Interleaved single and bursting spiking resonance in neurons

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Abstract

Under *in vivo* conditions, CA1 pyramidal cells from the hippocampus display transitions from single spikes to bursts. It is believed that subthreshold hyperpolarization and depolarization, also known as down and up-states, play a pivotal role in these transitions. Nevertheless, a central impediment to correlating suprathreshold (spiking) and subthreshold activity has been the technical difficulties of this type of recordings, even with widely used calcium imaging or multielectrode recordings. Recent work using voltage imaging with genetically encoded voltage indicators has been able to correlate spiking patterns with subthreshold activity in a variety of CA1 neurons, and recent computational models have been able to capture these transitions. In this work, we used a computational model of a CA1 pyramidal cell to investigate the role of intrinsic conductances and oscillatory patterns in generating down and up-states and their modulation in the transition from single spiking to bursting. Specifically, the emergence of distinct spiking resonances between these two spiking modes that share the same voltage traces in the presence of theta or gamma oscillatory inputs, a phenomenon we call interleaved single and bursting spiking resonance. We noticed that these resonances do not necessarily overlap in frequency or amplitude, underscoring their relevance for providing flexibility to neural processing. We studied the conductance values of three current types that are thought to be critical for the bursting behavior: persistent sodium current (I_{NaP}) and its conductance G_{NaP}, delayed rectifier potassium (I_{KDR}) and its conductance G_{KDR}, and hyperpolarization-activated current (I_h) and its conductance G_h. We conclude that the intricate interplay of ionic currents significantly influences the neuronal firing patterns, transitioning from single to burst firing during sustained depolarization. Specifically, the intermediate levels of G_{NaP} and G_{KDR} facilitate spiking resonance at gamma frequency inputs. The resonance characteristics vary between single and burst firing modes, each displaying distinct amplitudes and resonant frequencies. Furthermore, low G_{NaP} and high G_{KDR} values lock bursting to theta frequencies, while high G_{NaP} and low G_{KDR} values lock single spiking to gamma frequencies. Lastly, the duration of quiet intervals plays a crucial role in determining the likelihood of transitioning to either bursting or single spiking modes. We confirmed that the same features were present in previously recorded in vivo voltage-imaging data. Understanding these dynamics provides valuable insights into the fundamental mechanisms underlying neuronal excitability under *in vivo* conditions.

Author summary

Since discovering that neurons in the hippocampus can encode spatial position through phase precession, many experiments have explored how specific theta and gamma oscillations influence location specificity in the brain. However, the individual neuronal properties and dynamics behind these behaviors are still being uncovered. Previously, we found that stereotypical bursting and single-spike firing in pyramidal neurons are linked to these oscillations and further associated with an animal entering or leaving a place field. Advances in voltage-imaging techniques have enabled us to assess these properties more precisely. Our study shows that different frequencies can independently trigger these stereotypical spikes, demonstrating a complex pattern where the same cell can be double-coded: a phenomenon we called interleaved resonance. Additionally, we found that this coding can be modulated by persistent sodium and delayed-rectifier potassium currents. Moreover, these neurons are more likely to burst following long periods of silence. These findings provide new insights into the mechanisms underlying neural coding in the hippocampus and how it relates to behavior.

Introduction

Neurons from the CA1 hippocampus display a diversity of spiking dynamics. In the hippocampus, from the point of view of individual cells, *in vivo* voltage imaging [1, 2, 3], patch clamp recordings [4, 5, 6, 7], and multielectrode recordings [8, 9] reveals that CA1 pyramidal cells exhibit stereotypical single and burst spiking. While investigating their emergence and transitions it was noted that these spike modes appear with different probabilities of occurrence where bursts are more likely to occur after long periods of activity silence [9]. From the point of view of populations, gamma (30-100 Hz) and theta (3-12 Hz) oscillations coexist and have been well characterized in CA1 region of rodent hippocampus [10, 11]. Gamma is modulated by theta during working memory retention, and it is important to notice the cross-frequency coupling of these oscillations for working memory performance [12, 13]. Recent studies have tried to connect those transitions using *in vivo* data, as shown by shifts from single to bursting spikes in the same neurons that are preferentially observed during theta and gamma oscillations, respectively [1]. Behaviorally, these spiking modes have been proposed to encode an animal's position, where burst signals when animals enter a place. In contrast, single spikes signal when the animal leaves a place [1].

Admittedly, understanding spiking patterns transitions is synonymous to understanding ionic current's dynamics. For fixed synaptic input, the composition of ionic currents can produce distinguishable spiking outputs that vary from single to burst of spikes [14, 15, 16]. Combinations of ionic currents can determine these spike modes as in the case of calcium and burst [17, 18, 19], or in combination with the spatial location on the neuron as in the case of persistent sodium current (I_{NaP}) located at or near the soma [14, 20], or only the geometry [21], or combinations of both [22].

