1 The gene expression signature of electrical stimulation in the human brain

Snehajyoti Chatterjee^{1,2}, Yann Vanrobaeys^{1#}, Annie I Gleason^{3#}, Brian J. Park⁴, Shane
A. Heiney^{1,2,5}, Ariane E. Rhone⁴, Kirill V. Nourski⁴, Lucy Langmack^{1,2}, Budhaditya
Basu^{1,2}, Utsav Mukherjee^{1,2}, Christopher K. Kovach^{4,6}, Zsuzsanna Kocsis^{4,7}, Yukiko
Kikuchi⁷, Yaneri A. Ayala⁴, Christopher I. Petkov^{4,7}, Marco M. Hefti⁸, Ethan Bahl³, Jacob
J Michaelson³, Hiroto Kawasaki⁴, Hiroyuki Oya⁴, Matthew A. Howard III⁴, Thomas Nickl-Jockschat^{1,2,3,9,10,11}, Li-Chun Lin^{1,2,12}, Ted Abel^{1,2,3*}

9 Affiliations:

- ¹Department of Neuroscience and Pharmacology, Iowa Neuroscience Institute, Carver
 College of Medicine, University of Iowa, Iowa City, IA, USA
- 12 ²Iowa Neuroscience Institute, University of Iowa, Iowa City, IA, USA
- ¹³ ³Department of Psychiatry, University of Iowa Hospitals and Clinics, Iowa City, IA, USA
- ⁴Department of Neurosurgery, University of Iowa Hospitals and Clinics, Iowa City, IA,
 USA
- ⁵Neural Circuits and Behavior Core, Iowa Neuroscience Institute Carver College of
 Medicine, University of Iowa, Iowa City, IA, USA
- ⁶Department of Neurosurgery, University of Nebraska Medical Center, Omaha, NE, USA
- ¹⁹⁷Biosciences Institute, Newcastle University Medical School, Newcastle upon Tyne, UK
- ⁸Department of Pathology, University of Iowa Hospitals and Clinics, Iowa City, IA, USA
- ⁹Department of Psychiatry and Psychotherapy, Otto-von-Guericke University,
 Magdeburg, Germany
- ¹⁰German Center for Mental Health (DZPG), partner site Halle-Jena-Magdeburg,
 Germany
- ¹¹Center for Intervention and Research on adaptive and maladaptive brain Circuits
 underlying mental health (C-I-R-C), Halle-Jena-Magdeburg, Germany
- ¹²Iowa NeuroBank Core, Iowa Neuroscience Institute, Carver College of Medicine,
 University of Iowa, Iowa City, IA, USA

- [#]Authors contributed equally
- 31 *Corresponding author:
- 32 ted-abel@uiowa.edu
- 33

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36 Abstract

37 Direct electrical stimulation has been used for decades as a gold standard clinical tool to map cognitive function in neurosurgery patients¹⁻⁸. However, the molecular impact of 38 electrical stimulation in the human brain is unknown. Here, using state-of-the-art 39 transcriptomic and epigenomic sequencing techniques, we define the molecular 40 changes in bulk tissue and at the single-cell level in the human cerebral cortex following 41 42 direct electrical stimulation of the anterior temporal lobe in patients undergoing neurosurgery. Direct electrical stimulation surprisingly had a robust and consistent 43 44 impact on the expression of genes related to microglia-specific cytokine activity, an effect that was replicated in mice. Using a newly developed deep learning 45 computational tool, we further demonstrate cell type-specific molecular activation. which 46 47 underscores the effects of electrical stimulation on gene expression in microglia. Taken 48 together, this work challenges the notion that the immediate impact of electrical stimulation commonly used in the clinic has a primary effect on neuronal gene 49 expression and reveals that microglia robustly respond to electrical stimulation, thus 50 51 enabling these non-neuronal cells to sculpt and shape the activity of neuronal circuits in 52 the human brain.

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64 **Main**

65 Electrical stimulation of the human brain has become an indispensable clinical tool for diagnosis and therapy¹⁻⁸. In neurosurgical patients, cortical stimulation is the gold 66 standard diagnostic tool to identify the location of functionally relevant brain regions that 67 are critical for speech, language, or motor function^{9,10} to avoid potential damage to 68 these regions during surgery. Transient changes in gene expression underlie the ability 69 70 of the brain to adapt to changes in the environment or brain perturbation, such as electrical stimulation. Dynamic transcriptomic patterns are essential for cognition¹¹⁻¹³, 71 affective processing¹⁴, addiction¹⁵, and the initiation of behaviors^{16,17}. Thus, electrical 72 73 stimulation likely exerts neural effects via alterations in gene expression. Although 74 diagnostic and therapeutic brain stimulation is conducted daily in thousands of patients 75 worldwide, the molecular impact of electrical stimulation in the human brain remains 76 unknown. Recent advances in molecular sequencing techniques have revolutionized 77 transcriptomics by enabling mapping of changes in transcription and chromatin accessibility in single cells¹⁸. However, these experiments have only been conducted in 78 rodents^{19,20} and neurons derived from human induced pluripotent stem cells²¹. A recent 79 pioneering study utilized a single nuclei molecular approach to link human brain 80 transcriptomics signatures with oscillatory signatures of memory consolidation in 81 82 epileptic patients undergoing an episodic memory task, but these gene expression measures were obtained many days after assessment of intracranial recording data 83 during the task^{22,23}. Therefore, research is needed to evaluate the more immediate 84 effects on single-cell early gene expression, such as tens of minutes after neural system 85 86 perturbation with electrical stimulation. Studies investigating gene expression profiles 87 after electrical stimulation are particularly difficult to perform in human brains due to the requirement that tissue is sampled at precise temporal windows before and after 88 stimulation. 89

90 Here, in patients undergoing clinical neurosurgical resection of pathological 91 epileptogenic sites in the mesial temporal lobe, samples from tissue in the anterior

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92 temporal lobe (ATL) that required clinical resection to access the deeper epileptogenic 93 site for treatment were obtained before and after electrical stimulation. By analyzing 94 gene expression at baseline and minutes after stimulation from the same patient, along with our analysis of samples taken from patients that did not receive electrical 95 96 stimulation, we are able to distinguish genes responsive to electrical stimulation from 97 the genes altered nonspecifically by surgical or disease-based factors that would be 98 stable across samples. Our parameters for electrical stimulation were based on those regularly used in clinical mapping^{24,25}; moreover, we referenced the human results to a 99 100 similar electrical stimulation paradigm in mice, revealing a similar transcriptional profile. 101 Finally, using a single-nuclei multi-omics approach, we provide insight into the cell-type-102 specific transcriptomic and epigenomic responses to electrical stimulation in the human 103 brain, highlighting important effects beyond neurons, including microglia. This study 104 provides fundamental insights into changes in cell-type specific molecular signatures in 105 the human cortex after electrical stimulation, laying the groundwork for a molecular 106 understanding of the impact of this fundamental tool in clinical neurosurgery, diagnosis, 107 and treatment.

