In vivo patch-clamp recordings reveal distinct subthreshold signatures and threshold dynamics of midbrain dopamine neurons

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The in vivo firing patterns of ventral midbrain dopamine neurons are controlled by afferent and intrinsic activity to generate sensory cue and prediction error signals that are essential for reward-based learning. Given the absence of in vivo intracellular recordings during the last three decades, the subthreshold membrane potential events that cause changes in dopamine neuron firing patterns remain unknown. To address this, we established in vivo whole-cell recordings and obtained over 100 spontaneously active, immunocytochemically-defined midbrain dopamine neurons in isoflurane-anaesthetized adult mice. We identified a repertoire of subthreshold membrane potential signatures associated with distinct in vivo firing patterns. Dopamine neuron activity in vivo deviated from single-spike pacemaking by phasic increases in firing rate via two qualitatively distinct biophysical mechanisms: 1) a prolonged hyperpolarization preceding rebound bursts, accompanied by a hyperpolarizing shift in action potential threshold; and 2) a transient depolarization leading to high-frequency plateau bursts, associated with a depolarizing shift in action potential threshold. Our findings define a mechanistic framework for the biophysical implementation of dopamine neuron firing patterns in the intact brain.
he midbrain dopamine system is necessary for essential brain functions related to reward-based learning, motivation, action, and cognition. Loss or dysregulation of dopamine neuron subpopulations located in the substantia nigra pars compacta (SNc) and ventral tegmental area (VTA) perturbs crucial circuits leading to major brain disorders, including Parkinson Disease (PD), schizophrenia, depression, and addiction. However, describing the dynamics of subthreshold ionic conductances that cause dopamine neuron spiking activity in vivo has remained a challenge.

Single-unit extracellular recordings from non-human primates and rodents have provided major insights into the computational roles of dopamine neuron firing, most prominently in the context of reward-based learning. Individual dopamine neurons signal prediction errors by transient, sub-second changes in their firing rates. A recent study also suggested that prediction error coding by dopamine neurons might occur in a distributed manner. Transient cue- or reward-associated increases in firing rate – from a low tonic background firing frequency (in the range of about 0.5–8 Hz) – are thought to encode positive reward prediction errors (RPEs), while transient reductions or pauses in baseline firing are believed to encode negative RPEs, induced for instance by the omission of an expected reward. Increasingly, experimental evidence points to additional dopamine functions, such as signaling salience, novelty and action control, as well as aversion. This functional diversity might also be reflected by the emerging molecular, cellular, and anatomical diversity of the midbrain dopamine system. Even within canonical RPE signaling, the interpretation of the transient burst has become more complex. It is currently divided into a short-latency component, which might signal the sensory characteristics of the stimulus (e.g., intensity); and a long-latency component, which might represent the predictive properties of the stimulus. In addition, recent progress has been made in characterizing the underlying arithmetic of single dopamine neuron activity in the context of RPE signaling.

To approach in vivo electrophysiology from the perspective of the dopamine neuron, it is reasonable to assume that it integrates excitatory and inhibitory inputs that are derived from hetero-geneous afferents, thereby dynamically tuning its intrinsic pacemaker activity. This process enables the generation of burst-pause firing patterns, which in turn contribute to temporally resolved extracellular dopamine concentration transitions in axon target regions. However, how inputs are integrated in dopamine neurons in vivo has remained unknown for the last three decades due to the lack of available tools. While the in vivo extracellular recording method does reveal spontaneous firing patterns of dopamine neurons, it is blind to the causal subthreshold processes that drive them. Intracellular recordings are unique in their ability to directly monitor and manipulate membrane potentials with full temporal resolution. Yet, the only available in vivo intracellular data on dopamine neurons came from the pioneering studies by Grace and Bunney in the early eighties. Here, we provide the first large dataset on subthreshold membrane potential properties of identified dopamine neurons in vivo by establishing a robust method for deep in vivo patch-clamp recordings in anesthetized adult mice. Our dataset comprises 110 spontaneously active, immunocytochemically identified dopamine neurons, which were mapped to their respective anatomical positions within the SNc and VTA. We discovered two qualitatively distinct subthreshold membrane potential signatures associated with transient high-frequency firing, which we named rebound and plateau bursting. Moreover, we demonstrate that in vivo pacemaking of dopamine neurons exhibits tightly controlled action potential thresholds, while rebound and plateau bursts are characterized by opposing shifts in action potential thresholds. Thus, our study provides a framework for mechanistic studies into in vivo firing patterns of midbrain dopamine neurons.

Results

Stable in vivo whole-cell recordings of immunocytochemically identified dopamine neurons reveal subthreshold membrane potential.