Oscillations can cause transitions from low to high firing rate on a neuron. Cells may have amplified response for certain oscillatory frequencies compared to other frequencies. This is known as suprathreshold (spiking) resonance, and it has been observed widely across the brain [23, 24, 25, 26, 27], including CA1 pyramidal cells [28] and the mechanisms underlying this resonance have been intensively investigated using computational models [29, 30]. Similarly, subthreshold resonance can be seen as a transition voltage change of a neuron upon a particular

(non-zero) input frequency [31, 32]. Several ionic currents contribute to subthreshold resonance
and can be discussed in terms of their chord and derivative conductances [33]. For example,
persistent sodium current (I_{NaP}) is an amplifying ionic current and delayed rectifier potassium
(I_{KDR}) is a resonant current [31, 34, 35]. Less is known about the role of these ionic currents on
suprathreshold (spiking) resonance.

It is only natural to assume that CA1 pyramidal neurons embedded in a network can take advantage
of the spiking resonance to preferentially transmit information encoded in the spiking mode (single
or burst spiking) when a particular oscillatory synaptic input (e.g. theta or gamma oscillations) is
received [28, 23, 36, 37]. In addition to that, it is important to understand how transitions at these
different scales connect and modulate them. In this paper, we investigate suprathreshold resonance
of single and burst spiking in a CA1 pyramidal neuron model. We found that spiking resonance
can easily emerge when neurons are stimulated with theta and gamma synaptic inputs and ionic
currents are tuned. Variability of sodium and potassium currents differentially promoted the
presence of resonance according to the firing pattern (single vs. burst), a phenomenon we call
interleaved single and bursting spiking resonance. Our results provide insights that might help to
understand the mechanisms underlying single and burst spiking computation and its role in
connection to *in vivo* processing that is relevant for spike coding in CA1 pyramidal cells for spatial
navigation and memory.

Our paper is organized as follows: in the following section, we present studies of how ionic
currents promote transitions from single to burst firing under constant sustained input and later
under oscillatory inputs. For the oscillatory inputs, we show how resonance emerges and depends
on ionic currents. Then we move to a more systematic study where we characterize interleaved
single and bursting resonance in the presence of dendritic I_h current. For the same model, we
characterize phase-locking to gamma and theta and how these main ionic currents shape this
relationship. We conclude by systematically characterizing the probability of bursting spikes in
relation to the periods of silence in the voltage traces. Methods are included in the last section of
the paper.

Results

Ionic currents promote a transition from single to burst firing during sustained depolarization

CA1 pyramidal cells are known to exhibit transitions from single (SS) to bursting spikes (BS) *in*
vivo [1, 2, 3, 38]. This transition occurs in the presence of subthreshold oscillations and is
modulated by oscillatory inputs, mostly theta and gamma oscillations. Subthreshold activity is
characterized by transitions from hyperpolarized states to depolarized states (also known as down
and up-states, respectively). Firing mostly occurs on top of the depolarization phases. Input
patterns play a relevant role in the switching, but it is unclear how they can be promoted by the
neuron's intrinsic properties, e.g. ionic currents. We first decided to investigate the firing patterns
occurring during depolarization states and how SS and BS emerge under different ionic
conductances.

