



Bridges-2 Webinar

Scaling Up Ecological Monitoring with AI: How Supercomputing Is Unlocking the Value of Autonomous Acoustic Sensing

Sam Lapp
University of Pittsburgh



Pittsburgh Supercomputing Center

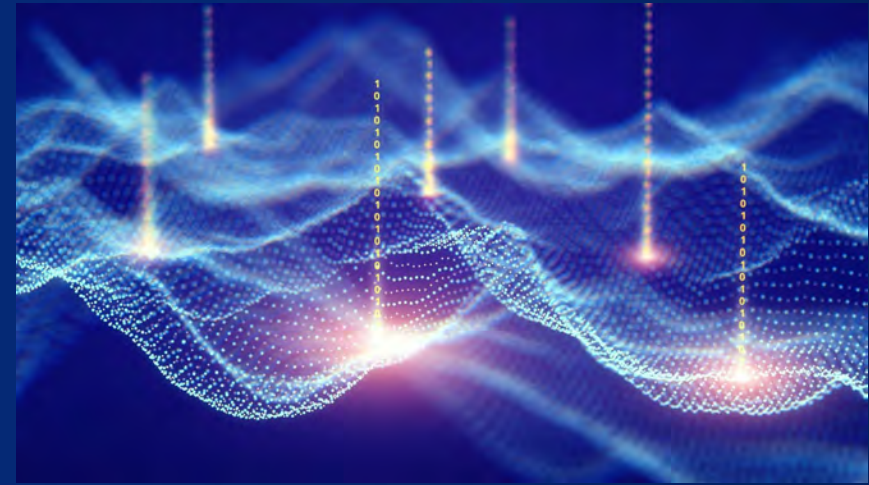
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Welcome!

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Bridges-2

Bridges-2 Overview

Bridges-2 is a unique computational platform for scientific and engineering research for national grid. It combines advanced artificial intelligence (AI), high performance computing (HPC), big data, and big data analytics.

Bridges-2 is available at no cost to researchers across the United States and their international counterparts, as well as to visitors.

The Bridges-2 project is led by principal investigators Bruce J. Berman, Florin A. Buzdugan, Sergio C. Beaulieu, and Robin W. Bustin, funded by NSF's highly competitive award.

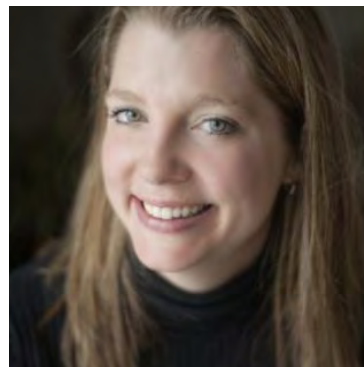
Bridges-2 is supported by National Science Foundation award 1008017. The Bridges-2 system was delivered by Hewlett Packard Enterprise.

 **Hewlett Packard Enterprise** is delivering *Bridges-2*

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Bridges-2 Webinars

- A forum for the Bridges-2 community to learn and share ideas and achievements: [Bridges-2 Webinar series | PSC](#)
- Topics and speakers of interest to work that is being done, or that may be done in future.
- Please suggest future speakers (including from your own team) and/or topics (including your own)!

Just email: sergiu@psc.edu

Introducing today's presenter: Sam Lapp

Sam Lapp is a machine learning and sound geek with a passion for biodiversity conservation. His current research focuses on developing and applying AI methods in bioacoustics, the strategy of studying ecology through the sounds produced by living things. His projects include developing the open-source Python package *OpenSoundscape* and applying machine learning methods to the conservation of birds, frogs, and insects. Sam studied engineering, music, and sound design at Penn State and is currently working on his Ph.D. in the Kitzes Lab at the University of Pittsburgh.

Q&A Logistics

- **We abide by <https://support.access-ci.org/code-of-conduct>**
- All of us except Sam will be muted during his presentation.
- Please type your questions into the Zoom chat.
- We may be able to address some questions in the chat while Sam is presenting.
- When Sam finishes his presentation, he will answer questions live during the final ~10 minutes of this webinar.

peeps, pops, and teraflops

discovering hidden patterns in nature by applying supercomputing to audio recordings



Sam Lapp – Kitzes Lab @ University of Pittsburgh

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Pittsburgh Supercomputing Center Bridges2 Seminar Series, August 2024





Sound carries
rich information

Song is a fundamental aspect of bird behavior



Acoustic sensors: systematically sample biological sounds in the wild

- Large spatiotemporal scales
- Remote areas
- Hard-to-survey species
- Acoustic “specimens”







Acoustic
Monitoring

Machine
Learning

How can supercomputing crack the code?

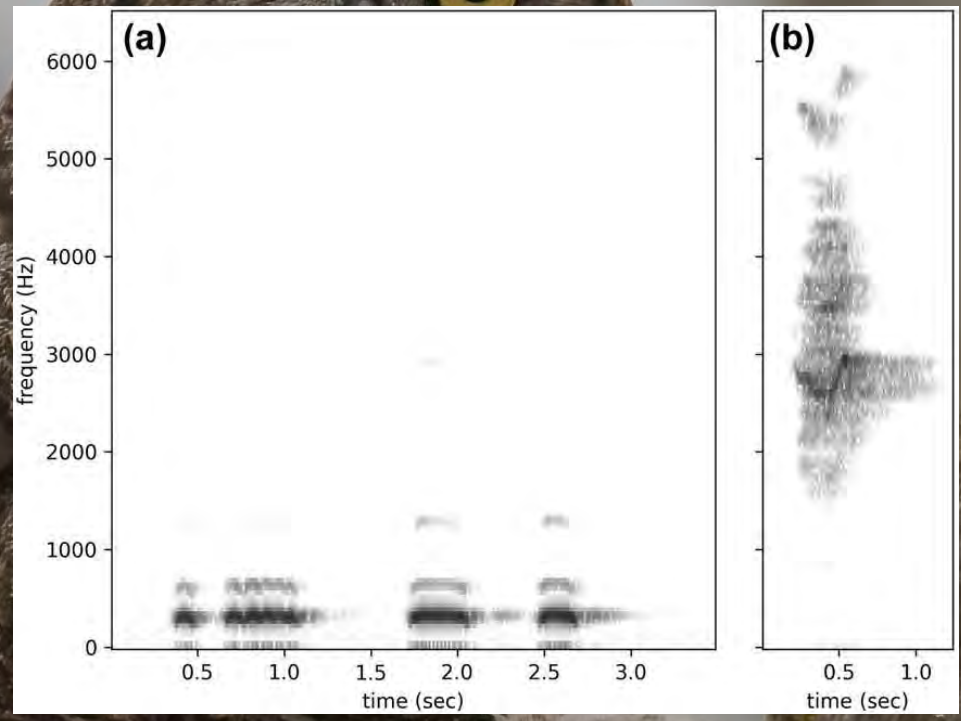
Three stories for today:

1. intelligent sound recognition
powered by GPU Nodes
2. probing the unknown with one-shot acoustic querying
enabled by Extreme Memory nodes
3. high-resolution spatial tracking
with super-parallelized CPU

A large owl with yellow eyes and brown feathers is perched on a tree branch. Below it, a fluffy, grey chick with yellow eyes is also perched on the same branch. The background is a soft, out-of-focus natural setting with green moss on the tree trunk.

