

Supplementary Materials for

1139 1140	Vocal learning-associated convergent evolution in mammalian proteins and regulatory elements
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1151	This PDF file includes:
1152 1153 1154 1155 1156	Materials and Methods Supplementary Text Figs. S1 to S7 Tables S1 to S2
1157 1158 1159 1160 1161 1162 1163 1164 1165 1166 1167 1168 1169	Other Supplementary Materials for this manuscript include the following:  Movie S1.mp4  DataS1.speciesAnnotations.4.1.7.1.csv  DataS2.vocalLearningRERConvergeMaster.csv  DataS3.HyPhyResults.1.xlsx  DataS4.GeneOntology.1.xlsx  DataS5.BatM1OcrS.2.xlsx  DataS6.BatM1OcrGo.1.xlsx  DataS7.VocalLearningTACITResults.1.xlsx  DataS8.TACIT_ttest.zip  DataS9.OCRMouseGeneAnnotation.1.xlsx  DataS10.CellTypeEnrichments.1.xlsx

#### **Materials and Methods**

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- 1171 Coding the vocal learning trait
- 1172 The vocal learning trait was annotated for a set of 215 mammalian species that satisfied the
- 1173 following conditions: (1) the species' genome was present in the Zoonomia whole-genome
- 1174 Cactus alignment (241 assemblies total, (17, 99)), (2) the species' protein-coding gene sequences
- were present in the TOGA gene alignment set described in (10) (427 species total), and (3) the 1175
- 1176 species was a member of the Boreoeutheria clade (Fig. 1A), given that our primary data were
- 1177 restricted to species within this Magnorder.
- 1178 Although, in many studies, vocal production learning is treated as a binary ("presence /
- 1179 absence") trait possessed by humans, bats, pinnipeds, and cetaceans alone among boreoeutherian
- mammals, 'gold standard' tests for the trait have been performed for only a handful of species 1180
- 1181 within these clades, and recent reevaluations of the evidence suggest that a more nuanced coding
- 1182 of this complex trait is needed, involving extensive behavioral testing of a greater taxonomic
- 1183 diversity of species (2, 51, 100). To attempt to account for this, boreoeutherian species were
- coded as vocal learners only if they belonged to an established vocal learning clade (i.e, bats, 1184
- 1185 pinnipeds, cetaceans, and humans) and presented evidence of song usage or a rich social acoustic
- 1186 repertoire in the literature. Species potentially falling somewhere outside a simplified binary
- coding were excluded from analyses based on the following criteria: (i) species with suggestive, 1187
- 1188 insufficient, or controversial evidence of vocal plasticity or learning (Callithrix jacchus,
- 1189 Heterocephalus glaber, Indri indri, Pan paniscus, Pan troglodytes) or (ii) domesticated species
- 1190 (or subspecies) given demonstrated connection between domestication and increased vocal
- 1191 variability (101) (Camelus bactrianus, Camelus dromedarius, Canis lupus familiaris, Capra
- 1192 hircus, Equus asinus asinus, Equus caballus, Felis catus, Mustela putorius furo, Vicugna pacos).
- 1193 "In sum, this totaled a set of 215 species with genome assemblies, aligned protein predictions,
- 1194 and vocal learning annotations (26 high-confidence vocal learners, 25 suspected vocal learners,
- 1195 labeled as "unsure", and 164 high-confidence non-learners, Data S1)
- 1197 Evaluating the relationship between relative evolutionary rate of protein-coding genes and vocal 1198 learning
- 1199 RERconverge (11, 65, 102) was used to evaluate the relationship between relative evolutionary
- rates (RERs) of protein-coding genes and the vocal learning trait across mammals. A set of 1200
- 1201 37,552 high-quality protein-coding amino acid alignments generated with TOGA using human
- reference sequences mapped across a human-referenced MAF alignment of 427 species was 1202
- 1203 obtained (103). These alignments were subsequently filtered to remove duplicated species,
- 1204
- poorly represented proteins, and low-scoring alignments. Specifically, alignments with fewer
- 1205 than 221 unique species (0.025 quantile of the distribution of unique species number for all
- 1206 alignments), alignments with fewer than 189 total species with ungapped coverage of 50% of the
- 1207 total alignment length (0.1 quantile), alignments with more than 97 duplicated species (0.95
- 1208 quantile), and alignments with ungapped length <267 bp (0.01 quantile) were excluded. In total,
- 1209 this resulted in excluding 4,723 transcripts representing 2,613 unique genes. Within the
- remaining alignments, any sequences that did not cover 50% of the total alignment length were 1210
- 1211 excluded, and, when there were multiple sequences for a species within an alignment, the
- 1212 sequence with the highest identity to the human reference sequence across the full alignment
- 1213 length was retained. For genes with multiple transcripts, only the alignment with the longest

- median ungapped coverage was retained. From the remaining 16,209 protein-coding gene
- alignments, branch lengths on the consensus species tree from (10) were estimated for each gene
- 1216 using approximate maximum likelihood estimation with the WAG substitution model, as
- implemented in the phangorn package in R (104, 105). RERs were generated from these
- topology-constrained gene trees using RERconverge version 0.3.0 (11) with R version 4.1.0.
- 1219 RER-to-phenotype correlations were generated using the vocal learning trait annotations
- described above and RERconverge's correlateWithBinaryPhenotype tool, considering all
- foreground branches and otherwise using default parameters. The p-values were corrected using
- the Benjamini-Hochberg procedure (106) and protein-coding genes were considered to have
- significant associations if their corrected p-values were less than 0.01. In addition, empirical p-
- values were computed using phylogenetic permulations (24), which accounts for the structure of
- the tree and the distribution of the trait across the tree, and only genes with permulation p < 0.01
- 1226 after Benjamini-Hochberg correction were considered for downstream analyses . To stratify
- which clade or clades are driving changes in the RERconverge results, a Bayes Factor approach,
- which measures the difference in evolutionary rates with that vocal learning clade relative to
- other species (18), was employed. The results of these analyses are found in Data S2.
- 1230 HyPhy Analysis
- 1231 The RELAX analysis, implemented in the HyPhy package, was applied to the same set of
- 1232 genomes for which the RERconverge analysis was run (10), with only the confident
- 1233 vocal learning species included in the foreground set. Briefly, RELAX
- infers and compares dN/dS distributions from test (vocal learners)
- 1235 and reference (all other) lineages, and estimates the
- 1236 intensification/relaxation parameter (K). When K≠1 with sufficient
- 1237 statistical support (LR test), test lineages were inferred to be
- 1238 relaxed/intensified compared to the reference lineages. Multiple
- 1239 ENSEMBL transcripts per gene were allowed to enable a more detailed
- 1240 exploration of each gene. Given the computational constraints of the
- 1241 software, HyPhy was run on the set of genes that showed greater
- 1242 conservation or acceleration based on the RERconverge analysis. A
- 1243 negative set of 100 transcripts randomly chosen from the set of
- transcripts not associated with vocal learning (p>0.5) set, which were
- 1245 matched to the RERconverge set to mirror the length distribution
- 1246 (Data S3). To best align with the assumptions of RERconverge
- 1247 analysis, transcripts with RELAX q<0.01 and the
- 1248 relaxation/intensification parameter (K) estimate were used; values
- of 0<K<1 indicate relaxed selection, and values K>1 intensified
- 1250 **selection** (12, 107). However, results from the BUSTED-PH analysis of associating
- episodic positive selection with a binary phenotype of vocal learning were also provided (12).
- For all analyses, synonymous substitution rates were allowed to vary from site to site because not
- doing so risks subjecting selection analyses to strong confounding biases (108).
- 1254 Gene Ontology Analysis

A set of genes were selected that were significantly accelerated or conserved in vocal learning species based on a number of criteria (RERconverge adjusted P < 0.01; Permulation Adjusted P < 0.01; HyPhy Q < 0.01). Those genes served as input to the EnrichR web interface with default parameters (91). The dataset for "GO\_Biological Process" was used to find the overall function of the genes in question. The "Human Phenotype" ontology was used to link the genes to specific human phenotypes or disease states. Specific gene ontology categories were chosen for visualization based on an adjusted p < 0.1 (Data S4).

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Subjects for tracing and electrophysiological experiments

1264 All experimental procedures were approved by the Institutional Animal Care and Use Committee 1265 of the University of California, Berkeley. All animals were adult Egyptian fruit bats (weight range 130-200g) maintained within the lab colony. For tracing experiments using 1266 channelrhodopsin injected into ofM1, subjects were three females born in the lab. For tracing 1267 1268 experiments using AAV<sub>DJ</sub>-hSyn-Synaptophysin as an anterograde tracer injected into ofM1, subjects were two males born in the lab. For the 3D mapping of ofM1 descending tracts using 1269 dextran amine (Figure S4A-B, F and Movie S1), the subjects were two females born in the lab. 1270 1271 For electrophysiological recordings into ofM1, the four implanted subjects were wild-caught bats. 10 other adult bats (seven males and three females) were used as companions during the 1272 recording sessions to elicit vocal interactions. While the age of the four wild-caught bats could 1273 1274 not be precisely estimated, all were greater than four years old. Prior to the start of experiments, 1275 bats were housed in a communal vivarium. After implantation for electrophysiology, the four 1276 implanted bats were initially single housed and subsequently, following recovery from surgery, 1277 were co-housed in pairs. All cages were kept in a humidity- and temperature-controlled room, on 1278 a 12-hour reversed light-dark cycle. Bats were given ad libitum access to water and fed with fruit 1279 mix supplemented with honey and vitamins.

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1290 1291 Anesthetic procedure for tracer injection and electrophysiological implant

The general anesthesia and surgical procedures used for Egyptian fruit bats were previously described (109, 110). Anesthesia was induced using a cocktail of ketamine, demedotomidine, and midazolam. The depth of anesthesia was evaluated by monitoring the breathing rate, body temperature, and toe pinch reflex. The body temperature was monitored continuously using a rectal probe and maintained around 34.5°C with a heating pad under the bat. If necessary to maintain anesthesia longer than one hour, the bat was injected with a cocktail of dexmedetomidine, midazolam, and fentanyl. Anesthesia was reversed with an injection of atipamezole and the bat was hydrated subcutaneously with lactated ringer's solution. For all surgical procedures, antibiotics were given for one week post-surgery and analgesic pain medication was given for three days post-surgery. For tracing experiments, the sutures were removed within five days post-surgery.