We first investigated the role of conductances that are mostly expressed in the soma, such as I_{NaP}
and I_{KDR}. We used a single compartment computational model of CA1 pyramidal cell (see
Methods) and we varied G_{NaP} (persistent sodium conductance) and G_{KDR} (delayed rectifier
potassium conductance) for the model receiving a pulse of current that maintains the voltage in a
constant depolarized state (Fig 1A). We observed that increasing G_{NaP} makes the firing transits
from SS to BS and finally to plateau potential, also known as depolarization block [39] (see voltage
traces Fig 1B). The phase plots (dV/dt) in the same figure reveals how the voltage series transits
between the different modes. In contrast, the effect of increasing G_{KDR} is a transition from BS to
SS (Fig 1C). Single spiking can be explained as a result of having long pauses between spikes due
to long afterhyperpolarization, which is facilitated by strong action potential repolarization due to
large G_{KDR} or by preventing fast depolarization due to small G_{NaP}.
144
145 **Fig 1. Ionic currents promote transition from SS to BS during sustained depolarization.**
146 Default values for fixed conductance while varying the other are $G_{NaP} = 0.3$ and $G_{KDR} = 6$. **A** Sketch
147 of CA1 model receiving constant current pulse. **B** Example voltage traces and phase plots (dV/dt)
148 when varying G_{NaP} or **C** G_{KDR}. **D** Firing rate and coefficient of variation (CV) heatmaps spanned
149 over G_{NaP} and **E** G_{KDR} vs. Input current.
150
The firing rate increases with the increase of stimulation amplitude, increase of G_{NaP} and decrease
of G_{KDR} (Fig 1D-E). The coefficient of variation (CV) above 2 (reddish to yellowish) determines
the area in which the combination of conductance promotes BS behavior (Fig 1D-E). As it is shown
in the heatmaps, BS is independent of the input amplitude and mostly occurs in areas around G_{NaP}
$= 0.35$ or $G_{KDR} = 6$. These values correspond to the edges where the transition from spiking to
depolarization block occurs (see dark regions in heatmaps of firing rate). Interestingly, BS occurs
in a very narrow range of conductances, meanwhile single spiking is likely to occur in a wide range
of values. This set of observations shows the opposite effects of the G_{NaP} and G_{KDR} ionic currents
in modulating a switch between SS to BS.
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Intermediate G_{NaP} and G_{KDR} generate spiking resonance
preferentially at gamma frequency input
We showed previously that G_{NaP} and G_{KDR} have opposite effects on the generated spiking patterns
during sustained depolarization. Nevertheless, CA1 pyramidal neurons receive oscillatory inputs
in the range from 3 to 100 Hz, where oscillatory gamma is between 30-100 Hz and theta 3-12 Hz
frequencies. This induces transitions from hyperpolarized to depolarized states where shorter
intervals occur at high input frequencies such as gamma frequencies and longer intervals at theta
frequencies. Under this condition, firing during depolarization is not sustained, but rather is
alternated by periods of silence. In our paper, we investigate the role of the silent states generated
by the interplay between input patterns and ionic currents to create a suprathreshold resonance and
its relation to SS and BS.
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For stimulation of the cell in this section, we choose varying oscillatory stimulation frequencies
covering the range from theta to gamma frequencies (Fig 2A). We investigate the firing rate
responses and their preference for a certain input frequency (i.e. suprathreshold resonance). We
observe low-pass filtering of the firing rate for low G_{NaP} or high G_{KDR}, suprathreshold resonance
for intermediate G_{NaP} or G_{KDR}, and a nearly absent frequency preference for high values of G_{NaP}
or low values of G_{KDR} (Fig 2C, E). Interestingly, regardless of the opposite effects of G_{NaP} and

G_{KDR}, we observed that theta input has a stronger effect on the firing rate response than gamma
when the neuron acts as a low-pass filter. In contrast, gamma is stronger than theta for resonance
(Fig 2C, E). These results unveiled an interaction between the input frequency and the filtering
properties that determine the firing output.

Fig 2. Intermediate G_{NaP} and G_{KDR} generates firing rate resonance. Default values for fixed
conductance while varying the other are $G_{NaP} = 0.3$ and $G_{KDR} = 6$. **A** Sketch of simplified CA1
model receiving oscillatory current stimulation. **B** Examples of voltage traces for varying values
of G_{NaP}. **C** Firing rate while varying stimulation frequency of an oscillatory input. Shaded red area:
theta frequencies. Shaded blue area: gamma frequencies. **D** Examples of voltage traces for varying
values of G_{KDR}. **E** Firing rate while varying stimulation frequency of an oscillatory input. **F** Firing
rate and CV heatmap spanned over G_{NaP} and oscillatory input frequency. **G** Firing rate and CV
heatmap spanned over G_{KDR} and oscillatory input frequency.

The CVs above 2 (reddish to yellowish) obtained while changing the oscillatory input frequency
and the conductance indicate the area where BS occurs. The heatmaps show similar patterns for
firing frequency and CV for both G_{NaP} and G_{KDR}, i.e. we observe high firing rates simultaneously
with high values of CV. The high firing rates obtained can be justified by the BS and low firing
rates are by the SS (Fig 2F, G).

We observed BS when G_{NaP} levels were intermediate or high (see the yellowish diagonal in Fig.
2F) and when G_{KDR} was at intermediate levels (Fig. 2G). Additionally, BS was noted at theta
frequencies with low G_{NaP} or high G_{KDR}. Similarly, we observed SS at low G_{KDR} and at gamma
frequencies with low or intermediate G_{NaP} or high G_{KDR}. Interestingly, BS in gamma frequencies
occurs simultaneously with resonance at intermediate values of G_{NaP} or G_{KDR}. These results
indicate that when suprathreshold resonance is present, a linear increase in input frequency and
G_{NaP} shifts the BS presence from theta to gamma frequencies. Similarly, a decrease in G_{KDR}, along
with increased input frequencies, also shifts the BS presence from theta to gamma frequencies (as
shown by the yellowish regions in the heatmaps). A similar pattern is observed for SS, with the
dark regions in the G_{NaP} heatmap and the dark middle-right region in the G_{KDR} heatmap illustrating
this shift.