Part 1: It takes a landscape to raise an owl

GPU nodes fuel automated recognition of rare
wildlife sounds

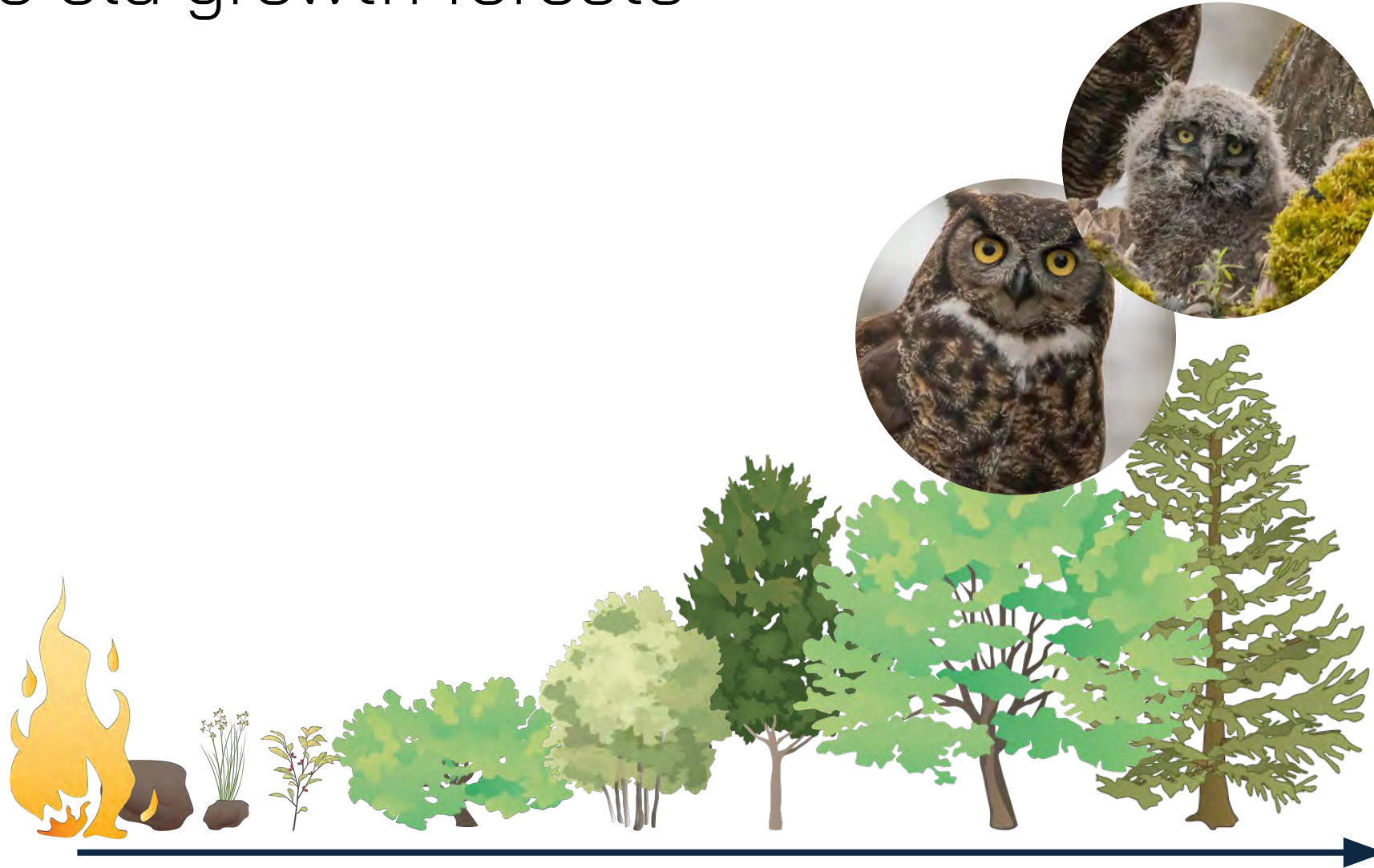


Adult hoots



Juvenile Begging

Through ecological succession, habitats mature from fields to old growth forests



Ecological succession over time

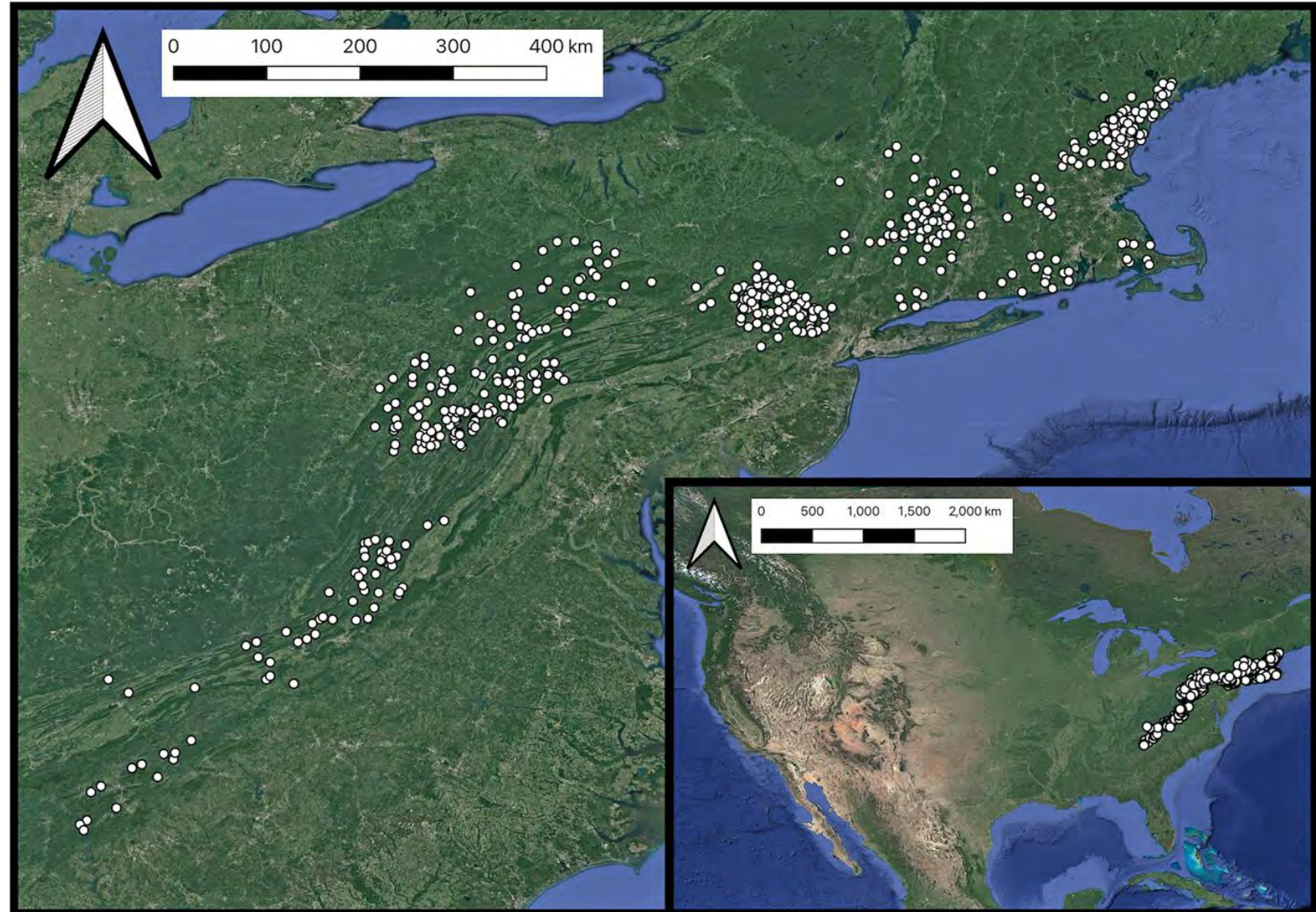
Season-long acoustic monitoring across the Northeast

- recorded over 100,000 hours of audio
- covered diverse landscapes
- produced 20 Terabytes of data

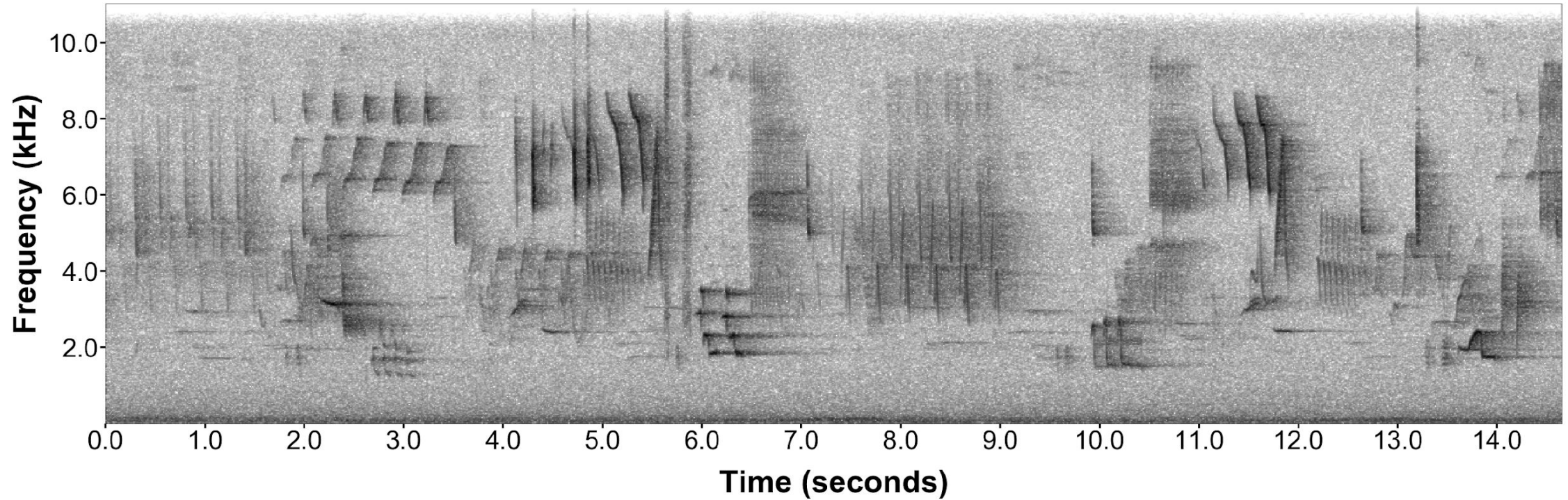
If Lauren had tried to listen to it all, she'd have spent

