

1 **Ketamine and sleep modulate neural complexity dynamics in cats**

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14

15 **Abstract**

16

17 There is increasing evidence that level of consciousness can be captured by neural informational
18 complexity: for instance, complexity, as measured by the Lempel Ziv (LZ) compression algorithm,
19 decreases during anesthesia and non-rapid eye movement (NREM) sleep in humans and rats, when
20 compared to LZ in awake and REM sleep. In contrast, LZ is higher in humans under the effect of
21 psychedelics, including subanesthetic doses of ketamine. However, it is both unclear how this
22 result would be modulated by varying ketamine doses, and whether it would extend to other
23 species. Here we studied LZ with and without auditory stimulation during wakefulness and
24 different sleep stages in 5 cats implanted with intracranial electrodes, as well as under
25 subanesthetic doses of ketamine (5, 10, and 15 mg/kg i.m.). In line with previous results, LZ was
26 lowest in NREM sleep, but similar in REM and wakefulness. Furthermore, we found an inverted
27 U-shaped curve following different levels of ketamine doses in a subset of electrodes, primarily in
28 prefrontal cortex. However, it is worth noting that the variability in the ketamine dose-response
29 curve across cats and cortices was larger than that in the sleep-stage data, highlighting the
30 differential local dynamics created by two different ways of modulating conscious state. These
31 results replicate previous findings, both in humans and other species, demonstrating that neural
32 complexity is highly sensitive to capture state changes between wake and sleep stages while adding
33 a local cortical description. Finally, this study describes the differential effects of ketamine doses,
34 replicating a rise in complexity for low doses, and further fall as doses approach anesthetic levels
35 in a differential manner depending on the cortex.

36

37 **Keywords**

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39 Ketamine; Sleep; Psychedelics; Complexity; Local Field Potential; Cortex; Thalamus; Cats

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46 **Introduction**

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48 There is increasing evidence for a strong association between neural information measures, such
49 as electrophysiological signal complexity, and level of consciousness (Abásolo et al., 2015;
50 Castro-Zaballa et al., 2019; Mateos et al., 2018; Schartner et al., 2015; Schartner, 2017; Schartner,
51 2017; Zhang et al., 2001). One of the most studied neural complexity metrics is Lempel-Ziv
52 complexity (LZ), capturing the number of distinct substrings or patterns within a sequence
53 (Lempel & Ziv, 1976; Ziv & Lempel, 1978). A decrease in complexity has been demonstrated for
54 anesthesia (Li & Mashour, 2019; Schartner et al., 2015; Zhang et al., 2001), and during non-rapid
55 eye movement sleep (NREM sleep) when compared to normal wakefulness. However, REM
56 complexity has consistently been shown to be above NREM sleep and below normal wakefulness
57 (Abásolo et al., 2015; Andrillon et al., 2016; Mateos et al., 2018; Schartner et al., 2017). The
58 increase in complexity during REM, where vivid dreaming often occurs, may lend credence to the
59 hypothesis that complexity may not only be modulated by consciousness level but also signal the
60 degree of contents of consciousness (Abásolo et al., 2015; Mateos et al., 2018).

61

62 Further evidence for LZ associated with an increase in the range of conscious contents comes from
63 higher LZ during resting state in humans under the effect of psychedelics, specifically LSD,
64 psilocybin, and subanesthetic doses of the dissociative NMDA-antagonist ketamine, compared to
65 placebo (Li & Mashour, 2019; Mediano et al., 2020; Schartner, et al., 2017). These drugs have
66 profound and widespread effects on conscious experiences, both internally and externally
67 generated. More specifically, they appear to “broaden” the scope of conscious contents, vivifying
68 imagination and positively modulating the flexibility of cognition (Carhart-Harris et al., 2016;
69 Carhart-Harris et al., 2014). For all three drugs, reliably higher spontaneous signal diversity was
70 reported. More recently, a higher level of complexity following a subanesthetic dose of ketamine
71 was also reported (Li et al., 2019; Farnes et al., 2020) in spontaneous high-density scalp
72 electroencephalography (EEG) signals in healthy volunteers, but no increase was observed when
73 auditorily stimulated.

74

75 Ketamine also appears to maintain spatiotemporal complexity, as measured through the
76 perturbational complexity index (PCI) (Sarasso et al., 2015). PCI is the result of applying LZ to

77 the spatiotemporal pattern of cortical activation evoked by transcranial magnetic stimulation
78 (TMS), and has proven to be a reliable classifier of level of consciousness (Casali et al., 2013).
79 PCI decreases during propofol, midazolam and xenon anesthesia (Casali et al., 2013), but
80 maintains wakefulness baseline level during ketamine anesthesia (Sarasso et al., 2015).

81
82 Despite this body of work, important questions remain unanswered. First, prior studies provide
83 only a disjointed picture by investigating the effect of anesthetic dose in TMS-evoked cortical
84 activation (Sarasso et al., 2015) or subanesthetic dose in spontaneous MEG signals (Schartner et
85 al., 2017). For a more complete understanding of ketamine's psychoactive effects, a systematic
86 investigation of the dose-dependent effects of ketamine on cortical complexity using the same
87 modality is required. Therefore, in this work we aimed to investigate the level of informational
88 complexity during different stages of sleep in the cat as well as under subanesthetic doses of
89 ketamine in a dose-dependent manner, compared to the control awake state. Additionally, we
90 determined how the complexity measures under ketamine compared to baseline conditions, with
91 or without the presence of sensory stimulation. Finally, we sought to understand the possible
92 differences in informational complexity between resting-state periods and sensory stimulation
93 periods across conscious states, in an attempt to characterize the interaction between psychedelic
94 states and perturbational states. Accordingly (Pascovich et al., 2019), the following hypotheses
95 were proposed: (1) LZ would reflect sleep level: LZ in wakefulness would be just above REM
96 sleep. REM sleep would be above light sleep (LS), and NREM sleep would have the lowest
97 complexity value; (2) LZ would be increased during the initial period of drug infusion compared
98 to baseline wakefulness; (3) the level of complexity would be higher under sensory stimulation
99 compared to baseline, for both conditions, with and without ketamine; and (4) stimulation-induced
100 complexity increase would be more evident under the effect of ketamine.

