

IESTI01 - TinyML

About the Course & Syllabus

Prof. Marcelo Rovai

April 28th, 2021



Dear students,

Welcome to the first class of the IESTI01 (TinyML) course. I am Professor Marcelo Rovai, a former student here at UNIFEI, and I am pleased to be with you this semester.

IESTI01 is a discipline that mixes Machine Learning (part of Artificial Intelligence) with small devices, such as microcontrollers and sensors, whose main characteristics are ultra-low power consumption, 32-bit CPUs, and a few kilobytes of memory.

We understand that the explosive growth of the Machine Learning field, the ease of use of software development platforms such as TensorFlow (TF), based on the Python language and the current generation of powerful microcontrollers, make TinyML an indispensable topic of study for Engineering students in the areas of Electronics, as well as Computing and Control & Automation.

This mix of expertise and the pioneering nature of this type of discipline (we are the second university worldwide to have this type of course, being Harvard School of Engineering the first) leads us to significant challenges concerning the necessary basis for the minimum understanding of the matter. Thus, it is essential to emphasize that the time we have available for the course is short. So, We must commit ourselves to do our utmost in class and with complementary activities such as readings, laboratories, and assignments. You can count on me outside the hours of the weekly classes to clarify doubts and to review the necessary concepts for a good understanding of the course.

Saludos from the south of the world!

Prof. Marcelo Rovai

Who I am

Marcelo José Rovai

- I am a Brazilian from São Paulo, **Data Science Master's degree by UDD, Chile**, and MBA by IBMEC (INSPER), Brazil.
- I graduated in 1982 as an **Engineer from UNIFEI** with Specialization from Poli/USP, both in Brazil.
- I worked as a **teacher, engineer, and executive** in several technology companies such as AVIBRAS Aeroespacial, SID Informática, ATT-GIS, NCR, DELL, COMPAQ (HP), and more recently at IGT, where I continue as a Senior Advisor for Latin America.
- I **write about electronics**, publishing my works in sites as MJRoBot.org (Editor/Writer), Hackster.io (#1 Contributor), Instructables.com, and Medium.com (TDS – Towards Data Science).



Course Overview

What is Tiny Machine Learning (TinyML)?

Tiny machine learning (TinyML) is a **fast-growing field of machine learning** technologies and applications including **algorithms, hardware, and software** capable of performing **on-device sensor data analytics** at **extremely low power consumption**, typically in the mW range and below, and hence enabling a variety of **always-on ML use-cases** on **battery-operated devices**.

Applications: EdgeML & TinyML

Autonomous Car Control



KeyWord Spotting



Environmental Control



Image Recognition



Motion & biometric



Image Spot



Predictive Maintenance



Motion, current, audio and camera

- Industrial
- White goods
- Infrastructure
- Automotive

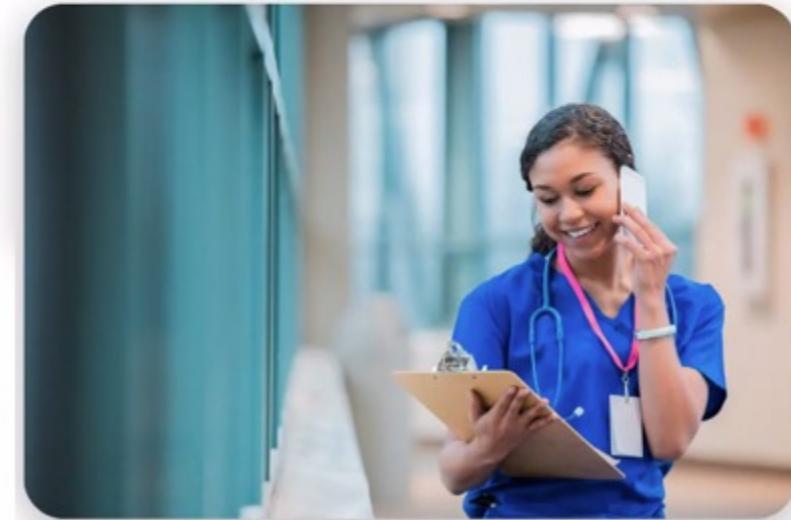
Asset Tracking & Monitoring



Motion, temp, humidity, position, audio and camera

- Logistics
- Infrastructure
- Buildings

Human & Animal Sensing



Motion, radar, audio, PPG, ECG

- Health
- Consumer
- Industrial

ElephantEdge

Building The World's Most Advanced **Wildlife Tracker**.



Source: <https://wildlabs.net/resources/competition/challenge-elephantedge>



EDGE IMPULSE

Microsoft

Western Digital



TAOGLAS

ublox



EARTH RANGER

AV

ElephantEdge

Risk Monitoring

“Know when an elephant is moving into a high-risk area and send real-time notifications to park rangers.”

Conflict Monitoring

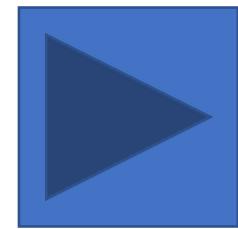
“Sense and alert when an elephant is heading into an area where farmers live.”

