D-β-hydroxybutyrate stabilizes hippocampal CA3-CA1 circuit during acute insulin
resistance.

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1. **Abstract**

The brain primarily relies on glycolysis for mitochondrial respiration but switches to alternative
fuels such as ketone bodies (KB) during low glucose availability. Neuronal KB uptake, which does
not rely on the glucose transporter 4 (GLUT4) and insulin, has shown promising clinical

applications in alleviating the neurological and cognitive effects of disorders with hypometabolic
components. However, the specific mechanisms by which such interventions affect neuronal
functions are poorly understood. In this study, we pharmacologically blocked GLUT4 to
investigate the effects of the exogenous KB D-β-hydroxybutyrate (D-βHb) on mouse brain
metabolism during acute insulin resistance (AIR). We found the impact of D-βHb to be distinct
across neuronal compartments: AIR decreased synaptic activity and LTP, and impaired axonal
conduction, synchronization, and action potential (AP) properties. D-βHb rescued neuronal
functions associated with axonal conduction, synchronization and LTP.

Keywords: hippocampus, ketone bodies, beta-hydroxybutyrate, ketogenic diet, insulin
resistance, GLUT4, hypometabolism, Schaffer collaterals, pyramidal neurons

2. **Introduction**

Over the past two decades, research on insulin signaling in the brain has become increasingly
important in the contexts of type 2 diabetes mellitus (T2DM) and Alzheimer's disease (AD) [1,2].
Insulin plays a critical role in memory formation processes in the hippocampus, such that systemic
insulin resistance can interfere with hippocampal metabolism and cognitive function, as shown by
various studies [3,4]. Additionally, insulin is an essential element in memory processing in the
hippocampus and a vital mediator of cognitive impairments in T2DM and AD, due to its function
in promoting GLUT4 translocation [5,6]. Although insulin from the blood crosses the blood-brain
barrier, research suggests supplementary local insulin synthesis and release in the brain [7] which
enhances the pro-cognitive effects of brain glucose [8]. Therefore, impaired glucose uptake in the
hippocampus may contribute to cognitive decline associated with aging, T2DM, and AD [9,10,11].

This relationship has led to the hypothesis that brain hypometabolism reflects encroaching
neuronal insulin resistance. The hippocampus is highly enriched with the GLUT4 receptor, which
is the primary neuronal insulin dependent glucose transporter, particularly in brain areas with high
concentrations of insulin receptors and relatively high neuronal activity [8,12,13]. Insulin acts by
enhancing GLUT4 translocation across neuronal membranes under regulation of
phosphatidylinositol 3-kinase (PI3K), which furthermore promotes memory formation [14,15].
During memory formation, GLUT4 is primarily expressed in the soma rather than in the more
metabolically active hippocampal neuropil [16,17], suggesting that insulin effects may be primarily
involved in maintenance of neuronal firing. Mouse models of AD or diet-induced obesity (DIO)
showed blunting of the pro-cognitive effects of intrahippocampal insulin injections, along with
decreased local glucose metabolism [18,19], coupled with neuronal hyperexcitability with
epileptiform spikes and impairment of GLUT4 translocation [20].

During periods of lower glucose availability, ketone bodies (KBs) are supplementing metabolism

as an alternative fuel [21,22]. Introduction of a ketogenic diet has shown pro-cognitive benefits in
patients suffering from T2DM or AD. These benefits may arise from a reduction in neuronal firing
rates during ketosis [23,24], increased ATP:ADP ratios [25], increased secretion of GABA due to
substantial formation of acetyl-CoA [26], equilibration of the NAD:NADH, and curtailed production
of free radicals, observed during ketogenic diet, together may help to temper neuronal
hyperexcitability and its metabolic consequences [27,28].

In addition, KBs play a crucial role in protecting mitochondria from acute metabolic stress by
preventing mitochondrial permeability transition (mPT) through their effects on intracellular
calcium levels [29,30]. Furthermore, KBs can affect neuronal firing by promoting the opening of
ATP sensitive K^+ channels (K-ATP), thus reducing the cytosolic pool of ATP generated from
glycolysis [24,30].

Exogenous KBs, such as D-β-hydroxy-butyrate (DβHb) ester, unlike a ketogenic diet, can raise
KB levels in circulation without severe dietary restrictions [31,32]. It is unclear whether
administration of exogenous KBs in the presence of normal plasma glucose levels maintains their
neuroprotective effects, but recent studies show that DβHb provided to healthy adults increased
their overall brain activity and stabilized functional networks [33,34].

Those observations led us to hypothesize that exogenous KB could be utilized to rescue neuronal
metabolism during cerebral insulin resistance. To test these hypotheses, we first established a
murine hippocampal model of acute insulin resistance (AIR) through the inhibition of GLUT4 by
administration of Indinavir [35,36]. We then studied how neuron-specific insulin resistance
affected hippocampal neurons in the CA3-CA1 circuit; a model circuit for studying learning and
memory. We tested the effects of D-βHb during AIR to establish the therapeutic potential of
exogenous KBs in hypoglycemic animals. In expectation of circuit-wide effects on synaptic and
axonal function, we obtained field potential recordings in hippocampus of the treated mice and
made patch-clamp recordings in hippocampal slices to assess the in vitro effects of AIR and D-
βHb on the electrophysiological properties of CA1 pyramidal neurons and CA1 fast-spiking
interneurons (FSI). Finally, we used computational modeling to establish a bridge between our
electrophysiological results and the dysfunction of Na^+/K^+ ATPase, as the potential reason for the
detrimental changes observed during AIR.

3. **Results**

Various paradigms of synaptic plasticity associated with learning and memory have been
identified in the hippocampus. There is extensive documentation of long-term potentiation (LTP)
and long-term depression (LTD), spike-timing-dependent plasticity, and EPSP-spike potentiation
in many hippocampal circuits, making the hippocampus a classic system for studying
neuroplasticity. Furthermore, the hippocampus's simple cytoarchitecture makes it an ideal model
system. In this study, we utilized Indinavir, a potent GLUT4 receptor blocker, to pharmacologically

induce AIR in the hippocampal slices Our investigation focused on the stratum radiatum of the
CA1, where numerous synapses are formed between Schaffer collaterals (SCO) and the apical
dendrites of the pyramidal neurons.

3.1 Synaptic activity and LTP, but not fiber volleys (FV), are adversely affected by AIR and are
not reversed by either 0.1 mM or 1 mM DβHb.

To test synaptic transmission within CA1 under a wide range of energetic demands, we measured
the circuit's response under 3 different stimulation paradigms. First, we applied mild physiological
stimulation (not imposing high energetic demands [37,38]), consisting of paired stimulation
applied at 25 Hz, repeated every 20 s over 30-60 trials [**Fig.1A-F**].

During the AIR condition, we observed a strong decrease in fEPSP amplitudes; -39.23 ± 8.72%
(1st stimulus) and -38.79 ± 8.93% (2nd stim.) compared to baseline (**Fig. 1A**). There was no
recovery of amplitude when D-βHb was applied at low (0.1 mM) (+13.10 ± 6.64% and +14.07 ±
6.76%, **Fig. 1A**, p=0.34, p=0.30) or high (1 mM) concentrations [**Fig.1A**, -5.98 ± 6.13% and -5.81
± 6.26%, p=0.88; p=0.56]. There were no effects of either AIR or D-βHb+AIR on PPR [**Fig.1B**,
Control: 1.89; AIR: 1.92; 0.1 mM D-βHb+AIR: 2.00; 1 mM D-βHb+AIR: 1.88; p=0.54], suggesting
that the facilitation at the synaptic sites remained largely unchanged.

