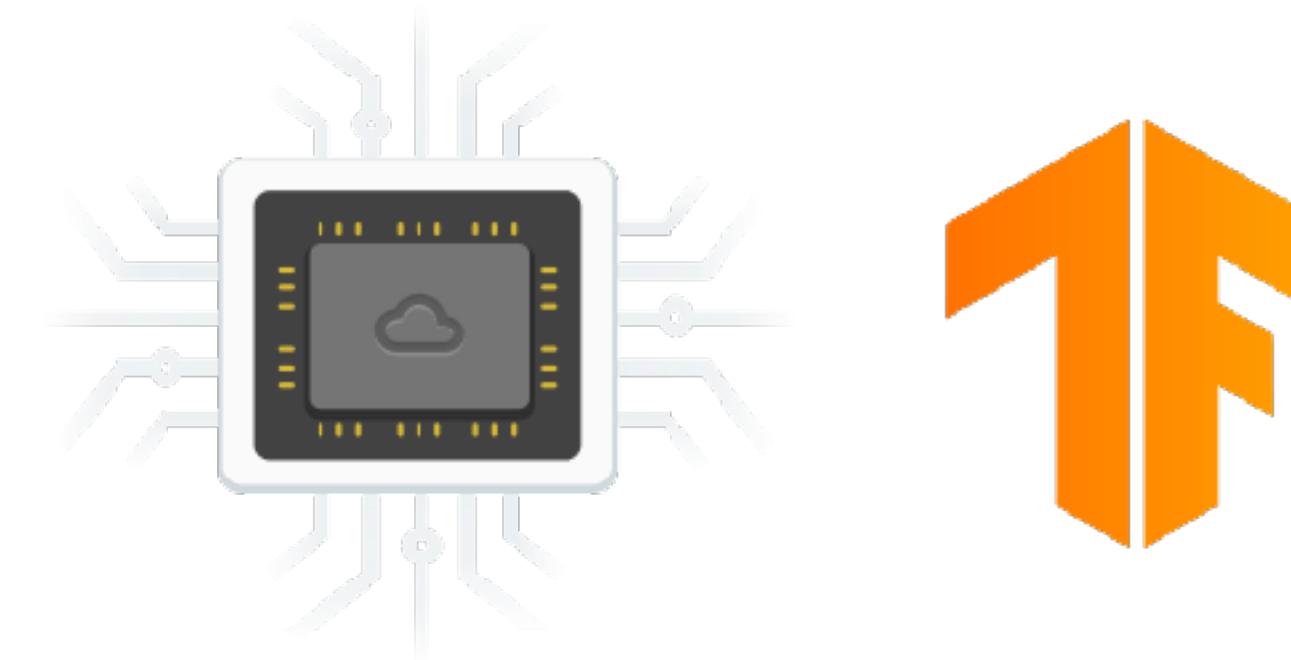


# MACHINE LEARNING EM IOT COM TENSORFLOW



**Alvaro Viebrantz**

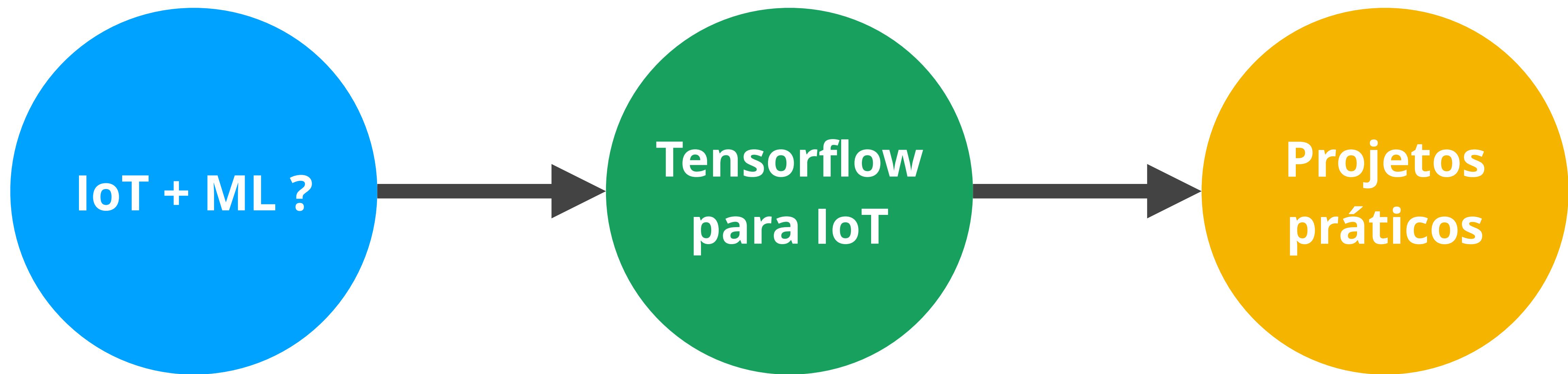
Google Developer Expert for IoT and Product Engineer at Leverege

[aviebrantz.com](http://aviebrantz.com)



[@alvaroviebrantz](https://twitter.com/alvaroviebrantz)

# O que vamos ver hoje ?

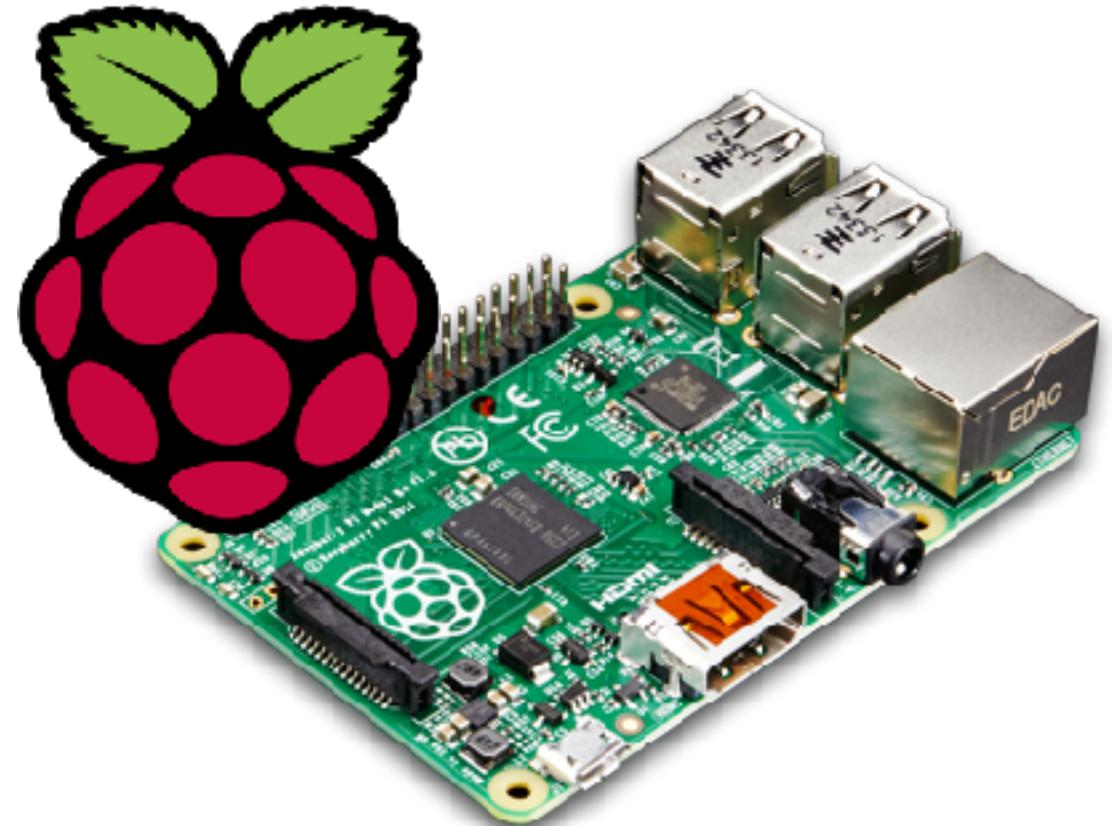


# Internet das Coisas

# Dispositivos IoT

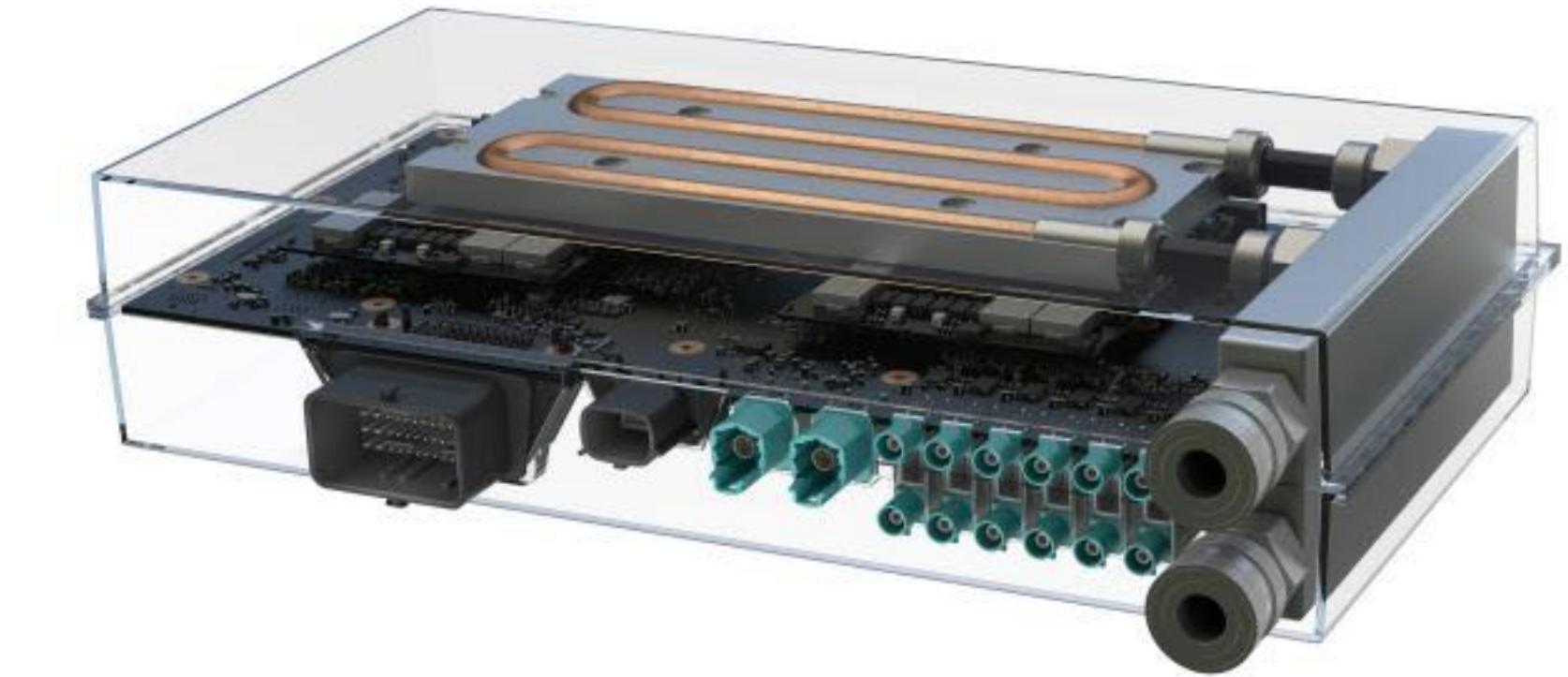
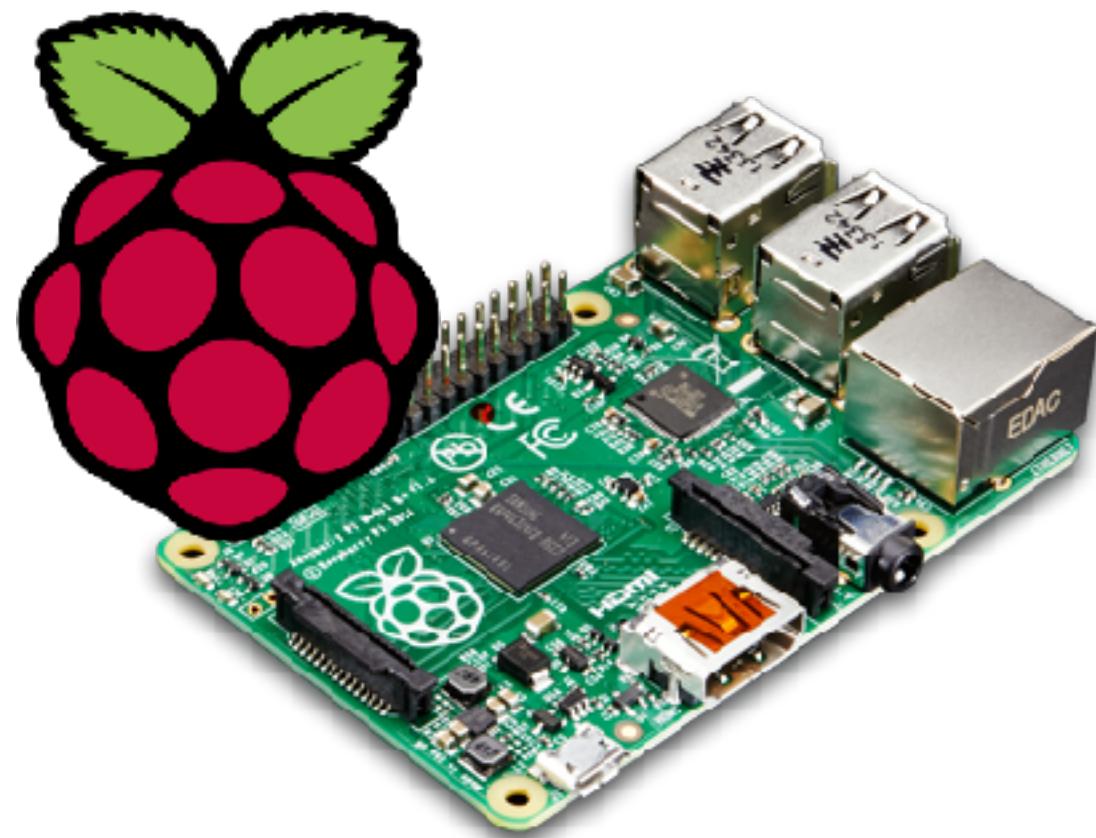
Apesar de contar, na maioria das vezes não estamos falando disso

