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## Abstract

- The **HumBug** project [1] aims to combat the spread of mosquito-borne diseases.
- We use a Bayesian convolutional neural network (**BCNN**) to successfully detect mosquitoes from their acoustics in challenging real-world conditions.
- Previous work has not quantified uncertainty, which we require for field deployment due to the variable nature of the target environment conditions.

## HumBug pipeline



(a) Bednet with four smartphones positioned to trial the best location for recording mosquitoes.



(b) Bednet pockets to hold smartphones for recording.

**Figure 1:** Deployment in Tanzania (Oct 2020) to trial the effectiveness of acoustic mosquito detection with low-cost non-invasive measures.

- Smartphones are placed in bednet corners to allow autonomous data collection.
- Anonymised phone recordings synchronise to a central server and database.
- The recordings are screened with **Algorithm 1** to identify at-risk areas.

### Algorithm 1 BCNN detection

- 1: Read 8 kHz mono wave file
- 2: Take sliding window log-mel transform ( $40 \times 128$  frames, each frame duration 64 ms)
- 3: Take  $N$  MC dropout samples from BCNN (2 conv/max-pool layers with  $3 \times 3$  kernels)
- 4: Calculate mean probability  $\bar{p}_C$ , predictive entropy (**PE**), mutual information (**MI**)
- 5: Output the mean of  $\bar{p}_C$ , PE, MI per continuous section with  $\bar{p}_{\text{mosquito}} > p_{\text{threshold}}$

## Test performance

	Duration (h)	Class acc. (%)
Signal A	2.8	$89.27 \pm 0.07$
Noise A	1.3	$94.05 \pm 0.11$
Noise B	3.0	$97.99 \pm 0.05$

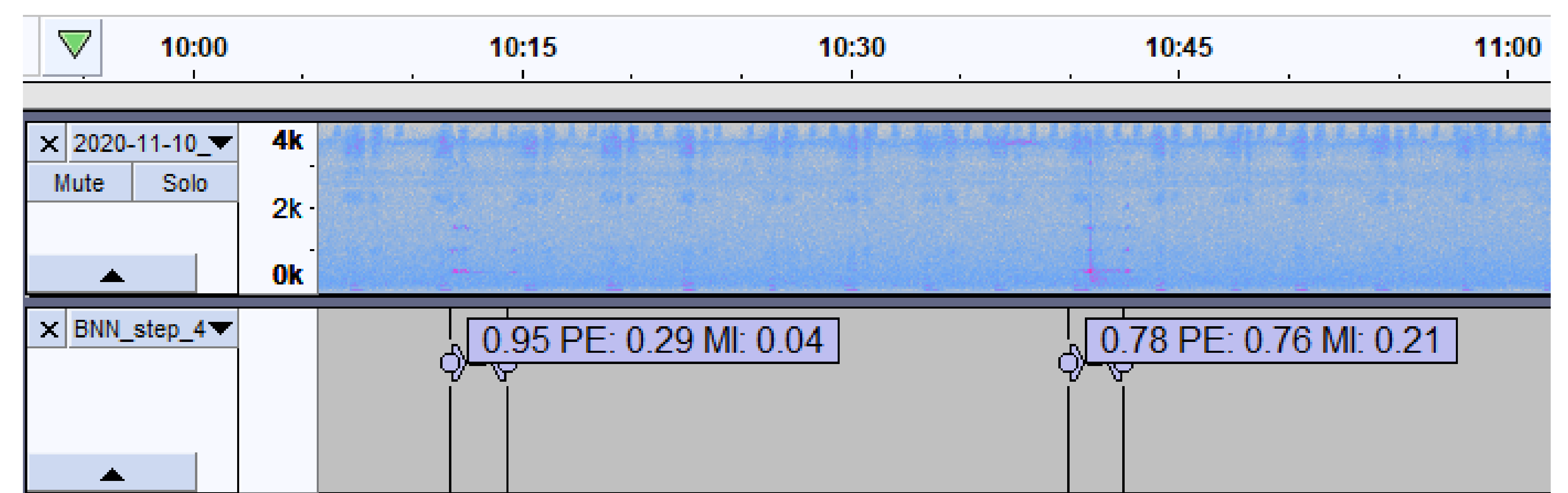
**Table 1:** Out-of-sample performance estimated with  $N = 10$  MC dropout samples (mean  $\pm$  standard deviation). Signal A: lab and home mosquito recordings; Noise A: corresponding background. Noise B: bednet (Figure 1a) field background noise.

True label	Predicted label	
	Noise	Mozz
Noise	$97.24 \pm 0.03$	$2.76 \pm 0.03$
Mozz	$10.73 \pm 0.07$	$89.27 \pm 0.07$

**Figure 2:** Out-of-sample performance on sources from Table 1 combined.

## Unlabelled data

- Import the audio and predicted labels to screen detections in Audacity [2].
- We can accept low entropy, high posterior predictive probability detections in the case where the audio background is noisy.
- Conversely, for good recording conditions we can accept higher entropy predictions to encourage higher recall of mosquito events.



**Figure 3:** BCNN predictions on unlabelled field data (Nov 2020) in Audacity in the form:  $\{\hat{p}_{\text{mosz}}, \text{PE}, \text{MI}\}$ . Two windows with mosquito present were correctly identified in this section of audio, recorded in a similar arrangement to that shown in Figure 1.

## References

- [1] The HumBug project homepage: <https://humbug.ox.ac.uk/>
- [2] Audacity. Open-source cross-platform audio editor. <https://www.audacityteam.org/>

## Acknowledgements

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