# tinyML. EMEA

Enabling Ultra-low Power Machine Learning at the Edge

June 26 - 28, 2023



# arm





# Powering Machine Learning Applications on Arm with algae

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28/06/2023



# Generating electricity from photosynthesis

# Part 1

Paolo Bombelli, Research associate @ Dept. Biochemistry, University of Cambridge

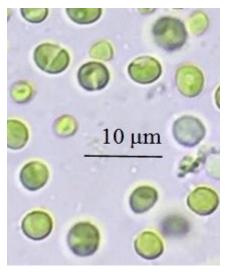


### **Plants:**

Grow with water and sun and, for this reason, they are called autotrophic

In simple terms, algae are aquatic plants







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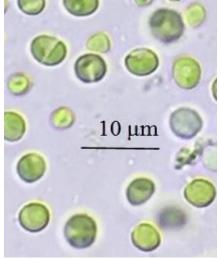
### Algae:

Large group of autotrophic organisms predominantly aquatic. Algae range in size from unicellular microscopic (e.g., Chlorella, ca. 5µm) to giant multicellular (e.g., Kelp, up to 60m)

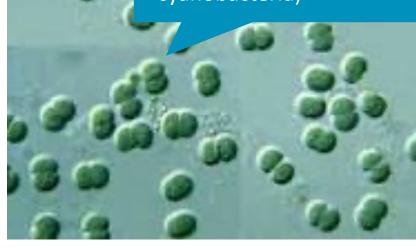
In simple terms, algae are **aquatic plants** 

In a simple terms, they are "old" algae (algae evolved from cyanobacteria)









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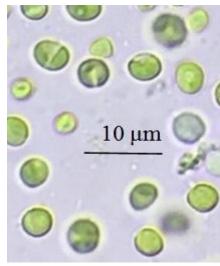
### **Cyanobacteria**:

Large group of unicellular, autotrophic organisms with microscopic size (e.g., Synechocystis ca.2µm ). They resemble the algae in many ways including morphology and ecological niches

In simple terms, algae are aquatic plants

In a simple terms, they are "old" algae (algae evolved from cyanobacteria)









**Plants:** 

Algae:

**Cyanobacteria:** 

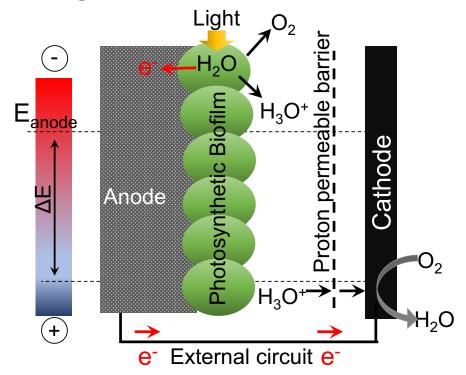
# They are all able to generate electrons from water (i.e., water-photo-lysis)

Water 
$$H_2O$$
 Light Electrons Protons Oxygen  $e^- + H^+ + O_2$ 

Can we generate electricity from photosynthesis?

# Can we generate electricity from photosynthesis?

Photosynthetic microorganisms (e.g., micro-algae) are able to generate electrons that can be harvested by a suitable electrochemical setup and be used as a source of electrical current. This concept forms the basis of Bio Photo Voltaic (BPV) devices<sup>1,2</sup>. The main components forming a BPV are shown in the diagram below



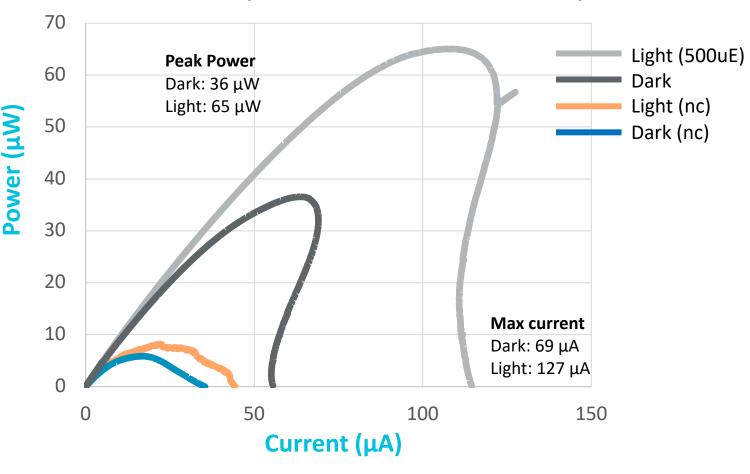
- [1] McCormick et al., (2015), Energy & Environmental Science 8 (4), 1092-1109.
- [2] Howe and Bombelli (2020) Joule 4 (10), 2065-2069

How much current can we generate?

