

LAMOC

**Development of an IoT-Based
Acoustic Signal Sensing Device
in Mosquito Monitoring using
Raspberry Pi**

Fianza | Manguan | Ponio



Mosquitoes bring **deadly diseases***



Anopheles sp.

malaria



Aedes aegypti

dengue



Culex sp.

west nile virus

* Tolle, M. A. (2009). Mosquito-borne diseases. *Current Problems in Pediatric and Adolescent Health Care*, 39(4), 97–140.
doi:10.1016/j.cppeds.2009.01.001
Images taken from <http://medent.usyd.edu.au/arbovirus/mosquit/photos/mosquitphotos.htm>

Existing methods are **not enough***

	Human landing catch	Bioacoustics
Contact	close-contact	none
Specialized knowledge	species identification	none, automated
Results obtained	after several days	real-time
Cost	\$	\$\$\$

* Dia, I., Diallo, D., Duchemin, J., Ba, Y., Konate, L., Costantini, C., & Diallo, M. (2005). Comparisons of Human-Landing Catches and Odor-Baited Entry Traps for Sampling Malaria Vectors in Senegal. *Journal of Medical Entomology*, 42(2), 104–109. <https://doi.org/10.1093/jmedent/42.2.104>

Mosquitoes are uniquely identified by their wingbeat frequencies*

Ae. aegypti (male)

900-1000 Hz**

Ae. aegypti (female)

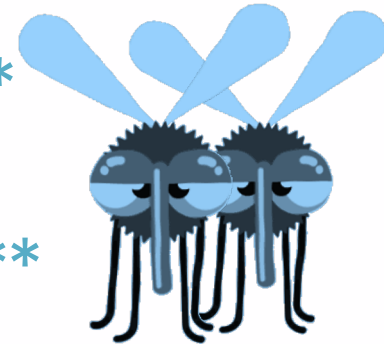
550-600 Hz**

Anopheles sp.

320-480 Hz***

Culex sp.

500-650 Hz**

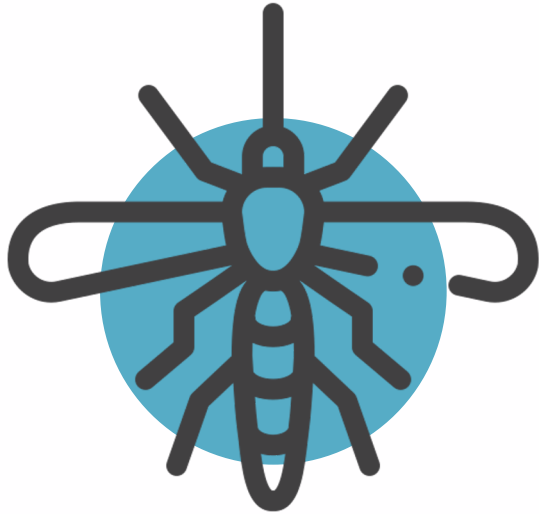


* Arthur, B. J., Emr, K. S., Wytenbach, R. A., & Hoy, R. R. (2014). Mosquito (*Aedes aegypti*) flight tones: Frequency, harmonicity, spherical spreading, and phase relationships. *The Journal of the Acoustical Society of America*, 135(2), 933–941. doi:10.1121/1.4861233

** Vasconcelos, D., Nunes, N., Ribeiro, M., Prandi, C., & Rogers, A. (2019). LOCOMOBIS: a low-cost acoustic-based sensing system to monitor and classify mosquitoes. 2019 16th IEEE Annual Consumer Communications & Networking Conference (CCNC), 1–6. <https://doi.org/10.1109/CCNC.2019.8651767>

*** Caprio MA, Huang JX, Faver MK, & Moore A. (2001). Characterization of male and female wingbeat frequencies in the *Anopheles quadrimaculatus* complex in Mississippi. *Journal of the American Mosquito Control Association*, 17(3), 186–189.

What we need is a device that can



detect the
presence of
mosquitoes



identify the
species of the
mosquitoes



inform
interested
individuals

LAMOC

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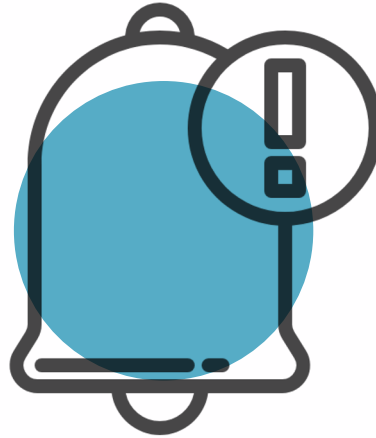
Fianza | Manguan | Ponio



