

# From Classroom to Community: Five Years of TinyML Academic Network Impact.

Marco Zennaro

The Abdus Salam International Centre for Theoretical Physics

[mzennaro@ictp.it](mailto:mzennaro@ictp.it)

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# TinyML4D: why?



ICT4D researchers claim that there are four technological requirements for an ICT4D project to be successful:

**Autonomous Connectivity**



**Low-cost equipment**



**Power resilience**



**Appropriate User Interface**



Brewer, Eric, et al. "The case for technology in developing regions." *Computer* 38.6 (2005): 25-38.



**SciTinyML: Scientific Use of Machine Learning on Low-Power Devices**

18-22 October 2021  
An ICTP Virtual Meeting  
Trieste, Italy

**Description:**  
TinyML enables low-cost learning technologies to perform on-device analysis of sensor data of extremely low power. This allows for new scientific applications to be developed on an extremely low cost and of large scale.

**Topics:**  
- In-situ sensor data  
- Sensor data analysis  
- Sensor data storage  
- Sensor data processing  
- Sensor data visualization

**Local Organiser:**  
Prof. Dr. Roberto Ruffini

**Registration:**  
Free of charge

**African Regional Workshop on SciTinyML: Scientific Use of Machine Learning on Low-Power Devices**

24-29 April 2022  
Online

**Description:**  
TinyML is a subset of Machine Learning focused on developing models that can be executed on small, real-time, low-power, and low-cost embedded devices. This allows for new scientific applications to be developed on an extremely low cost and of large scale.

**Topics:**  
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**Asian Regional Workshop on SciTinyML: Scientific Use of Machine Learning on Low-Power Devices**

4-10 June 2022  
Online

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**Registration:**  
Free of charge

**Latin American Regional Workshop on SciTinyML: Scientific Use of Machine Learning on Low-Power Devices**

13-19 July 2022  
An ICTP Virtual Meeting  
Trieste, Italy

**Description:**  
TinyML is a subset of Machine Learning focused on developing models that can be executed on small, real-time, low-power, and low-cost embedded devices. This allows for new scientific applications to be developed on an extremely low cost and of large scale.

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**Registration:**  
Free of charge

**Workshop on Scientific Use of Machine Learning on Low-Power Devices: Applications and Advanced Topics**

17-21 April 2023  
An ICTP Virtual Meeting  
Trieste, Italy

**Description:**  
TinyML is a subset of Machine Learning focused on developing models that can be executed on small, real-time, low-power, and low-cost embedded devices. This allows for new scientific applications to be developed on an extremely low cost and of large scale.

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- Sensor data processing  
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**Local Organiser:**  
Prof. Dr. Roberto Ruffini

**Registration:**  
Free of charge

**ICTP-UNU Workshop on TinyML for Sustainable Development**

AN ICTP 60<sup>th</sup> ANNIVERSARY SATELLITE EVENT

**Description:**  
TinyML is a new technology that allows machine learning models to run on low-cost, low-power microcontrollers. This technology has a significant impact on the Sustainable Development Goals (SDGs) and in facilitating scientific research in areas such as environmental monitoring and the physics of complex systems.

**Topics:**  
- In-situ sensor data  
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- Sensor data storage  
- Sensor data processing  
- Sensor data visualization

**Local Organiser:**  
Prof. Dr. Roberto Ruffini

**Registration:**  
Free of charge

**Workshop on Machine Learning on Low-Power Devices: Applications and Advanced Topics**

4-10 July 2024  
Online

**Description:**  
TinyML empowers machine learning technologies on low-cost, low-power microcontrollers. This technology has a significant impact on the Sustainable Development Goals (SDGs) and in facilitating scientific research in areas such as environmental monitoring and the physics of complex systems.

**Topics:**  
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- Sensor data processing  
- Sensor data visualization

**Local Organiser:**  
Prof. Dr. Roberto Ruffini

**Registration:**  
Free of charge

**Workshop on TinyML for Sustainable Development**

10-16 July 2024  
Site Specific, Brazil

**Description:**  
TinyML empowers machine learning technologies on low-cost, low-power microcontrollers. This technology has a significant impact on the Sustainable Development Goals (SDGs) and in facilitating scientific research in areas such as environmental monitoring and the physics of complex systems.

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**Local Organiser:**  
Prof. Dr. Roberto Ruffini

**Registration:**  
Free of charge

More than 2000 people trained!

