***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 (KBs) when less glucose is available. Neuronal KB uptake, which
does not rely on glucose transporter 4 (GLUT4) or insulin, has shown promising clinical
applicability 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 exogenous KB D-β-hydroxybutyrate (D-βHb) on mouse brain metabolism
during acute insulin resistance (AIR). We found that both AIR and D-βHb had distinct impacts
across neuronal compartments: AIR decreased synaptic activity and long-term potentiation (LTP)
and impaired axonal conduction, synchronization, and action potential (AP) properties, while D-
βHb rescued neuronal functions associated with axonal conduction, synchronization and LTP.

Keywords: beta-hydroxybutyrate, GLUT4, hippocampus, insulin resistance, ketone bodies

2. **Significance statement**

This study investigates the impact of acute insulin resistance on the functionality of the
hippocampal circuit and the potential protective effects of ketone body supplementation. By
inhibiting GLUT4 receptors to induce acute insulin resistance, we reveal several detrimental
changes caused by impaired neuronal glucose uptake. These changes include impairments in
synaptic activity, axonal conduction, and neuronal firing properties. The study further examines
the distinctive effects of acute insulin resistance (AIR) and the rescue agent D-βHb on synaptic
activity, long-term potentiation, axonal conduction, synchronization, and neuronal firing. By
shedding light on neuronal responses during insulin resistance, this investigation advances our
understanding of neurological disorders associated with hypometabolism and highlights the
potential therapeutic value of D-βHb.

3. **Introduction**

Over the past two decades, research on insulin signaling in the brain has become increasingly
important in studies 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, and 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 impairment in patients with 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 GLUT4 receptors, which are 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 the regulation of phosphatidylinositol 3-kinase (PI3K), which 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 the primary effect of insulin may be the maintenance of neuronal firing. Mouse models of AD or diet-induced obesity (DIO) showed blunting of the pro-cognitive effects of intrahippocampal insulin injections and decreased local glucose metabolism [18,19], coupled with neuronal hyperexcitability with epileptiform spikes and impairment of GLUT4 translocation [20].

During periods with lower glucose availability, ketone bodies (KBs) supplement metabolism as an alternative fuel source [21,22]. The ketogenic diet has shown pro-cognitive benefits in patients suffering from T2DM or AD. These benefits of the ketogenic diet may arise due to reduced 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 ratio, and reduced production of free radicals, which together may help to temper neuronal hyperexcitability and its metabolic consequences [27,28].

In addition, KBs are crucial 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 (D-βHb) ester, can increase KB levels in circulation without the severe dietary restrictions of the ketogenic diet [31,32]. Whether exogenous KBs maintain their neuroprotective effects in patients with normal plasma glucose levels remains unclear; however, recent studies have shown that in healthy adults, D-βHb led to increased overall brain activity and stabilized functional networks [33,34].

These observations led us to hypothesize that exogenous KBs could be utilized to rescue neuronal metabolism after cerebral insulin resistance. To test this hypothesis, we 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 determine the therapeutic potential of
exogenous KBs in hypoglycemic animals. To evaluate the circuit-wide effects on synaptic and
axonal function, we obtained field potential recordings in the hippocampi of the treated mice and
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
(FSIs). Finally, we used computational modeling approaches to relate our electrophysiological
results and Na^+/K^+ ATPase dysfunction as the potential reason for the detrimental changes
observed during AIR.

4. **Results**

Various paradigms of synaptic plasticity associated with learning and memory have been
identified in the hippocampus. Many studies have investigated long-term potentiation (LTP), long-
term depression (LTD), spike-timing-dependent plasticity, and excitatory postsynaptic potential
(EPSP)-spike potentiation in hippocampal circuits, making the hippocampus a classic system for
studying neuroplasticity. Furthermore, the simple cytoarchitecture of the hippocampus makes it
an ideal model system. In this study, we utilized indinavir, a potent GLUT4 receptor blocker, to
pharmacologically induce AIR in hippocampal slices. Our investigation focused on the stratum
radiatum of CA1, where numerous synapses are formed between Schaffer collaterals (SCOs)
and the apical dendrites of pyramidal neurons.