Activity Monitoring

“Classify the general behavior of the elephant, such as when it is drinking, eating, sleeping, etc.”

Communication Monitoring

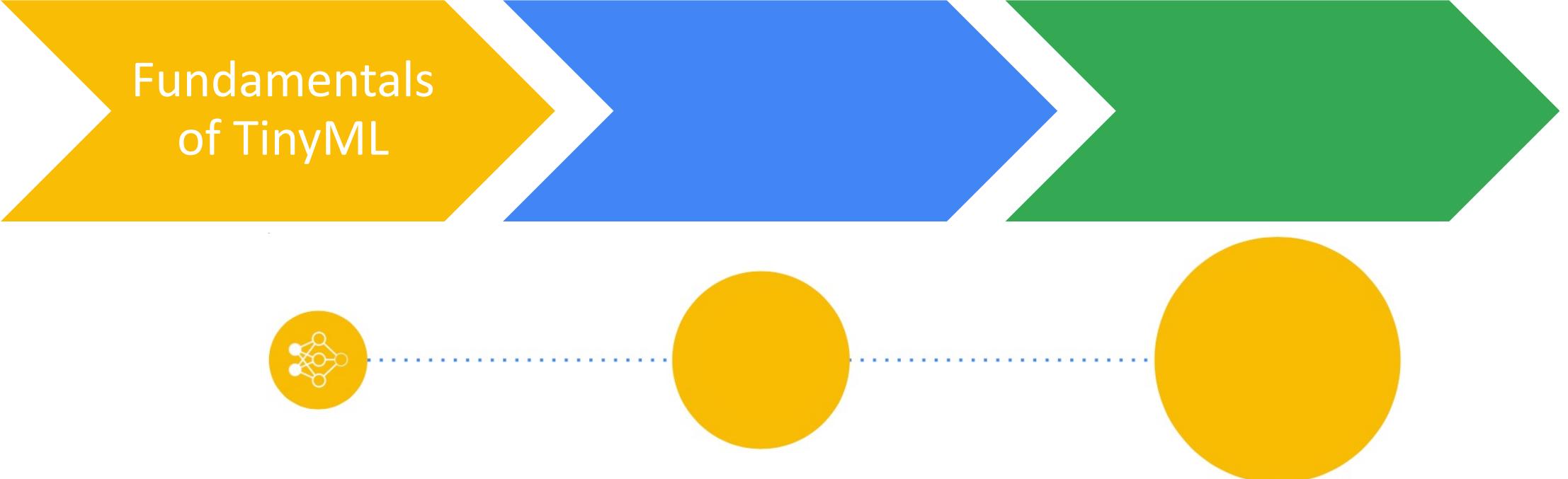
“Listen for vocal communications between elephants via the onboard microphone.”



[ElephantEdge](#)

What we will learn?