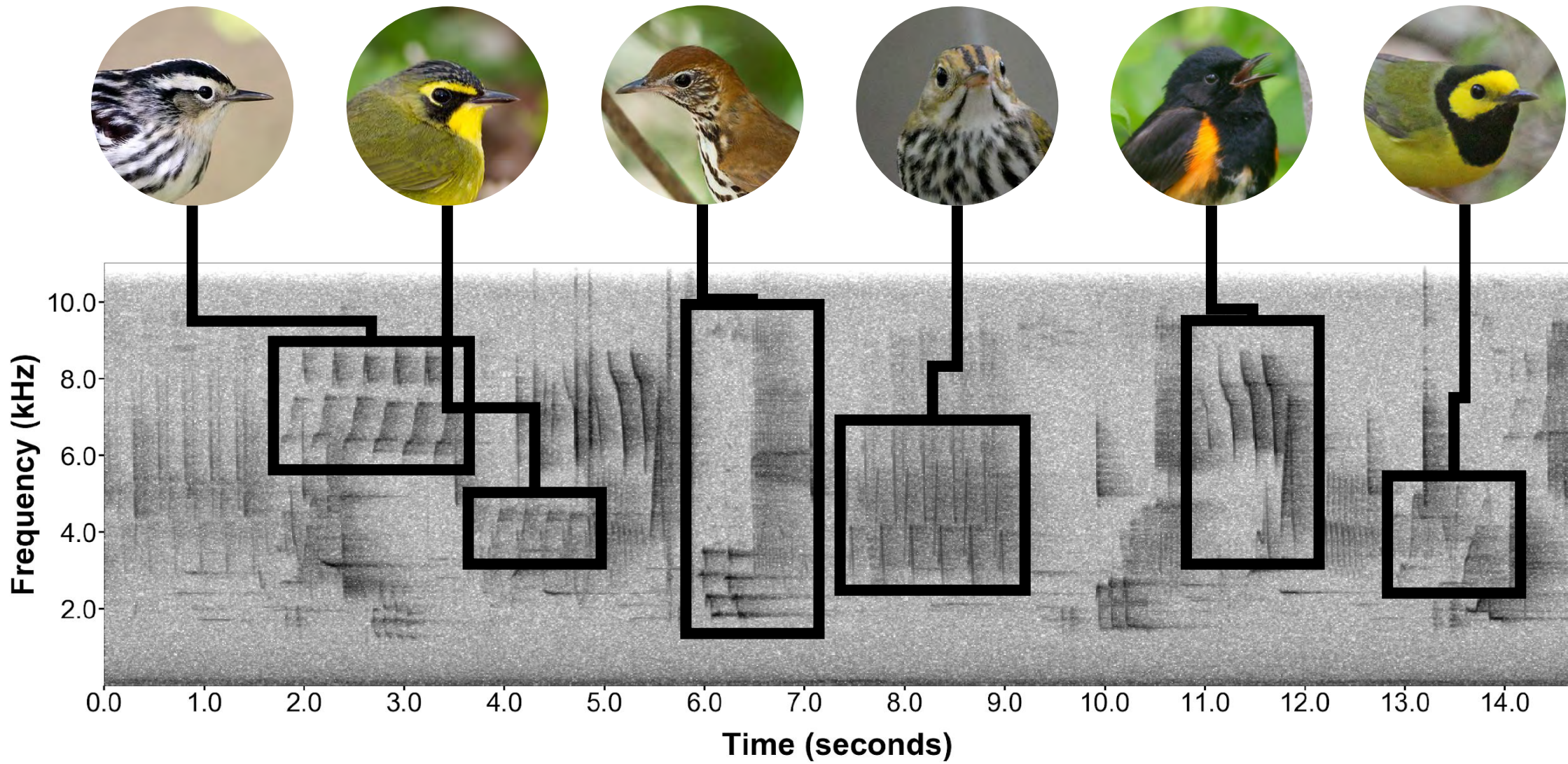
133 years

listening for 40 hrs/week

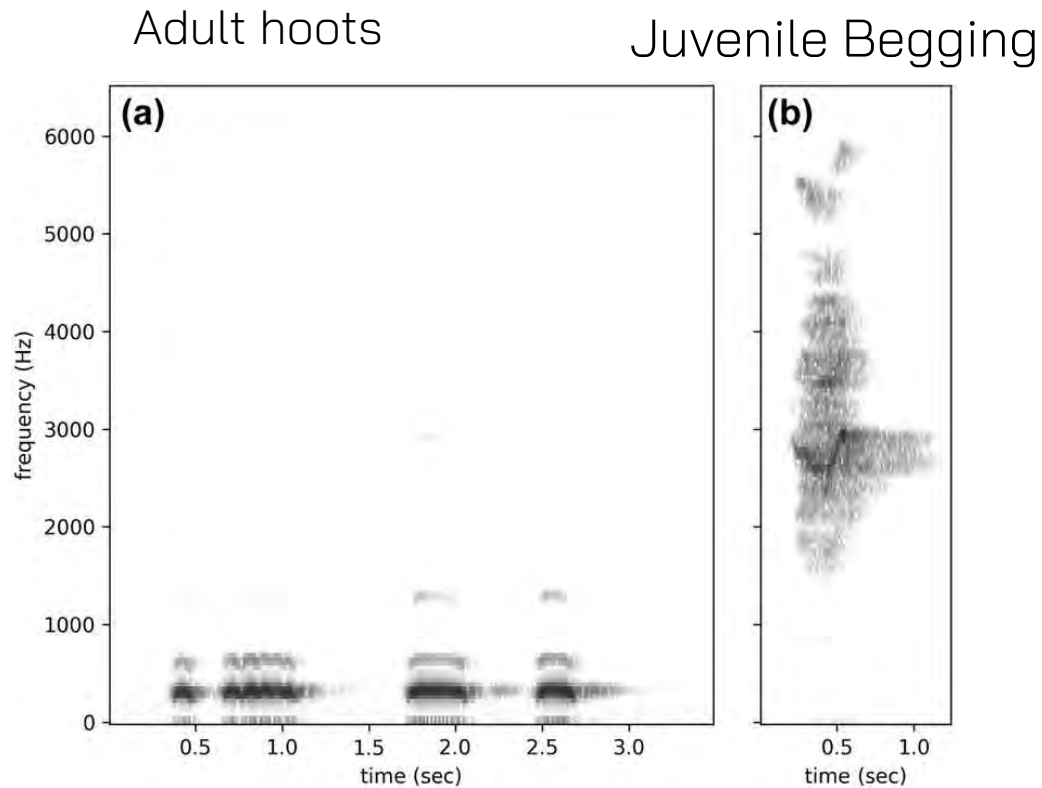


a spectrogram represents sound as an image





GPU computing accelerates deep learning sound recognition



compared to CPU, GPU
provides a **40x**
speed-up for training and
applying deep learning
models

for instance, training a
model could take 1 hour
instead of 40

Supercomputing-powered analysis of the massive dataset resulted in the **detection of rare juvenile begging events**

allowing Lauren to study **habitat needs** of juvenile owls



Findings: juvenile owls require diverse and distinct habitat types, including open unforested patches



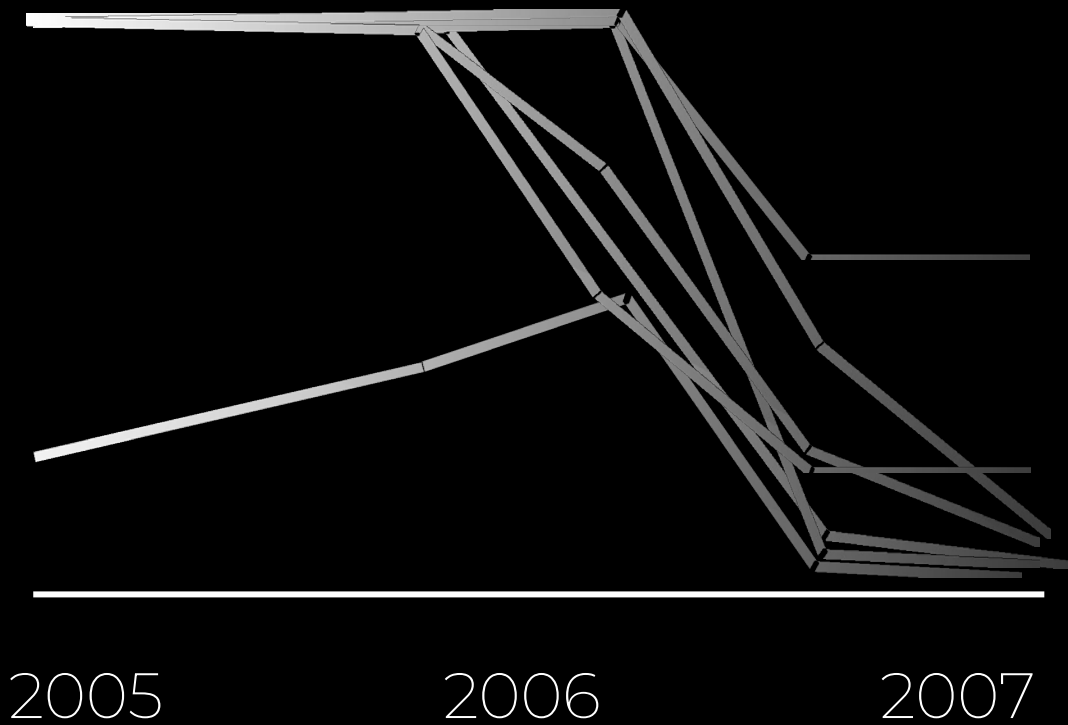
A photograph of a large owl and its fluffy chick perched on a mossy tree branch. The owl is the central focus, looking directly at the camera with its large yellow eyes. The chick is smaller and fluffier, also looking towards the camera. The background is a soft, out-of-focus landscape. The text "It takes a landscape to raise an owl" is overlaid in white, centered on the image.