108 **Results**

109 We recruited eight adult neurosurgical patients undergoing surgical resection of seizure 110 foci following clinical monitoring to treat epilepsy. The patients provided informed 111 consent to take part in this research and were informed that tissue samples would only 112 be taken from tissue that would require clinically resection for treatment. The patient participants underwent an anterior temporal lobectomy for access to a medial temporal 113 114 lobe epileptogenic site, during which samples were resected from the neocortex and 115 processed immediately after removal. The participants were evenly distributed between 116 two experimental paradigms. In the first group with the electrically stimulated paradigm (4 participants), a baseline tissue sample was resected from the ATL after the 117 118 craniotomy and durotomy, exposing the temporal lobe. Then, an adjacent region of the 119 cortex was stimulated using a stimulation protocol commonly used for bipolar electrical stimulation (50 Hz) for 2 minutes^{24,25} And resected approximately thirty minutes later 120 $(32.5 \pm 11.2 \text{ min})$, a sample was taken from the stimulated region (Fig. 1a; 121 122 Supplementary Fig. 1a, Supplementary data 1). In the second group with the

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123 unstimulated paradigm (4 participants), a sample was taken at baseline, and then a 124 second sample was taken without stimulation about 30 minutes later $(37.3 \pm 12.5 \text{ mins})$ 125 Fig. 1b; Supplementary Fig. 1a). The pre-stimulation sample and the unstimulated paradigm groups serves as controls for changes in baseline gene expression and 126 127 changes in gene expression occurring as a result of the surgical procedure. Among the 8 participants, five were under general anesthesia, whereas three participants were 128 129 awake during the surgical procedure. All the resected tissue samples from both the 130 stimulated and unstimulated paradigms were taken well outside the seizure focus (27.8 ± 7mm from the border of seizure focus, **Supplementary Fig. 1b**). 131

132 Tissue samples were first subjected to bulk whole-transcriptome RNA sequencing 133 (RNA-seq) to identify differentially expressed genes following electrical stimulation. Bulk 134 RNA-seq analysis from the stimulated and corresponding baseline samples revealed 135 124 differentially expressed genes following electrical stimulation, with 112 up-regulated 136 and 12 down-regulated genes (Fig. 1c; Supplementary data 2). Enrichment network 137 analysis was used to identify the pathways most represented among the differentially 138 expressed genes in the stimulated paradigm. The top significant pathways were 139 enriched with genes involved in cytokine activity, DNA-binding transcription activator activity (RNA Pol II), cytokine receptor binding, DNA-binding transcription activator 140 141 activity, nuclear glucocorticoid receptor binding, chemokine activity, CCR chemokine receptor binding, RNA pol II specific DNA binding, chemokine receptor binding, and 142 143 protein phosphatase activity (Fig. 1d). Notably, these genes were not significantly 144 enriched for previously identified genes induced in regions showing seizure activity in the human brain (Supplementary Fig. 2)²⁶. 145

Bulk RNA-seq analysis from the unstimulated paradigm comparing unstimulated samples with corresponding baseline samples identified differential expression of only 16 genes, with nine up-regulated and seven down-regulated genes (Fig. 1e, Supplementary data 3). Only one gene, *NR4A3*, was found to be differentially expressed in both the stimulated and unstimulated groups (Fig. 1g). The lack of overlap between the differentially expressed genes observed in our stimulated and unstimulated groups suggests that the changes in gene expression that we see following electrical

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stimulation do not reflect disease state or surgical effects such as craniotomy, brain
temperature, and anesthesia, but reflect changes due to electrical stimulation.

155 We used a similar electrical stimulation paradigm in the mouse non-primary auditory cortex to investigate whether these gene expression changes are unique to human 156 samples and to determine whether they are related to disease state ^{26,27}. Bulk RNA-sea 157 158 was performed from samples collected 30 minutes following electrical stimulation and 159 unstimulated samples collected at the same time from the contralateral side (Fig. 2a). 160 Bulk RNA-seq identified 44 upregulated and 107 downregulated genes (Fig. 2b, 161 **Supplementary data 4).** Pathway analysis identified from the upregulated genes were enriched for cytokine activity, chemokine activity, cytokine receptor binding, and CCR 162 chemokine receptor binding. (Fig. 2c). We also found upregulated pathways linked to 163 164 transcription and post-transcription regulatory pathways, such as mRNA 3'-UTR binding, 165 mRNA3'-UTR AU-rich region binding, and transcription co-repressor activity. Comparing 166 the stimulation-responsive genes from human bulk RNA-seg and mouse bulk RNA-seg revealed a significant correlation (R^2 =0.0415, p-value <0.00001), especially cytokine-167 168 related genes were commonly upregulated following electrical stimulation in both 169 humans and mice (Fig. 2d). Some of the pathways commonly altered by electrical stimulation between humans and mice are cytokine activity, chemokine activity, 170 171 chemokine receptor binding, cytokine receptor activity, and the CCR chemokine 172 receptor binding. Some common genes enriched in these chemokine and cytokine-173 related pathways are CCL3, CCL4, CXCL1, IL1A, and TNF. We further validated the 174 expression of the cytokine activity-related genes Ccl3 and Ccl4 using qPCR analysis 175 from mouse brain following electrical stimulation (Fig. 2e). The differentially expressed genes following electrical stimulation in the human cortex correlated significantly with 176 learning-induced genes in the mouse cortex (R^2 =0.2027, p-value <0.00001). Immediate 177 early genes such as Arc, Fos, Egr1, Nr4a1, and FosB were upregulated in both 178 179 datasets (Supplementary Fig. 3). However, learning did not induce expression of cytokine activity related genes in mouse cortex. Thus, our findings of gene expression 180 181 signature in response to electrical stimulation in the human and mouse brain reveal a 182 molecular signature that partially conserved across species.