4.1 Synaptic activity and LTP, but not fiber volleys (FVs), are adversely affected by AIR and are
not reversed by either 0.1 mM or 1 mM D-βHb.

To test synaptic transmission within CA1 for 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 strong decreases in field excitatory postsynaptic potential
(fEPSP) amplitudes, -39.23 ± 8.72% (1st stimulation) and -38.79 ± 8.93% (2nd stimulation),
compared to baseline [**Fig. 1A**]. The amplitude did not recover 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]. No changes in the
paired-pulse ratio (PPR) were observed in any of the conditions [**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 activity at the
synaptic sites remained largely unchanged.

When applied over sufficient time, our paired stimulation triggered LTP in CA1, as evidenced by
a gradual increase in the fEPSP amplitude [**Fig. 1C, top**] but no change in the PPR [**Fig. 1C,**
bottom]. Therefore, we compared the LTP magnitude between the groups by calculating the ratio
of the EPSP amplitudes during the first 10 min and the 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]; however, 1 mM D-βHb did reverse this effect [**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 (FVs) [**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 ensure
that the onset of LTP did not skew the results, we recorded all train stimulations at least 40 min
after the paired stimulation paradigm.

All compared groups showed potentiation of fEPSP amplitudes during the train, reaching a peak
at stimuli 5-6. Thereafter, the amplitudes decreased slightly but remained potentiated [**Fig. S1A,**
B]. Interestingly, throughout the entire stimulation period, the fEPSP amplitudes matched the
differences we observed during the paired stimulation experiments [**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 the paired stimulation experiments, no statistically significant differences were
observed between groups at any time point during the train stimulations [**Fig. S1C**; p=0.16 to
0.40].

Finally, we examined the hippocampal circuit under intense nonphysiological stimulation
conditions. This involved applying long trains of 50 stimuli at 25 Hz, spanning across 60 trials with
a 20 s break between each trial. The responses during this specific stimulation condition varied
among trials, making it inappropriate to average the values over the entire stimulation period. As
a result, we represent the fEPSP and FV amplitudes as heatmaps. In these heatmaps, 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. 0.1 mM D-βHb+AIR, p=0.59; -13.66 ± 9.08% AIR vs. 1 mM D-βHb+AIR]. Synaptic

depression was first observed during the final stimuli of the stimulation trains, approximately 3
min into the stimulation period, and became more prominent with each subsequent train
stimulation application [**Fig. S2A-D**]. In general, the fEPSP amplitudes in the control and AIR
groups differed substantially during the first 7 min of the recordings [**Fig. S2A**]. The fEPSP
amplitudes in the control and 1 mM D-βHb+AIR groups also differed throughout most of the
stimulation trains [**Fig. S2B**], but the fEPSP amplitudes in the 0.1 mM D-βHb+AIR and AIR groups
scarcely differed except at the end of the stimulation trains [**Fig. S2C**]. The fEPSPs in the AIR
and 0.1 mM D-βHb groups were generally comparable. The FVs remained statistically similar
among all the groups, with approximately the same response pattern [**Fig. S2E-G**], replicating the
results in Fig. 1 and Fig. S1.

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

4.2 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 and a noticeable reduction in fEPSP
amplitudes, with the effects further exacerbated by D-βHB. This difference suggests that AIR
and/or D-βHb treatment might have differential effects on cellular compartments. We then
investigated whether other axonal properties changed during AIR and whether D-βHB application
remedied such changes.