In summary, for low G_{NaP} and high G_{KDR}, we found high firing rate response to low input
frequencies (i.e. theta) due to weak filtering, and low firing rate response to high input frequencies
(i.e. gamma) due to strong filtering. This suggests that the tuning of ionic currents in interaction
with input frequencies are critical for determining the occurrence of resonance.

Single and burst firing display suprathreshold resonance with different amplitudes and different resonant frequencies

218 It is well known that CA1 pyramidal cells have a high expression of the hyperpolarization activated
219 I_h current [40]. This current is mostly expressed in the dendrites and plays a relevant role in
220 modulating synaptic inputs [40, 41, 42, 43] and resonance [33]. It prevents excessive
221 hyperpolarization and causes rebound spiking after hyperpolarizations [44, 45]. Thus, in this
222 section, we incorporated a dendritic compartment containing I_h current (Fig 3A).

Figure 3. Single and burst firing display suprathreshold resonance with different amplitudes and resonant frequencies. Default values for fixed conductance while varying the other are G_{NaP} = 0.93, G_{KDR} = 5.86, and G_h = 8.46. **A** Sketch of CA1 model with additional dendritic compartment receiving oscillatory current stimulation. **B** Exemplary voltage trace of the simulation with SS and BS indicated. **C1-2** Firing rate vs oscillatory input frequency for SS **C1** and BS **C2** when G_{NaP} values are 0, 0.5 and 1.5. **D1-2** Same as in **C**, but G_{KDR} values are 0, 5 and 10. **E1-2** Same as in **C**, but G_h values are 0, 2 and 5. **F** Representative sketch demonstrating how Δf and ΔA are calculated for the interleaved SS and BS resonances. **G1-2** Δf and ΔA for varying values of G_{NaP}. **H1-2** Same as in **G**, but varying G_{KDR}. **I1-2** Same as in **G**, but varying G_h. Colors mark the magnitude of ionic currents (blue = low, orange = medium, green = high).

We stimulated the cell model with varying oscillatory stimulation frequencies covering the range from theta to gamma frequencies (Fig 3A). Under this condition, the neuron is capable of switching its firing from SS to BS (Fig 3B). We separated our quantitative measures into SS and BS (see Methods). Overall, we observed resonance for all the explored conductances and for both spiking types (SS and BS), except for BS when G_{NaP} is large or when G_{KDR} is low (Fig 3 C2, D2). Finally, SS and BS have resonance for all G_h values (Fig 3 E1, E2).

We further compared the resonance frequencies and their respective firing rate amplitudes between SS and BS (Fig 3F). For low G_{NaP}, SS resonance frequency is larger than BS resonance frequency, but it is the opposite when G_{NaP} is large (Fig 3G1). For low G_{KDR}, SS resonance frequency is smaller than BS resonance frequency, but it is the opposite when G_{KDR} is large (Fig 3H1). Moreover, SS resonance frequency is larger than BS resonance frequency for all G_h values (Fig 3I1). Finally, as expected, firing frequency at the resonant peak was always higher for BS rather than SS (Fig 3G2, H2, I2).

The set of results in this section demonstrates an interesting coding behind SS and BS in our CA1 model. First, it shows that neither the resonance frequencies nor resonance amplitudes between SS and BS necessarily overlap. More interestingly, this phenomenon happens at the same voltage traces (interleaved) and can be modulated by ionic currents. In the next section, we show how these observations are connected to oscillations.

G_{NaP}, G_{KDR} and G_h determine phase locking of SS and BS to theta and gamma input respectively

It has been demonstrated that this neuron model can reproduce coexisting SS and BS phase-locked to gamma and theta oscillations, respectively. The model is derived from [1, 14] and has its conductances fitted to experimental voltage-imaging. In this paper, we modified the model for a similar fitting (see parameters in Methods). Based on the observations above for changing ionic conductances and the theta and gamma relationship to BS and SS, we question the ability of the ionic channels to modulate the phase-locking of the neuron model to these frequencies.

Since most of CA1 neuron inputs come with a mixture of theta and gamma frequencies, it is only natural to study the case where both are present. Thus, we injected both theta and gamma simultaneously into the neuron model (Fig 4A). As observed, the ionic currents embedded in the neuron can generate different phase-locking (Fig 4B). Low values of G_{NaP} are more prompt to BS at theta frequencies. Increasing the values of G_{NaP} removes this phase locking with the generation

of more SS (Fig 4C). In contrast, G_{KDR} had the opposite behavior (Fig 4D). We see only SS for
low values of G_{KDR}, BS at gamma for intermediate values of G_{KDR} and finally BS at theta is
generated for high G_{KDR}. Interestingly, only low and high values of G_h promoted a difference at
the gamma frequencies (Fig 4E).