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103 **Results**

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105 *Sleep shows a state-dependent effect on Lempel-Ziv complexity*

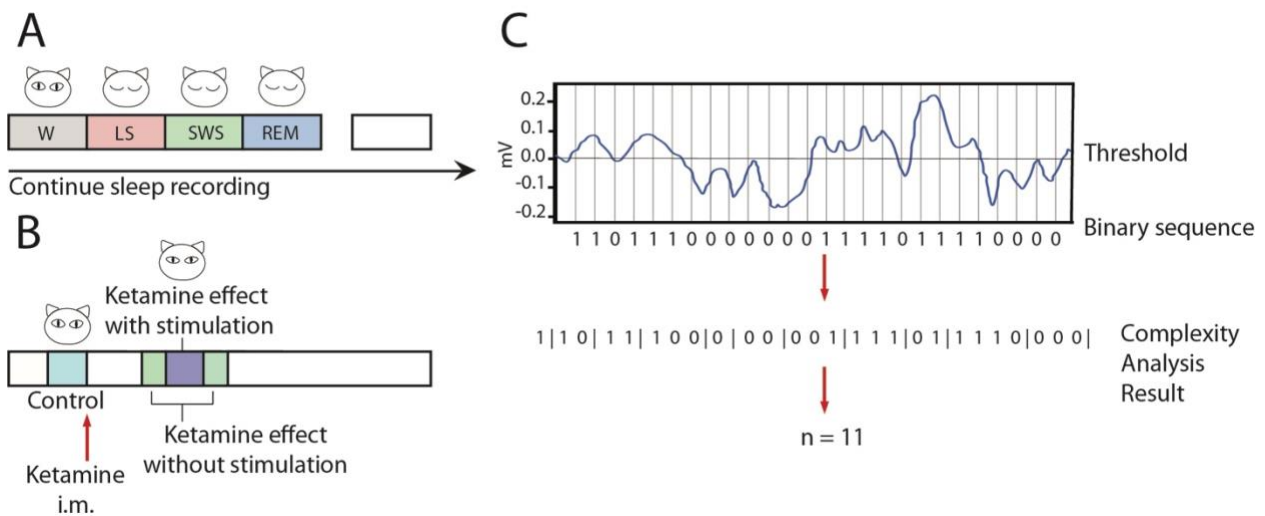
106 Cats underwent a polysomnographic recording in semi-restricted conditions where they were
107 adapted to sleep. Data were obtained during spontaneously occurring quiet wakefulness, LS,

108 NREM sleep and REM sleep (Figure 1A). LZ was computed using the LZ78 algorithm (Ziv &
109 Lempel 1978; Figure 1C) from the different sleep stages for all the cortices available (Figure 2A).
110 Effect sizes for differences between states at the single subject level are shown in Figure 2B. For
111 all animals, LZ scored higher for wakefulness than NREM sleep (Cohen's $d > 0.8$) for most of the
112 cortices. As predicted, LZ values were highest for REM and W, intermediate for LS, and lowest
113 for NREM.

114

115 Additionally, mixed effects models were formulated for each cortex including the cat as a random
116 effect when applicable. Thereafter, model selection was performed between linear and quadratic
117 models using Bayes Factors (BF) to decide between U-shaped and linear fits. All model
118 comparisons between linear and non-linear quadratic fits showed the supremacy of the non-linear
119 fit (Table 1) in agreement with our previous hypotheses, where the REM sleep showed higher
120 complexity values than the deep sleep - with the exception of the right somatosensory cortex,
121 where the results showed a flattening of the curve compared to all other cortices (Figure 2).

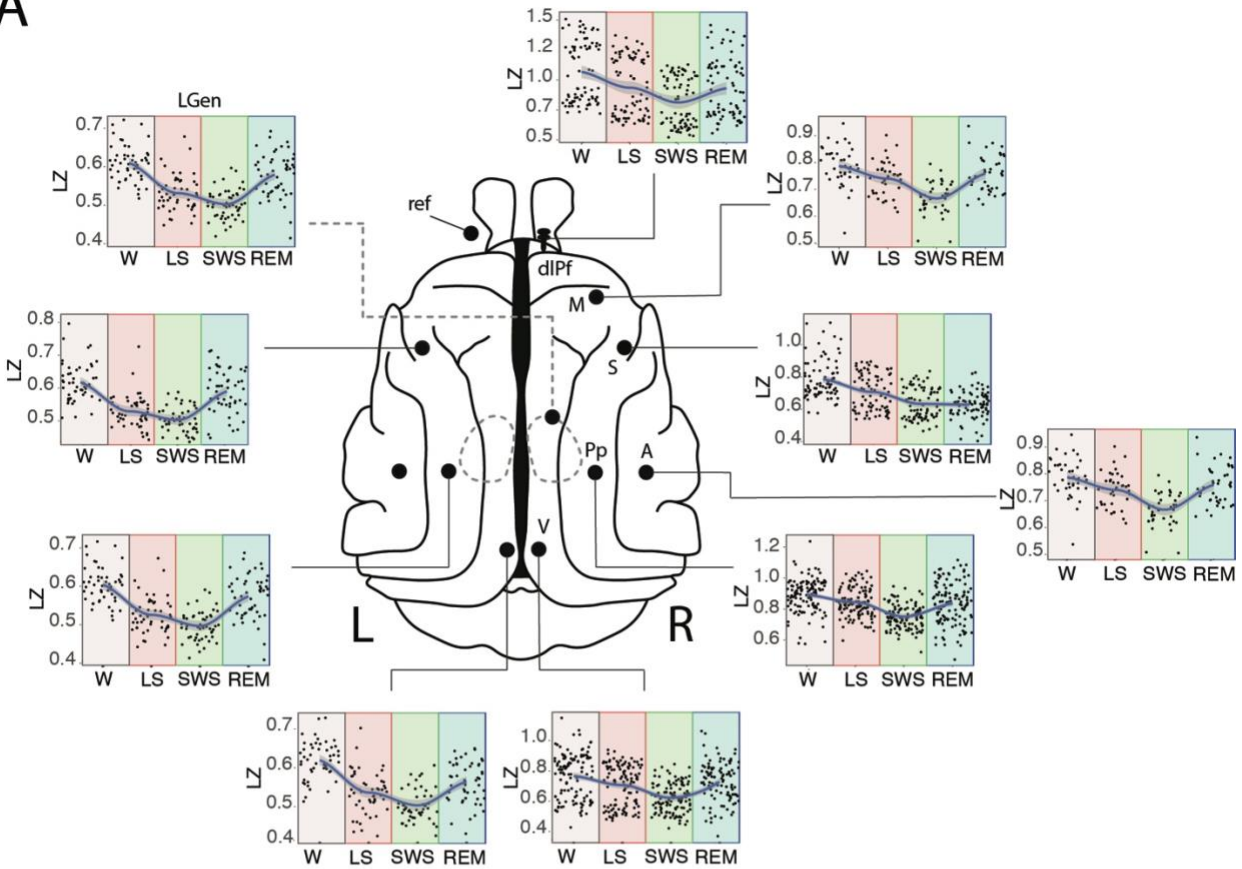
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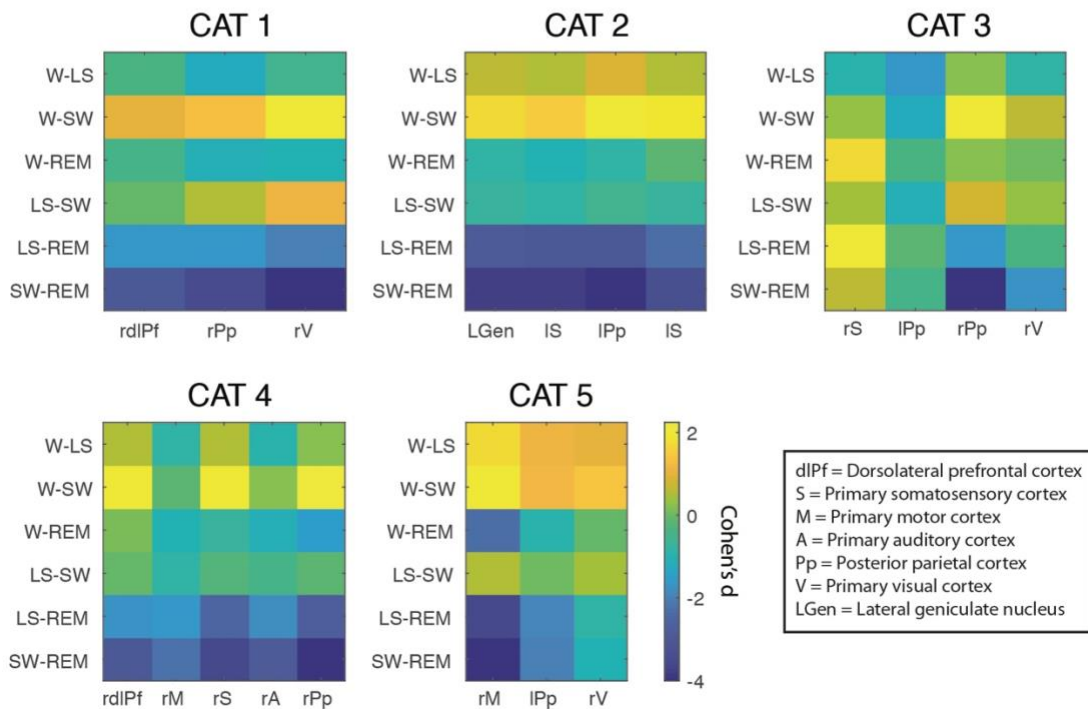
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124 FIGURE 1. Schematic illustrating the experimental design for electrocorticographic recordings
125 during the different states of sleep (A) and before, and after the different doses of ketamine (B).
126 i.m., intramuscular. (C) Illustration showing how to transform a segment of EEG signal series into
127 a binary sequence and the result of the LZ complexity analysis on the binary sequence.