We found that inducing AIR resulted in a strong reduction in the conduction velocities (CVs) of
SCOs [**Fig. 2C**; -13.07 ± 2.87% Control vs. AIR, p=0.027]. After the addition of 0.1 mM D-βHb
during AIR, the CV recovered 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. 0.1 mM D-βHb+AIR, p=0.99], and the application
of 1 mM D-βHb significantly improved the CV, surpassing the control level [**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 better interpret these findings, we constructed a computational model of the CV by employing
an approximation of axonal cable theory (see Methods). Our model examined three processes
that determine the CV: the 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, as this value primarily depends
on the 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 typical physiological ranges. We normalized the results according
to a reference state with V_R = -75 mV, V_A = 30 mV, and V_T = -50 mV. The modeling results showed
that decreased CV was associated with either hyperpolarized resting membrane potential or
decreased peak AP amplitude [**Fig. 2F-G**].

4.3 AIR adversely affects input timing, which is restored by high D-βHb concentrations.

Changes in the CV magnitude and the resulting differences in the input timing would likely
desynchronize the hippocampal circuit. Therefore, we investigated whether the 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.
Afterward, the AIR and 0.1 mM D-βHb+AIR groups displayed progressively more prominent
delays than the Control group, while the Control 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 increased delays during the stimulation suggest that the detrimental effects of AIR
are exacerbated over longer periods.

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

The above results [**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 D-βHb as a recovery agent. Based on the 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 the intrinsic membrane properties of CA1 pyramidal neurons (membrane
resistance, R_m; membrane capacitance, C_m, and resting membrane potential, V_{rest}) and their spike
thresholds. We found that the R_m of the CA1 pyramidal neurons in the AIR group was significantly
increased compared with that of the neurons in the control group [**Fig. 4B**, +43.24 ± 3.73%,
p=1.41E-07], and this increased value was observed even with the application of 1 mM D-βHb
[**Fig. 4B**, +27.86 ± 8.93%, p=0.030], while the results in the AIR and 1 mM D-βHb+AIR groups
were similar (p=0.31). Interestingly, neither the 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 the AIR or D-βHb+AIR treatments.

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

AIR significantly increased the frequency of spontaneous synaptic currents (sEPSCs), with
approximately twice as many events observed in the AIR group than in the control group [**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 sEPSC decay time did not differ significantly among the groups [**Fig. 5C-E**].

Given the change in R_m, the lack of significant changes in the amplitudes is surprising. Several
studies utilizing different metabolic states, such as food deprivation [41] and inhibition of glycolysis
[42], showed increased mEPSC/sEPSC amplitudes; therefore, we decided to investigate the
sEPSCs in more detail. In most central synapses, the distribution of mEPSC/sEPSC amplitudes
shows several peaks related to the simultaneous release of multiple neurotransmitters [43], with
each subsequent peak linked to an increasing number of concurrent sEPSCs. According to 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 in the number of cells with multipeak distributions among the groups [**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 among the groups [**Fig. S3E**]: the amplitudes in the AIR
group were significantly higher than those in the D-βHb+AIR group [p1: p=0.048; p2: p=0.0054;
p3: p=0.0038; p4: p=0.021] and differed from those in the control group at peak 3 [p=0.007]. The
Control and D-βHb+AIR groups did not show significant differences at any point [**Fig. S3E**],
despite the same sEPSC frequency as in the AIR group.

We validated these findings by comparing the normalized amplitude distributions among the
different groups [**Fig. S3F-I**]. Within the range correlating to peaks 2–4 (sEPSC of ~20 pA or
greater), significantly more large events were observed in the AIR group (15.19% of total) than in
the control (7.65%) or D-βHb+AIR (9.48%) groups [**Fig. S3F-I**], while the results in the control and
D-βHb+AIR groups were essentially comparable. This result suggests that during AIR, the
elevated frequency of events increased the probability of multivesicular events occurring, which
was further exacerbated by the ~20% increase in R_m. On the other hand, D-βHb+AIR produced a
distribution resembling the control, as if the vesicles had lower glutamate content, as suggested

by the literature [44,45].

