these findings support the hypothesis that extracellularly recorded gamma waves largely correspond to synchronous IPSPs in pyramidal cells, brought about by fast-spiking interneurons (Buzsáki et al. 1983, Bragin et al. 1995, Hasenstaub et al. 2005, Freund & Katona 2007, Hájos & Paulsen 2009).

Several other findings support the critical role of fast-spiking basket neurons in gamma oscillations. Basket cells have several distinctive features among the interneuron family, including (a) low spike threshold (Gulyás et al. 1993), (b) ability to fire rapidly without fatigue (Buzsáki et al. 1983, McCormick et al. 1985, Kawaguchi & Kubota 1997), (c) narrow spikes conferred by a large density of KV3.1/3.2 channels (Lien & Jonas 2003), (d) a unique spike-conductance trajectory (Tateno & Robinson 2009), and (e) resonance at gamma frequency in response to stochastic excitatory conductance inputs (Figure 4) (Pike et al. 2000, Cardin et al. 2009, Sohal et al. 2009). Overall, these findings support the hypothesis that gamma oscillations can be induced by activation of interconnected PV interneurons by multiple means.

The involvement of other interneuron types (Freund & Buzsáki 1996, Klausberger & Somogyi 2008) in gamma generation is understood less well. Chandelier cells are likely not critical in I-I models, because they innervate only principal cells. The somatostatin-containing O-LM interneurons and Martinotti cells mainly target distal dendrites, establish few connections among themselves (Gibson et al.
Figure 3

A critical role of parvalbumin (PV) basket cells in gamma oscillations. (a) Local field potential (LFP) recording from the CA1 pyramidal layer (top) and dentate hilus (bottom) and unit recording from a fast-spiking putative interneuron in the hilus (middle trace). Note the trains of spikes at gamma frequency, repeating periodically at theta frequency. (b) Power spectrum of the unit shown in panel a. Note the peak at theta and a broader peak at 50–80 Hz (gamma). (c) Spike-triggered average of the LFP in the hilus. Note the prominent phase locking of the interneuron to gamma wave phase and the cross-frequency coupling between gamma and theta waves. (d–g) Recordings from a behaving rat. (d) Camera-lucida reconstruction of the axon arbor of an immunocytochemically identified CA1 basket cell in vivo. The axon arbor outlines the CA1 pyramidal layer, showing (circles) putative contacts with other PV-positive neurons, (inset) averages of the intracellularly recorded V_m (membrane potential) and the LFP, (triangle) peak of the mean preferred discharge of the surrounding pyramidal cells, and (arrow) peak of the mean preferred discharge of the basket cell. Note the short delay between the spikes of pyramidal cells and the basket neuron. Current source density (CSD) map is superimposed on the pyramidal layer. Arrow points to current source of gamma wave (red). (e) Continuous display (110) of integrated and rectified gamma activity of the LFP and the fast intracellularly recorded V_m fluctuation (20–80 Hz; after digital removal of spikes) in a CA1 pyramidal neuron. V_m was biased by the intracellular current injection: (dashed line) resting membrane potential. Note the increase of the intracellular V_m gamma during both depolarization (inset) and hyperpolarization as well as the smallest V_m gamma power at resting membrane potential (asterisks) against the steady background of LFP gamma power. (d, e) In vivo recordings under urethane anesthesia. (f) Excitatory (E) and inhibitory (I) postsynaptic currents (PSCs) in a pyramidal cell, triggered by LFP gamma (top) and the spike timing of a pyramidal cell (P) and a basket interneuron (B) during carbachol-induced gamma oscillation in a hippocampal slice in vitro. Note that maximum discharge of the basket cell precedes the hyperpolarization of the pyramidal cell. (g) Intracellular recordings in a ferret prefrontal pyramidal cell in vivo illustrating the large amplitude, inhibition-dominated barrages recorded at 0 mV (brown) and smaller amplitude, excitation-dominated, synaptic barrages recorded at −80 mV (tan) for two representative UP states. Membrane potentials are expanded further (inset). EPSP, excitatory postsynaptic potential; IPSP, inhibitory postsynaptic potential. Reproduced with permission from (a–c) Buzsáki et al. (1983), (d–e) after Penttonen et al. (1998), (f) after from Mann et al. (2005), and (g) after Hasenstaub et al. (2005).
Figure 4

