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DATA DESCRIPTOR

A dataset of acoustic measurements from soundscapes collected worldwide during the COVID-19 pandemic

Samuel Challéat^{1,181}✉, Nicolas Farrugia^{2,181}✉, Jérémy S. P. Froidevaux^{3,4,5,181}, Amandine Gasc^{6,7,181}, Nicolas Pajusco^{2,8,181} & Silent Cities project consortium*

Political responses to the COVID-19 pandemic led to changes in city soundscapes around the globe. From March to October 2020, a consortium of 261 contributors from 35 countries brought together by the Silent Cities project built a unique soundscape recordings collection to report on local acoustic changes in urban areas. We present this collection here, along with metadata including observational descriptions of the local areas from the contributors, open-source environmental data, open-source confinement levels and calculation of acoustic descriptors. We performed a technical validation of the dataset using statistical models run on a subset of manually annotated soundscapes. Results confirmed the large-scale usability of ecoacoustic indices and automatic sound event recognition in the Silent Cities soundscape collection. We expect this dataset to be useful for research in the multidisciplinary field of environmental sciences.

Background & Summary

In response to the rapid spread of the coronavirus disease 2019 (COVID-19) around the world, governments of many countries adopted physical distancing measures in early 2020, including more or less drastically restricting individual travel and suspending many work and leisure activities deemed ‘non-essential’^{1–3}. Incidentally, these public health policy decisions opened a window of opportunity for many environmental scientists to investigate the effects of such a reduction in human activity on ecosystems at multiple spatiotemporal scales^{4–8}.

The modification of soundscapes, especially in urban and peri-urban areas, was among the most significant environmental changes observed during this period^{9–14}. The sudden decrease in individual travel and motorized transport of people and goods shaped extraordinary soundscapes in most cities of the world for a few weeks. This revealed the richness of animal sounds in urban areas, previously hidden by a multitude of anthropogenic sounds. Such a change, directly perceptible by the population, even raised interest outside the academic sphere, as reflected in numerous articles in the general press. Among the thousands of press articles on the subject, we will particularly mention the interactive publications produced by *The New York Times* (see, for example: [The Coronavirus Quieted City Noise. Listen to What’s Left](#); or: [The New York City of Our Imagination](#)).

¹UMR5602 GEODE (Géographie de l’environnement), CNRS, Université Toulouse 2 - Jean Jaurès, 31058, Toulouse, France. ²UMR6285 Lab-STICC (Laboratoire des Sciences et Techniques de l’information de la Communication et de la Connaissance), IMT Atlantique, CNRS, 29238, Brest, France. ³University of Stirling, Biological and Environmental Sciences, Faculty of Natural Sciences, FK9 4LA, Stirling, United Kingdom. ⁴UMR7204 CESCO (Centre d’Écologie et des Sciences de la Conservation), Muséum national d’Histoire naturelle, CNRS, Sorbonne Université, 75005, Paris, France. ⁵University of Bristol, School of Biological Sciences, Life Sciences Building, Bristol, BS8 1TQ, United Kingdom. ⁶UMR7263 IMBE (Institut Méditerranéen de Biodiversité et d’Écologie marine et continentale), Aix-Marseille Université, Avignon Université, CNRS, IRD, 13080, Aix-en-Provence, France. ⁷Department of Forest Ecosystems and Society, Oregon State University, Corvallis, OR, USA. ⁸UMR6613 LAUM (Laboratoire d’Acoustique de l’Université du Mans), Institut d’Acoustique - Graduate School (IA-GS), CNRS, Le Mans Université, 72085, Le Mans, France. ¹⁸¹These authors contributed equally: Samuel Challéat, Nicolas Farrugia, Jérémy S. P. Froidevaux, Amandine Gasc, Nicolas Pajusco. *A list of authors and their affiliations appears at the end of the paper. ✉e-mail: samuel.challeat@cnrs.fr; nicolas.farrugia@imt-atlantique.fr

From an academic point of view, several studies have already been carried out on these “soundscapes of a confined world”, at different scales and in different types of spaces and territories (sub-continent¹⁵, countries^{16–21}, regions^{22–25}, cities^{18,26–33}, neighborhoods^{29,34}, protected natural areas³⁵, semi-anthropized environments^{36–38}, tourist sites³⁹). Among these studies, some benefited from sensor networks predating the COVID-19 crisis, mobilizing for example underwater acoustics and/or seismic monitoring networks^{16,22,40}, permanent noise pollution monitoring networks in urban environments^{26,41,42}, or devices installed for pre-existing research projects⁴³. Beyond these physical measurement approaches, several studies have also investigated individual and subjective perceptions of changes in soundscape composition. More specifically, the perceived proportion between natural and anthropogenic events in the soundscape is regularly raised in investigations that more broadly address the changes induced by different periods of population containment on experiential relationships to nature^{21,29,44}.

In this paper, we present a global acoustic dataset⁴⁵, collected between March and October 2020 by 261 contributors, at 317 sites distributed in 35 countries (Fig. 1). Recordings were primarily collected using Open Acoustic Devices AudioMoth⁴⁶ or Wildlife Acoustics Song Meters SM4 (www.wildlifeacoustics.com) programmable recorders, which are widely used within the professional and amateur naturalist communities. This dataset is unique, with its international dimension, collaborative construction and open access availability. The acoustic data are presented in addition to climate classification and surrounding environment, offering a more comprehensive understanding of their significance and implications. In addition, we provide a set of descriptors based on ecoacoustic indices⁴⁷ and on automatic recognition of sound categories using a pretrained deep neural network⁴⁸. These descriptors were subsequently validated by collecting expert annotations on a small subset of the dataset, with which we derived statistical models to demonstrate their usability.

Methods

Silent Cities is a data collection that involved programmable audio recordings worldwide. The global scale of the project warranted us to not only gather acoustical recordings, but also contextualize them. We first describe the data collection procedures, the contributors’ network and contextual information related to the recording sites, such as location, urban density, climate classification or governmental policies related to human population containment in response to COVID-19. Next, we describe the processing of acoustic measurements computed on all recordings, including ecoacoustic indices, automatic sound event recognition, and voice activity detection.

Data collection. *Recording protocol.* On March 16, 2020, the French government announced the upcoming first containment of the population. A few days later, a first version of the Silent Cities protocol was submitted to professional networks. Feedback from researchers but also from journalists, artists and biological conservation practitioners interested in contributing were received. As requested, a more inclusive version, opening up the possibility of using different equipment and sampling efforts while preserving requirements for further robust statistical analyses was proposed (<https://osf.io/m4vnnw/>). This second and final version of the protocol was shared on March 25, 2020 and is described below.

Each contributor provided recording equipment. To homogenize the recordings collection, recording devices were configured to obtain a 1 minute-long recording every 10 minutes on a daily cycle schedule, with a sampling rate set at 48 kHz. All recorders were to be set in Coordinated Universal Time (UTC+00) with an output format in .wav. In order to have comparable data, the use of an audible SM4 (Wildlife Acoustics) or an AudioMoth (Open Acoustic Devices), which were the two most popular programmable recorders at the time, was recommended. However, any device with high quality recording, allowing the recording configuration requested, was accepted. To anticipate the return of high levels of anthropogenic sounds after the end of containment measures, the gain was to be set at “low” for the Audiomoth and at 31 dB for the SM4 (gain at 5 dB and preamplification at 26 dB). The final dataset includes 216 sites monitored by an Audiomoth, 47 by an SM4 and 54 by another device.

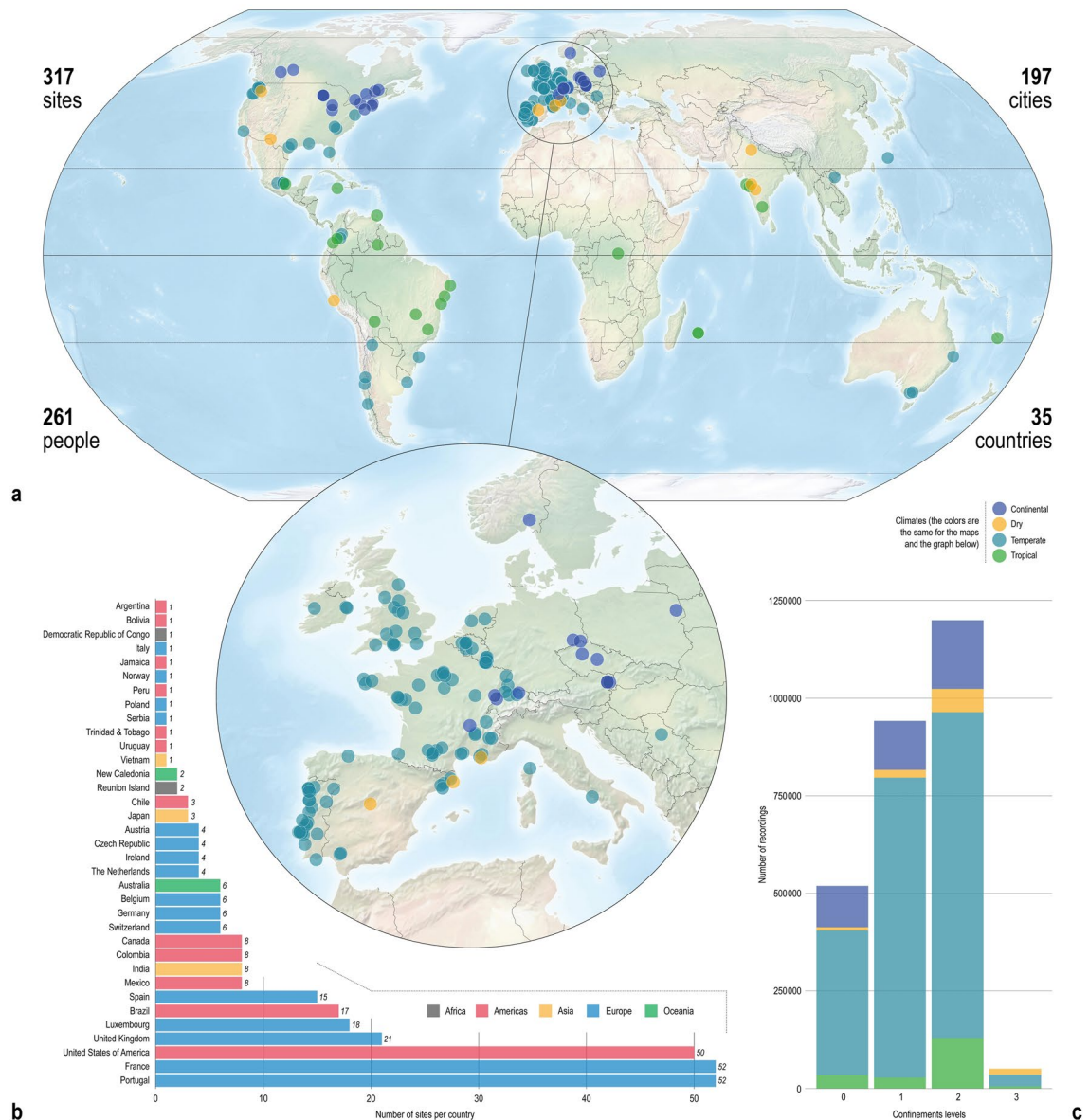
The sampling duration of the collection was locally dependent. The protocol recommended to continue recording a minimum of two weeks after the end of the total city shut down and restoration of “normal” activities. However, the expected scenario of the return to “normal” activity extended well beyond predictions as the magnitude of the pandemic became progressively realised. As containment measures were being lifted in many countries during the summer, the acoustic sampling was ended on July 31, allowing contributors to continue collecting data after this date based on local situations. To summarise, the entire recordings collection covers the period from March 16 to October 31, 2020, with the highest number of recordings between April and July (see Fig. 1d).