Significance



provides **safer**, more **sustainable**, and **efficient** mosquito monitoring



informs **health centers** in identifying the trend of mosquito population



allows LGUs to formulate **countermeasures** for a possible **outbreak**

Scope and Limitations

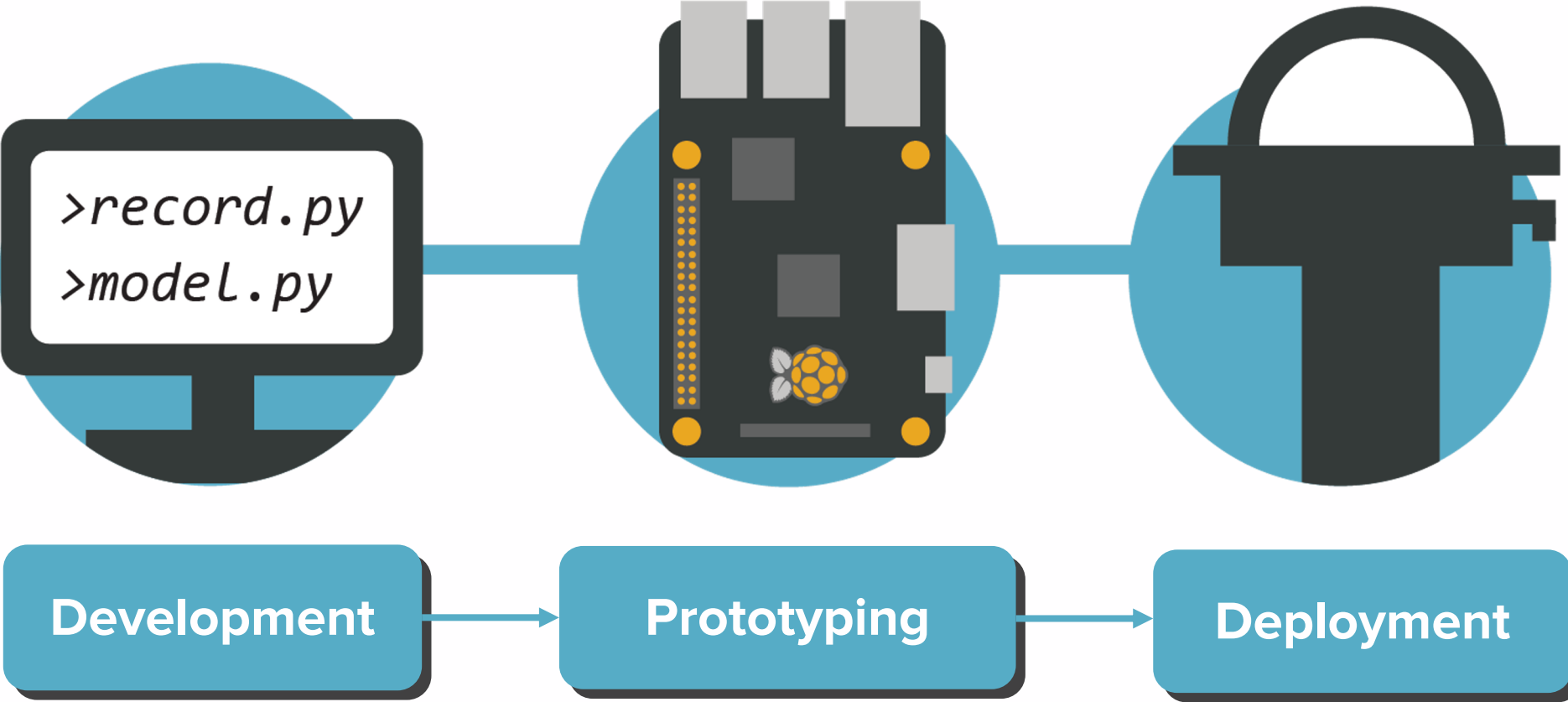
- create a **single** device
- use audio samples from **other researchers**
 - *Culex sp.* *
 - *Anopheles sp.* **
 - *Aedes aegypti* ***
- detect **three** mosquito species
- identify using **wing beat frequencies**

* Li, Y., Zilli, D., Chan, H., Kiskin, I., Sinka, M., Roberts, S., & Willis, K. (2017). Mosquito detection with low-cost smartphones: data acquisition for malaria research. NIPS Workshop on Machine Learning for the Developing World (ML4D), 1–5.

