



UNIVERSITY of WASHINGTON

Investigation of Duty Cycles in Passive Acoustic Bat Monitoring

By Aditya Krishna, mentored by Dr. Wu-Jung Lee

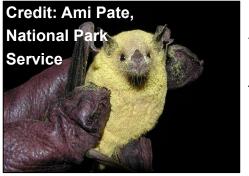
Introduction

 Why monitor bats using Passive Acoustic Monitoring (PAM)?

2) Why investigatesubsampling forbat-specific PAM?

Why are we monitoring bats with PAM?

- Bats have important environmental roles:
 - Pollinators
 - Natural Insecticides

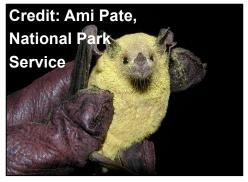


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(Russo et al. 2021)

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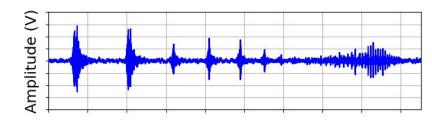
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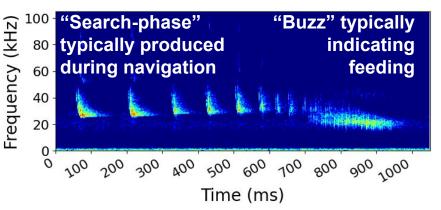
(Russo et al. 2021)

*Humans can only hear sounds less than 20 kHz so all played bat calls are slowed down by 1.2x to be **audible**. (Sugai et al. 2019)

(Surlykke et al. 2014)

- Bats echolocate for various activity!
- These calls can inform on behavior
- Thus, Passive Acoustic Monitoring!





- High sampling rate
- Long-term application

Large data volume and logistics

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Subsampling:

Combination of cycle length and duty cycle

Large data

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• Reduces the logistics and data issues

Cycle Length: 12-min, Duty Cycle: 16.7% or 2/12 ON ON ON 6 10 12 14 16 18 20 22 24 26 28 30 8 Minutes On for $12 \times 16.7\% = 2$ Cycle Length: 6-min, Duty Cycle: 33.3% or 2/6 ON ON ON ON ON 6 8 10 12 14 16 18 20 22 24 26 28 30 Δ Minutes On for $6 \times 33.3\% = 2$

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 Subsampled data did not capture the behaviors that were observed from continuous data (Rand et al. 2022)

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Research Question:

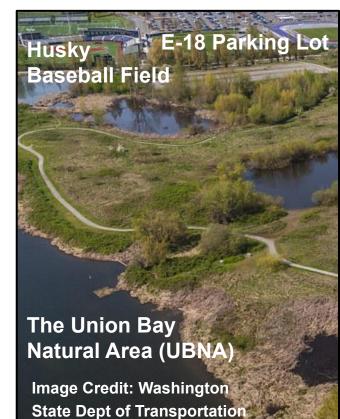
How does recording duty cycle and cycle length affect acoustic monitoring of bat activity?

Methods & Analysis

1) Data collection & analysis

2) Applied subsampling on collected data

Data was collected from UW's natural laboratory!



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Audiomoth

 Equipped w/ rechargeable batteries and an SD card

The Union Bay Natural Area (UBNA)

Husky

Baseball Field

E-18 Parking Lot

"Central Pond"

Image Credit: Washington State Dept of Transportation



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Collection Period:

 Daily from July - October 2022, decrease in activity due to hibernation.

Internal Settings:

- No subsampling
- Sampling rate: 192000 samples/sec needed for ultrasonic calls

Data rate: ~30GB / day

Lasts only 3-4 days before needing replacements

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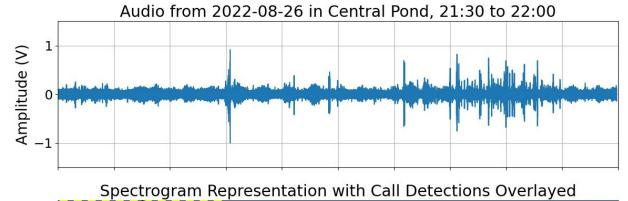
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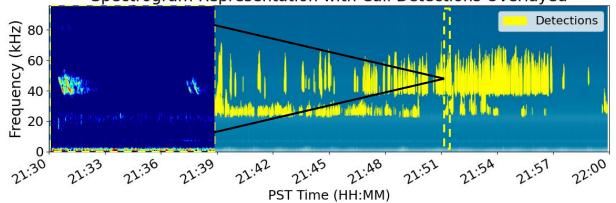
'BatDetect2' (Aodha et al. 2022)