Originally, contributors were able to choose between three levels of sampling effort based on their ability to record during the entire or partial duration of the project. Hereafter, we refine the definition of those levels to better fit the diversity of recording profiles represented in the final data set:

- expert - The daily cycle schedule, duration of files and sampling rate were set according to the recommendations, and the duration of the sampling period was at least two months;
- modified - The parameters are set as recommended but the sampling period is less than two months or some parameters such as the file duration, the sampling rate or the daily cycle schedule are different (*i.e.* every 3 hours), while conserving a fixed recording pattern along the sampling period;
- opportunistic - All other sites that do not show any type of recording patterns.

The expert protocol was applied by 228 contributors, while the modified protocol and the opportunistic protocol were followed respectively by 72 and 17 contributors.

Where?



When?

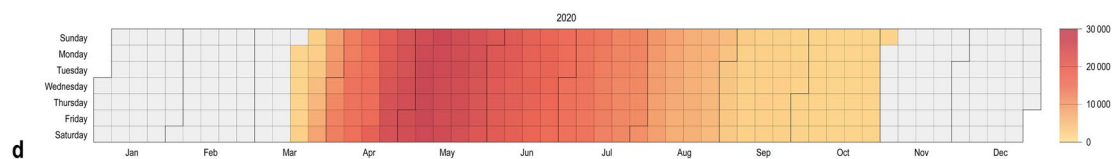


Fig. 1 Panel a. Global and European mapping of recording sites. Colors refer to climates. Panel b. Number of recording sites per country. Colors refer to continents. Panel c. Number of recordings by confinement level and climate. Panel d: temporal distribution of global sampling effort, in number of recordings.

International contributor network. The dataset⁴⁵ results from the collaborative work of 261 international contributors from various professional fields: 182 are academics, 37 are conservationist practitioners, 12 are artists and 30 do not recognize themselves in the three previous groups. An Open Science Foundation (OSF) project⁴⁵ was created to organize the data collection and guarantee its open access with no restrictions. Other tools used to manage the collaborative work were Framafoms (<https://framaforms.org/abc/fr/>) to collect metadata about sites and contributors from the consortium.

Site descriptions. The containment of a large number of citizens worldwide restricted the location of the recorders. Contributors deployed their recorder on private land or a balcony at their residency (example on Fig. 2a). We encouraged those living in (peri-)urban areas to participate, even though recordings were also collected in rural areas. The soundscape recordings were collected from 317 sites located in or around 197 cities and 35 countries (see Fig. 1). In order to protect citizens' privacy, the exact coordinates of the sites remain unknown and the location of the sites were based on the coordinates of their corresponding cities and approximate neighborhood. The sites cover four of the five climates defined by the Köppen climate classification⁴⁹, with a majority of sites located in the temperate and dry climates and a spatial sampling in favor of the European and American continents (see Fig. 1). For each site, we extracted information about the surrounding land cover (more specifically the percentage of built-up and tree cover within a 1 km radius buffer scale around the sites; 100 m resolution⁵⁰), human footprint (from 0 to 50, with the lowest score depicting the least human influence, 1 km resolution^{51,52}), and population density (no. of inhabitants per square kilometer, 1 km resolution⁵³) to document the degree of urbanization and human impact on the landscapes encompassing the recordings. In addition, contributors described in a few sentences the surroundings/context of their site. Thanks to the open data available on <https://aa.usno.navy.mil/data/AltAz>, we also extracted for each recording site the altazimuth coordinates of the Moon and Sun as well as the moon phase for each 10-second time interval during the days where soundscapes were collected. These data would be important for potential analysis about temporal soundscape dynamics. Finally, containment measures³ per country and date, summarized by the University of Oxford in the *Oxford COVID-19 Government Response Tracker* dataset, were downloaded from the web portal <https://ourworldindata.org/grapher/stay-at-home-covid>. These stay-at-home requirements are organized in 4 levels:

- 0 - No measures;
- 1 - Recommended not to leave home;
- 2 - Not allowed to leave home, with exceptions for daily exercise, grocery shopping, and other activities considered as essential;
- 3 - Not allowed to leave home, with rare exceptions (e.g. allowed to leave only once every few days, or only one person at a time).

Due to the limited data collected during the strictest containment period (level 3, see Fig. 1c), we combined data from the two periods when leaving home was not permitted (levels 2 and 3) when performing the technical validation.

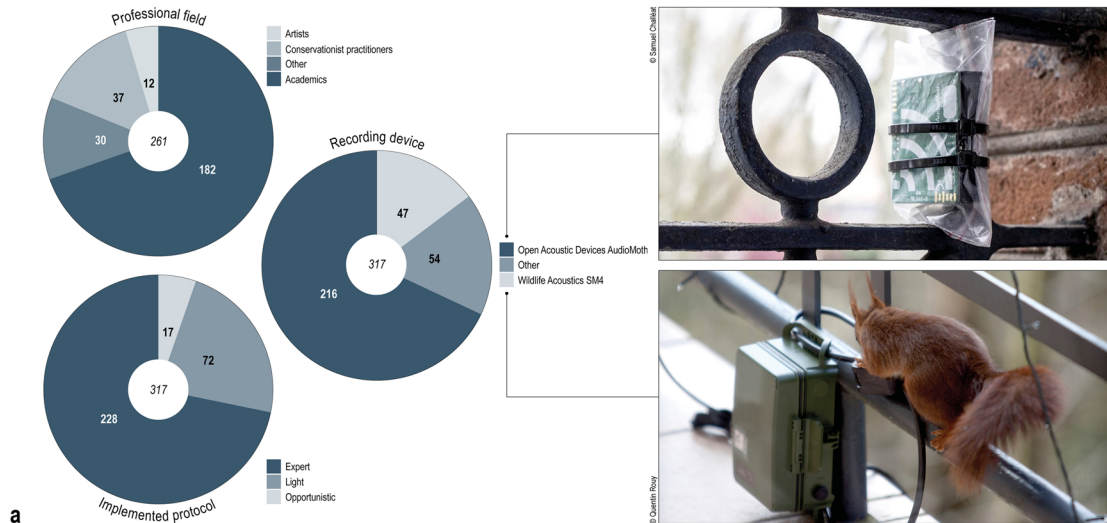
Acoustic measurements. All computations described here were performed with open-source packages or code from github, including scikit-maad (v1.4)^{54,55}, librosa⁵⁶ and pytorch⁵⁷. The analysis code used to prepare this dataset is available for reference at <https://github.com/brain-bzh/SilentCities>.

Preprocessing audio. Audio preprocessing was divided into two steps. First, the file name, sample rate, date and relative sound pressure level were extracted from each audio recording. Then, each file ($n = 2,701,378$) was divided into 10-second segments ($n = 16,252,373$) in order to have a meaningful duration for both acoustic index calculation and automatic sound event recognition. The sampling rate of audio segments were homogenised at 48 kHz for acoustic index calculation and resampled to 32 kHz for automatic sound event recognition. For acoustic indices, the signals were filtered using a bandpass filter from 100 Hz to 20 kHz to remove low frequency electronic noise inherent to some recorders.