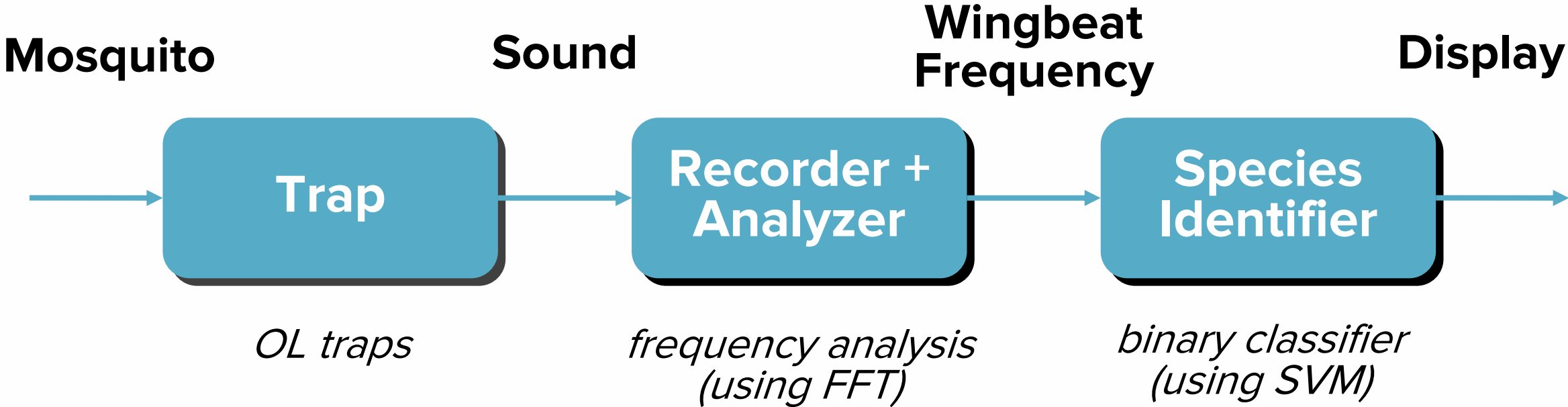
** Mukundarajan, H., Hol, F. J. H., Castillo, E. A., Newby, C., & Prakash, M. (2017). Using mobile phones as acoustic sensors for high-throughput mosquito surveillance. *ELife*, 6. <https://doi.org/10.7554/eLife.27854>

*** Homer, M., Champneys, A., Aldersley, A., & Robert, D. (2015). Lone & Pair Mosquito Auditory Interaction. Engineering and Physical Sciences Research Council.