By establishing deep in vivo patch-clamp recordings, we obtained in vivo whole-cell recordings from 110 spontaneously firing, identified midbrain dopamine neurons in isoflurane-anaesthetized, adult C57BL/6 mice (n = 110, N = 78; Fig. 1). All neurons included in this study were filled with neurobiotin or biocytin and successfully recovered for anatomical mapping and post-hoc neurochemical identification as dopaminergic using tyrosine hydroxylase (TH) immunocytochemistry (Fig. 1c). Our coverage included most sub-regions of midbrain dopamine regions, with the majority of our recorded neurons located in the rostro-dorsal (i.e., parabrachial) VTA and rostro-medial SNc, with fewer neurons in more lateral aspects of the SNc and caudal ventro-medial (i.e., paranigral) VTA regions (Fig. 1d). Due to the unbiased nature of in vivo patch-clamping, we also encountered TH-negative, electrically silent cells with membrane potentials below ~50 mV as well as TH-negative, fast-firing cells (>10 Hz) with narrow action potentials (data not shown). Moreover, in addition to the 110 spontaneously active dopamine neurons included in this study, we observed 2 electrically silent TH-positive cells that only fired in response to depolarizing current injections, and 3 other TH-positive neurons that fired spontaneously less than one spike per minute (Supplemental Fig. 1). Under isoflurane-anaesthesia, the technical and biological stability of recordings in the whole-cell configuration were not a major limiting factor (Supplemental Fig. 2), and recordings could last for up to 60 minutes with stable input resistances. However, we usually terminated the recording in a controlled fashion via the outside-out patch configuration to retrieve the cell for post-hoc immunohistochemistry. Our dataset includes dopamine neurons recorded in male (n = 100, N = 73) and female (n = 12, N = 7) mice from two different C57BL/6 substrains (C57BL/6N (n = 92, N = 65) and C57BL/6j (n = 20, N = 15)). We also used two different variants of the internal pipette solution with a small difference in estimated free calcium concentrations of 40 nM (n = 8, N = 6) and 80 nM (n = 104, N = 74). As there were no significant differences in the mean firing rates (FR) and firing variabilities (quantified as coefficient of variation of interspike intervals (ISI)) among these data sets, they were pooled (see Supplemental Table 1). The input resistances were calculated for a subset of these neurons, which had a median of ~300 MΩ (Median (IQR) = 295.2 (228.1–349.8) MΩ; n = 42, N = 56; Fig. 1g), an order of magnitude higher than the previously reported values using sharp microelectrodes and more similar to a median of ~450 MΩ for synaptically isolated dopamine neurons recorded in vitro with the same pipette solution (Median (IQR) = 446.6 (301.3–656.7) MΩ; n = 71, N = 11; see Supplemental Table 2). Assuming that intrinsic conductances are similar when recording either in vitro or in vivo, our results imply a contribution of ~1 nS synaptic and 2 nS intrinsic conductances in dopamine neurons in vivo. This ratio suggests that intrinsic conductances of dopamine neurons are less likely to be shunted in comparison to other types of central neurons. In addition, subthreshold membrane potential responses to in vivo hyperpolarizing current injections were recorded in a subset of cells, which revealed a diversity reminiscent of previous in vitro studies (Supplemental Fig. 3).

Figure 1b shows a typical in vivo patch-clamp recording of an identified midbrain dopamine neuron. The recording, when presented on an expanded timescale, displays episodes of...
overshooting action potentials interspersed by resting periods where the membrane potential is hyperpolarized below the threshold for action potentials (Fig. 1b, bottom trace). Post-hoc immunocytochemistry identified this cell as a dopamine neuron localized in the ventral-caudal VTA (Fig. 1c, bottom image; and 1d, red circle).

Figure 1e, f show the mean firing rates and CVs of spontaneous firing activity from 110 dopamine neurons, respectively (mean firing rates median (interquartile range, IQR) = 5.25 (3.56–6.76); CV median (IQR) = 42.49 (30.29–65.34)). Among all measured variables, only mean firing rate was found to be significantly different between SNc and VTA dopamine neurons (Supplemental Fig. 4). In accordance to previous in vivo extracellular studies, we also analyzed the firing patterns of dopamine neurons using the traditional burst firing heuristic established by Grace and Bunney28. The results, expressed as the percentage of spikes fired in bursts (%SFB), were again comparable to extracellular recordings using the same anaesthesia19 (see Supplemental
neurons displayed a wide spectrum of CVs of ISIs (ISI CV) — and, with an unbiased approach, we focused on the overall variability of each neuron, plotted in Fig. 1h, illustrates the wide range of potentials associated with distinct frequencies sampled with in vivo patch-clamp recordings.

Bimodal distributions in subthreshold voltage predict variability of action potential firing pattern. To best utilize the direct readout of subthreshold membrane potentials provided by in vivo whole-cell recordings, we focused on a systematic investigation of the subthreshold membrane potentials associated with distinct firing patterns. To achieve this with an unbiased approach, we focused on the overall variability of firing (i.e., coefficient of variation (CV)) to characterize the activity of the recorded cell. As shown in Fig. 1f, dopamine neurons displayed a wide spectrum of CVs of ISIs (ISI CV) ranging from 12.9 to 165.7%. Upon inspection of the firing activity of a dopamine neuron with an ISI CV at the low end of this spectrum (12.9%; Fig. 1f, cyan diamond), we found a cell discharging in a single-spike pacemaker pattern (i.e. 0% of spikes fired in bursts (%SFB) according to the 80/160-ms criterion24; Fig. 2a–c). To characterize the subthreshold membrane potentials in a comparable fashion, we identified the voltage minima ($V_{\text{min}}$) of each ISI, i.e., the most hyperpolarized membrane potential between two action potentials (Fig. 2a, blue circles); as well as the voltage maxima ($V_{\text{max}}$) for each ISI, i.e., the action potential threshold (Fig. 2a, green circles; see methods for action potential threshold definition). The respective distributions of $V_{\text{min}}$ and $V_{\text{thr}}$ for this example neuron were then fitted with a Gaussian function (Fig. 2b, c, respectively), both of which exhibited a narrow unimodal distribution with low variability ($V_{\text{min}}$ CV = 2.7%; $V_{\text{thr}}$ CV = 1.9%).

