Repository of the Max Delbrück Center for Molecular Medicine (MDC) in the Helmholtz Association

https://edoc.mdc-berlin.de/19928/

Cultural transmission of vocal dialect in the naked mole-rat

Barker A.J., Veviurko G., Bennett N.C., Hart D.W., Mograby L., Lewin G.R.

This is the final version of the manuscript. The original article has been published in final edited form in:

Science 2021 JAN 29 ; 371(6528): 503-507 doi: 10.1126/science.abc6588

Publisher: American Association for the Advancement of Science (AAAS)

Copyright © 2021 The Authors, some rights reserved; exclusive licensee American Association for the Advancement of Science. No claim to original U.S. Government Works.

Publisher's Notice

This is the author's version of the work. It is posted here by permission of the AAAS for personal use, not for redistribution. The definitive version was published in: Science **371**, 6528, (503-507), (2021).

https://doi.org/10.1126/science.abc6588

Cultural transmission of vocal dialect in the naked mole-rat

Authors: Alison J. Barker^{1*}, Grigorii Veviurko^{1,†}, Nigel C. Bennett², Daniel W. Hart², Lina Mograby¹, Gary R. Lewin^{1*}.

Affiliations: 5

¹Department of Neuroscience, Max Delbrück Center for Molecular Medicine, Berlin, Germany.

²Mammal Research Institute, Department of Zoology and Entomology, University of Pretoria, Pretoria, Republic of South Africa.

*Correspondence to: glewin@mdc-berlin.de, Alison.Barker@mdc-berlin.de

[†]Present address: Department of Electrical Engineering, Mathematics and Computer Science, Delft 10 University of Technology, Delft, Holland.

Abstract: Naked mole-rats form some of the most cooperative groups in the animal kingdom, living in multi-generational colonies under the control of a single breeding queen. Yet, how they maintain this remarkable social organization is unknown. Here we show that the most common naked mole-rat vocalization, the soft chirp, is used to transmit information about group membership, creating distinct colony dialects. Audio playback experiments demonstrate individuals make preferential vocal responses to home colony dialects. Pups fostered into foreign colonies in early postnatal life learn the vocal dialect of their adoptive colonies suggesting vertical transmission and flexibility of vocal signatures. Dialect integrity is in part controlled by the queen, as loss of a queen decreases dialect cohesiveness which remerges only with the ascendance of a new queen.

One Sentence Summary: Vocal culture in naked mole-rats supports colony identity and cohesion.

Main Text: The naked mole-rat (*Heterocephalus glaber*) was the first identified eusocial mammal 25 (1) and has received much attention for an array of extreme physiological traits (2-4). Yet often overlooked is their constant peeping, chirruping, and grunting (5, 6) (Audio S1-S2). Complex patterns of acoustic communication exist throughout the animal kingdom (7) and decades of study: notably in songbirds (8), bats (9), cetaceans (10), and primates (11) have generated debate about 30 the etiology of human language with compelling evidence for anatomical (12) genetic (13) and

15

cultural (14) drivers. The highly cooperative nature of naked mole-rat societies led us to investigate whether their vocalizations support social complexity.

The vocal repertoire of the naked mole-rat consists of 17-25 distinct vocalizations (6, 15). The most common vocalization, the soft chip, serves as a greeting call previously shown to occur in a stereotyped call and response, i.e. antiphonal manner (16).

We recorded 36,190 soft chirps from 166 animals (7 colonies), housed in Berlin, Germany or Pretoria, South Africa over a period of two years. We developed an algorithm to automatically segment, trace and extract acoustic features of individual soft chirps (Fig. 1A). In developing our analysis pipeline, we included established parameters for vocalization analysis (*17*) and whenever possible spectrogram-extracted features which minimized variable background noise from recordings made across different locations and days. Using a type of supervised machine learning, the Random Forest Classifier (*18*) we analyzed eight soft chirp features, three from the soundwave (pitch, wiener entropy, and zero-crossings rate) and five from the soft chirp spectrogram (asymmetry, peak frequency, height, duration and slope) (Fig. 1A, fig.S1). Training the classifier with soft chirps from individual mole-rats we found it could reliably predict the identity of individuals within a colony (Fig. 1B, fig.S2).

Within naked mole-rat colonies reproductive suppression of nearly all colony members is necessary to sustain the colony with limited food resources and leads to strong xenophobia (19). As such, multiple mechanisms for maintaining the social integrity of the colony and for detecting intruders might be necessary. We next tested for colony-specific signatures (16) using soft chirps recorded from three colonies in Berlin (Colonies, B, M and T) and a fourth colony which has always been located in South Africa, (Colony D). Again, using a Random Forest classifier, we found that soft chirp features were highly predictive of colony identity (Fig. 2A,B fig. S3-S5) with asymmetry and peak frequency found to be the best spectrogram-derived features for colony separation (Fig.2C,D, 16). While we did not find rank, age or sex to be strongly predicted by soft chirp features (fig. S6), we observed a positive correlation with body size and soft chirp pitch (Supplementary Text, fig. S7).

We next tested if animals recognize information communicated via soft chirps. To test this, we employed a place preference assay in which individual animals were given access to two interconnected chambers. (Fig.3A top, Video S1), each equipped for simultaneous audio playback

10

15

5



25

and recording. Animals preferred to spend most time in the chamber with sound presentation regardless of which colony soft chirp playback (home or foreign) was played (Fig.3A). Animals frequently vocalized in response to the audio playback stimulus with their own soft chirp, consistent with the antiphonal behavior previously described (Fig. 3B) (*16*, *20*). We observed very high responses rates when animals were presented with home colony audio playbacks, much higher and significantly different compared to responses to foreign colony playbacks (Fig. 3C).