- Celulares
- Carros inteligentes
- Roteadores/Gateways
- Raspberry Pi's e afins



# Dispositivos IoT

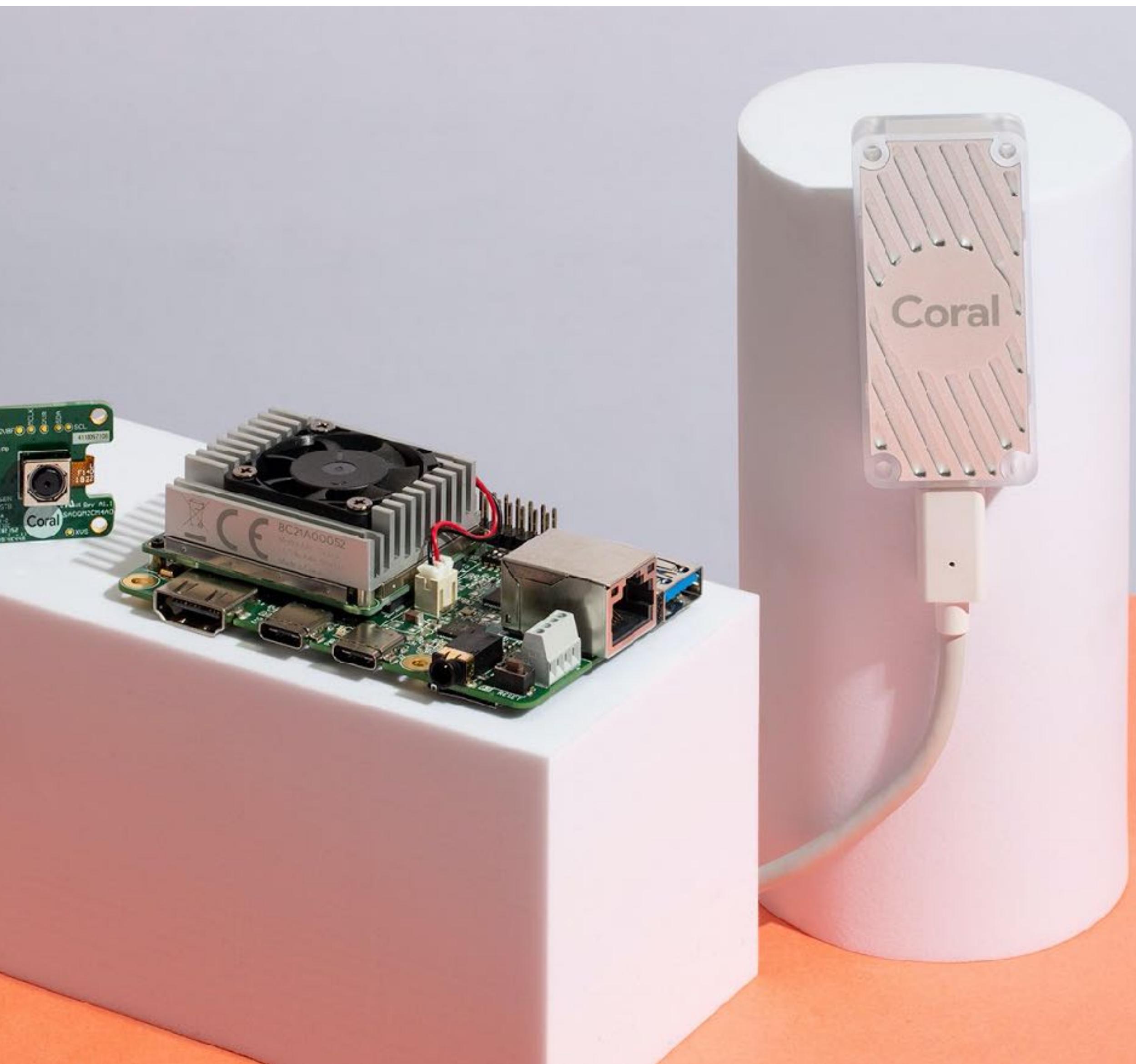
Essa classe de dispositivos são bem mais poderosos



	CPU	Cores	Memory	Storage	Power Consumption	Usage under Load	Price
Raspberry Pi 3	1.2 Ghz	4	1 GB	Up to 32 GB	6W / 400 mA @idle	700 mA @load	\$35
Pixel 4a	2.2 Ghz	8	6 GB	Up to 128 GB	5W / 80mA @idle	900 mA @load	\$350
Google Nest Mini	1.4 Ghz	4	512 MB	4 GB	2W / 150 mA @idle	210 mA @load	\$50
NVIDIA Drive PX2	1.4 Ghz	8 CPU 4 GPU	16 GB	Up to 128 GB	60 W / 16 A @idle	50 A @load	\$800

# Edge Devices

Performance otimizada para Machine Learning



# Dispositivos IoT

Bastante limitados em memoria, cpu e uso de bateria

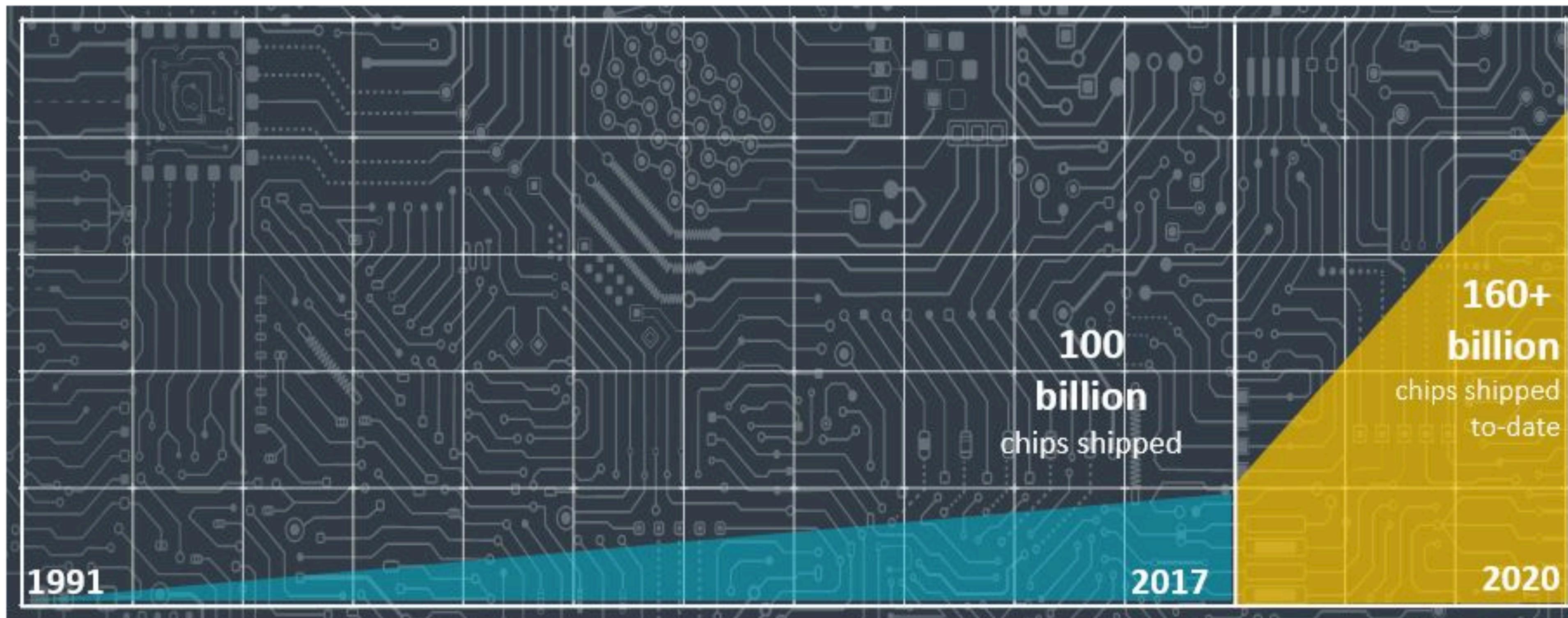
- Dispositivos bem menores e mais baratos
  - Lampadas, Sensores de porta, Fechaduras Smart, Termostatos
  - Smartwatches e Fitness bands
  - Rastreadores em geral - GPS, rede celular, LoRa, etc
  - Assistentes pessoais - Alexa/Google Assistant
  - Na maioria das vezes, não tem sistema operacional nenhum



In the fourth quarter of 2019 (Arm FY Q319):

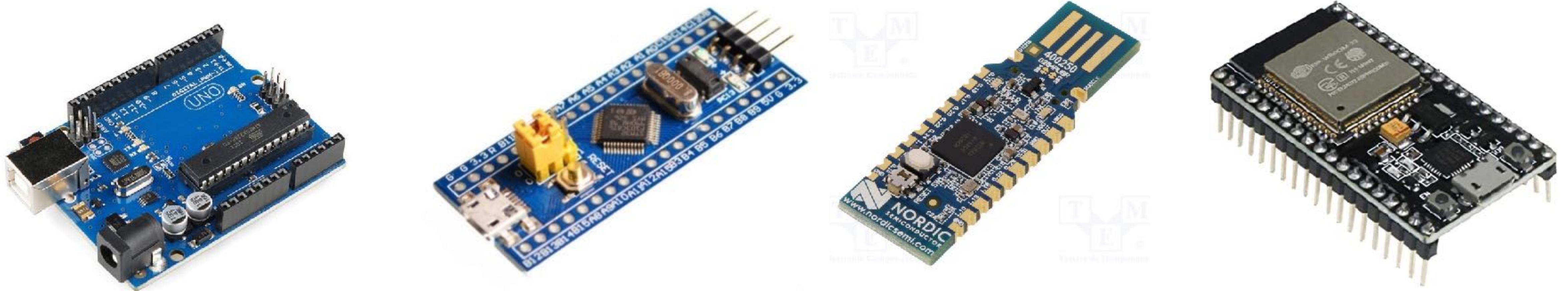
- Arm silicon partners shipped **a record 6.4 billion Arm-based chips**, the third record quarter for unit shipments in the past two years
- Arm saw growing demand for embedded intelligence in endpoint devices as demonstrated by the **record 4.2 billion Cortex-M processors** shipped

To-date, Arm partners have shipped more than **160 billion Arm-based chips**, and an average of more than **22 billion** over the **past three years**.