For a dopamine neuron at the high end of the ISI CV spectrum (88.97%; Fig. 1f, blue diamond), the discharge pattern was more irregular and also characterized by transient increases in firing rate that were interspersed by longer silent periods (Fig. 2d–f). During these non-spiking periods, the membrane potential was more hyperpolarized compared to bursting periods (Fig. 2d). The distributions of $V_{\text{min}}$ (Fig. 2e) and $V_{\text{thr}}$ (Fig. 2f) were also plotted and fitted with Gaussian functions. In contrast to the dopamine neuron from the low end of the ISI CV spectrum, the $V_{\text{min}}$ distribution exhibited an overall higher variability ($V_{\text{min}}$ CV = 13.4%) and showed a clear bimodal distribution (Fig. 2e). In contrast, the $V_{\text{thr}}$ distribution for this cell was unimodal with low variability ($V_{\text{thr}}$ CV = 5.6%; Fig. 2f). These two examples of distinct in vivo dopamine neuron firing patterns suggested that the variability of subthreshold membrane potential minima (i.e. $V_{\text{min}}$ CV) captured by in vivo patch-clamp recordings might be related to spike train statistics. Indeed, when comparing the $V_{\text{min}}$ CV versus the variability of spike firing pattern (i.e. ISI CV) of the entire dopamine neuron population, a strong positive correlation was detected (Fig. 2g; $V_{\text{min}}$ CV vs ISI CV $R^2 = 0.58$; $p = 5.66 \times 10^{-20}$, $n = 97$, $N = 72$). Moreover, $V_{\text{min}}$ CV scaled with ISI CV over the entire dynamic range. The variability of the threshold potentials also displayed a clear correlation with spike train regularity (Fig. 2h; $V_{\text{thr}}$ CV vs ISI CV $R^2 = 0.32$; $p = 6.65 \times 10^{-10}$, $n = 97$, $N = 72$). Other parameters of subthreshold membrane potentials (mean of $V_{\text{min}}$ and $V_{\text{thr}}$) and their distributions (skewness and kurtosis of $V_{\text{min}}$ and $V_{\text{thr}}$) did not correlate well with the ISI CV (Supplemental Fig. 5). However, when we calculated the bimodality coefficients (BC) of $V_{\text{min}}$ and $V_{\text{thr}}$ – a measurement in which higher values are an indication of a bimodal distribution – the resulting values showed a strong, significant correlation with the ISI CV (Fig. 2i; $V_{\text{min}}$ BC vs ISI CV $R^2 = 0.29$; $V_{\text{thr}}$ BC vs ISI CV $R^2 = 0.09$; $p = 0.001$). In summary, in vivo spike train variability of dopamine neurons was well-correlated, not just with the overall variability of $V_{\text{min}}$ and $V_{\text{thr}}$, but more specifically with the bimodality coefficient of the $V_{\text{min}}$ and $V_{\text{thr}}$ distributions. Like the spike train variability itself, the bimodality coefficient was distributed as a spectrum for the recorded dopamine neuron population (Fig. 2i–j).

Pacemaker firing pattern is characterized by unimodal distributions of voltage minima and action potential thresholds. Based on this analysis, we proceeded to further characterize the subthreshold membrane potentials of low ISI CV (i.e. pacemaker-like) firing patterns of in vivo dopamine neurons, which were now identified by their low BC (lower 10% of the population). Figure 3a, b shows a representative recording and immunocytochemical images of one such neuron located in the VTA. The bottom trace in Fig. 3a illustrates the stability of the in vivo firing pattern when identified by subthreshold membrane potential properties, $V_{\text{min}}$ and $V_{\text{thr}}$ (blue and green circles, respectively). Similar to the example neuron illustrated in Fig. 2a–c, the $V_{\text{min}}$ and $V_{\text{thr}}$ distributions of this neuron displayed unimodal distributions (Fig. 3c, d, respectively). Collectively, dopamine neurons with unimodal $V_{\text{min}}$ and $V_{\text{thr}}$ distributions (i.e. BC < 0.53, see Freeman and Dale34 for detailed discussion of criteria) had a mean $V_{\text{min}}$ BC of 0.29 ± 0.02 and mean $V_{\text{thr}}$ BC of 0.32 ± 0.05 ($n = 10$, $N = 9$). To further examine the relationship between these subthreshold membrane properties and spike firing properties, individual $V_{\text{min}}$ and $V_{\text{thr}}$ were plotted against their respective ISI durations (Fig. 3e). As the plot shows, this example dopamine neuron with a low $V_{\text{min}}$ BC had a narrow range of ISIs, creating a small cluster each for both $V_{\text{min}}$ and $V_{\text{thr}}$. To obtain a visual representation of a stereotypical subthreshold membrane potential signature of in vivo pacemaking activity, now defined as a combined profile of unimodal $V_{\text{min}}$ and $V_{\text{thr}}$ distributions, we utilized event-triggered averaging by superimposing pacemaking spikes (Fig. 3f). The resulting
avg.

averaged trace (Fig. 3f, bold trace) portrays a low-variance subthreshold signature with a ~400-ms gradual, ramp-like membrane potential depolarization of about 10 mV until reaching the action potential threshold. Figure 3g shows the narrow firing frequency distribution of this example cell. The frequency distribution of the population of pacemaker dopamine neurons identified by their low BCs is shown in Fig. 3h. The spread of this distribution is due to the wide range of pacemaking firing frequencies among all the recorded neurons.

For a more detailed insight, we analyzed the phase plots of action potentials in these pacemaker firing sequences. Figure 3i illustrates an example of a three-action potential sequence (black–orange–red) during pacemaker firing. A phase plot for all spikes occurring during pacemaker firing in this example dopamine neuron is shown in Fig. 3j, with the same three colors (black–orange–red) representing the average of each spike. Note the large degree of overlap throughout the different phases of the action potentials, from threshold to upstroke and overshoot, as well as during repolarization and afterhyperpolarization. Quantitative analysis of all dopamine neurons discharging in this in vivo pacemaker mode identified by their subthreshold properties confirmed a high degree of stability and low variability that extended from the subthreshold domain to the action potential. Indeed, we did not observe significant changes in threshold voltage, maximal depolarization (Max dV/dt), or repolarization (Min dV/dt) speeds across the sequence of three action potentials among all low V_{min} BC dopamine neurons (ΔV_{thr}; spike 1 vs spike 2 median (IQR) = -0.002 (-0.0043 to 0.0012) mV; spike 2 vs 3 = 0.0016 (-0.0035 to 0.0043) V; Δmax dV/dt: spike 1 vs 2 = -0.0124 (-0.021 to 0.13) V/s, spike 2 vs 3 = -0.0174 (-0.0387 to 0.002) V/s; p = 0.375, 0.2324, 0.4316, and 0.02 for all comparisons; n = 10, N = 9; range of 3-spike sequences per cell = 78–691; Fig. 3k and Supplemental Table 2). The low variability of subthreshold membrane potentials combined with stable action potential dynamics demonstrates the capacity of dopamine neurons to maintain their firing stability in vivo even when embedded within active circuits.