“Synthetic” gamma rhythm in vivo. (a) Local field potential (LFP) recordings in anesthetized mouse, expressing ChR2 selectively in either parvalbumin (PV) neurons (ChR2-PV-Cre) or pyramidal cells (ChR2-αCamKII-Cre). Stimulation at 8 Hz evoked rhythmic activity in the αCamKII-Cre but not the PV-Cre mouse. Conversely, stimulation at 40 Hz induced gamma oscillation in the PV-Cre but not in αCamKII-Cre mouse. (b) Mean LFP power ratio measured in multiple frequency bands in response to rhythmic light activation of ChR2-PV-Cre expressing neurons (blue) or ChR2-αCamKII-Cre expressing neurons (purple) at various frequencies. Reprinted from Cardin et al. (2009).

Do I-I and E-I Mechanisms Compete or Cooperate in the Brain?

Both I-I and E-I models have merits and disadvantages (Whittington et al. 2000, Tiesinga & Sejnowski 2009, Wang 2010). Because the oscillation frequency of individual neurons in the I-I model is at least partially determined by the amount of excitation, a heterogeneous input can result in a wide range of oscillation frequencies. In the face of such frequency dispersion, the population synchrony inevitably decreases. This shortcoming can be effectively compensated for by gap-junction-enhanced synchrony (Gibson et al. 1999, Hormuzdi et al. 2001, Buhl et al. 2003, Traub et al. 2004), resonant properties of basket cells, and fast and strong shunting inhibition between interneurons (Bartos et al. 2007). However, heterogeneity of neuronal firing rates may be beneficial. In networks consisting of neurons with different firing patterns and rates, gamma oscillation may function as a selection mechanism, because transient synchrony would emerge only among those neurons that are activated to approximately the same level.

In most E-I models, there is no need for I-I connections (Wilson & Cowan 1972, Whittington et al. 2000, Borgers & Kopell 2003, Brunel & Wang 2003, Geisler et al. 2005). In support of this prediction, experimentally disconnecting many I-I links in knockout mice did not strongly affect gamma power in the hippocampal CA1 region (Wulff et al. 2009). In the E-I models, the driving force of the oscillation is the activity of pyramidal cells. Note that gamma rhythms are also prominent in structures, which lack dense local E-I output (Csicsvari et al. 2003), but the IPSPs they produce in the dendrites of pyramidal cells may not be faithfully transferred to the soma (Lytton & Sejnowski 1991). These other types of interneurons appear better suited to contribute to slower oscillations and, by controlling basket cells, are likely critical in establishing cross-frequency coupling (see below) between gamma and slower rhythms.
connections, such as the basal ganglia or ventral tegmental area (Brown et al. 2002, Berke et al. 2004, Tort et al. 2008, Fujisawa & Buzsáki 2011). E-I models require a time delay between E spikes and I spikes, since timing of the interneurons is “inherited” from the pyramidal cells. In contrast, in I-I models the spike phase of pyramidal cells largely reflects the intensity of their tonic drive. In the hippocampal CA1 region, interneurons show both phase delay or advance relative to the spikes of pyramidal cells (Bragin et al. 1995, Csicsvari et al. 2003, Tü欺ker et al. 2007, Senior et al. 2008, Mizuseki et al. 2011). These results suggest that E-I and I-I hybrid gamma networks may work together to generate gamma frequency oscillations (Brunel & Wang 2003, Geisler et al. 2005, Tiesinga & Sejnowski 2009, Belluscio et al. 2012).