A



B



129 **FIGURE 2. Cortical dynamic of LZ during sleep.** Schematic representations of the cat brain are
 130 used to visualize the differential dynamics of LZ during wakefulness and different states of sleep
 131 (A), showing a U-shaped complexity curve with state progression from W to LS, SWS and REM.
 132 “L” indicates Left side and “R” right side. (B) The differences in average LZ between sleep states,
 133 as measured by ANOVA and Tukey post-hoc test. Effect sizes were calculated by Cohen’s d and
 134 represented in a colour scale, where yellow means a positive difference and blue means a negative
 135 difference between the effect sizes of the pair of states compared. W = wakefulness; LS = light
 136 sleep; SWS = slow wave sleep; REM = rapid eyes movements sleep; dlPf = dorsolateral prefrontal
 137 cortex; Pp = posterior parietal cortex; V = visual cortex; LGen = lateral geniculate nucleus; S =
 138 somatosensory cortex; M = motor cortex; A = auditory cortex; ref, reference electrode location. In
 139 Figure B, “r” indicates right and “l” indicates left cortex.
 140

Cortex	Nº of cats	Model	BF
Right dorsolateral prefrontal	2	linear	
		quadratic *	6.88x10 ¹⁸
Right primary motor	2	linear	
		quadratic *	1.90x10 ¹⁷
Right primary auditory	1	linear	
		quadratic *	4.42x10 ⁶
Right primary somatosensory	2	linear	
		quadratic	0.681
Left primary somatosensory	1	linear	
		quadratic *	8.53x10 ²⁰
Right posterior parietal	3	linear	
		quadratic *	1.76x10 ⁷
Left posterior parietal	1	linear	
		quadratic *	7.19x10 ⁹
Right primary visual	3	linear	
		quadratic *	3.70x10 ¹¹
Left primary visual	1	linear	
		quadratic *	3.77x10 ¹⁷
Right Lateral Geniculate	1	linear	
		quadratic *	7.00x10 ²⁰

141
 142 **TABLE 1. Selection between linear and non-linear models among different sleep stages for**
 143 **each cortex.** Mixed effects models were formulated for each cortex including the cat as a random

144 effect when applicable. Bayes Factors (BF) were used to decide between U-shaped and linear fits.
145 With the exception of right primary somatosensory cortex, all model comparisons showed the
146 supremacy of the quadratic fit. The asterisks indicate substantial evidence for a quadratic fit (BF
147 >5).

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149

150 *Heterogenous cortical dynamics across cortices under ketamine*

151

152 For this experiment, the data had been collected under the same experimental conditions than for
153 sleep recordings in the same cats, and i.m. injections of ketamine of 5, 10 or 15mg/Kg were
154 performed in separate non-consecutive days as schematized in Figure 1B (see Methods). Again,
155 LZ was calculated in epochs before and after the administration of the drug. To address dose-
156 response relationships, a multilevel model was used where LZ was predicted by dose (fixed effect),
157 and cat and session were considered as random effects (with sessions nested within cats, and each
158 dose of ketamine was repeated four times).