Rebound bursts are characterized by bimodal distribution in voltage minima and hyperpolarizing shift in action potential threshold at the beginning of the burst.

Next, to further explore the subthreshold membrane potential activity of dopamine neurons on the other end of the spectrum, we studied the dopamine neurons in which subthreshold V_{min} BC was in the top 10% (V_{min} BC range = 0.66–0.83). A representative recording and images of immunocytochemistry of such a dopamine neuron located in the SNc is shown in Fig. 4a, b. Similar to the dopamine neuron described earlier (Fig. 2d–f), the distribution of V_{min} from a bursting dopamine neuron was well-described by a bimodal Gaussian distribution (Fig. 4c, Mode 1 and Mode 2; V_{min} BC = 0.75 ± 0.06; Mode 1 V_{min} mean = -46.3 ± 2.8 mV, standard deviation (SD) = 2.7 ± 1.6 mV; Mode 2 V_{min}
mean $= -37.9 \pm 3.7$ mV, SD $= 1.5 \pm 0.4$; $n = 10$, $N = 10$). As in Fig. 2f, the $V_{\text{thr}}$ distribution was also unimodal (Fig. 4d; BC $= 0.38 \pm 0.09$; $V_{\text{thr}}$ mean $= -28.91 \pm 2.051$ mV, SD $= 1.73 \pm 0.44$ mV). When plotting the individual ISIs against the corresponding $V_{\text{min}}$, the more hyperpolarized membrane potentials (Fig. 4c, Mode 1) were almost exclusively associated with longer ISIs in the range of 600–1200 ms, forming an independent cluster in the plot (Fig. 4e, dark blue circles), whereas the more depolarized $V_{\text{min}}$ were associated with a wide range of shorter ISI durations (Fig. 4e, light blue circles). In contrast, the distribution of $V_{\text{thr}}$ in this dopamine neuron had a narrow membrane potential range (Fig. 4e, green circles).

We then applied event-triggered averaging to spikes that immediately followed the Mode 1 $V_{\text{min}}$ events, which revealed a
Fig. 3 Single-spine firing pattern is characterized by a stable and narrow subthreshold membrane potential dynamic range. a Representative recording of spontaneous in vivo electrical activity from a dopamine neuron in the VTA. The upper trace displays action potentials during a stable recording for >60 s. As in Fig. 2, blue dots in the lower trace represent $V_{min}$ while green dots represent $V_{thr}$. b Immunocytochemical identification demonstrates the recorded neuron (red) as TH-positive (green) within the VTA (arrow). Depiction shows the location of the recorded neuron (large red dot) in relation to all the recorded neurons in this study (gray and black dots). Black dots represent the locations of all the neurons that fired with a low $V_{min}$ BC (lower 10%). c Distribution of all $V_{min}$ from the cell in a, b, demonstrating a unimodal distribution (blue line = single Gaussian fit). d Distribution of all subthreshold voltage maxima ($V_{thr}$) from the cell in a, b, demonstrating an unimodal distribution (green line = single Gaussian fit). e Scatter plot of the interspike interval (ISI) versus $V_{min}$ (blue) and $V_{thr}$ (green). f High temporal resolution of single-spine firing pattern (upper panel), and a mean spike-triggered pacemaker pattern (black trace in lower panel) with overlaid (gray) traces. g Histogram of the instantaneous firing rate for the cell in a, b, h Histogram of instantaneous firing rate for all cells with a low $V_{min}$ BC. i Example trace from the cell firing in a single-spine pattern with three action potentials in order from black, then orange, then red. Inset illustrates where the threshold occurs for the black action potential. j The phase plots for each action potential are displayed in the same color scheme as in i (black, orange, and red; plots are nearly completely overlapping). k Summary data of the changes in threshold voltage (left) and maximal depolarization rate (Max dV/dt; right) across the sequence of three action potentials for all cells firing with a low $V_{min}$ BC (two-sided Wilcoxon signed-rank test).

...stereotypical, low-variance subthreshold membrane potential signature and associated firing (Fig. 3f). The resulting subthreshold signature was characterized by a curved, long duration (668.5 ± 213.1 ms) trajectory of initial membrane hyperpolarization (10.5 ± 2.7 mV of hyperpolarization to a $V_{min}$ of −47.99 ± 2.5 mV) followed by membrane depolarization back to threshold. The rebound firing occurring after large hyperpolarizations was associated with a transient increase in firing in the beta frequency range. Accordingly, we termed these events rebound bursts. In contrast, event-triggered averaging of spikes following the Mode 2 $V_{min}$ events did not result in a stereotypical low-variability subthreshold waveform. The frequency distribution associated with the rebound burst significantly widened the range of dopamine neuron firing (Fig. 4g, h) as compared to pacemaking (pacemaker firing rate SD: 5.5 Hz; rebound burst firing rate SD: 6.9 Hz; F-test $p = 1.064 \times 10^{-30}$), both on the level of individual cells and, in particular, on the level of dopamine neuron populations (compare Figs. 3g and 4g).