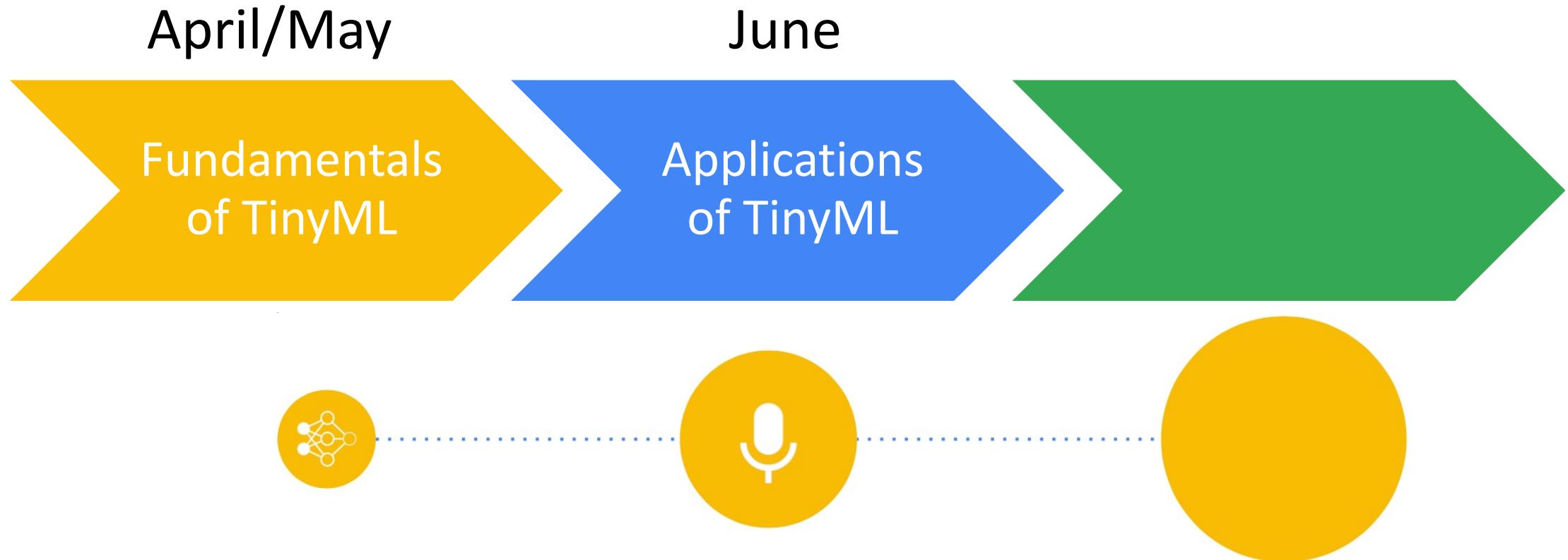
April/May



Fundamentals
of TinyML

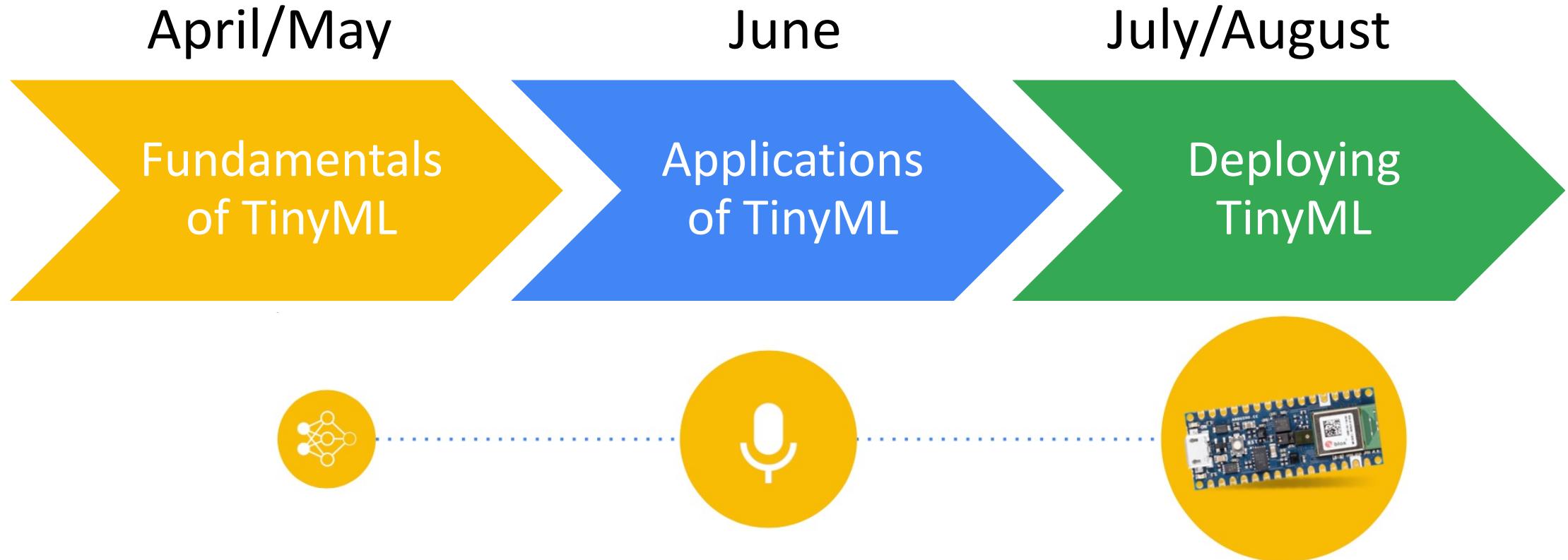
In April and May (Part 1) is all about talking about what is the language of machine learning

What we will learn?



In June (Part 2), We will get a sneak peek into the variety of different TinyML applications, as keyword spotting ("Alexa"), gesture recognition, understand how to leverage the sensors, and so forth.

What we will learn?



In July and August (Part 3), We will learn how to deploy models on a real microcontroller. Along the way we will explore the challenges unique to and amplified by TinyML (e.g., preprocessing, post-processing, dealing with resource constraints).