Naked mole-rats might recognize individual voices from home colonies rather than colony dialects. To test this, we designed artificial stimuli, using two features: asymmetry and peak frequency (*16*). Artificial stimuli were designed such that our colony classifier categorized these vocalizations as "mock colony members" but did not overlap with any known individual in the colony (fig.S8). Remarkably, response rates were again higher for the mock home stimulus suggesting that naked mole-rats can distinguish colony specific features in vocalizations (Fig.3D,E, fig.S8). To test if peak frequency or asymmetry alone were sufficient for behavioral preference, we used a pure tone of 4.5 kHz (mean colony peak frequency) and a frequency-doubled stimulus (9.0 kHz with mean colony asymmetry). We observed responses to the pure tone alone, but virtually none to the frequency-doubled stimulus (Fig.3E, fig.S8). The preferential response to home colony dialect was still found in the presence of a conflicting olfactory cue in the test chamber (Fig. 3F).

If naked mole-rats use distinct colony dialects to differentiate themselves from neighboring colonies or as a mechanism for ensuring conformity within the colony, such dialects must be maintained across generations. We cross-fostered three individuals between colonies, a non-trivial task as queens are rare breeders that cannot be synchronized across colonies. An abandoned pup (pup Mi) was cross fostered from Colony T to Colony M (Fig.4A,D) and we simultaneously tracked two surviving foster-siblings born in Colony M (pups Ob and Ny, Fig. 4, B, C). In a second experiment, two orphaned pups (pups Da and Jo, Colony S) were fostered into two different colonies (Colony M and Colony T, respectively) (Fig.4E-G). We observed that adult vocalizations fully develop by about ~ 3 months (Supplementary Text, fig. S9, S10), so we examined pup dialects at time points later then 6 months post-fostering. We tested foster pup vocalizations on our colony classifier, which classified the pups as belonging to one of five test colonies (including birth and foster colonies). In all three successful foster experiments, the new colony dialect was adopted with correct prediction rates between 59-95% (Fig. 4H).

Finally, we investigated if the queen's presence might influence the vocal signature of the colony (Supplementary Text). During the course of this study Colony S consecutively lost two queens (Fig.4I, fig. S11) allowing us to record soft chirps during queen epochs and subsequent periods of anarchy. Individual variability of several features including peak frequency was higher during periods of anarchy (Fig. 4J, fig. S12) and classification accuracy of the colony dialect decreased during periods of anarchy (Fig. 4K, fig. S12), suggesting the presence of the queen enhances dialect cohesiveness.

Acoustic communication of social information has been observed in multiple mammalian species: bats (9), primates (11), cetaceans (10), pachyderms (21), and carnivores (22) and here we expand this group to include a member of the order Rodentia. More work is needed to resolve if naked mole-rats are capable of true production learning as exemplified in songbirds or if like many nonhuman primates they are exceptionally good usage learners (23). With a simple vocal greeting, humans convey individual identity (unique voice) and cultural identity (dialect usage), and here we show that naked mole-rats also signal social membership with dialect usage. Dialect features can be transmitted across generations a remarkable and hitherto undescribed feat for a rodent species, supporting an accumulating body of evidence that social complexity evolved hand in hand with vocal complexity.

References

- 1. J. Jarvis, Eusociality in a mammal: cooperative breeding in naked mole-rat colonies. *Science*. **212**, 571–573 (1981).
 - T. J. Park, J. Reznick, B. L. Peterson, G. Blass, D. Omerbašić, N. C. Bennett, P. H. J. L. Kuich, C. Zasada, B. M. Browe, W. Hamann, D. T. Applegate, M. H. Radke, T. Kosten, H. Lutermann, V. Gavaghan, O. Eigenbrod, V. Bégay, V. G. Amoroso, V. Govind, R. D. Minshall, E. S. J. Smith, J. Larson, M. Gotthardt, S. Kempa, G. R. Lewin, Fructose-driven glycolysis supports anoxia resistance in the naked mole-rat. *Science*. **356**, 307–311 (2017).
 - 3. J. G. Ruby, M. Smith, R. Buffenstein, Naked Mole-Rat mortality rates defy gompertzian laws by not increasing with age. *Elife*. **7** (2018), doi:10.7554/eLife.31157.
 - 4. R. Buffenstein, S. Yahav, Is the naked mole-rat Hererocephalus glaber an endothermic yet poikilothermic mammal? *Journal of Thermal Biology*. **16**, 227–232 (1991).
 - 5. W. C. O. Hill, A. Porter, R. T. Bloom, J. Seago, M. D. Southwick. Field and Laboratory studies on the Naked Mole Rat, Heterocephalus glaber. *Proceedings of the Zoological Society of London.* **128**, 455–514 (1955).