The phase plot of the rebound burst action potentials shows different dynamics compared to that of pacemaking (Fig. 4i–k). In particular, the first action potential of the rebound burst (Fig. 4i, orange) had a significantly more hyperpolarized action potential threshold accompanied by an accelerated speed of depolarization (Max dV/dt, Fig. 4j, k). Once within the rebound burst, the threshold and depolarization speed return close to the pre-burst baseline already with the following spike ($\Delta V_{thr}$; spike 1 vs spike 2 median (IQR) = −3.1 (−5.6 to −2.4) mV, spike 2 vs 3 = 1.0 (0.9–1.1) mV; $\Delta$Max dV/dt; spike 1 vs 2 = 16.3 (13.6–21.9) V/s, spike 2 vs 3 = −5.3 (−8.3 to −3.2) V/s; $p = 0.02$ for all comparisons; $n = 10, N = 10$; range of 3-spike sequences per cell = 9–95; Fig. 4k and Supplemental Table 2). Thus, rebound burst dopamine neurons have distinct subthreshold membrane potential signatures with two modes of $V_{min}$ which widen their frequency range, and their observed threshold dynamics occur at burst onset.

Plateau bursts are characterized by bimodal distribution in action potential threshold and depolarizing shift in action potential threshold at the end of the burst. As apparent from Fig. 2j, some dopamine neurons possessed high BC values for $V_{thr}$. To explore the features of this particular subthreshold membrane potential signature, we analyzed the dopamine neurons in the top 10% of $V_{thr}$ BC. An example trace and its corresponding immunocytochemical images of an identified VTA dopamine neuron are shown in Fig. 5a, b. The Fig. 5a bottom trace illustrates that, in contrast to the rebound burst mode, the action potential threshold of this type of neuron showed a depolarizing shift (light green circles). As expected from this observation, the distribution of $V_{thr}$ was clearly bimodal (Fig. 5d; $V_{thr}$ BC: 0.61 ± 0.13; Mode 1 mean = −33.7 ± 8.9, SD = 1.3 ± 0.5; Mode 2 mean = −27.5 ± 9.8, SD = 2.0 ± 1.1; $n = 10, N = 10$), a unique feature that is not observed in pacemaking or rebound bursting. In this particular example, the distribution of $V_{min}$ also displayed a high BC and bimodal distribution (Fig. 1c), although this was not a consistent feature of cells displaying a bimodal distribution in $V_{thr}$ (Fig. 5c; $V_{min}$ BC: 0.59 ± 0.15; Mode 1 mean = −37.7 ± 5.2, SD = 2.7 ± 1.8; Mode 2 mean = −28.2 ± 5.9, SD = 1.2 ± 0.4). When plotting the ISI durations against their respective $V_{thr}$ (Fig. 5e), those $V_{thr}$ residing in the more depolarized distribution mode (Mode 2 in Fig. 5d) were exclusively associated with very short ISI durations in the range of 6–26 ms (Fig. 5e, light green circles). In other words, the bimodal $V_{thr}$ distribution identified high-frequency burst discharges.

Event-triggered averaging of action potentials associated with the $V_{thr}$ Mode 2 distribution revealed a distinct low-variability subthreshold membrane potential signature (Fig. 5f, bottom trace). What emerged was a sharply-rising membrane potential plateau of approximately 6 mV in amplitude and 35 ms in duration with a limited number of 2 to 3 high-frequency, gamma-range spikes riding on top (plateau amplitude median (IQR) = 5.2 (3.7–6.2) mV; plateau duration = 45.8 (17.9–63.3) ms; number of spikes = 2.1 (2.0–2.3); mean maximum spike frequency = 41.4 (27.3–83.4) Hz). As these spikes occurred on a depolarized “plateau” of the membrane potential, we termed this discharge pattern plateau bursts. As evident from the average trace (Fig. 5f, bold trace), plateau burst intraburst frequency was less variable compared to intraburst frequencies in rebound bursting (compare Figs. 4f and 5f). Figure 5g, h show the firing frequency distribution of the example cell and the entire population of dopamine neurons identified as discharging in the plateau burst mode, respectively. In contrast to rebound bursting, the firing frequency distribution of plateau bursts is almost exclusively in the gamma range.

When plotting the action potential trajectories of plateau bursts, in contrast to rebound bursts, we found no change in the action potential threshold voltage or waveform properties at the onset of plateau bursts (Fig. 5i–k). This implies that no systematic or predictive threshold dynamics precede the occurrence of plateau bursts in these dopamine neurons, unlike the burst-predictive threshold dynamics of rebound bursts. Or, stated from the neuron’s perspective, it was “surprised” by the burst. However, after the onset of plateau bursts, the action potential threshold and dynamics dramatically changed, as subsequent action potentials (in most cases the second action potential within bursts) displayed a large depolarizing shift in the threshold potential accompanied by reduced depolarization speed (Fig. 5j, k), both of which are indications of reduced excitability ($\Delta V_{thr}$; spike 1 vs 2 median (IQR) = 0.39 (−0.35–0.92) mV, $p = 0.19$; spike 2 vs 3 = 4.66 (3.62–4.43) mV, $p = 0.002$; $\Delta$Max dV/dt; spike...
1 vs 2 = −0.5 (−3.25–0.11) V/s, \( p = 0.11 \); spike 2 vs 3: −14.38 (−17.95 to −6.73) V/s, \( p = 0.02 \), \( n = 10 \), \( N = 10 \); range of 3-spike sequences per cell = 4–89; Supplemental Table 2). Thus, the threshold dynamics of the plateau burst not only occur in the opposite direction compared to those in rebound bursts, but also served to time the end, not the beginning, of the burst. Indeed, in over 90 percent of all plateau bursts, the second action potential with the large depolarizing shift in the threshold was also the terminating action potential of the burst (plateau bursts with 2 spikes, \( n = 455 \); 3 spikes, \( n = 48 \); 4 spikes; \( n = 11 \), \( N = 11 \)). In essence, our unbiased exploration of subthreshold membrane voltages from dopamine neurons in vivo uncovered two unique...
subthreshold burst signatures featuring opposing threshold dynamics, each associated with a distinct and non-overlapping frequency range.