Background Requirements

Part 1

Fundamentals of TinyML

- Python
- Google Colab
- Jupyter Notebook

Part 2

Applications of TinyML

- Python
- Google Colab
- Jupyter Notebook
- Edge Impulse Studio

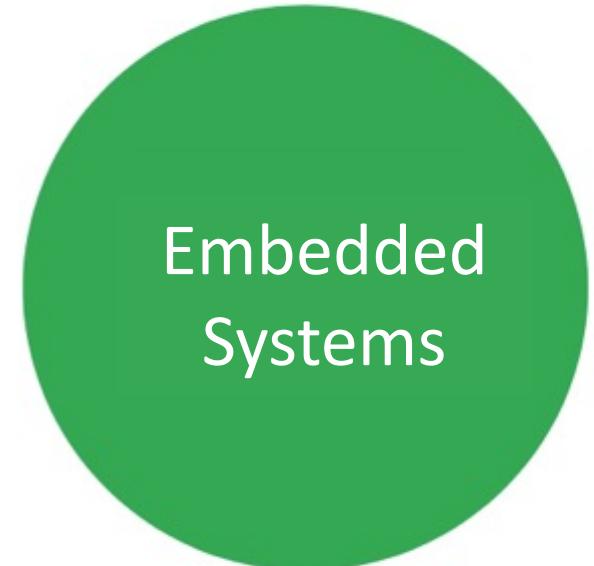
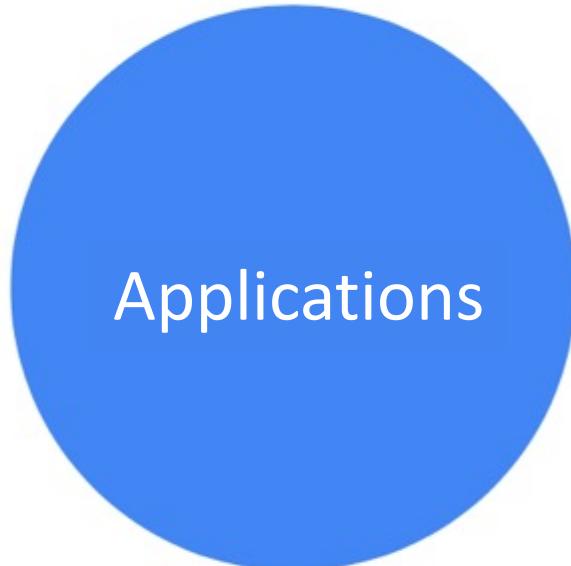
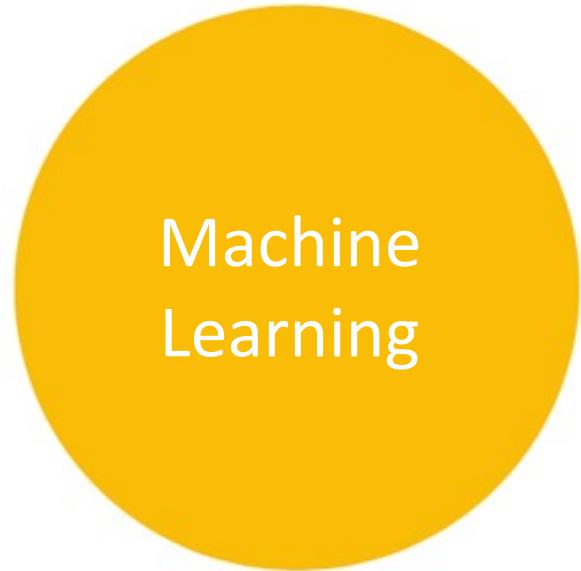
Part 3

Deploying TinyML

- Python
- Google Colab
- Jupyter Notebook
- Edge Impulse Studio
- IDE (as Arduino)
- C/C++

This course combines **computer science** with **engineering** to feature real-world application case studies that examine the challenges facing **TinyML deployments**.

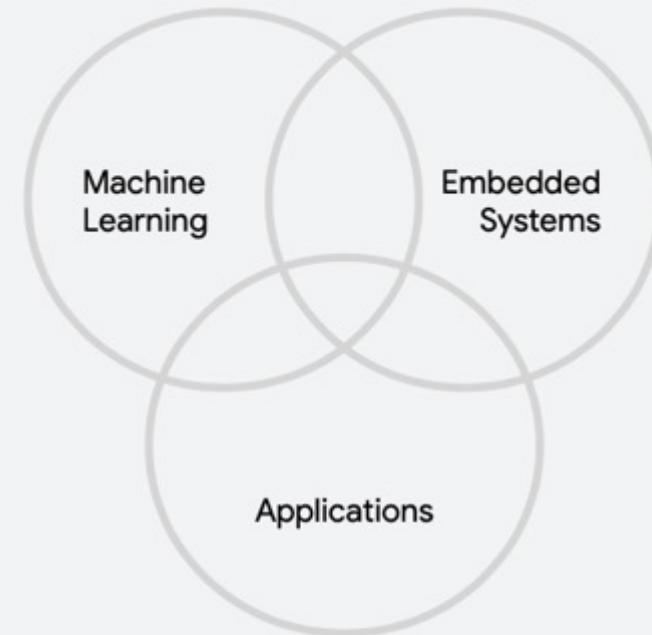
What areas we will learn?



We will learn the **fundamentals of each of these areas**, just enough to focus on the ultimate goal of being able to build TinyML applications.