It takes a landscape to
raise an owl

Part 2: searching in the dark

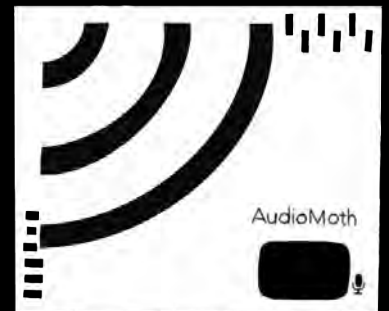
rediscovering lost frogs in Panama with Extreme Memory embedding querying



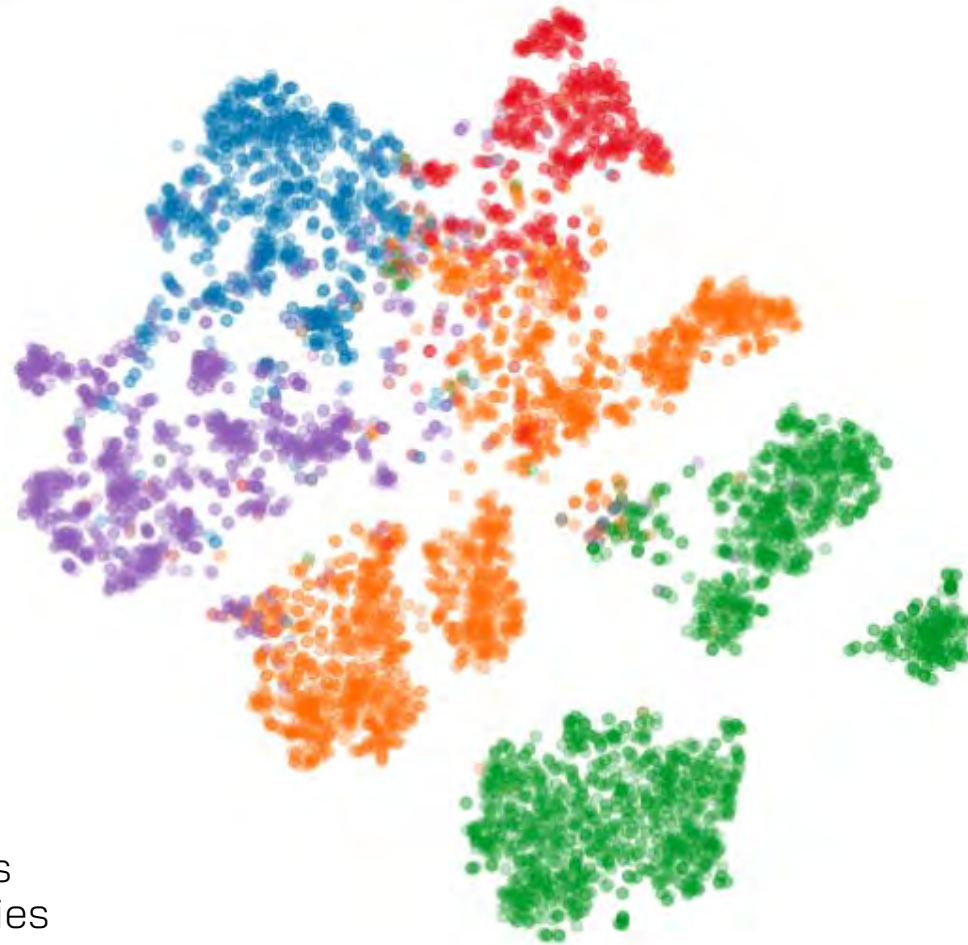


Chytridiomycosis and Amphibian Population Declines Continue to Spread Eastward in Panama
Woodhams et al, 2008. EcoHealth





Feature embeddings of a model trained to recognize bird vocalizations cluster by species

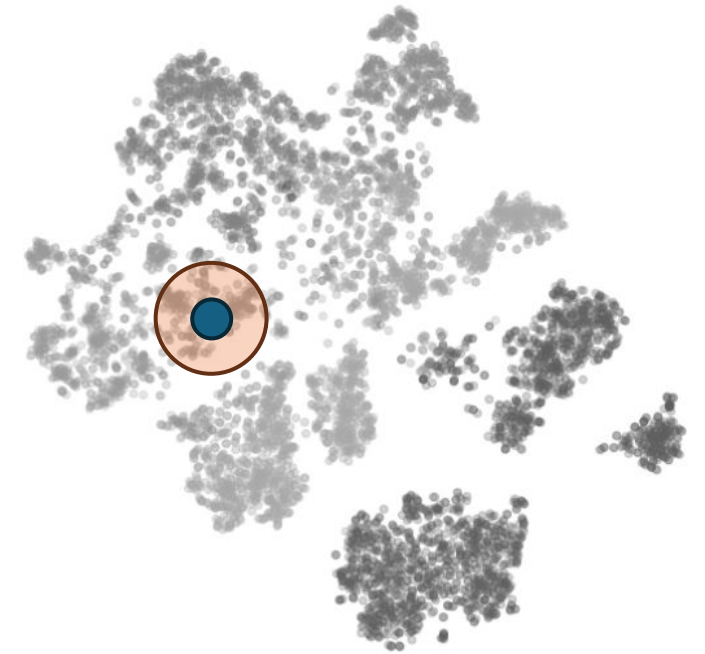


tSNE of feature embedding vectors
with samples colored by bird species

Ghani et al., 2023. Feature embeddings from large-scale acoustic bird classifiers enable few-shot transfer learning.

Using embedding model trained on birds to search unlabeled data for a specific frog sound

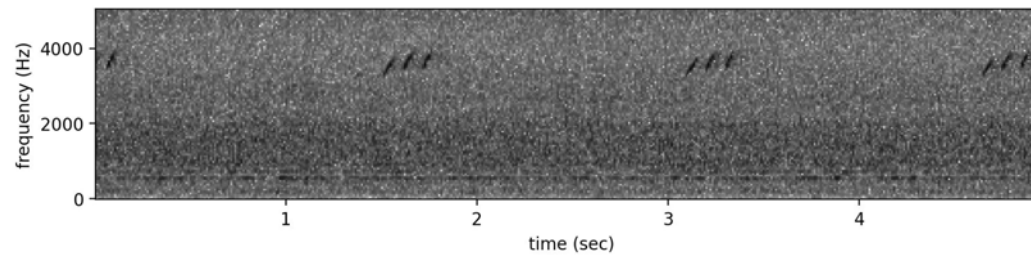
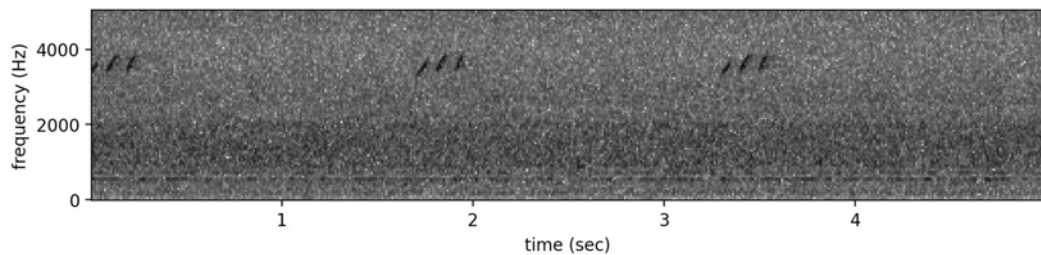
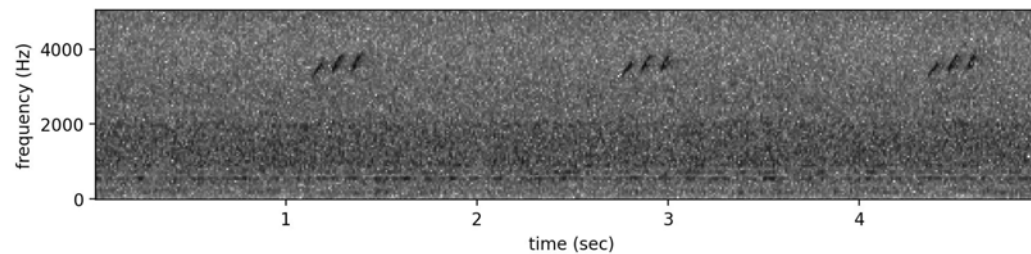
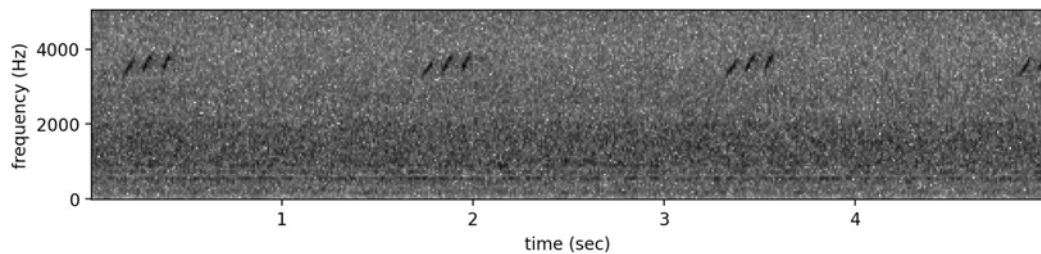
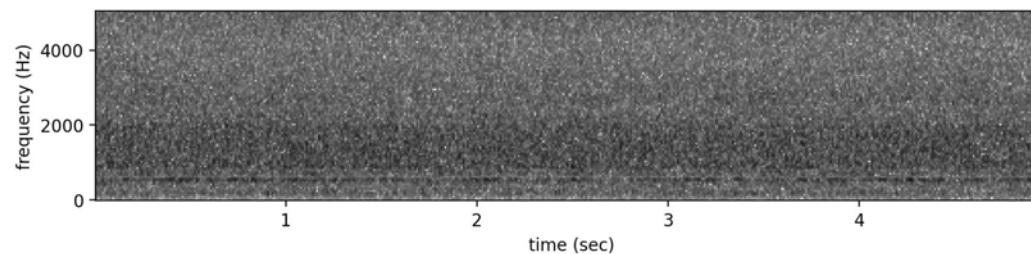
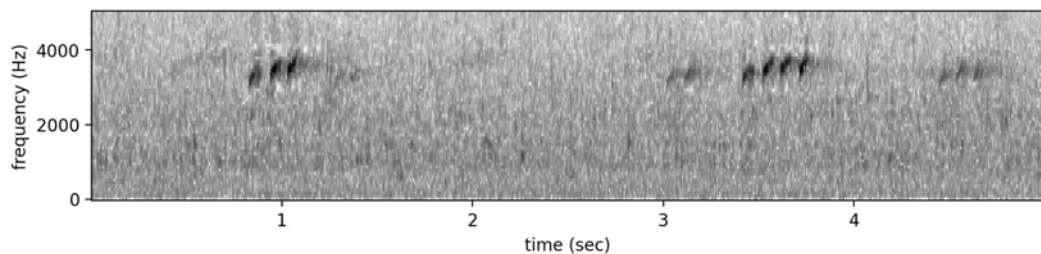
1. Generate embedding vectors for the new dataset (10 million samples)
2. Generate embedding vector for a sample of the sound you want to detect in the new dataset
3. Inspect the samples in the raw data with the most similar embedding vectors