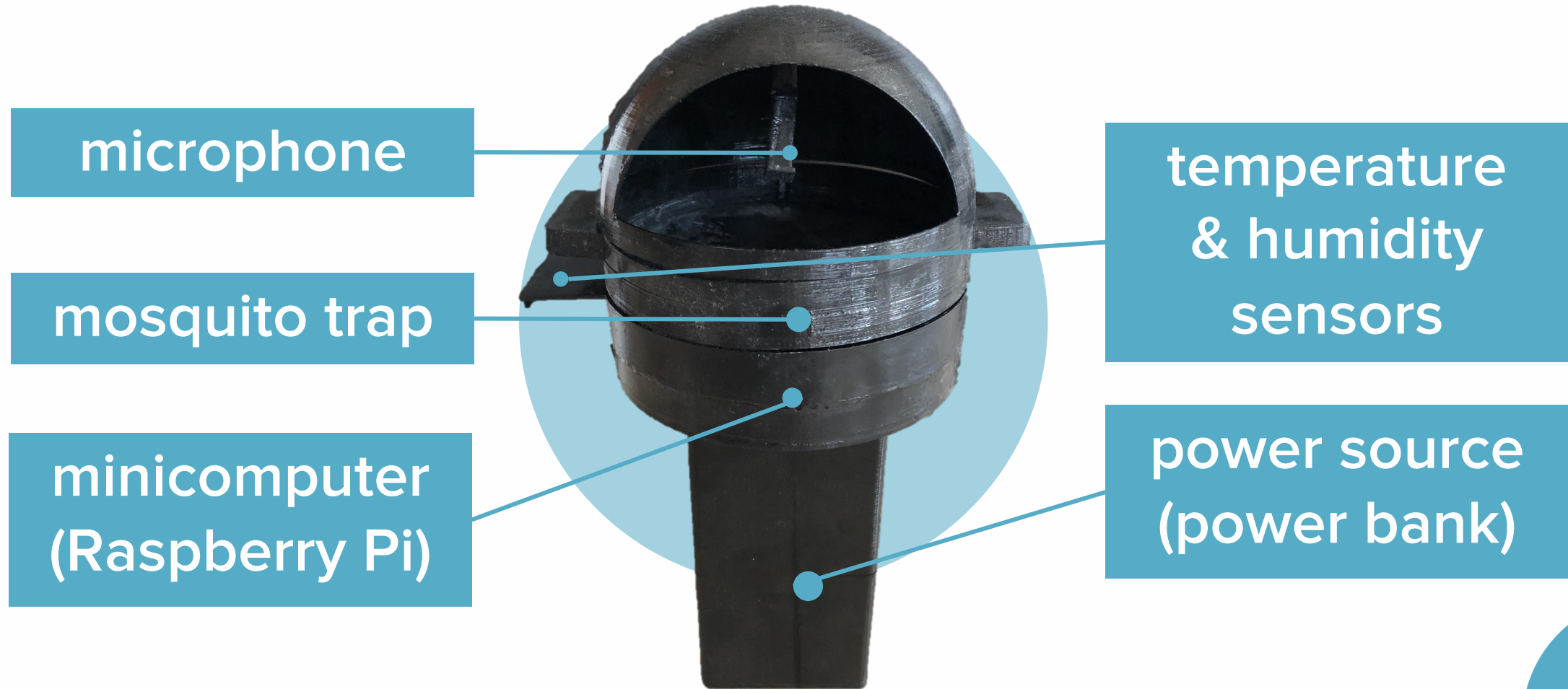
Methodology



How the device works



LAMOC Components



microphone

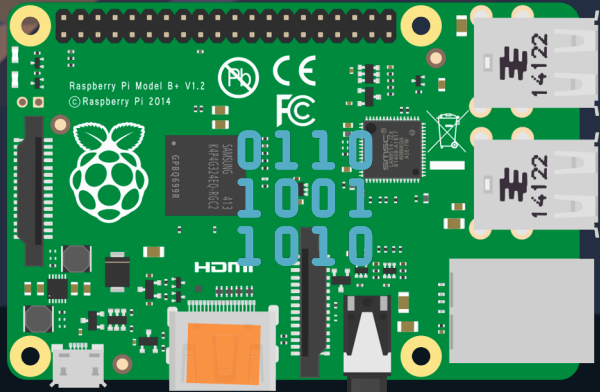
mosquito trap

minicomputer
(Raspberry Pi)

temperature
& humidity
sensors

power source
(power bank)





0110
1001
1010



identify species

record temperature
& humidity

send to server

Device 1

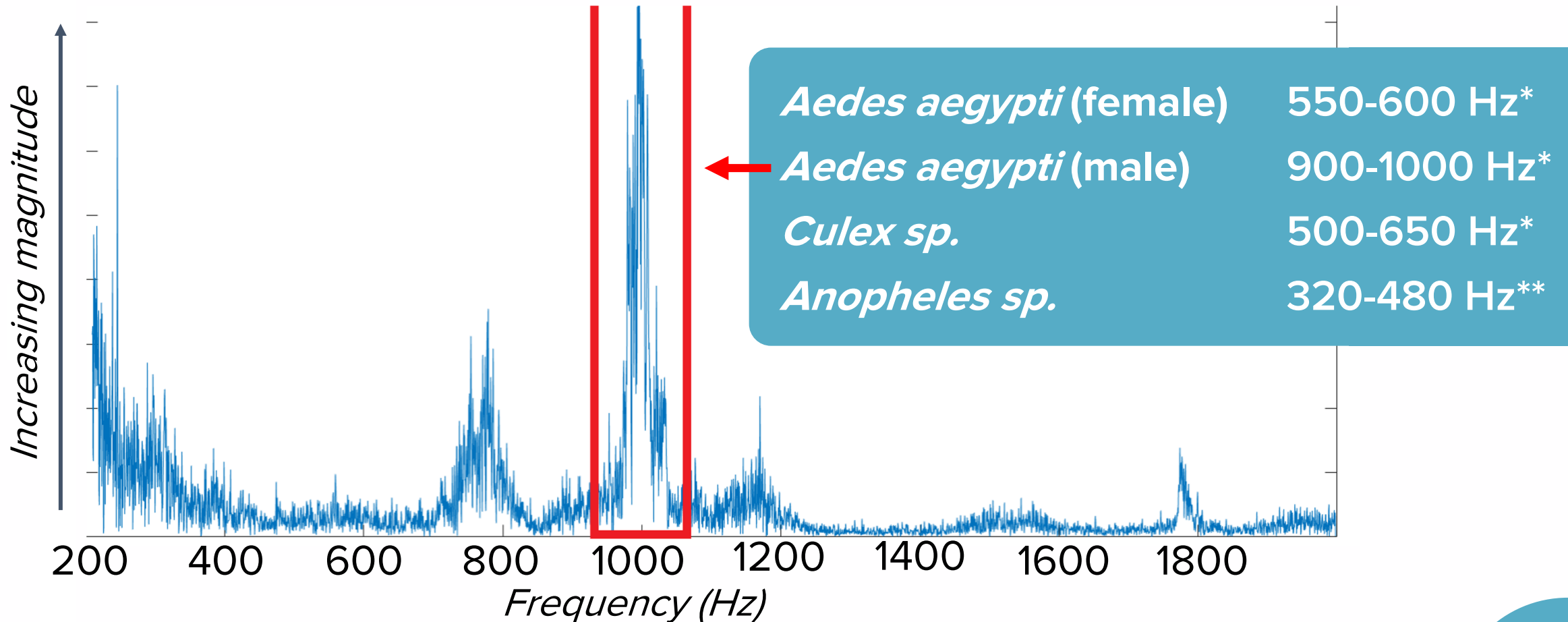
 PSHS-CLC Clark

Number

total number of audio recorded

0110
1001
1010

LAMOC detects mosquitoes



* Vasconcelos, D., Nunes, N., Ribeiro, M., Prandi, C., & Rogers, A. (2019). Locomobis: a low-cost acoustic-based sensing system to monitor and classify mosquitoes. 2019 16th IEEE Annual Consumer Communications & Networking Conference (CCNC), 1–6. <https://doi.org/10.1109/CCNC.2019.8651767>

**Caprio MA, Huang JX, Faver MK, & Moore A. (2001). Characterization of male and female wingbeat frequencies in the *Anopheles quadrimaculatus* complex in Mississippi. *Journal of the American Mosquito Control Association*, 17(3), 186–189.