20

5

10

15

30

- J. W. Pepper, S. H. Braude, E. A. Lacey, P. W. Sherman, in *The Biology of the Naked Mole-Rat*, P. W. Sherman, J. U. M. Jarvis, R. D. Alexander, Eds. (Princeton University Press, Princeton, 1991; pp. 243–274.
- 7. M. S. Brainard, W. T. Fitch, Editorial overview: Communication and language: Animal communication and human language. *Current Opinion in Neurobiology*. **28**, v–viii (2014).
- 8. A. J. Doupe, P. K. Kuhl, Birdsong and human speech: common themes and mechanisms. *Annu. Rev. Neurosci.* **22**, 567–631 (1999).
- 9. M. Knörnschild, Vocal production learning in bats. *Current Opinion in Neurobiology*. **28**, 80–85 (2014).
- 10 10. V. M. Janik, Cetacean vocal learning and communication. *Current Opinion in Neurobiology*. **28**, 60–65 (2014).
 - 11. R. M. Seyfarth, D. L. Cheney, The evolution of language from social cognition. *Current Opinion in Neurobiology*. **28**, 5–9 (2014).
 - 12. P. Lieberman, The Evolution of Human Speech: Its Anatomical and Neural Bases. *Current Anthropology*. **48**, 39–66 (2007).
 - G. Konopka, J. M. Bomar, K. Winden, G. Coppola, Z. O. Jonsson, F. Gao, S. Peng, T. M. Preuss, J. A. Wohlschlegel, D. H. Geschwind, Human-specific transcriptional regulation of CNS development genes by FOXP2. *Nature*. 462, 213–217 (2009).
 - 14. Dunbar, Robin, The social brain hypothesis. *Evolutionary Anthropology*. **6**, 178–190 (1998).
 - 15. Barker AJ, Koch U, Lewin GR, Pyott SJ. 6. Hearing and vocalizations in the naked molerat. In: The Extraordinary Biology of the Naked Mole-Rat. 2nd. edn. Springer *(in press)*.
 - S. Yosida, K. I. Kobayasi, M. Ikebuchi, R. Ozaki, K. Okanoya, Antiphonal Vocalization of a Subterranean Rodent, the Naked Mole-Rat (Heterocephalus glaber). *Ethology*. **113**, 703– 710 (2007).
 - D. Lipkind, O. Tchernichovski, Quantification of developmental birdsong learning from the subsyllabic scale to cultural evolution. *Proceedings of the National Academy of Sciences*. 108, 15572–15579 (2011).
- 30 18. Materials and methods are available as supplementary materials at the Science website.
 - 19. M. J. O'Riain, J. U. M. Jarvis, Colony member recognition and xenophobia in the naked mole-rat. *Animal Behaviour*. **53**, 487–498 (1997).
 - 20. S. Yosida, K. Okanoya, Naked Mole-Rat is Sensitive to Social Hierarchy Encoded in Antiphonal Vocalization. *Ethology*. **115**, 823–831 (2009).

15

20

25

35

5

- 21. A. S. Stoeger, P. Manger, Vocal learning in elephants: neural bases and adaptive context. Current Opinion in Neurobiology. 28, 101–107 (2014).
- 22. K. E. Holekamp, S. T. Sakai, B. L. Lundrigan, Social intelligence in the spotted hyena (Crocuta crocuta). Philos. Trans. R. Soc. Lond., B, Biol. Sci. 362, 523-538 (2007).
- 23. V. M. Janik, P. J. B. Slater, The different roles of social learning in vocal communication. Animal Behaviour. 60, 1–11 (2000).
 - 24. L. Buitinck, G. Louppe, M. Blondel, F. Pedregosa, A. Mueller, O. Grisel, V. Niculae, P. Prettenhofer, A. Gramfort, J. Grobler, R. Layton, J. Vanderplas, A. Joly, B. Holt, G. Varoquaux, API design for machine learning software: experiences from the scikit-learn project. arXiv:1309.0238 [cs] (2013) (available at http://arxiv.org/abs/1309.0238).
 - 25. T. Honkela, Ed., Artificial neural networks and machine learning -- ICANN 2011: 21st International Conference on Artificial Neural Networks, Espoo, Finland, June 14-17, 2011: proceedings (Springer, Heidelberg, 2011), Lecture notes in computer science.
- 26. B. McFee, M. McVicar, C. Raffel, Dawen Liang, O. Nieto, J. Moore, D. Ellis, D. Repetto, P. A. Holovaty, Librosa: Viktorin, J. F. Santos, *V0.4.0* (Zenodo, 2015; https://zenodo.org/record/18369).
- 27. G. Lindzey, H. Winston, M. Manosevitz, Social dominance in inbred mouse strains. *Nature*. **191**, 474–476 (1961).
- 28. F. M. Clarke, C. G. Faulkes, Dominance and queen succession in captive colonies of the eusocial naked mole-rat, Heterocephalus glaber. Proc. R. Soc. Lond. B. 264, 993-1000 (1997).
- 29. D. L. Bowling, M. Garcia, J. C. Dunn, R. Ruprecht, A. Stewart, K.-H. Frommolt, W. T. Fitch, Body size and vocalization in primates and carnivores. Sci Rep. 7, 41070 (2017).
- 30. F. M. Clarke, C. G. Faulkes, Hormonal and behavioural correlates of male dominance and reproductive status in captive colonies of the naked mole-rat, Heterocephalus glaber. Proc. R. Soc. Lond. B. 265, 1391–1399 (1998).
- Acknowledgments: We thank J. Reznick, U. Koch, A. Rossi, and O. Eigenbrod for helpful 30 comments on the manuscript, G. Pflanz, N. Schrim, and A. Mühlenberg for naked mole-rat husbandry, R. Hodge for illustrations and C.O. Lewin for Supplementary Movies. Funding: This work was supported by grants from the European Research Council (Advanced Grant 294678 to G.R.L) and a South African Research Chair for Mammalian Behavioral Research to N.C.B. We thank O. Daumke for additional salary funding. Author contributions: This project was 35 conceived by A.J.B. and G.R.L. who together wrote the manuscript with input from all authors. Behavioral experiments were performed by A.J.B., N.C.B., D.W.H., and L.M. Machine learning and data analysis pipelines were developed by G.V with input from A.J.B. Competing interests:

10

5

20

The authors declare no competing interests. **Data and materials availability:** All data is available in the manuscript or the Supplementary Materials.