# Dispositivos IoT

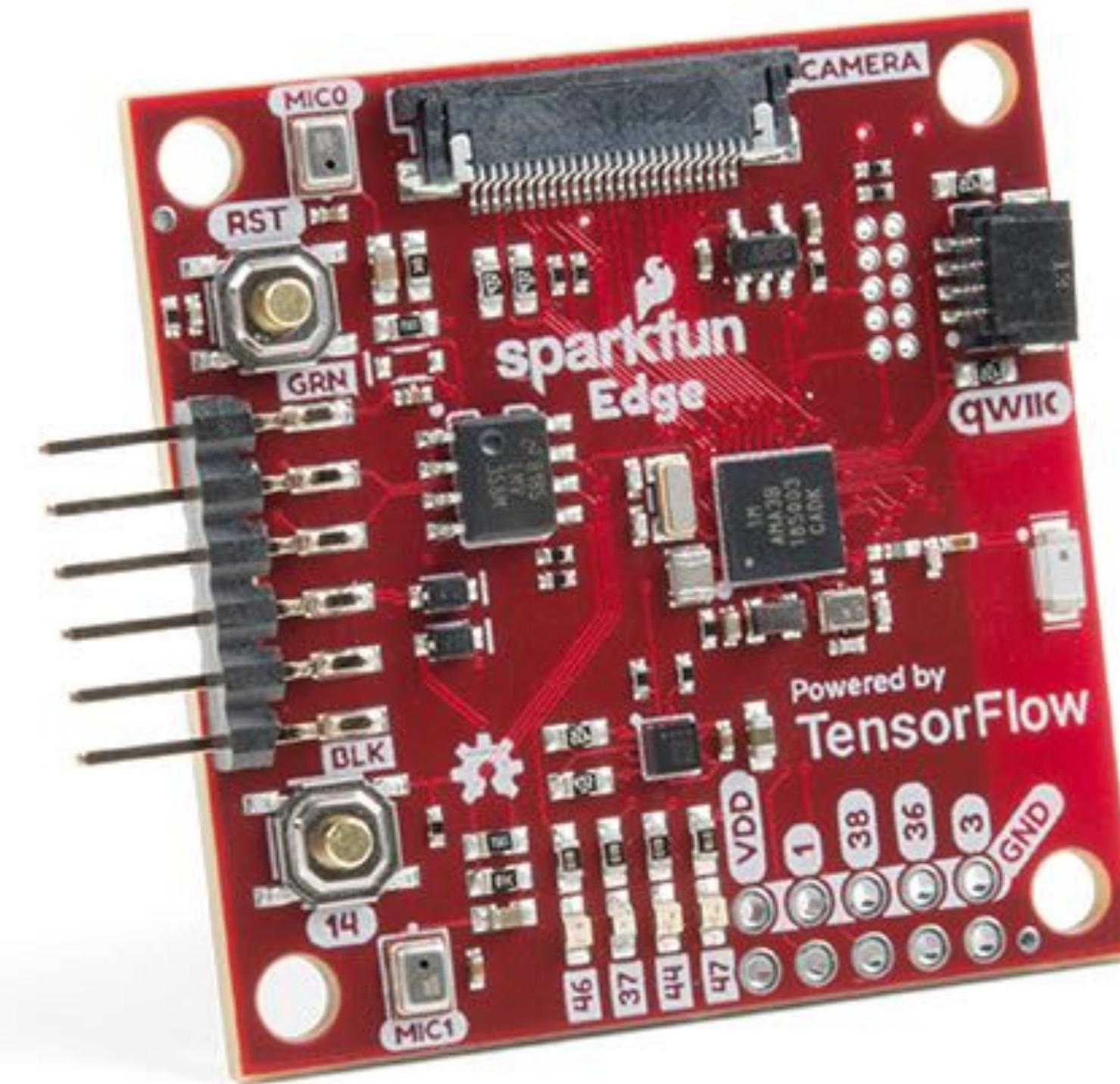
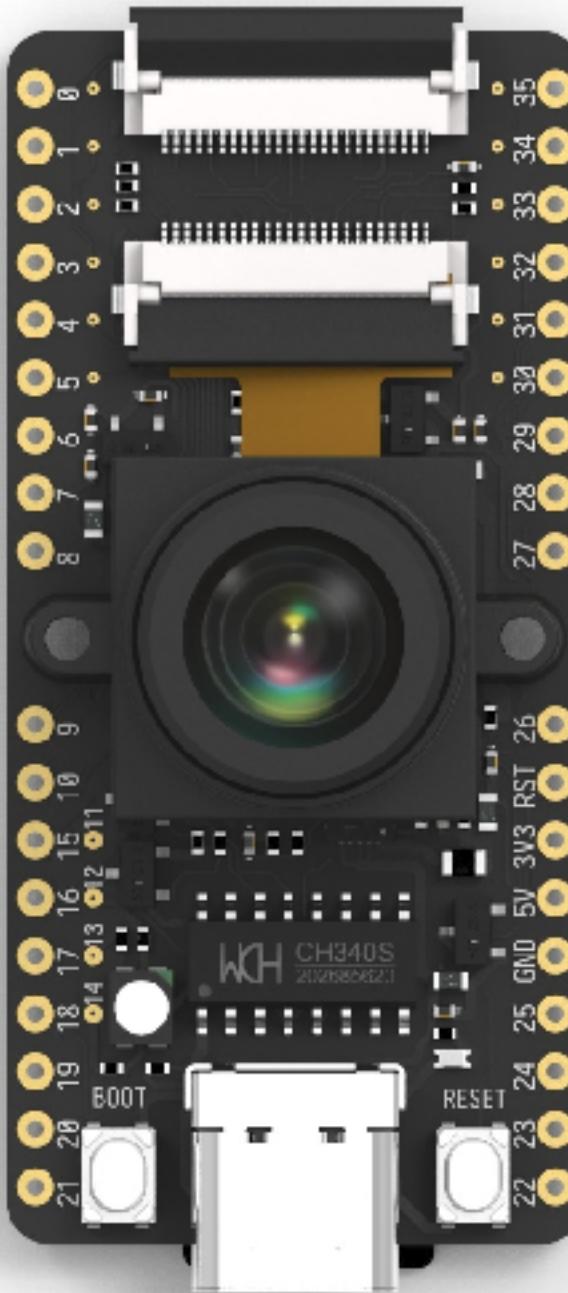
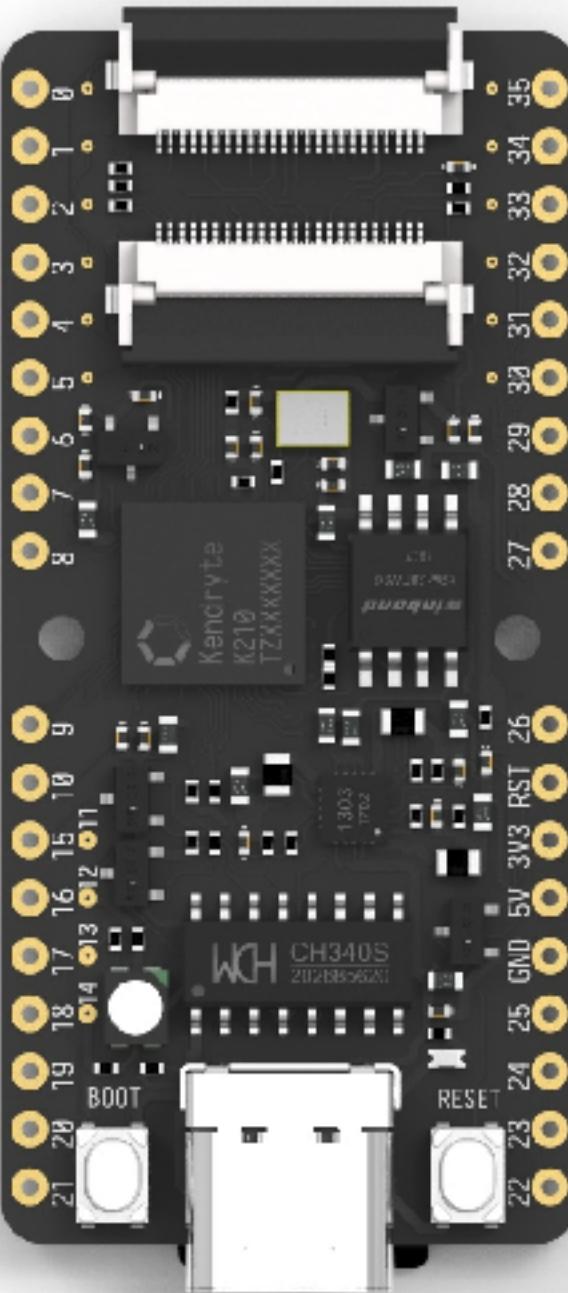
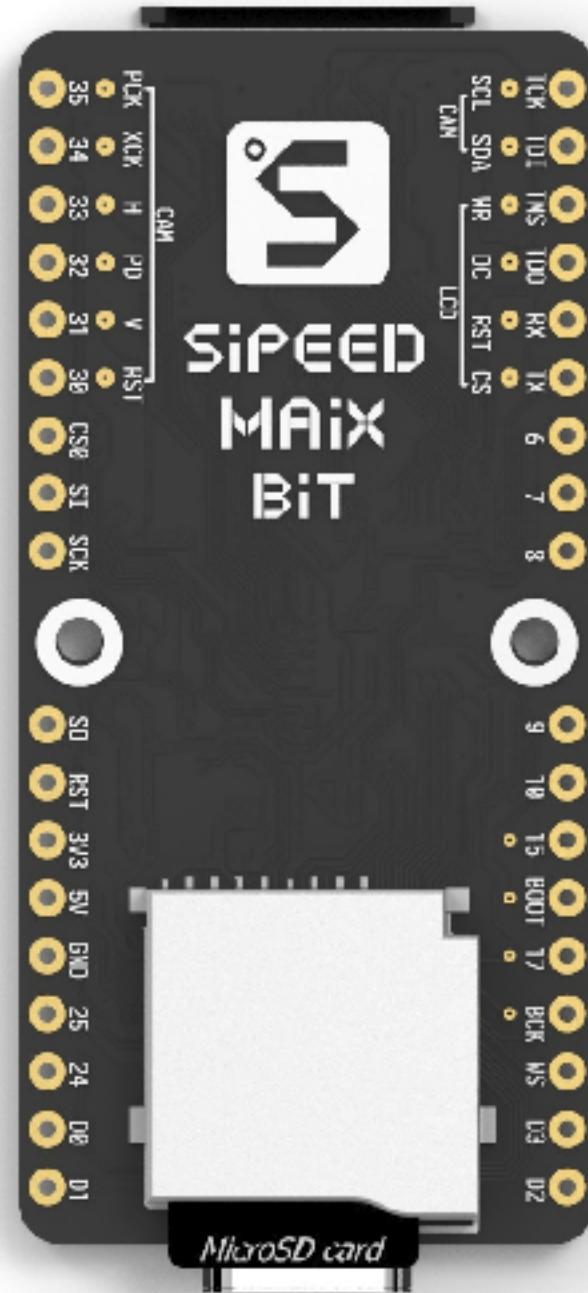
Bastante limitados em memoria, cpu e uso de bateria



	CPU	Memory	Flash	Connectivity	GPIO	Arch	Power Consumption	Deep Sleep	Price
<b>ATMega 328p</b>	16 Mhz	2 Kb	32 Kb	None	24	AVR RISC 8 bit	40mA	6uA	\$1.7
<b>STM32 “Blue Pill”</b>	72 Mhz	20 Kb	64 kb	None	32	ARM 32 bit Cortex M3	8mA	3uA	\$2.3
<b>Nordic nrf52840</b>	64 Mhz	256 Kb	1 MiB	Ble/Zigbee/ Thread	36	ARM 32 bit Cortex M4F	16mA	2uA	\$3
<b>ESP 32</b>	160 Mhz Dual Core	512 Kb	Up to 16MiB	Wifi/Ble	36	XTensa 32 bit	200mA	10uA	\$3

# Outros Edge Devices

# Performance melhor para machine learning também em micro controladores

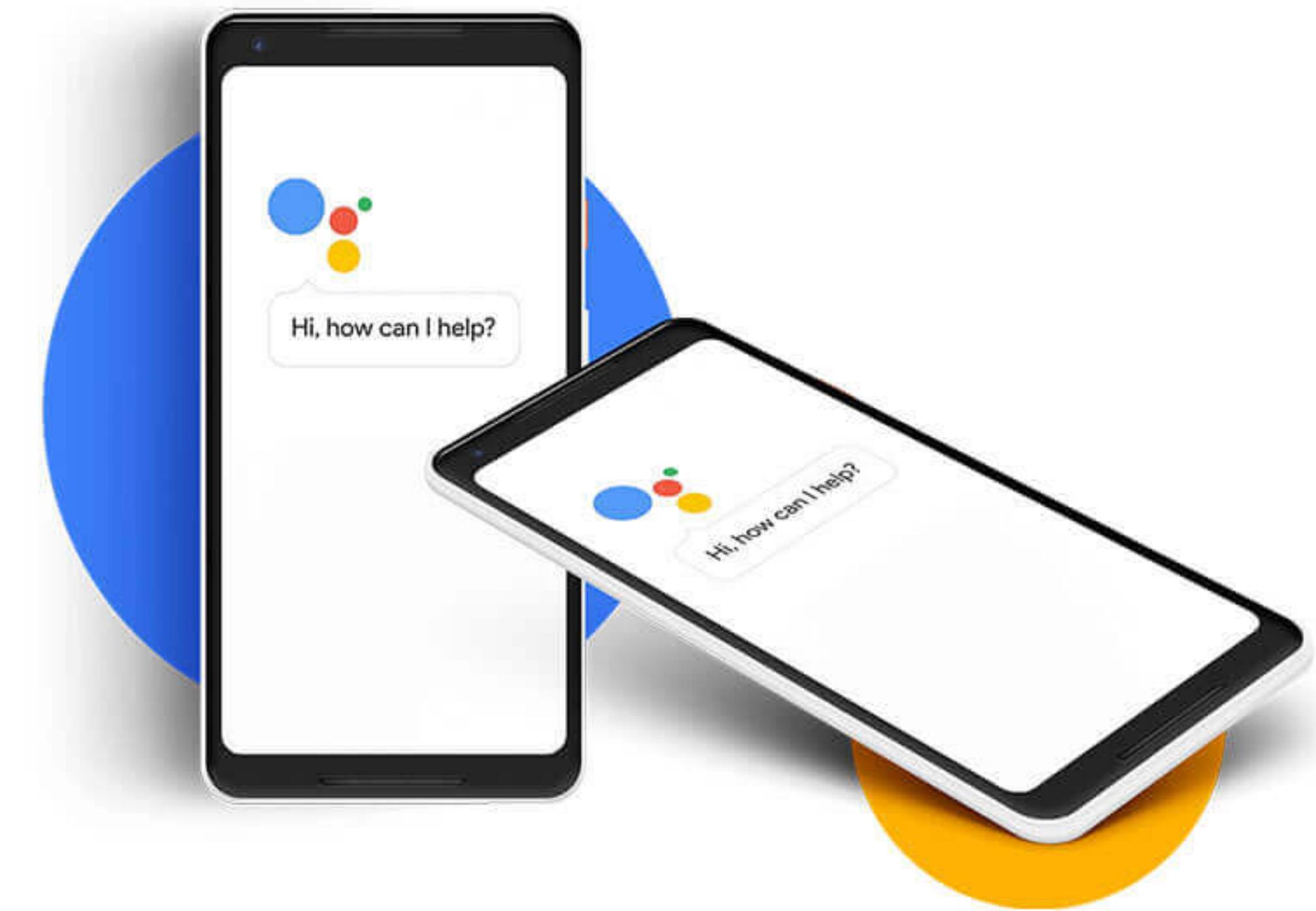


# Sipeed Maix Bit

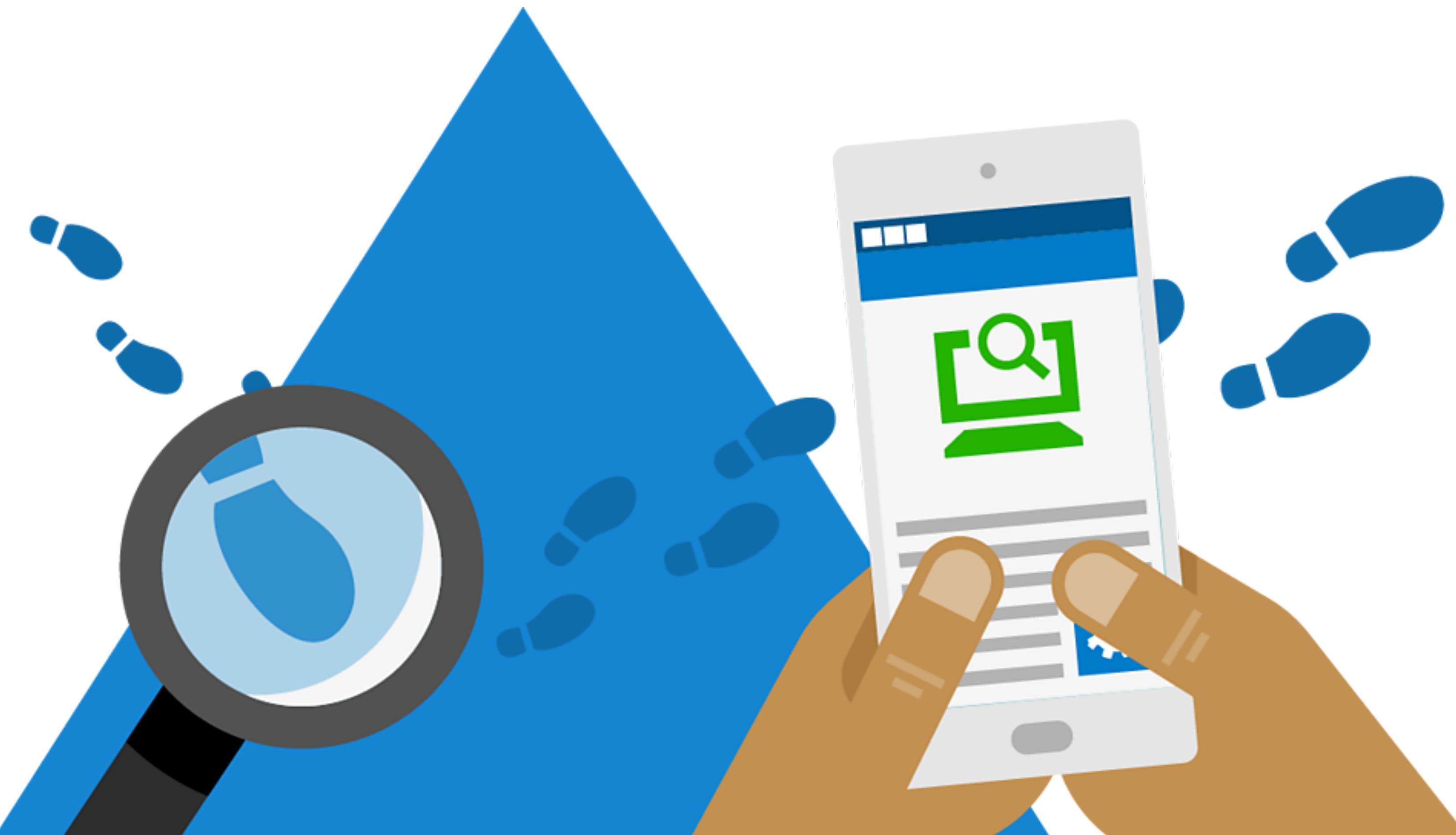
# Sparkfun Edge

IoT + Machine Learning  
Faz sentido ?