LAMOC detects mosquitoes

Table 1. Confusion matrix for mosquito detection

n = 5500		Predicted	
		Positive	Negative
Actual	Positive	TP: 4200	FN: 200
	Negative	FP: 105	TN: 995

Precision

$TP / (TP+FP)$

97.56%

Recall

$TP / (TP+FN)$

95.45%

Accuracy

$(TP+TN) / n$

94.45%

LAMOC identifies mosquito species

Table 2. Accuracy rates for the five classes using cross-validation

	<i>Aedes aegypti</i> (male)	<i>Aedes aegypti</i> (female)	<i>Anopheles sp.</i>	<i>Culex sp.</i>	Background noise
Accuracy	99.30%	96.74%	93.72%	92.09%	90.47%
n	1100	1100	1100	1100	1100

LAMOC has higher accuracy

Table 3. Comparison of LAMOC to Li et al. (2017)

	Smartphone detector (Li et al., 2017)	LAMOC
Species detected	<i>Ae. aegypti</i> <i>Ae. ealbopictus</i> <i>An. albimanus</i> <i>An. gambiae</i> <i>An. quadrimaculatus</i> <i>Cu. tarsalis</i> <i>Cu. quinquefasciatus</i>	<i>Aedes aegypti</i> <i>Anopheles sp.</i> <i>Culex sp.</i>
Samples	372	5500
Overall Accuracy	80.25%	94.45%

LAMOC Website

ID	Play	Species	Location	Time	Date	Temp (°C)	Humidity (%)
13	▶	<i>Aedes aegypti (female)</i>	PSHS-CLC Clark	09:13:53 am	03/03/2020	23	50
12	▶	<i>Aedes aegypti (female)</i>	PSHS-CLC Clark	09:13:34 am	03/03/2020	23	50
11	▶	<i>Aedes aegypti (female)</i>	PSHS-CLC Clark	10:03:05 pm	03/02/2020	23	50

ID	Play	Species	Location	Time	Date	Temp (°C)	Humidity (%)
13	▶	<i>Aedes aegypti (female)</i>	PSHS-CLC Clark	09:13:53 am	03/03/2020	23	50