- Detects 'search-phase' calls in recorded audio
- Uses a CNN model

Software Pipeline: 'bat-detector-msds'

 Developed by UW's Masters in Data Science students!





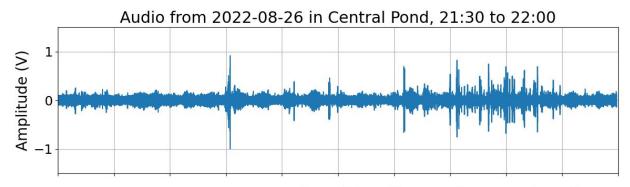
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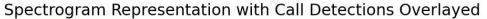
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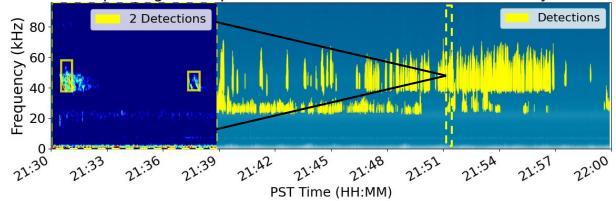
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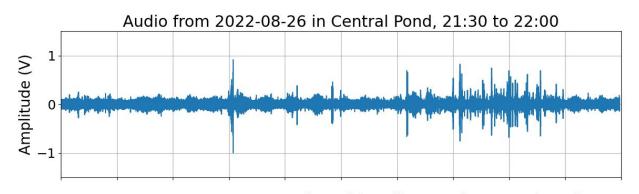
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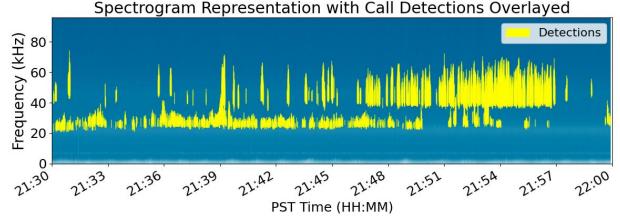
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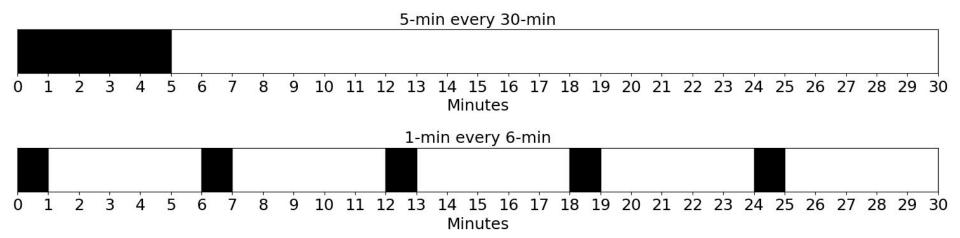
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2 subsampling schemes were applied on collected recordings

- Cycle Length: 30-min, Duty Cycle: 16.7%
 - ON for 30 × 0.167 = 5-min every 30-min
- Cycle Length: 6-min, 16.7%
 - ON for 6 × 0.167 = 1-min every 6-min

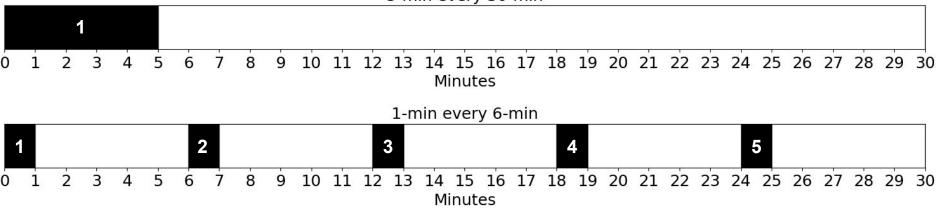


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Schemes have fixed sampling effort in observed time: 20:00 to 6:30 PST