Acoustic Indices calculation. Acoustic diversity indices aim to summarize the overall complexity of an acoustic recording in a single mathematical value. Numerous acoustic indices have been previously proposed^{47,58,59}, considering the time, frequency and/or amplitude dimensions of the recorded sound wave. We selected and calculated eight indices on all recordings; these indices were chosen based on their complementary and/or wide representation in the literature:

- dB represents the relative acoustic energy of a signal;
- dB Full Scale or dBfs represents the acoustic energy of a signal where the RMS value of a full-scale sine wave is defined as 0 dBfs⁶⁰;
- Acoustic Complexity Index or ACI⁶¹ measures the frequency modulation over the time course of the recordings. The value is calculated on a spectrogram (amplitude per frequency per time). ACI is described to be sensitive to highly modulated sounds, such as song birds, and less affected by constant sounds, such as background noise;
- Activity or ACT⁶² corresponds to the fraction of values in the noise-reduced decibel envelope that exceed the threshold of 12 dB above the noise level. This noise level was estimated for each site by seeking the audio file yielding the minimum dB value;
- Bioacoustic index or BI⁶³ measures the area under the frequency spectrum (amplitude per frequency) above a threshold defined as the minimum amplitude value of the spectrum. This threshold represents the limit between what can be considered acoustic activity (above threshold) and what could be considered background noise (under threshold);
- Entropy of the Average Spectrum or EAS⁶² is a measure of the 'concentration' of mean energy within the mid-band of the mean-energy spectrum;
- Entropy of the Spectrum of Coefficients of Variation or ECV⁶² is derived in a similar manner to EAS except that the spectrum is composed of coefficients of variation, defined as variance divided by the mean of the entropy values in each frequency bin;

Who and how?



Dataset technical validation

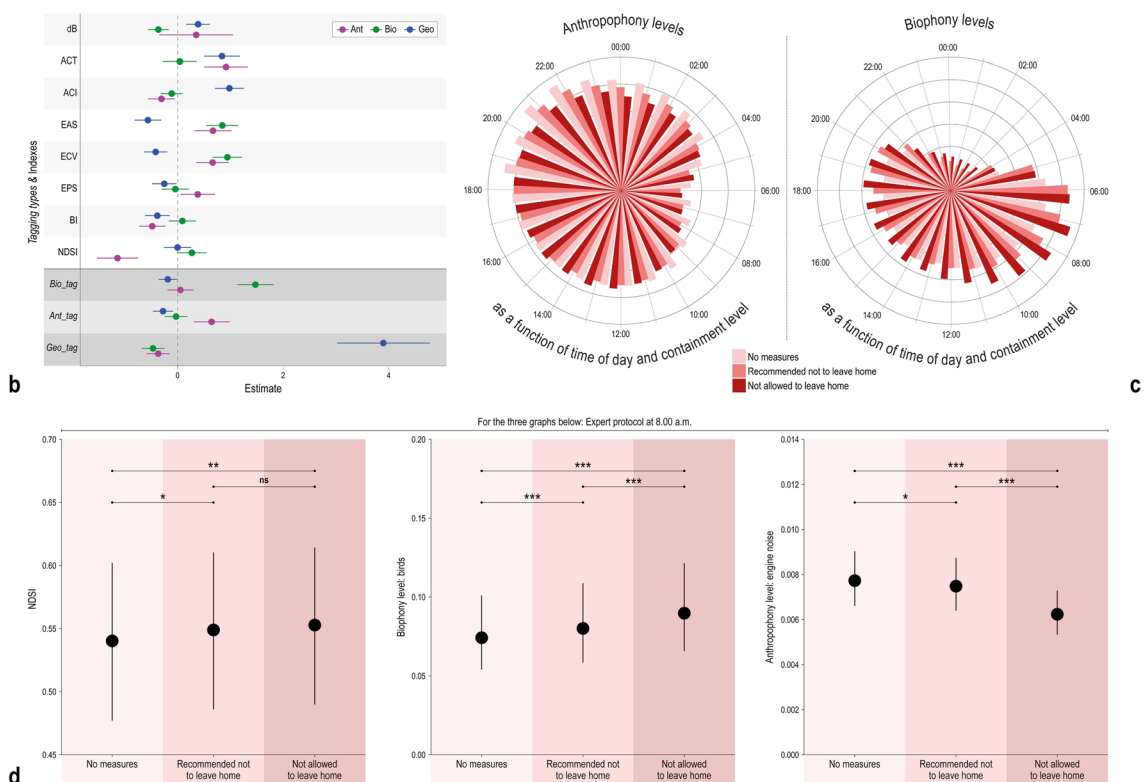


Fig. 2 Panel a. Top left: Professional fields of the 261 participants. Middle: Distribution of type of devices for the 317 recording sites. Bottom left: type of protocol implemented. Right: Photos of the two main recording devices used: Open Acoustic Devices AudioMoth (top) and Wildlife Acoustics SM4 (bottom). Panel b. Association between the measured acoustic indices and tagging types and the presence of geophysical (Geo), biophysical (Bio) and anthropophonical (Ant) events detected manually by the contributors. Model estimates and associated 95% confidence intervals are represented with points and bars, respectively. Positive and negative estimates with confidence intervals not overlapping zero indicate positive and negative associations, respectively. Panel c. Radial barplots depicting the mean anthropophony and biophony level values per site, combining all protocols, recorded hourly throughout each period of COVID-19 containment levels. Panel d. Model predictions and associated 95% confidence intervals for NDSI, biophony (birds) and anthropophony (engine noise) levels at 8:00 a.m. during the COVID-19 confinement measures, following the expert protocol only. *** $p < 0.001$, ** $p < 0.010$, * $p < 0.050$, ns: $p > 0.050$.

- Entropy of the Spectral Peaks or EPS⁶² is defined as a measure of the evenness or ‘flatness’ of the maximum-frequency spectrum, maximal frequencies being measured along the time of the recording. A recording with no acoustic activity should show a low EPS value, as all spectral maxima are low and constant over time;
- Normalized Difference Soundscape Index or NDSI⁶⁴ measures a ratio between biophony and anthropophony. The value of this index is calculated on a spectrogram and varies between -1, meaning the entire acoustic energy of the recording is concentrated under the frequency threshold of 2kHz and attributed to anthropophony only, and +1, meaning the entire acoustic energy of the recording is concentrated above the frequency threshold and attributed to biophony only.

Manual soundscapes description. In order to have a more thorough description of the recorded soundscapes, some contributors manually performed sound identification on a subset of their recordings. Two non-consecutive days of recordings were randomly selected for each site and each one-minute-long audio file recorded at the beginning of each hour was analysed (*i.e.* a total of 48 1-min files). Using software dedicated to sound analysis (e.g. Audacity: <https://www.audacityteam.org/>, Sonic visualizer: <https://www.sonicvisualiser.org/>, and Kaleidoscope: <https://www.wildlifeacoustics.com/uploads/user-guides/Kaleidoscope-User-Guide.pdf>), contributors were to (i) listen and view spectrograms of the recordings, (ii) estimate the percentage of time occurrence (0%, 1-25%, 25-50%, 50-75% and 75-100%) of geophonic, biophonic, and anthropophonic events in each audio file, and (iii) provide more information about the source/type (e.g. geophony: wind, rain and river; biophony: birds, mammals and insects; anthropophony: car, plane and music) of each event. They further indicated the strength/intensity (on a scale from 0 to 3) of the identified geophonic and anthropophonic events and to provide for each biophonic event the number of different song/call/stridulation types visible on the spectrogram. Scoring per recording was associated with a confidence level on a scale from 1 to 5 (see Table 3 for an example of the identification table, inspired from protocol proposed in⁶⁵).

A total of 1351 minutes of sounds were manually described from 30 sites. Contributors from Europe (Austria, Czech Republic, France, Germany, Ireland, Poland, Portugal, Serbia, and United Kingdom), the Americas (Canada, Colombia, Mexico, and United States of America), and Australia participated in the manual sound identification process. The number of audio files described varied slightly between participants (min: 9 minutes, max: 96, median: 48, mean: 45). Most audio files manually analyzed were recorded using AudioMoth (19 sites) and SM4 (8 sites).

Recordings were dominated by geophonic, biophonic and anthropophonic events (*i.e.* time occurrence >75% within 1-min files) in 20, 34 and 51% of the 1351 minutes of sounds recorded, respectively. The most detected geophonic sounds were from wind (26% of the total number of records, including 76 records with strong wind intensity) and rain (12%). Bird calls (63%) and insect stridulations (16%) were the most encountered biophonic sounds. Around one third of the recordings with bird calls contained at least four different bird call types. Noise from cars (61%) and people talking (26%) were responsible for most of the anthropophonic sounds.

Automatic sound event recognition. Automatic sound event recognition (SER) became an essential task due to the immense volume (around 20 Terabytes) of the Silent Cities dataset. We adopted the AudioSet ontology and dataset⁶⁶, which covers a wide range of everyday sounds. We explored the viability of utilizing PANNs (pre-trained audio neural networks) pretrained on the full AudioSet data (available online: https://github.com/qiuqiangkong/audioset_tagging_cnn). The choice of a pretrained model was driven by its generality, as it has been exposed to a wide range of sounds, rendering it suitable for recognizing various audio events. In implementing our methodology, we employ a zero-shot inference approach. This involves applying the pretrained model directly to the entirety of the Silent Cities recordings without the need for additional training or fine-tuning. By doing so, we can benefit from the model’s generalization capabilities and avoid the time-consuming process of manual annotation. To categorize the diverse audio events within our dataset, we leverage the AudioSet ontology and make necessary adaptations. Specifically, we classify the sounds into three main types: anthropophony (sounds produced by human activities), biophony (sounds originating from natural living organisms), and geophony (sounds resulting from non-living sources like weather or geological activities). The details of sound event grouping (*i.e.* audio tagging types) and corresponding labels are presented in Table 4 to provide clarity and consistency in the classification process. This grouping was also done to have the same categories than in the manual annotation described in the previous section.