When applied over sufficient time, our paired stimulation triggered LTP in the CA1, as evident by
a gradual increase in the fEPSP amplitudes [**Fig.1C, top**], but no change in PPR [**Fig.1C,**
bottom]. Therefore, we compared the magnitude of LTP between the groups by calculating a
ratio of EPSP amplitudes during the first 10 min with amplitudes at 40 ± 10 min [**Fig. 1D-F**]. The
control condition showed the strongest LTP ratio of 1.53 ± 0.13. LTP induction was almost
abolished during AIR [**Fig. 1E**, 1.06 ± 0.06 ratio; -30.66 ± 9.42% reduction vs Control, p=0.018]
and 0.1 mM D-βHb did not reverse this effect [**Fig. 1E**, 1.08 ± 0.08, +1.25 ± 6.61% AIR vs. D-
βHb+AIR 0.1 mM, p > 0.99], 1 mM D-βHb, however, did [**Fig. 1E**, 1.39 ± 0.10, +21.19 ± 7.64%
AIR vs. D-βHb+AIR 1 mM, p=0.044].

Moreover, we found no significant effects of either AIR or D-βHb+AIR on the amplitudes of the
fiber volleys (FV) [**Fig. 1F**, Control: 1.03 ± 0.08 mV; AIR: 0.89 ± 0.07; D-βHb+AIR 0.1 mM: 0.85 ±
0.08; D-βHb+AIR 1 mM: 1.03 ± 0.12; p=0.40]. Next, we tested whether the circuit would behave
differently under a stronger stimulation that imposed greater metabolic demands. We stimulated
the Schaffer collaterals with 20 pulses applied at 25 Hz, with 20 s breaks between trials. To avoid
the onset of LTP from skewing the results, we recorded all train stimulations at least 40 min after
the paired stimulation paradigm.

Figure 1: *Synaptic activity and LTP, but not neuronal firing or PPR, decrease during paired stimulation under AIR, and do not recover during D-βHb administration.*

A) Scatter plot of the fEPSP amplitudes evoked by paired-pulse stimulation of Schaffer collaterals. Each pair of circles is the average of 30-40 consecutive responses. Black bars represent mean ± SEM. Control fEPSPs in gray, AIR in red; 0.1 mM D-βHb+AIR in blue; 1 mM D-βHb+AIR in purple. Control: n=26 slices; AIR n=19; 0.1 mM D-βHb+AIR: n=20; 1 mM D-βHb+AIR: n=16. B) Scatter plot of the paired-pulse ratio (PPR) of fEPSP recorded in A. The labels, n, are identical to A). No significant differences (ρ=0.41, ANOVA). C) On top: Representative recording of a Control experiment where paired stimulation was applied for min. Each circle represents fEPSPs amplitude normalized to the average of the first 30 fEPSPs. The dashed orange line represents the mean amplitude at time -10 to 0 min. The black line represents the

*146   mean amplitude at +30 to +40 min. At the bottom: corresponding normalized PPR. **D)** fEPSPs waveforms*
147   recorded in C), showing LTP development. In gray, 30 fEPSP waveforms at t=0 ± 5 min and their average
148   (black). In yellow, 30 fEPSP at time +30 ± 5min and their average (orange). Stimulation artifacts and FVs
*149   are removed. **E)** Scatter plot of LTPs triggered by paired stimulation. Control: n=13; AIR n=11; 0.1 mM D-*
*150   βHb+AIR: n=12; 1 mM D-βHb+AIR: n=14. **F)** Scatter plot of the FV amplitudes recorded during A). The*
151   labels, n is identical to A). No significant differences (p=0.42, p=0.40; ANOVA). In all plots *p<0.05;
152   **p<0.01; ***p<0.001.

All compared groups showed potentiation of fEPSP amplitudes during the train, reaching peak at
stimuli 5-6. Thereafter, the amplitudes experienced a slight decline but remained potentiated [**Fig.**
S1A,B]. Interestingly, throughout the entire stimulation period, the fEPSP amplitudes matched the
differences we had seen during paired stimulation [**Fig. S1B**; at stimulus 5: -34.25 ± 11.80%
Control vs. AIR, p=0.031; -40.54 ± 11.48% Control vs. 1 mM D-βHb+AIR, p=0.000044; +3.37 ±
7.80 AIR vs. 0.1 mM D-βHb+AIR, p = 0.99; -25.21 ± 7.36% AIR vs. 1 mM D-βHb+AIR, p=0.0054].
The FV amplitudes followed a different pattern, with potentiation at stimulus 2, followed by a
consistent decline to approximately 70% of the initial amplitude [**Fig. S1C**]. Similar to paired
stimulation, no statistically significant differences were observed between groups at any time point
during the train [**Fig. S1C**; p=0.16 to 0.40].

Finally, we examined the hippocampal circuit under conditions of intense non-physiological
stimulation. This involved applying long trains of 50 stimuli at 25 Hz, spanning across 60 trials
with a 20-second break between each trial. The responses during this specific stimulation varied
between trials, making it inappropriate to average them over the entire stimulation period. As a
result, we represent the fEPSP and FV amplitudes as heat maps. In these heat maps, each point
represents the group's mean response to a single stimulus at a given time point.

Consistent with our previous results, we observed a period of fEPSP amplitude potentiation
followed by a steady decline [**Fig. S2A-D**; at maximum, trial 4-6, stimuli 7-9: -34.81 ± 10.56%
Control vs. AIR, p=0.0084; -48.27 ± 11.37% Control vs. 1 mM D-βHb+AIR, p=0.00065; +10.19 ±
11.39% AIR vs. 0.1mM D-βHb+AIR, p=0.59; -13.66 ± 9.08% AIR vs. 1mM D-βHb+AIR]. Synaptic
depression first emerged during the final stimuli of the trains, roughly 3 min into the stimulation
and became more prominent with each subsequent train application [**Fig. S2A-D**]. In general,
fEPSP amplitudes of Controls and AIR groups differed substantially through the first 7 min of the
recordings [**Fig. S2A**]. Control and 1 mM D-βHb+AIR groups also differed throughout most of the
stimulation trains [**Fig. S2B**], but the 0.1 mM D-βHb+AIR and AIR alone groups scarcely differed
except at the end of the trains [**Fig. S2C**]. The AIR and 0.1 mM D-βHb group fEPSPs generally
remained comparable. FV remained statistically similar between all the groups, with almost the
same response pattern [**Fig. S2E-G**], replicating the results from Fig. 1 and Fig. S1.

Taken together, these results strongly suggest that the effects of AIR and/or D-βHb could differ
between cellular compartments (i.e., dendritic versus somatic), have different specificity for
various cellular compartments, with synapses being particularly susceptible to metabolic
challenges induced by GLUT4 inhibition.

Figure 2: C*onduction velocity (CV) of Schaffer collaterals decreases under AIR are rescued by 0.1 mM D-βHb, and increase with 1mM D-βHb concentration.*

A) Representative hippocampal slice during CV recording. B) Representative averaged FV waveforms, recorded in CA1. The stimulus onset time (t=0) is marked with a dashed line. Stimulus artifacts are removed. C) Conduction velocity scatter plots. Each colored circle represents a CV replicate. Control in gray, AIR in light red, 0.1 mM D-βHb+AIR in blue, and 1 mM D-βHb+AIR in purple. Data are shown as means ± SEM. *p<0.05; ***p<0.001, nested ANOVA, Control: m=62, n=24, N=14; AIR: m=65, n=22, N=13; 0.1 mM D-βHb+AIR: m=64, n=21, N=13; 1 mM D-βHb+AIR: m=50, n=14, N=10. D) Diagram of conduction velocity of

*an action potential propagating along an axon: distance along the axon (Δx) and the time required for an AP to pass that distance (t_2-t_1). V represents membrane potential. **E)** Model of a propagating action potential. Change in V(t) is shown at a longitudinal increment along the axon (x_2) and at neighboring increments lying immediately ahead (x_1) and behind (x_3). Membrane potential at rest (V_R) changes over time proportionally to the sum of differences between its value and those of the preceding and superseding increments. Upon reaching the threshold potential (V_T), a spike occurs, and the membrane potential reaches peak value (V_A). **F-G)** CV is modulated by V_A and V_R. Our computational model predicts declining CV due to reductions in the peak of action potentials (F) and/or hyperpolarization of the resting membrane potential (G). CV quantified as percentage, relative to Control.*

3.2 AIR impairs axonal conduction speeds, but D-βHb treatment restores and enhances axonal conduction in a dose-dependent manner.