5-min every 30-min



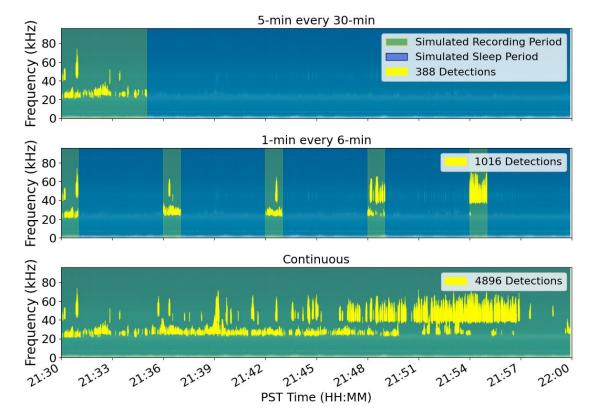
Results & Discussion

1) Recorded bat call features

2) Subsampling effects

Data consisted of:

- Low-Frequency Bat Calls (~20-45 kHz)
- High-Frequency Bat Calls (~35-75 kHz)



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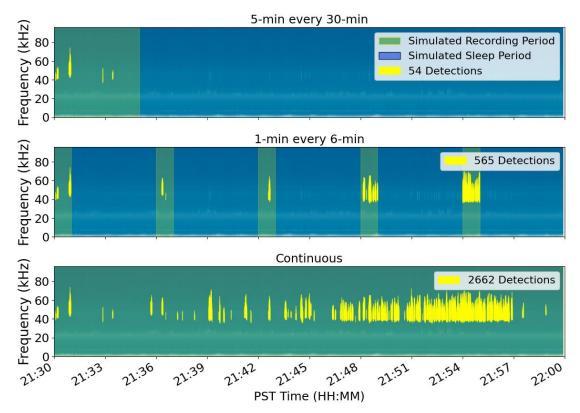
5-min every 30-min Frequency (kHz) 0 0 0 08 0 0 Simulated Recording Period Simulated Sleep Period 54 Detections 1-min every 6-min Frequency (kHz) 0 0 0 0 0 0 565 Detections Frequency (KHz) Continuous 2662 Detections 80 21:30 21:33 21:36 22:00 21:39 21:48 21:51 21:54 21:57 21:42 21:45 PST Time (HH:MM)

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Features of bat activity:

- Nightly distribution during the Summer
- Nightly presence during the Fall



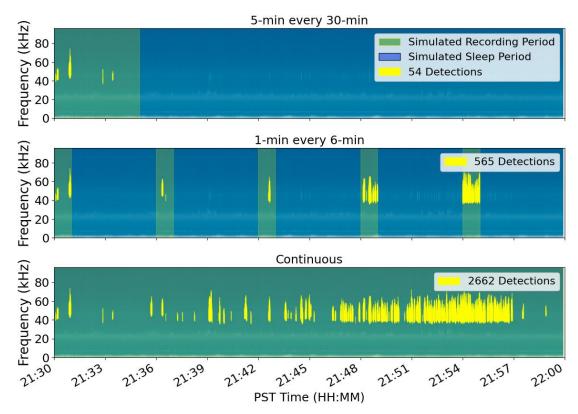
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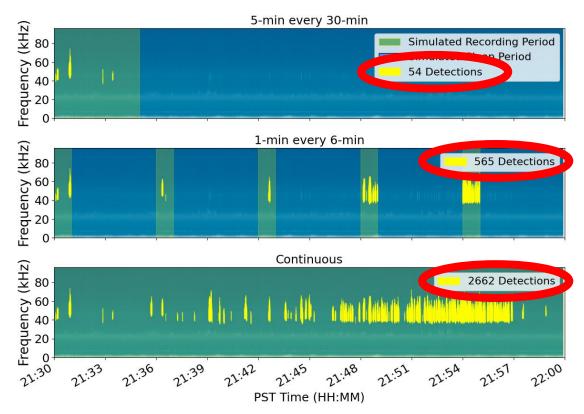
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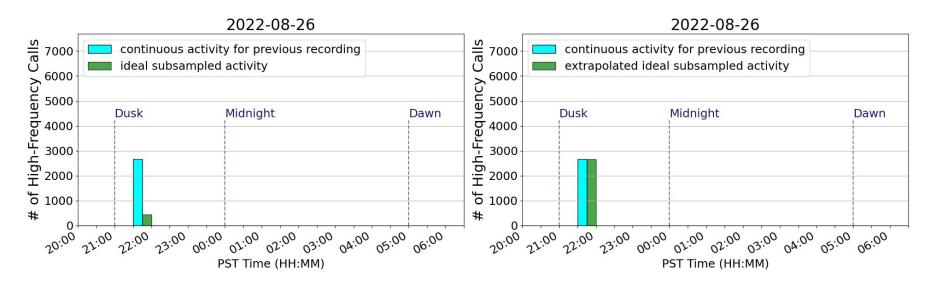
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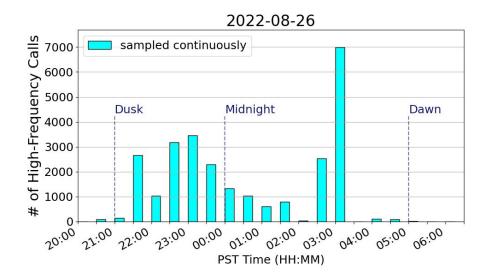


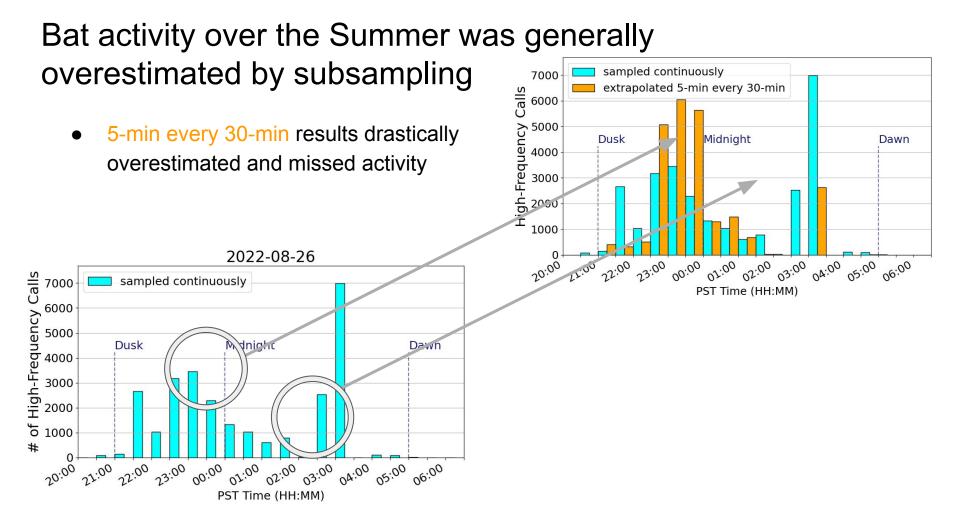
Number of calls will always be underestimated by subsampling

- Question: How representative is our sample of the recorded calling activity?
 - In other words, given features from 5-min samples, could a simple 6× extrapolation recover the features from 30-min of calling activity?



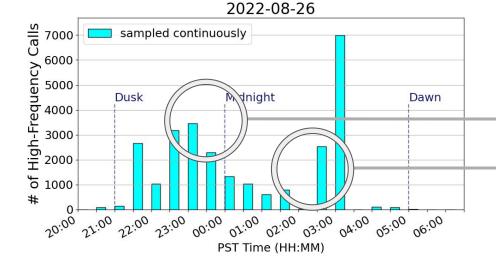
Bat activity over the Summer was generally overestimated by subsampling