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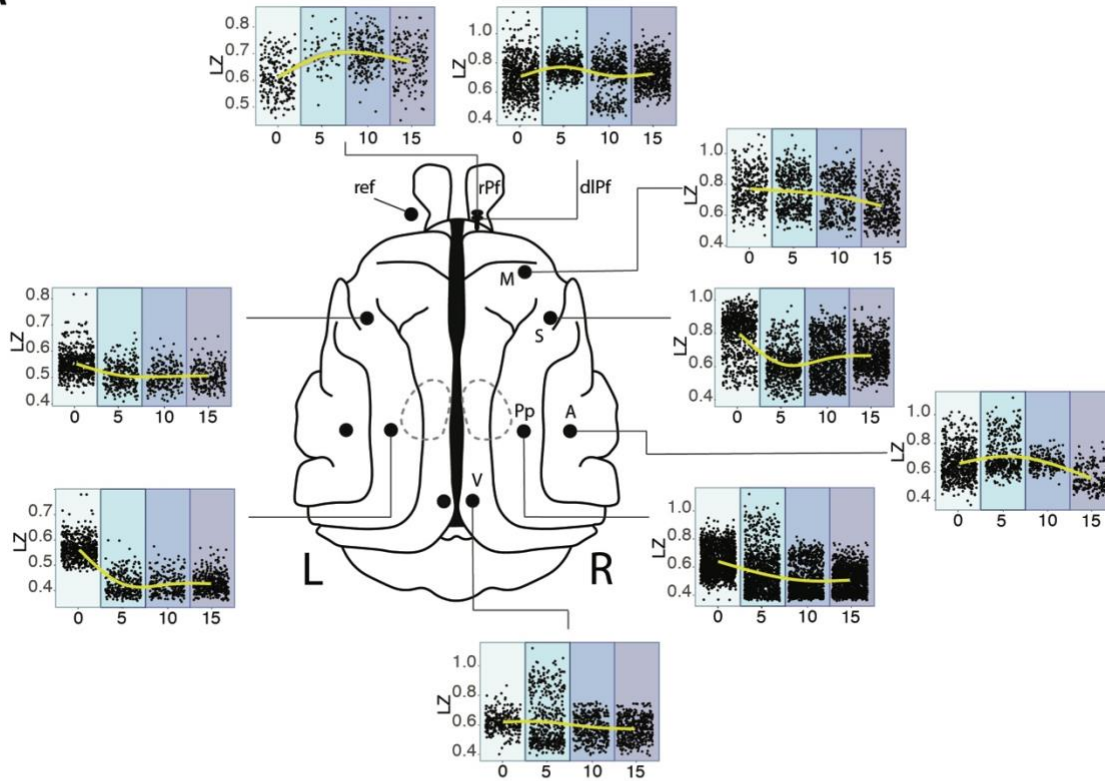
160 Considerably greater LZ variability was observed under ketamine than for the sleep results,
161 especially during the lowest doses explored (Figure 3A). In some regions, the results were in
162 agreement with our hypothesis, which predicted an increase in informational complexity after the
163 lowest ketamine dose, followed by a decrease after the higher dose showing an inverted U-shaped
164 relationship. This was observed clearly in right rostral and dorsolateral prefrontal cortices as well
165 as the right primary auditory cortex (BF = 1.70×10^{14} , BF = 2.57×10^5 , BF = 8.25×10^9 , respectively).
166 However, when we look at the individual effect per cortex in each animal, it can be seen that the
167 inverted U-shaped relationship is not systematic between cats and is present only in the cortices of
168 2 out of 5 cats (Figure 3B).

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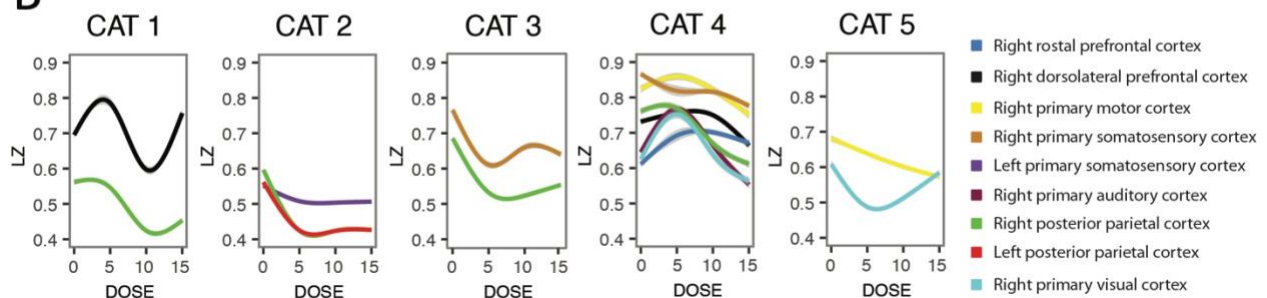
170 On the other hand, an opposite curve was obtained for somatosensory and posterior parietal
171 cortices. Finally, for the visual cortex the effects were less consistent among cats; in this last
172 example, the two cats tested had different responses to ketamine with opposite effects (Figure 3B).

173 As for sleep, we studied ketamine effects on LZ using model fitting of the individual mixed effects
174 models for each cortex. Model selection was performed in this case between linear, quadratic and
175 cubic models using BF (Table 2).
176

A



B



177

178

179 **FIGURE 3. Curves dose-response of the dose of ketamine on cortical dynamic of LZ.** (A)
180 Dose-response curve of subanesthetic doses of ketamine, showing an inverted U-shaped curve
181 only for prefrontal and auditory cortices, with monotonic decrease of complexity with
182 concentration for the other cortices. Each plot represents the sum of the different sessions for each
183 dose of the different cats which have that cortex, therefore the N° of cats is different per cortex.

184 (B) The curves are plotted per cat. It can be clearly seen that the variability in the informational
 185 complexity dynamic per cortex and per cat is evidenced more clearly when plotted individually,
 186 and shows that there is a dissociation of the dose-response and the anatomical location. The doses
 187 are represented in mg/Kg. W, wakefulness; LS, light sleep; SWS, slow wave sleep; REM, rapid
 188 eyes movement sleep; rPf, rostral prefrontal cortex; dlPf, dorsolateral prefrontal cortex; M, primary
 189 motor cortex; S, primary somatosensory cortex; A, primary auditory cortex, Pp, posterior parietal
 190 cortex; V, visual cortex. “L” indicates Left side and “R” right side.

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 192

Cortex	Nº of cats	Model	BF
Right dorsolateral prefrontal	2	linear	
		quadratic *	2.57x10 ⁵
		cubic	288.73
Right rostral prefrontal	1	linear	
		quadratic *	1.70x10 ¹⁴
Right primary motor	2	linear	
		quadratic	0.04
Right primary auditory	2	linear	
		quadratic *	8.25x10 ⁹
Right primary somatosensory	3	linear	
		quadratic	3.0x10 ⁻³
		cubic	0.66
Left primary somatosensory	1	linear	
		quadratic	1.08
Right posterior parietal	4	linear	
		quadratic	1.13x10 ⁻⁴
Left posterior parietal	2	linear	
		quadratic *	46057.12
Right primary visual	2	linear	
		quadratic	1.28
		cubic	3.0x10 ⁻³

193

194 **TABLE 2. Selection between linear and non-linear models among different doses of ketamine**
 195 **for each cortex.** Mixed effects models were formulated for each cortex including the cat as a
 196 random effect when applicable. Bayes Factors (BF) were used to decide between quadratic (U-
 197 shaped), cubic and linear fits. A clear evidence towards a quadratic fit was found for right

198 dorsolateral and rostral prefrontal cortices, right primary auditory cortex and left posterior parietal
199 cortex. The asterisks indicates substantial evidence for a quadratic fit ($BF > 5$).