Fig. 4. G_{NaP}, G_{KDR} and G_h determine phase locking statistic of SS and BS to theta and gamma
input. Default values for fixed conductance while varying the other are $G_{NaP} = 0.93$, $G_{KDR} = 5.86$,
and $G_h = 8.46$. **A** Sketch of CA1 model with somatic and dendritic compartments receiving theta
and gamma oscillatory current stimulation. **B** Examples of voltage traces with SS behavior
transitioning to a BS behavior (compare top to bottom). **C** Phase locking statistic to theta and
gamma input separated into SS and BS with varying G_{NaP}. **D** Phase locking statistic with varying
G_{KDR}. **E** Phase locking statistic with varying G_h. Each bar is the result of 10 different simulations.
t-test with 0.05 level of significance (asterisk).

Taken together, these results show that G_{NaP}, G_{KDR}, and G_h modulate the coexistence of BS and
SS phase locked to theta and gamma differently: low values of G_{NaP} and high values of G_{KDR} lock
BS to theta frequencies. G_h has little effect on theta frequencies and affects gamma frequencies
instead. Furthermore, high values of G_{NaP} and low values of G_{KDR} lock SS to gamma frequencies.

Quiet intervals determine the probability of bursting or single spiking
The *in vivo* activity of CA1 neurons is characterized by subthreshold oscillations that follows the
theta and gamma synaptic inputs. During these oscillations, spiking activity is observed at the
depolarization phase (up-states) and between them it is possible to observe silence intervals at
hyperpolarized phase (down-states). During the up-states, neurons can fire SS or BS. In this
section, we investigate whether the silence intervals modulate the probability of having SS or BS.

According to our observations in Fig 5, high probability of BS occurs for long intervals of silence
(above 100 ms) in spite of variation in the values of the conductances G_{NaP}, G_{KDR}, and G_h (Fig
5D1-F1). Interestingly, we observed that interevent intervals (ISI) are fractured and distributed in
more than one position in the diagram, and that different peaks show up mostly for large G_{NaP} or
low G_{KDR} (Fig 5A and B). We also observed that high probability of SS occurs for short intervals
of silence (below 60 ms) for low G_{NaP}, high G_{KDR} and regardless of the values of G_h (Fig 5D2-F2).
Moreover, we observed high probability of SS occurring for long intervals (above 100 ms) only
when G_{NaP} is high or when G_{KDR} is low (Fig 5D2-E2). In summary, we found that the probability
of BS is robust to variation of the ionic conductances and mostly depends on long silent periods
above 100 ms (i.e. when theta input is present). In contrast, SS is influenced by ionic conductances
and can occur at short or long intervals.

Figure 5. Quiet intervals determine burst and single spikes. Default values for fixed
conductance while varying the other are $G_{NaP} = 0.93$, $G_{KDR} = 5.86$, and $G_h = 8.46$. **A** Histogram of
interevent intervals (ISI) with varying G_{NaP}. **B** Histogram of interevent intervals (ISI) with varying
G_{KDR}. **C** Histogram of interevent intervals (ISI) with varying G_h. **D1-F1** BS and **D2-F2** SS
probabilities for varying G_{NaP}, G_{KDR}, and G_h, respectively. Colors mark the magnitude of ionic
currents (green = low, red = medium, blue = high).

To determine whether the spiking probability is the sum of independent effects of theta and gamma stimulation, or the result of the interaction between both inputs, we calculated the spiking probability when only theta or gamma inputs were used (Fig 6). As expected, when only gamma input is present, we observed that the probability of BS or SS is always larger at short intervals of silence (< 50 ms) (Fig 6A). In contrast, when only theta input is present, we observed that the probability of BS is larger for long intervals of silence (> 100 ms). For SS probability when only theta input is present, we observed the same tendency found above, where high probability of SS occurs for short intervals of silence below 60 ms for low G_{NaP}, high G_{KDR} and regardless the values of G_h (Fig 6B), and high probability of SS occurs for long intervals above 100 ms only when G_{NaP} is high or when G_{KDR} is low.

Figure 6. Burst probability follows the longest quiet intervals according to the slowest oscillatory input. Default values for fixed conductance while varying the other are G_{NaP} = 0.93, G_{KDR} = 5.86, and G_h = 8.46. **A** BS and SS probabilities with only gamma input. **B** BS and SS probabilities with only theta input. Colors mark the magnitude of ionic currents (green = low, red = medium, blue = high).