Voice activity detection. As many recordings in Silent Cities were performed at home (e.g. on a balcony) during periods of containment, human voices are likely to be heard and speakers may be easily identified. In order to prevent issues related to privacy, we identified audio segments containing speech and only shared in open access the audio segments without speech. Voice activity detection was conducted using a general purpose voice activity detector (GP-VAD) that was pretrained on noisy, natural speech recordings in the wild⁶⁷ (available online: <https://github.com/RicherMans/GPV>). We applied GP-VAD on a subset of 250000 one-minute recordings (approx. 24 weeks). Detections on this subset were considered as a ground truth speech label, that we set a reference to detect speech in the entirety of the Silent Cities dataset, for which we have a weak speech label from the AudioSet SER (described in the previous paragraph). More precisely, we used the GP-VAD predictions on the subset to estimate a receiver-operator characteristic curve, and by setting a true positive rate of detecting 75 % of speech recordings, we obtain an average false positive rate of 34 % false alarms when using the AudioSet SER. The corresponding threshold was applied on the raw probability from the AudioSet SER on the entirety of the dataset, which eventually resulted in a rejection of 2,868,098 10-second audio segments, representing approximately 18 % of the dataset.

Name	Description	Type	Number of files
Collection of acoustic recordings	Preprocessed 10-second audio files from soundscape recordings collected for each site (compressed in tar.gz archives)	FLAC	16,252,373
Glossary	Definitions of table elements	csv	1
dB	List of readable files uploaded by the contributors and their dB level (archived in a single zip file)	csv	317
Site	Information about each site including contributors' description about the recorder (e.g. type and serial number), the location (e.g. description of the surrounding area, city), and the description of the containment measures in place at the time of deployment. Also contains the metadata describing the landscape (e.g. population density, climate) corresponding to the cities of the dataset as well as the extracted information about the protocol used (e.g. type, sampling rate, file duration), and the amount of data collected	csv	1
ConfinementLevels	For each country and date covered by the acoustic collection, displays information about the levels of "stay-at-home requirements" according to the dataset built by the University of Oxford	csv	1
SunMoon	Information about the sun and moon azimuth and altitude for the dates and times covered by the Silent Cities dataset, with a 10-second increment, for each city (197 csv files in a zip file)	csv	197
AcousticMeasurements	List of preprocessed 10-second acoustic files and associated calculations of acoustic indices and categories of automatic sound event (all csv files compressed in a single zip file)	csv	317
AcousticMeasurements_nospeech	Same as AcousticMeasurements but only for recordings without speech (all csv.gz files in a single zip file)	csv.gz	317
ManualIdentification	Sound event identification made by contributors on a subsample of the original 1-min recordings	csv	1
AverageCompleteTable	For each unique site at a unique date and a unique hour, averaged values of acoustic indices and automated event recognition categories. This table also includes the corresponding Site, SunMoon, and ConfinementLevels information. Finally, given the original recording date and time in UTC+0 and knowing the associated timezone, a local date and time information was calculated	csv	1
AverageCompleteTable_nospeech	Same as AverageCompleteTable but the averaged values are only calculated on speech-filtered subsample of the acoustic collection	csv	1

Table 1. Silent Cities dataset description.

Response variable	Explanatory variable	Estimate	SE	Z value	P value
NDSI	Intercept	0.121	0.243	0.497	0.619
	Confinement level 1 vs no measures	0.035	0.012	2.832	0.005**
	Confinement level 2 and 3 vs no measures	0.051	0.014	3.535	<0.001***
	PCA axis: degree of anthropization	-0.190	0.052	-3.645	<0.001***
	Julian day: season	-0.083	0.005	-17.930	<0.001***
	Climate: dry vs tropical	0.498	0.366	1.358	0.174
	Climate: temperate vs tropical	0.018	0.230	0.077	0.938
	Climate: continental vs tropical	0.023	0.276	0.082	0.934
Birds	Intercept	-2.408	0.264	-9.137	<0.001***
	Confinement level 1 vs no measures	0.083	0.013	6.191	<0.001***
	Confinement level 2 and 3 vs no measures	0.207	0.015	14.074	<0.001***
	PCA axis: degree of anthropization	-0.115	0.044	-2.595	0.009**
	Julian day: season	-0.131	0.005	-28.449	<0.001***
	Climate: dry vs tropical	0.381	0.353	1.079	0.281
	Climate: temperate vs tropical	-0.153	0.238	-0.642	0.521
	Climate: continental vs tropical	-0.136	0.287	-0.476	0.634
Engine noise	Intercept	-4.844	0.207	-23.445	<0.001***
	Confinement level 1 vs no measures	-0.033	0.011	-2.887	0.004**
	Confinement level 2 and 3 vs no measures	-0.216	0.014	-15.130	<0.001***
	PCA axis: degree of anthropization	0.091	0.049	1.845	0.065
	Julian day: season	0.054	0.005	10.839	<0.001***
	Climate: dry vs tropical	-0.220	0.351	-0.626	0.531
	Climate: temperate vs tropical	0.022	0.220	0.100	0.920
	Climate: continental vs tropical	-0.152	0.265	-0.573	0.566

Table 2. Outputs of the full GLMMs relating the effects of COVID-19 confinement measures (alongside covariates) on NDSI, biophony (here, probability of bird calls) and anthropophony (here, probability of engine noise) levels at 08:00 am. Δ AIC values between the full and the null models are, from top to bottom, 593, 2056 and 1023, thus indicating that the full models were more informative than the null ones. SE: standard error of the estimate. *** $P < 0.001$, ** $P < 0.010$, * $P < 0.050$. Confinement level 1 calls for "recommended not to leave home" and Confinement level 2 and 3 calls for "not allowed to leave home".

Data Records

The dataset⁴⁵ comprises the entire collection of acoustic recordings in Free Lossless Audio Codec (FLAC) format and associated metadata spread across several Comma Separated Value (CSV) tables (see Table 1). In order to

Variable	Definition	Possible value and range
Geophony_TempLevel	range of occupancy	0%/1-25%/25-50%/50-75%/75-100%
Wind	strength	0/1/2/3
Rain	strength	0/1/2
Wave	strength	0/1/2
Thunder	strength	0/1
Biophony_TempLevel	range of occupancy	0%/1-25%/25-50%/50-75%/75-100%
Bird	range of song types number	0/1-3/4-6/7-8/9-11/>11
Amphibian	range of song types number	0/1-3/4-6/7-8/9-11/>11
Insect	range of song types number	0/1-3/4-6/7-8/9-11/>11
Mammal	range of song types number	0/1-3/4-6/7-8/9-11/>11
Reptile	range of song types number	0/1-3/4-6/7-8/9-11/>11
Antropophony_TempLevel	range of occupancy	0%/1-25%/25-50%/50-75%/75-100%
Walking	presence/absence	0/1
Cycling	presence/absence	0/1
Beep	presence/absence	0/1
Car	sound intensity	0/1/2
Car honk	presence/absence	0/1
Motorbike	sound intensity	0/1/2
Plane	presence/absence	0/1
Helicopter	presence/absence	0/1
Boat	presence/absence	0/1
Other_motors	sound intensity	0/1/2
Shoot	presence/absence	0/1
Bell	presence/absence	0/1
Talking	presence/absence	0/1
Music	presence/absence	0/1
Dog bark	presence/absence	0/1
Kitchen sounds	presence/absence	0/1
Rolling shutter	presence/absence	0/1
Confidence level	low (0) to high confidence (5)	0/1/2/3/4/5

Table 3. Summary table of the typology used to manually describe the recordings.

protect privacy, only the preprocessed 10-second audio files with no speech identified are in direct open access on the OSF website (<https://doi.org/10.17605/OSF.IO/H285U>).

Technical Validation

To validate the Silent Cities dataset⁴⁵, we verified the veracity of the metadata reported by the contributors and consolidated the acoustic recordings collections by checking for device malfunctions. We also verified whether the automated acoustic measurements conducted on the recordings were coherent with aural human observations. Finally, proof of validity of the dataset to reflect urban soundscape changes due to stay-at-home requirements is presented. The three steps of this technical validation are detailed below.

First, we verified the quality of the data by manually verifying that the recordings were correctly attributed to their dedicated site with the help of the contributors. We also ran a manual cleaning of information given by the contributors to remove any personal information, such as address or GPS coordinates, and to correct spelling mistakes to ensure interoperability between tables. In addition, we verified the conformity of the protocol by automatically extracting information from the recording collection (*i.e.* frequency range, schedule of recordings) and reported observed modification of the protocol. We also automatically and manually verified the proper calculation of acoustic measurements and identified 10,724 files for which the calculation failed, probably due to file-related issues; these files were excluded from the dataset without affecting an entire site (*i.e.* no sites were excluded because of this issue). Finally, we checked for recorder device malfunction by making sure of a temporal variation of the dB value for each recorder, only one site was identified with a flat dB response, leading to its exclusion from the dataset.