200

201 ***Informational complexity is not modulated by auditory stimuli under ketamine***

202

203 In 3 cats, modulation by auditory stimuli was studied. Under control conditions without ketamine,
204 an increase in LZ was observed during stimulation in dorsolateral prefrontal (0.66 ± 0.04 to
205 0.70 ± 0.007 , $p < 0.01$, $\eta^2 = 0.044$, $BF = 10619.07$) and auditory (0.63 ± 0.05 to 0.66 ± 0.007 , $p <$
206 0.01 , $\eta^2 = 0.035$, $BF = 75.27$) cortices, whereas the effect on other cortices studied were non-
207 reliable, including right posterior parietal cortex (0.53 ± 0.02 to 0.53 ± 0.001 , $p < 0.01$, $\eta^2 = 0.008$,
208 $BF = 7.0 \times 10^{-5}$), right somatosensory cortex (0.62 ± 0.05 vs 0.62 ± 0.002 with $p = 0.61$, $\eta^2 = 0.005$,
209 $BF = 1.0 \times 10^{-4}$), left somatosensory cortex (0.52 ± 0.005 vs 0.53 ± 0.002 with $p < 0.01$, $\eta^2 = 0.002$,
210 $BF = 0.09$), and left posterior parietal cortex (0.47 ± 0.01 vs 0.48 ± 0.001 , $p = 0.85$, $\eta^2 = 0.001$, BF
211 $= 1.0 \times 10^{-4}$, Figure 4A).

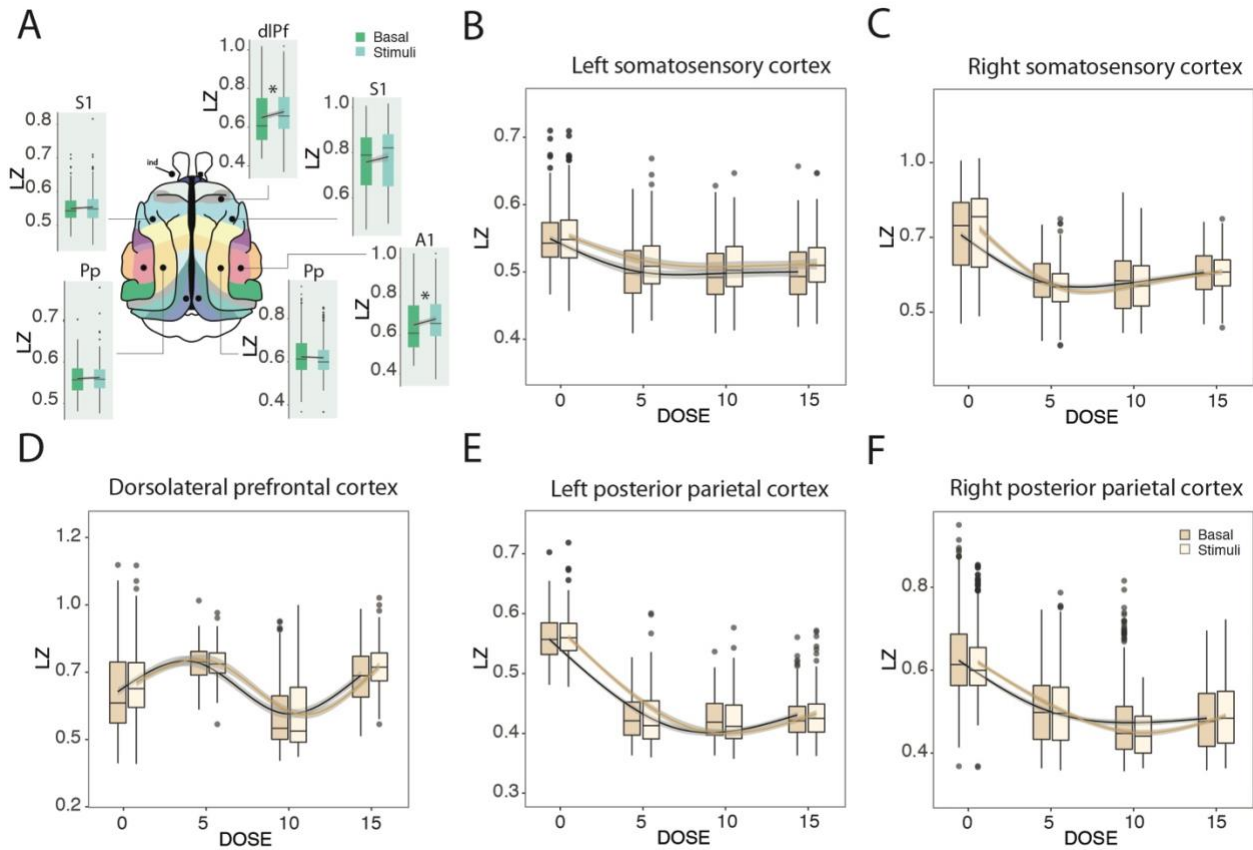
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213 Initially we hypothesized that the increment in complexity under the sensory stimulation versus
214 non-stimulation conditions would be more evident under the effect of ketamine. However, there
215 was no interaction between stimulation and ketamine. For left somatosensory cortex, non-reliable
216 effect was observed during basal conditions in response to the stimuli ($p = 3 \times 10^{-4}$, $BF = 0.09$), and
217 this effect remained unchanged under ketamine, and non-reliable interaction between dose and
218 stimuli was observed ($p = 0.16$; $BF = 3.5 \times 10^{-4}$, Figure 4B). For right somatosensory cortex, where
219 non-reliable increase was evidenced in control conditions (Figure 4A), no reliable interaction was
220 found during ketamine effect ($p = 2.0 \times 10^{-3}$; $BF = 0.012$) with no response to the stimuli ($p = 0.64$;
221 $BF = 1.5 \times 10^{-4}$; Figure 4C). For the prefrontal cortex, where an increase was observed in control
222 conditions, the same effect was found under ketamine ($p = 9.4 \times 10^{-8}$; $BF = 335.92$), with non-
223 reliable interaction between stimuli and ketamine ($p = 0.51$; $BF = 2.5 \times 10^{-8}$; Fig. 4D). For left
224 posterior parietal cortex, non-reliable effect was found under baseline conditions, there was no
225 effect of stimulation under ketamine ($p = 0.85$; $BF = 1.2 \times 10^{-4}$), and the interaction also remained
226 unchanged under ketamine ($p = 0.74$; $BF = 2.5 \times 10^{-6}$; Figure 4E). Finally, for right posterior parietal
227 cortex, no effect was found under basal conditions, there was no change with the stimuli under

228 ketamine ($p = 0.46$; $BF = 7.4 \times 10^{-5}$), and the modulation by stimulation was non-reliable ($p = 8 \times 10^{-3}$;
229 $BF = 8 \times 10^{-3}$; Figure 4F).

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235 **FIGURE 4. Modulation of LZ by auditory stimulation.** (A) The effect of stimulation was shown
236 without ketamine where an increase in LZ was observed during stimulation in dorsolateral
237 prefrontal and auditory cortices. * Statistically reliable ($p < 0.01$; $BF > 5$). dIPf = dorsolateral
238 prefrontal cortex; Pp = posterior parietal cortex; S1 = primary somatosensory cortex; A1 = primary
239 auditory cortex. (B-F) Modulation by stimulation under the effect of ketamine in 5, 10 and 15
240 mg/Kg doses. No stimulation by dose interaction was observed. The doses are represented in
241 mg/Kg.