In summary, we observed that when the input is only at gamma frequencies, the spiking probability is largest for short silent intervals for both SS and BS. Opposite to that, when the input is only oscillating at theta, BS is more likely to occur at long intervals and SS at short intervals. Interestingly, when both theta and gamma are present, the probabilities behave as if only theta was present, i.e., the gamma component is mostly disregarded. These results suggest that BS probability follows the input frequency with the longest intervals of silence, i.e. it follows theta even in the presence of gamma, but it follows gamma in the absence of any other slowest input. Meanwhile, SS would preferentially occur at short intervals.

Voltage imaging burst probability follows the longest quiet intervals

Despite the difficulties in reconciling our computational results with voltage-imaging recordings, where frequency-dependent stimulation is typically absent, it is possible to compare the bursting probability following the longest quiet intervals with spontaneous recordings. Thus, we analyzed voltage imaging traces of 7 neurons (20 s long) obtained from a public repository [3] . Fluorescence images were collected from CA1 pyramidal cells and somatostatin expressing interneurons located in the dorsal CA1 area at depths ~50 to 100 μm (Fig 7A). LED epifluorescence illumination allowed the collection of the voltage indicator fluorescence by a high-speed sCMOS camera at a rate acquisition of 600 Hz. Photobleaching was subtracted by fitting a single exponential decay. We selected the traces from pyramidal cells based on their high signal-to-noise ratio and high frequency of events suggesting good quality of recordings (Fig 7B). These criteria allowed us to properly detect and classify events as SS and BS as done in the computational simulations (Fig 7C). We calculated the probability of finding SS or BS for a certain ISI (Fig 7D). We performed a statistical comparison between BS and SS probabilities for 3 areas marked for low ISI (50 to 150 ms), intermediate ISI (150 to 250 ms), and long ISI (400 to 500 ms).

Figure 7. Voltage imaging burst probability follows the longest quiet intervals. Data from [3]. **A** Sketch of the voltage-imaging experiment. **B** Sample traces that were used for this analysis (right, n = 7). **C** Sample trace showing spike detection algorithm prior to division into BS and SS.

361 **D** Interevent interval (ISI) dependent probabilities of BS and SS (n = 7), and 3 areas marked for
362 low ISI (50 to 150 ms), intermediate ISI (150 to 250 ms), and long ISI (400 to 500 ms). **E** From
363 the marked areas, bar plots with averaged values in each ISI trace (n = 7). t-test with 0.05 level of
364 significance (asterisk).
365
The results showed that SS is more likely at low ISIs, while BS is more likely at long ISIs.
Observations of these probabilities across all traces (Fig. 7D) and statistical significance from the
traces and areas (Fig. 7E) confirm this. The voltage-imaging recordings agree with our simulation
observations, where burst probability follows the longest quiet intervals (Figs 5 and 6). An
independent study observed this phenomenon in behaving rats, focusing on CA1 pyramidal cell
burst activity from electrodes that were surgically implanted within the brain [9].
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Discussion

In this paper, we investigated the role of ionic conductances in causing suprathreshold resonance
for SS and BS in a CA1 pyramidal neuron model. When studying neuronal resonance, it is
important to assess input-output properties and combinations of intrinsic properties [29, 46, 47,
48, 31, 49]. Our focus lay on specific input and output parameters, with emphasis on oscillatory
theta (3-12 Hz) and gamma (30-100 Hz) frequencies for inputs, and SS and BS for outputs. In
particular, we systematically assessed these quantitates separately and identified interleaved SS
and BS resonance that does not match in frequency nor amplitude, but it is modulated by the ionic
currents. To the best of our knowledge, this is the first paper showing the ability of a single
neuronal voltage trace to carry a double-coded spiking resonance. We have also observed that the
burst probability is associated with longer periods of silence and confirmed this observation
through the analysis of in vivo voltage-imaging data from CA1 pyramidal neurons.
385
More specifically, in terms of ionic currents, we found opposite effects of persistent sodium and
delayed rectifier potassium currents. Regarding the ionic conductances, we investigate how G_{NaP}
and G_{KDR} ionic currents affect the emergence of SS to BS. Accordingly, whereas BS is observed
for large G_{NaP} or low G_{KDR}, SS is observed for low G_{NaP} or large G_{KDR}. We found that under
conditions that allow spiking resonance, BS occurs for theta while SS for gamma. Furthermore, at
low input frequencies (theta), the firing rate response is higher when G_{NaP} is low and G_{KDR} is high
due to weak filtering. Conversely, at high input frequencies (gamma), the firing rate response is
lower for the same conditions (low G_{NaP} and high G_{KDR}) due to strong filtering. This different
strength of filtering is a consequence of the modulation of intrinsic conductances (i.e. G_{NaP} and
G_{KDR}). Both conductances cooperate with each other to set the neuronal filtering at different levels
for the respective input frequency (i.e. theta and gamma). One consequence of these results is that
changes in expression levels in these conductances could allow shifting of the strength of the
output for a certain input frequency. As a result, the neuron can facilitate the transfer of low or
high frequency inputs according to the expression of its ionic conductances [50, 51, 52].
400
We found that the interplay between ionic currents and oscillatory input frequencies determines
the output strength of the firing rate. Specifically, we found that decreasing G_{NaP} increases the
firing at theta frequencies and decreases the firing at gamma. Again, G_{KDR} changes had opposite
results. This suggests that CA1 neurons might be able to tune their output preferences by
downregulating or upregulating the expression of their intrinsic conductances [53]. It also suggests

