

# Manual versus automatic identification of black-capped chickadee (*Poecile atricapillus*) vocalizations Vala Ingolfsson, William D. Service, Carolina Montenegro, & Christopher B. Sturdy Department of Psychology, Neuroscience and Mental Health Institute, University of Alberta

### Introduction

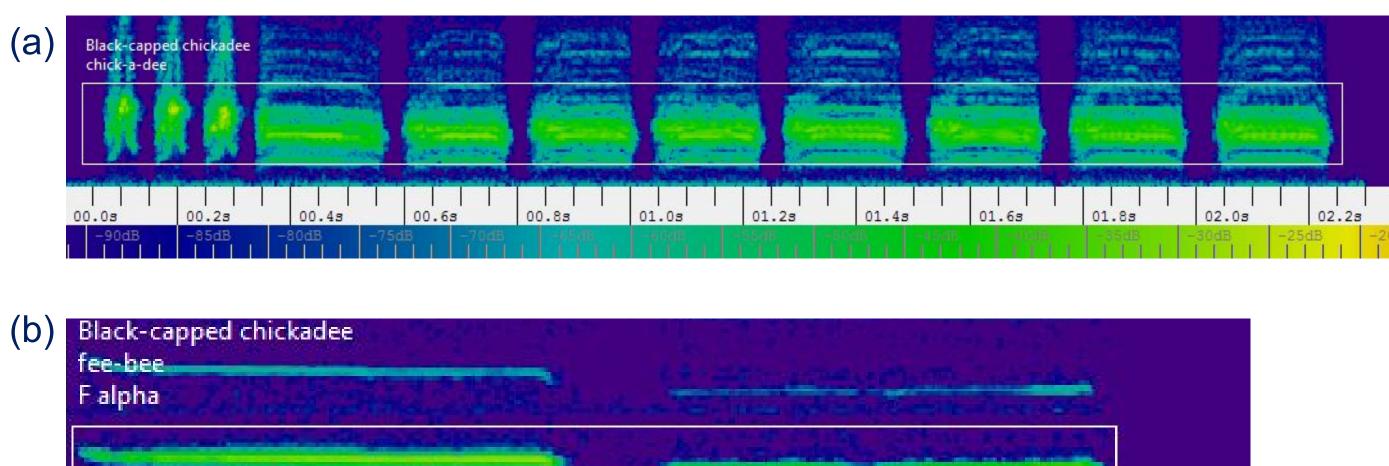
- Black-capped chickadees (BCCH) are ideal subjects for studying communication and vocal learning.
- Studies often involve "call cutting" (i.e., isolating and identifying vocalizations from hours of recordings).
- **SongScope** is a computer program used to create recognizers that identify specific animal vocalizations.
- The current study tests how recognizers built in SongScope compare to manual call cutting.
- In addition we assessed how the time of day and noise impacts vocalizations produced.



chickadee.

## Methods

• A recognizer was generated in SongScope for each chickadee vocalization (i.e., *chick-a-dee* call, *chick-a* call, tseet call, tseet cluster, gargle call, fee song, *fee-bee* song), using pre-existing samples.

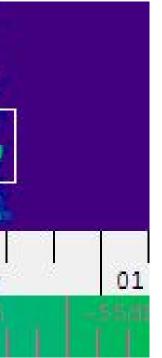


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Figure 2. Annotations in SongScope. (a) *chick-a-dee* call; (b) *fee-bee* song.