querying an entire dataset requires extreme memory nodes

Using embedding model trained on birds to search Panama data for a frog's sound

Query: *C. panamensis* call



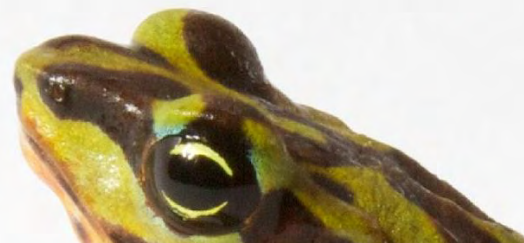
5 samples from field data with most similar embedding vectors

Atelopus varius

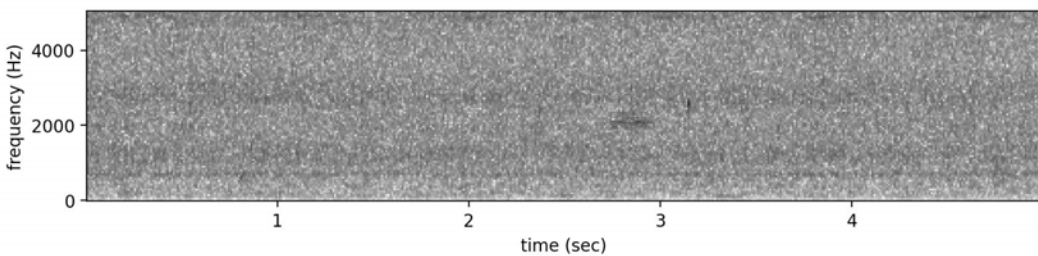
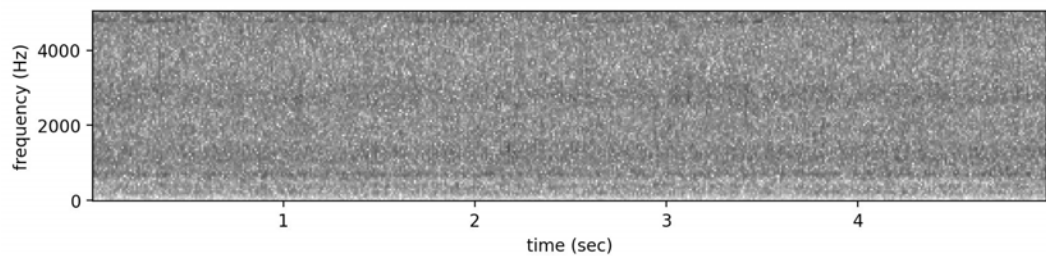
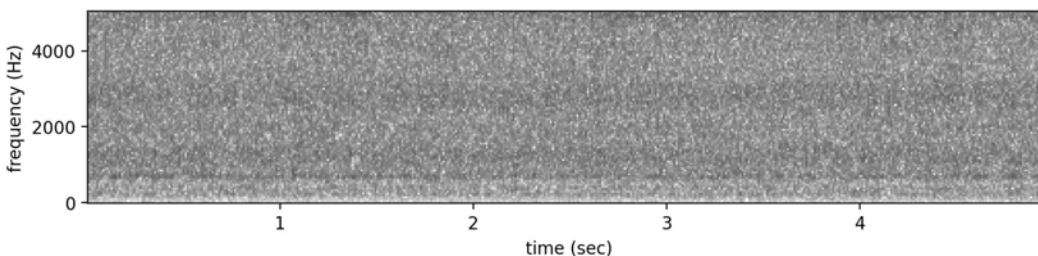
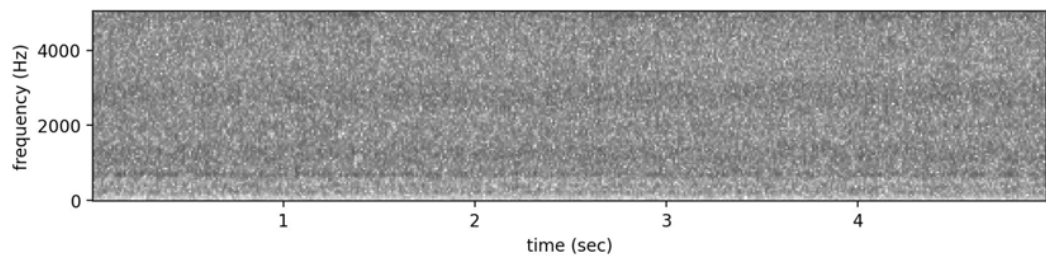
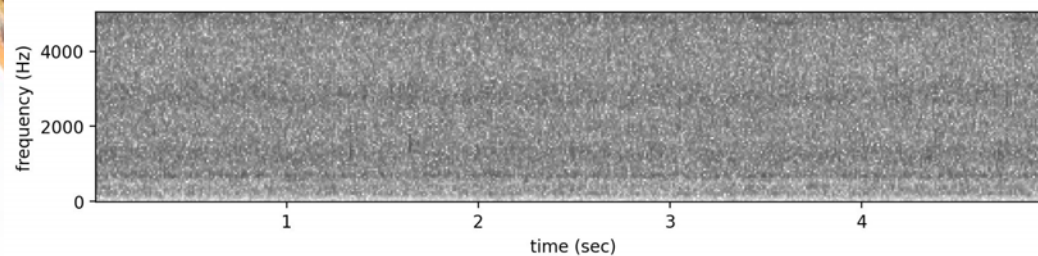
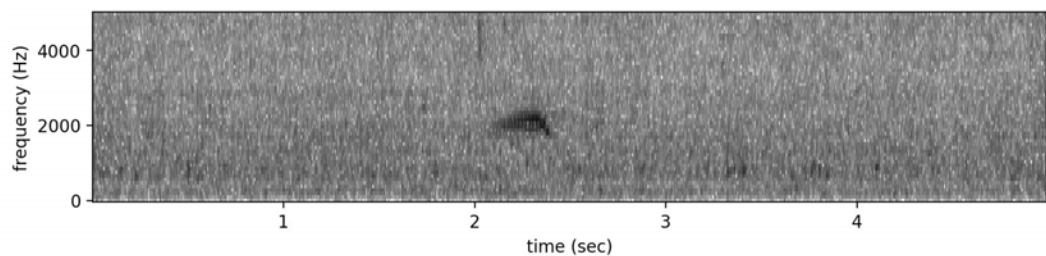
critically endangered