Supplementary Materials:

Materials and Methods

Supplementary Text Figures S1-S12 References (24-30)

Movies S1-S2

Audio Files S1-S2

10

5



Fig. 1. Naked mole-rat soft chirps encode individual identity (A) Soft chirp analysis and classifier training workflow. (B) Individuals can be identified with high accuracy using machine learning tools trained on vocal features (A).



Fig. 2. Naked mole-rat soft chirps signal colony identity. (A) Schematic of classifier training. (B) Across four colonies (Colonies B, M and T; Berlin, Germany; Colony D; Pretoria, South Africa) colony identity is predicted with high accuracy. (C) Contributions of each vocalization feature to the colony classifier. (D) Asymmetry and peak frequency (inset in C) of soft chirps for all colonies. Error bars SEM.



Fig.3. Vocal response rates are modulated by colony identity (A) Left, in a Place Preference Assay naked mole-rats spend more time in the chamber with sound presentation compared to silence (n =4 animals; N \geq 36 trials per animal, one-way ANOVA, P < 0.005). Right, no place preference to colony-specific audio playbacks was observed. (B) Soft chirp response rates were enhanced to home colony audio playbacks. Example responses from Colony T animals. (C) Response rate is greater to home colony audio playback versus no playback or foreign colony playback, (n = 9 animals; N \geq 36, one-way ANOVA, P < 0.0005). (D) Example responses to home colony and artificial stimuli, Colony B. (E) For artificially generated stimuli, soft chirp responses to home colony classified audio playbacks are significantly increased compared to foreign colony classified audio playbacks or when frequency and asymmetry features alone are tested (n = 4 animals, Colony B, n= 5 animals, Colony T, one-way ANOVA or unpaired t test * P < 0.05, ** P < 0.005, *** P < 0.0005) . (F) Colony-specific response rates were present when conflicting olfactory cues (bedding from a foreign colony) was placed in the test chamber. (n= 6 animals, N \geq 36 trials per animal, P < 0.05). Error bars, SEM. For all experiments a minimum of N= 36 behavioral trials performed per animal.



Fig.4. Cultural transmission of colony dialects. (A) Schematic of cross fostering. (B-D) Individual colony dialect predictions for each pup (non-fostered control pups, Ny and Ob and fostered pup Mi). Prediction accuracies: pup Mi = 95.5%, pup Ob = 99.2%, pup Ny = 99.0%). (E) Schematic of second cross-fostering. (F, G) Individual colony dialect predictions for foster pups Da and Jo (Da = 59.1%, pup Jo = 68.4% prediction rate for foster colony dialect). (H) All fostered pups adopt the dialect of their adoptive colonies. (I) Timeline of social upheaval in Colony S. J. During Anarchy periods variability in soft chirp frequency increases. (K). Colony classification accuracy decreases during anarchy periods compared to epochs with a stable queen (black circles).

Supplementary Materials for

Cultural transmission of vocal dialect in the naked mole-rat

Alison J. Barker^{1*}, Grigorii Veviurko¹, Nigel C. Bennett², Daniel W. Hart², Lina Mograby¹, Gary R. Lewin^{1*}.

*Correspondence to: grlewin@mdc-berlin.de; Alison.Barker@mdc-berlin.de

This PDF file includes:

Materials and Methods Supplementary Text Figs. S1 to S12 Captions for Movies S1 to S2 Captions for Audio S1 to S2

Other Supplementary Materials for this manuscript include the following:

Movies S1 to S2 Audio S1 to S2

Materials and Methods

Animal care and maintenance

Naked mole-rats (Heterocephalus glaber) were maintained in a humidity (50-70%) and temperature (30-32 °C) controlled environment, under low illumination levels. A diet of tubers (primarily sweet potatoes, celery root and carrots) was provided daily (ad libitum) and supplemented weekly with ProNutro (Bokomo). Animals were housed by colony in a series of custom designed interconnected plastic chambers (Fräntzel Kunstsstoffe, Rangsdorf, Germany). All experimental procedures and husbandry protocols were approved by the local governmental authorities in Berlin (Landesamt für Gesundheit und Soziales, Berlin, License: G0196/17). Colonies in South Africa were housed in tunnel systems with several plastic chambers serving as food storage, toilet and sleeping areas and connected by acrylic glass tunnels. They were fed a variety of chopped vegetables (primarily sweet potatoes, cucumbers and carrots) and supplemented weekly with ProNutro (Bokomo). Nesting material consisted of wood shavings. The ambient room temperature ranged from between 29 and 32 °C and humidity from 50-70%. The experimental protocol was approved by the Animal Ethics Committee of the University of Pretoria (ECO73-17). All the naked mole-rats used in this study are descended from multiple colonies captured by Prof. Jenny Jarvis primarily in Mtito Andei, and Lerata Kenya and constitute a mixed parentage (1). Detailed information related to animals from each colony can be found in fig.S3.