Dopamine neurons in vivo fire both Plateau and Rebound bursts during a single recording.

As highlighted in Fig. 2, subthreshold membrane potential features – like spike train statistics – were best described as a spectrum. Indeed, many of the recorded dopamine neurons in the intermediate BC range (10–90%) fired both single spikes and occasional burst-like events (i.e. “mixed mode”). In addition, a handful of dopamine neurons fired in both plateau and rebound bursts as well as in single spikes (Supplemental Fig. 6). A representative trace and immunocytochemical images of an identified VTA dopamine neuron that fired in single spikes with occasional burst events are shown in Fig. 6a, b, respectively. Figure 6a bottom trace illustrates the detail of single spikes interspersed with hyperpolarizing events leading to rebound bursts. This mode of firing was associated with wider unimodal distribution in the $V_{\text{min}}$ and $V_{\text{thr}}$ (Fig. 6c, d, respectively), sometimes with prominent tails on either or both sides of the distributions; in this case on the left side of the $V_{\text{min}}$ distribution. When plotting individual ISIs against their respective $V_{\text{min}}$ and $V_{\text{thr}}$ (Fig. 6e), wider clustering was observed with no clear segregation. Therefore, we identified events within spike trains by taking advantage of the two dimensions in the scatter plots (Fig. 6e, see Methods section for details). In this example, outlier detection for both ISI (ordinate) and membrane potential (abscissa) values isolated events with hyperpolarized $V_{\text{min}}$ and long ISI durations (Fig. 6e, dark blue circles), but not for $V_{\text{thr}}$ (Fig. 6e, green circles). Based on the detected events, the event-triggered averaging revealed a subthreshold membrane potential signature similar to the one observed for rebound bursting (Fig. 6f, bottom trace), indicating that the mechanism of burst generation might be similar between dopamine neurons with continuous bursting and those displaying a more “mixed” firing mode. Figure 6g, h show the firing frequency distribution of the example cell and the entire population of dopamine neurons identified as discharging in a mixed pattern, respectively. Note that event-triggered averaging captured fewer high frequency events compared to continuously bursting cells.

In this dopamine neuron firing in a mixed pattern, the phase plot of action potentials in rebound bursts shows similar properties compared to that of dopamine neurons locked in a rebound burst mode (Fig. 6i–k). Again, the first action potential of the rebound burst (Fig. 6i, orange) had a significantly more hyperpolarized action potential threshold coupled with an accelerated speed of depolarization (Max $dV/dt$; Fig. 6j, k). Once within the rebound burst, the threshold as well as depolarization speed quickly returned to baseline with the subsequent spike ($\Delta V_{\text{thr}}$; spike 1 vs spike 2 median (IQR) = $-2.3 (-2.6$ to $-1.8)$ mV, spike 2 vs 3 = 0.6 (0.5–0.9) mV; $\Delta$ max $dV/dt$; spike 1 vs 2 = 7.18 (6.04–8.98) V/s, spike 2 vs 3 = $-1.91 (-2.79$ to $-1.13)$ V/s; $p = 4.88 \times 10^{-4}$ for all comparisons; $n = 12$, $N = 12$; range of 3-spike sequences per cell = 1–15; Fig. 6k; and Supplemental Table 2 for cells firing a mixed pattern containing plateau bursts). Thus, the observed threshold dynamics of rebound and plateau bursts in dopamine neurons firing in a mixed pattern are similar to rebound and plateau bursts from dopamine neurons firing in a locked burst mode. Indeed, many of the intrinsic properties were similar among neurons firing in different patterns (Supplemental Fig. 7). The results indicate that firing activity of most dopamine neurons is not well represented by a single state, but is constantly changing in the in vivo environment.

In summary, the first systematic in vivo patch-clamp study of identified midbrain dopamine neurons demonstrated the presence of two qualitatively distinct burst firing mechanisms, identified by their bimodal subthreshold distributions, with contrasting subthreshold membrane potential signatures and diametrically-opposing action potential threshold dynamics. These findings present the first framework of how predictive coding might be biophysically implemented in midbrain dopamine neurons.

Discussion

For the first time in nearly four decades, our study provides new intracellular in vivo recordings of identified midbrain dopamine neurons. By establishing deep in vivo whole-cell patch-clamp recordings in anaesthetized mice with no noticeable perturbations on firing activity, our large dataset allowed the first systematic analysis of the subthreshold membrane potential range of dopamine neurons in the intact brain. We show that the overall regularity of action potential discharge, captured by the coefficient of variation (CV) of the spike train, correlated well with the CV of the minima ($V_{\text{min}}$) and maxima ($V_{\text{thr}}$) in the subthreshold membrane potential range. In addition, we demonstrate that bimodality in the distributions of $V_{\text{min}}$ and $V_{\text{thr}}$ enabled the identification of unique subthreshold signatures associated with distinct burst patterns. We defined (1) a prolonged membrane potential hyperpolarization leading to rebound bursts, which were associated with a hyperpolarizing shift in action potential threshold at the onset of the burst; and (2) a fast depolarization leading to plateau bursts, which were associated with a depolarizing shift in the action potential threshold at the end of the burst. While plateau bursts were strongly coupled to short, high-frequency (gamma range) discharge, rebound bursts occurred at lower frequencies (beta range), some of which would not have been detected by traditional heuristics.
Both plateau and rebound bursts substantially widen the dynamic range of dopamine neuron firing, providing attractive candidate mechanisms that enable predictive coding. The hyperpolarizing shift of the action potential threshold induced in rebound bursts also enhances the potential excitability of the neuron. At the same time, these dopamine neurons remain flexible to allow integration of incoming synaptic inputs. Moreover, the action potential thresholds of rebound burst spikes are distinct from those of the background single spikes in that they would enable the cell to compute a new prediction for spike timing. Future in vivo whole-cell studies in awake behaving animals will be necessary to explicitly test whether these candidate subthreshold mechanisms are indeed at work during ongoing reward prediction error computations.
In contrast, the observed threshold dynamics of the short, high-frequency plateau bursts might primarily serve temporal precision of firing by controlling the termination of high-frequency bursts. These features render the plateau burst mechanism a plausible candidate for coding unexpected and salient sensory events. While the subthreshold signature of the rebound burst suggests a timed series of initial inhibition followed by excitation, short plateau bursts might be generated by rapid fluctuations in the excitation/inhibition (E/I) balance. Indeed, our previous dynamic clamp studies in vitro revealed similar threshold and action potential waveform behavior when the E/I balance rapidly shifts toward more excitation or disinhibition. In essence, our study identified in vivo operating principles of discharging dopamine neurons and provides explicit candidate mechanisms, which can now be tested in awake animals.