In the previous sections, we highlighted how AIR and D-βHb affected fEPSPs and FVs, finding a surprising lack of adverse effects of AIR on FV amplitudes, along with a noticeable reduction in fEPSP amplitudes; all further exacerbated by D-βHB. This difference suggests that AIR and/or D-βHb treatment might have differential effects on cellular compartments. We decided to investigate whether any other axonal properties changed during AIR and whether D-βHB remedied such changes.

We found that inducing AIR resulted in a strong reduction in the conduction velocities (CV) of SCOs [**Fig. 2C**; -13.07 ± 2.87% Control vs. AIR, p=0.027]. Addition of 0.1 mM D-βHb during AIR recovered CV back to Control levels [**Fig. 2C**; +12.43 ± 4.42% AIR vs. 0.1 mM D-βHb+AIR, p=0.034; +0.01 ± 0.07% Control vs. 0.1mM D-βHb+AIR, p=0.99] and 1 mM D-βHb significantly improved CV, surpassing the Control [**Fig. 2C**; +31.79 ± 4.59% for AIR vs. 1 mM D-βHb+AIR, p=4.2E-8; +19.10 ± 4.53% for Control vs. 1 mM D-βHb+AIR, p=0.00057].

To enable a better interpretation of these findings, we constructed a computational model of CV by employing an approximation of axonal cable theory (see Methods). Our model examined three processes that determine CV: resting membrane potential (V_R), peak membrane potential (V_A, the amplitude of AP overshoot), and activation threshold potential (V_T, potential for Na_v activation). We considered V_T to be robust to various physiological conditions, given that it primarily depends on Na_v channel type and gating kinetics [39]; therefore, we fixed V_T at a constant value of -50 mV in our calculations. We varied V_R between -90 and -60 mV and V_A between 10 and 30 mV, both of which are considered physiological ranges. We normalized the results to a reference state where V_R = -75 mV, V_A = 30 mV, and V_T = -50 mV. Modeling showed that a decrease in CV required either hyperpolarized resting membrane potential or decreased peak AP amplitude [**Fig. 2F-G**].

3.3 AIR adversely affects input timing, restored by high D-βHb concentrations.

Figure 3: *Time delays in axonal firing during train stimulation are increased during AIR and are reversed by 1 mM DβHb.*

A) Top: Representative example of the FVs recorded at the first stimulus in the control group during stimulation with 20 pulses applied at 25 Hz every 20 s, over 20 min. Bottom: FVs recorded in response to the last stimulus in each train. The FVs recorded at t = 0 are marked in black; the FVs recorded at t = +20 min, are marked in dark gray. B) - D) The same as A) in the AIR, 0.1 mM D-βHb+AIR and 1 mM D-βHb+AIR groups. E) Heat map of the mean, normalized FV peak latencies recorded in the Control group. Each trial is represented by a new row. The columns represent the time points of successive stimuli. Normalized latencies (0 ms) are color-coded in yellow. Delays (>0 ms) are color-coded in green-violet, and latency improvements (<0 ms) are coded in orange-brown. n=24, N=22. F) - H) The same as E) for AIR, n=16, N=15; 0.1 mM D-βHb+AIR. n=17, N=15, 1 mM D-βHb+AIR. n=14, N=10. I) Statistical comparison of latencies for the Control (E) and AIR (F) groups, performed on means of 3 stimuli x 2 trials blocks. p<0.05 in green, p<0.1 in yellow, p≥0.1 in white. J) The same as I) comparing Control (E) and 1 mM D-βHb+AIR (H) groups. No significant differences, p=0.067 to p=0.99. K) The same as I) comparing AIR (F) and 0.1 mM D-βHb+AIR (G) groups. No significant differences, p=0.41 to p=0.99. L) The same as I) comparing AIR (F) and 1 mM D-βHb+AIR (H) groups.

Changes in the CV of the observed magnitude and resulting differences in input timing would likely desynchronize the hippocampal circuit. Therefore, we investigated whether latency of FVs changes during stimulation with 20 pulse trains.

In all treatment groups, the early FVs showed ~-0.2 ms improvement in latency. During the first 3 min, all experimental groups were comparable and exhibited the same pattern of latency changes. Afterwards, AIR and 0.1 mM D-βHb+AIR groups displayed progressively more prominent delays than Control, while Controls and 1 mM D-βHb+AIR groups remained comparable. During AIR, we observed the largest delays [**Fig. 3E-F, Fig. 3I, L**; at 15 min, stim. 9-11: +2.69 ± 0.84 ms, Control vs. AIR, p=0.0077; 2.64 ± 0.94 ms, AIR vs.1 mM D-βHb+AIR, p=0.020]. The escalating nature of the delays during the stimulation suggests that the detrimental effects of AIR are further exacerbated over longer periods of time.

3.4 AIR increases membrane resistance (R_m) without affecting other intrinsic membrane properties of CA1 pyramidal neurons. D-βHb treatment under AIR does not rescue R_m to normal levels.

Results presented above [**Fig. 2**] indicate that AIR might negatively impact the membrane properties of hippocampal neurons, therefore we sought to identify the properties most vulnerable to AIR and test DβHb as a recovery agent. Based on dose-response results from the field potential studies, we selected 1 mM D-βHb for use in patch clamp experiments, testing the effects of AIR and D-βHb on intrinsic membrane properties of the CA1 pyramidal neurons (membrane resistance, R_m; membrane capacitance, C_m, and resting membrane potential, V_{rest}), and their spike thresholds. We found significantly increased R_m of the CA1 pyramidal neurons in the AIR group compared with Controls [**Fig. 4B**, +43.24 ± 3.73%, p=1.41E-07], which remained increased in the presence of 1 mM D-βHb [**Fig. 4B**, +27.86 ± 8.93%, p=0.030], while AIR and 1 mM D-βHb+AIR groups were similar (p=0.31). Interestingly, neither C_m [**Fig. 4C**, 114.25 ± 9.73, 115.52 ± 9.26,

132.20 ± 9.46, Control, AIR, D-βHb+AIR, respectively, p=0.32], spike threshold [**Fig. 4E**, -46.31
± 0.76, -44.53 ± 0.80, 44.64 ± 0.73, p=0.17] nor $V_{rest.}$ [**Fig. 4F**, -67.26 ± 1.17, -63.57 ± 1.38, -65.11
± 0.99, p=0.093] were significantly affected by either AIR or D-βHb+AIR treatments.

Figure 4: *Membrane resistance (R_m) increases during AIR and is not reversed by D-βHb.*
Other intrinsic properties did not change under either condition.

A) On the left: A CA1 pyramidal neuron recorded in whole-cell patch clamp mode (outlined in red)
with resistances and capacitance within the cell. On the right: A simple electric circuit diagram of
the cell with the same resistances. **B)** Scatter plots of CA1 pyramidal neuron R_m, compared
between the experimental groups. Each colored circle represents a value recorded in a single
cell. Control in gray, AIR in light red, 1 mM D-βHb+AIR in purple. Vertical black bars represent
mean ± SEM. **C)** Scatter plots of CA1 pyramidal neurons C_m, compared between the experimental
groups. The plots and coloring are identical to B). No significant differences (p=0.32). **D)**
Representative example of a 300pA ramp current injection performed in a Control CA1 pyramidal
neuron. The dashed line marks the onset (spike threshold) of the first AP. The inset shows the
magnified AP with the arrow pointing to the onset. **E)** Scatter plots of CA1 pyramidal neuron spike
thresholds, compared between the experimental groups. The plots are identical to B). No
significant differences (p=0.17). **F)** Scatter plots of CA1 pyramidal neurons RMP (V_{rest}), compared
between the experimental groups. The plots and coloring are identical to B). No significant
differences (p=0.09). Control n=28, N=22; AIR n=25, N=17; 1 mM D-βHb+AIR; n=30, N=16. In
black, *p<0.05; **p<0.01; ***p<0.001; in gray *p<0.1.3.3 AIR adversely affects input timing,
restored by high D-βHb concentrations.