242

243

244 **Discussion**

245

246 In this work, using LZ as a measure of dynamical complexity on direct intracranial recordings, we
247 studied the effect of subanesthetic doses of ketamine in a dose-dependent manner. Ketamine
248 elicited a diverse set of dynamics, with the lower doses showing the most variable effects. For
249 prefrontal and auditory cortices an increase in LZ was observed from low to medium ketamine
250 dose. However, a decrease was evidenced at the maximum dose, drawing an inverted U-shape
251 dose-effect curve, whereas the opposite effect was observed for other cortices including
252 somatosensory and posterior parietal cortices, where an initial decrease was followed by an
253 increase in complexity at higher doses. Additionally, we also presented auditory stimulation to the
254 cats, which elicited an increase in LZ in prefrontal and auditory cortices, but this effect was not
255 modulated by ketamine. Finally, in the same animals, we studied LZ during sleep, which by
256 contrast show an homogeneous pattern among cortices. We demonstrate that informational
257 complexity in the cortex and thalamus of the cat decreases in light and deep sleep compared to
258 awake states and REM. For most of the cortex, there is only marginal complexity difference
259 between wakefulness and REM sleep. The results were consistent among cats and similar for all
260 the cortices studied and, more importantly, confirm previous results in humans and rats.

261

262 As measures of neural signal diversity are known to be sensitive to conscious level in natural state
263 changes (the sleep-wake cycle), they are also sensitive to the changes in brain dynamics associated
264 with psychedelic and anesthetic states. Specifically, Schartner et al. found increased global neural
265 signal diversity for the psychedelic state induced by ketamine, psilocybin and LSD, as compared
266 to placebo, across a range of measures (Schartner et al., 2017). Other recent MEG and EEG studies
267 have also demonstrated elevated signal diversity induced by canonical serotonergic psychedelics
268 and ketamine (Tagliazucchi et al., 2014; Timmermann et al., 2019).

269

270 From the perspective of its effects on EEG signal diversity, the dissociative NMDA-antagonist
271 ketamine diverges from traditional anesthetics at subanesthetic concentrations, as it induces
272 dissociative states characterized by a maintained or enhanced repertoire of brain states (Li &
273 Mashour, 2019; Schartner et al., 2017). This is in contrast to GABAergic anesthetics such as
274 propofol, which have been shown to degrade sensory integration and attenuate neural signal

275 diversity in a dose-dependent manner (Ferenets et al., 2006, 2007; Ishizawa et al., 2016). While
276 those studies were based on EEG signals that had been low-pass filtered at 55 Hz and lacked
277 cortical dynamics in higher gamma frequencies, Pal et al. have recently demonstrated that this part
278 of the signal is important. Using intracranial EEG data from frontal and parietal cortices of rats
279 receiving ketamine or propofol anesthesia, they demonstrated a reduction in broadband (0.5–175
280 Hz) EEG complexity during ketamine anesthesia that is comparable to that induced by the
281 GABAergic anesthetic propofol. Bandwidth-specific analyses restricted to higher gamma
282 frequencies showed that ketamine anesthesia is distinguished from propofol by suppression of
283 EEG complexity in high gamma frequencies in the range of 65–175 Hz, which previous human
284 studies using scalp EEG could not reveal (Pal et al., 2020). In the present study, by using
285 intracranial electrodes in cats, we were able to study broadband (>0.5Hz) signal complexity .

286

287 Contrary to the apparent convergence of psychedelics (LSD, DMT, psilocybin) reported (Schartner
288 et al., 2017), some of us (González et al., 2021) have shown that the effects of ibogaine, a
289 psychedelic alkaloid, induces high gamma power but are less coherent and less complex than
290 control condition, and similar to natural REM sleep. Although some differences in the complexity
291 measure or animal model may explain the difference, it is key to highlight that the ibogaine local
292 complexity patterns were more consistent than those found in the current study, pointing to a
293 different mode of action between alkaloid, serotonergic and NMDA psychedelics.

294

295 Ketamine's primary mechanism of action is as an NMDA antagonist whose receptors are located
296 quite ubiquitously across the cerebral cortex, as well as subcortically (Conti et al., 1994; Huntley
297 et al., 1994). A differential interaction with various subtypes of NMDA receptors could explain
298 the heterogeneity in cortical response under the effects of ketamine (Zanos et al., 2018). However,
299 the non-NMDA receptor effects of ketamine cannot be discounted, in particular its interactions
300 with opioid receptors and HCN channels (Chen et al., 2009; Zanos et al., 2018; Zhou et al., 2013).
301 Additionally, ketamine may indirectly exert effects through its interaction with other circuits.
302 Previous work reported that subanesthetic doses of ketamine increased the release of not only 5-
303 HT (Amargós-Bosch et al., 2006; López-Gil et al., 2012, 2019), but also noradrenaline (Lorrain et
304 al., 2003) as well as glutamate (Moghaddam et al., 1997) in the medial prefrontal cortex, which
305 may increase signal complexity. At the receptor level, ketamine blocks excitatory NMDA

306 receptors on fast-spiking cortical interneurons more effectively than those on pyramidal neurons.
307 This results in down-regulation of interneuron activity, and decreased GABA release at the
308 interneuron-pyramidal neuron synapse (Homayoun & Moghaddam, 2007; Seamans, 2008). This
309 decrease in inhibitory tone (decreased GABA release) results in markedly excited pyramidal
310 neurons. It has been proposed that this may explain why ketamine is associated with increased
311 cerebral glucose utilization and blood flow (Langsjo et al., 2005; Långsjö et al., 2004), and
312 increased EEG gamma oscillations (Blain-Moraes et al., 2014; Castro-Zaballa et al., 2019; Ferrer-
313 Allado et al., 1973; Lee et al., 2013; Schwartz et al., 1974) and may also help us understand the
314 changes observed in the complexity of the signal. However, our results show a decrease of LZ in
315 somatosensory and posterior parietal cortices after the lowest dose of ketamine (Figure 3B). As
316 both of these cortices process somatosensory information, our results may be due to a reduction in
317 the somatosensory information influx, as one of the main effects of subanesthetic doses of
318 ketamine is analgesia (Zanos et al., 2018).

319
320 Neural diversity, assessed by LZ, is an attractive measure because of simplicity, practical
321 applicability, and consistency with both complexity-based (Tononi et al., 2016; Tononi &
322 Edelman, 1998) and entropy-based (Carhart-Harris, 2018; Carhart-Harris et al., 2014) theories of
323 neural integration and consciousness. The measure is also useful in questions regarding local
324 processing as it is computed at the electrode level, thus was able to demonstrate differential effects
325 in distinct thalamic and cortical brain regions. Indeed, according to the dynamic core hypothesis
326 (Tononi & Edelman, 1998) and subsequent theoretical developments such as Information
327 Integration Theory (Tononi et al., 2016), only certain distributed subsets of the neuronal groups
328 that are activated or deactivated in response to a given task are associated with conscious
329 experience, therefore a large cluster of neuronal groups that together constitute, on a time scale of
330 hundreds of milliseconds, a unified neural process of high complexity can be termed the “dynamic
331 core”. In line with this idea, our results could be interpreted as the prefrontal and auditory cortices,
332 where an increase in LZ was observed under the 5mg dose of ketamine, constituting a part of the
333 “dynamic core”, and somatosensory and posterior parietal cortices playing a different role in neural
334 integration.