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Next, we investigated the cell types exhibiting differential gene expression following 183 electrical stimulation in the human brain by utilizing single nuclei multiomics (RNA and 184 185 ATAC) on samples from the stimulated paradigm from 3 participants (Fig 3a, 186 Supplementary Fig. 4). Cell clustering analysis identified seven major cell types in our 187 samples-excitatory neurons, microglia, VIP-Sncg-Lamp5 inhibitory neurons, PValb-Sst 188 inhibitory neurons, oligodendrocytes, oligodendrocytes precursor cells (OPC), and 189 astrocytes (Fig. 3b). To determine the differentially expressed genes in each cell type following electrical stimulation, we performed a pseudobulk RNA-seq analysis due to its 190 191 superior performance for detecting differential expression in single-cell RNA-sequencing analyses²⁸ and the stringent nature of this analysis. The pseudobulk analysis revealed 192 193 that the microglia displayed the highest differential gene expression of all cell types with 194 31 upregulated genes (Fig. 3c, Supplementary data 5). Genes related to cytokine activity (CCL3, CCL4, CCL3L1, and IL1B,) were upregulated exclusively within 195 196 microglia. Five genes were upregulated in oligodendrocytes (FOS, HSPA1A, JUNB, 197 GADD45B, and FOSB), and only two genes were differentially expressed (one upregulated: LHFPL3, and one downregulated: ROBO2) in astrocytes (Supplementary 198 199 data 5). Surprisingly, no differentially expressed genes were detected within the 200 neuronal cell types (excitatory and inhibitory neuronal cell types). Comparing the 201 upregulated genes in microglia with human and mice bulk RNA seg revealed 5 genes 202 (Fos, Dusp1, Ccl3, Ccl4, Zfp36) that are commonly upregulated following electrical 203 stimulation (Fig. 3d). Atf3, Cd83, Egr3, Nr4a1, II1b, McI1, Nedd9, Spp1, Nfkbid, Rgs1, 204 and Ccl311 were found to be human microglia-specific genes induced by electrical 205 stimulation that do not change in the mice cortex (Fig 3d). Pathways enriched among 206 the upregulated genes in microglia included cytokine-related pathways such as cytokine 207 activity, chemokine activity, CCR chemokine receptor, and cytokine receptor binding (Fig. 3e). Microglia exhibited activation of transcription regulatory pathways such as 208 209 DNA-binding transcription activator activity and RNA polymerase-specific DNA binding 210 TF binding (Fig. 3e). These results reveal that electrical stimulation increases the expression of genes that are involved in cytokine activity and transcription regulation in 211 212 microglia.

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213 To investigate if electrical stimulation has an impact on the epigenome, we assessed chromatin accessibility using the single nuclear assay for transposase-accessible 214 chromatin with sequencing (snATAC-seq)²⁹ on human samples from the stimulated 215 216 paradigm. We focused on promoter accessibility (-2 to +2 kb from TSS) to analyze 217 changes in chromatin accessibility in snATAC-seq data (Supplementary data 6). The 218 differentially accessible regions that were either enriched or depleted following electrical stimulation in microglia significantly correlated (R²=0.0262, p-value<0.002) with the 219 220 microglia gene expression from snRNA-seq (Fig. 3f), suggesting that the genes that 221 exhibit induced expression following electrical stimulation also exhibit increased 222 chromatin accessibility. Genes related to transcription regulators, such as NR4A1 and 223 FOS, showed a positive correlation between transcriptome and chromatin accessibility. 224 We also found enriched promoter accessibility and increased gene expression for CCL4 225 in microglia following electrical stimulation (Fig 3g). A recent study demonstrated that 226 ambient RNA from neurons may contaminate non-neuronal cells in single-nuclei 227 transcriptomic data³⁰. Our snRNA seq data revealed the upregulation of several IEGs, 228 such as FOS, EGR3, JUNB, and NR4A1 within microglia. These genes are often seen 229 to be upregulated in neurons following neuronal activation. However, our snATAC-seq 230 data from the same samples also showed increased chromatin accessibility in microglial 231 populations within IEG promoters, thus suggesting that the microglia-specific IEG 232 expression is not due to neuronal ambient RNA contamination. Lastly, we analyzed 233 genome-wide transcription factor (TF) binding motifs from the upregulated peaks following electrical stimulation in microglia using chromatin accessibility data. 234 235 Genomewide TF motif analysis revealed enrichment of binding motifs for ELK3, ELK1, 236 YY2, NRF1, HINFP, and ELK4 (Fig. 3h, Supplementary data 7). Interestingly, ELK4, a 237 transcription factor downstream of MAPK signaling, is predicted to positively regulate the expression of Cc/3 and Cc/ 4^{31} . Thus, our sn-ATAC-seq results reveal a signature of 238 239 transcription factor motifs and chromatin accessibility underlying electrical stimulation-240 induced gene expression in human microglia.

Our experimental design included internal controls (baseline/unstimulated samples) from each participant, enabling us to investigate the impact of electrical stimulation across individual participants. Therefore, we next examined the cell type-specific

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response to electrical stimulation from within participants using our newly developed 244 deep-learning computational tool, NEUROeSTIMator³², to map the active population of 245 246 cells following electrical stimulation in each participant, comparing stimulated samples 247 with corresponding baseline controls. We estimated cell activity from the single nuclei 248 RNA data from individual populations of cells with either baseline or stimulated from 249 each participants (Fig. 4a). We used paired linear regression with bootstrap sampling to 250 assess differences in estimated activity between conditions for each cell type (Fig. 4b). 251 Interestingly, we observed that the microglial cluster showed a significant increase in 252 activity scores following stimulation, and this effect was consistent across all three 253 participants [β = 0.100301, p-value < 2e-16] (**Fig. 4c**). This data supports our bulk RNA 254 seq and snRNAseq data. However, we observed variability in electrical stimulation-255 mediated activation of excitatory neurons across participants. We detected significant 256 differences in activity when using all three participants [$\beta = -0.053920$, p-value < 2e-16]. One pair of samples showed a significant increase in activity [$\beta = 0.032528$, p-value = 257 258 0.000807], while 2 out of 3 pairs showed a significant decrease in activity [β = -259 0.022086, -0.083254, p-values = 1.58e-8, 4e-9] (**Fig. 4c**). Variability in activity estimates 260 was also observed in inhibitory neuronal sub-types across participants (Fig. 4c).