Second, we confirmed that the automated soundscape measurements informed and aligned with real soundscape events. More specifically, we investigated whether the acoustic indices and audio tagging categories were representative of geophonic, biophonic and anthropophonic events detected manually by the contributors. To do

Final tag name	Corresponding labels in AudioSet Ontology	Category
Wind	Wind	Geophony
Rain	Rain	
River	Stream/Waterfall	
Wave	Ocean	
Thunder	Thunderstorm	
Bird	Bird vocalization, bird call, bird song/Pigeon, dove/Crow/Owl/Gull, seagull	Biophony
Amphibian	Frog	
Insect	Insect	
Mammal	Rodents, rats, mice/Canidae, dogs, wolves	
Reptile	Snake	Anthropophony
Walking	Run/Walk, footsteps	
Cycling	Bicycle/Bicycle bell	
Beep	Reversing beeps	
Car	Car passing by/Tire squeal	
Car honk	Vehicle horn, car horn, honking	
Motorbike	Motorcycle	
Plane	Aircraft engine/Fixed-wing aircraft, airplane	
Helicopter	Helicopter	
Boat	Motorboat, speedboat/Ship/Sailboat, sailing ship	
Other motors	Traffic noise, roadway noise	
Shoot	Gunshot, gunfire	
Bell	Chime/Jingle bell/Cowbell/Church bell/Change ringing (campanology)	
Talking	Speech/Hubbub, speech noise, speech babble/	
Music	Music	
Dog bark	Dog	
Rolling shutter	Power windows, electric windows	
Kitchen sounds	Door/Cupboard open or close/Drawer open or close/Dishes, pots, and pans/ Cutlery, silverware/Chopping (food)/Sink (filling or washing)/Water tap, faucet/Kettle whistle/ Microwave oven/Blender	

Table 4. Mapping between the Silent Cities tags and the labels from the AudioSet ontology. Each tag is computed using the maximum probability output from the pretrained network among the corresponding Audioset labels. Finally, the three tags Anthropophony, Geophony and Biophony are computed using the maximum tag probability in the category.

so, we conducted a series of univariate generalized linear mixed-effect models (GLMMs; ‘glmmTMB’ package⁶⁸) in R v4.2.1. We tested independently the presence/absence of geophonic, biophonic and anthropophonic events within the 1351 1-min recordings (*i.e.* response variables) in relation to acoustic indices and tagging types (*i.e.* explanatory variables). Models were fitted with a binomial error distribution and a logit link function. We considered the identity of contributors as a random effect to avoid pseudoreplication. We also implemented a first-order autoregressive function to account for serial autocorrelation in residuals. Statistical assumptions were visually assessed using model diagnostics (*i.e.* Quantile-Quantile plot, residuals vs fitted plot) with the DHARMA package⁶⁹. The acoustic indices were linked to geophonic, biophonic, or anthropophonic events, albeit to varying degrees (Fig. 2b). For instance, the presence of biophonic events was associated with greater values of EAS and ECV and lower values of dB. Audio tagging categories effectively captured the intended soundscapes they aimed to portray (Fig. 2b).

Third, we assessed the validity of the dataset in evaluating the impact of stay-at-home requirements on soundscapes. In a first step, we plotted the mean values of biophony and anthropophony levels (here defined as the maximum probability of having a biophonic and anthropophonic event in the 1-min recording, respectively) per site recorded at each hour (all protocols combined). As expected, we observed temporal patterns in biophony and anthropophony levels throughout the day (Fig. 2c). Regardless of the time of day, biophony levels were greater during the period when leaving home was not permitted (*i.e.* confinement level 2 or 3) compared to the other periods, while the opposite pattern was true for anthropophony. In a second step, we modeled changes in the values of acoustic indices as well as biophony and anthropophony levels (*i.e.* response variables) in relation to the containment measures (*i.e.* explanatory variables) using GLMMs with a beta distribution and a log link function. We aimed to provide a proof of validity and therefore limited the analysis to the expert protocol and all recordings collected at 8:00 am (*i.e.* peak of biophonic and anthropophonic events; Fig. 2d). We focused on NDSI for the acoustic index and the probability of bird calls and engine noise indicated by the automatic sound event recognition in the recordings as proxies of biophony and anthropophony levels, respectively. We added as covariates in the models: (i) Julian day to consider seasonal changes in biological and anthropogenic sounds, (ii) the first Principal Component Analysis axis depicting the level of anthropization in the landscape

Response variable	Explanatory variable	Estimate	SE	t ratio	P value
NDSI	Confinement level 1 vs no measures	0.035	0.012	2.832	0.013*
	Confinement level 2 and 3 vs no measures	0.051	0.014	3.535	0.001**
	Confinement level 2 and 3 vs Confinement level 1	0.016	0.011	1.464	0.309
Birds	Confinement level 1 vs no measures	0.083	0.013	6.191	<0.001***
	Confinement level 2 and 3 vs no measures	0.207	0.015	14.074	<0.001***
	Confinement level 2 and 3 vs Confinement level 1	0.124	0.011	11.272	<0.001***
Engine noise	Confinement level 1 vs no measures	-0.327	0.011	-2.887	0.011*
	Confinement level 2 and 3 vs no measures	-0.216	0.014	-15.130	<0.001***
	Confinement level 2 and 3 vs Confinement level 1	-0.183	0.01	-16.999	<0.001***

Table 5. Results of the post hoc pairwise comparisons applied to the GLMMs relating the effects of COVID-19 confinement measures (alongside covariates) on NDSI, biophony (here, probability of bird calls) and anthropophony (here, probability of engine noise) levels at 08:00 am. SE: standard error of the estimate. *** $P < 0.001$, ** $P < 0.010$, * $P < 0.050$. Confinement level 1 calls for “recommended not to leave home” and Confinement level 2 and 3 calls for “not allowed to leave home”.

surrounding the recordings, and (iii) the climate type. Continuous covariates were scaled (mean = 0; SD = 1) to avoid convergence issues. We considered as random effects site identity nested within country to account for hierarchical clustering within data and recorder type, due to potential sensitivity differences between devices. Due to the limited data collected during the strictest containment period, we combined data from the two periods when leaving home was not permitted. The same approach as outlined previously was employed for model validation (note that the validity of the statistical assumptions, assessed using Quantile-Quantile and residuals vs fitted plots, was only partially met for the engine noise model). Full models were more informative than the null ones with differences in Akaike Information Criterion scores > 500 . Finally, we conducted Tukey’s post hoc multiple comparison test to investigate pairwise differences in NDSI values and biophony and anthropophony levels between the three COVID-19 containment measures investigated. Overall, we found that COVID-19 lockdown had positive effects on NDSI values and biophony levels and negative effects on anthropophony levels. After accounting for seasonal and landscape effects, our models suggest that NDSI values and biophony levels were significantly greater during the periods when leaving home was not recommended or permitted, compared to the period with no measures (Fig. 2d; Table 2). There were also higher biophony levels during the period when leaving home was not permitted than during the period with when leaving home was not recommended. The opposite patterns were found for the anthropophony levels, with significantly lower values measured during the periods when leaving home was not permitted compared to the other periods, albeit the differences were of smaller magnitude (Fig. 2d; Table 5). Altogether, our preliminary analysis revealed potential changes in soundscape patterns that can be attributed to containment policies, these changes being above expected differences due to climate.

Usage Notes

The Silent Cities dataset could be considered for multiple applications. In the specific fields of bio/ecoacoustics, it could be used to study the effect of containment measures on urban soundscapes²⁹, to improve the performance of acoustic indices in urban environments⁷⁰ and to gain a deeper understanding of the interplay between biophony and urban environment characteristics⁷¹. In the field of machine learning (machine listening, deep learning), it will allow the testing of difficult cases of generalization in sound event recognition from one site to another, due to the variety of sampled sites⁷². In the interdisciplinary field of territorial sciences (e.g. economic geography, territorial economics, spatial planning, urban engineering sciences), it will make it possible to analyze the links between the levels of economic activity of a city and the levels of noise pollution. Finally, for environmental sciences interested in well-being and relationships between humans and non-humans within urban socio-ecosystems (e.g. environmental and health psychology, landscape design, environmental geography, etc.), this dataset opens up opportunities for the qualitative study of individual and subjective perceptions of the different soundscape configurations collected. More broadly, we aim for this international and collaborative dataset to be usefully mobilized in any research working to make better coexistence between humans and non-humans possible, and thus working to maintain the Earth’s habitability conditions for all of them.

The Silent Cities dataset⁴⁵ is available under the terms of a Creative Commons Attribution 4.0 International waiver (CC-BY 4.0, <https://creativecommons.org/licenses/by/4.0/>). The CC-BY-4.0 waiver facilitates the discovery, re-use, and citation of the dataset. When using all or part of the dataset, we require anyone to cite both the dataset⁴⁵ and this publication.

Code availability

The recording manipulation and acoustic measurements were run using Python, <https://github.com/brain-bzh/SilentCities> and the analyses were run on R <https://github.com/agasc/SilentCities-R>.

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Author contributions

S.C. introduced the concept; S.C., A.G., J.S.P.F. and N.F. designed the protocol; S.C., A.G., J.S.P.F., N.P., N.F. and the Silent Cities Consortium collected the soundscape recordings; A.G., N.P. managed the data; A.G., J.S.P.F., N.P. and N.F. conducted the data analysis; A.G. and J.S.P.F. conducted the technical validation. N.F. was in charge of high performance computing; S.C., A.G., J.S.P.F., N.P. and N.F. wrote the initial draft; all authors including the Silent Cities Consortium reviewed the manuscript.

Competing interests

The authors declare no competing interests.

Additional information

Correspondence and requests for materials should be addressed to S.C. or N.F.