Audio recordings

Audio recordings were made from a total of 166 animals from seven colonies; six housed at the Max Delbrück Center for Molecular Medicine in Berlin, Germany and one at the University of Pretoria, South Africa. Audio recordings were acquired using Sennheiser MKH8020 microphones and a Behringer U-Phoria UMC1820 audio interface with Avisoft-Recorder Software (Avisoft Bioacoustics, Glienicke/Nordbahn) (sampling rate 32 kHz, 24 bit resolution). Of recorded soft chirps, 2,526 soft chirps did not pass our quality control checks and were removed as outliers, the remaining 36,190 soft chirps were used for analysis. Raw sound recordings were visualized as spectrograms for further analysis using Avisoft-SASLab Pro software (Avisoft Bioacoustics, Glienicke/Nordbahn) or with custom python scripts.

Code availability

All custom-written scripts are archived here: DOI 10.5281/zenodo.4104396.

Audio data processing and feature extraction

Raw audio recordings were digitized as wav files which were then segmented into individual sounds using a custom developed "SoundSplitter" program. Following the segmentation of putative individual sounds, wav file segments relating to single soft chirps were converted into spectrograms as described in detail below and manually confirmed by the experimenter. Soft chirp spectrograms were then traced using a custom script to identify and generate a new spectrogram file containing only the fundamental (lowest) harmonic frequency. Feature extraction was

performed on both raw sound files and spectrograms and combined parameters used for training machine learning classifiers. A schematic of feature extraction can be found in fig.S1.

SoundSplitter

Audio recordings (wav files) were processed with a custom python "SoundSplitter" script to segment individual vocalizations from a longer recording often containing multiple vocalizations, and some background noise. In brief, the algorithm parses the entire audio recording into bins of 23 ms, extracts forty Mel-frequency cepstral coefficients to use as a features vector, and designates each bin as sound or noise using a Random Forest Classifier adapted from the Python3 sklearn module (with default parameters and 256 trees) (24). All splits were manually validated by experimenters with a built-in validation step.

Spectrogram generation

Audio recordings (wav files) were converted into spectrograms using the short time Fourier transform (STFT) function from the Python librosa library using default parameters and bin size set to 512.

SoftTrace

We developed a "SoftTrace" algorithm for extracting the features of individual soft chirps from a spectrogram image, using a convolutional auto encoder (CAE) trained on manually traced soft chirps. CAE is a neural network composed of three convolution layers with max-pooling followed by symmetrical deconvolution and up-sampling layers (25). All soft chirp spectrograms were traced using our CAE script to generate a new image file of the first harmonic of the soft chirp, which was used for feature extraction. Soft traces were passed through an automated filter to remove outliers (i.e. other non-soft chirp sound types that occasionally occurred during the recording) and all soft chirp traces were manually validated for accuracy before proceeding to feature extraction. (see fig. S1).

Feature extraction

For comparative analysis of soft chirp sound parameters we extracted eight features: (1) *pitch*, (2) *wiener entropy*, and (3) *zero crossing rate* using the librosa library (4) and (4) *asymmetry*, (5) *duration*, (6) *height*, (7) *peak frequency* and (8) *slope* using a custom script designed to work with soft chirp traces made from spectrogram images. Features (1)–(3) were extracted from the original soundwave files (down sampled from 32,000 Hz to 22,050 Hz) using the librosa library for audio data processing (26). (1) *Pitch* was computed using the piptrack function from the librosa library. The frequency range was set to [2kHz, 8kHz] and window size 10ms. Other parameters were set at default values. Determination of pitch differed from the extraction of the fundamental frequency in that the pitch determination weighted the contribution of all harmonics. (2) *Wiener entropy* measures how close the soundwave signal is to white noise. It was computed using implementation from the librosa library with bin size set to 10ms. (3) *Zero crossing rate* is a commonly used feature for audio data analysis counting how many times the soundwave changes in sign. The built-in

librosa zero_crossings function was used to compute this feature. Parameters from the soft chirp trace were computed directly from the spectrogram image and determined as follows: (4) *Asymmetry* describes the vertical (frequency) axis difference between the left- most and rightmost points of the soft chirp harmonic, (*Start frequency – End frequency*). (5) *Duration* is defined as the length of the soft chirp on the horizontal (time) axis and calculated by subtracting the start point from the end point of the soft chirp in (s) (*End time – Start time*). (6) *Height* describes the difference in the vertical (frequency – (*End frequency – (End frequency – (End frequency / Start frequency)*). (7) *Peak frequency* and (8) *slope* were computed from a parabolic fit to the soft chirp spectrogram image. Using the Python3 numpy.linalg module, a parabola $y = -(a(x - h)^2 + k)$ was fit to each soft chirp such that the mean square error was minimized. The frequency of the vertex (k) was taken as the peak frequency and slope as the coefficient, *a* preceding x^2 .