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

The AP generation and subsequent recovery of V_m to resting levels require 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) during AIR and how D-
βHb alleviates AIR effects.

First, we focused on the firing frequency [**Fig. 6A**]. Cells in the AIR group showed increased firing
at lower current injections than those in the D-βHb+AIR group [**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
in the AIR condition than in the control or D-βHb+AIR conditions. To test this further, we measured
the maximum depolarization obtained 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 control neurons [**Fig. 6E, F**; p=0.0027 to 0.048] or D-βHb+AIR [**Fig. 6E, F**; p=0.0029
to 0.042].

Next, we investigated the average AP peak amplitudes [**Fig. 6G**], per step [**Fig. 6H**]. Surprisingly,
D-βHb+AIR neurons showed larger amplitudes, which were consistently elevated throughout the
injection period, than control or AIR neurons, while AIR neurons remained comparable to control
neurons [**Fig. 6I**, p=0.0017 to 0.036, AIR vs. Control; p=0.0093 to 0.044, AIR vs. 1 mM D-
βHb+AIR].

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

We next proceeded to investigate other AP properties: the 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 than AIR [**Fig. 7B, C**; p=0.047 to
0.0019] and D-βHb+AIR neurons [**Fig. 7B, C**; p=0.041 to 0.0023] for almost all current injection
steps. Moreover, 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 than control cells [**Fig. 7E, F**;
p=0.000094 to 0.0025] and AIR cells [**Fig. 6E, F**; p=0.047 to 0.0052] for most injection steps.
Interestingly, AIR cells exhibited slightly faster AP rise times exclusively at the lowest current
injection times [**Fig. 4N, 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**].

Finally, as we did not observe significant differences in the 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 interevent interval, AP IEI) [**Fig. 7G**], but we did not observe obvious
changes in the AP IEI [**Fig. 7H, I].**

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

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

To test this observation, we compared the 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 represent neuronal firing after rest (10 s between subsequent injections), while the
average and final APs represent firing after energy expenditure. As expected, the first APs
displayed minimal changes in amplitude during injections [**Fig. S4D**], with larger amplitudes
observed in the D-βHb+AIR group than in the control group. Moreover, although elevated
amplitudes were observed in the AIR group, the results did not differ significantly from those in
the control or D-βHb+AIR groups. For the final APs, we observed a sharp decline in amplitude,
which scaled with the injected current [**Fig. S4F**], in both the AIR and D-βHb+AIR groups. Lower
amplitudes were observed in the control group than in the D-βHb+AIR group for 100-300 pA
injections (p=0.040 to p=6E-05), and the results remained comparable to those in the AIR group
until the final, large, 400-500 pA injections (p=0.032 to p=0.0011), when the APs in the AIR group
decreased significantly [**Fig. S4F**]. The results in the AIR and D-βHb+AIR groups remained
consistently different. Linear fits to the average AP amplitudes revealed significantly steeper
declines in the AIR and D-βHb+AIR groups [**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 and
D-βHb does not provide direct recovery.

4.9 *Changes in AP decay and rise times correspond to differences in the widths of FVs.*

In our initial analysis of the field potential (FV) data, we focused on the amplitudes of the biological
signals. Consistently, we found no significant changes in the FV amplitudes [**Figs. 1F, 2C, 3E-G**],
which supports our findings on AP overshoot amplitudes [**Fig. 6G-I**] (except in the D-βHb+AIR
group). Based on these results, we investigated whether the changes in the AP decay and rise

times corresponded to changes in the FV widths during train stimulation [**Fig. S5**].

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

4.10 *With abolished GABAergic inhibition, AIR strongly decreased the firing rates of CA1*
pyramidal neurons and fast-spiking interneurons (FSIs), which were not rescued by D-βHb.