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Silent Cities project consortium

Carlos R. Abrahams⁹, Orlando Acevedo-Charry^{10,11}, Ludmilla M. S. Aguiar¹², Zachary R. Ahlin¹³, Franz Aiple¹⁴, Cécile H. Albert⁶, Irene Alcocer^{15,16}, Ana Sofia Alves¹⁷, Francisco Amorim^{18,19,20}, Ludmila B. Andrade²¹, Pedro M. Araújo^{20,22}, Fernando Ascensão²³, Serge Aucoin²⁴, Elias Bader²⁵, Diego Balbuena²⁶, Luc Barbaro²⁷, Eder Barbier²⁸, Eliana Barona Cortés¹¹, Luis Emilio Barrie²⁹, José L. Bartheld³⁰, Henry Bates¹³, Alice Baudouin³¹, Richard D. Beason³², Christa Beckmann^{32,33,34}, Amy Beeston³⁵, Gvan Belá, Kristen M. Bellisario³⁶, Simon Belshaw, Juan F. Beltrán³⁷, Raone Beltrão-Mendes³⁸, Enrico Bernard²⁸, Thierry Besche³⁹, Peter A. Biro³⁴, Cathie Boléat⁴⁰, Mathieu Bossaert, Ally Bradley¹³, Paulo Branco⁴¹, Wijnand Bredewold, Philip A. Briggs⁴², Sylvio Romério Briglia-Ferreira⁴³, Emily Buckner¹³, Ivana Budinski⁴⁴, Albane Burens¹, Rachel T. Buxton⁴⁵, Andrés Canavero⁴⁶, Paulo Cardoso⁴⁷, Farah Carrasco-Rueda⁴⁸, Paula C. Caycedo⁴⁹, Frédéric Cazaban⁵⁰, Lara R. Cerveira²², Ada Ceuppens, Alain Challéat, Angela Chappa Larrea¹³, Adrien Charbonneau⁶, Mina Charnaux⁵¹, Pooja Choksi⁵², Jan Cibulka, Julián Clavijo-Bustos^{53,54}, Zuania Colón-Piñeiro⁵⁵, Sofia Conde⁴¹, Maria João Costa⁵⁶, António Cotão⁵⁷, Clément Couturier, Marina D. A. Scarpelli⁵⁸, Luis P. da Silva^{18,20}, Tom Davis⁵⁹, Nathalie de Lacoste, Sarah L. Deans, Serge Dentin, Krzysztof Deoniziak⁶⁰, Sarah R. Dodgin, Ivo dos Santos²², Tudor I. Draganoiu⁶¹, Bruno Drolet⁶², Marina H. L. Duarte^{63,64}, Gonçalo Duarte⁴¹, Chloé Dubset, Frank Dziock⁶⁵, Alice Eldridge⁶⁶, Simon Elise^{67,68}, David R. Elliott⁶⁹, Arthur Enguehard⁷⁰, Karl Esztl, Darren M. Evans⁷¹, Daniel M. Ferreira^{72,73}, Sonia A. F. Ferreira^{18,20}, Diogo F. Ferreira^{18,20}, Ana Margarida Ferreira^{22,74}, Penelope C. Fialas^{5,42}, Lauren Foster-Shaner¹³, Bárbara Freitas⁷⁵, Nicholas R. Friedman^{76,77}, Susan Fuller⁵⁸, Didier Galop¹, Daniel Garside⁷⁸, Jean-Christophe Gattus, Sylvain Geoffroy, Louis Godart, Laurent Godet⁷⁹, Inês Gomes Marques⁴¹, Fernando González-García⁸⁰, Paul Griesberger⁸¹, Bilal Habib⁸², Madeline E. Hallet¹³, Meena M. Haribal⁸³, Jennifer Hatlauf⁸¹, Sylvain Haupt⁸⁴, José M. Herrera^{85,86}, Sierra E. Herzberger¹³, Frederico Hintze Oliveira⁸⁷, Kathy H. Hodder⁸⁸, Isabelle Hoehler¹³, Mark F. Hulme⁸⁹, Emilia Hyland⁹⁰, Michel Jacobs, Akash Jaiswal⁹¹, Laurent Jégou⁹², Steve Jones⁹³, Hervé Jourdan⁹⁴, Tomáš Jůnek⁹⁵, Leili Khalatbari^{18,20,96}, Sarika Khanwilkar^{52,82,97}, James J. N. Kitson⁷¹, Amanda H. Korstjens⁸⁸, Kim Krähenbühl-Künzli²⁵, Natalija Lace⁹⁸, Sébastien Laguet⁹⁹, Hedwig Lankau¹⁰⁰, Thiago O. Laranjeiras¹⁰¹, Gregoire Lauvin¹⁰², Samuel Lavin¹⁰³, Matthieu Le Corre⁶⁸, Monica León¹⁰⁴, Judah J. Levenson^{105,106}, Pavel Linhart¹⁰⁷, Juliette Linossier¹⁰⁸, Diego J. Lizcano¹⁰⁹, Diego Llusia^{15,16,110}, Marty Lockett¹¹¹, Pedro B. Lopes¹¹², Ricardo Jorge Lopes^{23,113}, José Vicente López-Bao¹¹⁴, Adrià López-Baucells¹¹⁵, David López-Bosch¹¹⁵, Ricardo B. Machado¹², Claude Mandé¹¹⁶, Guillaume Marchais, Fabio Marcolin^{19,117}, Oscar H. Marn Gómez^{118,119}, Carina B. Marques¹²⁰, J. Tiago Marques^{86,121}, Tilla Martin, Vanessa Mata^{18,20}, Eloisa Matheu-Cortada¹²², Vincent Médoc¹²³, Kirsten E. Miller^{71,124}, Basile Montagne¹²⁵, Allen Moore, JoMari M. A. Moreno¹³, Felipe N. Moreno-Gómez³⁰, Sandra Mueller¹²⁶, Daniela Murillo-Bedoya¹²⁷, Luciano N. Naka¹²⁸, Adrian C. Newton⁸⁸, João T. Nunes¹²⁹, Pierrette Nyssen^{130,131}, Fionn Ó Marcaigh¹³², Darren P. O'Connell^{71,133}, M. Teague O'Mara^{134,135,136}, David Ocampo¹³⁷, Meryem Ouertani¹³⁸, Jan Olav Owren, Vitor

H. Paiva²², Stéphane Paris, Marion Parisot, Swaroop Patankar¹³⁹, Jorge M. Pereira²², Slvia Pereira Barreiro⁸⁶, Cédric Peyronnet, Magali Philippe¹, Bryan C. Pijanowski^{36,140}, Nuno Pinto^{141,142}, Zach Poff¹⁴³, Jonathan M. Poppele, Andrew Power¹⁴⁴, Victoria Pratt¹⁴⁵, Darren S. Propp¹⁴⁶, Raphaël Proulx¹⁴⁷, Laura Prugh¹³, Sebastien J. Puechmaile^{148,149}, Xavier Puig-Montserrat¹¹⁵, Lorenzo Quaglietta^{18,150}, John E. Quinn⁹⁰, Nancy I. Quiroga¹⁵¹, Mariana Ramos¹⁵², Rebecca Rasmussen¹⁵³, Georges Reckinger¹⁵⁴, Mimi Reed¹³, Jean-Benoît Reginster¹⁵⁵, Vanesa Rivera⁸⁶, Clara F. Rodrigues¹⁴¹, Patricia Mara Rodriguez-González⁴¹, Eduardo Rodriguez-Rodriguez³⁷, Luke Romaine⁵, Andrei L. Roos¹⁵⁶, Joao Rosa, Samuel R. P.-J. Ross^{132,157}, Quentin Rouy^{158,159}, Alyssa M. Ryser¹³, Sougata Sadhukhan^{82,160}, Robin Sandfort^{81,161}, José M. Santos⁴¹, David Savage^{36,140}, Stéphanie C. Schai-Braun⁸¹, Michael Scherer-Lorenzen¹²⁶, Mathilde Schoenauer Sebag, Pedro Segurado¹¹⁷, Ana M. Serronha^{18,20}, Taylor Shaw¹²⁶, Brenda Shepherd²⁴, Cáról Sierra-Durán¹⁶², Bruno M. Silva⁸⁶, Victoire Simon, Peter F. Sinclair¹⁰², Carolina Soto-Navarro¹⁶³, Anne Sourdil¹⁶⁴, Jérôme Sueur⁸⁴, Larissa S. M. Sugai^{15,16,165}, Ian B. Tarrant, Fran Tattersall¹⁶⁶, Christopher N. Templeton¹⁶⁷, Michelle E. Thompson⁴⁸, Marcela Todd¹³, Juan D. Tovar-García^{168,169}, Karina Townsend¹⁷⁰, Amaro Tuninetti¹⁷¹, Paul A. Ullrich¹⁶⁶, Juan S. Vargas Soto¹⁷², Kevin Vega¹⁷³, Gabriella Ventrice, Pierre J. Victor¹⁷⁴, Josep Vidal Oliveras¹⁷⁵, Sara Villén-Pérez¹⁷⁶, Olivier Vinet¹⁷⁷, Agnés Vivat¹⁷⁸, Jean-Do. Vignault, William D. J. Walton, Christopher J. Watson¹⁴⁷, Oliver R. Wearn¹⁷⁹, Damion L. Whyte¹⁸⁰, Fredric M. Windsor⁷¹, Yanchen Wu¹³, Selena Xie¹³, Ignacio Zeballos Puccherelli & Vera Zina⁴¹