LAMOC was able to



detect the
presence of
mosquitoes



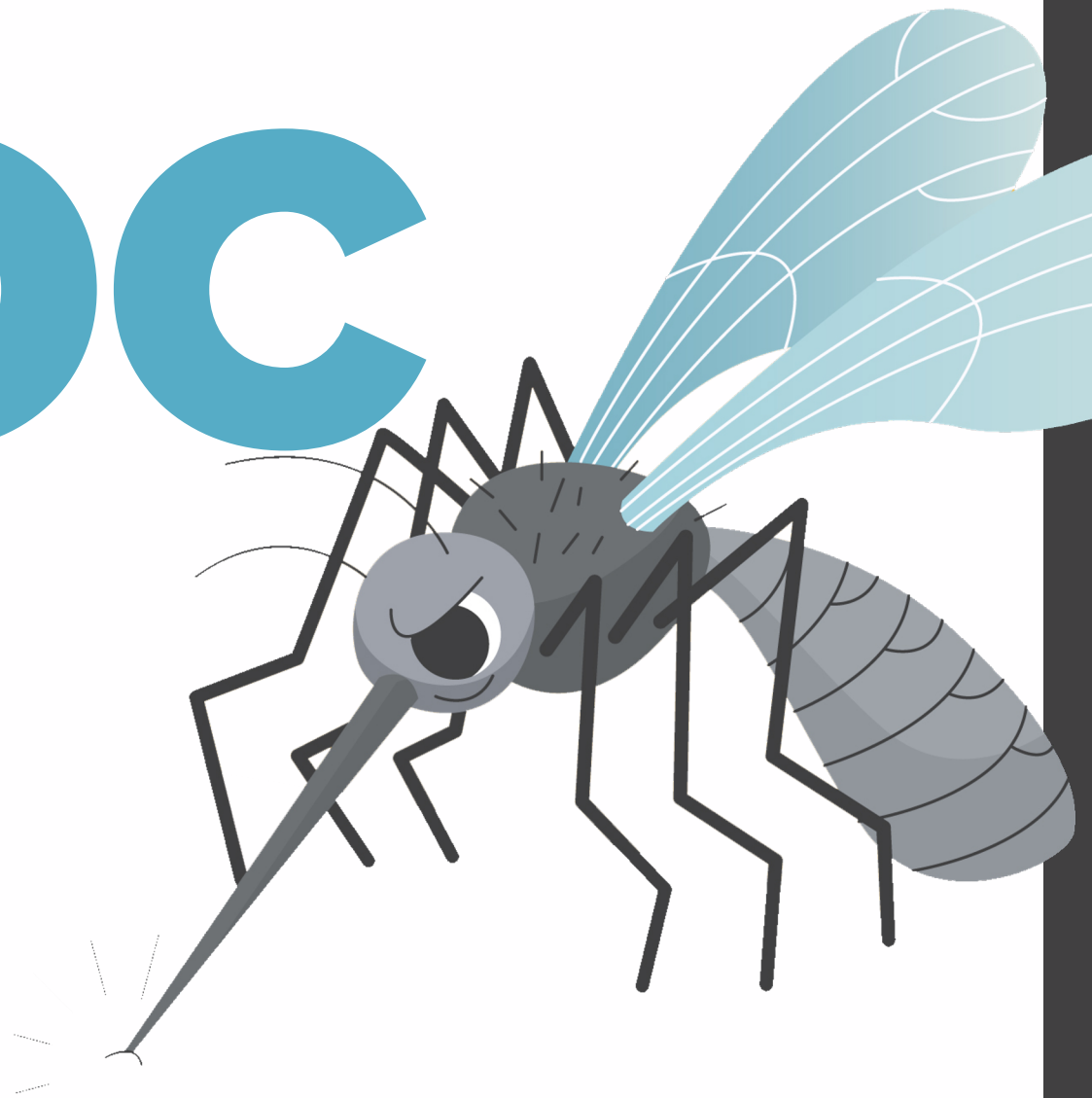
identify the
species of the
mosquitoes



inform
interested
individuals

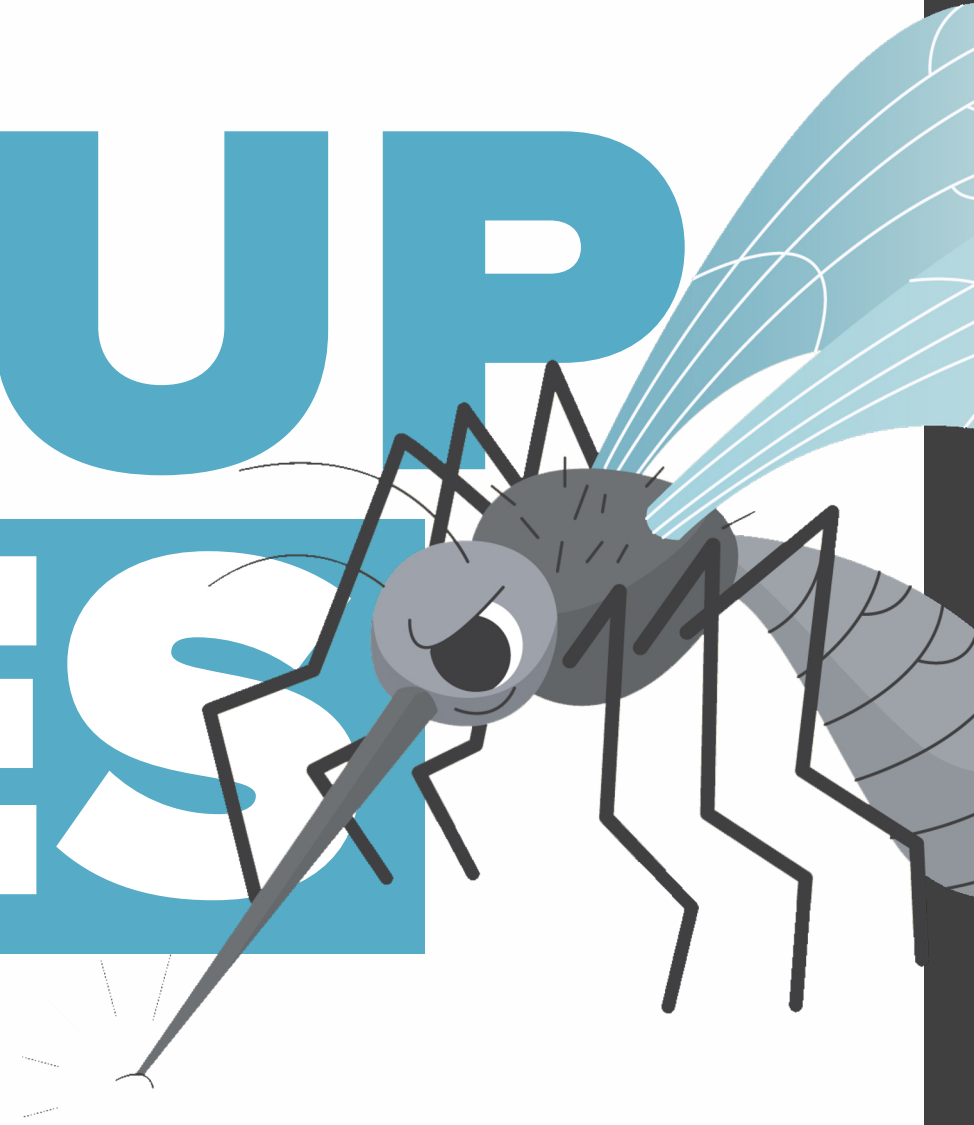
LAMOC

IS HERE
CAN HEAR



BACKUP

SLIDES



Mosquitoes vs orders of insects

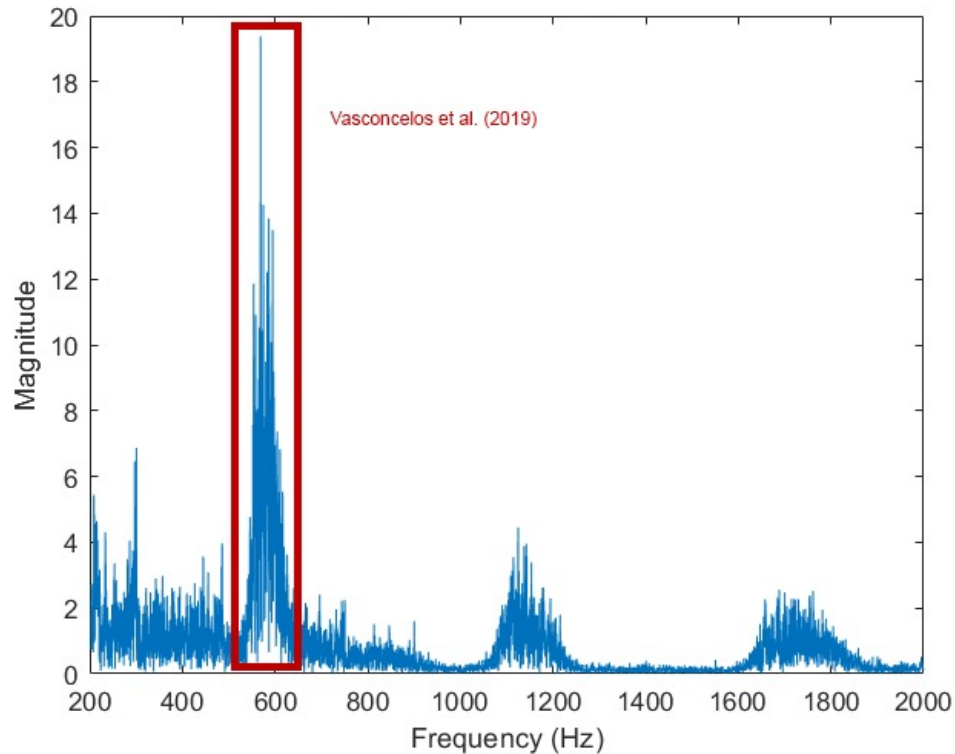
Table 4. Wingbeat frequencies of different insect orders

	Wingbeat frequency (mean Hz)		Bodymass (g)		Wing length (cm)		Wing area (cm ²)		Wing loading (g/cm ²)		Number of specimens
	range	mean	range	mean	range	mean	range	mean	range	mean	
Coleoptera	79 - 123.396	97.512	0.0061 - 0.117	0.0539	0.521 - 1.188	0.898	0.19 - 0.982	0.545	0.0321 - 0.141	0.085	10(10)
Diptera	59.567 - 557.351	208.244	0.0005 - 0.162	0.0268	0.172 - 1.739	0.729	0.022 - 1.17	0.327	0.0119 - 0.168	0.0554	28(28)
Ephemeroptera	n/a	75.0454	n/a	0.0027	n/a	0.634	n/a	0.306	n/a	0.00882	1(1)