## Design and Development of a Family of Integrated Devices to Monitor Animal Movement in the Wild

by Laila Daniela Kazimierski<sup>1,\*,†</sup>, Andrés Oliva Trevisan<sup>2,3,†</sup>, Erika Kubisch<sup>4</sup>, Karina Laneri<sup>1</sup> and Nicolás Catalano<sup>2,3</sup>

<sup>1</sup> Consejo Nacional de Investigaciones Científicas y Técnicas, Centro Atómico Bariloche (CONICET), Comisión Nacional de Energía Atómica (CNEA), San Carlos de Bariloche R8402AGP, Argentina

<sup>2</sup> Centro Atómico Bariloche, Comisión Nacional de Energía Atómica (CNEA), San Carlos de Bariloche R8402AGP, Argentina

<sup>3</sup> Instituto Balseiro, Universidad Nacional de Cuyo, Comisión Nacional de Energía Atómica (CNEA), San Carlos de Bariloche R8402AGP, Argentina

<sup>4</sup> Instituto de Investigaciones en Biodiversidad y Medioambiente, Consejo Nacional de Investigaciones Científicas y Técnicas (INIBIOMA, CONICET-Universidad Nacional del Comahue), San Carlos de Bariloche R8400AGP, Argentina

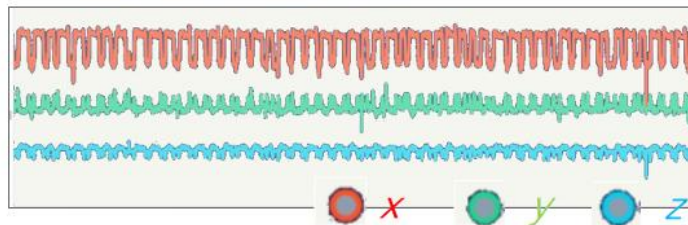
\* Author to whom correspondence should be addressed.

† These authors contributed equally to this work.

<https://www.mdpi.com/1424-8220/23/7/3684>



Example of  
accelerometer  
signal of a female  
digging a nest to  
lay eggs:



## Classifying mosquito wingbeat sound using TinyML

Moez Altayeb  
University of Khartoum, Sudan  
ICTP, Trieste, Italy  
mohedahmed@hotmail.com

Marcelo Rovai  
Universidade Federal de Itajubá  
Itajubá, Brazil  
rovai@unifei.edu.br

Marco Zennaro  
ICTP  
Trieste, Italy  
mzennaro@ictp.it

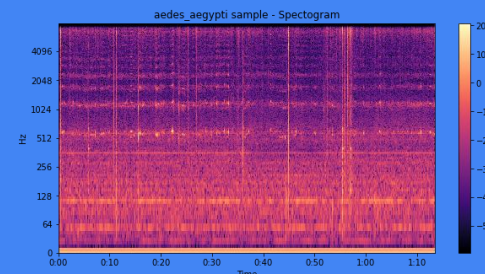
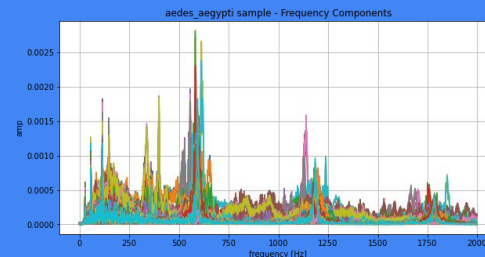
### ABSTRACT

Every year more than one billion people are infected and more than one million people die from vector-borne diseases including malaria, dengue, Zika and Chikungunya. Mosquitoes are the best known disease vector and are geographically spread worldwide. It is important to raise awareness of mosquito proliferation by monitoring their incidence, especially in poor regions. Acoustic detection of mosquitoes has been studied for long and ML can be used to automatically identify mosquito species by their wingbeat. We present a prototype solution based on an openly available dataset, on the Edge Impulse platform and on three commercially-available TinyML devices. The proposed solution is low-power, low-cost and can run without human intervention in resource-constrained areas. This insect monitoring system can reach a global scale.

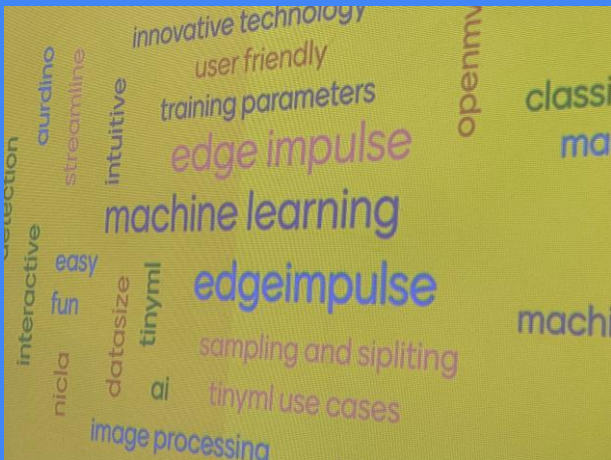
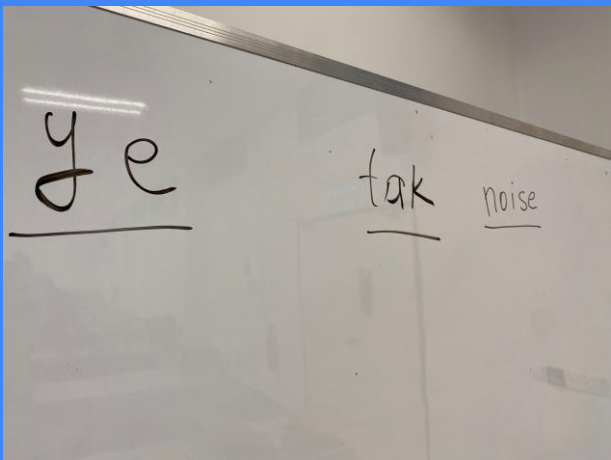
affected. People from poor communities with little access to health care and clean water sources are also at risk. Although anti-malarial drugs exist, there's currently no malaria vaccine.