Atelopus varius



Query: *A. varius* call



5 samples from field data with most similar embedding vectors

searching in the dark

rediscovering lost frogs in Panama with Extreme Memory embedding querying





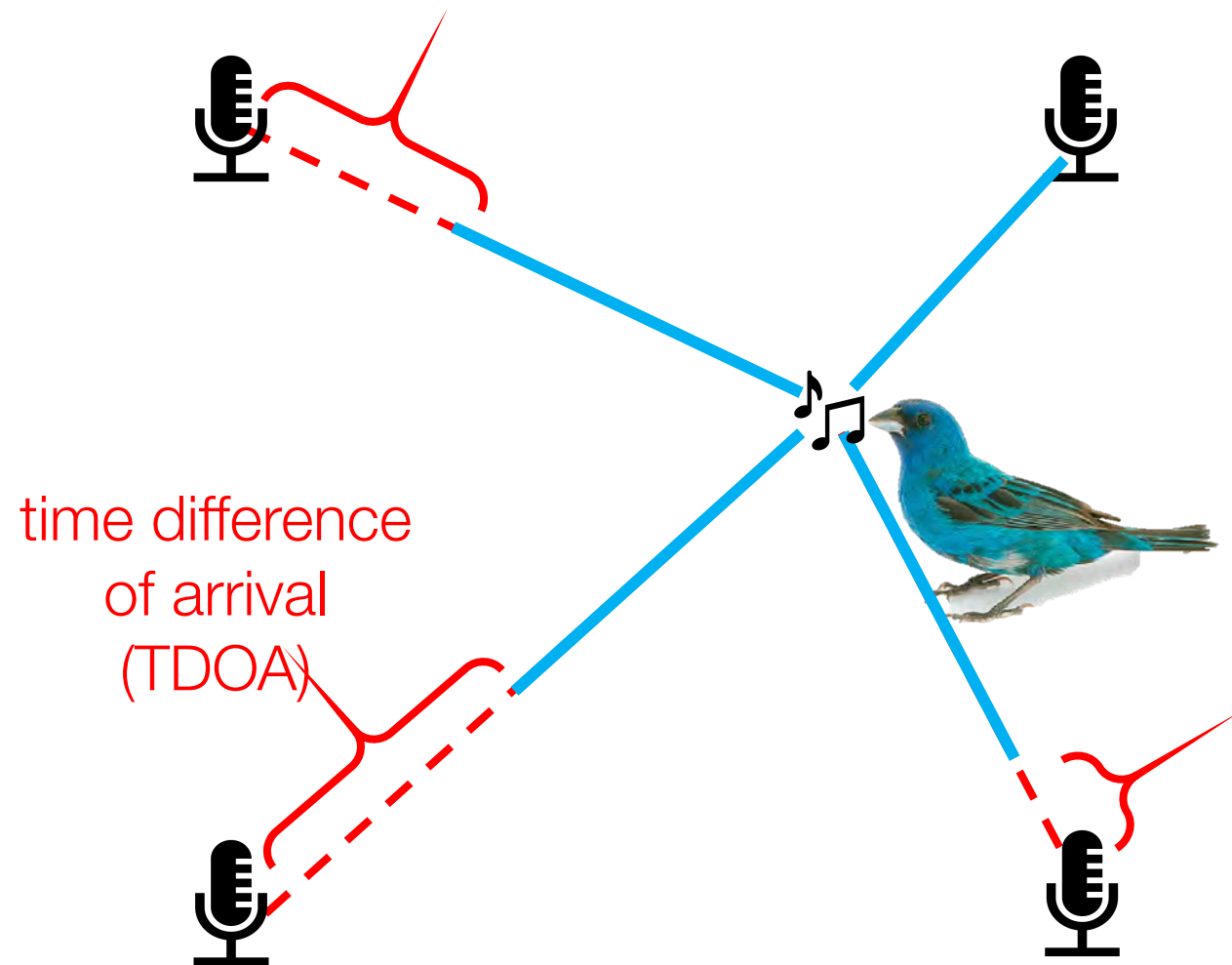
Part 3: A day in the life of a warbler



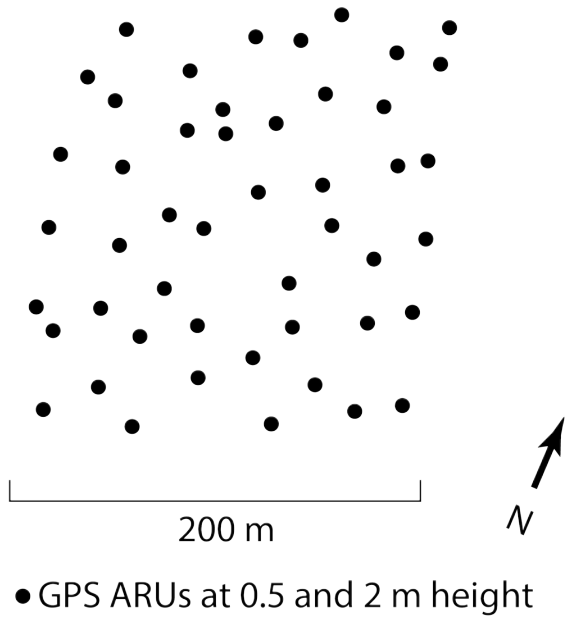
Automated Acoustic Localization



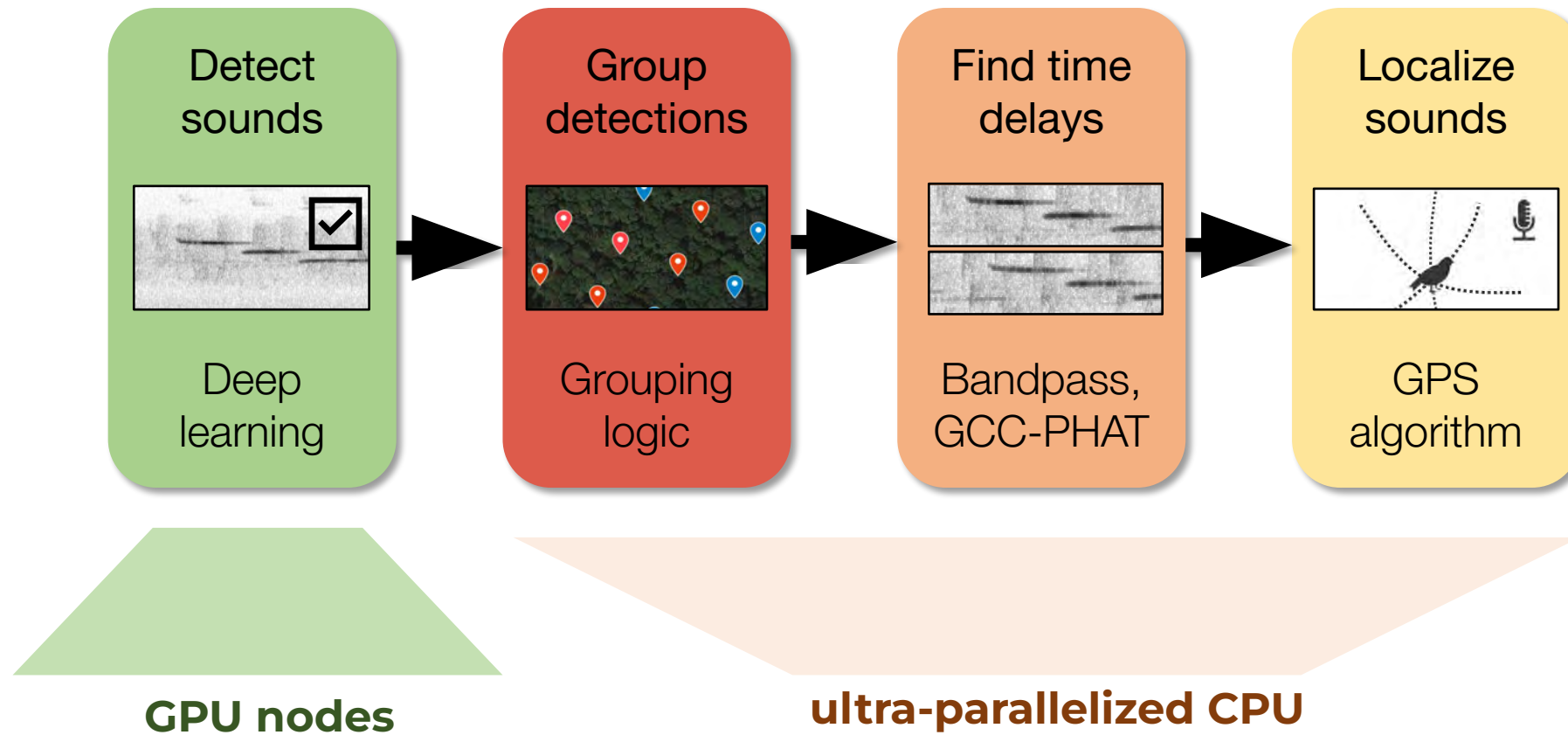
Acoustic localization: spatially triangulating sound positions



Array of time-synchronized automated recorders (ARUs)