261 To further investigate the epigenomic signature of cellular activity following electrical 262 stimulation, we estimated activity from snATAC-seq data using a similar approach (Fig. 263 5a). Activity was estimated for cells using gene-level counts in promoter regions and 264 paired linear regression with bootstrap sampling was used to assess activity between 265 baseline and electrical stimulation (Fig. 5b). Consistent with the transcriptomics data, 266 microglia showed a significant increase in promoter-informed activity estimates, and this 267 effect was consistent across all three participants [β = 0.060977, p-value < 2e-16] (Fig. 268 **5c**). Neuronal clusters showed variability in activation in activity when considering all 269 three participants [$\beta_{\text{Excitatory}} = 0.016163$, p-value_{Excitatory} = 0.000247; $\beta_{\text{Inhibitory}} = 0.028302$, 270 p-value_{Inhibitory} = 2.283e-7]. Only one participant showed a significant increase in the 271 excitatory[$\beta = 0.04712$, p-val = 0.000314] and inhibitory neurons (Sst-Pvalb) [$\beta =$ 0.08863, p-value = 0.00201]. These findings suggest that microglia exhibit robust 272 273 transcriptional and epigenomic responses to electrical stimulation and that neuronal cell types exhibit greater variability between patient participants. 274

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275 Discussion

276 Electrical stimulation is a model for studying the human cortex with protocols that are 277 used clinically for mapping the function of specific brain regions. Here, we take a molecular approach, defining the transcriptomic and epigenomic signatures of electrical 278 279 stimulation at the single cell level in the human cortex in neurosurgical patient 280 participants. Researchers have correlated human brain transcriptomics with prior recorded oscillatory signatures of memory consolidation^{22,33}, but have not previously 281 282 examined the changes in gene expression that are rapidly and directly driven by 283 electrical stimulation. Most work thus far has emphasized the role of neurons in 284 responding to external stimuli, but our work reveals the critical role of microglia in 285 sculpting the activity and function of brain circuits. In microglia, genes related to 286 cytokine signaling showed the greatest induction pattern, and this was conserved 287 across species. As previously hypothesized, we observed the induction of activity-288 dependent genes in our bulk RNA sequencing results, but to our surprise, single nuclei 289 multiomics experiments revealed pronounced microglia-specific transcriptomic 290 activation following electrical stimulation that was supported by analysis using NEUROeSTIMator³², a deep-learning computational model. Identifying a microglial 291 292 transcriptomic response following electrical stimulation represents an important 293 conceptual advance in our understanding of the impact of this common form of electrical 294 stimulation used clinically.

295 Microglia are critical modulators of neuronal function, acting to suppress excessive 296 activity by inhibiting surrounding neurons, including excitatory neurons³⁴. Although classified as non-excitable cells, microglia exhibit electrophysiological stimulus-297 298 response features, and changes in their membrane potential affect crucial microglial functions including phagocytosis, chemotaxis, and cytokine release³⁵. Our analyses 299 300 identified critical molecular components within microglia that change with electrical 301 stimulation, including chemokine-encoding genes such as CCL3 and CCL4, which act to 302 alter microglial motility, influence neuronal-microglial interactions, and shape neuronal connectivity^{34,36-39}. These chemokines are ligands for C-C chemokine receptor type 5 303 (CCR5), which regulates neuronal excitability and memory allocation in the 304 hippocampus⁴⁰. Previous rodent studies have shown that neuronal activation using 305

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chemogenetic approaches leads to distinct gene expression changes in microglia, 306 including genes encoding chemokines³⁴. Our work also revealed that genes encoding 307 308 transcription factors such as FOS and NR4A1 are induced in microglia in response to 309 electrical stimulation. Further, we observed increased promoter accessibility at these 310 genes, underscoring the conclusion that these genes are transcriptionally upregulated in 311 microglia after stimulation and broadening the potential functional impact of these 312 immediate early genes beyond their more frequently studied role in neurons. Indeed, NR4A1 acts in neurons as a transcription factor essential for memory consolidation^{41,42} 313 314 and functions in microglia as a molecular rheostat, contributing to the maintenance of a threshold that prevents microglial activation⁴³. Furthermore, our study of chromatin 315 316 accessibility defines a cell-type-specific signature of transcriptional motifs driving these processes that includes motifs for ELK4 and NRF1, which have been identified in other 317 318 studies as transcriptional motifs associated with the differential regulation of molecular pathways in specific subsets of microglia^{44,45}. 319

320 This study represents a rare opportunity to define the molecular impact of electrical 321 stimulation in patients undergoing anterior temporal lobectomy due to therapy refractory 322 seizures. Although presenting a unique opportunity, it is important to note the caveats 323 that come from this research carried out in a setting focused on clinical efficacy in 324 patients that have epilepsy. Thus, we can not control all of the parameters as we might 325 in an experiment carried out in a model organism, although we have worked to address 326 three potential limitations of our study. First, our experimental design includes a group 327 that does not receive electrical stimulation, thus controlling for the impact of the surgical procedure, craniotomy, brain temperature changes, and levels of anesthesia when 328 329 present. We also show that electrical stimulation in a mouse model, in which we can 330 control many additional parameters, gives rise to similar molecular changes. Importantly, our mouse experiment included the use of an inactive electrode, thus 331 332 controlling for mechanically induced changes in gene expression. Second, the 333 stimulation pattern used is safe and efficacious for brain mapping in human patients but is not optimized to modulate activity in specific circuits and cell types in the brain. 334 Indeed, the stimulation protocol used leads to the functional disruption of brain regions¹, 335 336 and thus, we may expect to see reductions in activity-dependent gene expression in

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excitatory neurons, as seen in two patients in our NEUROeSTIMator analysis. Third, it goes without saying that these patients have severe therapy refractory epilepsy. However, the resected samples were taken a considerable distance from the seizure foci (see Suppl. Fig. 1b), our gene induction signatures do not resemble those seen in actively spiking tissue²⁶, and our molecular changes are seen in a mouse model that does not exhibit seizures.