335

336 Another useful framework for understanding these results is the neuroscience of arousal, including
337 wakefulness, sleep, circadian rhythms, responsiveness and alertness (Bekinschtein et al., 2009;
338 Brown et al., 2011). Sleep shows a clear change in arousal throughout the day cycle; the intensity
339 of the stimuli needed to wake up a person is maximal in deep sleep and lower in light sleep and
340 REM. This pattern partially mimics the results obtained for informational complexity in this study
341 using local field potentials (LFPs), and several other nonlinear measures such as fractal dimension
342 and other entropy methods (Ma et al., 2018), but not to other measures such as power in different
343 bands and connectivity methods. This finding allows us to interpret that LZ may index
344 behaviourally defined wakefulness, or arousability by stimuli (Bonnet et al., 1978). Although
345 ketamine is used as an anesthetic and creates unconsciousness in high doses and hence can be
346 framed in terms of consciousness as wakefulness and arousal, the effects at lower doses require a
347 multidimensional framework, able to accommodate neurological symptoms (dizziness, slurred
348 speech), mood modulations, and psychedelic experiences. In principle, if ketamine had the classic
349 profile of a sedative, responsiveness would monotonically decrease (Brown et al., 2011) and a
350 similar profile would be expected for molecular and neural measures. However, ketamine has an
351 interesting profile as it belongs to a group of hypnotics that show hallucinatory capacities and an
352 hormetic or U-shaped curve (Calabrese & Baldwin, 2001) in EEG and blood flow (Cavazzuti et
353 al., 1987; Tsuda et al., 2007). The hormesis of the dose response allows for the comparison of not
354 only conscious level in the sense of wakefulness but in terms of contents of consciousness in low
355 ketamine and REM sleep. From humans we know that the likelihood of increased richness in
356 mental content during the sleep-wake cycle occurs during REM (Windt & Noreika, 2011) after a
357 decrease in NREM (U-shaped); and we know that the richness of mental content, including
358 hallucinations, peaks early with ketamine before decreasing into sedation and anesthesia (Powers
359 et al., 2015) (an inverted U-shaped curve). In both cases the higher levels of content agree with the
360 higher (or recovering) levels of informational complexity as measured by LZ (Abásolo et al., 2015;
361 Mateos et al., 2018; Schartner et al., 2015; Schartner, Carhart-Harris, et al., 2017; Schartner,
362 Pigorini, et al., 2017). In this study, we compare the consistency of the complexity in the cortex in
363 sleep and the diversity in the ketamine challenge as two putatively very different mechanisms of
364 reaching a higher level of content in consciousness.
365

366 Recent findings by Mediano et al. (2020) provide strong quantitative evidence on how
367 environmental conditions have a substantial influence on neural dynamics during a psychedelic
368 experience in humans. This work showed how brain entropy is modulated by stimulus
369 manipulation during a psychedelic experience by studying participants under the effects of LSD
370 or placebo, either with gross state changes (eyes closed vs. open) or different stimulus (no stimulus
371 vs. music vs. video). Results showed that while brain entropy increased with LSD in all the
372 experimental conditions, it exhibited largest changes when subjects have their eyes closed,
373 whereas the entropy enhancing effects of LSD were less marked when participants opened their
374 eyes or perceived external stimuli — such as music or video (Mediano et al., 2020). In the present
375 work, we studied the modulation of auditory stimulation on brain complexity in basal conditions
376 and under increasing doses of ketamine in 3 cats using LFP recordings with the hypothesis of
377 observing a higher level of complexity under stimulation. However, only a slight increase in LZ
378 was evidenced during stimulation in dorsolateral prefrontal and auditory cortices, whereas a
379 complete lack of or very weak effect were found in the other cortices studied (Figure 4). This weak
380 effect may be explained by the low relevance of the stimulus, as it failed to catch the attention of
381 the animals, compared to extremely salient or meaningful stimuli such as music or video. Further
382 evidence that stimulation studies should exploit more complex stimuli also comes from a recent
383 study where TMS pulses also failed to increase complexity in low doses of ketamine in humans
384 (Farnes et al., 2020). New experiments using more appropriate stimuli in terms of relevance and
385 salience are needed to better address this hypothesis and further the experimental understanding
386 neural dynamics of information theory, complexity and entropy as the system is modulated
387 pharmacologically.

388
389 Our sleep results are consistent with previous results in humans (Andrillon et al., 2016; M.
390 Schartner et al., 2017), as well as in rats (Abasolo et al., 2015). However, a closer read shows some
391 differences: Andrillon et al. (2016) reported a small but reliable decrease in LZ during REM sleep
392 compared to the waking state, possibly due to participants engaged in a task during the waking
393 state, whereas the participants in the Schartner et al. study were simply at rest with eyes closed and
394 not engaged or externally driven by task or stimuli. In our study the animals were also at rest but
395 with eyes open and showed a decrease in LZ during LS and further decrease in SWS, which was
396 similar for all cortices (Figure 2A) in line with previous findings (Andrillon et al., 2016; M.

397 Schartner et al., 2017). However, a greater variability was evident for REM sleep state where in
398 some cortices LZ was equal in level of complexity to wakefulness whereas in others it was similar
399 to LS or to SWS (Figure 2B). The complexity pattern among sleep stages observed in the cortex
400 was also evidenced the lateral geniculate nucleus (Figure 2A), lending clear convergent evidence
401 to the common effects of informational complexity in the brain beyond the cortex for the sleep
402 wake cycle.

403
404 In summary, our data demonstrate that there is a dose-dependent ketamine effect on neural
405 complexity. An increase in complexity compared to baseline was found for some cortices
406 (prefrontal, motor, auditory and visual) only in the lowest doses, while the higher dose frequently
407 showed the lowest informational complexity. However, a decrease in complexity was also seen in
408 somatosensory and posterior parietal cortex in the low doses. The heterogeneity of the ketamine
409 effects between cats and cortices contrasts with the homogeneity of the changes in complexity seen
410 for different stages of sleep, further highlighting the differences between natural and
411 pharmacologically induced changes in consciousness. The individual and cortical variability in the
412 neural complexity dynamics revealed by ketamine highlights the intricacy of the brain when
413 perturbed by dissociatives and psychedelics, pushing for a multidimensional framework beyond
414 simple arousal and alertness parameters to characterize the change in the states of consciousness
415 from a neuropharmacological perspective.