that pharmacological interactions targeting these ionic channels have the potential to affect
oscillations and BS behavior [54, 55].
408
We previously found that BS and theta frequencies are preferentially linked to the animal entering
a place field, while SS and gamma frequencies are linked to the animal leaving a place field [1].
Thus, BS and SS can be considered coding mechanisms for this type of behavior. Our simulations
demonstrate that different oscillatory frequencies can enhance (resonate with) BS and SS
separately in a complex frequency-dependent manner within this double-coded signal. This
indicates a higher-order interaction where the neuron does not need to code either BS or SS
exclusively but can encode both jointly. It would be interesting to find ways in which one
resonance is kept fixed while the other is varied. This would prove whether they are dependent or
independent.
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An interesting finding of our work was that ionic currents embedded in the neuron can generate
different phase-locking. These results showed that G_{NaP}, G_{KDR}, and G_h modulate the coexistence
of BS and SS phase locked to theta and gamma differently: while low values of G_{NaP} generate BS
at theta, it requires high values of G_{KDR} to generate BS at theta. G_h has little effect on theta
frequencies and affects gamma frequencies instead. I_h current modulation of up/down states is still
under investigation [56], but clearly a target of a diverse range of neuromodulators [57, 58].
425
Unexpectedly, we found that the likelihood of BS remains consistent despite changes in ionic
conductances and primarily relies on extended silent periods [9], meanwhile SS is influenced by
ionic conductances changes and can occur at any intervals. This BS behavior can be explained by
the degeneracy of ionic conductances. In these circumstances, multiple combinations of different
conductance values, even with significant heterogeneity, can produce the same BS behavior in the
neuron model [53, 59]. Another possible interpretation is that the various conductances within the
neuron model (such as transient sodium, muscarinic-sensitive potassium, A-type potassium, high-
threshold calcium, fast calcium-activated potassium, and slow calcium-activated potassium
currents) might respond differently to prolonged hyperpolarization, primarily inducing a state
conducive to bursting rather than single spiking. Conversely, shorter hyperpolarizations could
result in bistability, where both bursting and single spiking scenarios are plausible [60, 61].
437
The probability of spiking is influenced by the states of the conductances (i.e., the levels of
activation/inactivation) before the onset of depolarization [62]. For the conductances examined in
this paper, it is well known that during extended hyperpolarization (such as under the stimulation
of theta frequency oscillations), G_{KDR} and G_{NaP} remain inactive. However, during the initial phase
of depolarization, both G_{KDR} and G_{NaP} are rapidly activated [34], leading to the onset of bursting.
Similarly, it has been observed that the interplay between potassium and sodium conductances
triggers bursting after extended periods of silence [9, 60]. These findings align with previous
research demonstrating that these currents (e.g., G_{KDR} and G_{NaP}) are crucial for generating bursting
[14, 63]. Finally, it has been observed in cortical neurons that bursts can trigger switches from
intermittent bursting spiking to persistent single spiking activity and vice-versa [64]. These
findings bring new insights into the neural coding mechanisms in the CA1 hippocampus.
449

450 # Methods

Computational model

We use a Hodgkin-Huxley model of a CA1 pyramidal neuron described by conductance-based ionic currents [14, 1, 65, 66, 67, 68, 69]. The following differential equation describes how the voltage evolves in time:

$$C\frac{dV}{dt} = -g_L(V - E_L) - \sum I_j,$$

with the currents I_j

$$I_j = g_j m^M h^N (V - E_j).$$

The currents included are I_{NaT} the transient sodium, I_{NaP} the persistent sodium (M = 3, N = 0), I_{Kdr} the delayed rectifier K$^+$ (M = 0, N = 4), I_{KM} the muscarinic- sensitive potassium (M = 1, N = 0), I_{KA} the A-type K$^+$ (M = 3, N = 0), I_{HCa} the high-threshold Ca^{2+} (M = 3, N = 0), I_{fKCa} the fast Ca^{2+}-activated potassium current (M = 1, N = 0), and I_{sKCa} the slow Ca^{2+}-activated potassium current (M = 1, N = 0). The reversal potentials are $E_L = -70$ mV, $E_K = -90$ mV, $E_{Na} = 55$ mV, and $E_{Ca} = 120$ mV. The remaining parameters are $C = 1$ μF/cm^2, $g_L = 0.04$ mS/cm^2, $g_{NaT} = 35$ mS/cm^2, g_{NaP} =0.9 mS/cm^2, $g_{Kdr} = 5.8$ mS/cm^2, $g_{KA} = 2.8$ mS/cm^2, $g_{KM} = 2.4$ mS/cm^2, $g_{HCa} = 0.17$ mS/cm^2, $g_{sCa} = 2.8$ mS/cm^2. Equations for m and n follow