343 Despite these caveats, we were able to identify a reliable cell-specific signature of the 344 impact of electrical stimulation in the living human brain. This study demonstrates for the 345 first time a unique transcriptomic and epigenomic signature in microglia following direct 346 electrical stimulation, representing a conceptual advance in our understanding of how 347 the brain responds to an important clinical tool used to map brain function. These 348 findings have the potential to inform the clinical practice of diagnostic and therapeutic 349 brain stimulation. Clinical and translational research has focused mainly on the 350 immediate effects of stimulation in neuronal populations, while microglia receive little attention, often being discussed with regard to neuroimmunological aspects⁴⁶. Although, 351 352 for clinical mapping, clinicians rely on the immediate neuronal effects on behavior (e.g., 353 speech arrest), there is growing evidence that therapeutic effects of electrical stimulation can accumulate gradually over time⁴⁷. We could not test a range of 354 355 stimulation frequencies and intensities or examine a broad range of brain areas 356 because of clinical limitations, but our results raise the intriguing hypothesis that 357 microglia, not just neurons, shape plasticity after repetitive stimulation. Our work further 358 highlights that cytokine and chemokine-based mechanisms enable microglia to respond 359 to electrical stimulation and sculpt circuit function, making these potential targets to modify circuit activity with pharmacological approaches. Microglia exist in different 360 subpopulations depending on transcriptional state and age^{48,49}, and our epigenetic work 361 362 has revealed particular transcriptional motifs that mark microglial subtypes important for 363 cortical plasticity. It will be especially interesting in future work to extend our studies to 364 stimulation protocols used therapeutically as well as natural stimuli to probe the critical 365 role of microglial signaling in mediating the response of the human brain to experience.

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382 Author contributions

383 S.C., M.A.H., and T.A. conceived the project, designed the experiments, and interpreted 384 the data. S.C. and T.A. wrote the manuscript with assistance from C.I.P., L.L., and 385 T.N.J., and inputs from other co-authors. H.K. performed human surgery and electrical 386 stimulation and resected the tissue. A.E.R and C.K.K. processed with IRB and MRI 387 imaging registration. H.O, M.A.H, T.N-J, and K.V.L provided advice on surgical 388 procedures and electrical stimulation. M.M.H. provided a pathological assessment. 389 L.C.L., K.V.N., and B.J.P. collected the tissue and images during resection. S.C., B.J.P., 390 L.C.L., U.M., L.L. performed human or mouse tissue samples experiments. C.I.P. and 391 Y.A.A. performed distance to the seizure site analysis. U.M. and L.L. performed mouse brain imaging and data analysis. Y.V., A.I.G., and B.B. performed the bioinformatics 392 393 analysis with inputs from E.B. and J.J.M. S.A.H. performed the mouse surgery and electrical stimulation experiments. K.V.N, C.K.K, Z.K. Y.K, C.I.P, M.H, J.J.M, E.B, U.M, 394 395 and T.N.J. provided input on data analysis and interpretation. All authors discussed the 396 results and commented on the manuscript.

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397 Competing interests

398 T.A. is a scientific advisor to Aditum Bio and Radius Health and serves on the scientific

advisory board of Embark Neuro. The other authors declare no conflicting interests.

400 Figure legends

401 Figure 1. Electrical stimulation of the human cortex induces changes in genes associated with cytokine activity and transcription regulation. a, b. Schematics of 402 stimulated (a) and unstimulated (b) paradigms in the human anterior temporal lobe. For 403 both paradigms sample A was taken at T = 0 minutes then an adjacent sample B was 404 405 taken at either 30 min after stimulation (a, stimulated paradigm, 4 participants) or 30 min after sample A was taken (b, unstimulated paradigm, 4 participants). c. Volcano plot 406 407 showing gene expression changes in the stimulated samples. The most significant 408 genes (FDR < 0.05) are labeled in red (upregulated) or blue (downregulated). d. Cnet 409 plot showing pathway enrichment analysis of the genes significantly (FDR < 0.05) 410 differentially expressed in the stimulated samples. e. Volcano plots showing gene 411 expression changes in the unstimulated samples. The most significant genes (FDR < 0.05) are labeled in red (upregulated) or blue (downregulated). 412

413 Figure 2. Electrical stimulation of mouse cortex induces changes in genes associated with cytokine activity and transcription regulation. a. The mouse 414 auditory cortex was stimulated, tissue was collected 30 min later, and the contralateral 415 auditory cortex was obtained as baseline control. b. Volcano plot showing gene 416 417 expression changes after stimulation. The most significant genes (FDR < 0.05) are labeled in red (upregulated) or blue (downregulated). c. Cnet plot showing pathway 418 enrichment analysis of the genes significantly (FDR < 0.05) affected by electrical 419 420 stimulation. **d.** Quadrant plot comparing genes induced by electrical stimulation in mice with genes induced by electrical stimulation in human brain. Genes upregulated in both 421 422 mice and human brains after stimulation are labeled. The size, opacity, and color 423 intensity of each data point denotes the minimum false discovery rate value for a gene 424 between each transcriptomic datasets. e. qPCR analysis of the genes related to cytokine activity comparing stimulated versus baseline controls in mice cortex. 425 426 n=7/group.

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Figure 3. Single nuclei multiomics reveal cell type-specific transcriptomic and 427 epigenomic changes following electrical stimulation. a. Single nucleus multiomic 428 429 experimental approach. b. UMAP shows the specific cell types from each cluster. c. Volcano plot showing differentially expressed genes in microglia using pseudobulk 430 431 analysis comparing stimulated vs baseline samples from human brain (FDR<0.1). d. Upset plot comparing DEGs in mouse, human bulk RNA seq, and human microglia 432 433 snRNA-seq after electrical stimulation. e. Cnet plot showing pathway enrichment analysis of the genes significantly (FDR < 0.1) affected by electrical stimulation in 434 435 human microalia. f. Quadrant plot comparing DEGs from snRNA-seq following electrical stimulation in human microglia and genes with open chromatin accessibility 436 437 from snATAC-seq in microglia following electrical stimulation. Genes upregulated in both mice and human brains after stimulation are labeled. g. Quadrant plot comparing gene 438 439 expression changes (log10(FDR)*log2fold-change) between snRNA-seg (x-axis) and 440 snATAC-seg (y-axis) data in microglia after stimulation. Genes in red in the top-right corner of the plot are significantly upregulated at the gene expression level and exhibit 441 442 increased chromatin accessibility. The dotted lines represent FDR thresholds of 0.01. 443 The dashed line represents the linear regression applied to this guadrant plot. **g.** DNA motifs are overrepresented in the set of peaks differentially accessible in microglia after 444 445 stimulation. Motifs are ranked based on significance from the most significant left to 446 right. h. The coverage plot shows ATAC peaks at the CCL4 locus. Each track represents 447 a normalized chromatin accessibility signal from the ATAC assay for each cell type and 448 condition (baseline or stimulated).