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Anterograde tracer injection procedures

The procedures for the injection of the anterograde tracer into the brain and retrograde tracer into the cricothyroid muscle were performed at separate times to allow for optimal expression of both tracers (111). Two different tracers were used in order to validate the anatomical results across

- multiple techniques: AAV mediated channel-rhodopsin (ChR2) conjugated to GFP
- 1299 (rAAV5/CamkII-hChR2(H134R)-EYFP, Lot#AV4316LM; UNC Vector Core, NC) (112) and a
- 1300 custom synaptophysin/synapsin virus (SYN) from Byungkook Lim's lab that simultaneously
- labeled fibers in eGFP and boutons in mRuby2 (AAV<sub>DJ</sub>-hsyn-mRuby2-T2S-Synap-eGFP; Lim
- 1302 Lab, UCSD) (113, 114). Each anterograde tracer type was injected bilaterally into of M1. The 3D
- 1303 reconstruction of ofM1 projections and the localization of the decussation of the pyramidal tract
- 1304 (Supp Fig 1 A-B and Supp Fig 1 F) were obtained in two females injected bilaterally in ofM1
- with Dextran amine conjugated to Alexa Fluor 555 (ThermoFisher Scientific; D34679).
- For anterograde tracers injected intracranially into ofM1, bats were anesthetized and head fixed
- in a stereotaxic device (Model 942; Kopf, CA). After opening the scalp with a scalpel, the tissue
- was retracted to expose the skull. The center of three injection coordinates for ofM1 (AP:
- +10.72mm, 10.22mm, 9.72mm; ML:+/- 3.2mm) were bilaterally measured from a common
- reference point above the confluence of the sinus. A small craniotomy (1.2 mm long x 0.6mm
- wide) was made above of M1 to expose the surface of the brain while leaving the dura intact.
- Bilateral injections were made along the anterior-posterior axis into each hemisphere using a
- NanoFil syringe (36GA beveled needle; WPI, FL) attached to the stereotaxic device. The syringe
- was slowly lowered to -1.2mm below the surface of the brain around layers V/VI of ofM1 and
- allowed to rest for three minutes above the deep target. After pausing, 0.5 µL of one of the two
- anterograde tracers (ChR2, or SYN) were injected at a rate of 4nl/sec using a microinjection
- pump (UMP3; WPI, FL). The needle was left in place for five minutes at each site. A total
- volume of  $1.5\mu$ L was delivered in each hemisphere. Upon completion of the six injections,
- 1319 Kwik-Sil (WPI, FL) was used to fill the craniotomy and protect the brain and the tissue was
- 1320 sutured.

# 1322 Retrograde injections into cricothyroid muscles

- 1323 The retrograde tracer injection was performed approximately one month following the
- anterograde tracer injection to optimize maximal expression of the virus and propagation of
- 1325 cholera toxin B (CTB). Approximately one week before the planned perfusion time, the bats
- were anesthetized according to the same procedures above. Once anesthetized, the neck was
- shaved and the bat was placed on its back on a heating pad to facilitate access to the larynx. The
- skin overlying the larvnx was incised using a scalpel to reveal the larvnx below the sternohyoid
- and infrahyoid muscles. The tissue was retracted to expose the cricothyroid muscle caudal to the
- inferior border of the thyroid cartilage and medial to the cricothyroid joint.
- Bilateral injections of cholera toxin B conjugated to fluorescent labels (ThermoFisher, C34778,
- 1332 AlexaFluor 647) were made in the cricothyroid muscle at six different points, three on each side.
- 1333 A NanoFil syringe (35GA beveled needle; WPI, FL) attached to the stereotaxic device was
- slowly lowered approximately -0.4mm below the surface of the muscle. After waiting one
- minute to allow the tissue to settle, 2µL of CTB was injected at a rate of 16nl/sec into each
- injection site. The needle was left in place for one minute before moving to the next injection
- site. Upon completion of the six injections of  $2\mu L$  (12 $\mu L$  total) the tissue was sutured and two
- 1338 surgical staples were placed over the sutures.

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Electrophysiological implant surgery

1341 The surgical procedures for the implantation of four tetrode microdrives followed those 1342 described previously in detail for Egyptian fruit bats (46, 110, 115). Each bat was implanted with a lightweight microdrive (Harlan 4-Drive, Neuralynx; weight 2.1 g) loaded with four tetrodes 1343 1344 (~45μm diameter; four strands of platinum-iridium wire, 17.8μm in diameter, HML-insulated) that could be moved independently. The tetrodes exited the microdrive through a guide cannula 1345 1346 with ~300μm horizontal spacing between tetrodes. On the day before surgery, each tetrode's tip 1347 was cut flat using high-quality scissors (tungsten-carbide scissors ceramic coating, CeramaCut; 1348 FST) and plated with Gold Plating Solution (Neuralynx) to reduce the impedance of individual wires to 0.20- $0.55M\Omega$  (at 1kHz). The principal surgical steps to implant the microdrive were as 1349 1350 follows: after scoring the skull to improve adhesion and mechanical stability, a circular 1351 craniotomy of 1.7 mm diameter was made in the skull over the left hemisphere 3.2mm lateral to 1352 the midline and 10.7 mm anterior to the transverse sinus that runs between the posterior part of the cortex and the cerebellum; after a durotomy, the microdrive was placed vertically such that 1353 1354 the tip of the microdrive's guide tube was flush with the brain's surface; the exposed craniotomy 1355 was then filled with a biocompatible elastomer (Kwik-Sil, World Precision Instruments) to protect the brain; a bone screw (FST) with a soldered stainless-steel wire was fixed to the skull in 1356 1357 the frontal plate contralateral to the microdrive to serve as a ground screw; an additional set of 3 bone screws were fixed to the skull to serve as anchor screws for maintaining mechanical 1358 1359 stability of the implant; finally the exposed skull and anchor screws were covered with a thin 1360 layer of quick adhesive cement (C&B Metabond, Parkell) to provide a substrate for the adhesion 1361 of dental acrylic that was added as final layer to secure the entire microdrive to the screws and to 1362 the skull.

# Electrophysiological and audio recording devices

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1364 Electrophysiological recordings were conducted using a wireless neural data logging device 1365 ("neurologgers"; MouseLog-16 (vertical version), Deuteron Technologies) that interfaced with 1366 the microdrive of each implanted animal. The neurologger amplified the voltage signals from the 16 channels of the four tetrodes referenced to the ground screw, performed analog-to-digital 1367 conversion at a sampling rate of 31.25kHz, and stored the digitized data on a removable 32GB 1368 SD card that can hold up to 9 hours of recording. The system has a bandwidth of 1Hz - 7kHz, 1369 1370 records voltage with a fine resolution of 3.3µv, and has a low level of noise generally close to the 1371 limit of Johnson noise from the impedance of a given source. The neurologger and its lithium-1372 polymer battery were encapsulated in a house-made 3D-printed plastic casing to prevent damage 1373 to the electronics, and weighed a total of 9.9g.

1374 The audio recordings of each individual bat vocalizations were performed using a call detector, 1375 as previously described (46). In brief, a single-axis, low mass, piezo-ceramic accelerometers 1376 (BU-27135, Knowles Electronics, sensitivity 0-10kHz) was mounted on a flexible rubber 1377 necklace placed against the throat of the subject in a way that did not restrict normal behavior to 1378 detect laryngeal vibrations produced during vocalizations. The signal of the accelerometer was 1379 recorded, digitized at a sampling rate of 50Hz, and saved on removable SD cards with a wireless audio data logging device ("audiologgers"; Audio Logger AL1, Deuteron Technologies) 1380 mounted on the necklace on the back of the subject. The audiologger and its lithium-polymer 1381 1382 battery were encapsulated in a house-made 3D-printed plastic casing to prevent damage to the 1383 electronics. All audiologgers and neurologgers were controlled and synchronized by a single

transceiver. The Egyptian fruit bats in our experiment weighed more than 140g and could fly

with ease while equipped with both the neurologger and the audiologger.

1386 Vocalizations and motor actions recording sessions

1387 The four implanted bats were divided into two pairs that were independently recorded for 1-3

hours per day over multiple days (16 and 32 sessions) with two or three peers. These peers were

randomly chosen from a pool of ten bats (seven males, three females) and were used as

1390 companions to increase the probability of vocal interactions implicating the subjects. During the

daily electrophysiological and audio recording sessions these groups of four to five bats (2)

implanted + 2-3 companions) were housed in a rectangular prism (180 x 60 x 60cm) that had two

sides made of plexiglass, thereby permitting clear remote visual monitoring of bats behavior via

2 cameras (Flea 3 FLIR). The remaining sides of the enclosure were made of plastic mesh,

allowing bats to easily perch and crawl on the surface. The enclosure was placed in an

electromagnetically and acoustically shielded room and all recording sessions were conducted

during the dark cycle under red LED light. All bats were equipped with audiologgers such as to

record and identify their vocalizations. Water was given *ad libitum* and fruits were placed into

the cage such as to engage the animals into chewing and licking behavior. The experimenter was

monitoring the behavior of the animals in an ante-chamber via video cameras and an ambient

1401 ultrasonic microphone (Earthworks, M50) centered 20cm above the cage ceiling and connected

to the main computer unit via an analog to digital convertor (MOTU, 896mk3). Using a house-

made keystroke annotation code written under Matlab, the experimenter was manually

annotating chewing, licking (self-grooming using licking movements), and quiet (staying still in

a relaxed position, wing closed) behaviors. The audio was recorded throughout the session

1406 (sampling rate of 192kHz) from the ambient microphone using an in-house Matlab GUI

1407 (VocOperant; https://github.com/julieelie/operant\_bats). The synchronization between the

microphone recording, the manual annotation of behaviors and the transceiver controlling the

audiologgers and neurologgers was achieved using transistor-transistor logic (TTL) pulses

1410 generated by an UltraSoundGate Player 216H (Avisoft Bioacoustics) and sent via coaxial cables.

1411 After each recording session, tetrodes were connected to a wired recording system (Digital Lynx,

Neuralynx) to monitor the neural signals and advance the tetrodes. Tetrodes were moved

downward once every one to two days (generally by 100 µm) in order to record single units at

1414 new sites.