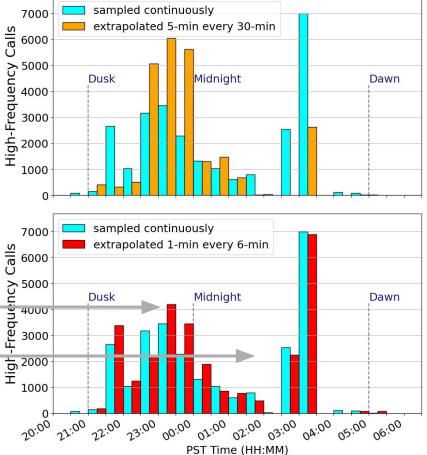




Bat activity over the Summer was generally overestimated by subsampling

- 5-min every 30-min results drastically overestimated and missed activity
- 1-min every 6-min results slightly overestimated and captured activity

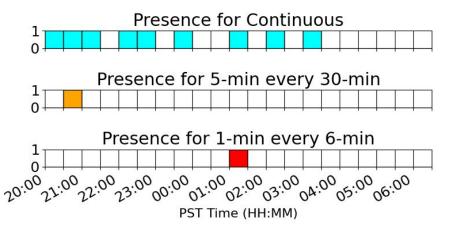




Bat presence over the Fall was underestimated by subsampling

For Oct-17:

- Bat presence was detected in continuous data for 9 separate instances
- Each scheme detected only 1 instance

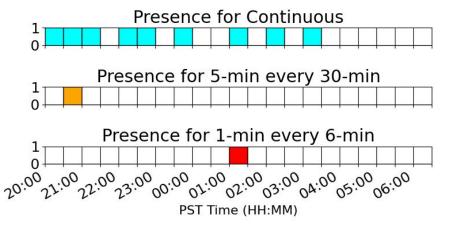


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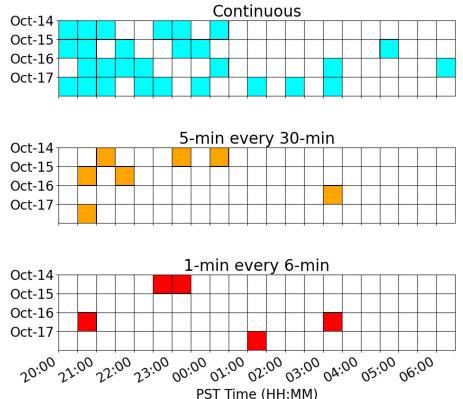
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High-Frequency Bat Presence Comparisons



1) Subsampling can leave out behavioral information found in continuous data (Rand et al. 2022)

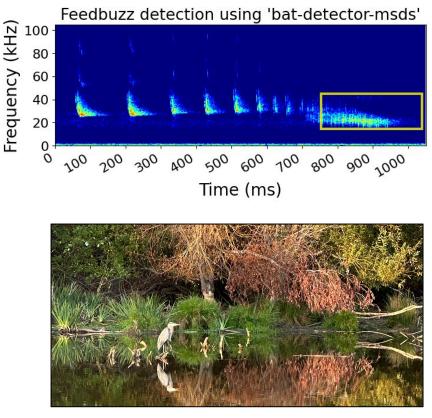
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- Our experiments support this and show that the same subsampling schemes can have varying effects on different features of seasonal call activity
- 3) PAM still requires subsampling so further investigation is required to identify schemes to capture desired activity features.
- 4) Fortunately, the UBNA bat population remains feeding and healthy so we can continue to noninvasively study these effects!

Future steps

- Simulating more schemes and looking at seasonal variations!
- Developing tools for detecting specific bat acoustic behaviors:
 Feeding buzzes
- Continuing our PAM program at the Union Bay Natural Area this Summer!



Resident Great Blue Heron at the UBNA

Acknowledgements

Contributors:







Wu-Jung Lee

Josie Sachen YeonJoon Cheong







Echospace @ UW

uw-echospace.github.io



Applied Physics Laboratory UNIVERSITY of WASHINGTON



Authors of

'bat-detector-msds':

- Corbin Charpentier
- Kirsteen Ng
- Ernesto Cediel

MARY GATES ENDOWMENT FOR STUDENTS