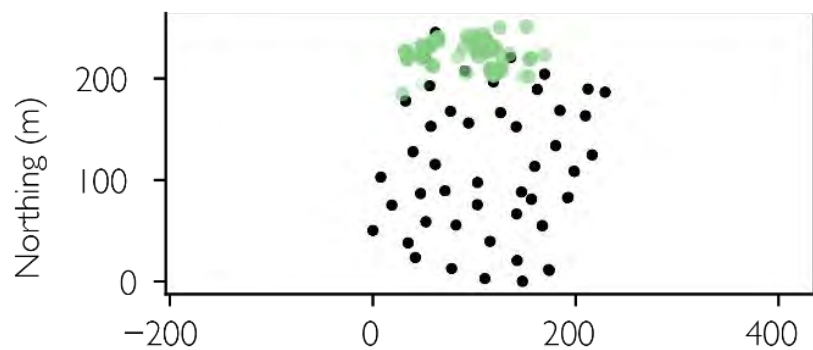


Automated detection and localization of bird song

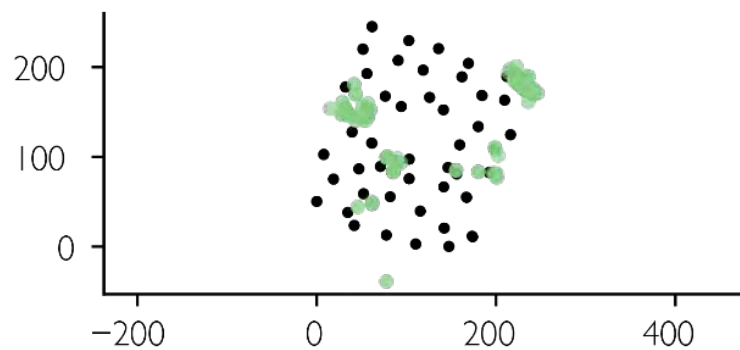


A day of singing activity for six species

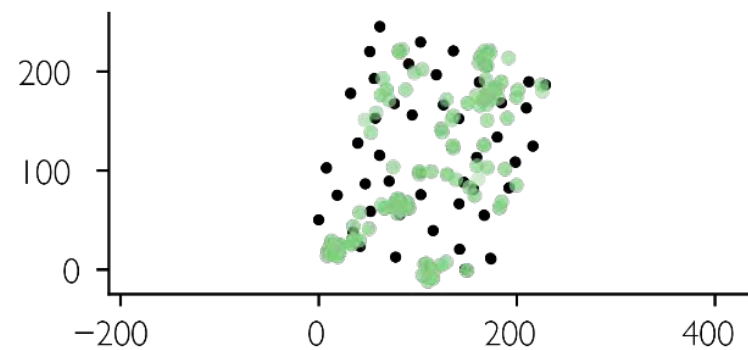
American Redstart



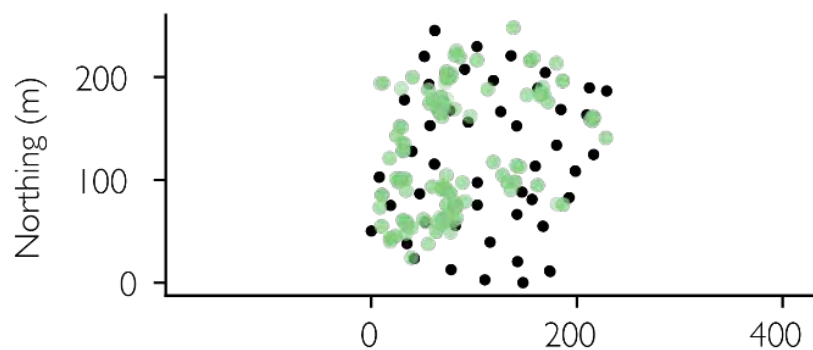
Common Yellowthroat



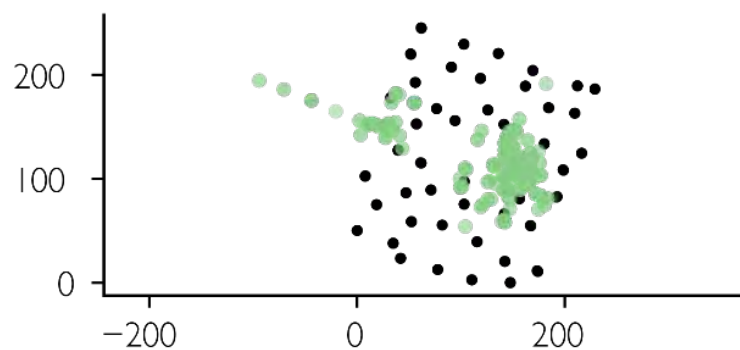
Chestnut-sided Warbler



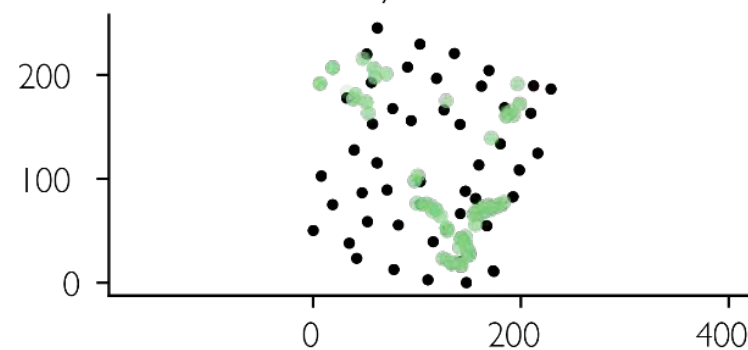
Tennessee Warbler



Hooded Warbler



Red-eyed Vireo

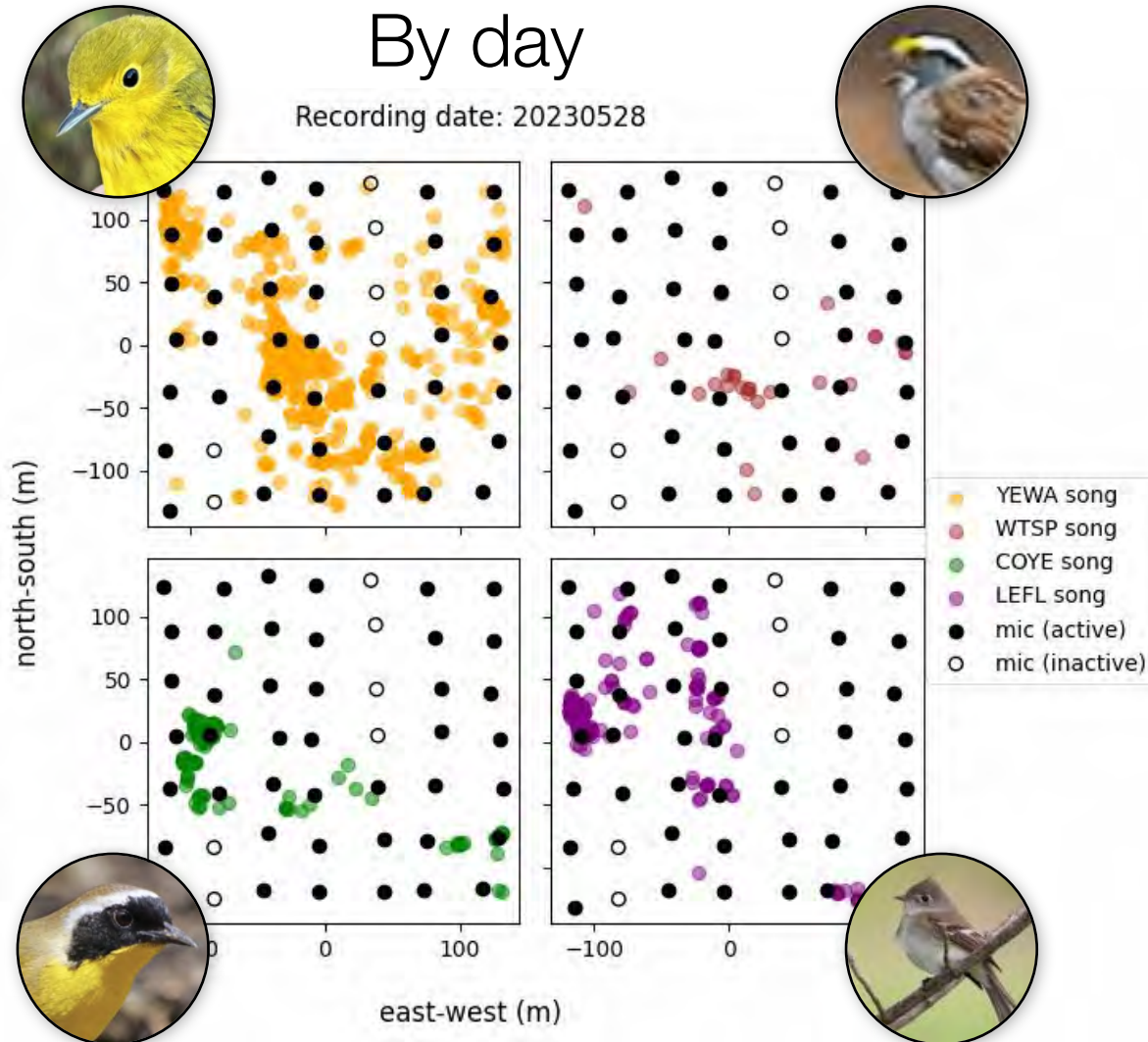


- Recorder
- Localized song

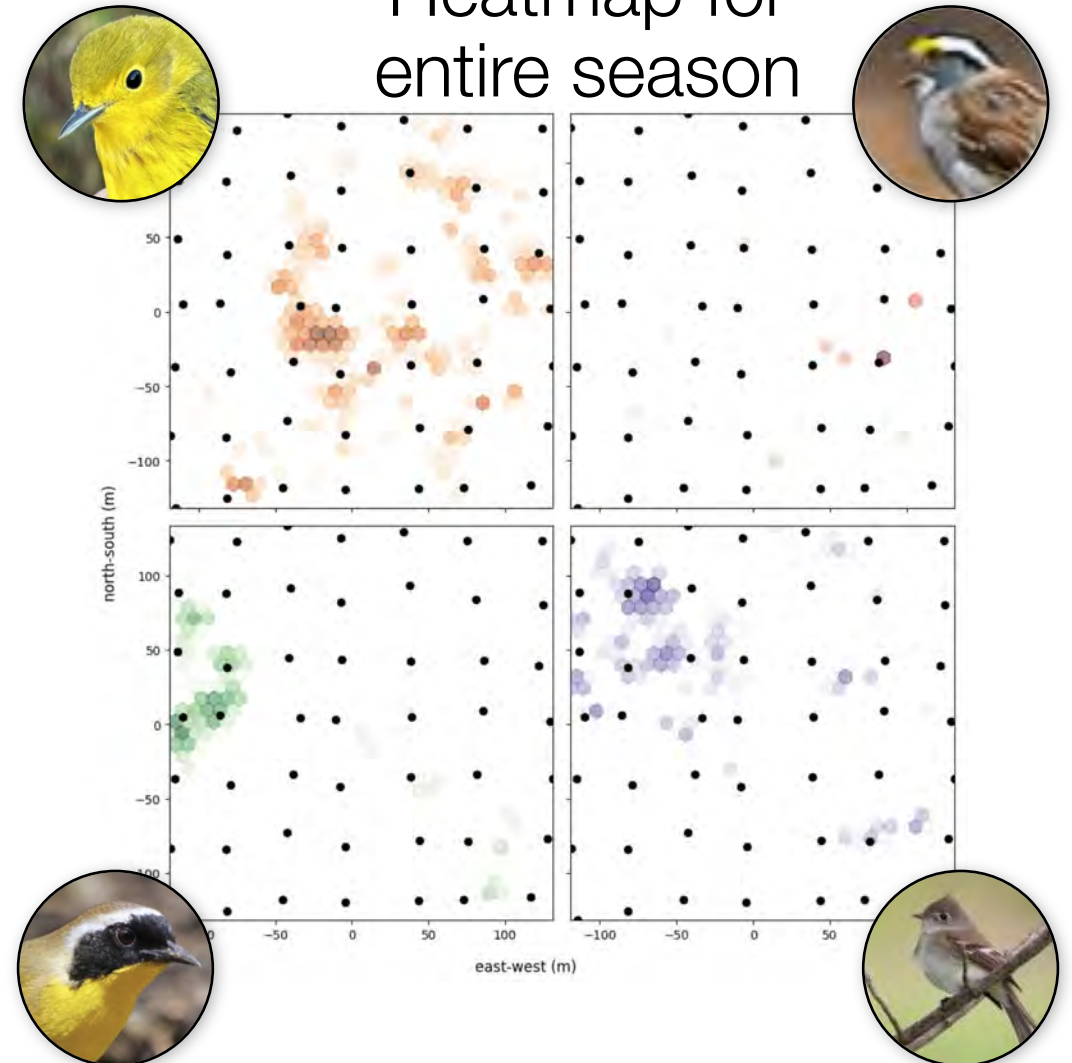
A season's worth of localization