⁹Nottingham Trent University, Nottingham, UK. ¹⁰School of Natural Resources and Environment, Department of Wildlife Ecology and Conservation, & Florida Museum of Natural History, University of Florida, Gainesville, FL, USA. ¹¹Environmental Sound Collection, Instituto de Investigación de recursos biológicos Alexander von Humboldt, Boyacá, Colombia. ¹²University of Brasilia, Department of Zoology, Campus Darcy Ribeiro s/n, Asa Norte, Brasilia, DF, Brazil. ¹³School of Environmental and Forest Sciences, University of Washington, Seattle, WA, USA. ¹⁴NABU Freiburg, Freiburg im Breisgau, Breisgau, Deutschland. ¹⁵Terrestrial Ecology Group, Departamento de Ecologia, Universidad Autónoma de Madrid, Madrid, Spain. ¹⁶Centro de Investigación en Biodiversidad y Cambio Global, Universidad Autónoma de Madrid, Madrid, Spain. ¹⁷Instituto Gulbenkian de Ciência, Oeiras, Portugal. ¹⁸CIBIO (Centro de Investigação em Biodiversidade e Recursos Genéticos), InBIO Laboratório Associado, Universidade do Porto, Vairão, Portugal. ¹⁹CIBIO (Centro de Investigação em Biodiversidade e Recursos Genéticos), InBIO Laboratório Associado, Universidade de Lisboa, Lisboa, Portugal. ²⁰BIOPOLIS (Program in Genomics, Biodiversity and Land Planning), CIBIO (Centro de Investigação em Biodiversidade e Recursos Genéticos), Vairão, Portugal. ²¹Núcleo de Pesquisa Flora Tristán: representações, conflitos e direitos. Instituto de Ciência Política, Universidade de Brasilia (UnB), Brasilia, DF, Brasil. ²²University of Coimbra, MARE (Marine and Environmental Sciences Centre), ARNET (Aquatic Research Network), Department of Life Sciences, Calçada Martim de Freitas, 3000-456, Coimbra, Portugal. ²³cE3c (Centre for Ecology, Evolution and Environmental Changes), CHANGE (Global Changes and Sustainability Institute), Faculdade de Ciências da Universidade de Lisboa, Lisbon, Portugal. ²⁴Parks Canada Agency, Jasper National Park, Alberta, Canada. ²⁵Bat Conservation Switzerland, Zürichbergstrasse, Zürich, Switzerland. ²⁶Wildlife Consulting & Equipment S.R.L, Sachaca, Arequipa, Peru. ²⁷UMR1201 DYNAFOR (Dynamiques et écologie des paysages agriforestiers), INRAE, Université de Toulouse, 31326, Castanet-Tolosan, France. ²⁸Laboratório de Ciência Aplicada à Conservação da Biodiversidade, Department of Zoology, Universidade Federal de Pernambuco, Recife, PE, Brazil. ²⁹Programa de Doctorado Territorio, Espacio y Sociedad de la Facultad de Arquitectura y Urbanismo, Universidad de Chile, Santiago, Chile. ³⁰EcoBioAcústicaLab, Departamento de Biología y Química, Facultad de Ciencias Básicas, Universidad Católica del Maule, Talca, Chile. ³¹LPO (Ligue pour la protection des oiseaux) Auvergne-Rhône-Alpes, Lyon, France. ³²School of Science, Western Sydney University, Locked Bag 1797, Penrith, NSW, 2751, Australia. ³³Hawkesbury Institute for the Environment, Western Sydney University, Locked Bag 1797, Penrith, NSW, 2751, Australia. ³⁴School of Life and Environmental Sciences, Deakin University, Geelong, Victoria, 3216, Australia. ³⁵Art and Design Department, University of the Highlands and Islands, Orkney, UK. ³⁶Center for Global Soundscapes, Purdue University, West Lafayette, IN, USA. ³⁷Department of Zoology, University of Seville, Seville, Spain. ³⁸Post-Graduate Program in Ecology and Conservation, Universidade Federal de Sergipe, São Cristóvão, Brazil. ³⁹Passerelle Arts Sciences Technologies, Albi, France. ⁴⁰CEN Occitanie (Conservatoire d'espaces naturels d'Occitanie), Montpellier, France. ⁴¹Forest Research Centre, School of Agriculture, University of Lisbon, Tapada da Ajuda, Lisbon, Portugal. ⁴²Bat Conservation Trust, London, UK. ⁴³ICMBio (Instituto Chico Mendes de Conservação da Biodiversidade), Boa Vista, Roraima, Brasil. ⁴⁴Department of Genetic Research, Institute for Biological Research "Sinisa Stankovic" - National Institute of Republic of Serbia, University of Belgrade, Belgrade, Serbia. ⁴⁵Carleton University, Ottawa, ON, Canada. ⁴⁶Departamento de Ecología y Gestión Ambiental, Universidad de la República, Montevideo, Uruguay. ⁴⁷Bioinsight, Lda, Odivelas, Portugal. ⁴⁸Keller Science Action Center, Science & Education, Field Museum of Natural History, Chicago, IL, USA. ⁴⁹Biodiversa Foundation US, Fundación Biodiversa Colombia, Bogotá, Colombia. ⁵⁰CPPIE (Centre permanent d'initiatives pour l'environnement) Seignaux et Adour, Saint-Martin-de-Seignanx, France. ⁵¹Eurometropolis of Strasbourg, Strasbourg, France. ⁵²Department of Ecology, Evolution and Environmental Biology, Columbia University, New York, NY, USA. ⁵³Sección de Entomología, Colecciones Biológicas, Centro Colecciones y Gestión de Especies, Instituto de Investigación de Recursos Biológicos Alexander von Humboldt, Villa de Leyva, Boyacá, Colombia. ⁵⁴Grupo de estudio en Aves GEA-UT, Universidad del Tolima, Ibagué, Tolima, Colombia.

⁵⁵Department of Biology, University of Florida, Gainesville, FL, USA. ⁵⁶Civil Engineering for Research and Innovation for Sustainability, Instituto Superior Técnico, University of Lisbon, Lisboa, Portugal. ⁵⁷Associação Vita Nativa - Conservação do Ambiente, Olhão, Portugal. ⁵⁸Faculty of Science, Queensland University of Technology, Brisbane, Australia. ⁵⁹Department of Creative Technology, Bournemouth University, Poole, UK. ⁶⁰Laboratory of Insect Evolutionary Biology and Ecology, Faculty of Biology, University of Białystok, Białystok, Poland. ⁶¹UR3456 LECD (Laboratoire éthologie cognition développement), Université Paris Nanterre, 92001, Nanterre, France. ⁶²CWS (Canada Wildlife Services), Environment and Climate Change Canada, Québec Region, Québec, QC, Canada. ⁶³Laboratory of Bioacoustics, Post graduate program of Vertebrate Biology, Pontifical Catholic University of Minas Gerais, Belo Horizonte, Minas Gerais, Brazil. ⁶⁴School of Science, Engineering, and Environment, University of Salford, Manchester, UK. ⁶⁵University of Applied Sciences HTW Dresden, Dresden, Germany. ⁶⁶Experimental Ecologies Group, Sussex Humanities Lab, University of Sussex, Brighton, UK. ⁶⁷Reef Pulse S.A.S., Sainte-Clotilde (La Réunion), France. ⁶⁸UMR9220 ENTROPIE (Écologie marine tropicale dans les Océans Pacifique et Indien), Université de La Réunion, IRD, CNRS, IFREMER, Université de la Nouvelle-Calédonie, 97744, Saint-Denis (La Réunion), France. ⁶⁹Environmental Sustainability Research Centre, University of Derby, Derby, UK. ⁷⁰Département de Géosciences, École Normale Supérieure, 75005, Paris, France. ⁷¹School of Natural and Environmental Sciences, Newcastle University, Newcastle Upon Tyne, UK. ⁷²Guimarães Landscape Laboratory, Guimarães, Portugal. ⁷³Laboratory of Applied Ecology, CITAB (Centre for the Research and Technology of Agro-Environment and Biological Sciences), University of Trás-os-Montes and Alto Douro, Vila Real, Portugal. ⁷⁴Cascais Municipality, Cascais, Portugal. ⁷⁵UMR5175 CEFE (Centre d'écologie fonctionnelle et évolutive), Université de Montpellier, CNRS, IRD, EPHE, 34090, Montpellier, France. ⁷⁶Museum of Nature Hamburg, Leibniz Institute for the Study of Biodiversity Change, Hamburg, Germany. ⁷⁷Environmental Informatics Section, Okinawa Institute of Science and Technology Graduate University, Onna, Okinawa, Japan. ⁷⁸National Institutes of Health, Bethesda, MD, USA. ⁷⁹UMR6554 LETG (Littoral - Environnement - Télédétection - Géomatique), CNRS, Nantes Université, 44312, Nantes, France. ⁸⁰Instituto de Ecologia, A. C. Biblioteca de Sonidos de las Aves de Mexico, Xalapa, Veracruz, Mexico. ⁸¹Department of Integrative Biology and Biodiversity Research, Institute of Wildlife Biology and Game Management, University of Natural Resources and Life Sciences, 1180, Vienna, Austria. ⁸²Wildlife Institute of India, Dehradun, India. ⁸³Cornell University, Ithaca, NY, USA. ⁸⁴UMR7205 ISYEB (Institut de Systématique, Évolution, Biodiversité), Muséum national d'Histoire naturelle, CNRS, Sorbonne Université, EPHE, Université des Antilles, 75005, Paris, France. ⁸⁵Departamento de Biología-IVAGRO, Universidad de Cádiz, Campus Ro San Pedro, Puerto Real, Spain. ⁸⁶MED (Mediterranean Institute for Agriculture, Environment and Development), University of Évora, Évora, Portugal. ⁸⁷Universidade Federal de Pernambuco, Recife, Brasil. ⁸⁸Department of Life and Environmental Sciences, Bournemouth University, Poole, UK. ⁸⁹Department of Life Sciences, University of the West Indies, St Augustine, Trinidad and Tobago. ⁹⁰Department of Biology, Furman University, Greenville, SC, USA. ⁹¹School of Life Sciences, Jawaharlal Nehru University, New Delhi, India. ⁹²UMR5193 LISST-CIEU (Laboratoire Interdisciplinaire Solidarités, Sociétés, Territoires), CNRS, Université Toulouse 2 - Jean Jaurès, 31058, Toulouse, France. ⁹³School of Art and Creative Industries, Teesside University, Middlesbrough, UK. ⁹⁴UMR7263 IMBE (Institut Méditerranéen de Biodiversité et d'Écologie marine et continentale), Aix-Marseille Université, Avignon Université, CNRS, IRD, Nouméa, New Caledonia. ⁹⁵Department of Ecology, Faculty of Environmental Sciences, Czech University of Life Sciences Prague, Prague, Czech Republic. ⁹⁶Mohitban Society, Tehran, Iran. ⁹⁷Ecology, Evolution, and Environmental Biology Department, Project Dhvani, Columbia University, Wildlife Institute of India, New York, NY, USA. ⁹⁸Cetalingua Project, Tampa, FL, USA. ⁹⁹Office National des Forêts, Maisons-Alfort, France. ¹⁰⁰University of Alberta, Edmonton, AB, Canada. ¹⁰¹Sylvio Romério Briglia-Ferreira's, i.e. the Instituto Chico Mendes de Conservação da Biodiversidade (ICMBio), Boa Vista, RoraimaBrasil. ¹⁰²Locus Sonus Vitae, École Supérieure d'Art d'Aix-en-Provence, Aix-en-Provence, France. ¹⁰³Department of Biology, University of Washington, Seattle, WA, USA. ¹⁰⁴IES Gabriel Garcia Marquez, Madrid, Spain. ¹⁰⁵University of Massachusetts Boston, Boston, MA, USA. ¹⁰⁶Oceans Forward, Boston, MA, USA. ¹⁰⁷Faculty of Science, University of South Bohemia, Ceske Budejovice, Czechia. ¹⁰⁸BioPhonia, Oletta, France. ¹⁰⁹The Nature Conservancy, Bogotá, Colombia. ¹¹⁰Laboratório de Herpetologia e Comportamento Animal, Departamento de Ecologia, Instituto de Ciências Biológicas, Universidade Federal de Goiás, Campus Samambaia, Goiânia, Goiás, Brazil. ¹¹¹School of Biosciences, University of Melbourne, Melbourne, Australia. ¹¹²Centre for Functional Ecology - Science for People & the Planet, Department of Life Sciences, University of Coimbra, Coimbra, Portugal. ¹¹³MHNC-UP (Museu de História Natural e da Ciência da Universidade do Porto), University of Porto, Porto, Portugal. ¹¹⁴Biodiversity Research Institute (CSIC - Oviedo University - Principality of Asturias), Oviedo University, Mieres, Spain. ¹¹⁵BiBio Research Group, Natural Sciences Museum of Granollers, Granollers, Spain. ¹¹⁶Department of Ecology and Wildlife Management, University of Kisangani, Kisangani, Democratic Republic of the Congo. ¹¹⁷CEF (Centro de Estudos Florestais), Instituto Superior de Agronomia, Universidade de Lisboa, Tapada da Ajuda, Lisbon, Portugal. ¹¹⁸Colección Ornitológica, Universidad del Quindo, Programa de Biología, Armenia, Quindo, Colombia. ¹¹⁹Facultad de Estudios Superiores Iztacala, Universidad Nacional Autónoma de México, Tlalnequintla de Baz, México. ¹²⁰Sociedade Portuguesa de Vida Selvagem, BragaPortugal. ¹²¹CHANGE (Global Change and Sustainability Institute), Institute for Advanced Studies and Research, Universidade de Évora, Évora, Portugal. ¹²²Museu de Ciências Naturals de Barcelona, Barcelona, Spain. ¹²³ENES Bioacoustics Research Lab, UMR5292 CRNL (Centre de Recherche en Neurosciences de Lyon), Université de Saint-Étienne, CNRS, Inserm, 42023Saint-Étienne, France. ¹²⁴Department of Ecology, The Swedish University of Agricultural Sciences, Uppsala, Sweden. ¹²⁵Groupe Mammalogique Breton, Sizon, France. ¹²⁶University of Freiburg, Faculty of Biology, Geobotany, Freiburg, Germany. ¹²⁷Colección Oológica Cornelis Johannes Marinkelle, Instituto de Investigación de Recursos Biológicos Alexander von Humboldt, Villa de Leyva, Boyacá, Colombia. ¹²⁸Laboratório de Ecologia & Evolução de Aves. Departamento de Zoologia, Universidade Federal de Pernambuco, Recife, Pernambuco, Brazil. ¹²⁹Associação ALDEIA, Vimioso, Portugal. ¹³⁰Ecofirst SRL, Awenne, Belgium. ¹³¹Natagora asbl, Namur, Belgium. ¹³²Department of Zoology, School of Natural Sciences, Trinity College Dublin, Dublin, Ireland. ¹³³School of Biology and Environmental Science, University College Dublin, Dublin, Ireland. ¹³⁴Department of Biological Sciences, Southeastern Louisiana University, Hammond, LA, USA. ¹³⁵Bat Conservation International, Austin, TX, USA. ¹³⁶Department of Migration, Max