$$\frac{dx}{dt} = \frac{x_\infty(V) - x}{\tau_x(V)},$$

where $x_\infty(V) = \left\{1 + \exp\left[-\frac{(V - \theta_x)}{\sigma_x}\right]\right\}^{-1}$.

The remaining parameters for these equations are $\theta_m = -30$ mV, $\sigma_m = 9.5$ mV, $\theta_h = -45$ mV, $\sigma_h = -7$ mV, $\theta_n = -35$ mV, $\sigma_n = 10$ mV, $\theta_{nt} = -27$ mV, $\sigma_{nt} = -15$ mV, $\theta_a = -50$ mV, $\sigma_a = 20$ mV, $\theta_b = -80$ mV, $\sigma_b = -6$ mV, $\theta_z = -39$ mV, $\sigma_z = 5$ mV, as in [14].

For the compartmental modeling, we add gap junction currents between i and $i+1$ following $I_{gap} = g_{gap}(V_i - V_{i+1})$, with $g_{gap} = 0.2$ mS/cm^2.

For external currents, we use sinusoidal currents with specific frequencies or combinations thereof, as illustrated in the respective figures. Gamma frequencies are defined as ranging from 30 to 100 Hz, and theta frequencies range from 3 to 12 Hz. In our simulations, we use either 40 Hz for gamma or 10 Hz for theta.

Experimental recordings

To compare the model results with experimentally recorded in vivo voltage-imaging using genetically encoded voltage indicator (acemNeon2) in awake moving mice, we used freely available data from [3] in https://tinyurl.com/kannanEtAlScience2022 (last accessed in June 2024). The recordings are 20 seconds long and the data for this paper contain 7 CA1 pyramidal cells from depths ranging from approximately 50 to 100 μm. Image segmentation was performed using custom-written Matlab scripts as used in previous publications [38, 70, 71, 72]. CA1 pyramidal cells were identified by their expression of the negative polarity relationship of

fluorescence emission to the membrane voltage as expected from the voltage indicator (acemNeon2), whereas somatostatin interneurons were identified by the positive polarity that expressed a different voltage indicator (pACe). The voltage indicator was injected in the entorhinal cortex (EC) and expressed retrogradely in CA1, thus, all CA1 pyramidal neurons reported here are projecting to EC. Data acquisition was performed using LED epifluorescence illumination, and the voltage indicator fluorescence was recorded at a rate of 600 Hz with a high-speed sCMOS camera. To correct for photobleaching, we applied a single exponential decay fit and subtracted it from the data. We specifically selected traces from pyramidal cells that exhibited a high signal-to-noise ratio and a high frequency of events, which indicated good recording quality.

Firing rate and coefficient of variation

We compute the firing rate as the sum of spikes divided by the time window considered. Unless stated otherwise, we adopted 4 s in each simulation. The inter-spike intervals (ISI) are taken as the intervals between spikes in the trace. We also divide intervals between single-spikes and bursts accordingly. ISI probabilities are calculated using the same criteria for separating ISIs from single spikes and bursts. The intervals are measured from the last spike to the next single spike or burst spike, regardless of whether the last spike was part of a single spike or a burst.

Burst detection

Following [4] that showed through intracellular studies that individual spikes with complex spikes can have ISIs of more than 10 ms, we adopted a 14ms ISI time window to separate single-spikes from bursts. As for the detected burst, we separate data from the first single spike only to account for the number of bursts and not the intraburst frequency.

Spike phase-locking computation

We quantify the relative phase of a spike to an oscillation using the phase-locking value (PLV):

$$PLV(f) = \left| \frac{1}{N} \right| \sum_N e^{i\phi(f,n)},$$

with f the frequency and N the number of spikes [73]. For theta oscillations we assume 3-12 Hz and for gamma oscillations we assume 30-100Hz. The phase ϕ was obtained with the help of the Hilbert transform. The phase-locking statistics (PLS) accounts for uneven size in the data between bursting and single-spikes [74, 75]:

$$PLS(f) = \frac{1}{N-1} (PLV(f)^2 \times N - 1).$$

Declaration of competing interests

The authors have declared that no competing interests exist.

Data availability

Data will be made available on request.

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Figure 1

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