Figure 4. NEUROeSTIMator for snRNA-seq. Cell-type specific differences in activity estimates. a. UMAP of cells representing the activity estimated from the RNA gene counts in the baseline and stimulated conditions. Darker blue points are cells with higher estimated activity. b. The bootstrap distribution of the difference between activity estimates (stimulated – baseline) by cell type. Horizontal bars indicate the 95% confidence interval of the mean difference in activity estimate. c. Activity estimates in selected cell types in baseline (green) and stimulated (orange) conditions.

Figure 5. NEUROeSTIMator for snATAC-seq. Cell-type specific differences in
 activity estimates. a. UMAP of cells representing the activity estimated from the ATAC

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458 counts in the baseline and stimulated conditions. Darker blue points are cells with
459 higher estimated activity. **b.** The bootstrap distribution of the difference between activity
460 estimates (stimulated – baseline) by cell type. Horizontal bars indicate the 95%
461 confidence interval of the mean difference in activity estimate. **c.** Activity estimates in
462 selected cell types in baseline (green) and stimulated (orange) conditions.

463 Materials and Methods:

Patient participants: The study participants were 8 adult neurosurgical patients (6 female, 2 male, age 19-63 years old, median age 41 years old) with medically refractory epilepsy. The patients were undergoing surgical resection of seizure foci following noninvasive electroencephalography (EEG) or invasive iEEG monitoring. All patients were diagnosed with intractable epilepsy. All patients underwent ATL resection surgery for epilepsy of various etiologies (Extended data fig 1).

470 Two of the patient participants, L472 had a cavernoma and L475 had a dysembryoplastic neuroepithelial tumor. X out of Y participants were required to be 471 472 awake during the surgical resection, and the rest were anesthetized. All patient 473 participants except #6 were non-smokers. Participant's age, sex, surgery, and awake or 474 sedative information were recorded (Extended Data Table 1). All participants were 475 native English speakers, 7 were right-handed, 1 was left-handed, and all had left 476 language dominance as determined by Wada tests. All participants underwent audiometric evaluation before the study, and none were found to have hearing deficits 477 478 or word recognition scores deemed sufficient to affect the findings presented in this 479 study. The vision was self-reported as normal or corrected to normal with glasses or 480 contact lenses. As determined by standard neuropsychological assessments, cognitive function was in the average range in all participants. 481 Research protocols were 482 approved by the University of Iowa Institutional Review Board (IRB 201910791, 201911084) and the National Institutes of Health. Written informed consent was 483 obtained from all participants. 484

485 Procedure: Surgery was performed awake, under general anesthesia or monitored 486 anesthetic care. Standard craniotomy was performed by the same senior epilepsy 487 neurosurgeon in all patients to reach the epilepsy focus for resection, which involved the

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488 anterior and medial temporal lobe in all patients except one who had a temporal 489 encephalocele and surrounding anterolateral temporal cortical focus. Before 490 neurosurgical ATL resection, cortical tissue from the anterior temporal lobe was sampled by the neurosurgeon and handed over to the research team for analysis. The 491 492 experimental condition was defined by an electrical "stimulation paradigm" and a control "no-stimulation paradigm". The location of the sampled tissue is plotted on anatomic 493 494 brain reconstructions (Extended data fig 1). The 30 minutes in between the control and 495 experimental samples was clinically required for the clinical EEG team to record interictal activity with surface recording grids placed gently on the brain surface. 496

There were 4 participants (1 iEEG and 3 EEG patients) who underwent the "stimulation paradigm" and 4 (2 iEEG and 2 EEG patients) who underwent the "no-stimulation paradigm" **(Extended Data Table 5)**.

500 In the "stimulation paradigm", a baseline sample was obtained from the anterior 501 temporal cortex that would be resected in the planned surgical resection. The area 502 directly adjacent to where the baseline sample was collected was stimulated with direct 503 bipolar electric stimulation (50 Hz frequency, 0.2 ms pulse duration, 2 min stimulation duration, and 10 V voltage). The stimulated area was then sampled after a period of 30 504 minutes to allow for gene expression^{50,51}. In the "no stimulation paradigm", no direct 505 506 electric stimulation was performed and the area directly adjacent to the baseline sample 507 was collected 30 minutes after initial baseline sampling.

After sampling the tissue samples were immediately placed in a sterile container on dry ice. The average weight of the baseline sample was 88.0 ± 33.2 mg (mean, standard deviation) and the adjacent sample was 113.4 ± 62.7 mg. After the collection of all samples in this fashion, they were weighed and transferred to a freezer at -80° C for storage until further testing.

513 Sample Localization to MNI space: All samples were from the same cortical region.
514 Intraoperative photos of the sample sites were obtained during the time of surgery.
515 Using patient participant matched preoperative T1 sequence MRI, the sample sites
516 were mapped onto their anatomic brain reconstructions (Extended data fig. 1). They
517 were also mapped onto MNI space coordinates.