# Banda de Rede



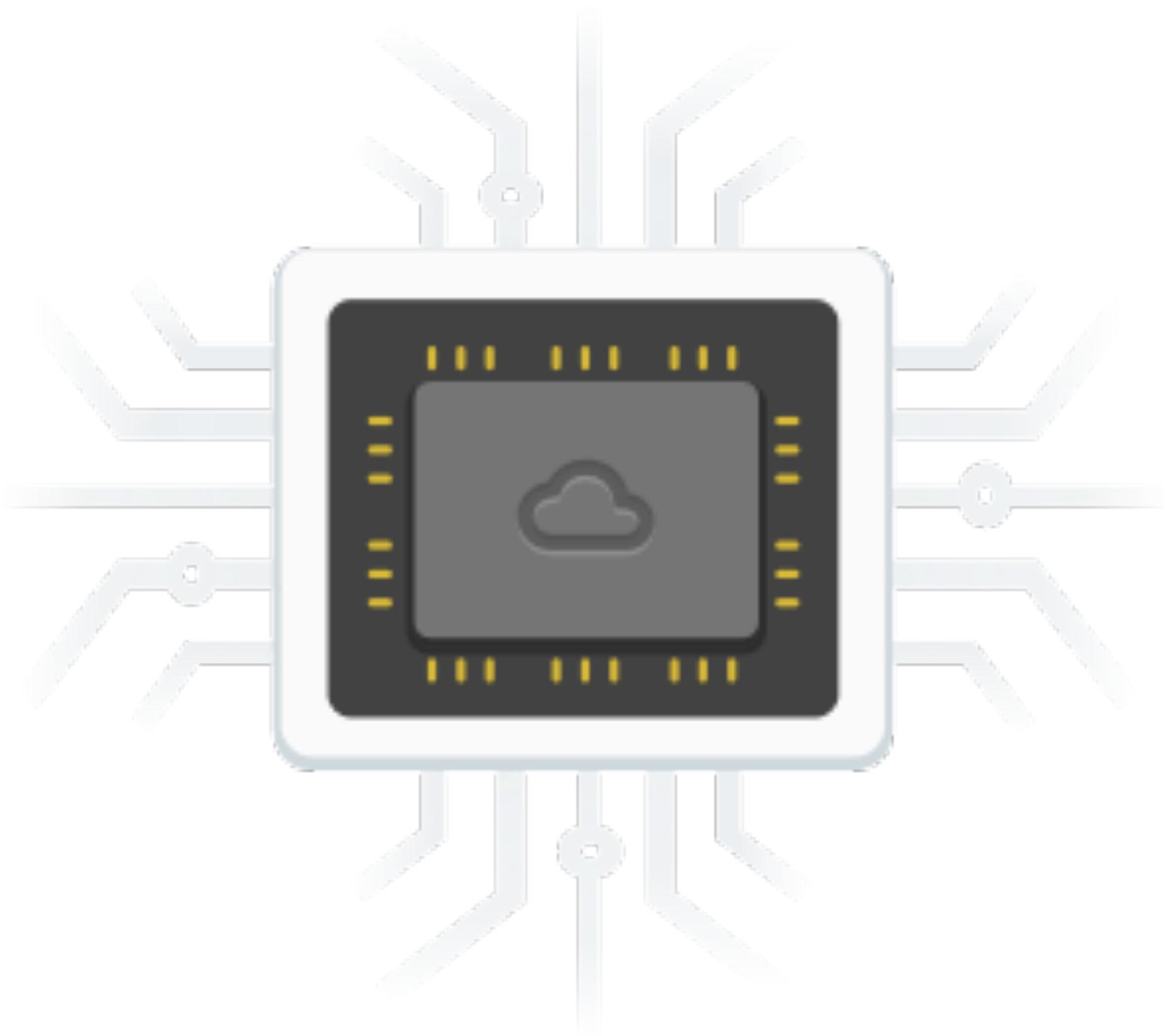
# Privacidade



# Latência



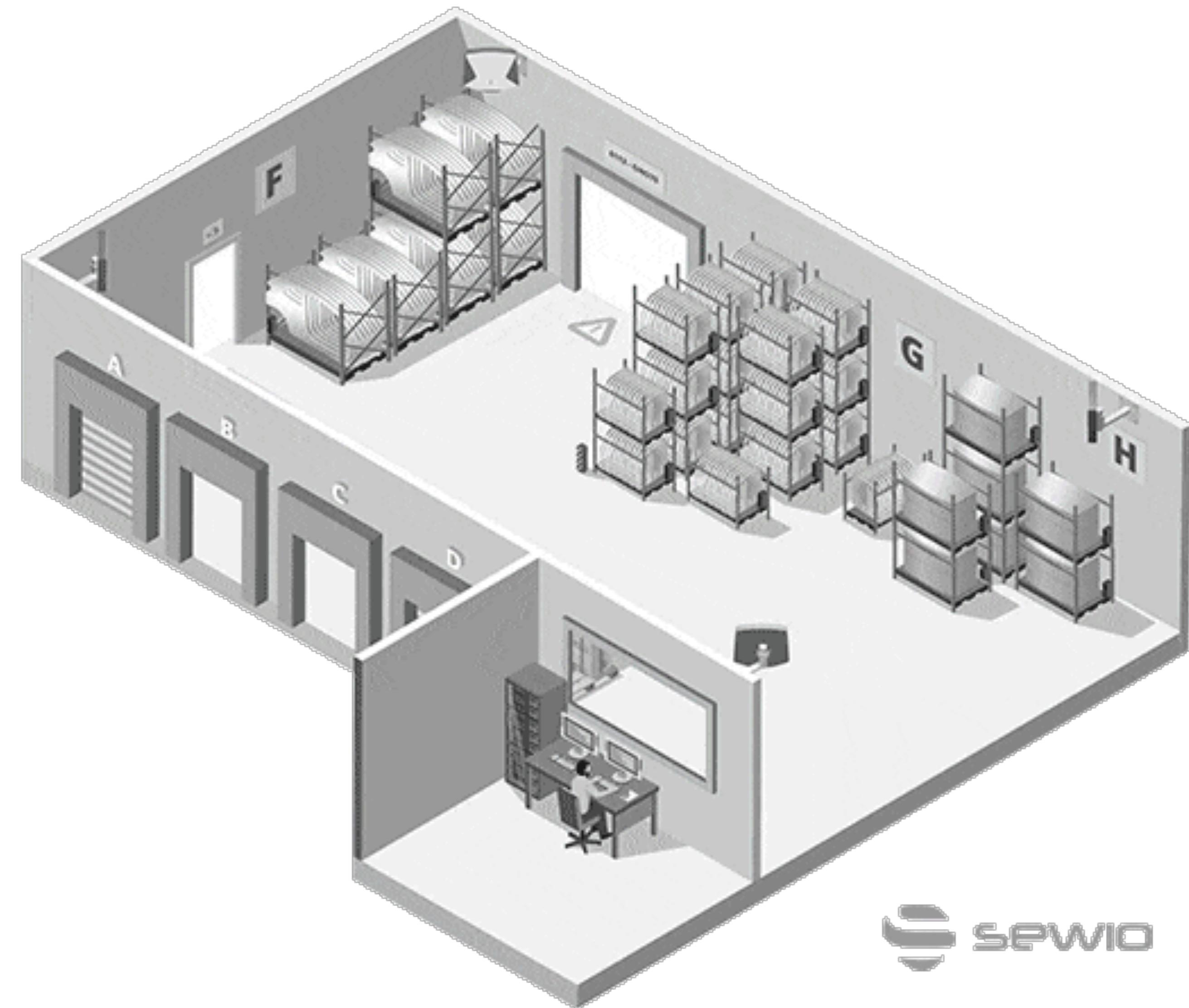
# Custos



# Mais casos de uso IoT + Machine Learning

# Monitoramento de ativos indoor

Dados bastante ruidosos e volumosos



# Comportamento de animais

Padrões de movimento - Internet of Cows/Cattle or Internet of Horses



# Problemas ambientais

Animais em extinção, queimadas e desflorestamento



---

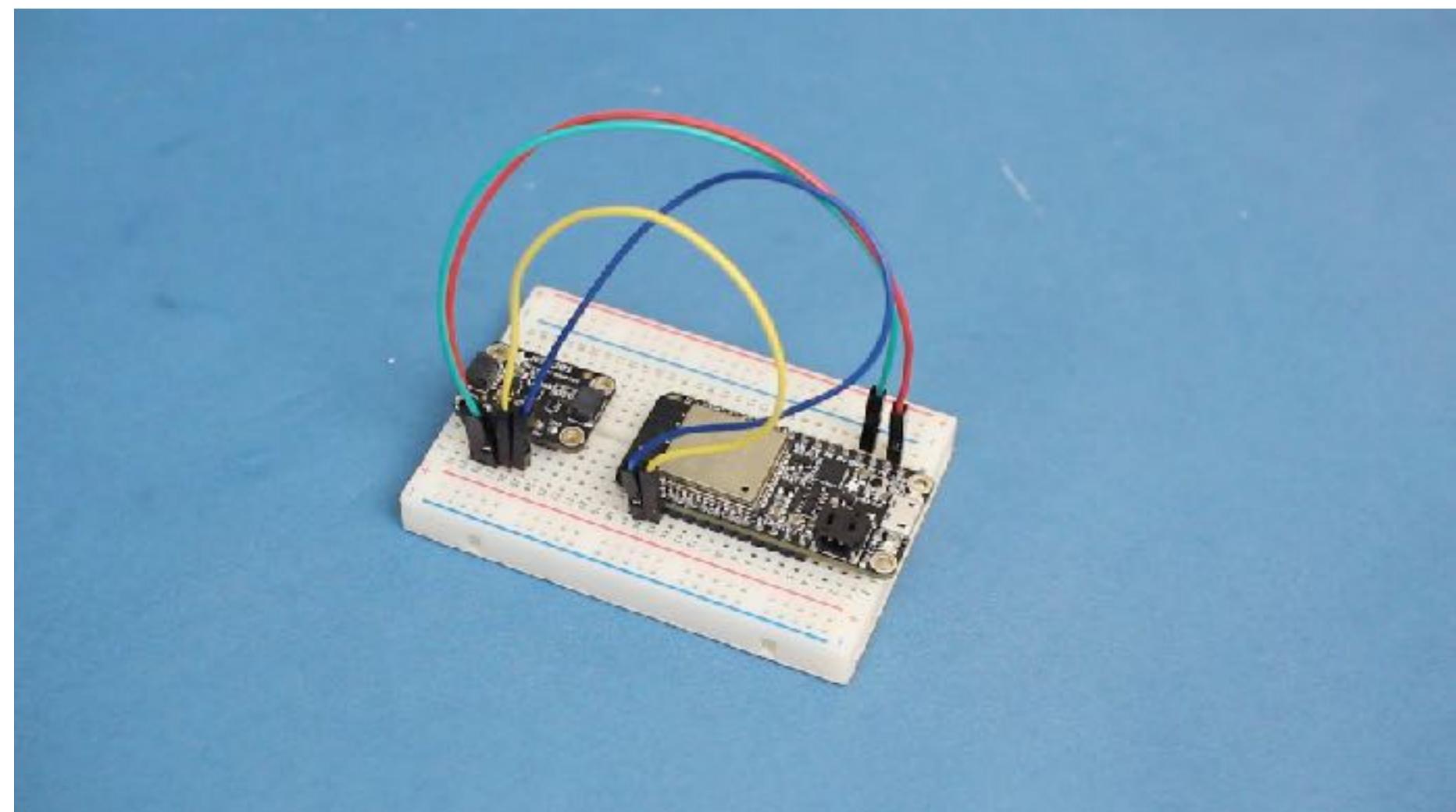
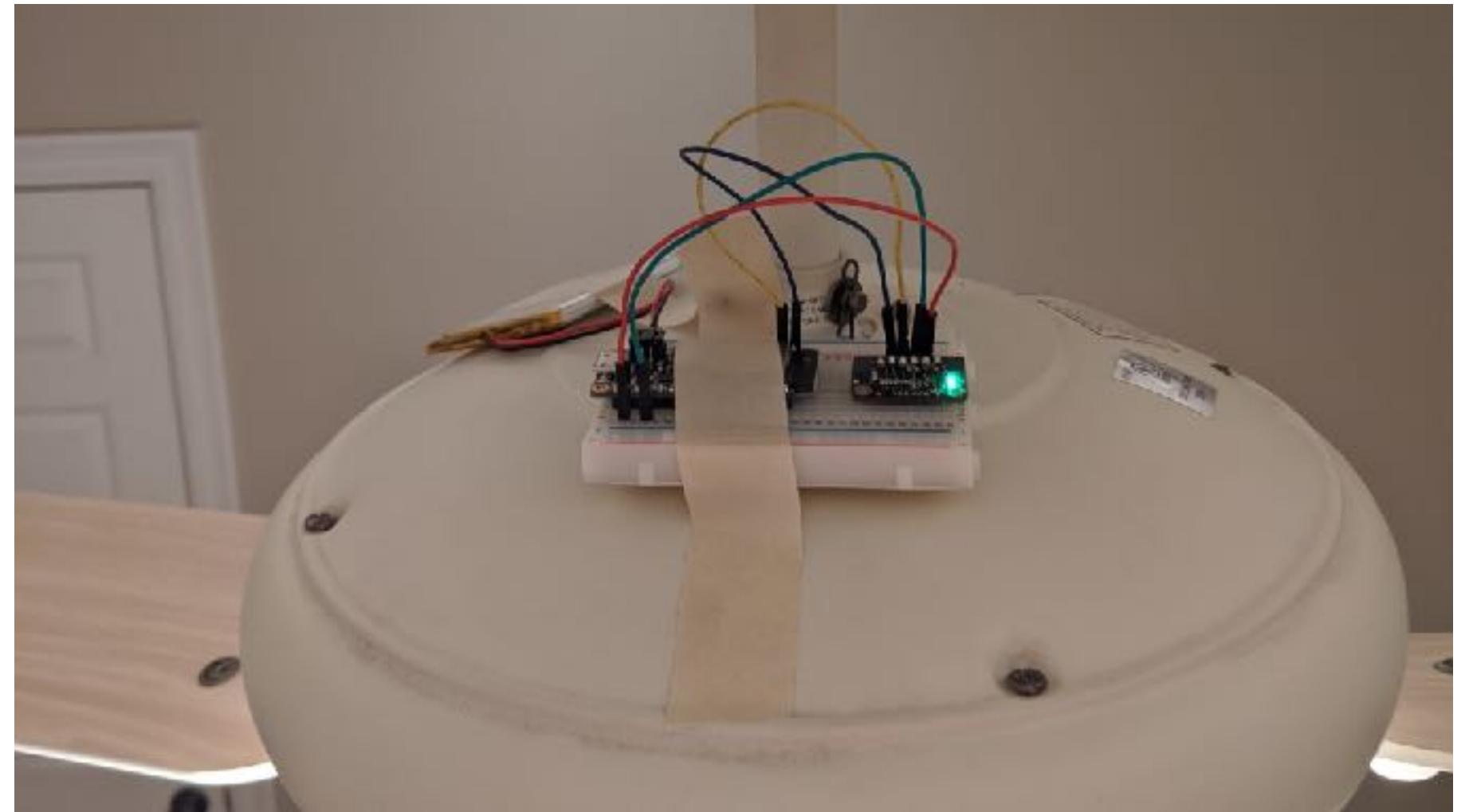
Building The World's Most Advanced  
**Wildlife Tracker.**