Interactions

In addition, we will bring these diverse topics together to reveal the interesting learnings at the various **intersections**

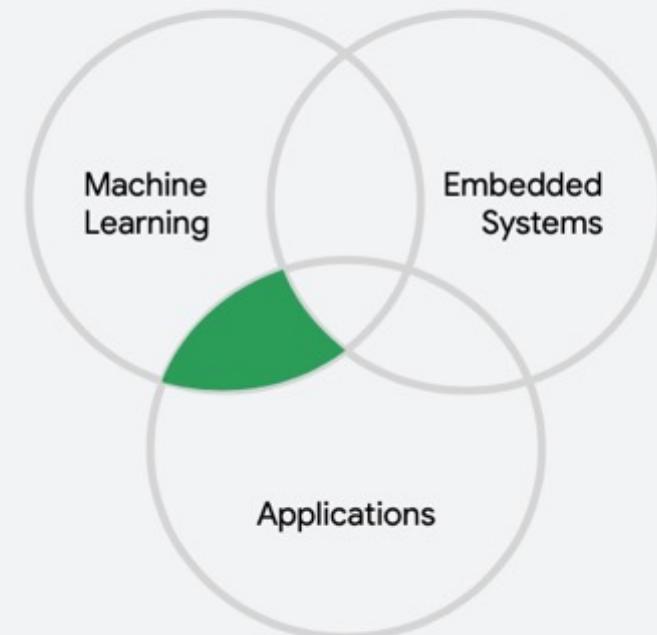


Interactions

How machine learning can
enable new and interesting
TinyML applications?

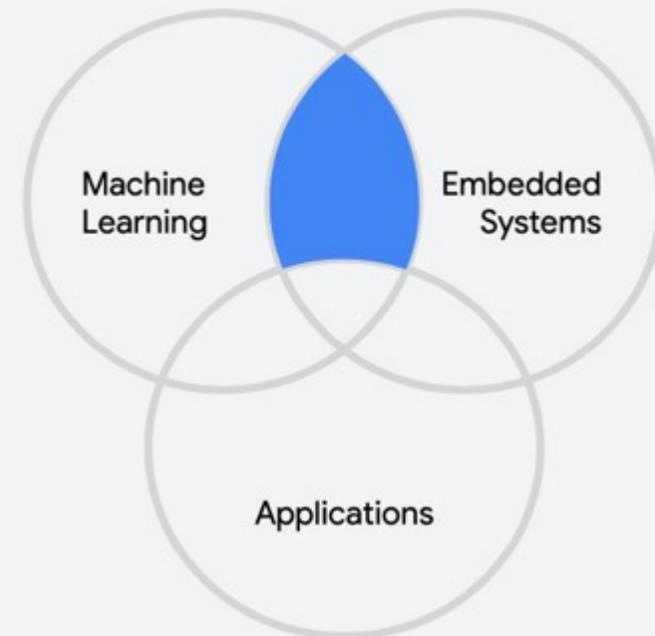


Source: <https://wildlabs.net/resources/competition/challenge-elephantedge>



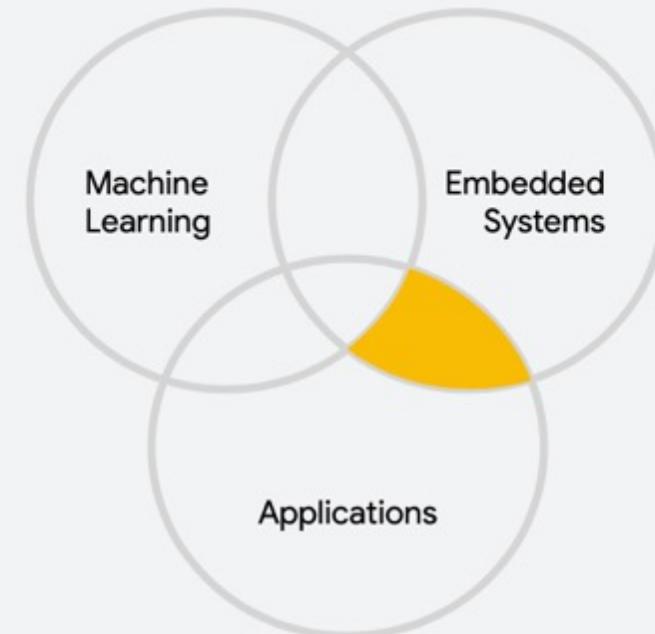
Interactions

What are the **challenges** with
enabling **machine learning**
on **tiny**, resource-constrained
embedded devices?

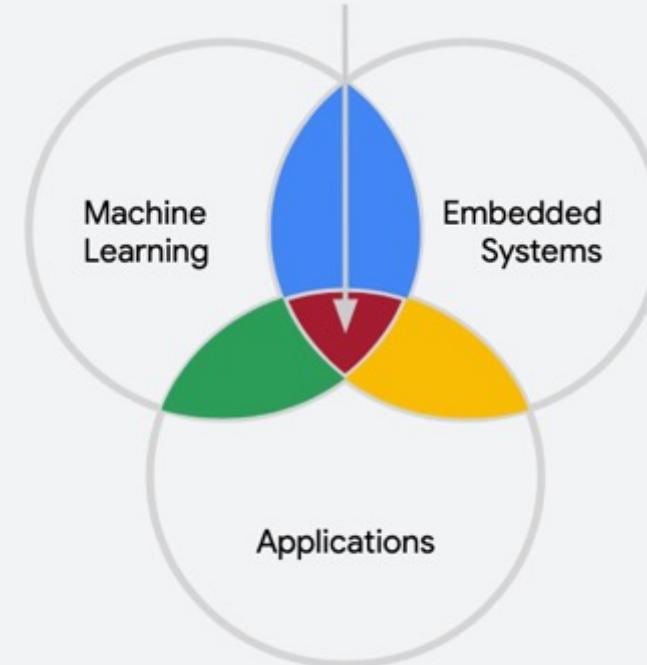


Interactions

What type of new **use cases** can we possibly enable on **embedded systems** that we could not otherwise do before?