We next 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**] and found that AIR prominently decreased their firing rate when
compared with the 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 the firing rate than in the control group [**Fig.**
S6E, p=0.000027 to 0.035]. The 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 the highest firing rate observed in the control group (31.03 ± 3.43) and the lowest firing
rate observed in the D-βHb+AIR group (13.17 ± 2.05).

Next, we examined CA1 FSIs using the same drug conditions [**Fig. S6E, F**]: AIR decreased the
firing rate of FSIs compared with the control [**Fig. S6G**, p=0.012 to 0.048], which was not reversed
by D-βHb [**Fig. S6H**, p=0.00017 to 0.035]. For pyramidal neurons, the FSIs in the 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 The *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**). **Fig. S7A, B** (at two
different temporal scales) shows the predicted changes in neuronal firing dynamics when the

Na$^+$/K$^+$ ATPase is impaired. This impairment predicts the depolarization of the resting membrane
potential [**Fig. S7C**] and is consistent with the trends of our I/O curves [**Fig. 4F**]. Furthermore, in
this depolarization model, the magnitude of the membrane potential shift needed for an AP is
reduced. 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 amplitudes of the subsequent APs to decline progressively with time [**Fig. S7E,**
F]. This amplitude decrease is consistent with the trends in the measured CV [**Fig. 2**].

5. **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 hippocampus-mediated memory tasks, thus suggesting
that GLUT4 is critical for the acquisition and consolidation of memory [8,13]. Memory formation
promotes GLUT4 incorporation into the membrane, resulting in increased glucose flux into
neurons [13,49]. This process enables increased metabolic support during high-demand
activities. Crucially, most GLUT4s localize in the perikaryon [50,51], likely near axosomatic
synapses, which are associated with synaptic plasticity and long-term memory (LTM). Indeed,
prolonged blockade of GLUT4 glucose transporters 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 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 presynaptic terminals
[41,55]. These studies also suggest that presynaptic glycolysis is essential for maintaining
synaptic transmission, even at low frequencies, and that mitochondrial respiration cannot
compensate for these effects. Normally, glycolysis requires approximately one-third of the energy
used at presynaptic terminals, thus sustaining low-frequency transmission. In addition, impaired
glycolysis leads to slower, broader, and smaller AP waveforms and depolarization of the resting
membrane potential [42], which is consistent with our results. Interestingly, D-βHb+AIR treatment,
which showed the same increase in the frequency of sEPSC events as AIR, led to the lowest
sEPSC amplitudes [**Fig. S3E**]. This result suggests a reduction in the number of synaptic vesicles
and is consistent with reports of reduced conversion of glutamate to aspartate during ketosis
[44,45].

Although the results of our computational models are consistent with the observed changes in the
resting-state membrane potential and firing frequency during AIR, there were seemingly
contradicting results with respect to the AP peak amplitude. Specifically, we found no significant

changes in the average AP overshoot amplitudes during AIR, and 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 such 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 depleted and, provided that the neuron rapidly fires multiple APs, the negative
effects of hypoglycemia overcome compensatory R_m increases; then, the AP amplitudes
decrease below control levels [**Fig. S4**]. Furthermore, we used the maximal Na^+/K^+ ATPase
activity to represent ATP availability and considered literature estimates of kinetic and
thermodynamic rate constants. A more direct approach might consider ATP concentration as an
input variable, in cases in which such experimental data are available.

Our model aims to establish the underlying causes of the changes in CV, and we 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, as the measured AP peak
amplitudes remain consistently elevated and V_{rest} changes are negligible, but fails in the case of
AIR alone, with the AP amplitudes remaining comparable to the control levels and V_{rest} tending to
depolarize. This discrepancy might reflect the simplistic nature of our model, which neglects
complex ion dynamics, altered membrane permeability and synaptic processes.