# Manutenção preditiva

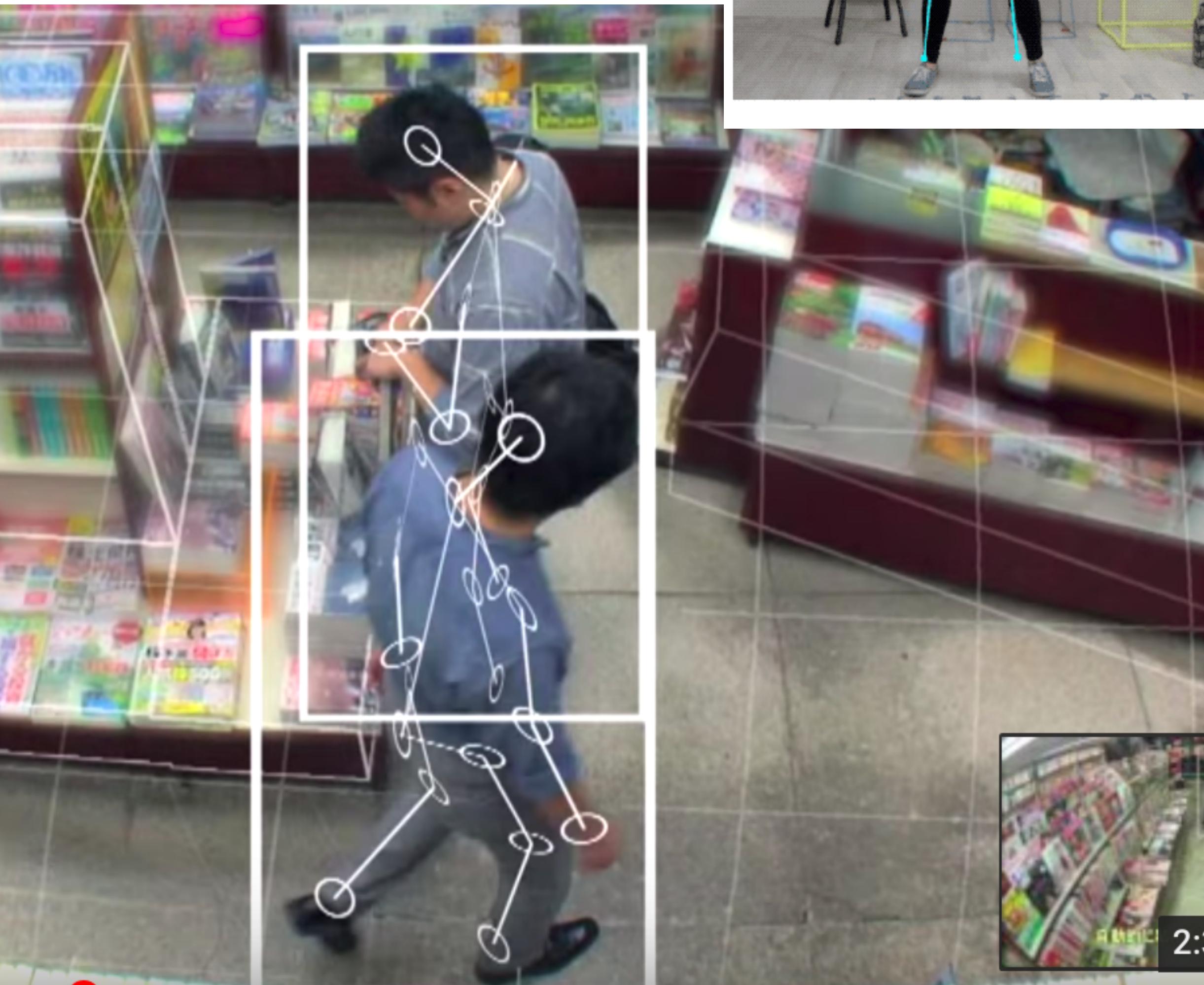
Padrões de vibração em motores

- Detectar anomalias no funcionamento de um equipamento
- Projeto usando um ESP32 + Accelerometro
- Diferentes modelos testados rodando no embarcado
  - Auto Encoder
  - Mahalanobis Distance
- [github.com/ShawnHymel/tinyml-example-anomaly-detection](https://github.com/ShawnHymel/tinyml-example-anomaly-detection)

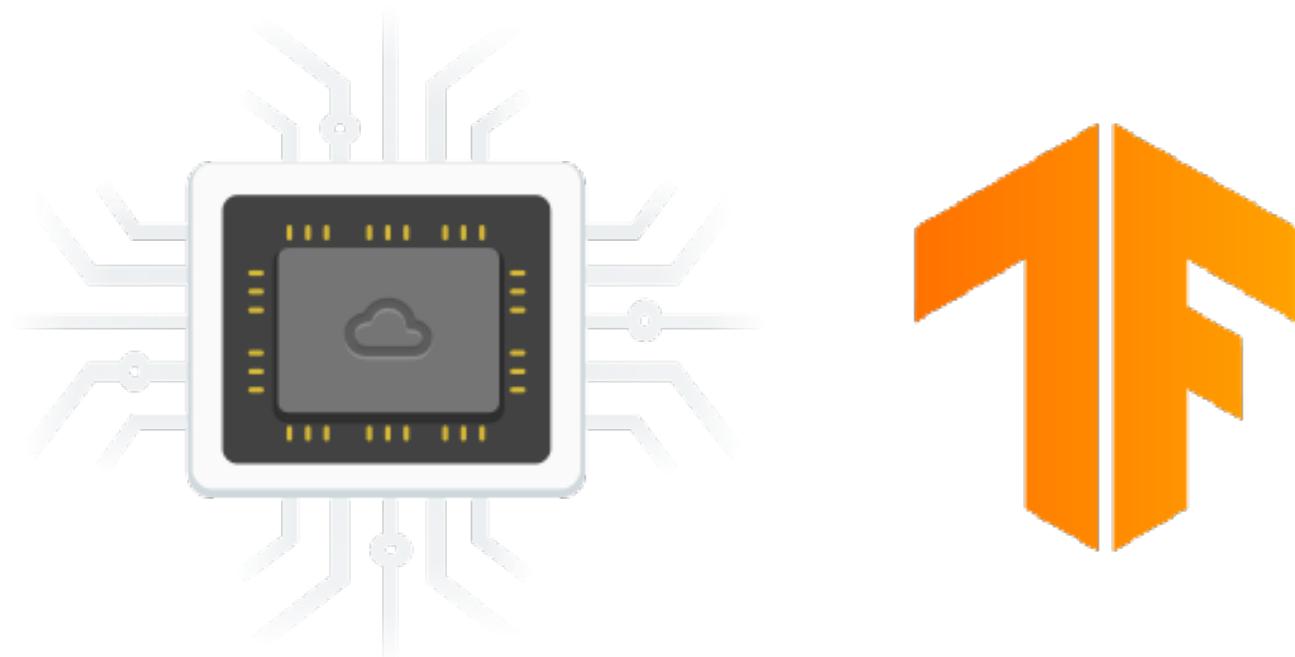


# Segurança e Proteção

Detectar máscaras e possíveis acidentes no trabalho



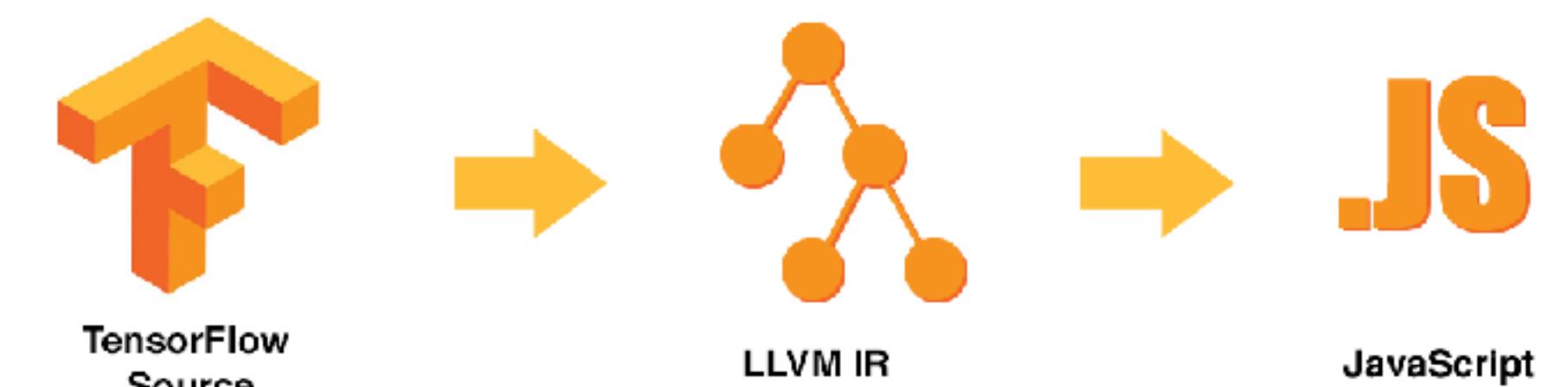
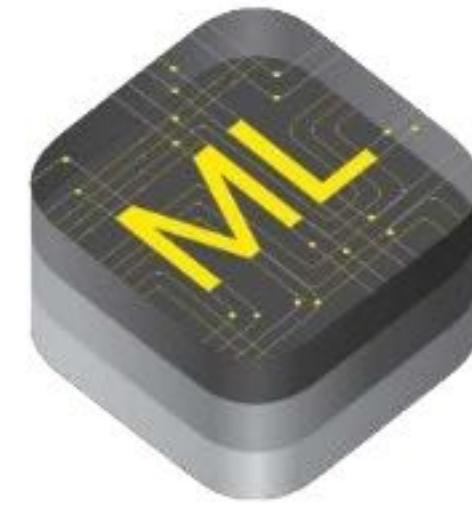
# O que o TensorFlow oferece para IoT ?



# TensorFlow Lite

Mobile e Internet das Coisas

- Plataformas
  - Android e iOS
  - Linux Embarcado
  - Web com TensorFlowJS
- Backend/Delegate
  - CPU
  - GPU WebGL / OpenCL / OpenGL
  - iOS Core ML / Metal
  - Web Assembly
  - Edge TPU 



# Modelos Pré treinados

Podem ser usados sem muitas alterações



## Classificação de imagens

Identifique centenas de objetos, incluindo pessoas, atividades, animais, plantas e locais.

[Ver modelo →](#)



## Detecção de objetos

Detecte diversos objetos com caixa delimitadora. Sim, isso inclui cães e gatos.

[Ver modelo →](#)



## Estimar poses

Estime poses para uma ou mais pessoas. Imagine as possibilidades, como festas de bonecos-palito.

[Ver modelo →](#)



## Resposta inteligente

Gere sugestões de respostas para mensagens de chat.

[Ver modelo →](#)



## Segmentação

Determine a forma de objetos com acurácia de localização exata e rótulos semânticos. Treinado para locais, pessoas, animais e muito mais.

[Ver modelo →](#)



## Transferência de estilo

Aplique qualquer estilo a uma imagem de entrada para criar uma nova imagem artística.

[Ver modelo →](#)



## Classificação de texto

Categorize texto livre em grupos predefinidos. Possíveis aplicativos: moderação de conteúdo abusivo, detecção de tom e muito mais.

[Ver modelo →](#)



## Resposta a perguntas em linguagem natural

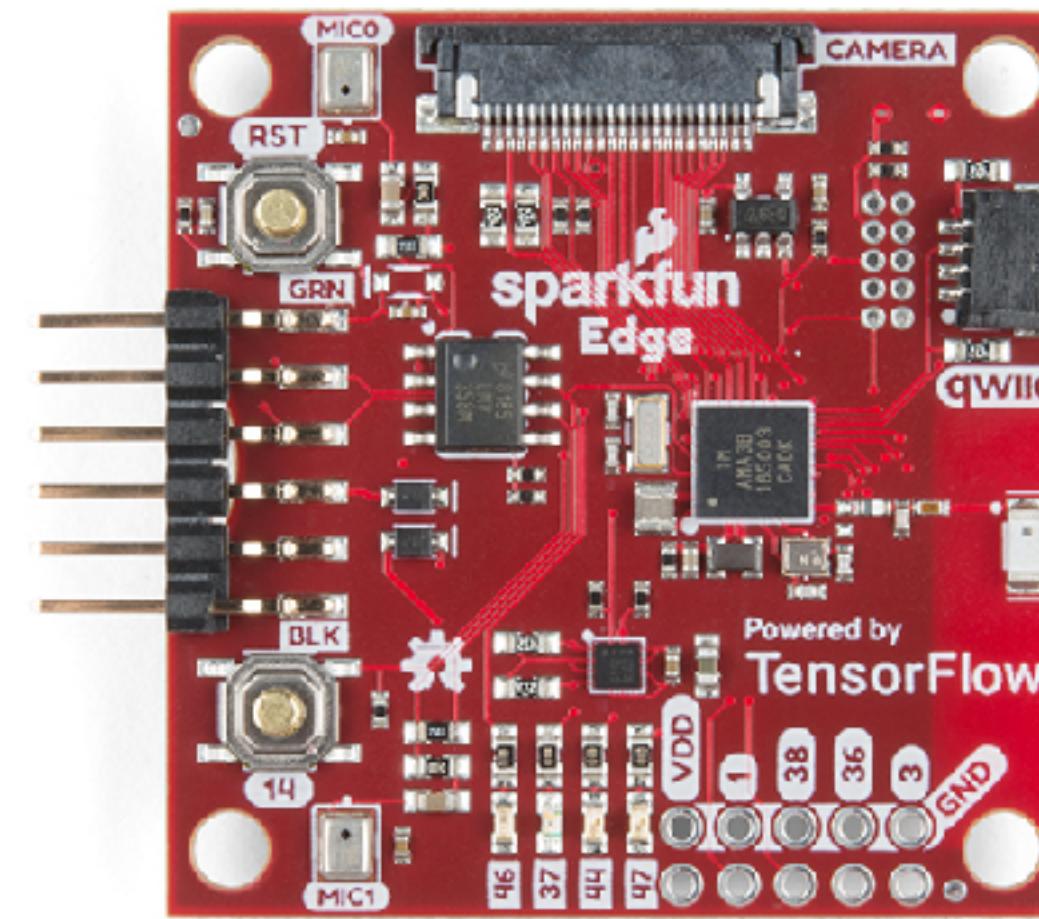
Responda a perguntas com base no conteúdo de um trecho de texto usando BERT.