416

417 **Methods**

418

419 *Animals*

420

421 Five adult cats were used in this study; all of whom were also utilized in a previous report (Castro-
422 Zaballa et al., 2019). The animals were obtained from and determined to be in good health by the
423 Institutional Animal Care Facility of the Faculty of Medicine (University of the Republic,
424 Uruguay). All experimental procedures were conducted in accordance with the Guide for the Care
425 and Use of Laboratory Animals (8th edition, National Academy Press, Washington DC, 2011) and
426 were approved by Institutional and National Animal Care Commissions of the University of the
427 Republic in Uruguay (Protocol N° 070153000089-17). Adequate measures were taken to minimize

428 pain, discomfort or stress to the animals. In addition, all efforts were made to use the minimum
429 number of animals necessary to produce reliable scientific data.

430

431 *Surgical procedure*

432

433 Following general anesthesia, the head was positioned in a stereotaxic frame and the skull was
434 exposed. Stainless steel screw electrodes (1.4 mm diameter) were placed on the surface (above the
435 dura matter) of different cortical areas including prefrontal, primary motor, primary somatosensory
436 and posterior parietal cortices. Note that because the animals were not prepared specifically for
437 this work, we did not analyse the same cortices in all of them. The electrodes were connected to a
438 Winchester plug, which together with two plastic tubes were bonded to the skull with acrylic
439 cement in order to maintain the animals' head in fixed position without pain or pressure. After
440 recovery from surgical procedures, they were adapted to the recording environment for a period
441 of at least 2 weeks.

442

443 *Data acquisition and preprocessing*

444

445 Experimental sessions of 4 h were conducted between 11 a.m. and 3 p.m. in a temperature-
446 controlled environment (21–23 °C). During these sessions (as well as during the adaptation
447 sessions), the animals' head was held in a stereotaxic position by four steel bars that were placed
448 into the chronically implanted plastic tubes, while the body rested in a sleeping bag (semi-restricted
449 condition).

450

451 The ECoG activity was recorded with a monopolar (referential) configuration, utilizing a common
452 reference electrode located in the left frontal sinus. The electromyogram (EMG) of the nuchal
453 muscles, which was recorded by means of an acutely placed bipolar electrode, was also monitored.
454 The electrocardiogram (ECG), by electrodes acutely placed on the skin over the pre-cordial region,
455 and respiratory activity by means of a micro-effort piezo crystal infant sensor were also recorded.
456 Each cat was recorded daily for ~30 days in order to obtain complete basal and treatment data
457 sets.

458

459 Bioelectric signals were amplified ($\times 1,000$), filtered (0.1 - 500 Hz), sampled (1,024 Hz, 2^{16} bits)
460 and stored in a PC using the Spike 2 software (Cambridge Electronic Design).

461
462 Data were obtained after ketamine administration as well as during spontaneously occurring quiet
463 W, LS, NREM sleep and REM sleep (Fig. 1). Five, 10, and 15 mg/kg i.m. of ketamine (Ketonal
464 ®, Richmond Veterinaria S.A.) were administered to five animals. These three doses were
465 administered four times in each animal in different experimental sessions in a counterbalanced
466 order (each animal received 12 doses of ketamine). The recovery time between consecutive
467 ketamine experiments was 72 h. Ketamine (50 mg/ml) was diluted in benzethonium chloride,
468 hydrochloric acid, and water (solution for veterinary use). Basal recordings (without injections)
469 were used as control. Sound stimuli during 300 s were applied 10 min after ketamine injection in
470 three cats. These sound stimuli had the same characteristics as those used to induce active W
471 (Castro et al., 2013).

472
473 Sound stimuli was applied for a period of 300 s; in drug-free recordings and following ketamine
474 administration. In drug-free condition, the stimuli were introduced ~ 30 min after the beginning
475 of the recording sessions. The sound consisted of continuous stimulus of 60–100 dB SPL with a
476 variable frequency of presentation (1–500 Hz, modified at random) in order to avoid habituation
477 (Castro et al., 2013; Torterolo et al., 2003). Sound stimuli during 300 s were also applied 10 min
478 after ketamine injection in three cats.

479
480 For preprocessing, sleep stages were scored off-line by visual inspection of 5-s epochs in Spike2
481 software, where the ECoG and electromyogram (EMG) were displayed simultaneously. In order
482 to analyze LZs during sleep, a total of 300 artifact-free seconds data were selected from each
483 behavioral state. Additionally, to study LZs during the Ketamine effect 300 s duration segments,
484 with and without stimulation, were selected before and after ketamine administration.

485
486 After scoring, for both experiments, the selected epochs were exported to matlab for further
487 preprocessing. The Matlab toolbox eeglab was used to filter the data (0.5-200 Hz band-pass). Each
488 epoch was visually inspected, and those with gross artifacts (e.g. movements) were removed from
489 the analysis.

490

491 *Lempel-Ziv complexity*

492

493 In this study we used Lempel-Ziv (LZ) complexity to compute the complexity of measured neural
494 signals (Lempel & Ziv, 1976). In particular, we used the LZ78 algorithm (Ziv & Lempel, 1978),
495 which corresponds to the standard word-dictionary implementation: given a binary string, the
496 algorithm scans it sequentially looking for distinct structures or “patterns.” The more diverse the
497 binary string, the more patterns are included in the dictionary (a sequence containing only zeros or
498 only ones would lead to the minimal number of patterns being obtained). The total number of these
499 patterns is a measure of signal diversity.

500

501 To compute LZ from our experimental data, the recording of each channel was split into segments
502 of 5120 samples (5s sampled at 1024Hz). Then, to generate a discrete sequence from a real-valued
503 signal X of length T , X is detrended and binarized with a threshold of 0, and the resulting binary
504 sequence is fed to the LZ78 algorithm. Finally, the resulting dictionary length L is normalized as

505

$$C = \frac{\log_2 L}{T} L$$

506

507

508 to yield a measure of complexity C .

509

510 *Statistics*

511

512 One way ANOVA, with Tukey post-hoc test were used to compare LZ between sleep stages per
513 cortex per animal (Fig. 2B) where Cohen’s d was used to address the size of the effect.
514 Additionally, a multilevel approach as well as Bayesian Informational Criterion (BIC) were used
515 to find the most likely explanatory model within the hierarchical model in the group statistical
516 analysis comparing linear, quadratic and cubic models. For sleep study, the state of sleep was used
517 as fixed effect and the cat identity as random effect. The same type of approach was used to study
518 the ketamine effect among different cortices under control and stimulus conditions. In this case the
519 dose and stimulus (if present) were used as fixed effects; and cat identity and session as random

520 effects. The interaction between stimuli and ketamine dose was also included in the model when
521 studying the modulation by stimulus. All models were estimated via restricted maximum
522 likelihood, using the open-source packages lme4 v.1.1-21 (Bates et al., 2015) and lmerTest v.3.1-
523 1 (Kuznetsova et al., 2017) on R v.3.6.1.

524

525

526

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528

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532

533

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