The role of recurrent excitatory (E-E) connections between principal cells in gamma models is not well-understood (Kopell et al. 2000, Whittington et al. 2000, Brunel & Wang 2003, Geisler et al. 2005). In the cortex, gamma oscillations are more prominent in the superficial, rather than the deep, layers where local recurrent connections are abundant (Chrobak & Buzsáki 1998, Quilichini et al. 2010, Buffalo et al. 2011). By contrast, the largest-amplitude gamma rhythm in the hippocampus is observed in the dentate gyrus (Buzsáki et al. 1983), even though granule cells lack recurrent excitation onto themselves. Decreasing recurrent excitatory synaptic currents in dynamic clamp studies had little effect on gamma power (Morita et al. 2008). The less critical role of E-E recurrent excitation may liberate the pyramidal cells from the timing constraints of the rhythm; therefore, they could fire spikes stochastically at various cycle phases in an input drive-dependent manner without interrupting rhythm.

**LONG-RANGE SYNCHRONIZATION OF GAMMA OSCILLATIONS**

Although gamma oscillations typically arise locally, patches of gamma networks can interact with each other. Synchronization of transient gamma bursts has multiple meanings, including phase-phase, phase-amplitude, and amplitude-amplitude coupling (Figure 5) (see Cross-Frequency Phase Coupling, sidebar below). Phase-phase synchrony between identical frequency oscillators that emerges at two (or multiple) locations can occur by phase locking (Figure 5b). The magnitude of such synchrony is typically measured by phase coherence. A second form of synchrony refers to the co-variation of gamma power at two (or multiple) locations, also known as amplitude or power comodulation (Figure 5c). In this latter case, phase constancy between the gamma waves may or may not be present (Figure 5c,d). Instead, the power (amplitude) envelopes of the gamma bursts are correlated (comodulation of power). Power-power synchrony of gamma rhythms can be effectively brought about by joint phase biasing of the power of gamma oscillations by a slower rhythm, known as cross-frequency phase-amplitude (CFPA) coupling or nested oscillations (Figure 5c,d,e) (Bragin et al. 1995, Schroeder & Lakatos 2009, Canolty & Knight 2010, Fell & Axmacher 2011). The third type of synchrony occurs when there is a relatively constant relationship between the gamma phase and the phase of a modulating slower rhythm (Figure 6c), known as cross-frequency phase-phase (CFPP), or n:m, coupling (Tass et al. 1998). Cross-frequency coupling can take place within or across structures. In practice, each relationship should be investigated with care because even stochastic signals can occasionally yield spurious coupling.

**Phase Coherence of Gamma Rhythms in Distant Networks**

If multiple cell assemblies in disparate brain areas need to be synchronized, how can they be engaged in coherent gamma oscillations given the long axon conduction delays of pyramidal cells? Solid evidence for coherent gamma oscillations in distant networks is scarce; perhaps the best-established case is interhemispheric synchronization. Multiple units with similar receptive fields in the left and right primary
Figure 5
Oscillatory coupling mechanisms. (a) Schematic view of the human brain showing hot spots of transient gamma oscillations (i–iv) and theta oscillation in the hippocampus (HI); entorhinal cortex (EC). Oscillators of the same and different kind (e.g., theta, gamma) can influence each other in the same and different structures, thereby modulating the phase, amplitude, or both. (b) Phase-phase coupling of gamma oscillations between two areas. Synthetic data used for illustration purposes. Coherence spectrum (or other, more specific, phase-specific measures) between the two signals can determine the strength of phase coupling. (c) Cross-frequency phase-amplitude coupling. Although phase coupling between gamma waves is absent, the envelope of gamma waves at the two cortical sites is modulated by the common theta rhythm. This can be revealed by the power-power correlation (comodugram; right). (d) Gamma phase-phase coupling between two cortical sites, whose powers are modulated by the common theta rhythm. Both gamma coherence and gamma power-power coupling are high. (e) Cross-frequency phase-phase coupling. Phases of theta and gamma oscillations are correlated, as shown by the phase-phase plot of the two frequencies. (f) Hippocampal theta oscillation can modulate gamma power by its duty cycle at multiple neocortical areas so that the results of the local computations are returned to the hippocampus during the accrual (“readiness”) phase of the oscillation. a and f, after Buzsáki (2010); b–e, after Belluscio et al. (2012).
Figure 6