# How much current can we generate?



### Power curve (derived from a I/V curve)



Volumetric power density: dark: ~2 μW / cm<sup>3</sup>; light: ~4 μW / cm<sup>3</sup>

Bombelli et al. (2022) Energy & Envtl Sci.

# BPV's domain of application





# BPV's domain of application

# **Power Consumption**



Micro and small electronic devices

Small domestic appliances

Large domestic appliances and industrial electric devices









mWs







kW

kWs

Powered by disposable and/or removable batteries

Powered by built-in butteries charged through the national grid

Powered by national grid

# BPV's domain of application

# **Power Consumption**



BPV's application tomorrow

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Small domestic appliances

Large domestic appliances and industrial electric devices









mWs







kWs

kW

μWs

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Powered by national grid Example of electronic devices powered using a BPV

# Example of electronic devices powered using a BPV

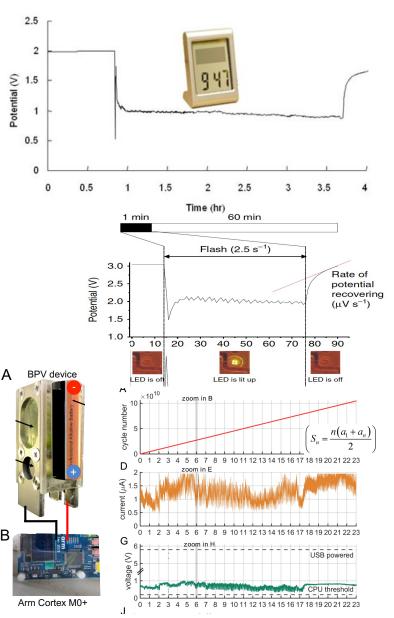
## Power a digital clock



Power an intermittent
LED from digitally printed
cyanobacteria



Power an Arm Cortex-M0+ for > 6 months, in a domestic environment under ambient light





From the journal: Energy & Environmental Science

Photosynthetic biofilms in pure culture harness solar energy in a mediatorless bio-photovoltaic cell (BPV) system†

Alistair J. McCormick,\* <sup>a</sup> Paolo Bombelli, <sup>b</sup> Amanda M. Scott, <sup>b</sup> Alexander J. Philips, <sup>b</sup> Alison G. Smith, <sup>c</sup> Adrian C. Fisher <sup>b</sup> and Christopher J. Howe <sup>a</sup>

### nature communications

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Article | Open Access | Published: 06 November 2017

# Electricity generation from digitally printed cyanobacteria

Marin Sawa, Andrea Fantuzzi, Paolo Bombelli, Christopher J. Howe, Klaus Hellgardt & Peter J. Nixon □

Nature Communications 8, Article number: 1327 (2017) Cite this article

### Energy & Environmental Science



### **PAPER**

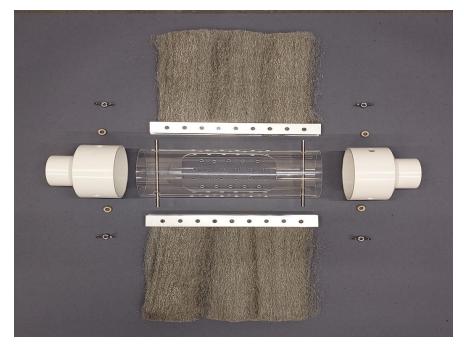


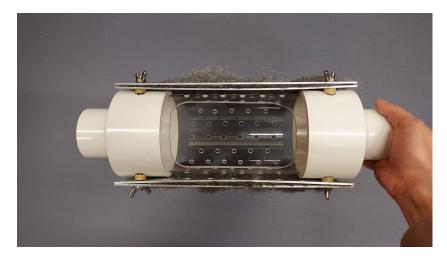
### Powering a microprocessor by photosynthesis†