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1415 Histology

#### 1416 Perfusion

1417 Approximately one week following the injection of the retrograde CTB tracer, or after the last

day of electrophysiological recording, bats were administered an overdose of pentobarbital and

perfused transcardially with 250ml of pH 7.4 phosphate buffered saline (PBS) spiked with 0.5ml

heparin (1000 USP units/ml) ---followed by 250ml of fixative (3.7% formaldehyde in phosphate

buffered saline). When the brain was implanted with electrodes, tetrodes were left in the brain

for 30 minutes before extracting them. The brain was then carefully removed from the skull and

post-fixed overnight in the same fixative. To avoid over-fixation, the brain was removed from

1424 fixative after 24 hours and switched into a 30% sucrose solution for cryoprotection. After

approximately two days or once the brain had sunk to the bottom, 40 µm coronal sections were

1426 made using a microtome (HM450; ThermoFisher, MA) with a freezing stage.

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# Staining and Immunocytochemistry

1429 The sections from the implanted bats were Nissl-stained with cresyl violet. Slides were imaged 1430 using a light microscope to verify tetrode positions.

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1432 The sections of the brains from bats injected with tracers were stained for VGLUT1 using a fresh 1433 tissue floating immunohistochemistry protocol. Immunohistochemistry was conducted in 12-well plates filled with 4ml of solution for washes and blockings and 48 well plates filled with 1 ml of 1434 1435 solution for primary and secondary incubation. Fifteen brainstem slices were selected from each series centered around the NA region and three brainstem slices anterior to the target region were 1437 selected for antibody control staining. Briefly, the tissue was placed in floating wells on a lab 1438 rotator in a cold room at 4°C and washed in three separate PBS (0.025M, pH 7.4) solutions for 1439 five minutes each wash. The tissue was then moved to a blocking solution containing 10% goat 1440 serum (Sigma-Aldrich, G9023) in 0.3% triton-PBS (Triton X-100 - ACROS Organics, 21568-2500 in 0.025M PBS) and rotated for 90 minutes at 4°C. The tissue was incubated overnight at 1442 4°C in rabbit anti-VGLUT1 primary antibody (provided by Eiman Azim, Salk Institute for 1443 Biological Studies and produced in Tom Jessell's lab at Columbia University) (116), which was 1444 prepared in a 1:16,000 dilution in 5% goat serum and 0.3% triton-PBS. Control slices were incubated in an antibody buffer without primary antibody. Approximately 16-24 hours after the 1445 start of the primary antibody incubation, the tissue was moved into three separate 0.3% triton-1446 1447 PBS washes for 10 minutes in each wash at 4°C. The secondary antibody was goat anti-rabbit 1448 conjugated to a fluorescent protein (ThermoFisher A27012, A27018) that did not conflict with the anterograde or retrograde tracers (selected wavelength 594 nm for bats injected with ChR2). The tissue was incubated in the secondary solution diluted 1:500 in 5% goat serum and 0.3% triton-PBS at room temperature for 90 minutes before three final washes in PBS (0.025M, pH 7.4) for 10 minutes each. DAPI was added to the secondary solution for the final 10 minutes of 1452 1453 incubation at 1:10,000 dilution (D1306; ThermoFisher, MA). The sections were then mounted on 1454 glass slides and cover-slipped using ProLong Gold Antifade Mountant (P36934; ThermoFisher, 1455 MA).

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### Imaging and anatomical quantification

#### 1458 Fluorescent imaging

- 1459 All imaging was conducted at the University of California, Berkeley Cancer Research
- 1460 Laboratory Molecular Imaging Center and the Henry H. Wheeler Jr. Brain Imaging Center at UC
- 1461 Berkeley. Preliminary imaging at magnification of 10x/20x using Plan-Apochromat 10x/20x
- 1462 objective was conducted on Zeiss Axio Scan Z1 Slide Scanner. Slices were imaged in four
- 1463 fluorescent channels – AF647 (Excitation: 653nm, Emission: 668nm, Gain: 0, Exposure time:
- 1464 25ms, Filter Cube: 50 Cy5), AF488 (Excitation: 493nm, Emission: 517nm, Gain: 0, Exposure
- 1465 time: 20ms, Filter Cube: 38 HE Green Fluorescent Protein), DAPI (Excitation: 353nm,
- 1466 Emission: 465nm, Gain: 0, Exposure time: 5ms, Filter Cube: 49 DAPI), and AF598 (Excitation:
- 570nm, Emission: 618 nm, Gain: 0, Exposure time: 15ms, Filter Cube: 64 HE mPlum). All 1467
- 1468 images were taken with 100% fluorescent lamp strength on Hamamatsu Orca Flash Camera with
- 1469 1x Camera Adapter.

- More specific imaging of nucleus ambiguus area and other target and control regions was
- 1471 conducted at magnification of 63x with Plan Apochromat 63x/1.4 Oil DIC M27 Objective with
- target area (135μm x 135μm) on Zeiss LSM 880 Confocal Microscope at the UC Berkeley
- 1473 Molecular Imaging Center. The target region was localized by finding retrogradely labeled cells,
- 1474 centering them, and taking a z-stack to encompass the entire cell volume. Each z-stack was taken
- over 1.5µm depth with the above mentioned area and 1µm between each z-stack slice with 4
- 1476 fluorescent channels AF 647 (Laser wavelength: 633 nm, Excitation: 633nm, Emission:
- 1477 697nm, Laser Power: 6.0%, Detector Gain: 650, Detector Digital Gain: 1, Detector Offset: 0),
- 1478 AF 594 (Laser wavelength: 594nm, Excitation: 594nm, Emission: 659nm, Laser Power: 8.0%,
- 1479 Detector Gain: 675, Detector Digital Gain: 1, Detector Offset: 0), AF 488 (Laser wavelength:
- 488nm, Excitation: 488nm, Emission: 552nm, Laser Power: 5.0%, Detector Gain: 625, Detector
- Digital Gain: 1, Detector Offset: 0), and DAPI (Laser wavelength: 405nm, Excitation: 405nm,
- Emission: 462nm, Laser Power: 6.0%, Detector Gain: 625, Detector Digital Gain: 1, Detector
- Offset: 0). If relevant, AF 561 (Laser wavelength: 561nm, Excitation: 561nm, Emission: 621nm,
- Laser Power: 8.0%, Detector Gain: 650, Detector Digital Gain: 1, Detector Offset: 0) was taken
- on a separate z-stack due to maximum channels/image microscope limitations. All confocal
- images were taken as 12-bit images with line-by-line imaging and 2x averaging.
- 1487 Deconvolution was conducted using HuygensPro software from the Biological Imaging Facility
- at UC Berkeley. Theoretical PSF was generated using LSM 880 confocal microscope and 63x
- objective parameters. Image histograms were created using a logarithmic mapping function and
- background was generated using automatic estimation with an area radius of  $0.7\mu m$ .
- Deconvolution was conducted for each channel with 200 maximum iterations, 30 signal to noise
- ratio, 0.01 quality threshold, and with optimized iteration mode.

### 3D model of ofM1 tracing and localization of decussation

- Series of coronal slices were manually stacked to build a 3D model of the fluorescent pathways
- of fluorescently labeled dextran amine (DA) tracing to/from ofM1. Bats were injected with
- 1497 fluorescent dextran amine in ofM1 and, following perfusion, coronal slices of the whole brain
- separated by 200 µm were stained for DAPI to create a uniform background marker. After slide
- scanning the entire brain (see scanning settings above), the color levels for the DAPI and DA
- 1500 channels were equalized for every coronal slice and exported into lossless tiffs from ZEISS ZEN
- 1501 Microscope Software. The coronal slices were cut out from against the background and manually
- aligned using Adobe Photoshop so that the edges lined up in a full stack from the rostral tip of
- the olfactory bulbs to the caudal tip of the spinal cord. All images were then stacked in Imaris to
- 1504 create a 3D model of the brain that can be rotated to observe the whole structure of the
- 1505 corticobulbar pathway from ofM1 to NA (Fig. S4A-B and Movie S1).

#### 1507 <u>Image quantification</u>

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- 1508 Total numbers of cells, fibers, boutons, and DAPI-labeled cells of brains injected with SYN and
- 1509 ChR2 were quantified using Imaris software (Version 9.2.1) at the UC Berkeley Cancer
- 1510 Research Laboratory Molecular Imaging Center. Retrogradely labeled cells and boutons were
- automatically counted using Imaris Spot Tracker. Initially, the diameter of retrogradely labeled
- 1512 cells (15-30µm), boutons (1-3µm), and DAPI labeled nuclei (10-15µm) was measured using

- 1513 automatic measuring tools in Imaris in a 2D slice and imputed as a parameter for Spot Tracker.
- After computing, x, y, and z, diameter and position was manually adjusted until the spot covered 1514
- the entire retrogradely labeled cell. Fibers were counted and traced using semi-automatic Imaris 1515
- 1516 AutoPath Tracer by manually tracing along the length of the fiber. Fiber diameter was set to
- 1517 1.4um across all z-stacks.
- 1518 Overlap was quantified manually by determining points of colocalization between fiber/cell and
- 1519 fiber/cell/bouton. Z-stacks were exported as single multi-channel TIFF images and then opened
- each Z-stack individually in Adobe Photoshop. Points of co-localization were individually and 1520
- 1521 manually marked. TIFF images were compared to the original z-stack to confirm the existence of
- 1522 fiber/cell/bouton/DAPI-labeled cells over multiple slices and ensure real signal. The number of
- retrogradely labeled cells with CTB that had at least one overlap with fibers and the number of 1523
- 1524 retrogradely labeled cells with CTB that had at least one overlap with fibers and boutons were
- counted. If overlap on the same cell spanned multiple TIFF images, it was only counted in the 1525
- 1526 first slice in which it appeared.

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Acoustic data processing

Acoustic data logged as voltage traces on the SD cards of the audiologgers were extracted into Matlab files and aligned across bats simultaneously recorded using a custom-made Matlab code (https://github.com/NeuroBatLab/LoggerDataProcessing/). Potential vocalizations were then detected and segmented from these piezo recordings using an in-house series of Matlab scripts (https://github.com/NeuroBatLab/SoundAnalysisBats). The whole process consisted in three major steps: 1) detection of sound events on call detector signals, 2) automatic classification between vocalizations and noise, 3) manual curation of potential vocalizations.