Long-range synchrony of gamma oscillations. (a) Neurons sharing receptive fields in left (LH) and right (RH) primary visual cortex of the anesthetized cat fire coherently with zero time lag at gamma frequency. (b) Local field potential (LFP) traces from the left (L) and right (R) hippocampal CA1 pyramidal layer of the mouse during running and coherence spectra between the traces during running (orange) and REM sleep (blue). (c) LFP coherence map of gamma (30–90 Hz) in the rat hippocampus during running. Coherence was calculated between the reference site (star) and the remaining 96 recording sites. Note the high coherence values within the same layers (outlined by white lines) and rapid decrease of coherence across layers. (d) Distribution of distances between the unit and LFP recording sites with maximum spike-LFP coherence in the gamma band. Note that, in a fraction of cases, maximum coherence is stronger at large distances between the recorded unit and the LFP. (e) Spike-LFP coherence in the human motor cortex. The probability of spiking correlates with frequency-specific LFP phase of the ipsilateral (blue) and contralateral (green) motor area and contralateral dorsal premotor area (red). (f) The phase-coupling-based spike rate (generated from the preferred LFP–LFP phase-coupling pattern) predicts the measured spike rate. Panels reproduced after (a) Engel et al. (1991), (b) Buzsáki et al. (2003), (c) Montgomery & Buzsáki (2007), (d) Sirota et al. (2008), and (ef) Canolty et al. (2010).
visual cortex can display coherent gamma-range oscillations (Figure 6a). Similarly, gamma oscillations in homologous hippocampal layers in the two hemispheres display high coherence (Figure 6b). In both cases, phase synchrony is mediated by interhemispheric axon tracts, given that severing these conduits abolishes the synchrony. The high interhemispheric coherence and task-dependent inter-regional gamma synchrony (Engel et al. 1991, Roelfsema et al. 1997, Chrobak & Buzsáki 1998, Rodríguez et al. 1999, Tallon-Baudry et al. 2001, Montgomery et al. 2008) can be contrasted with the fast decrease of gamma coherence across different layers (Figure 6c), owing to the noncoherent relationships among the inputs.

The importance of anatomical connectivity, as opposed to physical distance, can explain the occasionally high gamma coherence between spikes and LFP at distant sites (Figure 6d) and the gamma timescale covariations of firing rates of spatially distant neurons (Figure 6e,f).

Temporal coordination between spatially separated oscillators can be established by axon collaterals of pyramidal cells (Traub et al. 1996a, Whittington et al. 2000, Bibbig et al. 2002), interleaving assemblies (Vicente et al. 2008), or long-range interneurons (Buzsáki et al. 2004). In each case, conduction delays are the primary problem because the differing delays between the different gamma inputs can destabilize the rhythm (Ermentrout & Kopell 1998), and the extra interneuron spikes brought about by the excitatory collaterals from the oscillating regions can decelerate the oscillation frequency in the target network. Reciprocal coupling between oscillators in the two hemispheres (Figure 6b) can alleviate the phase-shift problem and result in 0 phase-lag synchrony, provided that the conduction delays are short enough (<4–8 ms) and that synchrony is assessed over multiple cycles (Traub et al. 2002).

Long-range interneurons may be another candidate substrate for establishing gamma synchrony (Figure 7) (Buzsáki et al. 2004). These interneurons distribute their axon terminals over multiple regions and layers of the cortex and even across the hemispheres (Sik et al. 1994, Gulyás et al. 2003, Tomioka et al. 2005, Jinno et al. 2006). Importantly, distally projecting axons of long-range interneurons have several-fold thicker axons and larger diameter myelin sheaths than do pyramidal cells (Figure 7c,d), allowing for considerably faster axon conduction velocity (Jinno et al. 2006). In the I-I gamma model, replacing just 10–20% of the basket synapses with synapses of fast-conducting long-range interneurons could achieve global-phase synchrony (Figure 7a,b) (Buzsáki et al. 2004). An obvious advantage of the hybrid basket, long-range interneuron network is that synchrony among local and distributed cell assemblies can be tuned selectively by differentially targeting the two interneuron types.