Machine learning classifiers

Using the Random Forest Classifier module from the librosa library classifiers were trained on the eight extracted soft chirp features (described above, fig. S1). The classifier used 256 trees with additional parameters set to default settings (24,26). Confusion matrices were obtained by averaging cross validation results and display the prediction strength of the classifier as a fraction (i.e. the fraction of times the predicted data label matches the actual data label, where a value of 1 denotes a 100% correct prediction rate). Data splits for cross-validation testing colony dialects were performed on individual animals, i.e. all sounds from each animal were either in the training or testing dataset. This step was necessary in order for cross validation results to test the ability of the classifier to learn intercolonial differences rather than memorize individuals. For prediction of other features (i.e. age, rank, sex, body mass and body length) cross-validation was also split by individual animal id. For cross-validation testing of individual soft chirp recognition, soft chirps were split by recording date. To classify colony dialects of foster pups, the trained colony classifier was used to test soft chirps recorded from foster pups and output a classification prediction rate (as a percentage) for each colony. For each foster pup the colony classifier was trained to distinguish between 5 colony dialects (including foster and birth colonies). In addition to confusion matrices, we provide a list of feature importance which show the impact each feature has on the soft chirp classification (Fig. 2C). Importance calculation was implemented in the scikit-learn library. Prediction accuracy for Colony M was on average 76.18% + 6.7, SEM).

Statistics

All statistics were performed using GraphPad Prism 8 or custom python scripts. Data were tested for normality (D'Agostino-Pearson test) and for homoscedasticity (Bartlett's test.) For multiple comparisons normally distributed datasets were tested with a one-way ANOVA and post-hoc tests performed with Tukey's multiple comparisons test. For nonparametric data Kruskal-Wallis test with Dunn's multiple comparisons test were used. For single comparisons of normally distributed data unpaired t tests were used. Confusion matrices were generated using standard python libraries. For data shown in fig.S7 (A-D) linear regressions were plotted in Prism. Residuals for all features are also plotted in fig. S7 (E-H). Significance values are reported as: * P value ≤ 0.05 ; *** P value ≤ 0.005 ; *** P value ≤ 0.0005 . All error bars are standard error of the mean (SEM).

Generation of playback and artificial stimuli

<u>Playback stimuli</u>: Playback soft chirps were selected to be representative of each colony. A 10 s excerpt of soft chirps were compiled from a rank and weight matched individual from each colony and the temporal sequence of soft chirps was maintained from the original recording. A single trial consisted as presentation of the entire playback stimuli (between 10-18 soft chirps). Soft chirp sequences were followed by 10s of silence before repetition. <u>Artificial stimuli</u>: Artificial stimuli were generated using the graphical synthesizer function in SASLab Pro software (Avisoft Bioacoustics, Glienicke/Nordbahn). Duration was kept constant at 120 ms for all stimuli. Artificial stimuli were designed for both Colony B and Colony T using empirically determined mean frequencies and mean asymmetries for each colony. We also generated a pure tone of 4.5 kHz (Colony B, mean peak frequency) and peak frequency-doubled stimuli (Colony B peak frequency, 9.0 kHz). For frequency – doubled artificial stimuli, the mean asymmetry values for the colony were maintained. For all artificial stimuli, a sequence of 8 soft chirps constituted a single trial. Individual soft chirps were presented at intervals ranging from 0.5- 3s. Soft chirp sequences were followed by 10s of silence before repetition.

Behavioral experiments

Place Preference Assay

Four subordinates from Colony B and five from Colony T were used for all behavioral experiments. Animals were habituated to the behavioral chamber (as shown in Fig. 3A top) consisting of three interconnected chambers for ~ 10 minutes prior to the start of the experiments. Each chamber (left, right) contained a microphone (Sennheiser MKH8020) and loudspeaker (UKHONK Mini USB Speaker, HK-5002) connected to a laptop for simultaneous audio recordings and audio playback presentation. All individuals tested were in the worker class (rank 3-5). For each animal a minimum of 36 trials for each audio stimulus was performed with each trial containing between 8-18 soft chirp presentations. During the behavioral trial the animal was allowed to move freely throughout all chambers, and sound presentation begun only when the animal was in the central chamber. The chamber with sound presentation (left or right speakers) was pseudo-randomly alternated. Each playback stimulus was presented both in the right and left chambers with equal frequency and trials from all chambers averaged to control for any inherent chamber preferences. Time spent in each chamber was logged manually by a partially-blinded experimenter (via a handheld timer) as the animal transitioned into the chamber during times of audio stimulus presentation. Soft chirp responses made by experimental animals were analyzed offline from audio recordings made during the behavioral trial. All experiments were performed on two separate experimental days to ensure repeatability. A Place Preference Index was calculated as the (amount of time spent in sound chamber – the amount of time spent in non-sound chamber) divided by (the total amount of time spent in either sound presentation chamber, left or right). Thus, maximal avoidance of sound stimuli would be scored with a value of -1 and maximal preference for the sound stimulus would be scored with a value of 1. A soft chirp occurring within 0.5 s of a presented playback stimulus was counted as a response. Response rate was calculated per animal per day (averaging all trials for each stimulus) and normalized to the maximal response

per animal. Response rates were then averaged across animals and across days for final statistical analysis. As part of the habituation process, bedding from the home colony was evenly distributed throughout the behavioral apparatus. To test that bedding was not providing an olfactory cue that may influence soft chirp responses, we performed control behavioral experiments with the six animals (four from Colony B and two from Colony T) with bedding from a foreign colony (Fig.3F). All playback stimuli were presented at the same volume for all trials and measured at 81.9 dB (+/-1.5dB, SEM) in the sound presentation chamber.

Forced Choice Assay

A modified version of the Place Preference Assay was also performed in which audio playback from two colonies was presented simultaneously. We quantified the place preference in each chamber using the place preference index as described above (see fig. S8).