- 1. Detection: To focus on the detection of vocalizations emitted by the bat wearing the collar, the signal of each call-detector was first band passed between 1 and 5kHz. As previously described, this frequency range is not contaminated by airborne vocalizations from other bats standing close to the collar wearer (46). After determining a noise threshold from sections of silence during the recording session for that call-detector, potential vocalizations were detected by threshold crossing on the amplitude envelope (RMS, sampling frequency 1kHz) and any sound event above threshold for longer than 7ms was kept.
- 2. Automatic step of data sorting between actual vocalizations and noise: Sound events closer than 50ms were merged as a single sound sequence and a battery of 20 acoustic measurements were applied on them (see (117) for mathematical definitions): RMS, maximum amplitude, the five first momentum of the amplitude envelope taken as a distribution (mean, standard deviation, kurtosis, skewness, and entropy), the five first momentum of the frequency spectrum taken as a distribution, the three quartiles of the frequency spectrum, the mean pitch saliency, and four parameters that pertain to the sound as recorded from the ambient microphone (RMS, mean and maximum of the amplitude envelope, and maximum value of cross-correlation between the microphone and the call detector signals). Potential vocalizations among the detected sound events were then identified using these 20 acoustic parameters as input to a support vector machine trained on the data of two sessions manually sorted between vocalizations and noise by an expert (JEE). This automatic sorting was set to be very conservative of vocalizations by setting the threshold on the posterior probability of a vocalization at 0.2.

3. Manual curation: to further eliminate noise and check the identity of the bat producing vocalizations, each potential vocalization was aurally and visually scrutinized by an expert (JEE) based on the inspection of the spectrograms of its signal as recorded from the ambient microphone and from the call detectors of all the bats. After this step, vocalizations further than 200ms apart were considered as distinct, while the others were merged into a single sequence.

## Neural data processing

## Spike detection and sorting

Neural data logged as voltage traces on the SD cards of the neurologgers were extracted into Matlab files and aligned across simultaneously recorded bats and with the audio data, using custom Matlab code (<a href="https://github.com/NeuroBatLab/LoggerDataProcessing/">https://github.com/NeuroBatLab/LoggerDataProcessing/</a>). Spike detection and sorting was achieved using a generative algorithm (Kilosort 2.0,

https://github.com/MouseLand/Kilosort/releases/tag/v2.0, (118)) with the parameters set as indicated under <a href="https://github.com/julieelie/Kilosort2\_Tetrode/configFiles/configFile16.m">https://github.com/julieelie/Kilosort2\_Tetrode/configFiles/configFile16.m</a>. Clusters were further manually curated using Phy (<a href="https://github.com/cortex-lab/phy">https://github.com/cortex-lab/phy</a>).

Units obtained after manual curation were sorted between multi-units and single units by applying thresholds on two metrics: 1) the consistency of the spike snippet (signal over noise ratio of the spike snippet, SNR, estimated at the four contact points of the tetrode) and 2) the respect of the refractory period (percent of violation of the refractory period, VRP). A unit had to have at least one out of four SNR values above 5 and VRP<0.1% to be considered as a single unit.

For each unit, the spike snippet SNR was calculated for each electrode of the tetrode as:

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$$SNR = \frac{max(MeanSpike) - min(MeanSpike)}{\sqrt{(StdSpike_{tMax}^2 - StdSpike_{tMin}^2) \div 2}}$$

with *MeanSpike* the average spike snippet over all spikes, *StdSpike* the time-varying standard deviation of the spike snippet over all spikes, and *tMax* and *tMin* the time point at which the average spike snippet reaches maximum and minimum values, respectively.

Applying the above mentioned thresholds on SNR and VRP of units manually curated with Phy yielded 381 single units (SNR = 8.15 + -0.12; VRP = 0.0308 + -0.0015 %). 94.2% of single units (359/381) had an index of contamination Q (118–120) below 0.2 (Q = 0.0531 + -0.0042, N=381) and a significant test of refractoriness against a Poisson distribution at p<0.05.

#### Firing rate calculations and analysis

For this analysis, we selected the single units that were recorded during the sessions when the subject had produced a minimum of ten vocalizations longer than 100 ms and had displayed chewing, licking and quiet behaviors (n=237 units). For each single unit, the time average firing rate during the production of vocalizations longer than 100ms was estimated for the duration D of each vocalization starting 10ms before vocalization onset as:

- $R = \frac{N_{AP}}{D + I0}$  with  $N_{AP}$  the number of action potentials occurring during the time window D + 10 ms.
- For all other behaviors (chewing, licking, and quiet), the annotated period of time where the bat was demonstrating that behavior was reduced by 1 second (true offset considered as 1s before

1600 keystroke) to conservatively accommodate for the annotator time response. Then the firing rate 1601 was estimated in time segments of the same durations as those used to estimate firing rate during vocalizations by randomly sampling without replacement into the period of time where the bat 1602 1603 was demonstrating the behavior of interest.

Firing rate comparisons between pairs of behaviors were achieved by applying a test on the deviance of the Poisson Generalized Linear Model (GLM) predicting the rate R as a function of the category of behavior (function fitglm of Matlab). P-values were corrected for false detection rate using the Benjamini-Hochberg procedure.

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## Information on coherence during vocalization perception and production

The relationships between each single unit activity and vocal production or vocalization perception were quantified by calculating the coherence between the time varying amplitude of vocalizations and the time varying firing rate of the unit during the vocalizations respectively produced or heard. For this analysis, we selected the single units that were recorded during the sessions when the subject had both produced and heard a minimum of twenty vocalizations (n=219 units). Vocal activity, as measured by the piezoelectric sensor of the call detector, and neural activity, as represented by the arrival times of action potentials, were collected from 200ms prior to the onset to 200ms after the offset of each vocalization. The time-varying amplitude of each sound extract was taken as the amplitude envelope calculated with a frequency cut-off at 150Hz and sampled at 1000Hz (BioSound python package,

1618 1619 1620 https://github.com/theunissenlab/soundsig, (117)). The spike pattern corresponding to each sound extract was convolved with a Gaussian window of 1ms standard deviation to obtain a 1621 1622

time-varying firing rate sampled at 500 Hz. The coherence between the time-varying amplitude 1623 and the time-varying firing rate was then calculated across all vocalizations that the animal had 1624 produced/heard during the session to give the motor/auditory coherence. A multitaper approach

1625 was implemented in the estimation of coherence to obtain an error measure (121). The

information on coherence was obtained by integrating all values of positive coherence up to the 1626 1627 Nyquist limit. Information was calculated on the estimate of coherence and on its lower and

1628 upper bounds both for vocalizations produced and vocalizations heard, yielding a value of motor

information  $Info_{motor}$  and its corresponding lower and upper bounds  $Info_{motor}^{low}$  and  $Info_{motor}^{up}$  as well as a value of auditory information  $Info_{aud}$  and its lower and upper bounds  $Info_{aud}^{low}$  and 1629 1630

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Inf  $o_{aud}^{up}$ . For each unit, the information D-Prime was then calculated as:  $DPrime = \frac{Info_{motor} - Info_{aud}}{\sqrt{E_{motor}^2 + E_{aud}^2}}, \text{ with } E_{motor} = Info_{motor}^{up} - Info_{motor}^{low} \text{ and } E_{aud} = Info_{aud}^{up} - Info_{aud}^{low}$ 1632

 $Info_{aud}^{low}$ . 1633

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The significance of the difference in information between produced (motor) and heard (auditory) vocalization was assessed by a likelihood ratio test between two linear mixed effect (LME) models predicting the information value with or without the type of information (motor vs auditory) as a fixed effect and with the identity of the subject as a random variable.

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### Animals and sample collection for epigenomics

- 1641 All animal procedures were in accordance with the National Institutes of Health Guide for the
- 1642 Care and Use of Laboratory Animals and approved by the Institutional Animal Care and Use
- 1643 Committees the University of California, Berkeley. Two adult (>1 year) male Egyptian fruit bats

1644 (Rousettus aegyptiacus), one male and one female, were housed socially in a large free-flight 1645 vivarium. Bats were acoustically and socially isolated in a sound recording chamber (external 1646 dimensions: 61 X 65 X 61cm; internal dimensions: 51 X 61 X 61cm) the night prior to 1647 experiments. Bats were acoustically monitored to confirm non-vocalizing status prior to experiments in order to control for the effects of activity-induced expression. To control for 1648 1649 circadian effects, all experiments were performed between 8 and 10am. Bats were administered 1650 with an overdose of pentobarbital with an intraperitoneal injection. We then rapidly opened the 1651 skull and removed the brain using round-tipped safety scissors. Brains were sliced coronally into 1652 300 µm sections in a vibrating microtome (Leica VT 1200) in ice-cold, oxygenated artificial 1653 cerebrospinal fluid [119mM NaCl, 2.5mM KCl, 1mM NaH2PO4 (monobasic), 26.2mM 1654 NaHCO3, 11mM glucose] and regions of interest were excised under a dissection microscope. 1655 Liver, muscle, and gonads were collected immediately. Tissues were preserved in a 1656 cryoprotectant medium (CryoStor CS10, Biolife Solutions) in cryovials, which we placed in a 1657 foam freezing container (CoolCell, Corning) and transferred to a -80°C freezer in order to ensure 1658 a controlled freezing rate of -1°C per minute.

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# CryoATAC-seq protocol

Tissue samples were processed as described previously (61, 122, 123) with the following minor differences in procedure and reagents. Cryopreserved samples were warmed in a 37°C water bath for 2 minutes and then transferred into 12 mL PBS supplemented with a protease inhibitor cocktail (Roche) and gently mixed by inversion. Samples were centrifuged at 300 rcf for 5 minutes at 4°C before aspirating all supernatant and resuspending the samples in ice cold lysis buffer (61). Nuclei were isolated from dissected tissues using 30 strokes of homogenization with the loose pestle (0.005 in. clearance) in 5mL of cold lysis buffer placed in a 15 mL glass Dounce homogenizer (Pyrex #7722-15). Nuclei suspensions were filtered through a 70 µm cell strainer, pelleted by centrifugation at 2,000 x g for 10 minutes, resuspended in water, and filtered a final time through a 40 µm cell strainer. Sample aliquots were stained with DAPI (Invitrogen #D1206), and nuclei concentrations were quantified using a manual hemocytometer under a fluorescent microscope. Approximately 50,000 nuclei were input into a 50 µL ATAC-seq tagmentation reaction as described previously (61, 122). The resulting libraries were amplified to 1/3 qPCR saturation, and fragment length distributions estimated by the Agilent TapeStation System showed high quality ATAC-seq fragment length periodicity. We shallowly sequenced barcoded ATAC-seq libraries at 1-5 million reads per sample on an Illumina MiSeq and processed individual samples through the ENCODE ATAC-seq pipeline (version 1.8.0, accessed at https://github.com/ENCODE-DCC/atac-seq-pipeline) for initial quality control. We used the QC measures from the pipeline (clear periodicity, library complexity, and minimal bottlenecking) to filter out low-quality samples and re-pooled a balanced library for paired-end deep sequencing on an Illumina NovaSeq 6000 System through Novogene services to target >30 million uniquely mapped fragments per sample after mitochondrial DNA and PCR duplicate removal.