Brain-Wide Synchronization of Gamma Oscillations by Slower Rhythms

Slower temporal coordination among gamma oscillators may be achieved by modulating the gamma power by the phase of slower rhythms (Figure 6). Compared with faster oscillators,
4,000 neurons
Local neurons only (p = 0)

Rate (Hz)
Time (ms)

0 50 100 150 200

4,000 neurons
Local and long-range neurons (p = 0.2)

Rate (Hz)
Time (ms)

0 50 100 150 200

0.5 μm
slower oscillators involve more neurons in a larger volume (Von Stein & Sarnthein 2000) and are associated with larger membrane potential changes because in longer time windows spikes of many more upstream neurons can be integrated (Hasenstaub et al. 2005, Quilichini et al. 2010).

CFPA coupling between gamma and other rhythms within the same and different brain regions has been well documented, including modulation by theta (Figure 3c) (Buzsáki et al. 1983; Soltesz & Deschénes 1993; Bragin et al. 1995; Chrobak & Buzsáki 1998; Wang 2002; Mormann et al. 2005; Canolty et al. 2006; Demiralp et al. 2007; Tort et al. 2008, 2010; Colgin et al. 2009; Griesmayer et al. 2010), alpha (Palva et al. 2005, Cohen et al. 2009), spindle (Peyrache et al. 2011), delta (Lakatos et al. 2005), slow (Hasenstaub et al. 2005, Isomura et al. 2006), and ultralow (Leopold et al. 2003) oscillations (Buzsáki 2006, Jensen & Colgin 2007, Schroeder & Lakatos 2009, Canolty & Knight 2010). Because perisomatic basket cells contribute to both gamma and theta rhythms by firing theta-rhythm-paced bursts of spikes at gamma frequency, it has been hypothesized that fast-firing basket cells may play a key role in cross-frequency coupling (Buzsáki et al. 1983, Bragin et al. 1995). This is plausible because several other types of interneurons are often entrained by slower oscillations and they inhibit basket cells (Freund & Buzsáki 1996, Klausberger & Somogyi 2008). A prediction of this hypothesis is that temporal coordination by the basket cells also introduces a CFPP (i.e., phase-phase or n:m) coupling relationship between theta and gamma oscillations (Figure 5e). It may well be that CFPP mechanisms underlie CFPA coupling in most situations, but convincing demonstration of clear phase-phase coupling is hampered by the lack of adequate methods to quantify cross-frequency interactions and reliably track the true phase of nonharmonic oscillators (Tort et al. 2010, Belluscio et al. 2012).


Several studies have examined the relationship between cross-frequency coupling of gamma oscillations and cognitive processes. The magnitude of theta-gamma coupling in the hippocampal region varied with working memory load in patients implanted with depth electrodes (Axmacher et al. 2010). The strength of theta-gamma coupling in the hippocampus and striatum of the rat was affected by task demands (Tort et al. 2008, 2009). Similarly, the magnitude of CFPA coupling between a 4-Hz oscillation and gamma power in the prefrontal cortex increased in the working memory phase of a choice task (Fujisawa & Buzsáki 2011). In an auditory task, gamma power in the frontal and temporal sites was phase-locked mainly to theta oscillations, whereas over occipital areas phase modulation was strongest by the alpha rhythm in a visual task (Voytek et al. 2010). Increased CFPP coupling between alpha and beta/gamma oscillations correlates with the difficulty of arithmetic mental tasks in the human magnetoencephalogram (Palva et al. 2005), whereas in another study working memory was correlated with theta-gamma synchrony (Griesmayer et al. 2010). Cross-frequency coupling between slow rhythms and gamma oscillations can support a “reader-initiated” mechanism for information exchange (Sirota et al. 2008). For example, the hippocampal theta rhythm can entrain local gamma oscillations in multiple cortical areas. During its duty cycle, the theta output can phase align gamma oscillations that emerge in numerous activated neocortical local circuits (Figure 5f). In turn, the cell assemblies
Multiple gamma sub-bands. Wavelet power between 30 and 150 Hz as a function of waveform-based theta cycle phases. Note the different theta-phase preference of mid-frequency (M) (gamma_M, 50–90 Hz, near theta peak) and slow (S) (gamma_S, 30–50 Hz on the descending phase of theta) gamma oscillations. Note also the dominance of fast (F) (gamma_F, or epsilon band, 90–150 Hz) at the trough of theta. After Belluscio et al. (2012).