3.5 AIR increases the frequency of spontaneous vesicular release at CA1 synapses (sEPSCs)
and mildly increases sEPSC amplitudes. D-βHb does not reverse the increase in frequency but

mildly decreases sEPSC amplitudes.

Recovering the membrane back to resting potential during synaptic transmission is the most
energy-consuming neuronal process (~% of total ATP produced [40]), therefore, we tested for
changes in the frequency and magnitude of quantal EPSCs that might explain the ~35% decrease
in fEPSP amplitudes reported above (**Fig.1-3**).

AIR significantly increased the frequency of sEPSCs, nearly doubling the number of events
compared with the Control [**Fig. 5A,B**; 0.53 vs. 0.96 Hz, +0.42 ± 0.12 Hz, p=0.0031]. Interestingly,
D-βHb failed to rectify the increased qEPSC frequency [**Fig. 5A,B**; D-βHb+AIR 0.93 vs. AIR 0.96
Hz, 0.021 ± 0.15 Hz, p=0.99; D-βHb+AIR vs. Control, +0.40 ± 0.12 Hz, p=0.0091]. However, the
amplitude, rise-time, and sEPSCs decay time showed no significant differences between the
groups [**Fig. 5C-E**].

Figure 5: *AIR increased sEPSC frequency without altering sEPSC properties and D-βHb (1*
mM) does not reverse this effect.

A) *Representative examples of the first 2 min of sEPSC recorded in CA1 pyramidal neurons, voltage*
clamped at V_h= -70 mV (Control in gray; AIR in red; 1.0 mM D-βHb+AIR in purple). Below: averaged sEPSC
waveforms from individual cells (light color) and the group mean (full color). ***B)*** *Scatter plots of the sEPSCs*
frequency recorded in CA1 pyramidal neurons, compared between the experimental groups. Each colored
circle represents frequency recorded in a single cell. Labels as in A) Vertical black bars represent mean ±
*SEM. *p<0.05.* ***C)*** *Scatter plots of the sEPSC 20-80% rise times, as in B). No significant differences*
(p=0.51). ***D)*** *Scatter plots of the sEPSC decay tau times, as in B). No significant differences (p=0.98).* ***E)***
Scatter plots of the sEPSC amplitudes, as in B). No significant differences (p=0.21). Control n=19, N=19;
*AIR n=19, N=14; 1 mM D-βHb+AIR, n=23, N=16. **p<0.01.*

Given the change in R_m, the lack of significant changes in the amplitudes is surprising. Several
studies utilizing different metabolic challenges, such as food deprivation [41] and inhibition of

glycolysis [42], showed an increase in mEPSC/sEPSC amplitudes, therefore, we decided to
investigate our sEPSCs in more detail. In most central synapses, the distribution of
mEPSC/sEPSC amplitudes shows several peaks related to simultaneous release of multiple
quanta of neurotransmitter [43], with each subsequent peak linked to an increasing number of
concurrent sEPSCs. Capitalizing on the multiplicative nature of this process, we assumed that
small changes in amplitudes [**Fig. 5E**] would scale up. We detected 2-4 distinct peaks in ~85% of
the analyzed cells [**Fig. S3A-C**], with no differences between the groups in the number of cells
with multi-peak distributions [**Fig. S3C**; 89.5% in Control; 84.2% in AIR, 87.5% in D-βHb+AIR]. In
line with our initial assumptions, the scaled amplitudes revealed differences between the groups
[**Fig. S3E**]: AIR group had significantly higher amplitudes than D-βHb+AIR [p1: p=0.048; p2:
p=0.0054; p3: p=0.0038; p4: p=0.021] and also tested as different from Controls at peak 3
[p=0.007]. Control and D-βHb+AIR groups did not show significant differences at any point [**Fig.
S3E]**, despite the same sEPSC frequency as AIR.

We further validated those findings by comparing the normalized amplitude distributions [**Fig.
S3F-I**]. Within the range correlating to peak 2 – peak 4 (sEPSC of ~20pA or more), the AIR group
had a significantly higher number of large events (15.19% of total) when compared to Control
(7.65%) and D-βHb+AIR (9.48%) [**Fig. S3F-I**], while Control and D-βHb+AIR groups showed little
consistent difference. This result suggests that during AIR, the elevated frequency of events
increased the probability of multi-vesicular events occurring, further exacerbated by the ~20%
increase in R_m. On the other hand, D-βHb+AIR produced a distribution resembling Control, as if
the vesicles had lower glutamate content, as suggested by literature [44,45].

3.6 *AIR slightly changes the firing pattern of hippocampal pyramidal neurons and causes*
increased membrane depolarization with input. Those effects are reversed by D-βHb, which also
increases AP amplitudes.

The action potential generation and subsequent recovery of V_m to resting levels entail a large part
of the energy budget of pyramidal neurons (~41% of all produced ATP [40]), therefore, we
investigated whether pyramidal neurons retain their input-output curves (I-O curve) under AIR
and how D-βHb alleviates AIR effects.

First, we focused on the firing frequency [**Fig. 6A**]. Cells under AIR showed increased firing at
lower current injections than D-βHb+AIR [**Fig. 6B**, 75–150 pA, p=0.0038 to 0.050], but with the
same maximal firing rate. These results suggest that the cells activate earlier under AIR than in
the Control or D-βHb+AIR conditions. To test this further, we measured the maximum
depolarization attained by each neuron during the final 5 ms of each current injection step [**Fig.
6D**]. AIR neurons had slightly higher depolarization with each injection step compared to Controls
[**Fig. 6E,F**; p=0.0027 to 0.048] or D-βHb+AIR [**Fig. 6E,F**; p=0.0029 to 0.042].

Figure 6: *Neuronal firing is mildly affected by AIR and reversed by DβHb, with DβHb increasing AP overshoot amplitudes.*

A) Representative examples of responses to 400 pA square current injections into Control (gray), AIR (red), and 1 mM D-βHb+AIR (purple) CA1 pyramidal neurons, held at V_h=-70 mV. B) I-O curves of CA1 pyramidal neurons receiving 21, Δ+25pA injections (max. 0 pA; V_h=-70 mV). Circles represent group mean ± SEM. * Control vs. AIR, + Control vs. 1 mM D-βHb+AIR, ^ AIR vs. 1 mM D-βHb+AIR. *,+,^ in black p<0.05; **,++,^^ in black p<0.01; *,+,^ in gray p<0.1; ANOVAC) Maximum firing rate of the pyramidal neurons in B). Circles represent single neurons; black bars represent group means ± SEM. No significant differences (p=0.778, ANOVA). D) Example of responses to Δ+pA square current injections into a control pyramidal neuron. The arrowhead and inset mark the final membrane depolarization (V_m, last 5 ms of the step). E) V_m at the current injections from B). F) Group means ± SEM of V_m from E). Statistical tests and labels are identical to B). Control n=23, N=20; AIR n=20, N=14; 1 mM D-βHb+AIR n=22, N=16. G) Example of AP overshoot measurement. Black line marks V_m= 0 mV. H) Averaged AP overshoot amplitudes from B). I) Group means ± SEM of overshoot amplitudes from H). Statistical tests and labels are identical to B). In all comparisons Control n=23, N=20; AIR n=20, N=14; 1 mM D-βHb+AIR, n=22, N=16.

Next, we investigated the averages of AP peak amplitudes [**Fig. 6G**], per step [**Fig. 6H**]. Surprisingly, D-βHb+AIR neurons showed the largest amplitudes, consistently elevated

throughout the injections, when compared to Controls or AIR, while AIR neurons remained
comparable to Controls [**Fig. 6I**, p=0.0017 to 0.036, AIR vs. Control; p=0.0093 to 0.044, AIR vs.
1 mM D-βHb+AIR].

3.7 *AIR increases AP decay time, while D-βHb accelerates the AP rise without reversing the AIR-*
induced slower AP decay.

Figure 7: *AP decay times change during AIR and DβHb + AIR but DβHb + AIR APs have*
faster rise times.