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#### ATAC-seq data processing

Raw FASTQ files of ATAC-seq experiments were processed with the ENCODE ATAC-seq pipeline (version 1.8.0, accessed at https://github.com/ENCODE-DCC/atac-seq-pipeline) to

identify open chromatin region (OCR) peaks from sequenced samples. The ENCODE pipeline

was run using the mRouAeg1.p assembly (124). The pipeline was run with the default

parameters except for "atac.multimapping": 0, "atac.cap num peak": 300000,

"atac.smooth\_win": 150, "atac.enable\_idr": true, and "atac.idr\_thresh": 0.1. Filtered bam files,

peak files, and signal tracks were generated for each biological replicate and the pool of

replicates for each tissue. To account for differences in sequencing depth between samples,

reproducible peak sets were identified, which were defined as peaks with an irreproducible

discovery rate (IDR, (125)) < 0.1 across pooled pseudo-replicates. All of our samples except for

the two liver samples were high-quality, displaying high fragment length periodicity, low PCR

duplicate rates, and concordance between biological replicates.

1698 In addition to identifying peak sets for individual tissues, consensus peak sets were identified to 1699 serve as genome-wide background sets representing the intersection of the reproducible open chromatin peaks identified from all processed tissues, from all cortical samples, or from all 1700 motor cortex samples. These background sets were obtained using bedtools (version 2.25.0, 1701 1702 (126)) intersect with the -wa and -u options to combine reproducible peak sets. OCRs were prepared for downstream analysis in the following way:peaks within 50 bp of one another were 1703 1704 combined using bedtools merge, preserving the summit location as the average of the summits of 1705 all merged peaks, bedtools subtract with option -A was used to remove those peaks that were

within 1.5 kb from any annotated coding or noncoding exons, enabling us to exclude promoters,

1707 coding sequences, and noncoding RNAs from our background set. Peaks greater than 1.5 kb in

1708 width were also removed. In order to identify the complete set of exonic exclusion regions for

Egyptian fruit bat, the complete set of mRouAeg1.p annotations was used (124).

1710 To identify OCR peaks differentially active between tissues, the number of reads from each

1711 tissue that aligned to the consensus peaksets described above was quantified using featureCounts

1712 (127). The read counts at each peak between tissues were contrasted using the negative binomial

model in the DESeq2 R package (128). Differential peaks were identified based on their Wald

statistic value, with a statistical cutoff of unadjusted p < 0.05 (Data S5).

1715 In order to identify OCR orthologs across species, the consensus peaksets were aligned as well as

their peak summits across all of the species present in the Zoonomia Cactus alignment (17, 99)

using halLiftover (93) with default parameters. In the case of the bat ofM1 and wM1 data, we

1718 used consensus peaks that were common across subregions for both training and mapping. This

1719 was done so that the bat M1 data would most closely mirror the bulk motor cortex tissue

1720 collected in other species. Our conservative procedure led to strong training and evaluation of bat

motor cortex models. However, this also means that we have not directly traced the evolutionary

history of the ofM1 or wM1-specific open chromatin regions. The motor cortex was not

subdivided in any other species and so it would be impossible to validate the accuracy of these

models. The raw outputs of halLiftover were filtered and assembled into contiguous OCRs using

1725 HALPER (94) with parameters -max\_frac 2.0, -min\_len 50, -protect\_dist 5, and -narrowPeak.

The sequences underlying these OCR summit orthologs' +/- 250bp were obtained using

1727 fastaFromBed in bedtools (126).

1728 Single-nucleus ATAC-seq data for model training and evaluation was processed as described in

our previous work (23), and orthologs across species and their sequences were identified in the

same way that they were identified for OCRs from bulk ATAC-seq data. For evaluating house

mouse bulk motor cortex OCR ortholog overlap with OCRs from specific cell types, we re-

1732 processed the data (downloaded from NEMO archive:

1733 <a href="https://data.nemoarchive.org/biccn/grant/u19">https://data.nemoarchive.org/biccn/grant/u19</a> cemba/cemba/epigenome/sncell/ATACseq/mouse/

1734 ; metadata from Supplemental Table 1 of (71)) to include all cortical regions (instead of only

motor cortex) so that we could have enough reads to obtain sufficiently many peaks from each

- 1736 cell type to have the power for overlap enrichment analysis.
- 1737 The raw sequencing reads were realigned with chromap with default parameters (129). The
- 1738 fragment files from chromap were next processed with the ArchR comprehensive suite of
- algorithms for processing snATAC data (130). Doublets and empty droplets were identified with
- default recommended parameters for each biological replicate. The data were clustered together,
- and cell clusters were identified, some of which contained predominantly low-quality cells. The
- high-quality cell clusters contained cell types that were not previously annotated, cell types in
- this data were re-annotated using the published mouse whole cortex snRNA-seq cell type
- taxonomy defined by the Allen Brain Institute (131). Clusters were also visually inspected for
- 1745 quality concerns, which led to one cluster being eliminated. The addGeneIntegrationMatrix()
- 1746 function in ArchR was used to perform cell type label transfer from the mentioned snRNA-seq
- dataset to the snATAC-seq dataset, followed by manual correction of minor mapping mistakes.
- 1748 ArchR was then used to call peaks for distinct cell types across each biological replicate, and
- 1749 reproducible peaks that were detected in more than half of the biological replicates were
- identified. Finally, these peaks were filtered to retain enhancer regions, defined as regions >20k
- base pairs away from transcription start sites (TSS). This was done using a reference mouse
- genome annotation file containing TSS locations from Gencode version 15 (132) and applied to
- the ArchR reproducible peaks.
- 1755 Bat OCR Gene functional enrichment analyses
- Gene ontology analyses was performed using GREAT version 4.0.4 (133) and g:Profiler (134).
- 1757 GREAT was run on the genomic coordinates of ofM1 versus wM1 differential bat OCR peak
- sets mapped to human (hg38) using halLiftover (93) and HALPER (94) with the same
- parameters that were used for other analyses (Data S4). g:Profiler was also run to identify
- 1760 functional enrichments in genes demonstrating relative evolutionary rate convergence in vocal
- 1761 learners using default parameters (Data S6).

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- 1764 Predicting OCR ortholog open chromatin activity across species and associating predictions
- 1765 with vocal learning
- 1766 The Tissue-Aware Conservation Inference Toolkit (TACIT) was used to identify open chromatin
- 1767 regions (OCRs) whose predicted open chromatin differences between species are associated with
- differences in vocal learning (23). Specifically, the association between vocal learning and open
- 1769 chromatin predictions across 222 boreoeutherian species from Zoonomia from models trained on
- 1770 Egyptian fruit bat (this study), Brown Norway rat (20), C57BL/6J mouse (21), and Rhesus
- macaque motor cortex was investigated. If open chromatin data was available from multiple
- motor cortical regions, the intersection of OCRs was used, as described in our previous work
- 1773 (23). In addition to using the intersection of OCRs across motor cortical regions, OCRs were
- 1774 limited to regions < 1kb, > 20kb from the nearest protein-coding TSS, and not overlapping
- protein-coding exons (20), and M1-PV+ OCRs were filtered in the same way neuron (22, 23,

71). From the cross-species mapped OCR consensus peak sets, motor cortex OCRs that did not have usable orthologs in at least half of the species with vocal learning annotations (provided in Data S5) and at least three vocal learning species were removed, as these OCRs were unlikely to have sufficient power to identify associations. Likewise, motor cortex OCRs that did not have a usable ortholog in at least one species in Chiroptera (bats), the order with the largest number of mammalian vocal learners, were removed. In addition, motor cortex OCRs that did not have at least one usable ortholog in a non-chiropteran vocal learner were removed, as any association found for such OCRs would likely be explainable by phylogeny (i.e., driven by bats) and therefore not be able to be found through phyloglm (72, 73), the method that was used for associating predicted open chromatin with phenotypes while correcting for phylogeny. After applying these filtering steps, empirical p-values were calculated using phyloglm (72, 73) with phylogenetic permulations (24) and the conditional p-value method (23, 135) and were corrected using Benjamini-Hochberg (106) (Data S8).

To determine if M1-PV+ OCRs with human orthologs near autism-associated genes tended to be more significantly associated with vocal learning than other M1-PV+ OCRs with human orthologs, the human orthologs of all M1-PV+ OCRs were filtered for those of OCRs tested by TACIT. Then, autism-associated genes were obtained by downloading the complete human autism gene list from the SFARI Gene database (https://gene.sfari.org/database/human-gene/) on September 18, 2023 (82). These genes' transcription start sites were identified using human GENCODE version 27 (132). OCRs with human orthologs within 1Mb of these genes transcription start sites were found using window from bedtools version 2.29.2 (126). The uncorrected empirical p-value distribution for the OCRs tested by TACIT with human orthologs within 1Mb of the these genes was compared to that for the other OCRs tested by TACIT with human orthologs using the Wilcoxon rank-sum test. This process was repeated for motor cortex OCRs with human orthologs that were tested by TACIT, though the result was not statistically significant. Both p-values were multiplied by two for a Bonferroni correction.

To identify whether a specific clade was driving TACIT results, whether each of the vocal learning clades showed differential predicted open chromatin relative to other closely related species was tested. A t-test between each vocal learning clade within Laurasiatheria (bat, cetacean, pinniped) was used to identify differences in the distribution relative to vocal non-learning Laurasiatheria. For the human lineage, a t-test to identify whether that specific data point was likely to be drawn from a normal distribution of the Euarchonta open chromatin predictions was used (Data S10).

From a broader set of motor cortex OCRs associated with vocal learning (permulations unadjusted p < 0.05), whether or not there was a bias to overlap open chromatin from specific cortical cell types was tested. Cell type-specific open chromatin based on single-nucleus ATAC-Seq experiments from the mouse motor cortex (74) was measured and processed as described in our previous work (23). The bulk motor cortex data that the models are based on, seem to contain a higher proportion of neurons relative to glial cells. To account for potential biases in representation, an empirical p-value for the enrichment of each cell type comparing the number of overlaps observed in the true data with the number of overlaps observed in 1,000 random samples of the same number of OCRs that were not associated with vocal learning was calculated.