Applying TinyML



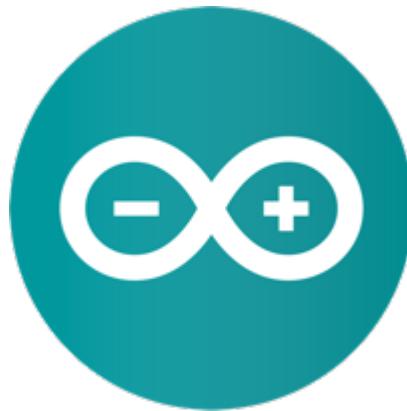
At the End of the Day

Given your understanding of things at these various intersections, you will have a deep understanding for **how to apply TinyML**

How are we going to get there?

Hands-on Learning

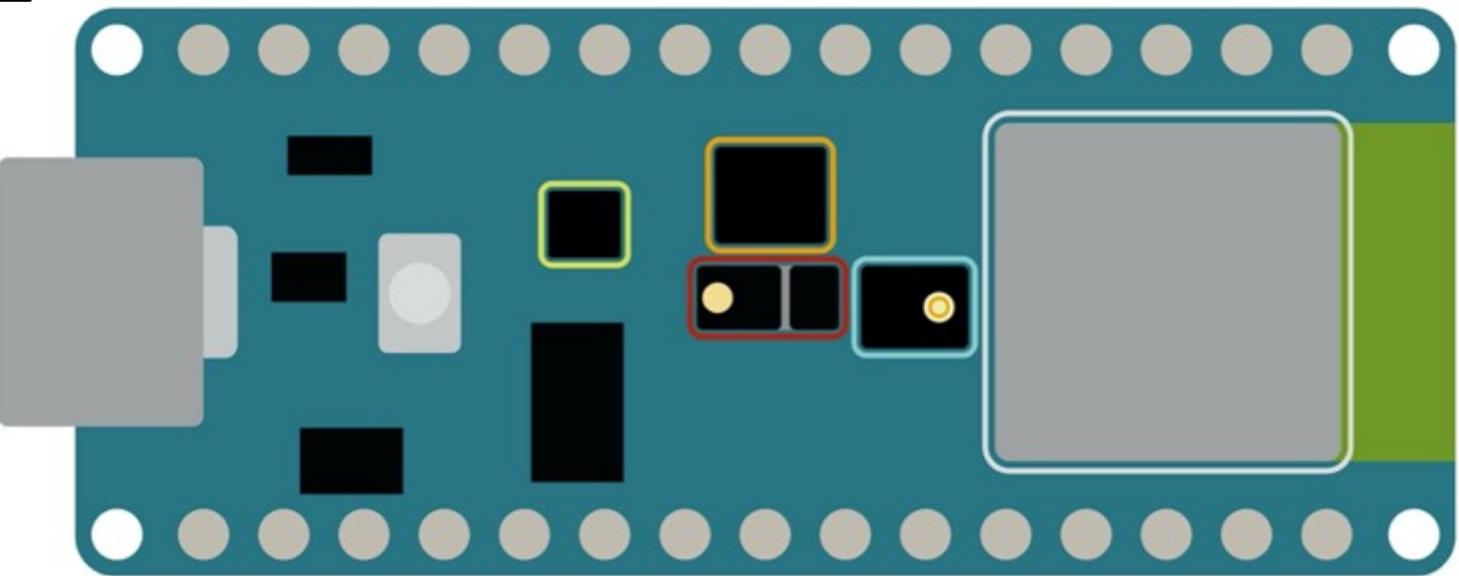
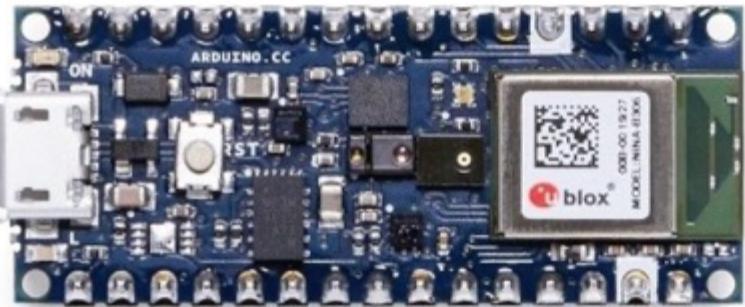
- Software
 - Machine Learning
(TensorFlow)
 - Programming environments
(Jupyter and Colab)
 - Edge Impulse Studio
- Hardware (Part 3)
 - Arduino 33BLE Sense
 - Sensors