A) Example of AP decay time measurement. Black line marks V_m= 0mV. B) Mean AP decay times for the
current injections. Circles represent averaged amplitudes in single neurons. Black lines represent group
means. C) Group means ± SEM of AP decay times from B). Statistical tests and labels are identical to B).
D) Example of AP decay time measurement. Black line marks V_m= 0mV. E) Averaged AP rise times for the
current injections in B). Circles represent averaged amplitudes in single neurons. Black lines represent
group means. F) Group means ± SEM of AP rise times from B). Statistical tests and labels are identical to

We next proceeded to investigate other properties of the APs: decay time (from the overshoot peak to 0 mV) [**Fig. 7A-C**] and rise time (from 0 mV to the overshoot peak) [**Fig. 7D-F**]. We found that Control neurons had significantly lower average AP decay times when compared to AIR [**Fig. 7B,C**; p=0.047 to 0.0019] and D-βHb+AIR conditions [**Fig. 7B,C**; p=0.041 to 0.0023], across almost all current injection steps. AIR and D-βHb+AIR cells did not significantly differ at any point [**Fig. 7B,C**; p=0.32 to 0.99]. D-βHb+AIR cells demonstrated faster rise times compared to Control [**Fig. 7E,F**; p=0.000094 to 0.0025] and AIR cells [**Fig. 6E,F**; p=0.047 to 0.0052] in the majority of the injection steps. Interestingly, AIR cells exhibited slightly faster AP rise times exclusively at the lowest current injections [**Fig. 4 N,O**; p=0.012 to 0.035]. These slightly faster APs were likely due to earlier activation of the cells, as shown in [**Fig. 6B-C, E-F**]. **Fig. 7Fig. 6**

Finally, as we did not observe significant differences in firing frequency among the groups [**Fig. 6B,C**], we examined variations in the timing of AP firing based on the intervals between consecutive APs (AP Inter Event Interval, AP IEI) [**Fig. 7G**], but we found no clear changes in the AP IEI [**Fig. 7H,I**].

3.8 *AP adaptation is impaired during AIR, with AP amplitudes showing steeper declines, and is not reversed by D-βHb.*

The amplitudes of the AP overshoot, decays, and rise times demonstrated steeper slopes in the AIR and D-βHb+AIR groups. The properties of APs during neuronal firing undergo dynamic changes, commonly referred to as spike adaptation or spike frequency adaptation [46] and in our experiments, we see evidence that GLUT4 antagonism intensified these changes.

To test this observation, we compared peak amplitudes of the first APs [**Fig. S4A-D**], the averaged APs [**Fig. 6G-I**, **Fig. S4E**] and the final APs during current injections [**Fig. S4F**]. The first APs represents neuronal firing after rest (10s between subsequent injections), while the average and final AP – after energy expenditure. As expected, the first APs displayed minimal changes in amplitude during injections [**Fig. S4D**], with D-βHb+AIR exhibiting consistently larger amplitudes than the Control. AIR, also showing elevated amplitudes, did not differ significantly from either the Control or D-βHb+AIR groups. For the final APs, we observed a sharp decline in amplitude scaling with injected current [**Fig. S4F**] in both AIR and D-βHb+AIR groups. Control showed lower amplitudes than D-βHb+AIR at 100-300pA injections (p=0.040 to p=6E-05) and remained comparable to AIR up until the very last, large, 400-0pA injections (p=0.032 to p=0.0011), when AIR APs significantly declined [**Fig. S4F**]. AIR and D-βHb+AIR remained consistently different.

Linear fits to the average AP amplitudes revealed significantly steeper declines [**Fig. S4G,H**;
p=0.00039, AIR vs Control; p=0.0012, D-βHb+AIR vs Control], indicating that prolonged neuronal
activity is heavily affected by AIR.

3.9 *Changes in AP decay and rise times translate into differences in the widths of FVs.*

In our initial analysis of the field potential (FV) data, our focus was solely on the amplitudes of the
biological signals. Consistently, we found no significant changes in the FV amplitudes [**Fig. 1F,**
2C, 3E-G], which supports our findings on AP overshoot amplitudes [**Fig. 6G-I**] (with the exception
of D-βHb+AIR group). Based on these results, we further investigated whether the changes in AP
decay and/or rise times also corresponded to changes in FV widths during train stimulation [**Fig.**
S5].

Consistent with our findings on AP decay/rise times [**Fig. 7A-F**], AIR consistently exhibited the
widest FVs throughout the stimulation followed by the Control group, with 1mM D-βHb+AIR FVs
as the narrowest [**Fig. S5D**]. We found differences between AIR and Control (p=3.94E-06 to
0.049), and AIR and 1mM D-βHb+AIR (p=4.04E-05 to 0.046) throughout the whole stimulation
period, while Control and 1mM D-βHb+AIR were not different (p=0.079 to 0.99), supporting our
AP decay/rise time findings [**Fig. 6G-I**].

3.10 *With abolished GABAergic inhibition, AIR strongly decreased firing rates of CA1 pyramidal*
neurons and fast-spiking interneurons (FSI), which were not rescued by DβHb.

As the next step, we investigated the properties of CA1 pyramidal neurons and FSIs after
pharmacologically abolishing AMPA/kainate excitatory (NBQX) and fast GABAergic inhibitory
(gabazine) transmission [**Fig. S6**].

We first recorded the firing frequency of CA1 pyramidal neurons [**Fig. S6A,B**] at different
activation levels [**Fig. S6C,D**], finding that AIR prominently decreased their firing rate, when
compared with Control [**Fig. S6C**, p=0.024 to 0.0053]. Surprisingly, the application of 1 mM D-
βHb during AIR resulted in a stronger reduction in firing rate, when compared with Control [**Fig.**
S6E, p=0.000027 to 0.035]. AIR and D-βHb+AIR results remained similar throughout the
injections [**Fig. S6E**]. However, all groups differed significantly with respect to the maximum firing
rate, with Control having the highest (31.03 ± 3.43) and D-βHb+AIR having the lowest (13.17 ±
2.05) firing rates.

Next, we examined CA1 FSIs using the same drug conditions [**Fig. S6E,F**]: AIR decreased the
firing rate of FSIs compared with Control [**Fig. S6G**, p=0.012 to 0.048], not reversed by D-βHb
[**Fig. 5S6H**, p=0.00017 to 0.035]. As for pyramidal neurons, the FSIs for AIR and D-βHb+AIR

groups remained mostly comparable [**Fig. S6G**]. None of the groups differed in the maximum
firing rate [**Fig. S6H**, p=0.068].

3.11 *Hodgkin-Huxley model predicts that impairments in Na$^+$/K$^+$ ATPase activity result in more*
depolarized V$_{rest}$, lower AP overshoot (V$_a$), and increased neuronal firing.

Here we aim to explain the previous observations of dysregulated firing dynamics in response to
AIR. We hypothesize that Indinavir restricts the energy supply of neurons and thus impairs the
neuronal Na$^+$/K$^+$ ATPase. To test this hypothesis, we developed a computational model of an
isolated CA1 neuron in the absence of inhibition and toggled Na$^+$/K$^+$ ATPase activity through the
enzyme kinetic parameter v_{ATPase}^{max} (see supplementary methods, **Fig. S7**). **Figs. S7A,B** (at two
different temporal scales) shows the predicted changes in neuronal firing dynamics when the
Na$^+$/K$^+$ ATPase is impaired. Such an impairment predicts the depolarization of the resting
membrane potential (**Fig. S7C**) and is consistent with trends of our I/O curves (**Fig. 4F**).
Furthermore, this modeled depolarization reduces the magnitude of membrane potential shift
required for an action potential. Consequently, the model predicts increased firing frequency (**Fig.**
S7D). Finally, the model predicts that the Na$^+$/K$^+$ ATPase should act more slowly to recover ion
gradients after an AP, causing the amplitude of subsequent APs to decline progressively with time
(**Figs. S7E,F**). Such an amplitude decrease is consistent with the trends in measured conduction
velocity (**Fig. 2**).