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Human cortical tissue RNA extraction, library preparation, and sequencing: Total 518 RNA was extracted from sampled human brains using miRNeasy Mini Kit (Qiagen, CA, 519 520 USA). The tissue samples were homogenized in QIAzol (Qiagen, CA. USA) 521 stainless steel beads (Qiagen, CA, USA). Chloroform was then used for phase 522 separation. RNA containing an aqueous layer was further purified using the 523 RNeasy MinElute spin column. RNA was finally eluted in RNase-free water. RNA 524 concentrations were estimated using a Nanodrop (Thermo Fisher Scientific, MA, USA) and Qubit (Thermo Fisher Scientific, MA, USA). RNA libraries were prepared at 525 526 the Iowa Institute of Human Genetics (IIHG), Genomics Division, using the Illumina 527 Stranded Total RNA Prep, Ligation with Ribo-Zero Plus (Illumina Inc., San Diego, CA). 528 The KAPA Illumina Library Quantification Kit (KAPA Biosystems, Wilmington, MA) was used to measure library concentrations. Pooled libraries were sequenced on Illumina 529 NovaSeq6000 sequencers with 150-bp paired-end chemistry (Illumina) at the Iowa 530 Institute of Human Genetics (IIHG) core. 531

532 Bulk RNA sequencing analysis: RNA-seq data were processed with the bebio-533 nextgen pipeline (https://github.com/bcbio/bcbio-nextgen, version 1.2.9). The pipeline uses STAR⁵² to align reads to the hg38 or mm10 reference genome and guantifies 534 expression at the gene level with feature Counts ⁵³. All further analyses were performed 535 using R. For gene-level count data, the R package EDASeq was used to account for 536 sequencing depth (upper quartile normalization) ⁵⁴. Latent sources of variation in 537 538 expression levels were assessed and accounted for using RUVSeq (RUVs mode using all features) ⁵⁵. Appropriate choice of the RUVSeq parameter k was guided through 539 540 inspection of principal components analysis (PCA) plots. Specifically, the smallest value k was chosen where PCA plots demonstrated replicate sample clustering in the first 541 three principal components⁵⁶. Differential expression analysis was conducted using the 542 edgeR package⁵⁷. Codes to reproduce the RNA-seq differential gene expression 543 544 analysis are available at https://github.com/YannVRB/Human-brain-stimulation.git.

545 All the transcriptomics data have been deposited in NCBI's Gene Expression Omnibus 546 and are accessible through GEO Series accession number <u>GSE224952</u>.

547 Downstream pathway analysis:

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The enrichment analysis of differentially expressed genes-associated pathways and molecular functions from the stimulated and unstimulated RNA-seq was performed with the Gene Ontology (GO–molecular function) databases using clusterProfiler package in R. Only the pathways with an adjusted p-value ⊂ 0.05 were considered as significant and displayed. Further, the enrichment data were visualized using 'cnetplot' function of clusterProfiler.

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555 Single-nuclei multiomics (nuclei isolation, library preparation, sequencing: Nuclei were isolated from brain tissue using the Chromium Nuclei Isolation Kit (10X 556 557 Genomics). Briefly, frozen tissue was dissociated with pestle in lysis buffer, passed through nuclei isolation column and spun at 16,000 rcf for 20 sec at 4^oC. Flowthrough 558 was vortexed and spun at 500 rcf for 3 mins at 4^oC. Pellet was resuspended with debris 559 removal buffer and centrifuged at 700 rcf for 10 mins at 4^oC, nuclei resuspended in 560 wash buffer and centrifuged again at 500 rcf for 5 mins at 4^oC. Pellet was resuspended 561 in resuspension buffer and nuclei were counted using a hemocytometer. Nuclei were 562 563 directly processed for droplet capture for single cell multiome ATAC + gene expression using a chromium controller (10X Genomics). Chromium Next GEM Single Cell 564 565 Multiome ATAC + Gene v1 chemistry was used to create single nuclei ATAC and RNA 566 libraries from the same cell. Two baseline and two stimulated samples were used for 567 independent replicates. Libraries were sequenced on an Illumina Novaseg 6000 with a 568 150 bp paired end read setup.

569 Single-nuclei multiomic data processing and analysis: To analyze the RNA part of 570 the human brain stimulation multiomic data, gene counts were normalized and log 571 transformed (LogNormalize), and the top 2,000 most variable features between each 572 nuclei were identified using FindVariableFeatures (selection.method $\Box = \Box$ 'vst'). Features that are repeatedly variable across nuclei and datasets were selected for integration 573 574 (SelectIntegrationFeatures). We then identified anchors (FindIntegrationAnchors), which took the list of 4 individual Seurat objects for each sample as input and used these 575 576 anchors to integrate the four datasets together (IntegrateData). The following analyses 577 were performed on the integrated Seurat object. Linear dimensionality reduction was

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performed by principal component analysis (runPCA, npcs = 25). A k-nearest-578 579 neighbors graph was constructed based on Euclidean distance in PCA space and 580 refined (FindNeighbors, npcs = 30), then nuclei were clustered using the Louvain algorithm (FindClusters, resolution $\Box = \Box 0.5$). Clusters were visualized with UMAP 581 582 (runUMAP, dims = 30). Cell types were annotated by label transfer cell labels from an 583 existing human primary motor cortex reference dataset from the Allen Institute (doi: 584 10.1038/s41586-021-03465-8) (FindTransferAnchors and TransferData). Cell types 585 identification was validated by expression of specific biomarkers. Prior of running 586 differential gene expression analysis, as recommended by recent publications (https://doi.org/10.1038/nmeth.4612 and https://doi.org/10.1038/s41467-020-19894-4), 587 588 we used an aggregation-based (pseudobulk) workflow. We aggregated all cells within the same cell type and sample using the AggregateExpression function. This returns a 589 590 Seurat object where each 'cell' represents the pseudobulk profile of one cell type in one 591 individual. After we aggregated cells, we performed celltype-specific differential 592 expression between stimulated and baseline samples using DESeq2.

593 To analyze ATAC part of the human brain stimulation multiomic data, prior to integrating 594 the four Seurat object, the default assay was switched to ATAC, and peak calling was performed. Since the set of peaks identified by Cellranger often merges distinct peaks 595 596 that are close together, creating a problem for motif enrichment analysis and peak-to-597 gene linkage, we identified a more accurate set of peaks by calling peaks using MACS2 598 (CallPeaks) on all cells together. Peaks on nonstandard chromosomes and in genomic 599 blacklist regions were removed (keepStandardChromosomes and subsetByOverlaps). 600 Normalization was performed with a frequency-inverse document frequency 601 normalization which normalizes across cells and peaks (RunTFIDF). Then, a feature 602 selection was performed using all the peaks as input (FindTopFeatures). The 603 dimensional reduction was performed on the TF-IDF normalized matrix with the 604 selected peaks using a singular value decomposition (RunSVD). To mimic the open 605 chromatin conformation of a gene, a gene activity matrix was calculated using a window of 1000bp before and after the transcription start site of each protein coding gene 606 607 (GeneActivity). Differentially accessible transcription start sites in individual clusters 608 between baseline and stimulated samples were calculated using a logistic regression

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framework (FindMarkers, test.use = 'LR', latent.vars = 'nCount_peaks', Padj < 0.05). Motif and transcription factor enrichment analysis for Microglia cluster was performed using FindMotifs on genome-wide all peaks assay of the Seurat object. The top six enriched motifs in microglia cluster are shown. Genomic locations of typical genes like CCL4 were presented (CoveragePlot). It also includes co-accessibility between peaks and transcription start site of genes. Codes to reproduce the multiomic data analysis are available at <u>https://github.com/YannVRB/Human-brain-stimulation.git</u>.