Cross fostering experiments

Two separate cross fostering experiments were performed involving three animals. For all crossfostering experiments foster pups were washed with warm water and coated in a slurry of fecal matter removed from the foster colony before being placed in the new colony, and identifying marks made on their forepaw digits which could be continually tracked until they were old to chip with RFID chips at ~ 6 months of age. In the first experiment two orphaned pups (pups Jo and pup Da) born into Colony S, were cross fostered into two separate colonies that both had litters within one week of the birth of pups J and D (Colonies T and M respectively). Audio recordings were made at periodic intervals and pups were frequently monitored to ensure survival in the colonies. In a second cross foster experiment a pup (pup Mi) that was abandoned from Colony T was fostered into Colony M within the first postnatal week of life. Several foster siblings from Colony M also survived, namely pups Ob and Ny, and were tracked along with pup Mi with periodic audio recordings. After six months of fostering, colony dialects were tested.

Queen transitions

In Colony S, the first queen, who had reigned stably, producing several litters died due a pregnancy related complication, which was confirmed upon autopsy. The ascending queen was considered established once she gave birth to her first litter. She was attacked following the birth of the first litter by several males within her colony and had to be euthanized due to her severe injuries (~1 month after the birth of the pups). Two surviving pups from this queen that were cross-fostered (pups Jo and Da in Fig. 4). In Colony B, the queen was also attacked and overthrown by colony members (fig.S12). Additional soft chirp recordings were made in the three months following her overthrow and constitute the anarchy phase described in fig.S12.

Hierarchy assessment

To reliably assess the rank of individuals within colonies, we modified an assay for dominance in rodents (28, 29). When two naked mole-rats approach each other head on in a tunnel, the more dominant individual is often observed to climb over the more subordinate individual (29). Using this principle, we established a behavioral ranking test, which was reliably able to predict the

queen, who is morphologically distinct and therefore easily visually confirmed. In brief, two plastic chambers were connected via a transparent plastic tube. Two naked mole-rats were placed in each chamber and allowed to move freely between the two chambers, when the animals entered the tube simultaneously the identity of the mole-rat that climbed over the other was recorded. Hierarchy was assessed in a winner take all, single elimination strategy (with a minimum of three trials for each pairing). Animals were pseudo-randomly selected for pairing and hierarchies were tested over multiple months to ensure animals were tested with equal frequency. In some cases, the colony was split into lower and higher-ranking individuals based on previous hierarchy tests to perform a more fine-grained ranking of individuals. A Ranking index (R.I.) was defined as follows: (number of wins) divided by (the total number of behavioral trials). Ranking indices were normalized to maximum values for each colony to allow for comparisons across colonies. Thus, the winner, the individual with the highest fraction of wins would have ranking index of 1. Ranks were assigned as rank 1, R.I. ≥ 0.8 , rank 2, R.I. ≥ 0.6 , rank 3, R.I. ≥ 0.4 , rank 4, R.I. ≥ 0.2 , rank 5, R.I. < 0.2 (as shown in fig. S6). The queen was always assigned a rank of 1 when testing rank in our dialect classifiers.

Supplementary Text

Body size and its influence on soft chirp sound features

Previous reports have suggested that the frequency of the soft chirp may be highly dependent on body mass (*16*, *20*) and inverse correlations between vocal pitch and body length have been reported in numerous mammalian species (*29*). Surprisingly, we observed a significant positive correlation between soft chirp vocal pitch and both body length and body mass (fig.S7A,B; R^2 = 0.189, P = 0.0006, body length and R^2 = 0.295, P = 0.0007, body mass). In some cases, the queen, who is consistently the largest individual (in both body mass and length), displayed the highest frequency soft chirps (fig. S7K,L).

Queen influence on colony dialect

We were able to examine a second colony, Colony B which experienced an overthrow of the queen. In this case we could confirm that the queen was killed by other colony members and we observed a similar loss of soft chirp colony dialect cohesion in the three months following her death (fig.S12A,B) as also observed in Colony S (Fig.4I-K). This new result provides further evidence that the queen's presence is necessary for the maintenance of colony dialects. Comparisons of individual soft chirp variability between queens, breeding males and representative subordinates from five colonies revealed the highest variability in individual soft chirps (comparing peak frequency and asymmetry, fig.S5) from the breeding males, which except for the queen are the only reproductively active members of the colony (1, 28, 30). Increased vocal variability in the breeding males suggests a link between vocal plasticity and hormonal state. We hypothesize that a similar mechanism may be at work when the queen is lost, as colony-wide reproductive suppression is also lost (28, 30) and thus accompanying physiological changes may contribute to overall variability in individual soft chirps.

Developmental acquisition of soft chirps

Young animals (from 1-3 months of age) produce at least three distinct sounds (pup combo, pup squawk, pup cheveron), not observed in adults. Additionally, several juvenile versions of adult sound types namely the upsweep, downsweep, phee and soft chirp begin to appear around \sim 1 month of life but do not recapitulate adult sound features until \sim 6 months of age (6, 15; fig.S9-S10). Developmental transitions in sound type usage may have a purely anatomical basis, due to an immature vocal tract or may represent a type of vocal learning: either as an active sensory motor learning process where practice of vocalizations is required for refinement (i.e. vocal production learning), or from context learning (i.e. usage learning, where pup sounds may still be vocally possible for adults but animals learn to use only adult sounds as they mature) (23). While the underlying mechanisms are not yet known, it is clear that during early life the vocal repertoire of the naked mole-rat differs from the adult, suggesting the possibility of a developmental window for acquisition of colony dialects.