TinyML Kit



Nano 33 BLE SENSE



- ◆ Color, brightness, proximity and gesture sensor
- ◆ Digital microphone
- ◆ Motion, vibration and orientation sensor
- ◆ Temperature, humidity and pressure sensor
- ◆ Arm Cortex-M4 microcontroller and BLE module

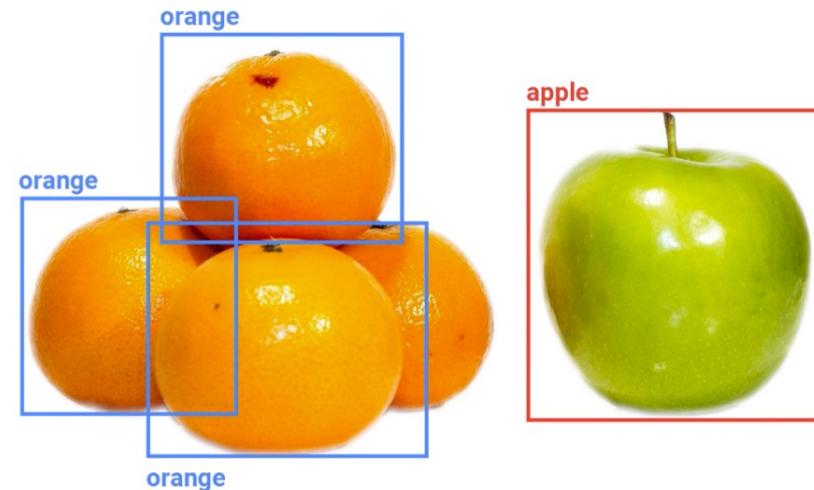
Hands-on Activity (Part 1 & 2)

Speech

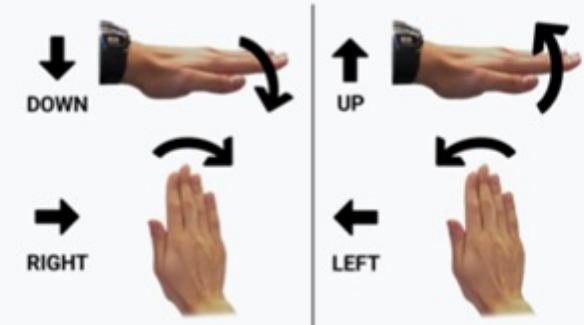


Okay, Google.

Vision



IMU



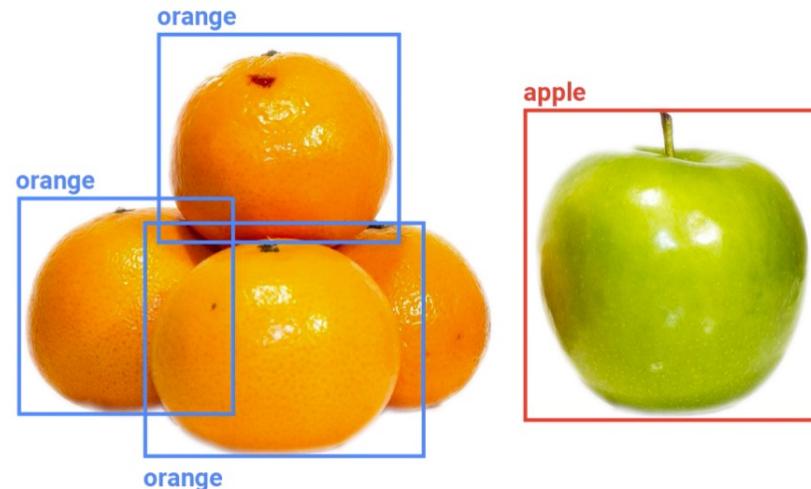
Hands-on Activity (Part 3 – On-Device)

Speech

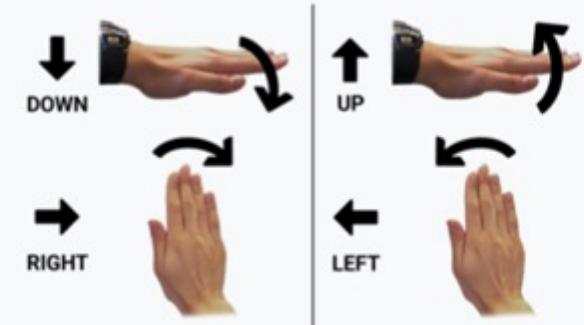


Okay, Google.

Vision



IMU



How is the course structured?