While the in vivo intracellular recordings with sharp microelectrodes of identified dopamine neurons from the 1980s did not provide systematic analyses of the subthreshold membrane potential range in relation to firing patterns, those results are mostly in accordance with our findings. There is, however, one important exception. Input resistances in our in vivo patch-clamp recordings of dopamine neurons are about 10-fold higher than those reported previously, indicating that the in vivo "high conductance" state might have been overestimated and, in part, be caused by impaling dopamine neurons with sharp microelectrodes. In addition, it was only now possible to determine that the in vivo intracellular recordings did not alter the firing pattern of intact cells – both on-cell recordings in this study and by comparison to previous in vivo extracellular recordings of identified dopamine neurons. Importantly, all dopamine neurons included in our study were filled, recovered, and identified as TH-immunopositive. Given these quality control measures and the size of the dataset, we can begin to also address the in vivo subthreshold diversity of dopamine neurons. The previously reported differences in in vitro intrinsic subthreshold properties, such as sag amplitudes and rebound latencies, were also evident in our in vivo dataset, although dopamine neurons in the paragangal aspect of the VTA were underrepresented. Also, our present in vivo patch-clamp study does not provide information about axonal projections of the recorded dopamine neurons. However, the observed diversity might contribute to subpopulation-specific computations, as also recently identified by in vivo GCaMP imaging during a complex behavioral task.

Despite the strengths, our study has a limitation in that we recorded under anesthesia and were therefore unable to study subthreshold membrane potential signatures of dopamine neuron firing patterns in response to sensory cues or reward-related signals. However, the stability of brain states in controlled anesthesia is also an advantage as it enabled the unbiased identification of subthreshold membrane potential signatures. Another limitation of our study is that we did not go beyond the identification of subthreshold patterns to explicitly target underlying synaptic and intrinsic biophysical mechanisms. For instance, the deep hyperpolarization that precedes rebound bursting could be activated by different types of synaptic input involving distinct receptors ranging from GABA_{A}, GABA_{B}, dopamine receptor D2 (D2R), or metabotropic glutamate receptors (mGluR) coupled to small conductance calcium-activated potassium (SK) channels. In addition, the rebound kinetics are likely to be controlled by intrinsic conductances in the subthreshold range. Here, rebound depolarization accelerating (e.g. HCN channels) and delaying (e.g. Kv4 channels) conductances are likely to be relevant. Some dopamine neurons even respond to the termination of a hyperpolarizing current injection with a robust recruitment of T-type calcium channels. The plateau burst, on the other hand, is driven by a rapid onset depolarization, which is likely to be caused by a rapid change in the E/I balance. Fundamentally, rebound and plateau bursts represent responses within dopamine neurons comprised of a sequence of synaptic inputs that are active at any moment in the intact brain in addition to the intrinsic cellular processes that are particular to the individual neuron. Future studies exploring detailed intrinsic biophysical properties of in vivo dopamine neurons in the context of synaptic input will be necessary to determine the degree to which synaptic inputs contribute to rebound and plateau bursts in dopamine neurons. In addition, dynamic clamp and pharmacological manipulations could be performed to further characterize intrinsic activity underlying these burst patterns. In essence, in vivo patch-clamp studies will facilitate future mechanistic studies of dopamine neuron firing in anesthetized as well as awake behaving animals.
Customized head-plates were anchored to the skull with superglue and dental cement (Paladur, Kulzer), as well as with screws mounted on the skull.

**In vivo electrophysiology.** Electrophysiology was performed using glass electrodes (5–10 MΩ) filled with internal solution containing (in mM): 135 K-gluconate, 10 HEPES, 5 KCl, 5 MgCl₂, 0.1 EGTA, 0.075 CaCl₂ (variant 1) or no added CaCl₂ (variant 2), 5 ATP (Na), and 1 GTP (Li), neurobiotin (0.1%; Vector Laboratories) or biocytin (0.3%), at pH 7.35. The estimated free calcium concentrations based on a calcium contamination of 15 µm (for details see Woehler et al.49) were about 80 nM for variant 1 and ~40 nM for variant 2 (variant 1: $n = 104$, $N = 74$; variant 2: $n = 8$, $N = 6$). Recording electrodes were made with borosilicate glass capillaries (G120F-4, Warner Instruments) pulled with a pipette puller (DMZ Universal Electrode Puller, Zeitz-Instruments).