Cite this: Energy Environ. Sci., 2022, 15, 2529 P. Bombelli, [0] \*ab A. Savanth, \*a A. Scarampi, [0] \*a S. J. L. Rowden, \*a D. H. Green, [0] \*d. Erbe, \*a E. Årstøl, \*l. Jevremovic, [0] \*M. F. Hohmann-Marriott, \*fg S. P. Trasatti, \*b E. Ozer (0] \*\*a and C. J. Howe (0] \*\*a

# The next step: floating-BPV for powering a tinyML application













# Designing an Arm-based microcontroller board from scratch

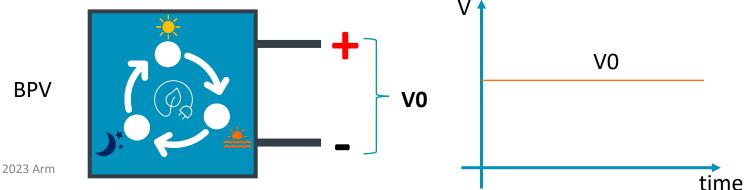
# Part 2

Gian Marco Iodice, Tech lead for ML @ Arm



# When powering tinyML with algae

- → Monitoring environment through many sensors
- Nural areas where a small amount of power might be beneficial to power environmental sensors.
- Provide continuous electricity for longer than traditional batteries
  - It can last years!







# Goal

Demonstrating that people, <u>regardless of their</u> <u>expertise</u>, has the power to address environmental challenges using <u>today's</u> <u>technology</u> in a <u>sustainable</u> and <u>affordable</u> manner.



How

By powering the <u>first ML application</u> with algae on an Arm-based microcontroller



# **Project**

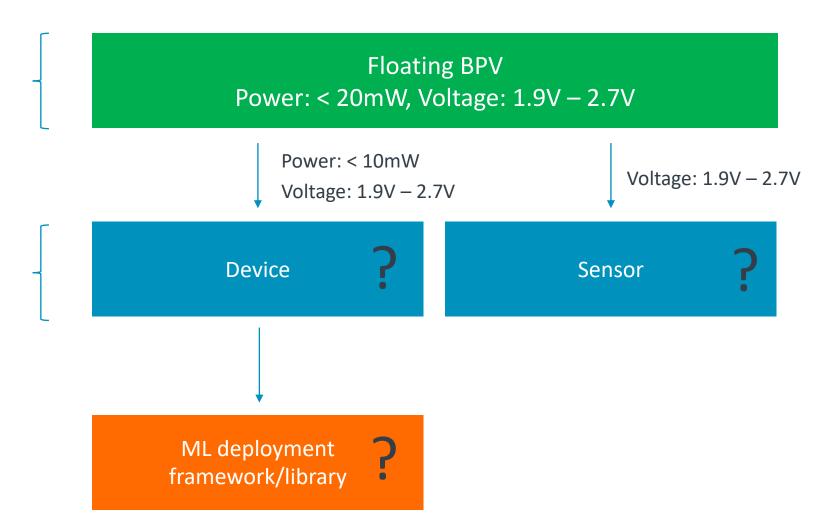
Monitoring weather conditions and water quality with tinyML in the river Thames (London) using native algae of the river.



# Ingredients for a tinyML application

The battery defines the power and voltage constraints

The device selection has implication on the ML framework adoption





# Device selection considerations

- When aiming for a user-friendly development experience, microcontrollers are typically the preferred option.
- The microcontroller must be low-power (less than 10mW) and operating with voltages between 1.9V – 2.7V
  - No floating-point hardware acceleration
  - Small program (32 Kbytes) and data memory (8 Kbytes)
  - Low CPU clock frequency (16 MHz)
  - Only a few external HW components should be used. Mainly sensors.
  - Voltage regulator has to be very efficient



# Today's microcontroller challenges

- Many microcontroller development boards do not operate with voltage below 3.3V
  - Most platforms operates with input voltages between 3.3 5V
- Therefore, we may need to build a custom board to target the specific voltage requirements
  - Nowadays, it can be very easy and fun developing a custom board with a microcontroller, thanks to detailed datasheets and plenty of free supporting online material
  - For prototyping, consider microcontroller that are easy to solder and can fit into breadboards
    - For example, SO8N or DIP footprint