4. **Discussion**

Central insulin resistance impairs cognitive performance and promotes hippocampal
neurodegeneration and memory loss [47,48]. Our present results highlight the adverse effects of
GLUT4 inhibition on synaptic function and LTP, in accord with findings that acute and chronic
blockade of GLUT4 are detrimental to performance on hippocampal-mediated memory tasks, thus
suggesting that GLUT4 is critical for acquisition and consolidation of memory [8,13]. Memory
formation is promoting GLUT4 incorporation into the membrane, resulting in increased glucose
flux into neurons [13,49]. This process enables increased metabolic support during high demand.
Crucially, most GLUT4s localize in the perikaryon [50,51], likely in close proximity to axo-somatic
synapses, which are associated with synaptic plasticity and long-term memory (LTM). Indeed,
prolonged blockade of the GLUT4 glucose transporter in the brain results in impaired formation
of LTM [13] and decreased levels of hippocampal BDNF [52]. Moreover, hypoglycemia negatively
affects memory performance and cognition [53,54, however, see 13]

In this study, we found that AIR increased the frequency of sEPSC events, in line with the
previously published data showing similar increases during food deprivation [42], inhibition of
glycolysis [41] and ischemia [55], likely due to hypoglycemia-driven Ca^{2+} accumulation at the
presynapse [41,55]. The same studies suggest that presynaptic glycolysis is essential for

maintaining synaptic transmission, even at a low frequency, and that mitochondrial respiration
cannot substitute it. Normally, glycolysis meets approximately one-third of the energy budget at
presynaptic terminals, thus sustaining low-frequency transmission. In addition, impaired
glycolysis leads to slower, broader, and smaller AP waveforms, and a depolarization of the resting
membrane potential [42]; results which are consistent with ours. Interestingly, D-βHb+AIR
treatment, which showed the same increase in the frequency of sEPSCs as AIR, had the lowest
sEPSC amplitudes [**Fig. S3E**]. This result suggests a reduction in the quantal content of synaptic
vesicles and is consistent with reports of lowered conversion of glutamate to aspartate under
ketosis [44,45].

Although the results of our computational models are consistent with the observed changes in
resting state membrane potential and firing frequency during AIR, there were seemingly
contradicting results with respect to AP peak amplitude. Specifically, we measured no significant
changes to average AP overshoot amplitudes during AIR, our model predicted a decrease due to
energy constraints. In the model we did not consider a progressive decline in available energy,
which is suggested by the steeper AP adaptation [**Fig. S4**]. Moreover, neurons may compensate
for the energetic limitations by increasing R_m (as reported in 42), while our model assumes that
R_m remains unchanged. When those factors are accounted for, the model and experimental data
agree: during periods of extensive activity, energy reserves in neurons are progressively depleting
and, provided that the neuron fired multiple APs in short succession, the negative effects of
hypoglycemia overcome compensatory R_m increases and AP amplitudes decline below Control
levels [**Fig. S4**]. Furthermore, we used the maximal Na^+/K^+ ATPase activity as a surrogate for
ATP availability, and we relied on literature estimates of kinetic and thermodynamic rate
constants. A more direct approach might consider ATP concentration as an input variable, were
such experimental data available.

Our model aiming to establish the underlying causes of the changes in conduction velocity
investigated resting membrane potential (V_{rest}) and AP peak amplitude as two critical factors. The
model explains CV recovery and increases during D-βHb+AIR treatment, since the measured AP
peak amplitudes remain consistently elevated and V_{rest} changes are negligible, but fails for AIR
alone, where AP amplitudes remain comparable to Control and V_{rest} tends to depolarize. This
discrepancy might reflect the simplistic nature of our model, which neglects complex ion
dynamics, altered membrane permeability or synaptic processes.

The ketogenic diet has served for decades to treat epilepsy and has an emerging wide range of
therapeutic applications [30]. Possible mechanisms include modulation of ATP-sensitive K^+
channels (K_{ATP}), free fatty acid receptor 3 (FFAR3/ GPR41) activation, promotion of GABA
synthesis, and epigenetic modifications [56]. The action of D-βHb in modulating potassium flux
through K_{ATP} channels may be most relevant in the present context [57], since K_{ATP} activation
results in decreased neuronal firing rates [58,59] and prevented picrotoxin-induced epileptiform
activity in the hippocampus [60] and may explain present findings that firing rates and membrane
depolarization remained at control levels in the presence of 1 mM D-βHb during AIR (despite
elevated R_m).

D-βHb is a precursor of glutamine, which promotes increased GABA levels, as seen in studies of
the ketogenic diet in clinical epilepsy [61]. D-βHb also directs glutamate towards GABA production
by reducing the conversion of glutamate to aspartate [44,45]. In addition, D-βHb has a direct effect
on the vesicular glutamate transporter VGLUT2, by inhibiting Cl^- dependent glutamate uptake.
VGLUT2 is critical for glutamate excitatory transmission in CA3-CA1 and its conditional VGLUT2
knock-out results in impaired spatial learning and reduced LTP [62].   Together with the increased
formation of GABA and/or GABA derivatives, VGLUT2 inhibition could further reduce glutamate
excitatory transmission [63].

The present results have important implications for the treatment of conditions such as type 2
diabetes, metabolic syndrome, and seizure disorders that involve insulin hyposensitivity.
According to NIH estimates, ~30% of Americans today display symptoms of prediabetes, a
percentage expected to increase in the nearby future. Together with the longer average lifespans
and demographic changes leading to an expansion in the aging population, obtaining insights into
the outcomes of metabolic dysfunction and its potential treatment becomes a more pressing
matter each day. Our results provide a unique insight into the potential use of KBs as an
inexpensive and relatively risk-free treatment for metabolic disorders and insulin resistance.
Medical researchers can use this information to discover new targets for treatment and develop
therapies that are more effective. Further research in this area could inform novel treatment
approaches that address the complex effects of metabolic disorders on brain health.

5. **Materials and Methods**

5. 0 *Mice*

Breeding pairs of transgenic GAD2Cre/GCamp5gTdTomato, Thy1/GCamp6fTdTomato, and
C57BL/6J mice were originally obtained from The Jackson Laboratory (stocks 010802, 024477,
024339) or Charles River Laboratories and bred in-house under standard conditions of 12-12
hours light-dark cycle, with food and water available *ad libitum*. The Institutional Animal Care and
Use Committee approved all experiments. Mice of both sexes aged between 30 and 60 days
(P30-P60) were used in all experiments.

5.1 *Electrophysiology – slice preparation.*

Coronal brain slices containing the hippocampus were used in all recordings. Mice were deeply
anesthetized with a mixture of isoflurane and air (3% v/v), decapitated, and their brains extracted
and cut using a Leica VT1200S vibratome into 300 μm-thick slices in ice-cold ACSF solution
containing (in mM): 230 sucrose, 1 KCl, 0.5 $CaCl_2$, 10 $MgSO_4$, $NaHCO_3$, 1.25 NaH_2PO_4, 0.04 Na-
ascorbate, and 10 glucose; 310 ± 5 mOsm, pH adjusted to 7.35 ± 5 with HCl, and, gassed with

carbogen (95% O_2, 5% CO_2) for at least 30 min before use. The slices were then transferred to a
Zbicz-type BSC-PC submerged slice chamber (Harvard Apparatus, USA) filled with artificial
cerebrospinal fluid (ACSF) containing (in mM): 124 NaCl, 2.5 KCl, 2 $CaCl_2$, 2 $MgSO_4$, 26 $NaHCO_3$,
1.25 NaH_2PO_4, 0.04 Na-ascorbate, and 10 glucose; 305 ± 5 mOsm/kg, pH adjusted to 7.35 ± 0.05
mOsm/kg, gassed with carbogen and prewarmed to 32 $^{\circ}$C. The slices were allowed to recover in
the warm ACSF for 15 min, whereupon the heating was switched off, and the chamber gradually
cooled down to room temperature (RT; 23 ± 2 $^{\circ}$C) over 45 min.