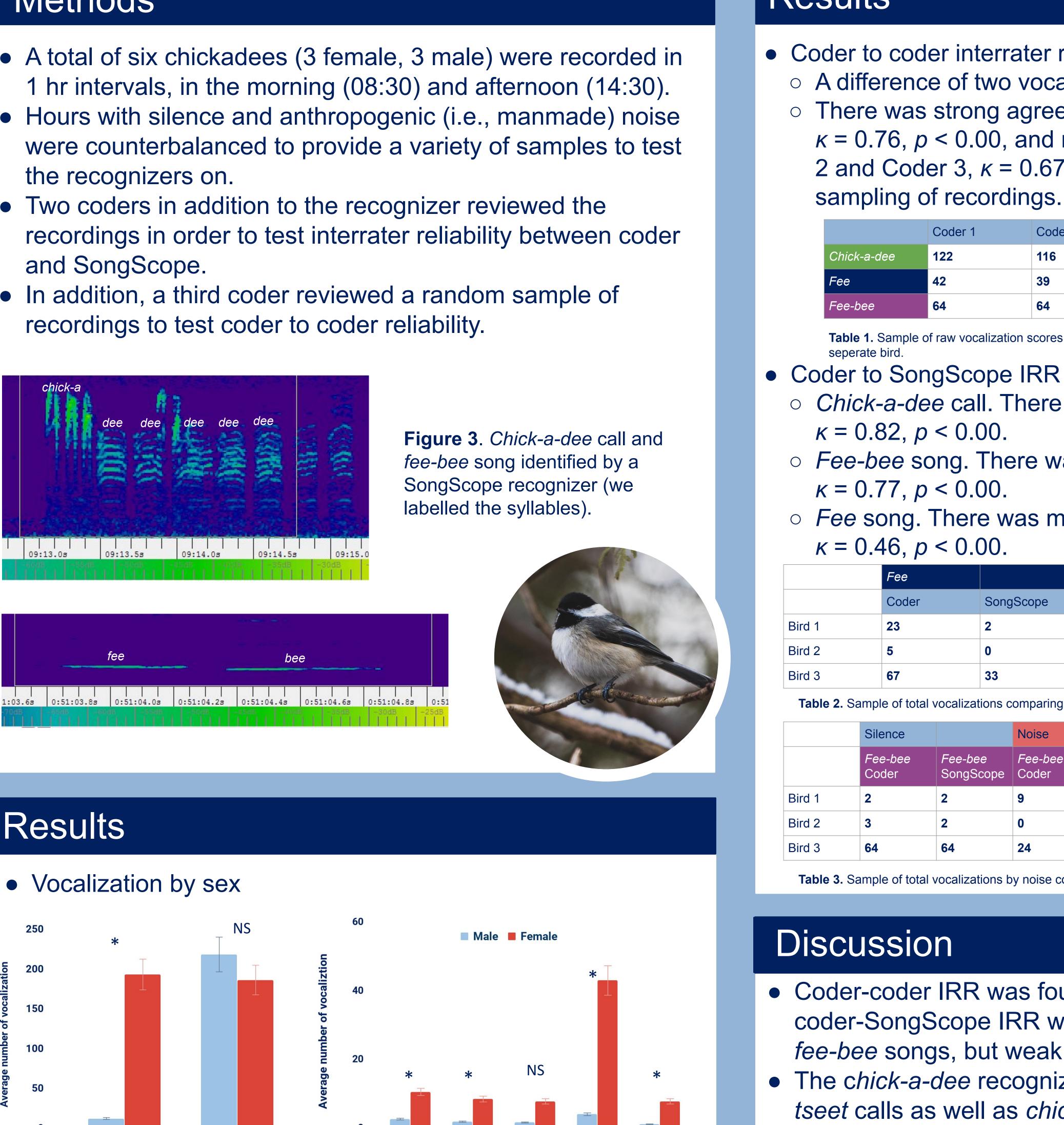
- By annotating samples from individuals the program develops a model of what each vocalization should look like.
- By adjusting variables such as frequency range, maximum song length, and the length of syllables and the gaps between them, the recognizer can be made more accurate and specific.

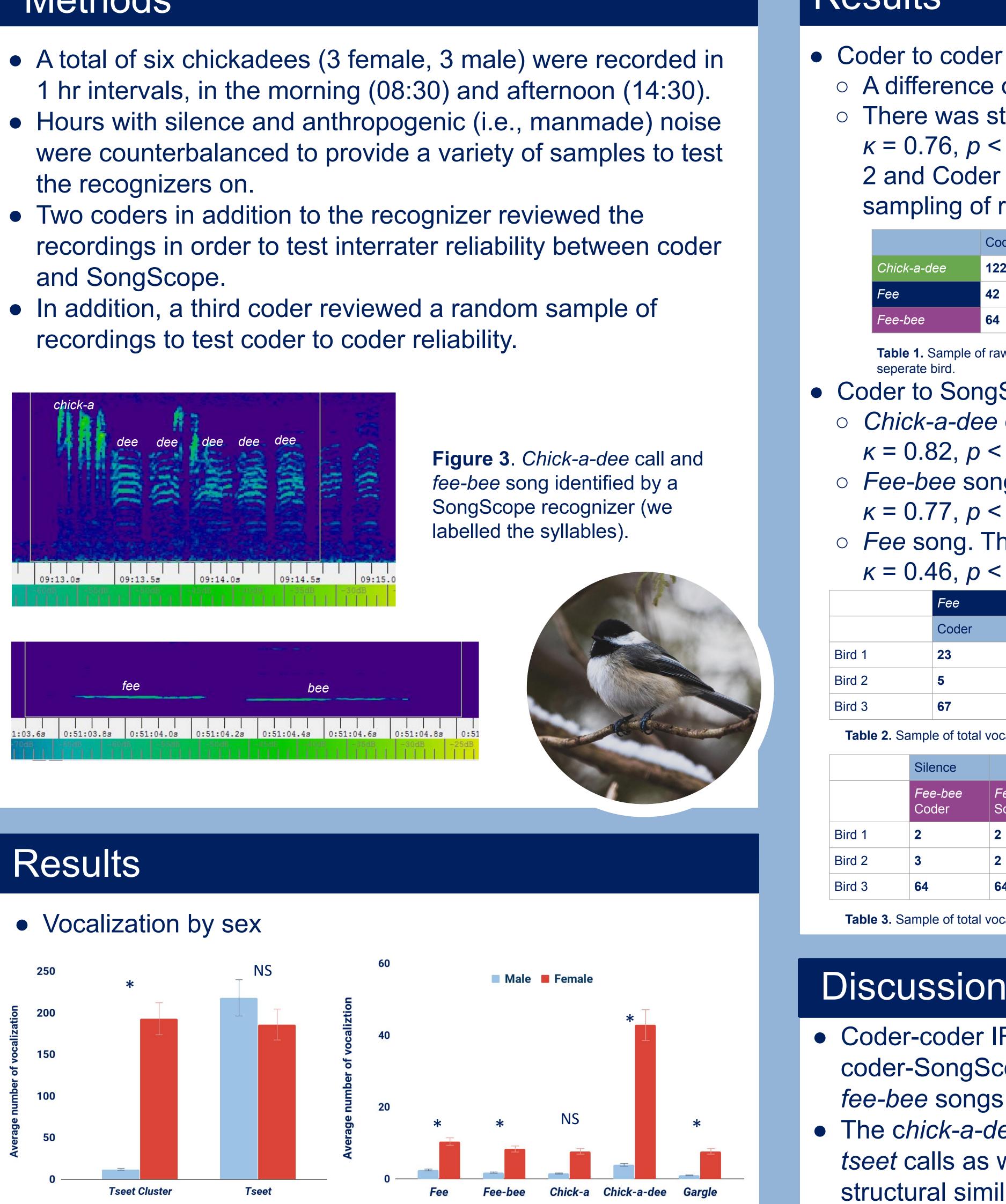
Figure 1. Black-capped ÚSTICS (Version 4.1.5. Wildlife Acoustics, Inc.)