Workflow for soft chirp feature extraction. (A) Features from both the soundwave and spectrogram are used. (B) Example of automated soft chirp spectrogram feature detection from custom -written scripts.



Fig. S2.

Soft chirps contain information that is sufficient to distinguish individuals within a colony. Confusion matrices show prediction success rates for Random Forest classifier trained on individual identity with eight soft chirp features. Results from two additional colonies are shown here. (A) Colony A. (B) Colony T. S denotes subordinates from each colony.



Fig. S3.

Colony biometrics. (A-C) Distribution of body mass (A), body length (B) and age (C) across colonies. (D) Number of animals recorded from each colony. (E) Sex distribution by colony (one-way ANOVA, Tukey's multiple comparisons test in (A, B), Kruskal-Wallis test with Dunn's multiple comparisons test in (C); * P <0.05, ** P < 0.005, *** P,< 0.0005 . Error bars, SEM.



Fig. S4.

(A-I) Analysis of individual soft chirp features for each colony display significant differences across multiple spectrogram and soundwave extracted features , * P <0.05, ** P < 0.005, *** P,< 0.0005 . Error bars, SEM. The eight features displayed here were used to train Random Forest classifiers in Fig. 2.



Fig. S5.

Breeding males produce the most variable soft chirps within colonies. (A-C) Soft chirp peak frequency vs. soft chirp asymmetry graphs provide a snapshot of variability across individuals for queens (A), breeding males (B) and subordinates (C). Colony indicated by color bar. Ellipses display 95% confidence intervals.



Fig. S6.

Rank, age and sex are not strongly predicted by soft chirp features. (A-B) Example of hierarchy test for assessing individual rank in a colony, data shown from Colony B and previously described in (28). More dominant individuals will climb over less dominant individuals (A, and see Movie S2) an assay which successfully predicts the queen (B). (C-E) Rank (C), age (D), and sex (E) are not strongly encoded in soft chirp features. Confusion matrices for Random Forest classifier training with rank, age and sex respectively.



Fig. S7.

Naked mole-rat soft chirp pitch but not peak frequency shows a positive correlation with body size. (A-B) Pitch vs. body length (A) and body mass (B). (C-D) Peak soft chirp frequency vs. body length (A) and body mass (B). (E-H) Residuals for pitch, body length, body mass and peak frequency. (I,J) Confusion matrices for Random Forest classifier trained on body length (I) and body mass (J). (K, L) Despite having one of the longest body lengths and largest body masses relative to the rest of the colony, the queen does not produce the lowest frequency vocalizations when compared across the entire colony, as predicted by an inverse correlation between pitch or frequency and body size, data plotted for Colony M.



Fig. S8.

Recognition of colony dialects drives preferential behavioral responses. (A) Schematic of forced choice behavioral assay (top). When soft chirp audio playbacks from two colonies are presented simultaneously animals prefer to spend time in the chamber presenting soft chirps from their own colony * P < 0.05, *** P < 0.005, *** P, < 0.0005. Error bars, SEM (B) Representative examples of responses to additional audio stimuli: pure tone (4.5 kHz) and peak frequency-doubled stimuli from Colony B. (C) Artificial playback stimuli generated for Colony B overlap with Colony B features but not with any individual Colony B member (peak frequency is plotted vs. asymmetry). (D-F) Artificial stimuli generated for Colony T are correctly assigned to Colony B and T respectively when tested with our colony classifier. Error bars are SEM, ellipses display 95% confidence intervals.





(A-F) Representative examples of developmental sound type usage in a single pup, Ce, from Colony L. Note transition from exclusively pup sounds to soft chirps begins around P60-P90.



Fig.S10.

(A-E) Distribution of sound type usage across early development for four pups in Colony L (pups Ce, Ja, Sa, and Ty. Note transition to soft chirp primary soft chirp usage over time. ‡, denotes exclusively pup produced sound types.



Fig. S11.

(A-F) Hierarchy changes in Colony S can be reliably tracked over time. Note ascendance of new queen (ID 8318) which corresponds to Queen Epoch 2 in Fig. 4 I-K.



Fig. S12.

Individual variability is increased during periods of anarchy, when no queen is present. (A) Colony B prediction accuracy is high when the colony has a stable Queen and is reduced when the Queen is lost (B). Black circles indicate changes in prediction accuracy. (C-F) Representative examples of soft chirp variability (peak frequency versus asymmetry) are shown in four subordinates (S1-S4) from Colony S during the epoch of Queen I (C), the subsequent phase of Anarchy (I, D), the epoch of Queen II (E) and a second phase of Anarchy (II, F). Ellipses display 95% confidence intervals.

Movie S1. Example of audio playback experiments. Spectrograms with audio stimulus and vocal responses are superimposed on video. Naked mole-rats show preferential responses to home colony playbacks (here Colony B) and artificially generated home colony playbacks but not to playbacks from a foreign colony, here Colony T. Video shown in real time.

Movie S2. Example of dominance interaction assessed with hierarchy test for rank as previously described in (28). When two naked mole-rats encounter one another in a tunnel, the dominant individual will climb over the subordinate individual.

Audio S1. Excerpt of a vocal interaction between two naked mole-rats (queen and breeding male) provides a snapshot of the vocal complexity produced by these animals.

Audio S2. Audio excerpt from a whole colony recording demonstrates the near constant volley of soft chirps heard during normal colony interactions.