Following the recovery, individual slices were transferred to another BSC-PC chamber filled with
gassed ACSF and a mixture of drugs (Indinavir 0.1 mM; Indinavir 0.1 mM and either 0.1- or 1-mM
DβHb), where they were incubated for 60 ± 15 min. Next, the slices were transferred to a
submerged recording chamber mounted on an upright microscope stage (E600FN, Nikon, Japan)
equipped with infrared differential interference contrast (IR-DIC) filters. The slices were kept at
RT and superfused continuously with gassed ACSF delivered by a peristaltic pump at a steady
rate of ~2 ml/min.

5.2 Field potential recordings.

Evoked field potential responses were triggered with an isolated pulse stimulator (Model 2100,
AM Systems, USA) using a CBARC75 concentric bipolar electrode (FHC, USA) placed between
CA3 and CA1. The recording electrode was placed 300 ± μm away for the recordings of CA1 FV
and fEPSP. During the recordings of axonal conduction velocity (CV), the electrodes were 300 to
900 ± 50 μm apart, and up to 6 different sites were recorded. All stimulation paradigms were
applied every 20 s as biphasic rectangular pulses of 240 μs total duration, 50 ± 5 V strength (~60
± 10% of the maximum response size), and a between-stimuli frequency of 25 Hz (40 ms).

During non-CV recordings, three different types of stimulation were applied: paired, 20 stimuli,
and stimuli trains, repeated every 20 s for a total period of 20 min in each case. After each train
stimulation, the slice was allowed to recover for 20 min under paired stimulation.

Recording pipettes were pulled from borosilicate glass capillaries (ID 0.86, OD 1.5, BF1-86-10,
Sutter Instruments, USA) on a vertical PC-100 puller (Narishige, Japan) and filled with 6% NaCl
(in H_2O, 1.02 M). Only pipettes with a resistance of 0.2-1.0 MΩ were used. Recording pipettes
were allowed to rest at the recording site for 10 min to equilibrate before the start of a recording.
All recorded signals were acquired using a MultiClamp 700B amplifier (Molecular Devices, USA),
Bessel-filtered at 2 kHz, and digitized at a sampling frequency of 10 kHz using an Axon Digidata
1440A digitizer (Molecular Devices, USA).

5.3 Whole-cell patch-clamp recordings.

Pyramidal neurons of the hippocampal CA1 region were selected for recordings based on their
general morphology and position within the pyramidal layer. As an additional verification, recorded
cells were screened for GAD2$^+$-driven tdTomato fluorescence. Cells that displayed any level of
tdTomato fluorescence were rejected.

Patch pipettes were pulled from borosilicate glass capillaries (BF1-86-10, Sutter Instruments,
USA) on a vertical puller (PC100, Narishige, Japan). Pipettes had a resistance of 6–8 MΩ when
filled with an internal solution containing (in mM): 136 K-gluconate, 4 Na$_2$ATP, 2 MgCl$_2$, 0.2 EGTA,
10 HEPES, 4 KCl, 7 di-triphospho-creatine, 0.3 Na$_3$GTP; 280–290 ± 5 mOsm/kg, titrated to pH
7.3 ± 0.05 with KOH.

During recordings in voltage clamp (VC), cells were clamped at the holding potential (V_h) = −70
mV with a Multiclamp 700B amplifier (Molecular Devices, USA). Liquid junction potential was not
corrected. All cells were compensated for pipette capacitance. During VC experiments, series
resistance (R_s) was not compensated. During recordings in current clamp (CC), a holding current
was applied to the cells to maintain membrane potential at V_h = -70 ± 3 mV. All cells were bridge-
balanced at % of R_s. Cell capacitance was not compensated in either mode. Immediately after
establishing the whole-cell configuration, the cell was kept in I=0 mode to monitor the resting
membrane potential (V_{rest}) over a period of ~120 to 180 s. The average value of V_{rest} over this
period is reported as the neuron's V_{rest}.

During the recordings, every 5 to 10 min a series of 10 square voltage steps of −10 mV was
apllied to monitor changes in the R_s (Vh = -70 mV). All whole-cell currents in response to voltage
steps were low-pass filtered at 10 kHz with a Bessel filter and sampled at 20 kHz frequency (Axon
Digidata 15B, Molecular Devices, USA). For inclusion in the final dataset, recordings had to meet
3 criteria: 1) R_s after any recorded protocol could not increase by more than 20% compared to the
R_s pre-protocol; 2) R_s during any of the recorded protocols could not exceed 30 MΩ; 3) the offset
drift by the end of the recordings could not be higher than ± 5 mV.

Spontaneous synaptic currents (sEPSC) were recorded in VC at V_h = -70 mV in a single,
continuous sweep for 10 to 15 min, low-pass filtered at 2 kHz with a Bessel filter and sampled at
10 kHz frequency.

To establish the spiking threshold, each neuron was current-clamped at V_h ~ -70 ± 3 mV and
subject to a ramp-shaped 200-0 pA current injection with a 1s duration. The spike threshold was
determined as the membrane potential at which the first fully developed action potential (AP) was
visible. Each ramp injection was repeated 10x, and the averaged threshold value is reported for
the neuron.

To establish the firing properties of recorded neurons, each neuron was current-clamped as
described previously and subjected to a square-shaped series of current injections with 25pA Δ
per step to a maximum of 0 pA, with each step of 1 s duration and inter-step interval (ISI) of 10 s.
If necessary, the holding current was corrected during the protocol to maintain V_h at ~ -70 ± 3 mV.

During analyses, only fully-formed APs that crossed the 0 mV threshold were included.

All data acquisition was performed using pCLAMP software (Molecular Devices, USA).

5.4 Field potential recordings – analyses.

During field potential recordings, the biologically relevant signals are superimposed onto the
decaying stimulation artifact. To isolate the stimulation artifacts for subtraction from the signal at
the end of each recording, we applied 1 µM TTX to the slice and recorded 30 repetitions of each
stimulation paradigm were recorded in the presence of TTX.

Recorded traces were analyzed in IgorPro (Wavemetrics, USA). First, an average of the traces
recorded with TTX was subtracted from each trace. The traces were then filtered using a rolling
five-point average to remove noise, and the baseline was adjusted to the 1 s period before the
stimuli. The amplitudes and latencies of FVs and fEPSPs were measured at their peaks.
Amplitudes were measured against 0 mV baselines, while latencies were measured against the
onset of the stimulation. During high-frequency stimulation, cells often do not have sufficient time
to fully recover their membrane potentials before the next stimulus is applied, leading to a gradual
shift in the baseline, therefore, as a correction, we subtracted the median voltage at +35 to +40
ms after each stimulus from each FV or fEPSP amplitude. FV widths were measured at the peaks
of positive deflections in the waveform, first after the artifact subtraction and second directly before
fEPSP.

All traces were screened for occasional electrical interference and the affected data points were
manually removed from the analysis.

To measure axonal conduction velocity (CV), we calculated the time shift (Δt, ms) between the
stimulus start and the time of the FV peak. Next, the linear distance between the recording site
and the stimulation electrode was divided by the measured Δt. We recorded CV at 3-6 sites over
a span of 300-900 µm, with ~1 µm steps between sites. Each CV recording site served as a
replicate for statistical analysis (see Statistics). During CV recording, stimulation remained
stationary, and the sites were recorded in randomized order.

To calculate changes in FV peak latency during train stimulations the median Δt at stimulus 1-3
of traces 1-2 (Δt_N) was subtracted from each Δt (normalization).

5.5 Field potential recordings – data visualization.

For constructing "heatmap" visualizations of the data, numerical matrices of FV or fEPSP
amplitude means were built for each of the four treatment groups. In every matrix each new row

represented responses during each subsequent train. Then, two additional columns were added
at the end as a color normalization in the heatmap (0, 1 mV), and the matrix was saved as a text
file. Afterwards, the text files were imported as text images to ImageJ (NIH, USA), converted from
32- to 16-bit, saved as TIFF images and imported to IgorPro (Wavemetrics, USA), whereupon
their look-up table (LUT) was changed from "Grays" to "Spectrum."

The heat maps for latency changes during train stimulation were generated similarly but using
different normalizations: 20 pulse train recordings had normalization boundaries of -1 and +1 ms.