Planck Institute of Animal Behavior, Radolfzell, Germany. ¹³⁷Department of Ecology and Evolutionary Biology, Princeton University, Princeton, NJ, USA. ¹³⁸Museum de Toulouse, Toulouse, France. ¹³⁹Azim Premji University, Bengaluru, Karnataka, India. ¹⁴⁰Department of Forestry and Natural Resources, Purdue University, West Lafayette, IN, USA. ¹⁴¹CESAM (Centre for Environmental and Marine Studies), Department of Biology, University of Aveiro, Aveiro, Portugal. ¹⁴²Associação BioLiving, Albergaria-a-Velha, Portugal. ¹⁴³Cooper Union For The Advancement of Science and Art, New York, NY, USA. ¹⁴⁴Atlantic Technological University, Galway, Ireland. ¹⁴⁵Invisible Flock, Yorkshire Sculpture Park, West Bretton, UK. ¹⁴⁶Wild Basin Creative Research Center, St. Edward's University, Austin, TX, USA. ¹⁴⁷RIVE (Centre de recherche sur les interactions bassins versants écosystèmes aquatiques), Département des sciences de l'environnement, Université du Québec à Trois-Rivières, Trois-Rivières, QC, Canada. ¹⁴⁸SEM (Institut des Sciences de l'Évolution de Montpellier), Université de Montpellier, CNRS, IRD, EPHE, CIRAD, INRAP, 34095, Montpellier, France. ¹⁴⁹Institut Universitaire de France, Paris, France. ¹⁵⁰ANP/WWF (Associação Natureza Portugal in association with World Wildlife Fund), Lisboa, Portugal. ¹⁵¹SonarAndino Project, Jujuy, Argentina. ¹⁵²The atelier Desenho Aberto, project Extremophilium, Lisboa, Portugal. ¹⁵³School of Mathematics and Statistics, University of Melbourne, Melbourne, Australia. ¹⁵⁴Schroeder & Associés, Kockelscheuer, Luxembourg. ¹⁵⁵Plecotus Pole, Natagora, Bruxelles, Belgium. ¹⁵⁶CEMAVE, Centro Nacional de Pesquisa e Conservação de Aves Silvestres, Instituto Chico Mendes de Conservação da Biodiversidade, Florianópolis, Brazil. ¹⁵⁷Integrative Community Ecology Unit, Okinawa Institute of Science and Technology Graduate University, Onna, Okinawa, Japan. ¹⁵⁸Azimet230, Les Ulis, France. ¹⁵⁹Alcathoé, Orsay, France. ¹⁶⁰Bharati Vidyapeeth Deemed University, Institute of Environment Education and Research, Pune, India. ¹⁶¹Capreolus e.U., Orth an der Donau, Austria. ¹⁶²Laboratorio de Ecologia y Conservación de Vertebrados Terrestres, Instituto de Ecología, Universidad Nacional Autónoma de México, Tlalnepantla de Baz, México. ¹⁶³UNEP-WCMC (UN Environment Programme World Conservation Monitoring Centre), Cambridge, UK. ¹⁶⁴UMR7533 LADYSS (Laboratoire dynamiques sociales et recomposition des espaces), CNRS, Université Paris Nanterre, 92001, Nanterre, France. ¹⁶⁵K. Lisa Yang Center for Conservation Bioacoustics, Cornell Lab of Ornithology, Cornell University, Ithaca, NY, USA. ¹⁶⁶Wildlife Acoustics Inc., Maynard, MA, USA. ¹⁶⁷Department of Biology, Western Washington University, Bellingham, WA, USA. ¹⁶⁸Laboratorio de Ornitología y Bioacústica, Unidad de Ecología y Sistemática (UNESIS), Departamento de Biología, Facultad de Ciencias, Pontificia Universidad Javeriana, Bogotá, Colombia. ¹⁶⁹Laboratorio de Ecología Funcional, Unidad de Ecología y Sistemática (UNESIS), Departamento de Biología, Facultad de Ciencias, Pontificia Universidad Javeriana, Bogotá, Colombia. ¹⁷⁰University of East London, London, UK. ¹⁷¹Department of Cognitive, Linguistic, and Psychological Sciences, Brown University, Providence, RI, USA. ¹⁷²Department of Biological Sciences, University of Toronto Scarborough, Scarborough, ON, Canada. ¹⁷³École polytechnique fédérale de Zürich, Zürich, Switzerland. ¹⁷⁴Pôle Aves Natagora, Liéges, Belgium. ¹⁷⁵Grup de Natura Sterna, Vall d'Aro, Girona, Catalunya, Spain. ¹⁷⁶Universidad de Alcalá, GloCEE - Global Change Ecology and Evolution Research Group, Departamento de Ciencias de la Vida, Madrid, Spain. ¹⁷⁷Groupe Chiroptères Languedoc-Roussillon, Montpellier, France. ¹⁷⁸Réseau Natura 2000 des Hautes-Alpes, Communauté de communes de Serre-Ponçon, Embrun, France. ¹⁷⁹Fauna & Flora International - Vietnam Programme, Tay Ho, Hanoi, Vietnam. ¹⁸⁰Department of Life Sciences, University of the West Indies, Mona, Jamaica.