The ketogenic diet has been used to treat epilepsy for decades and has been considered in many
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 prevents picrotoxin-induced epileptiform
activity in the hippocampus [60]; thus, this action may explain the 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 on
the ketogenic diet in patients with epilepsy [61]. D-βHb also directs glutamate toward GABA
production by reducing the conversion of glutamate to aspartate [44,45]. In addition, D-βHb
directly affects the vesicular glutamate transporter VGLUT2 by inhibiting Cl^- dependent glutamate
uptake. VGLUT2 is critical for glutamate excitatory transmission in CA3-CA1, and conditional
VGLUT2 knockout results in impaired spatial learning and reduced LTP [62]. Together with the
increased formation of GABA and/or GABA derivatives, VGLUT2 inhibition could 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 display symptoms of prediabetes, and this
percentage is expected to increase. With longer average lifespans and demographic changes
leading to an increase in the aging population, obtaining insights into the outcomes of metabolic
dysfunction and its potential treatment has become a crucial issue. Our results provide 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 more effective therapies. Further research in this
area could inform novel treatment approaches that address the complex effects of metabolic
disorders on brain health.

6. **Materials and Methods**

6.0 *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 a 12-12
hour 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.

6.1 *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) and decapitated, and their brains were
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 °C. The slices were allowed to recover in
warm ACSF for 15 min; then, the heating was switched off, and the chamber was gradually cooled
to room temperature (RT; 23 ± 2 °C) over 45 min.

Following 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
D-βHb), 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.

6.2 *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 from the stimulation electrode
for the CA1 FV and fEPSP recordings. During the axonal conduction velocity (CV) recordings, the
electrodes were spaced 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 with a total duration
of 240 µs, 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 conditions.

Recording pipettes were pulled from borosilicate glass capillaries (ID 0.86, OD 1.5, BF1-86-10,
Sutter Instruments, USA) with 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).

6.3 *Whole-cell patch-clamp recordings.*

Pyramidal neurons in 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) with 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_2ATP, 2 $MgCl_2$, 0.2 EGTA,
10 HEPES, 4 KCl, 7 di-triphospho-creatine, 0.3 Na_3GTP; 280–290 ± 5 mOsm/kg, titrated to pH

7.3 ± 0.05 with KOH.

During the voltage clamp (VC) recordings, cells were clamped at a holding potential (V_h) of −70
mV with a Multiclamp 700B amplifier (Molecular Devices, USA). The liquid junction potential was
not corrected. The pipette capacitance was compensated for in all recordings. During VC
experiments, the series resistance (R_s) was not compensated. During the current clamp (CC)
recordings, a holding current was applied to the cells to maintain the membrane potential at $V_h =$
-70 ± 3 mV. All cells were bridge-balanced at 50 % of R_s. The cell capacitance was not
compensated for 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
applied to monitor changes in R_s (V_h = -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) The R_s value after any recording protocol could not be more than 20% larger than
the R_s value before the protocol; 2) the R_s value during any of the recording protocols could not
exceed 30 MΩ; and 3) the offset drift by the end of the recordings could not be higher than ± 5
mV.

Spontaneous synaptic currents (sEPSCs) 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
subjected to a ramp-shaped 200-500 pA current injection with a 1 s 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 the recorded neurons, each neuron was current-clamped as
described previously and subjected to a series of square current injections with a change of 25
pA per step to a maximum of 0 pA, with each step having a 1 s duration and an interstep 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).

6.4 *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 in the presence of TTX.

Recorded traces were analyzed in Igor Pro (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 the 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, with the first peak after the artifact subtraction and the
second peak directly before the fEPSP.

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

To measure the 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 values 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, the stimulation
site remained stationary, and the sites were recorded in randomized order.

To calculate changes in the FV peak latency during train stimulations, the median Δt for stimuli 1-
3 and traces 1-2 (Δt_N) was subtracted from each Δt (normalization).

6.5 *Field potential recordings – data visualization.*

To construct "heatmap" visualizations of the data, numerical matrices of the 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. Next, the text files were imported as text images to ImageJ (NIH, USA), converted from 32-
to 16-bit, saved as TIFF images and imported into Igor Pro (Wavemetrics, USA); then, their look-
up table (LUT) was changed from "Grays" to "Spectrum."