Vector-borne diseases also exacerbate poverty. Illness prevents people from working and supporting themselves and their families, impeding economic development. Countries with intensive malaria have much lower income levels than those that don't have malaria.

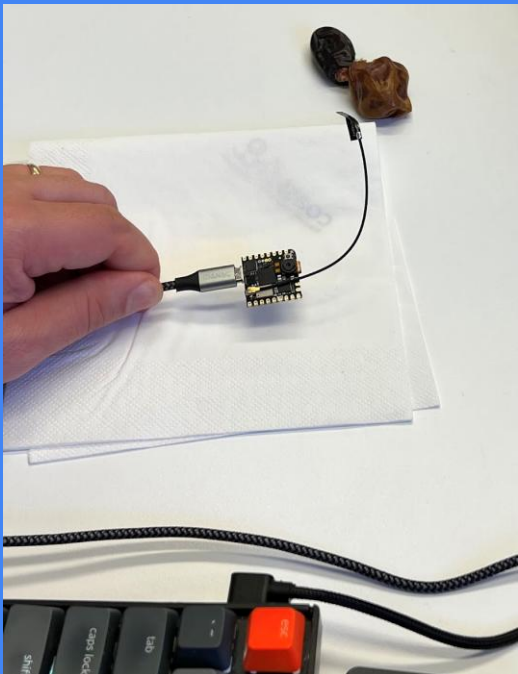
Countries affected by malaria turn to control rather than eradication. Vector control means decreasing contact between humans and disease carriers on an area-by-area basis. It is therefore crucial to be able to detect the presence of mosquitoes in a specific area. This paper presents an approach based on TinyML and on low power embedded devices.





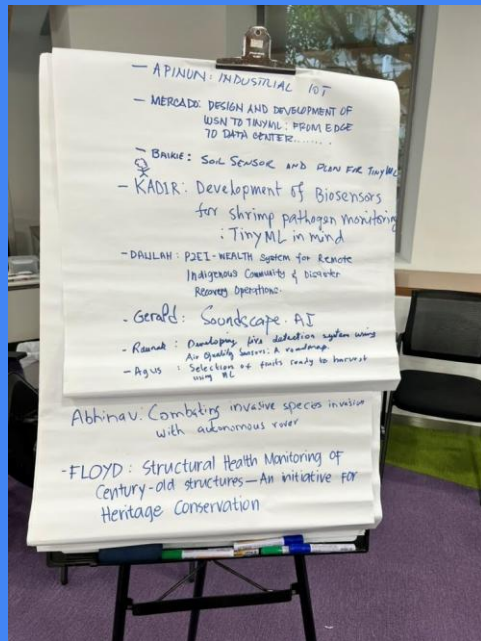


Malaysia 2023



KAUST 2023

# Macau 2024





# Brazil 2024



# Malawi 2025



# Policy recommendations for TinyML

## **1. Invest in Local Capacity Building for Embedded AI**

Support national and regional programs that develop local expertise in embedded AI through universities, vocational training, and continuing education. Prioritize funding for curricula co-developed with global partners like the TinyML Academic Network.

## **2. Promote Open, Low-Cost Infrastructure for TinyML Deployment**

Subsidize access to open-source TinyML tools and low-power hardware for use in schools, research centers, and community projects. This lowers entry barriers and stimulates local innovation for SDG-related challenges.

# Policy recommendations for TinyML

## **3. Integrate TinyML into National Digital and Innovation Strategies**

Recognize TinyML as a key enabler of Edge AI for agriculture, health, climate monitoring, and education. Include TinyML in broader policies on digital transformation, AI governance, and smart infrastructure.

## **4. Fund Context-Aware Pilot Projects in Key Development Sectors**

Support the scaling of pilot projects that apply TinyML to real-world needs—e.g., disease-vector monitoring, air quality sensing, or precision irrigation. Fund projects that demonstrate impact, replicability, and local ownership.



# Policy recommendations for TinyML

## **5. Facilitate Regional Collaboration and Knowledge Sharing**

Encourage South-South cooperation by funding regional centers of excellence and networks that share training resources, datasets, and deployment best practices, building on the TinyML4D community model.

# Learning and Teaching



[https://github.com/harvard-edge/cs249r\\_book](https://github.com/harvard-edge/cs249r_book)

# TinyML4D on Science!



<https://www.science.org/doi/epdf/10.1126/science.adw7713>

# Thanks!