[Ver modelo →](#)

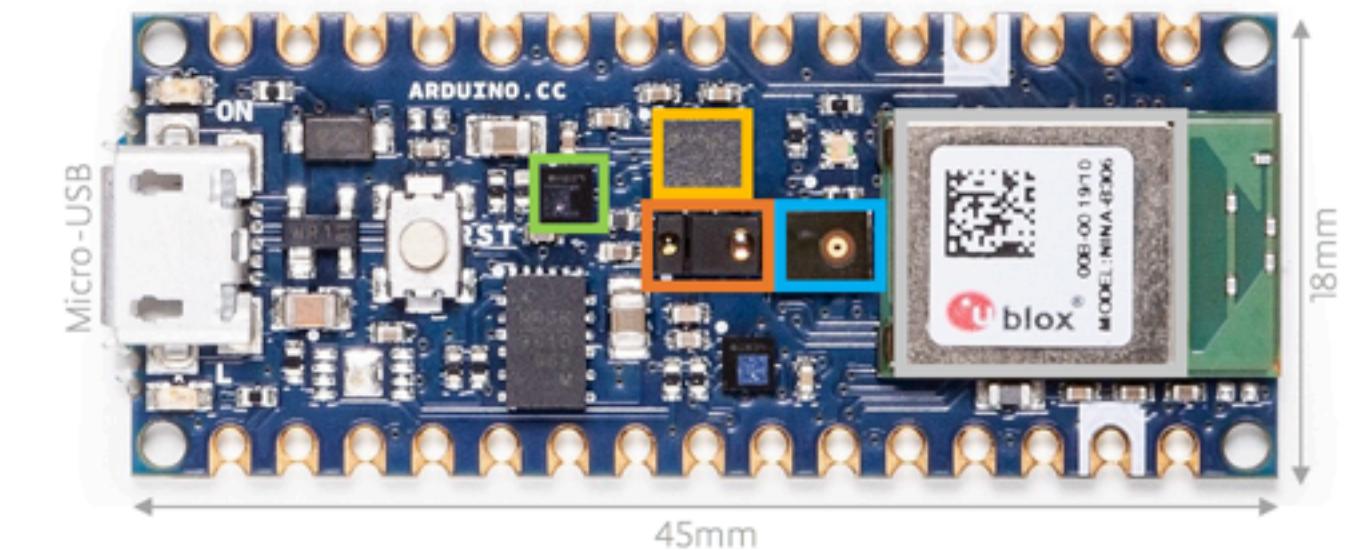
# TensorFlow Lite Micro

Micro-controladores e dispositivos com 16kb de ram ou mais

- Arquiteturas
  - Arm Cortex-M e CMSIS DSP
  - XTensa - ESP32
- Biblioteca em C/C++
- Demos
  - Detecção de hotwords - Sim/Não
  - Varinha mágica - Detecta diferentes movimentos
  - Deteção de Pessoas com Camera
  - Conjuntos de instruções limitado
  - [www.tensorflow.org/lite/microcontrollers/build\\_convert#operation\\_support](http://www.tensorflow.org/lite/microcontrollers/build_convert#operation_support)



ARDUINO NANO 33 BLE SENSE

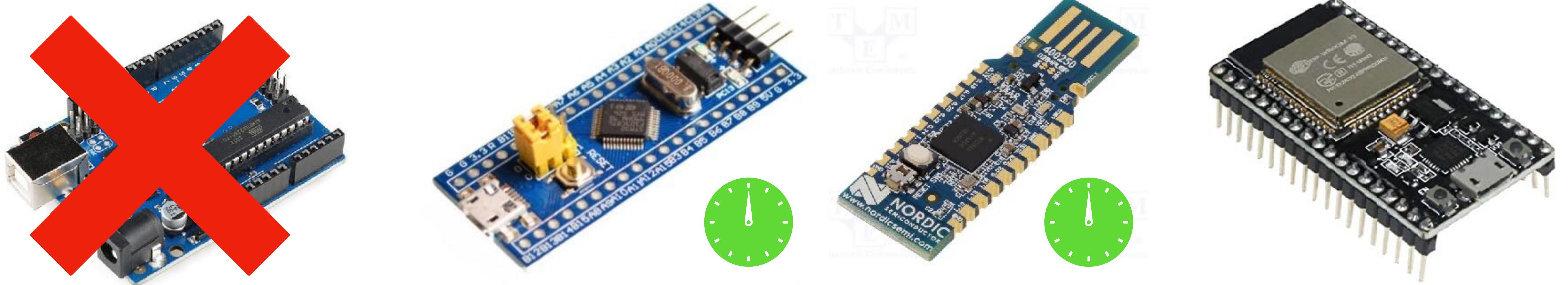


- Color, brightness, proximity and gesture sensor
- Digital microphone
- Motion, vibration and orientation sensor
- Temperature, humidity and pressure sensor
- Arm Cortex-M4 microcontroller and BLE module



# Relembrando alguns devices

Varios tem suporte ao TensorFlow Lite Micro



	CPU	Memory	Flash	Connectivity	GPIO	Arch	Power Consump	Deep Sleep	Price
<b>ATMega 328p</b>	16 Mhz	2 Kb	32 Kb	None	24	AVR RISC 8 bit	40mA	6uA	\$1.7
<b>STM32 "Blue Pill"</b>	72 Mhz	20 Kb	64 kb	None	32	ARM Cortex M3	8mA	3uA	\$2.3
<b>Nordic nrf52840</b>	64 Mhz	256 Kb	1 MiB	Ble/Zigbee/Thread	36	ARM Cortex M4F	16mA	2uA	\$3
<b>ESP 32</b>	160 Mhz Dual Core	512 Kb	Up to 16MiB	Wifi/Ble	36	XTensa 32 bit	200mA	10uA	\$3

# Digital Signal Processing - DSP

Instruções otimizadas para cálculos complexos em Micro Controladores

- “GPU” para micro controladores
- Instruções especializadas para determinadas tarefas

 CMSIS-DSP Version 1.8.0

CMSIS DSP Software Library

General	Core(A)	Core(M)	Driver	DSP	NN	RTOS v1	RTOS v2	Pack	Build	SVD	DAP	Zone	
Main Page	Usage and Description			Reference									

The library has generally separate functions for operating on 8-bit integers, 16-bit integers, 32-bit integer and 32-bit floating-

### Using the Library

The library installer contains prebuilt versions of the libraries in the `Lib` folder.

Here is the list of pre-built libraries :

- `arm_cortexM7lfdp_math.lib` (Cortex-M7, Little endian, Double Precision Floating Point Unit)
- `arm_cortexM7bfdp_math.lib` (Cortex-M7, Big endian, Double Precision Floating Point Unit)
- `arm_cortexM7lfsp_math.lib` (Cortex-M7, Little endian, Single Precision Floating Point Unit)
- `arm_cortexM7bfsp_math.lib` (Cortex-M7, Big endian and Single Precision Floating Point Unit on)
- `arm_cortexM7l_math.lib` (Cortex-M7, Little endian)
- `arm_cortexM7b_math.lib` (Cortex-M7, Big endian)
- `arm_cortexM4lf_math.lib` (Cortex-M4, Little endian, Floating Point Unit)
- `arm_cortexM4bf_math.lib` (Cortex-M4, Big endian, Floating Point Unit)
- `arm_cortexM4l_math.lib` (Cortex-M4, Little endian)
- `arm_cortexM4b_math.lib` (Cortex-M4, Big endian)
- `arm_cortexM3l_math.lib` (Cortex-M3, Little endian)
- `arm_cortexM3b_math.lib` (Cortex-M3, Big endian)
- `arm_cortexM0l_math.lib` (Cortex-M0 / Cortex-M0+, Little endian)
- `arm_cortexM0b_math.lib` (Cortex-M0 / Cortex-M0+, Big endian)
- `arm_ARMv8MBLI_math.lib` (Armv8-M Baseline, Little endian)
- `arm_ARMv8MMLI_math.lib` (Armv8-M Mainline, Little endian)
- `arm_ARMv8MMLIfsp_math.lib` (Armv8-M Mainline, Little endian, Single Precision Floating Point Unit)
- `arm_ARMv8MMLld_math.lib` (Armv8-M Mainline, Little endian, DSP instructions)
- `arm_ARMv8MMLldfsp_math.lib` (Armv8-M Mainline, Little endian, DSP instructions, Single Precision Floating Point Unit)

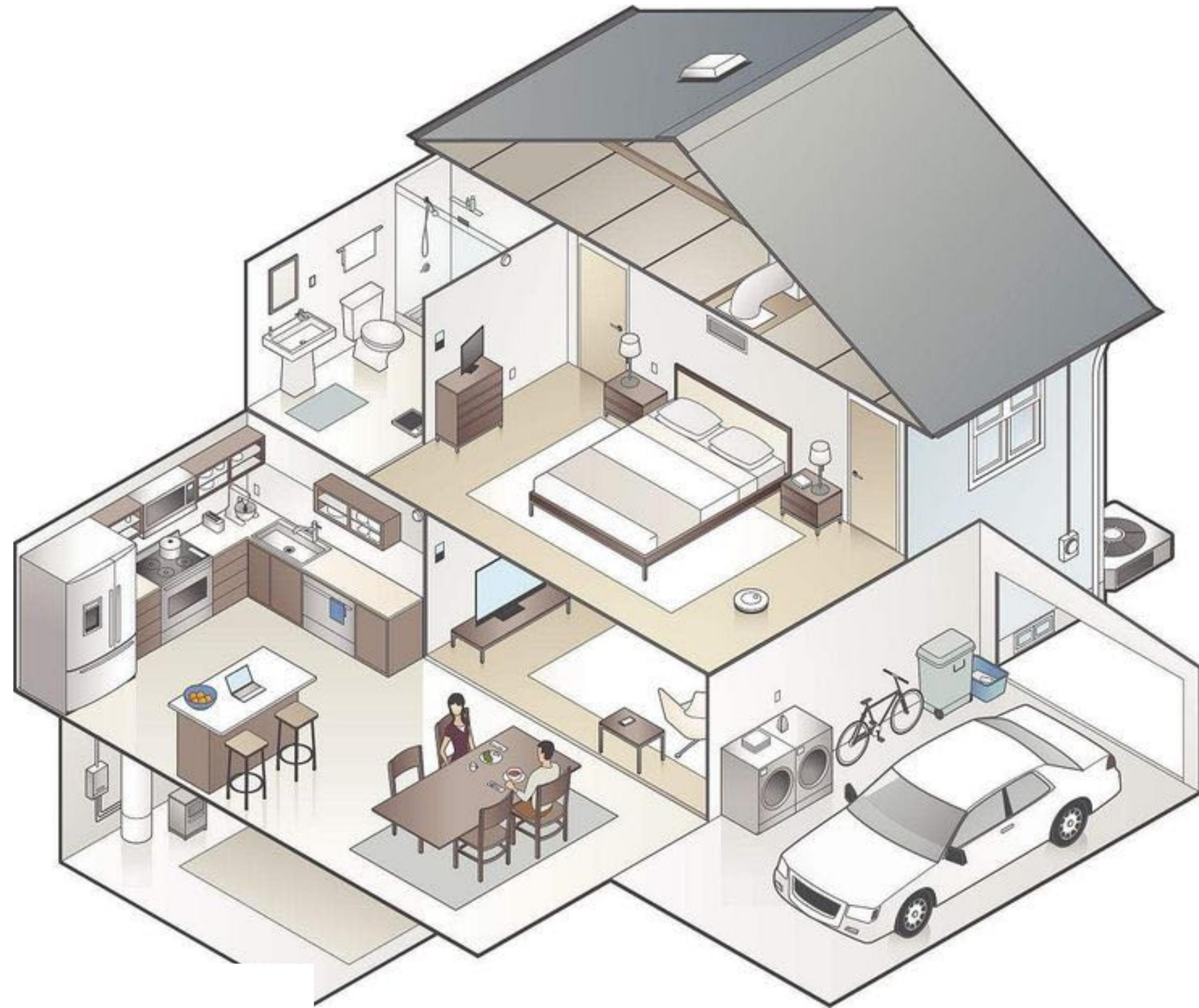
The library functions are declared in the public file `arm_math.h` which is placed in the `Include` folder. Simply include this file at the top of your application code and begin calling the library functions. The Library supports single public header file `arm_math.h` for C and C++.