The heatmaps for latency changes during train stimulation were generated similarly but using

different normalizations: the 20 pulse train recordings had normalization boundaries of -1 and +1
ms.

6.6 *Whole-cell patch-clamp – sEPSC analyses.*

sEPSC detection was automatized using a convolution-based algorithm in Fbrain 3.03 [64], a
customized program in Igor Pro 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 convolved trace was passed through a digital bandpass filter at 0.001 to 200 Hz.
The event detection template had a rise-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. Events that clearly did not show the fast rise times and
exponential decay kinetics of a typical hippocampal EPSC were manually removed. 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 segment
after the sEPSC onset to capture the full decay of the waveform. Before averaging, all sEPSCs
were baseline-adjusted to the pre-onset time by 5 ms. The peak amplitude, 20-80% onset-to-peak
rise time, and decay constant τ were measured for the averaged sEPSCs. The decay constant 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
for each cell and averaged the results.

6.7 *Statistics.*

All data were acquired in randomized sequences, 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 preincubation 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 were neither normally distributed nor had equal variance, we applied 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 n, replicates are
labeled m, and the numbers of animals used are labeled N. If the groups tested are 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 the 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 the mean ± SEM. The heatmaps represent
the group means.

6.8 *Model of conduction velocity*

Here, we model how the 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 is activated and
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 $V_2(t) = V_t$ and
$V_2(0) = V_{rest}$). Over the discrete lattice described 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 can approach zero. Thus:

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

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

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

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

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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. **Conflicts of interest**

The authors declare no conflicts of interest.

11. **Data and materials availability**

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

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13. **Figures.**

Figure 1: Synaptic activity and LTP, but not neuronal firing or PPR, decrease during paired stimulation under AIR conditions 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. The black bars represent the 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 the fEPSPs recorded in A. The labels, n, are identical to those in A). There were no significant differences (p=0.41, ANOVA). C) On top: Representative recording of a control experiment in which paired stimulation was applied for 50

min. Each circle represents the fEPSP 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 mean amplitude at +30 to +40 min. Bottom: corresponding normalized PPR. D)
fEPSP waveforms recorded in C), showing LTP development. In gray, 30 fEPSP waveforms at
t=0 ± 5 min and their average (black). In yellow, 30 fEPSP waveforms at t=+30 ± 5 min and their
average (orange). Stimulation artifacts and FVs are removed. E) Scatter plot of LTP signals
triggered by paired stimulation. Control: n=13; AIR n=11; 0.1 mM D-βHb+AIR: n=12; 1 mM D-
βHb+AIR: n=14. F) Scatter plot of the FV amplitudes recorded during A). The labels, n, are
identical to those in A). There were no significant differences (p=0.42, p=0.40; ANOVA). In all
plots, *p<0.05; **p<0.01; ***p<0.001.

Figure 2: Conduction velocity (CV) of Schaffer collaterals decreases under AIR conditions, is rescued by 0.1 mM D-βHb, and increases with 1 mM D-βHb.

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) CV 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 the mean ± 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 the CV of an AP propagating along an axon: distance along the axon (Δx)
and the time needed for an AP to pass that distance (t_2-t_1). V represents the membrane potential.
E) Model of a propagating AP. The change in V(t) is shown at longitudinal increments along the
axon (x_2) and at neighboring increments immediately before (x_1) and behind (x_3). The membrane
potential at rest (VR) 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
(VT), a spike occurs, and the membrane potential reaches the peak value (VA). F-G) CV is
modulated by VA and VR. Our computational model predicts declining CV due to reductions in
the peak AP (F) and/or hyperpolarization of the resting membrane potential (G). The CV is
quantified as a percentage relative to the control.