#### **Supplementary Text**

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transcription factors mediate behavioral differences in vocal learning across clades. At the protein-coding level, we found evidence of vocal learning-associated convergent evolution in 200 genes. The 126 genes which are evolving more slowly in vocal learning species are enriched for transcriptional regulatory function. At the noncoding level, TACIT, a machine learning approach, identified 50 candidate enhancers whose predicted open chromatin states across

Multiple lines of evidence suggest a model in which differences in gene regulatory networks of

- species is associated with vocal learning behavior, including OCRs near genes whose expression was previously associated with vocal learning (4) (Table 1 and 2). The predicted and measured
- was previously associated with vocal learning (4) (Table 1 and 2). The predicted and measured differences in open chromatin are likely due to gains and losses of transcription factor binding
- sites at those regions of open chromatin (23). These findings are consistent with previous work
- linking vocal learning to transcription factors like FOXP2(62), NEUROD6, and the MEF2 family
- 1832 (44). Although our *cis*-regulatory results clearly implicate motor cortex specializations, we
- cannot rule out that these candidate enhancers and their nearby vocal learning-associated genes
- impact a number of different brain regions and cell types.
- 1835 For both protein-coding genes and for open chromatin regions, we find greater concordance
- between vocal learning species within Laurasiatheria (bats, cetaceans, and pinnipeds) than with
- any of those clades and humans (Fig. 1B,C; Fig. 4B,C). These differences could arise because
- humans are the most distantly related vocal learning species within our analysis. In addition, it
- may be the case that our analysis is finding genetic signals of traits that are correlated with but
- not sufficient for vocal learning (2, 100)Exciting extensions to this work would be to apply our
- methodology to a continuous measure of vocal learning or to multiple vocal learning
- 1842 components.
- 1843 Across our various approaches (protein-coding analysis, bat orofacial specialized, and OCR
- analysis), the results broadly support the role of specific neurodevelopmental transcriptional
- regulatory networks in the convergent evolution of vocal learning behavior. At the pathway
- 1846 level, NFAT transcription factor signaling emerged as a common theme. The NFATC3 protein
- showed increased conservation in vocal learning mammals (RERcConverge Tau p.adj. p =
- 1848 0.0027; permulations p.adj. p = 0.019). An OCR near the NFATC2IP coding sequence showed
- differential open chromatin between wingM1 and ofM1 in Egyptian fruit bat (Data S6). The
- 1850 TACIT analysis also implicated NFAT signaling, with one of the candidate enhancers lying near
- the NFATC2 locus. Previous studies have demonstrated a potential role for NFAT signaling in
- motor degeneration in Parkinson's disease, as well as in with gene expression specializations in
- the songbird brain for vocal learning (136, 137). Further research is needed to identify the
- specific role that NFAT signaling could play in specialization of motor control regions.
- 1855 These results from mammals show similarities to results from avian vocal learners at specific
- genes. The behavior of vocal learning is associated with shared patterns of gene expression,
- especially decreases in expression, in the vocal motor control region (4). Using TACIT, we were
- able to find several candidate enhancers with lower predicted motor cortex open chromatin levels in vocal learning mammals relative to vocal non-learning mammals near these convergently
- evolved genes. These regulatory elements could contribute to the convergent evolution in the
- expression of *DAAM1*, *VIP*, and *SCNA* genes that are shared between human and song-learning
- birds. Out of the 50 candidate convergently evolved enhancers, several others showed
- 1863 concordance with expression patterns currently observed only in songbirds, including OCRs
- predicted to be more open in vocal learners near KCNQ5 and CNTNAP4 and OCRs predicted to

1865	be more closed in vocal learners near PRKCE, MAGI2, CDH8, and ATXNI (137). Finer- grained
1866	single -cell profiling of gene expression between vocal motor and other subregions of M1 may
1867	provide insight into whether any of these specializations are shared with human or other
1868	mammalian vocal learners.

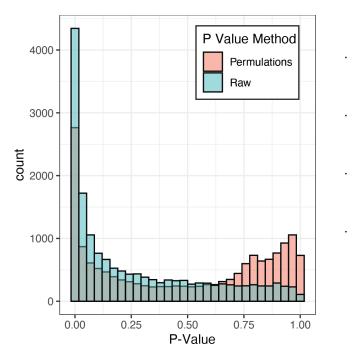


Fig. S1. Histogram comparison of raw and permulation RERconverge P-values. RERconverge was run for aligned proteins across the Zoonomia mammal dataset. A histogram of RERconverge p-values is shown, which was obtained by applying to the Kendall rank correlation of the gene relative evolutionary rates to the trait of vocal learning (*blue*). After running RERconverge, phylogenetic permulations was applied to obtain a p-value adjusted for the distribution of the vocal learning trait across the phylogenetic tree (*red*). The application of permulations reduced the number of candidate vocal learning-associated genes, as indicated by the shift of p-values towards less significance.

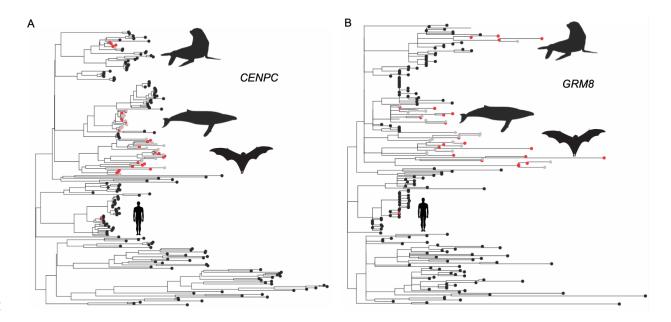


Fig. S2. Example genes implicated by RERConverge include both CENPC (Conserved) (A) and GRM8 (Accelerated) (B), which exhibit relative evolutionary rate (RER) shifts in vocal learning mammals (red) relative to vocal non-learners (black). Branch lengths illustrate the evolutionary rate of the protein-coding genes on the branches. Species with non-confident annotations of behavior are shown in gray.

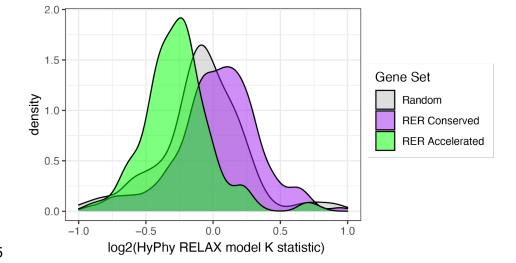
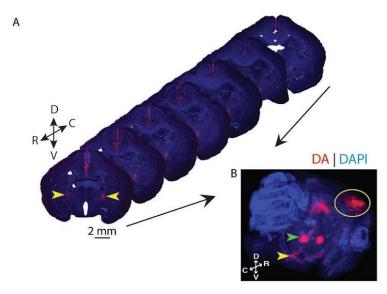
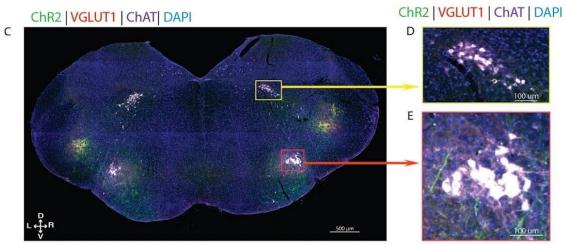
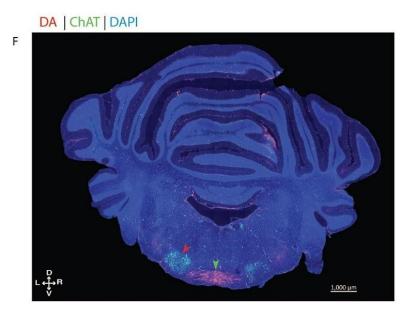


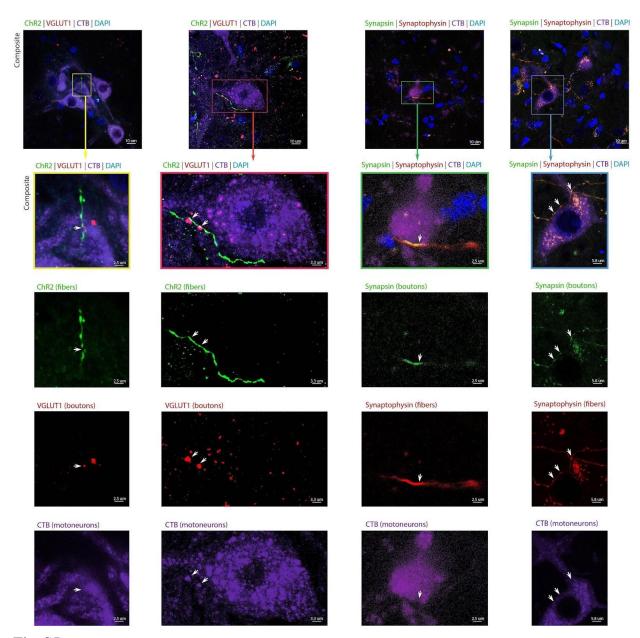
Fig. S3. 1896 1897 Comparison of genes implicated by RERconverge and HyPhy RELAX. 1898 RERconverge was run for aligned proteins across the Zoonomia mammal dataset (see methods). 1899 Genes were split between conserved (purple, RERconverge p < 0.01; permulations p < 0.01, tau 1900 < 0) and accelerated (green, RERconverge p < 0.01; permulations p < 0.01, tau > 0). 1901 Additionally, a set of genes outside those groups with the same length distribution as genes from 1902 vocal learners and non-learners was chosen as a control (gray). For each group, the density plot 1903 shows the distribution of the HyPhy RELAX model K statistic, which will be positive if there is 1904 more constraint within vocal learning clades and negative if there is more constraint outside 1905 vocal learning clades. 1906 1907