**Falar é fácil  
Vamos ver na prática**

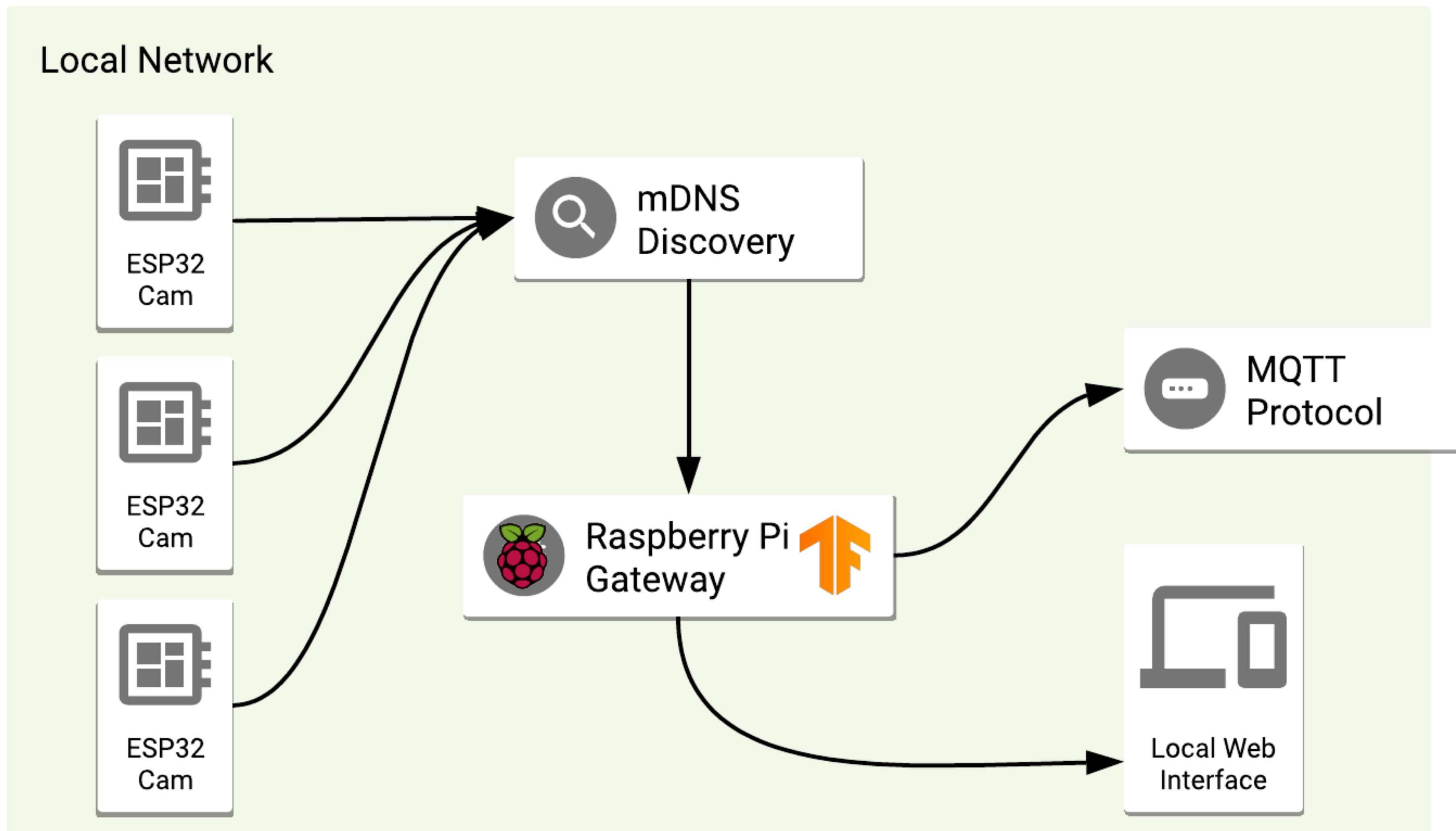
ATENÇÃO  
EU NÃO SOU NENHUM EXPERT  
EM MACHINE LEARNING

Mas talvez essa seja a **MELHOR**  
parte desses projetos

**Monitoramento indoor usando  
cameras como sensores**



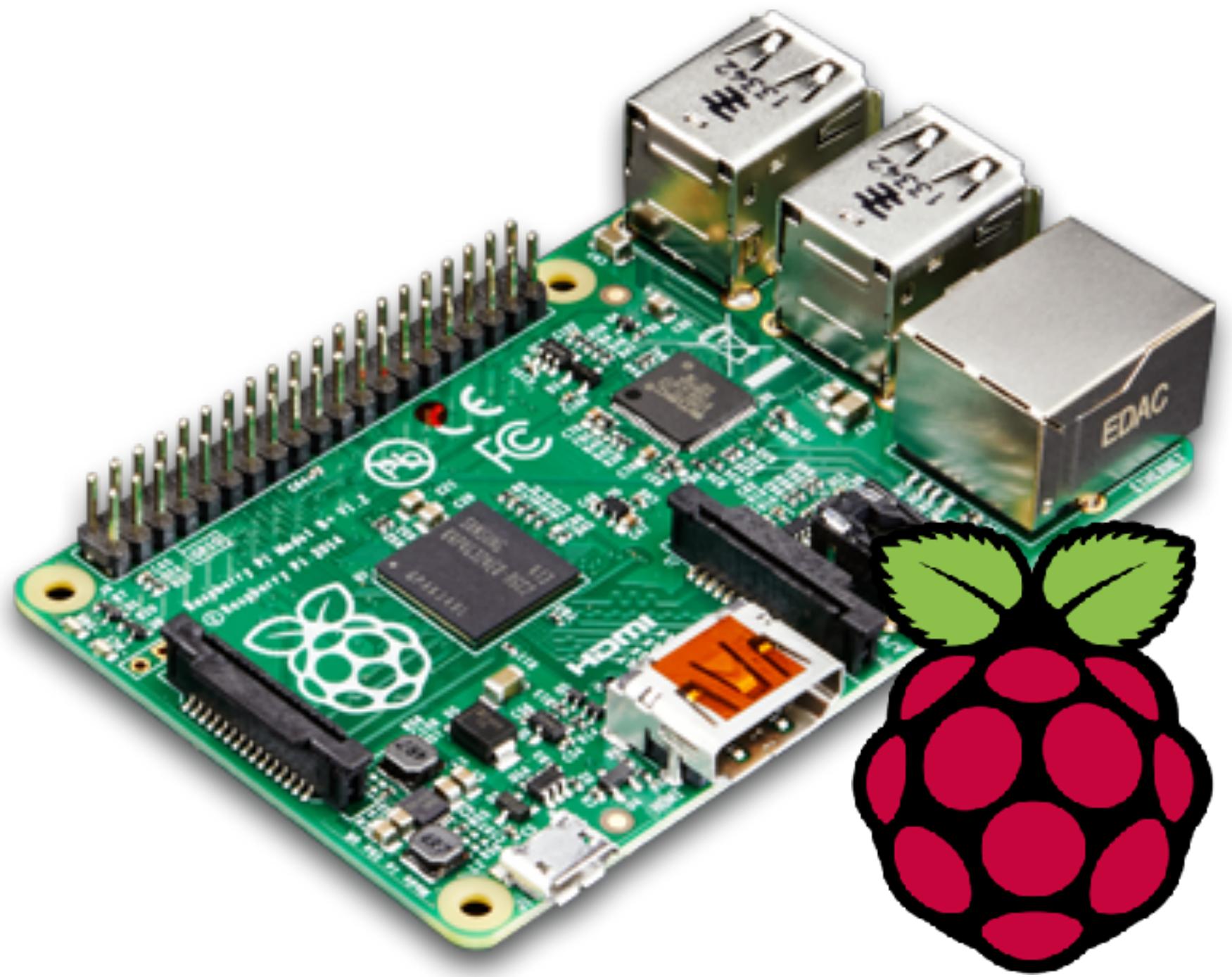
# Arquitetura de Rede Local



10\$



**ESP32 Cam**



**Raspberry Pi 3**

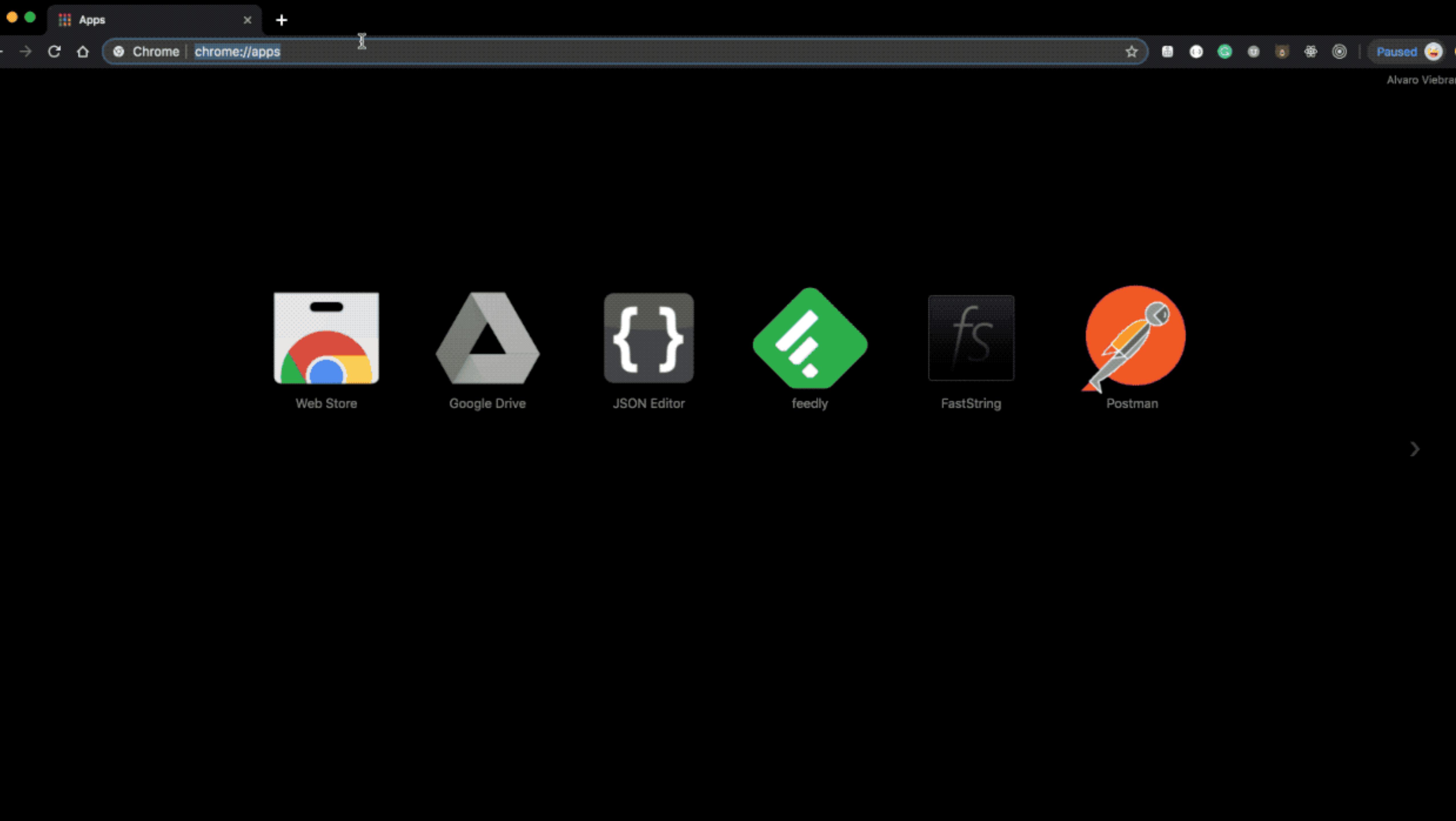
35\$

# Encontrando cameras locais

Podemos testar localmente using a ferramenta dns-sd

- http://indoor-camera-ec5d.local/jpg
- http://indoor-camera-60d8.local/jpg

```
# alvaroviebrantz @ MacBook-Pro-Alvaro-Viebrantz in ~/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/edge-server on git:master ✘ [15:24:31]
$
```



Apps

+

Chrome | chrome://apps



Alvaro Viebran



Web Store



Google Drive



JSON Editor



feedly



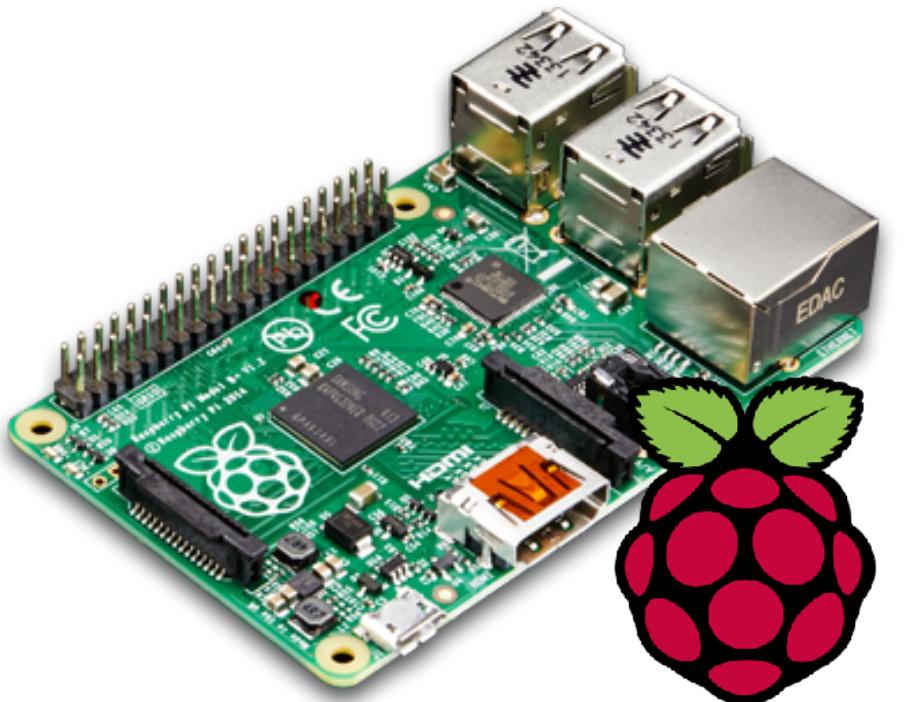
FastString



Postman



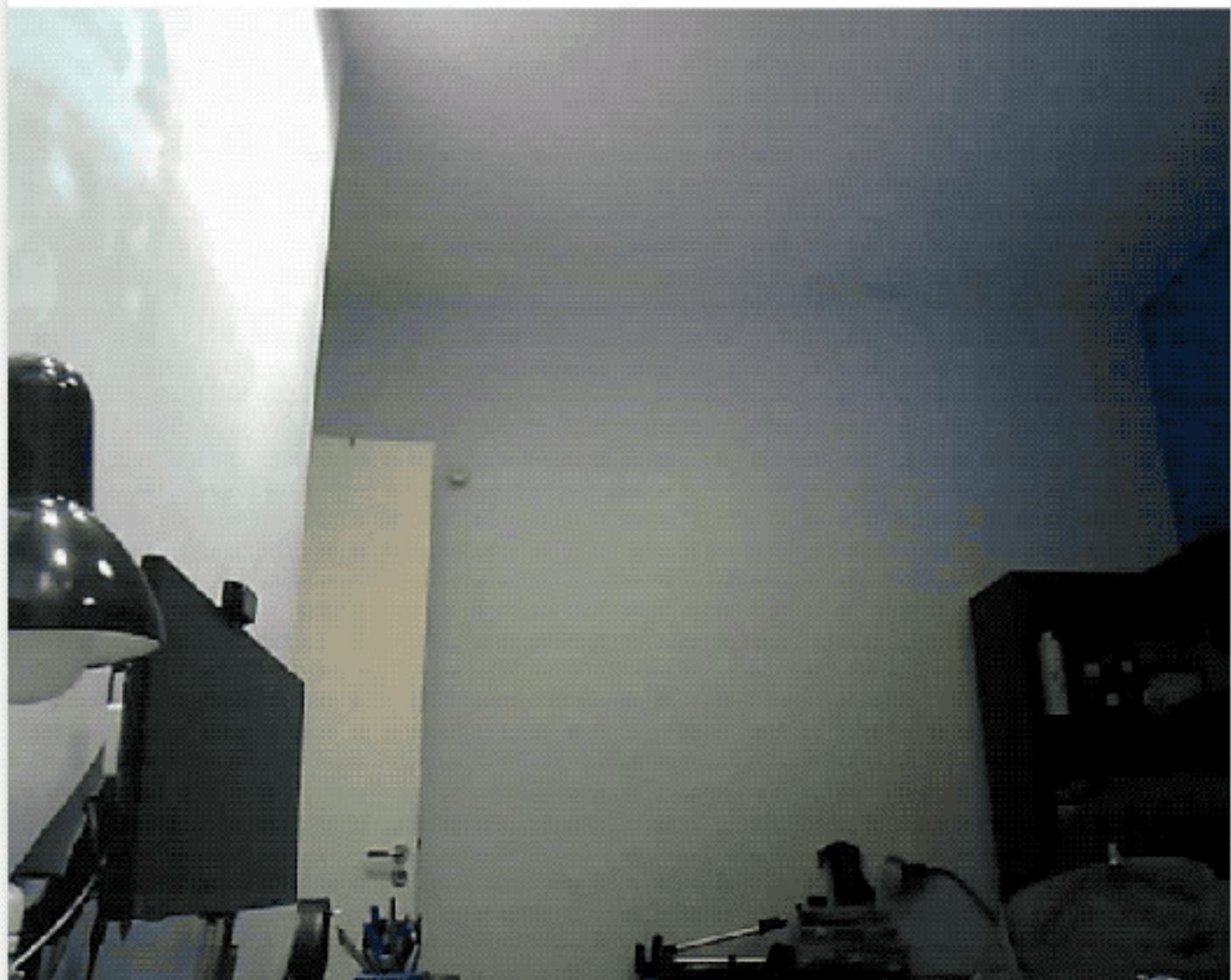
Baixar imagens, classificar  
e enviar classes para a Nuvem