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

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) Heatmap 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
color-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 based 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 the results of the
Control (E) and 1 mM D-βHb+AIR (H) groups. There were no significant differences, p=0.067 to
p=0.99. K) The same as I) comparing the results of the AIR (F) and 0.1 mM D-βHb+AIR (G)
groups. There were no significant differences, p=0.41 to p=0.99. L) The same as I) comparing the
results of the AIR (F) and 1 mM D-βHb+AIR (H) groups.

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) Left: A CA1 pyramidal neuron recorded in whole-cell patch-clamp mode (outlined in red), with
resistances and capacitance denoted within the cell. Right: A simple electric circuit diagram of a
cell with the same resistances. B) Scatter plots of CA1 pyramidal neuron Rm compared among
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 the mean ±
SEM. C) Scatter plots of CA1 pyramidal neurons Cm compared among the experimental groups.
The plots and coloring are identical to B). There were no significant differences (p=0.32). D)

Representative example of a 300 pA 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 appreciate, with the arrow pointing to the onset. E) Scatter plots of CA1 pyramidal
neuron spike thresholds compared among the experimental groups. The plots are identical to B).
There were no significant differences (p=0.17). F) Scatter plots of CA1 pyramidal neurons RMP
(V_{rest}) compared among the experimental groups. The plots and coloring are identical to B). There
were 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, which is restored by high D-βHb concentrations.

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Figure 5: AIR increased sEPSC frequency without altering sEPSC properties, and D-βHb
administration did not reverse this effect.

A) Representative examples of the first 2 min of sEPSC waveforms 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). Bottom: averaged sEPSC waveforms from individual cells (light color) and the group
mean (full color). B) Scatter plots of the sEPSC frequency recorded in CA1 pyramidal neurons
compared among the experimental groups. Each colored circle represents the frequency
recorded in a single cell. The labels are the same as in A). Vertical black bars represent the mean
± SEM. *p<0.05. C) Scatter plots of the sEPSC 20-80% rise times, labeled as in B). There were
no significant differences (p=0.51). D) Scatter plots of the sEPSC decay tau times, labeled as in
B). There were no significant differences (p=0.98). E) Scatter plots of the sEPSC amplitudes,
labeled as in B). There were 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.

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Figure 6: Neuronal firing is mildly affected by AIR and reversed by D-βHb, with D-βHb 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+25 pA injections (max. 0 pA; V_h=-70 mV). Circles represent the 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; ANOVA) Maximum firing rate of the pyramidal neurons in B). Circles represent single neurons; black bars represent group mean ± SEM. There were no significant differences (p=0.778, ANOVA). D) Example of responses to Δ+25 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) Vm at the current injections from B). F) Group mean ± 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. The black line marks V_m= 0 mV. H) Averaged AP overshoot amplitudes from B). I) Group mean ± SEM of overshoot amplitudes from H).

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Figure 7: AP decay times change during AIR and D-βHb + AIR conditions, but D-βHb + AIR APs have faster rise times.

A) Example of AP decay time measurement. The black line marks V_m= 0 mV. B) Mean AP decay times for the current injections. Circles represent averaged amplitudes in single neurons. The black lines represent group means. C) Group mean ± SEM of AP decay times from B). Statistical tests and labels are identical to B). D) Example of AP decay time measurement. The black line marks V_m= 0 mV. E) Averaged AP rise times for the current injections in B). Circles represent averaged amplitudes in single neurons. The black lines represent group means. F) Group mean ± SEM of AP rise times from B). Statistical tests and labels are identical to B). G) An example of the AP IEI measurement. The double arrow marks the latency between AP peaks. H) Averaged IEI recorded during the injections in B). Each point represents the average IEI recorded at each step in a single neuron. The black line represents the group mean. I) Group mean ± SEM of

1046   averaged IEI. *,+,^ p<0.05; ANOVA. In all comparisons, Control n=23, N=20; AIR n=20, N=14; 1

1047   mM D-βHb+AIR, n=22, N=16.