### Methods

- the recognizers on.
- and SongScope.





**Figure 4.** Differences of average vocalizations by sex. \* indicate *ps* < 0.05, NS show no significance. Error bars represent 10% error amount for each data point.

- Males produced more *tseets* during Silence, *p* < 0.01, and females produced more *gargles* during Noise, *p* < 0.00.
- There were no significant differences in vocalizations by time of day.

### Results

• Coder to coder interrater reliability (IRR) sampling of recordings.

	Coder 1	Coder 3		Coder 2	Coder 3
Chick-a-dee	122	116	Chick-a-dee	22	20
Fee	42	39	Fee	5	3
Fee-bee	64	64	Fee-bee	3	5

 Table 1. Sample of raw vocalization scores between Coder 1 and 2 for one bird and between Coder 2 and 3 for a

- $\kappa = 0.82, p < 0.00.$
- *κ* = 0.77, *p* < 0.00.
- $\kappa = 0.46, p < 0.00.$

	Fee		Fee-bee		Chick-a-dee		
	Coder	SongScope	Coder	SongScope	Coder	SongScope	
Bird 1	23	2	11	6	0	2	
Bird 2	5	0	3	2	40	42	
Bird 3	67	33	88	81	250	249	

Table 2. Sample of total vocalizations comparing Coder and SongScope coding.

	Silence	Silence		Noise		Silence		Noise	
	<i>Fee-bee</i> Coder	<i>Fee-bee</i> SongScope	<i>Fee-bee</i> Coder	<i>Fee-bee</i> SongScope	<i>Chick-a-dee</i> Coder	<i>Chick-a-dee</i> SongScope	<i>Chick-a-dee</i> Coder	<i>Chick-a-dee</i> SongScope	
Bird 1	2	2	9	4	0	0	0	2	
Bird 2	3	2	0	0	22	22	18	22	
Bird 3	64	64	24	17	122	120	128	129	

Table 3. Sample of total vocalizations by noise condition, type by Coder and SongScope coding.

- fee-bee songs, but weak for fee songs.
- structural similarity. accuracy.

### Acknowledgements

Special thanks to all the Neuroethology Laboratory members.





• A difference of two vocalizations was allowed for agreement • There was strong agreement between Coder 1 and Coder 3,  $\kappa$  = 0.76, p < 0.00, and moderate agreement between Coder 2 and Coder 3,  $\kappa$  = 0.67, p < 0.00, based on a random

• *Chick-a-dee* call. There was strong agreement by recording,

• *Fee-bee* song. There was strong agreement by recording,

• *Fee* song. There was moderate agreement by recording,

### Coder-coder IRR was found to be satisfactory, and

coder-SongScope IRR was strong for *chick-a-dee* calls and

• The chick-a-dee recognizer was able to identify gargle and tseet calls as well as chick-a-dee calls, possibly due to

• Recognizers can be continuously improved for greater

• Call cutting by SongScope was found to be much faster (48) hours versus approximately 12 hours) than human call cutting.