616 NEUROeSTIMator analysis. Gene-level counts for each cell were used as input for the NEUROeSTIMator model³³ to estimate transcriptional signatures of cell activity. Activity 617 was also estimated for cells using gene-level counts in promoter regions from the ATAC 618 619 assay. Estimated activity was assessed for significant differences between the baseline 620 and stimulated conditions across the donors and cell types using paired linear 621 regression. P-values were adjusted for multiple testing with the Benjamini-Hochberg 622 correction. To account for variable sample sizes across cell types and participants, 623 bootstrap resampling was performed (n = 100 cells per donor, replicates = 1000) for 624 each cell type in each condition to generate a distribution of the mean difference in 625 estimated activity between conditions.

To account for variation in sample size across cell types and participants, bootstrap resampling was performed (n = 100 cells per participant, replicates = 1000) for each cell type of the mean difference in estimated activity scores between conditions (mean stimulated – mean baseline). The distributions were analyzed by mean and 95% confidence interval.

Animals: Adult male C57BL/6J mice were purchased from The Jackson Laboratory
 were 3 to 4 months of age during experiments. All mice had free access to food and
 water; lights were maintained on a 12-hour light/12-hour dark cycle.

Mouse electrical stimulation: Stimulation experiments were performed in anesthetized adult male C57BL6/J mice. Anesthesia was induced with 5% isoflurane by inhalation and maintained at 1.8-2% for the duration of the experiment. The mouse was placed in a stereotax (Kopf) and a midline incision was made and the skin retracted to expose the temporal muscle bilaterally. The dorsal insertion of both temporal muscles was

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639 removed, and the muscles retracted. A 2-3 mm craniotomy was made over area AuV 640 (centered at 2.9 mm posterior, 4.2 mm lateral, 2.8 mm ventral from Bregma based on 641 Paxinos atlas) bilaterally to expose the cortical surface and a small square of gel foam 642 soaked in ACSF was applied on top of the dura to prevent it from dehydrating. For each 643 mouse the side of electrical stimulation and sham stimulation were alternated, with the 644 sham side serving as the baseline control for gene expression profiling. Electrical 645 stimulation was delivered through a bipolar ball electrode constructed from two silver 646 wires in which the uninsulated tips were melted under a butane flame (1 mm tip size, 2 647 mm tip spacing). For both electrical and sham stimulations, the gel foam was removed 648 and the electrode was gently lowered to make contact with the cortical surface. For 649 electrical stimulation, a biphasic pulse train was then delivered for two minutes (8 mA, 650 50 Hz, 200 µs pulse width). For the sham stimulation no current was delivered but the 651 electrode was left in place for 2 minutes. Following electrical or sham stimulation the 652 electrode was slowly retracted, and the exposed dura was covered with gel foam. 653 Selection for hemisphere was performed randomly, which resulted in the order of 654 electrical and sham stimulation alternating from mouse to mouse. No more than 5 655 minutes elapsed between electrical or sham stimulation of both sides. After both sides 656 were stimulated (electrical or sham) the mouse was left in the stereotax under 657 anesthesia for 30 minutes before euthanasia and tissue collection. Tissue samples were immediately stored at -80°C in RNA later solution (Ambion). 658

659 **RNA extraction, cDNA preparation and gPCRs from mouse auditory cortex:** Tissue 660 samples were homogenized in Qiazol (Qiagen) using stainless steel beads (Qiagen). 661 Chloroform was added and centrifuged at 12,000g at room temperature for 15 min to 662 separate RNA in the aqueous phase. RNA was precipitated in ethanol and cleared using 663 RNeasy kit (Qiagen). RNA eluted in nuclease-free water was then treated with DNase (Qiagen) at room temperature for 25 min to remove genomic DNA. RNA was further 664 665 precipitated in ethanol, sodium acetate (pH 5.2) and glycogen overnight at -20°C. RNA was precipitated by centrifugation at 30,000g for 20 min, precipitate washed with 70% 666 667 ethanol and the dried RNA pellet was resuspended in nuclease-free water. RNA 668 concentration was measured using NanoDrop (Thermo Fisher Scientific). 1 µg of RNA 669 was used for complementary DNA (cDNA) preparation using the SuperScript IV First-

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Strand Synthesis System (Ambion). Real-time reverse transcription polymerase chain reactions (RT-PCRs) were performed on the QuantStudio 7 Flex Real-Time PCR System (Applied Biosystems, Life Technologies). Data were normalized to housekeeping genes (*Tubulin*, *Pgk1*, and *Actin*), and $2^{(-\Delta\Delta Ct)}$ method was used for gene expression analysis.

675 Mouse cortex spatial transcriptomics analysis (Yann): Spatial transcriptomics data 676 were obtained from the Visium Spatial Gene Expression platform, as described 677 previously⁴². This dataset provides comprehensive gene expression profiles across various brain regions, including the mouse cortex. To focus on the mouse auditory 678 679 cortex, the closest homolog to the human anterior temporal lobe, we manually selected 680 spatial spots from the Visium data corresponding to this brain region. This selection was 681 based on anatomical landmarks and known spatial coordinates for the auditory cortex. 682 For each of the 14 mice in the dataset, we subsetted the data to include only these 683 selected spots, thereby isolating the gene expression profiles specifically from the 684 auditory cortex.

The dataset of 14 mice was divided into two experimental conditions: 7 mice underwent Spatial Object Recognition after 1 hour (SOR-1h), while the other 7 mice were kept in their homecage environment without specific tasks. Based on the Visium spots data overlaying the mouse auditory cortex, differential gene expression analysis was performed to identify genes that were significantly differentially expressed between the SOR-1h and homecage conditions.

691 *Statistics*: For the qPCR analysis, the Wilcoxon matched pairs signed rank test and 692 one sample Wilcoxon test was performed.

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