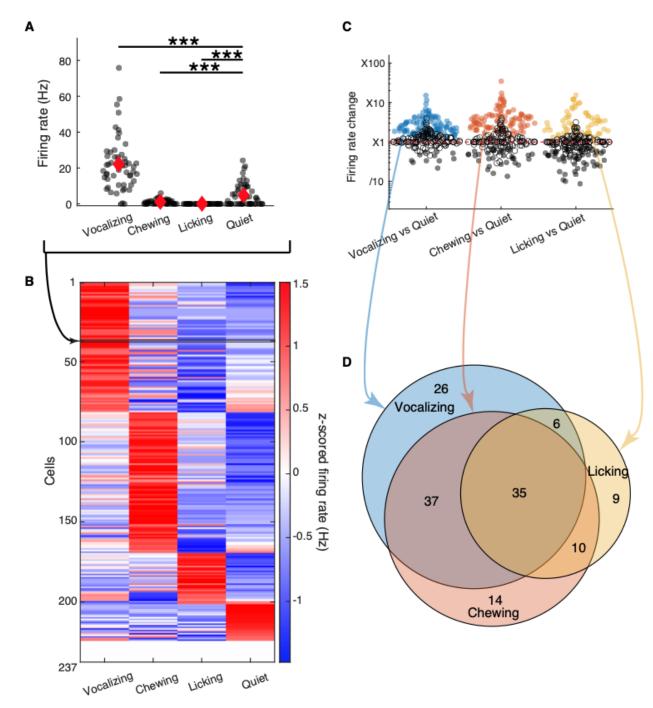


1909 Fig. S4. 1910 Reconstruction of anterograde and retrograde tracing from orofacial motor cortex (ofM1). (A) 1911 Example images from one bat injected bilaterally with dextran amine tracer in ofM1 showing 1912 seven sequential coronal planes separated by 240 µm, aligned by hand and stacked into a 3D 1913 volume shown in (B). The anterograde propagation of the tracer (red) is visible in the pyramidal 1914 tract (yellow arrows) against the background DAPI stain (blue). (B) 3D side view of the stacked 1915 images showing the bilateral anterograde propagation of the tracer (red) projecting from ofM1 1916 (*yellow circle*) down the pyramidal tract (*yellow arrow*) to the brainstem. Retrograde propagation 1917 of the tracer shows cell bodies in the thalamus (green arrow) that send afferents into ofM1. See 1918 Movie S1 for a 3D rotational view of the image stack. (C-E) Anterograde tracers are not found in any other brainstem motor nuclei. (C) Coronal slice in brainstem following bilateral injections in 1919 1920 ofM1 with rAAV5/CamkII-hChR2(H134R)-EYFP (ChR2) labeling fibers in green. Synaptic boutons were histologically labeled with VGLUT1 in red. Bilateral retrograde injections of CTB 1921 1922 (white) were delivered within the same bat into the cricothyroid muscles to label laryngeal 1923 motoneurons in NA, and in nearby neck and tongue muscles, which enabled the labeling of cells 1924 in the hypoglossal nucleus. (**D**) Magnification of hypoglossal nucleus with motoneurons 1925 controlling the neck and tongue muscles (white). (E) Magnification of NA with larvngeal motoneurons (white). Note the absence of cortical axons labeled with the anterograde tracer 1926 1927 within the hypoglossal nucleus or any other motor nuclei within the medulla except for NA. (F) 1928 Example coronal section from Egyptian fruit bat medulla showing the decussation of the 1929 pyramidal tract (green arrowhead) labeled by anterograde injections of fluorescent dextran 1930 amine into ofM1. Note the decussation occurs at the level of the facial nucleus (red arrowhead), 1931 rostral to the location of nucleus ambiguus. Motoneurons are labeled with choline 1932 acetyltransferase (ChAT) in green. 1933 1934 1935



**Fig. S5.** Examples of triple colocalization between corticobulbar fibers, synaptic boutons, and laryngeal motoneurons.

(First 2 columns) Example confocal images of nucleus ambiguus (NA) showing triple colocalization (*white arrows*) between corticobulbar axons from ofM1 labeled with rAAV5/CamkII-hChR2(H134R)-EYFP (ChR2, *green*), CTB-labeled cricothyroid motoneurons (*purple*), and VGLUT1-labeled presynaptic boutons (*red*) with a background DAPI stain (*blue*). (Last 2 columns) Example confocal image of NA showing triple colocalization (*white arrows*) between corticobulbar axons from ofM1 labeled with AAV<sub>DJ</sub>-hsyn-mRuby2-T2S-Synap-eGFP (*fibers in red, boutons in green*), and CTB-labeled cricothyroid motoneurons (*purple*), with a background DAPI stain (*blue*). Across all columns, the first two rows are composite images, with the second row being insets of the first row. The three last rows depict three of the four individual channels merged to obtain the insets of the second row.



**Fig. S6.**ofM1 neurons activity is modulated by vocalizing, chewing and licking. **(A)** Time average firing rate for an example ofM1 neuron during vocalization production (n=53 vocalizations) and duration matched period of time when the bat was chewing, licking, or quiet and immobile. Black dots are individual events, red diamonds are mean +/- SEM. The neuron was significantly excited during vocal production as compared to quiet and significantly inhibited during chewing and licking as compared to quiet (Anova on Poisson GLM, all p<0.001). **(B)** Z-scored firing rate of ofM1 neurons, the activity of which could be evaluated

during vocalizing, chewing, licking, and quiet (n=237). (C) Change in firing rate during orofacial 1961 1962 motor actions as compared to quiet for ofM1 neurons. Each circle represents a neuron. The significance of the Anova on Poisson GLM is signified by filled circles (p<0.001 after FDR 1963 1964 correction; n=237). As expected, neurons in the orofacial motor cortex are modulated during orofacial motor actions as compared to quiet periods (D) Venn Diagram displaying the number 1965 of neurons that were excited (significant increase in firing rate as compared to quiet; Anova on 1966 Poisson GLM and FDR corrected p-value < 0.001) by vocalizing, chewing, or licking (n=137). 1967 1968 26 neurons were excited by vocal production only and not by chewing or licking.



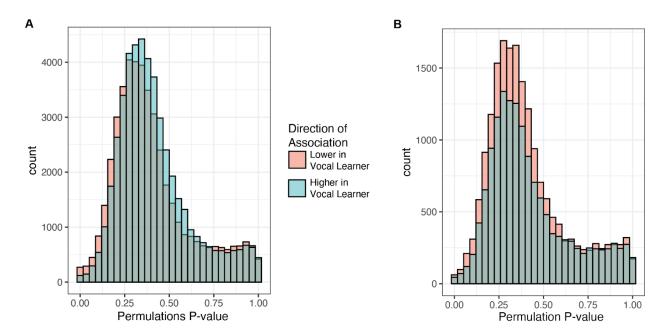


Fig. S7.

TACIT identified stronger associations between regions that are predicted to have lower chromatin activity and the vocal learning ability of species. Histograms of TACIT permulation p-values are plotted separately for OCRs whose chromatin is predicted to be less open (*red*) vs. more open (*blue*) in vocal learners as opposed to vocal non-learners. Results are plotted separately for motor cortex (A) and parvalbumin positive inhibitory interneuron (B).

OCR Ortholog Coordinates (mm10)	TACIT P-Value	TACIT Coef.	Permu- lations P-Value	Permu- lations BH Adj. P-Value	Gene Name	RER Conv. Result	Additional Genes
chr15:16107524-16107960	1.71E-02	-10.32	1.40E-05	0.05	Cdh9 (138)	No	-
chr18:65510387-65511111	3.99E-02	-9.46	3.70E-05	0.09	Zfp532	No	Malt1
chr2:131650717-131651399	1.44E-02	-10.89	3.70E-05	0.09	Adra1d	No	Smox
chr9:8176928-8177666	1.56E-02	-9.66	4.20E-05	0.09	Cep126	No	
chr18:68703945-68704585	1.46E-02	-13.68	3.50E-05	0.09	4930546 C10Rik	No	Tcf4 (139),
chr13:48323987-48324750	2.84E-02	15.06	2.90E-05	0.08	ld4	No	
chr10:5735065-5735721	2.73E-02	-10.25	2.50E-05	0.07	Fbxo5	No	VIP
chr19:56747655-56748356	1.70E-02	-9.31	1.10E-05	0.05	Adrb1	No	
chr18:15238033-15239015	2.36E-02	-9.96	6.00E-06	0.04	Kctd1	No	AQP4
chr16:65629047-65629933	2.51E-02	-10.04	9.00E-06	0.04	Chmp2b ( <i>140</i> )	No	VGLL3
chr17:94243749-94244469	1.35E-02	-7.06	2.00E-05	0.07	Mettl4	No	
chr8:60393577-60394004	2.54E-02	-9.83	2.60E-05	0.07	Gm10283	No	
chr2:168519584-168519969	3.82E-02	-7.50	3.00E-06	0.04	Nfatc2	No	
chr17:72819735-72820272	6.78E-06	-5.90	4.40E-05	0.10	Ypel5	No	ALK
chr7:37573064-37573594	4.61E-03	-11.53	1.00E-06	0.02	Zfp536	No	Tshz3 (141–143)
	1.33E-02	-12.10	9.00E-06	0.04	NA	No	ZNHIT6, CCN1

	1.10E-02	10.44	2.80E-05	0.08	NA	No	
chr12:75463900-75464505	3.10E-02	12.42	9.00E-06	0.04	Gphb5	No	PPP2R5E (144)
chr2:145057020-145057789	2.14E-02	-5.83	2.00E-06	0.03	Slc24a3 ( <i>145</i> )	No	Dtd1
chr15:61299599-61300306	1.91E-02	-10.06	4.50E-05	0.10	A1bg	No	Мус
chr18:25702100-25702785	1.86E-03	-6.47	1.00E-06	0.02	Celf4 (76)	p<0.01	
chr8:100212225-100213081	1.44E-02	-14.07	1.50E-05	0.05	<b>Cdh8</b> ( <i>146</i> )	p<0.1	
chr13:26914286-26915174	1.77E-02	-11.20	1.50E-05	0.05	Prl	No	HDGFL1
chr7:96710377-96711352	2.92E-02	-9.48	3.70E-05	0.09	Tenm4	No	Nars
	3.39E-02	-6.45	1.00E-06	0.02	NA	No	ZNF704
chr14:28101682-28102178	2.39E-02	-10.64	3.20E-05	0.08	Erc2	No	Wnt5a
chr2:61494077-61494853	2.69E-03	15.75	1.00E-06	0.02	Tank (147)	No	Tbr1 (148, 149)
chr17:63914921-63915759	2.01E-02	-11.71	1.40E-05	0.05	Fer	No	
chr12:97833162-97833781	5.23E-03	-14.31	1.50E-05	0.05	Galc ( <i>150</i> )	p<0.01	
chr2:33543268-33544008	2.30E-02	13.70	1.00E-05	0.05	Zbtb43	No	<u>Lmx1b</u> (151)
chr8:18790099-18790817	4.64E-02	-10.55	2.50E-05	0.07	Angpt2	No	
	3.85E-02	-9.41	4.00E-06	0.04	NA	No	SORCS3
chr5:20750587-20751246	1.33E-02	-10.20	5.00E-06	0.04	Phtf2	No	<b>Magi2</b> (152)
	5.68E-03	10.75	8.00E-06	0.04	NA	No	DPPA4
chr12:86795973-86796457	3.78E-07	5.92	5.00E-06	0.04	Lrrc74a	No	
	2.48E-03	6.51	8.00E-06	0.04	NA	p<0.1	Galc (150)