Hemiptera	90.222 - 152.247	116.39	0.0011 - 0.14	0.0226	0.345 - 1.185	0.624	0.112 - 1.186	0.445	0.009 - 0.118	0.034	11(11)
Hymenoptera	87.129 - 230.987	163.89	0.0024 - 0.223	0.103	0.356 - 1.48	1.006	0.038 - 1.234	0.64	0.022 - 0.245	0.136	24(15)
Lepidoptera	12.468 - 64.566	39.606	0.0044 - 2.24	0.203	0.646 - 5.214	1.792	0.318 - 23.362	5.031	0.004 - 0.096	0.025	22(22)
Mecoptera	n/a	48.885	n/a	0.0398	n/a	1.387	n/a	1.492	n/a	0.027	1(1)
Neuroptera	25-923 - 94.413	52.801	0.0003 - 0.0065	0.0035	0.352 - 1.393	0.757	0.106 - 1.972	0.701	0.003 - 0.007	0.005	6(6)
Odonata	17.847 - 40.665	32.331	0.0278 - 1.23	0.27	1.795 - 5.158	3.002	1.964 - 22.784	8.768	0.0112 - 0.054	0.022	8(6)
Trichoptera	n/a	27.515	n/a	0.159	n/a	2.267	n/a	4.738	n/a	0.0336	1(1)

Tercel, M.P.T.G., Veronesi, F., & Pope, T.W. (2018). Phylogenetic clustering of wingbeat frequency and flight-associated morphometrics across insect orders. *Physiological Entomology*. <https://doi.org/10.1111/phen.12240>