WHEN GAMMA POWER IS NOT A RHYTHM

A caveat in many studies is the lack of a disciplined and quantified analysis of gamma oscillations. To identify true gamma oscillations, appropriate statistics should be applied to demonstrate periodicity (Muresan et al. 2008, Burns et al. 2011, Ray & Maunsell 2011), and additional experiments are needed to distinguish between a power increase resulting from genuine oscillations and an increase resulting from greater spiking activity (Jarvis & Mitra 2001, Crone et al. 2006, Montgomery et al. 2008, Whittingstall & Logothetis 2009, Quilichini et al. 2010, Belluscio et al. 2012, Ray & Maunsell 2011). This is especially important for higher frequencies, such as the epsilon band, but spike-afterdepolarization and -hyperpolarization components can also contribute to the gamma band power. Although spike contamination to oscillatory power can be a nuisance, by using proper analytical methods, spike power can be exploited as a proxy for the assessment of neuronal outputs even in recordings of subdural local field potentials. Studying the temporal features of such high-frequency events may provide clues about oscillatory events that modulate them, even in situations when invasive unit recordings are not an option.

MULTIPLE GAMMA RHYTHMS

Cross-frequency coupling can assist with the separation of gamma sub-bands (Tort et al. 2008, Colgin et al. 2009). In the hippocampal CA1 region, wavelet analysis identified three distinct gamma bands: (a) slow gamma (gamma_S, 30–50 Hz) on the descending phase, (b) mid-frequency gamma (gamma_M, 50–90 Hz) near the peak, and (c) fast gamma (gamma_F, or epsilon band, 90–140 Hz) near the trough of the theta cycle (Figure 8) (Tort et al. 2010, Belluscio et al. 2012). Support for the different origins of gamma sub-bands is provided by their differential distribution in the different depths of the CA1 pyramidal layer and in different segments of the subiculum (Belluscio et al. 2012, Jackson et al. 2011). It is likely that the slow and mid-gamma band distinction applies to other brain regions as well (Kay 2003).

Previous works have distinguished only low and high gamma sub-bands (Csicsvari et al. 1999, Ray & Maunsell 2011) with the high sub-band defined as 60–140 Hz (Canolty et al. 2006, Colgin et al. 2009). Because power in the mid-gamma (50–90 Hz) and epsilon (90–150 Hz) bands is associated with different phases of theta oscillation (Figure 8) and is likely generated by different mechanisms (Belluscio et al. 2012), lumping these bands together is not justified on physiological grounds. Future studies, therefore, should distinguish sub-bands of gamma oscillations and carefully separate true and spurious gamma rhythms (see When Gamma Power is Not a Rhythm, sidebar above).

To conclude, although the word “rhythm” readily conjures up the picture of a clock, gamma rhythms occur in relatively short bursts and are quite variable in frequency, typically
associated with stochastic firing of single neurons. The LFP gamma reflects largely the balancing act of excitation and inhibition, i.e., the active mode of a local circuit. Future studies on gamma oscillations will continue to inform us about the complex dynamics of brain circuits.

**SUMMARY POINTS**

1. Transient cell assemblies may be organized into gamma-wave cycles.
2. Perisomatic inhibition by PV basket cells is essential for gamma oscillations.
3. Gamma oscillations are short-lived and emerge from the coordinated interactions of excitation and inhibition. Thus, LFP gamma can be used to identify active operations of local circuits.
4. Network gamma oscillations may coexist with highly irregular firing of pyramidal neurons.
5. Different sub-bands of gamma oscillations can coexist or occur in isolation.
6. Long-range interneurons may be critical for gamma-phase synchrony in different brain regions.
7. Cross-frequency coupling is an effective mechanism for functionally linking active cortical circuits.
8. Genuine gamma oscillations should be distinguished from mere increases of gamma-band power and/or increased spiking activity.

**DISCLOSURE STATEMENT**

The authors are not aware of any affiliations, memberships, funding, or financial holdings that might be perceived as affecting the objectivity of this review.

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