# Sensor selection considerations

• The sensors must be low-power (a few mW) and operating with voltages between 1.9V-2.7V

- Some suitable sensors are:
  - IMU
  - Temperature/humidity
  - Optical flow
  - Light
  - Pressure
  - Gas
  - Turbidity



# Today's sensor challenges

 Many sensor modules do not operate with a voltage below 3.3V

 We may need to design a board from scratch. This may reduce the number of options for easy prototyping



# ML deployment framework/library selection considerations

- It depends on:
  - The target device (microcontroller?)
  - The features of the target device (on-board memory)
- There are many great tools (and often free and/or open source) for microcontrollers.
  - However, we should consider the feasibility of fitting the ML model onto the chosen microcontroller.
  - Frameworks with ahead-of-time (AOT) capabilities may be preferred to reduce program memory usage.
    - The model should NOT be loaded at runtime



### Our solution **~£0.9** single unit **~£1.7** single **~£2.6** single unit or **~£0.5** unit on large STM32G0J6 order STM32G0J6 SparkFun STM32G031J6 SOIC to DIP adapter MCU

Feature	
CPU	Arm Cortex-M0+ @16 MHz
Operating voltage - Power	2 - 3.6V - 93uA/MHz
FLASH - SRAM	32 Kbytes – 8 KBytes
Peripherals	I2C, SPI, UART, USART, Timers, DMA, RTC, ADC, PWM



# Programming and debugging with SWD on Raspberry Pi

# Target Host + debugger probe SWDCLK SWDIO

### **Serial Wire Debug (SWD):**

Arm interface for flashing and debugging an Arm-based microcontroller with only two wires (SWDCLK and SWDIO)

- We can use the SWD IO pins on Raspberry Pi to communicate with the microcontroller
- Raspberry Pi can be used as a standard computer for programming the microcontroller.



STM32G0J6

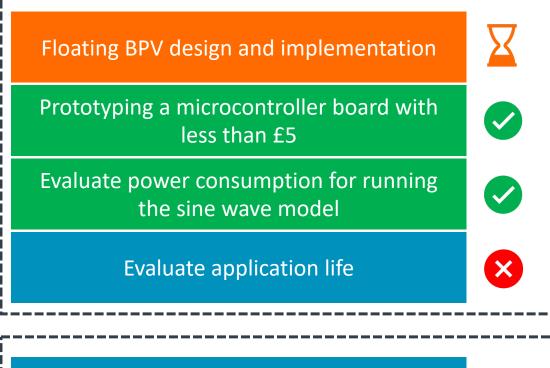
# Designing and deploying the first ML model powered by algae!

- As first model powered by algae, we have considered the sine wave model.
  - This model represents the "Hello, World" for the tinyML community.
  - We express our deepest gratitude to Pete Warden and Daniel Situnayake for their contribution.
- The model was trained and quantized (8-bit) directly in Google Colaboratory
- The tinyML application was written in C++
  - The application has been compiled and uploaded from the Raspberry Pi
- The model has been accelerated on the microcontroller using CMSIS-NN

Stats	Values
Model params	3 Fully Connected layers with a total of 321 params
Program memory	9.7 Kbytes
SRAM	2.1 Kbytes
Current consumption	1.8 mA



# State of the project and some pics!



Phase 1

Expand the microcontroller board with lowpower sensors for water monitoring



Design a tinyML model for weather/water monitoring



Phase 2





# Acknowledgments



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Lorraine Archer
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University of
Cambridge



Omar Al Khatib Engineer for ML, Arm



Howe's lab

### <u>Alphabetical order:</u>

SHUTTLEWORTH





The Leverhulme Trus

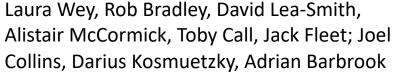








Collaborations with: Jenny Zhang & group, Alasdair Davies (London Zoo), Anand Savanth & Emre Ozer, Tuomas Knowles & group (esp. Akhila Denduluri)







arm Thank You Danke Gracias Grazie 谢谢 ありがとう **Asante** Merci 감사합니다 धन्यवाद Kiitos شکر ً ا ধন্যবাদ תודה

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