5.6 Whole-cell patch-clamp – sEPSC analyses.

sEPSCs detection was automatized using a convolution-based algorithm in Fbrain 3.03 [64], a
customized program running under IgorPro 6 (WaveMetrics, Lake Oswego, USA), kindly provided
by the Peter Jonas Lab (IST, Klosterneuburg, Austria). Recorded traces were smoothed by
subtracting a 20-term polynomial fitted to the trace and digital Notch (60 ± 0.5 Hz) filtered in FBrain
before the analysis. The trace subject to convolution was passed through a digital band-pass filter
at 0.001 to 200 Hz. The event detection template had a rise-time time constant τ = 1.5 ms, a
decay τ = 6 ms, and an amplitude of −10 pA. The event detection threshold (θ) was set to 4.2
times the standard deviation of a Gaussian function fitted to the all-point histogram of the
convolved trace. All events detected by the algorithm were inspected visually. Those that clearly
did not show fast rise time and exponential decay kinetics of a typical hippocampal EPSC were
removed by hand. Only cells with a rejection ratio lower than 20% were included in the analysis.
The analysis was performed using custom-written macros provided by Dr. Maria Kukley
(Achucarro Basque Center for Neuroscience, Bilbao, Spain).

Afterwards, individual sEPSC waveforms were extracted from the trace. Each waveform
consisted of a 5 ms segment of the trace before the onset of the sEPSC and a 50 ms piece after
the onset to capture the full decay of the waveform. Before averaging, all sEPSCs were baseline-
adjusted to the pre-onset 5 ms. The peak amplitude, 20-80% onset-to-peak rise time, and decay
constant τ were measured for the averaged sEPSC. The decay τ was calculated from the
monoexponential fit from the sEPSC peak to the final ms of the averaged waveform.

To avoid selection bias during sEPSC extraction, we repeated the Fbrain analysis at least twice
in each cell and averaged the results.

5.7 Statistics.

All data acquisitions were made in randomized sequence, with a maximum of 2 slices (field
potential recordings) or 2 cells (patch clamp recordings) from a single animal per experimental
condition to avoid pseudoreplication. All experimental groups had comparable distributions of

ages (close to P50 ± 6) and sex. The order of slice pre-incubation with drugs was randomized for
each animal. The numbers of slices or cells and animals used in each experiment are indicated
in the figure legends.

Statistical analysis was performed using GraphPad Prism 9.5.1 (GraphPad Software, USA).
Significant outliers were removed with the Prism ROUT method at Q = 5%, and the normality of
residuals and homoscedasticity were tested for each comparison. If the datasets had normal
residuals and equal variances, we used ordinary one-way ANOVA with *post hoc* Holm-Šídák's
test. If the datasets had normal residuals but unequal variances, we used Brown-Forsythe
ANOVA with *post hoc* Dunnett's T3 test. If the datasets were not normally distributed but had
equal variances, we used Kruskal-Wallis ANOVA with *post hoc* Dunn's test. In rare cases where
the data was neither normally distributed nor had equal variance, we applied the Brown-Forsythe
ANOVA with *post hoc* Dunnett's T3 test. If the dataset consisted of multiple replicates, we used
nested one-way ANOVA with *post hoc* Holm-Šídák's test. The tested groups and p values are
indicated in the text or the figure legends. Individual cells or slices are labeled as n, replicates are
labeled as m, and the numbers animals used are labeled as N. If the groups tested as being
different, we report the p-values of *post hoc* tests but omit the p-value of the omnibus test. If the
groups did not test as different, only the p-value of an omnibus test is reported. For scatter plots,
each point represents an individual data point (cell, slice, or replicate) and the horizontal bars
represent the group mean ± SEM. In all other graphs, we present mean ± SEM. The heat maps
represent the group means.

5.8 Model of conduction velocity

Here we model how AP shape affects the conduction velocity (CV) along an unmyelinated axon.
We model an infinitesimally small unit of the axon consisting of three distinct units in a row,
separated by a space h. The first and last of these units are kept "fixed" at voltages $V_1 = V_a$ (the
peak membrane potential) and $V_3 = V_{rest}$, respectively; this potential difference leads to a voltage
flow across the axon which is generally described by the cable equation [65]:

$\lambda^2 \frac{\partial^2 V}{\partial x^2} = \tau \frac{\partial V}{\partial t} + V$

Once the intermediate unit V_2 is excited above a threshold voltage V_t, the unit activates and is
self-excited up to a voltage V_a. This process then repeats *ad infinitum* along the entirety of the
axon (assuming no backflow of the charge). The conduction velocity (i.e., the rate of this voltage
flow) is thus inversely proportional to the reset time t_{reset} (the first time when $V_2(t) = V_t$, with
$V_2(0) = V_{rest}$ as well). Over the discrete lattice above, the cable equation is given by:

$\lambda^2 \frac{V_3 - 2V_2 + V_1}{h^2} - \tau \frac{dV_2}{dt} = V_2$

Solving this for $V_2(t)$ and substituting in the boundary conditions, we find:

$$V_2(t) = \frac{\lambda^2(V_{rest} + V_a)}{h^2 + 2\lambda^2} + Ke^{\frac{-1}{\tau}\left(1+\frac{2\lambda^2}{h^2}\right)t} = V_B + Ke^{\frac{-1}{\tau}\left(1+\frac{2\lambda^2}{h^2}\right)t}$$

where K is an integration constant needed to satisfy the initial condition $V_2(0) = V_{rest}$. Substituting
in the initial condition and solving for t_{reset}, we find:

$$t_{reset} = \frac{-\tau}{1+\frac{2\lambda^2}{h^2}} \log \frac{V_t - V_B}{V_{rest} - V_B}$$

Thus:

$$CV \propto \frac{1}{\log \frac{V_{rest} - V_B}{V_t - V_B}}$$

However, h is an arbitrarily small spacing that we can take to approach zero. Thus:

$$CV \propto \frac{1}{\log \frac{V_{rest} - V_a}{2V_t - V_{rest} - V_a}}$$

5.9 *Power and sample size calculations.*

The sample sizes for all statistical comparisons in this work were determined based on the means
and pooled standard deviations from preliminary recordings of 8-10 cells or slices per group, α =
0.05, β = 0.8, corrected for the number of pairwise comparisons (τ), based on the following
equations:

$$n = 2\left(\sigma \frac{z_{1-\alpha/(2\tau)} z_{1-\beta}}{\mu A - \mu B}\right)^2,$$

$$1 - \beta = \phi\left(z - z_{1-\alpha/(2\tau)}\right) + \phi\left(-z - z_{1-\alpha/(2\tau)}\right),$$

$$z = \frac{\mu A - \mu B}{\sigma\sqrt{\frac{2}{n}}},$$

Where n is the sample size, σ is standard deviation, ϕ is standard normal distribution function, α
is Type I error rate, τ is the number of pairwise comparisons, and β is Type II error rate. During
the calculations, the normality of residuals and equality of variances were assumed *a priori*.

All of the calculations were performed in an online calculator available at:

http://powerandsamplesize.com/Calculators/Compare-k-Means/1-Way-ANOVA-Pairwise

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985 9. **Contributions**

Conceptualization: BK, BA, CW, LMP, NAS

Methodology: BK, BA, FG, AB, JMH, CW, MK, LMP, NAS

Software: FG, AB, JMH, BA, CW, MK

Validation: BK, BA, CW, LMP, NAS

Formal Analysis: BK, BA, CW, FG, AB, JMH

Investigation: BK, BA, CW, FG, AB, JMH, LMP, NAS

Writing – Original Draft: BK

Writing – Review & Editing: BK, BA, CW, FG, AB, VV, JMH, MK, LMP, NAS

Visualization: BK, BA

Supervision: LMP, NAS

Funding Acquisition: LMP, NAS

10. **Competing Interests**

Authors declare no competing interest.

1000

1001   11. **Data and materials availability**

1002   Raw data presented in this work, data tables with analyses and tables with statistics are available
1003   upon request. All code is available on GitHub, by following the links provided in the text.

1004

1005   12. **Supplementary materials**

1006   Figures S1 to S8, Supplementary Methods.