					Prkce		
chr17:86429786-86430568	6.81E-03	-11.21	2.50E-05	0.07	(153)	No	Epas1
chr1:21773189-21773501	1.27E-03	6.73	2.00E-05	0.07	Kcnq5 (154)	No	
chrX:87178627-87179266	3.10E-02	5.49	2.00E-06	0.03	ll1rapl1 ( <i>155</i> )	No	
	5.14E-05	7.45	2.00E-06	0.03	NA	p<0.01	<b>DAAM1</b> (15 6), DACT1 (157)
chr12:67321023-67321586	5.17E-02	-15.62	1.70E-05	0.06	Mdga2 (158)	No	
chr17:86352256-86352626	7.58E-02	-6.33	5.00E-06	0.04	<b>Prkce</b> (153)	No	Epas1
chr5:18822320-18822911	1.62E-01	-3.66	7.00E-06	0.04	<b>Magi2</b> ( <i>152</i> )	No	
chr8:35222316-35223646	1.23E-01	-10.39	3.40E-05	0.09	Ppp1r3b	No	

# Table S1. M1 OCRs whose predicted open chromatin in boreoeutherian mammals is significantly associated with vocal learning.

Coordinates are reported for the mouse genome (mm10). The coefficients shown are those reported by phyloglm for the association between the vocal learning trait and the OCR orthologs' open chromatin prediction. The permulations p-values represent the adjusted p-values after Benjamini-Hochberg correction. Nearest genes are indicated, with genes previously associated (in some cases specifically, in other cases as part of larger deletions) with speech delay or disability underlined (with associated references) and genes previously shown to be convergently regulated in humans and song-learning birds (4) in bold red. Genes that have vocal learning associated specialization in the songbird brain are in bold black (137). A summary of the RERConvege unadjusted p-value for the nearest annotated mouse gene is provided. Human coordinates in the hg38 assembly are used when a mouse ortholog does not exist.

OCR Ortholog Coordinates (mm10)	TACIT P-Value		Permu- lations P-Value	Perm. BH Adj. P-Value	Gene Name	RER Converge Result	Additio nal Genes
chr13:45770571.45771071	2.32E-02	-8.44	1.40E-05	0.08	<b>Atxn1</b> (159, 160)	No	GMPR
chr2:42511067.42511567	1.26E-03	-14.22	6.00E-06	0.05	Lrp1b (161)	No	
chr2:56025027.56025527	3.21E-02	-7.40	3.00E-06	0.03	Kcnj3 (162)	No	
chr2:148111903.148112403	5.08E-04	-14.11	1.00E-06	0.03	Foxa2	No	
chr6:61939127.61939627	5.49E-04	-14.67	2.00E-06	0.03	Ccser1	p<0.1	<b>Snca</b> (163, 164)
chr8:112642458.112642958	2.45E-02	7.48	1.20E-05	0.08	<b>Cntnap4</b> (81)	No	Mon1b

Table S2. PV OCRs whose predicted open chromatin state in boreoeutherian mammals is significantly associated with vocal learning.

1994 Coordinates are provided for the mouse (mm10) OCR orthologs; otherwise this table is in the same format as Table S1.

- 1996 **Movie S1.**
- 1997 3D rotational view of fluorescent dextran amine bilaterally injected in ofM1. See legend of
- 1998 Figure S4A-B for details. The anterograde propagation of the fluorescent tracer dextran amin
- 1999 (red) is visible from the injection point to the pyramidal tract and down to the medulla. A
- background DAPI stain (blue) is used to better visualize the brain. Note, that the decussation of
- 2001 the pyramidal tract, as further depicted in Fig. S4F, is rostral to the location of NA.

- 2003 DataS1. (separate file: speciesAnnotations.4.1.7.1.csv)
- Table of vocal learning annotations across mammals considered in this study. Species are
- referenced based on their ID in the TOGA alignment, propername, and NCBI accession ID.
- 2006 DataS2. (separate file: vocalLearningRERConvergeMaster.csv)
- 2007 This file contains the output from running RERconverge for the vocal learning trait on the
- 2008 Zoonomia genomes. The "Rho" column corresponds to a rank sum test between vocal learners
- and non-learners. Positive values have higher evolutionary rates in vocal learning species. The
- 2010 "permP" column includes p-values after conducting permutations. Both the RERconverge and
- 2011 permutations p-values were adjusted using Benjamini Hochberg. The final four columns
- 2012 correspond to the Bayes Factor calculated for that group of species. Bayes Factors > 5 are
- 2013 considered a strong signal of selection.
- 2014 DataS3. (separate file: HyPhyResults.1.xlsx)
- This file contains the output from running HyPhy on selected set of vocal learning genes.
- 2016 1 ID: gene
- 2017 2 Sequences: number of sequences in the alignment
- 2018 3 Sites: length of the alignment
- 2019 4 FG q-value: q-value (Benjamini-Hochberg) of the BUSTED test on foreground branches
- 2020 5 BG q-value: q-value (Benjamini-Hochberg) of the BUSTED test on background branches
- 2021 6 DIFF q-value: q-value (Benjamini-Hochberg) for the difference between FG and BG
- 2022 7 S.Sites (FG): Sites that are showing support for positive selection in the Foreground branches
- with evidence ratio of 100 or higher
- 8 S.Sites (BG): Sites that are showing support for positive selection in the Background branches
- with evidence ratio of 100 or higher
- 2026 9 L (FG): tree length foreground
- 2027 10 L (BG): tree length background
- 2028 11 RELAX q-value
- 2029 12 RELAX K: the foreground is modeled as (omega) K so it matters if K<1 or K>1
- 2030 13 -18 Foreground parameters for BUSTED-PH
- 2031 19 PH: Selection associated with phenotype
- 2032 20 TYPE: Binary summary of 4,5,6.

2033 2034 DataS4. (separate file: GeneOntology.1.xlsx) This file contains the result of running gene set enrichment analysis using ENRICHR. Different 2035 tabs correspond to different categories (GO Biological Process; Human Phenotype) and different 2036 2037 gene sets (Conserved and Accelerated). The input genes are the intersection of significant genes 2038 from RERconverge and HyPhy. 2039 2040 2041 DataS5. (separate file: BatM1OcrS.2.xlsx) 2042 The file contains a table of 348 open chromatin regions (OCRs) discovered from ATAC-seq 2043 analyses with differential activity between the bat orofacial and wing subregions of primary 2044 motor cortex (ofM1 and wM1, respectively). The first four columns provide locations in the 2045 Rousettus aegyptiacus genome assembly (mRouAeg1.p), with the fourth column presenting the relative position of the ATAC peak summit in the OCR. Columns 5-9 provide output of DEseq2 2046 2047 analysis (128). Proximal genes were identified from the mRouAeg1.p assembly gene annotations 2048 (124). The full list of TACIT vocal learning associated peaks (intersections from which are 2049 provided in column 10) can be found in Tables S1 and S2. The full list of genes under RER 2050 convergence in vocal learning mammals (intersections from which are provided in column 11) 2051 can be found in Data S1. 2052 2053 DataS6. (separate file: BatM1OcrGo.1.xlsx) 2054 This file contains the results of the gene functional enrichment analyses performed using 2055 GREAT on the OCRs differential between ofM1 and wM1 (provided in Data S3) mapped to the 2056 human genome assembly (hg38), prepared as described in the Materials and Methods section "Gene functional enrichment analyses." The file contains two tabs, one for the enriched terms 2057 2058 from GO Biological Processes (Data S4A GOBiological Process) and the other for those from the set of GO Molecular Functions (Data S4B GOMolecularFunction). 2059 2060 2061 DataS7. (separate file: VocalLearningTACITResults.1.xlsx) 2062 The results of running TACIT for vocal learning on motor cortex and PV+ inhibitory interneurons. The "Exp Pvalue" is calculated using permutations and then adjusted using 2063 2064 Benjamini Hochberg (bh column). The number of permutations is the "Trials" column. 2065 2066 DataS8. (separate file: TACIT ttest.zip) 2067 This archive contains a series of .csv files that hold the results of t-test applied across motor 2068 cortex (mcx) and PV+ interneuron (pv) open chromatin. Each vocal learning clade is calculated 2069 separately. 2070 2071 DataS9. (separate file: OCRMouseGeneAnnotation.1.xlsx)

This file contains a subset of results from Data S7 that pass the significance threshold. The HOMER pipeline was used to annotate the nearest mouse gene. Coordinates are reported for the mouse genome (mm10). The coefficients shown are those reported by phyloglm for the association between the vocal learning trait and the OCR orthologs' open chromatin prediction. The permulations p-values represent the adjusted p-values after Benjamini-Hochberg correction. Nearest genes are indicated, with genes previously associated (in some cases specifically, in other cases as part of larger deletions) with speech delay or disability underlined (with associated references) and genes previously shown to be convergently regulated in humans and song-learning birds (4) in bold red. Genes that have vocal learning associated specialization in the songbird brain are in bold (137). A summary of the RERConvege pvalue for the nearest annotated mouse gene is provided. Human coordinates are used when a mouse ortholog does not exist.

# DataS10. (separate file: CellTypeEnrichments.1.xlsx)

This file contains a subset of results of exploring cell type-specific open chromatin intersecting peaks of interest from the bulk result. The number of overlapping peaks is given separately for the sets that have higher or lower predicted open chromatin in vocal learners (vl\_down\_num, vl\_up\_num). A permutation test is used to identify the significance of the overlap. The permutation p-value is provided along with the log2 of the enrichment over the expected value from the permutation test.

#### **References and Notes:**