Course Structure

- Weekly live lectures
 - Slides
 - Hands-on coding (by teacher & students)
- Readings
- Assignments
 - Quizzes
 - Notebooks with codes
 - Hands-on lab reports
- Final Project (in group)
 - Report
 - Presentation

UNIFEI-ESTI01-T01-2021.1

Course Repository - TinyML - Machine Learning for Embedding Devices



Instituto de Engenharia de Sistemas e Tecnologias da Informação –
ESTI - Campus de Itajubá

NOTA: O curso será ministrado majoritariamente em português, mas uma base de inglês será fundamental para o acompanhamento do mesmo, pois todo o material de leitura, vídeos e slides, serão disponibilizados no idioma inglês.

Visão geral do curso

TinyML é um curso introdutório na interseção entre o Aprendizado de Máquina (Machine Learning) e dispositivos embarcados (Embedded Devices). A difusão de dispositivos embarcados com ultra-baixo

<https://github.com/Mjrovai/UNIFEI-ESTI01-T01-2021.1>

Class schedule and assignment deadlines

| | | | | |
|--------------|----|----------|------------------------------------------------------------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Fundamentals | 1 | 28-04-21 | About the Course and Syllabus | Coding Exercises List 1 (2 weeks to delivery) Deadline delivery List 1 Coding Exercises List 2 (2 weeks to delivery) Deadline delivery List 2 Coding Exercises List 3 (2 weeks to delivery) Deadline delivery List 2 Coding Exercises List 4 (2 weeks to delivery) Deadline delivery List 3 |
| | 2 | 28-04-21 | Introduction to TinyML | |
| | 3 | 05-05-21 | TinyML - Challenges | |
| | 4 | 05-05-21 | Jupyter Notebook, CoLab and Python Review | |
| | 5 | 12-05-21 | The Machine Learning Paradigm | |
| | 6 | 12-05-21 | The Building Blocks of Deep Learning | |
| | 7 | 19-05-21 | Exploring Machine Learning Scenarios | |
| | 8 | 19-05-21 | Building a Computer Vision Model | |
| | 9 | 26-05-21 | Responsible AI Design | |
| | 10 | 26-05-21 | TinyML Fundamentals Summary - Group Project rules | |
| Applications | 11 | 02-06-21 | Preview of TinyML Applications | Deadline delivery List 4 Deadline delivery Group Project Proposal Deadline delivery KWS Lab Report |
| | 12 | 02-06-21 | AI Lifecycle and ML Workflow | |
| | 13 | 09-06-21 | Machine Learning on Mobile and Edge IoT Devices | |
| | 14 | 09-06-21 | Introduction to Edge Impulse Studio | |
| | 15 | 16-06-21 | Keyword Spotting | |
| | 16 | 16-06-21 | Lab KWS | |
| | 17 | 23-06-21 | Visual Wake Words | |
| | 18 | 23-06-21 | Lab VWW | |
| | 19 | 30-06-21 | Anomaly Detection | |
| | 20 | 30-06-21 | Lab Gesture Classification and Anomaly Detection | |
| Deploying | 21 | 07-07-21 | Deploying TinyML Applications on Embedded Devices | Deadline delivery VWW Lab Report Deadline delivery Anomaly Detection Lab Report Deadline delivery Project Report and Presentation |
| | 22 | 07-07-21 | Demo/Lab | |
| | 23 | 14-07-21 | Collecting a Custom TinyML Dataset | |
| | 24 | 14-07-21 | Demo/Lab | |
| | 25 | 21-07-21 | Pre and Post Processing for Keyword Spotting, Visual Wake Words, and Gesture Recognition | |
| | 26 | 21-07-21 | Demo/Lab | |
| | 27 | 28-07-21 | Profiling and Optimization of TinyML Applications | |
| | 28 | 28-07-21 | Responsible AI | |
| | 29 | 04-08-21 | Group Presentations | |
| | 30 | 04-08-21 | Group Presentations | |

Class planning and approval process

- Minimal suggested Workload (4 hours per week):
 - 30 hours (Weekly online classes of 2h, for 15 weeks)
 - 15 hours of assignments/coding/labs
 - 15 hours in research, individual studies, and final project (in a group)
- Approval process:
 - 1st Evaluation:
 - Individual **Quizzes**: 10%
 - Individual **Exercise Lists** (Notebooks): 25%
 - Group **Project Proposal**: 15%
 - 2nd Evaluation
 - Individual **Quizzes**: 10%
 - Individual **Practical Projects** (Lab reports): 25%
 - Group **Project Presentation (*)** and Final Report: 15%

[\(*\) Example: Harvard CS249r: Tiny Machine Learning – Final Projects](#)

Main references

- [Harvard School of Engineering and Applied Sciences - CS249r: Tiny Machine Learning](#)
- [Professional Certificate in Tiny Machine Learning \(TinyML\) – edX/Harvard](#)
- [Introduction to Embedded Machine Learning \(Coursera\)](#)
- [Text Book: "TinyML" by Pete Warden, Daniel Situnayake](#)

I want to thank Harvard professor [Vijay Janapa Reddi](#), Ph.D. student [Brian Plancher](#) and their staff for putting together a great material regarding TinyML that is the base of this course at UNIFEI.

Thanks
And stay safe!