Animals were maintained under isoflurane (1.0–2.5% in 100% O₂, 0.35 L/min) and head-fixed to a customized recording platform (Luigs & Neumann). The body
temperature (37–38 °C) and breathing rate (1-2 Hz) were continuously monitored. Cranioptomies and duratomies were performed at the reference locations. The location of all neurons that fired in an intermediate V_{rmin} BC (10–90%) ([c, d) Distribution of V_{rmin} (c) and V_{thr} (d) from the cell in a, b, e Scatter Plot of ISI versus V_{rmin} and V_{thr}. Blue dashed lines indicate outlier identification in the upper left plane (dark blue circles) for V_{rmin}. Green dashed lines indicate outlier identification in the lower left plane for V_{thr} with no events detected. f High temporal resolution of a rebound burst (upper panel), and event-triggered rebound burst trace (lower panel). In both trace, gray trace and overlaid (gray) trace. g Histogram of the cell firing rate associated with rebound bursts. h Histogram of firing associated with rebound bursts for all cells that fired in a mixed pattern. i Example trace of the cell firing a rebound burst with three action potentials in order from black prior to the burst, then orange beginning the burst, and red. Inset illustrates where the action potential threshold occurs prior to the burst (black). j The phase plots for each action potential are displayed in the same color scheme as in i. k Summary data of the changes in action potential threshold voltage (left) with significant hyperpolarization in threshold of the first action potential of the burst (orange; p = 4.88 × 10^{-5}), and change in maximal depolarization rate (Max dV/dt; middle; p = 4.88 × 10^{-5}; two-sided Wilcoxon signed-rank test) with significant increase in Max dV/dt, across the sequence of three action potentials for all rebound bursts from cells firing in a mixed pattern.]}

Detection of spike threshold was dependent on the dV/dt and d2V/dt^2 crossing certain thresholds. To make the algorithm flexible and robust, these thresholds were indexed to the maximal values of the first and second voltage derivative for each detected spike. Specifically, the spike threshold is detected when the following conditions are met within a 2-ms window ending at the spike peak: Both first and second derivative of voltage cross 5% of their respective within-window maximal values for two consecutive samples. This ensured that spike threshold determination was accurate. A threshold was said to be terminated when the instantaneous firing rate fell below the median. In non-bimodal cells, deep hyperpolarizations showed up as outlier in the ISI vs V_{thr} scatter plot. As such, we used Tukey’s method to detect outlier ISI and V_{thr} values. If a V_{thr} was more hyperpolarized than q1-2.5x (q3-q1), where q1 = 25th quantile and q3 = 75th quantile in the V_{thr} distribution; and if the corresponding ISI was longer than q1-2.5x (q3-q1), where q1 = 25th quantile and q3 = 75th quantile in the ISI distribution, it was classified as a deep hyperpolarization. Identification of the subsequent burst followed the same criteria as for bimodal cells. The factor 2.5 was derived from analysis of bimodal cells and to optimize detection in non-bimodal cells.

Identification of rebound and plateau bursts. Rebound bursts: A Gaussian mixture model with two Gaussians was fitted to the V_{min} distribution of neurons with the highest (top 10%) V_{thr} BC values. The voltage value midway between the crossing point of the two individual Gaussians and the peak of the leftward Gaussian was chosen as the threshold for classifying V_{min} values as deep hyperpolarizations. If either of the two ISIs following a deep hyperpolarization were faster than the cell’s median firing rate, they were identified as a rebound burst. The burst was said to be terminated when the instantaneous firing rate fell below the median. In non-bimodal cells, deep hyperpolarizations showed up as outliers in the ISI vs V_{thr} scatter plot. As such, we used Tukey’s method to detect outlier ISI and V_{thr} values. If a V_{thr} was more hyperpolarized than q1-2.5x (q3-q1), where q1 = 25th quantile and q3 = 75th quantile in the V_{thr} distribution; and if the corresponding ISI was longer than q1-2.5x (q3-q1), where q1 = 25th quantile and q3 = 75th quantile in the ISI distribution, it was classified as a deep hyperpolarization. Identification of the subsequent burst followed the same criteria as for bimodal cells. The factor 2.5 was derived from analysis of bimodal cells and to optimize detection in non-bimodal cells.

Phenotype: For dopamine neurons with the highest (top 10%) V_{thr} BC values, the voltage criteria for classifying V_{min} values as plateau potentials was calculated in a similar way to deep hyperpolarizations, with the only difference being that it was calculated on V_{min} distributions and the second Gaussian in the V_{min} distribution lay to the right of the main Gaussian. All ISIs whose V_{min} values were higher than this criterion were considered part of the plateau burst. In non-bimodal cells, plateau potentials were detected in a parallel fashion to deep hyperpolarizations, with 2.5 used as the factor for V_{min} outlier detection and 0.5 as the factor for ISI outlier detection, in addition to the V_{min} being 3.5 mV more depolarized than its immediate predecessor for it to be identified as a plateau burst. Also, the corresponding ISI has to be shorter than q1-0.5x (q3-q1), where q1 = 25th quantile and q3 = 75th quantile in the ISI distribution. Similar to rebound burst detection, these criteria were derived from bimodal cells and with an aim to optimize detection in non-bimodal cells.

Quantification and statistical analysis. Data processing and quantifications were performed using MATLAB (MathWorks). Wilcoxon rank-sum tests were used to determine statistical differences using MATLAB (MathWorks). Graphs and maps were created using MATLAB or Prism 6. Bimodality coefficient was used for dual process detection.4 Measurements of effect size were also calculated.40 Parametric data were presented as mean ± standard deviation. Non-parametric data were presented as median with interquartile range. Wilcoxon rank-sum test and Wilcoxon sign rank tests were used for unpaired and paired data. Adjusted R squared values are reported for linear correlation.

Reporting summary. Further information on research design is available in the Nature Research Reporting Summary linked to this article.

Data availability. The data used to generate the results are available at a Harvard Dataverse repository at https://dataverse.harvard.edu/dataset.xhtml?persistentId=doi:10.7910/DVN/4Q2X1LZ. Source data are provided with this paper.
Code availability
The custom code used for analyzing this data, along with sample data and documentation, has been uploaded to a Github repository at https://github.com/annandsku/DA_invivo.

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