By day

Recording date: 20230528

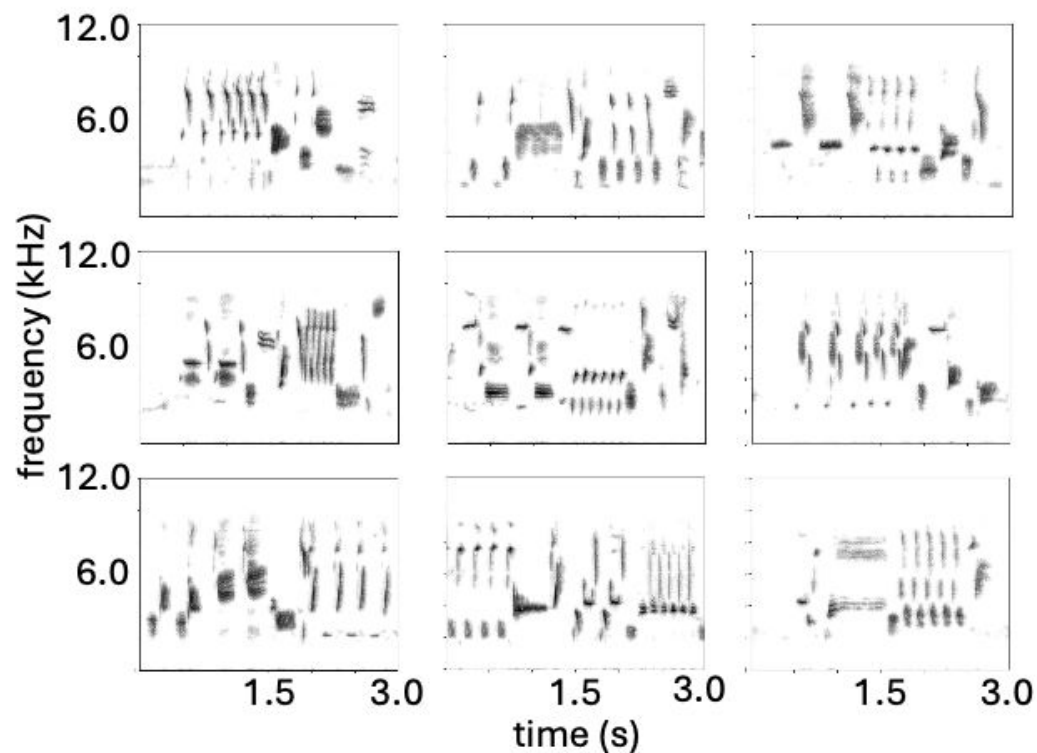


Heatmap for entire season

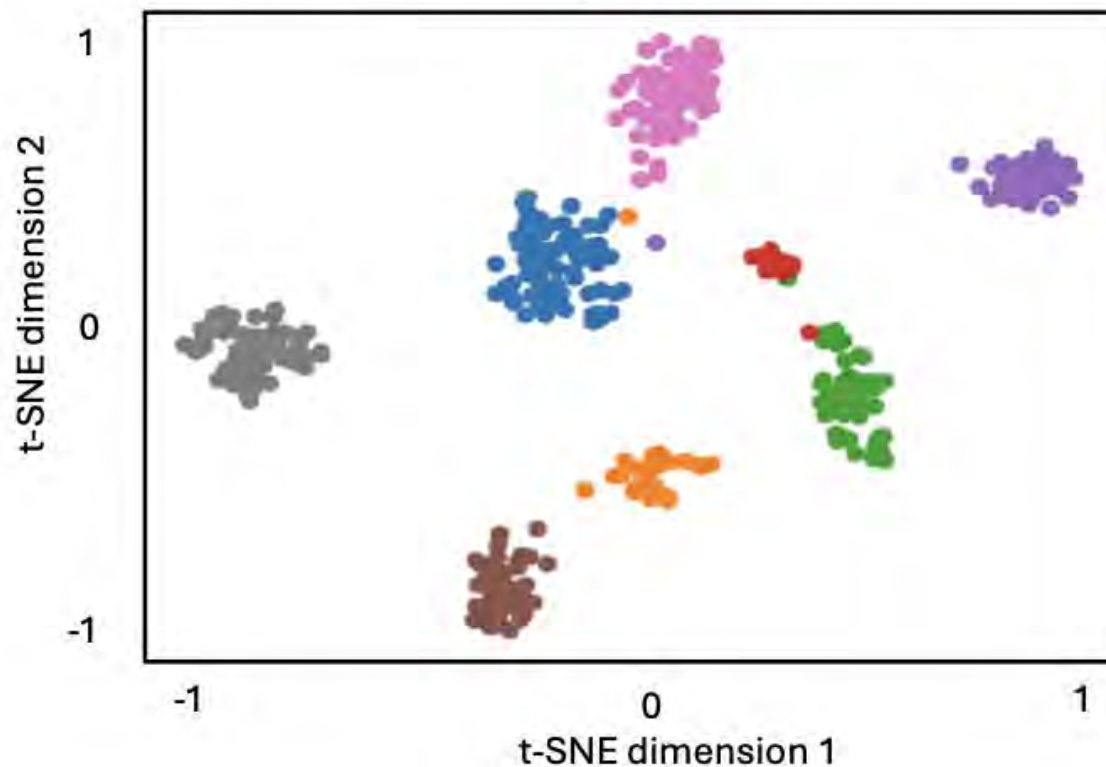


recognizing individual Song Sparrows by song

Example spectrograms of sampled song sparrow songs, each representing a different song variant.



Automatically separated versus hand separated song sparrow song variants. Points are assigned by automated separation and colors indicate each variant determined by a trained observer.



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Thank you



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