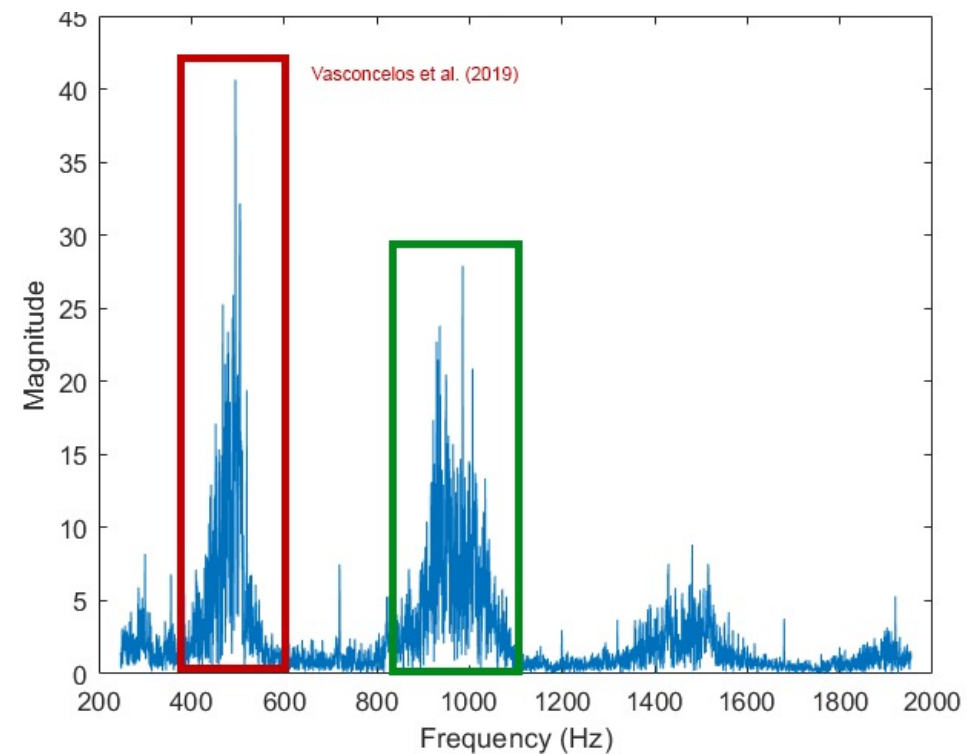
Fundamental Frequencies

Frequency spectrum of wing beat of *Culex*



Culex sp. 500-650 Hz*

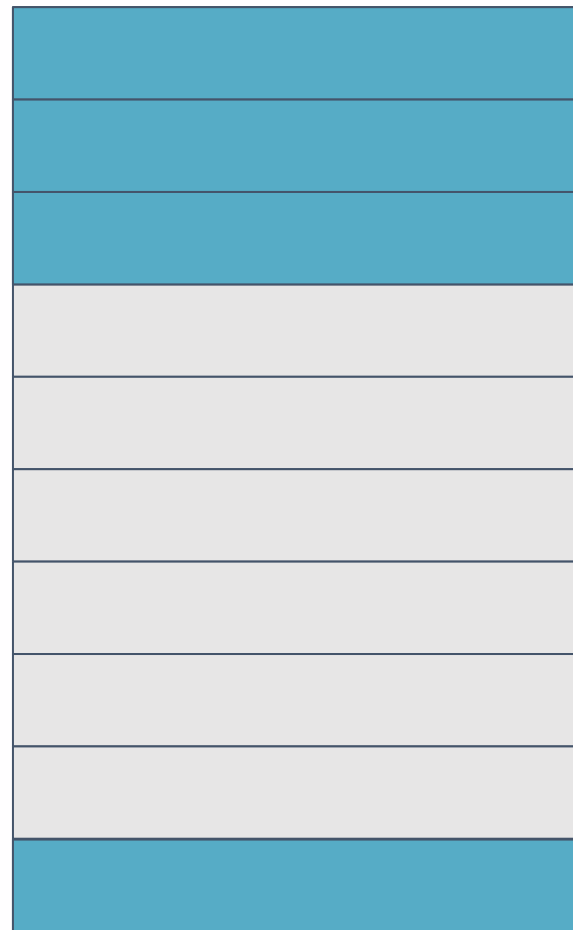
Frequency spectrum of wing beat of *Aedes aegypti* female





Aedes aegypti (female) 550-600 Hz*

* Vasconcelos, D., Nunes, N., Ribeiro, M., Prandi, C., & Rogers, A. (2019). LOCOMOBIS: a low-cost acoustic-based sensing system to monitor and classify mosquitoes. 2019 16th IEEE Annual Consumer Communications & Networking Conference (CCNC), 1–6. <https://doi.org/10.1109/CCNC.2019.8651767>

Cross validation



-  Testing data set
-  Training data set

ANALYSIS OF DATA

Table 2. Day 1 (February 26, 2020)

Accuracy 86.91%		n 657	Predicted	
			Positive	Negative
Actual	Positive	0	0	
	Negative	86	571	

ANALYSIS OF DATA

Table 3. Day 2 (February 27, 2020)

Accuracy		n	Predicted	
			Positive	Negative
Actual	Positive	0	0	
	Negative	33	4956	

ANALYSIS OF DATA

Table 4. Day 3 (February 28, 2020)

Accuracy		n	Predicted	
			Positive	Negative
Actual	Positive	13	3	
	Negative	93	4787	

Figure 1. Website Ratings