# IoT Edge - Local Monitor

indoor-camera-b021

5/25/2019, 6:03:51 PM



Nothing found.

indoor-camera-ec5d

6/1/2019, 12:03:58 PM



cat(1)

indoor-camera-8455

6/1/2019, 12:11:37 PM



person(1)

cat(1)

# Performance

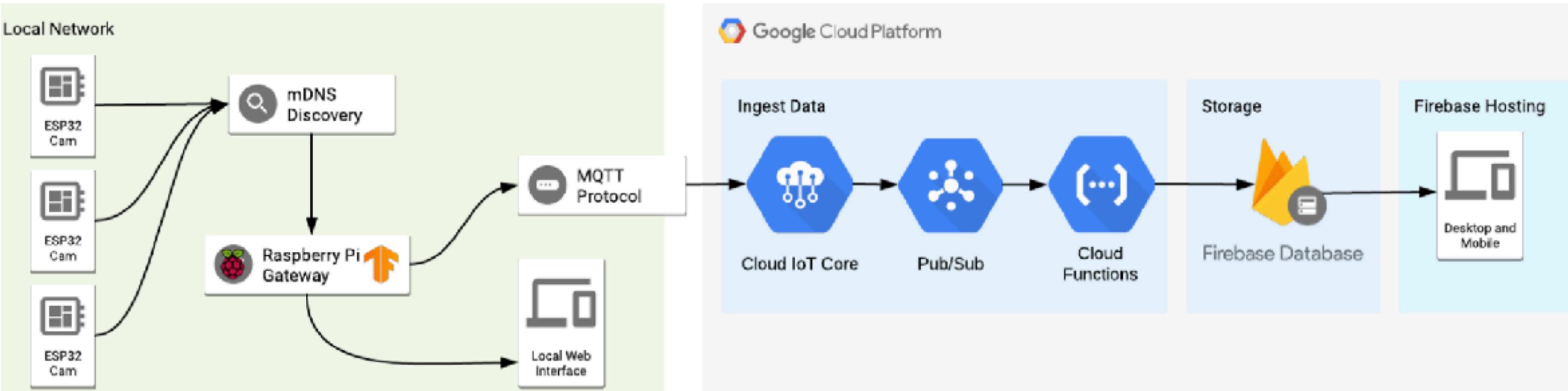
Dica - Use tfjs-node ou tfjs-node-gpu se possível

	Meu Computador	Raspberry Pi 3
TensorFlow.js Core	<b>8 seconds per frame</b>	<b>45 seconds per frame</b>
TensorFlow.js Node	<b>200 milliseconds per frame</b>	<b>1 second per frame</b>

# Artigo descrevendo tudo que mostrei até agora

[bit.ly/gcloud-iot-edge](https://bit.ly/gcloud-iot-edge)

 Alvaro Viebranz in Google Cloud Platform - Community  
Feb 19 · 13 min read

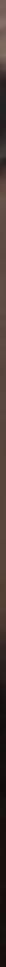
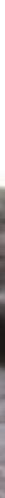


**Asset Tracking using Cameras, IoT, Machine Learning and Edge Computing.**

Read more...  149  2 responses

# Eu amo gatos

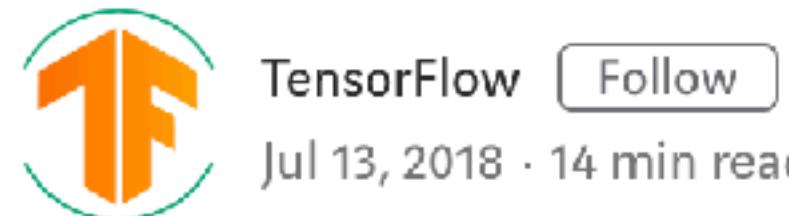
E se eu pudesse rastrear meus gatos em casa ?



# Treinando um modelo customizado

Primeira tentativa

Training and serving a realtime mobile object detector in 30 minutes with Cloud TPUs



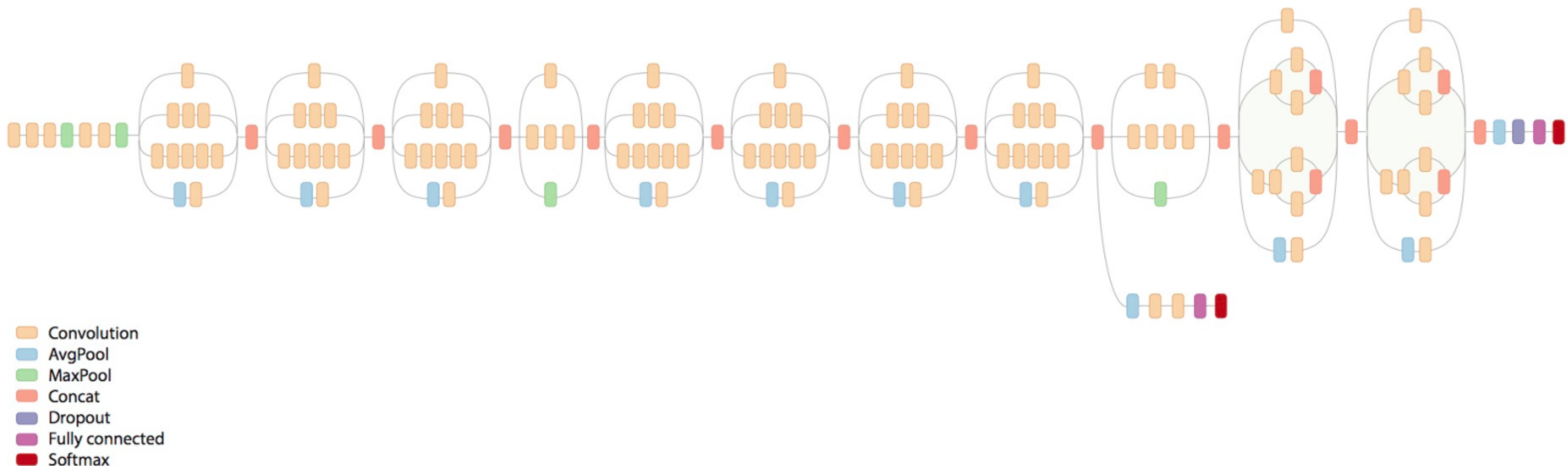
By [Sara Robinson](#), [Aakanksha Chowdhery](#), and [Jonathan Huang](#)

How to train your own Object Detector with TensorFlow's Object Detector API



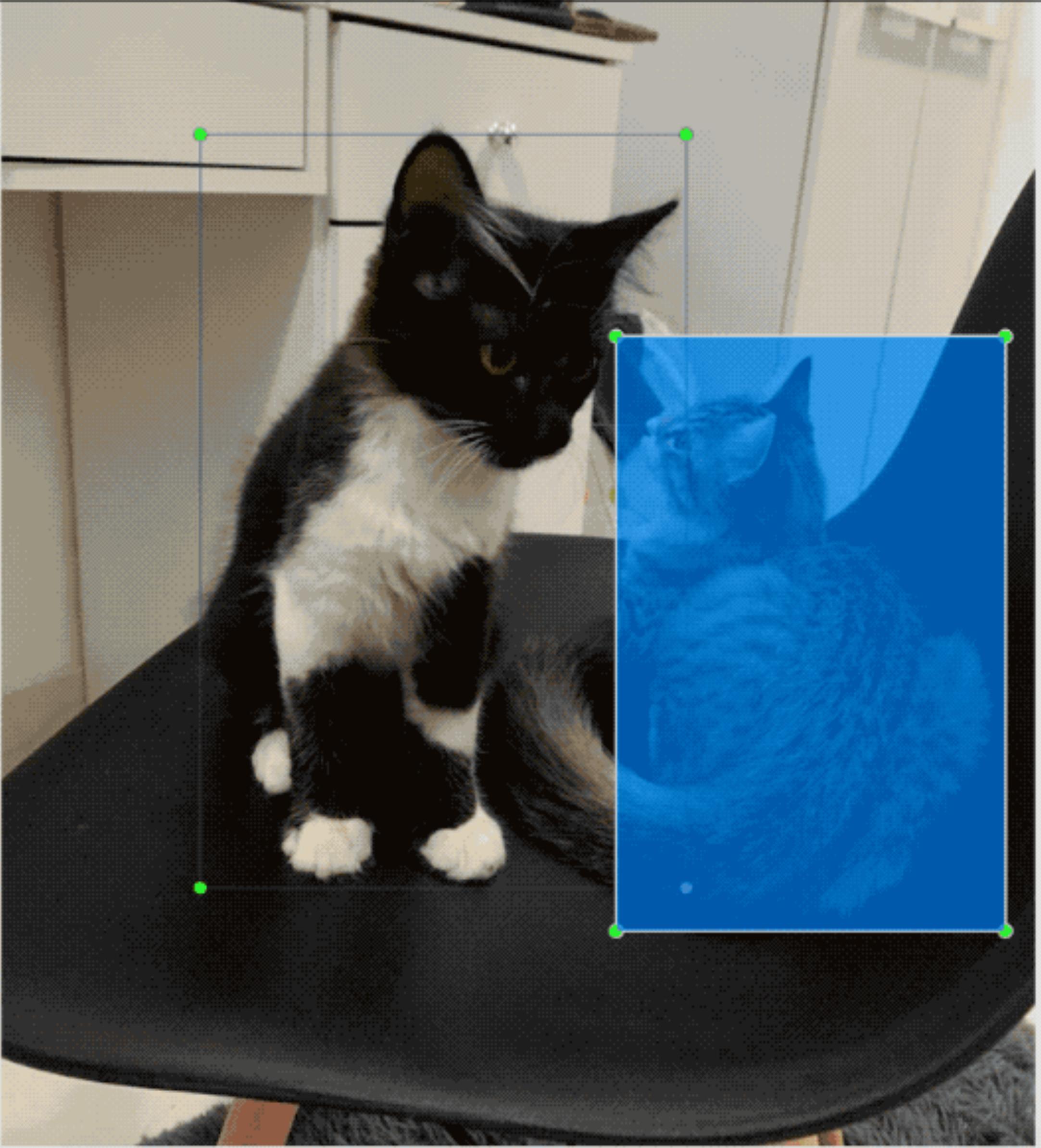
# Transfer Learning

Usando modelos pré treinados para treinar novos modelos com datasets menores



# Construindo o dataset

Geralmente a parte mais difícil de qualquer projeto de ML



Box Labels

Edit Label

difficult

Use default label

muffin

berry

PascalVOC

Create RectBox

Duplicate RectBox

Delete RectBox

Zoom In

89 %

File List

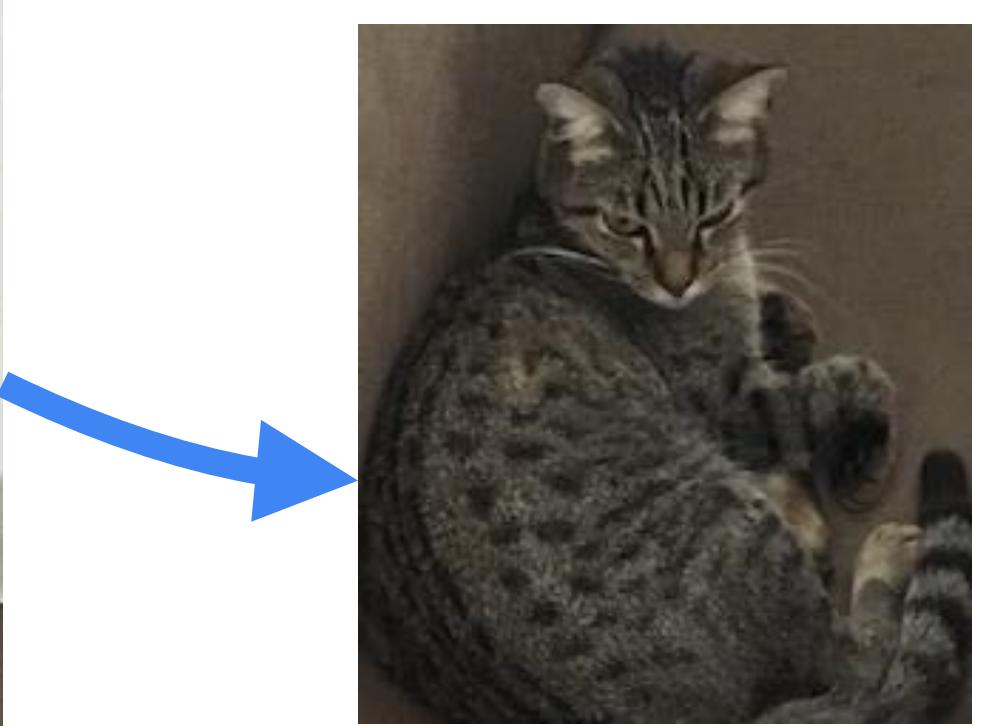
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00002.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00003.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00004.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00005.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00006.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00007.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00008.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00009.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00011.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat00012.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200001.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200002.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200004.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200005.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200006.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200007.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200008.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200009.jpg
- /Users/alvaroviebrantz/Documents/Desenvolvimento/IoT/gcloud-iot-edge-tensorflow/custom\_model/images/cat200010.jpg

**BAIXA ACURÁCIA NO MEU MODELO**



# Treinar modelo de classificação

Usando o mesmo dataset





berry

jam

muffin

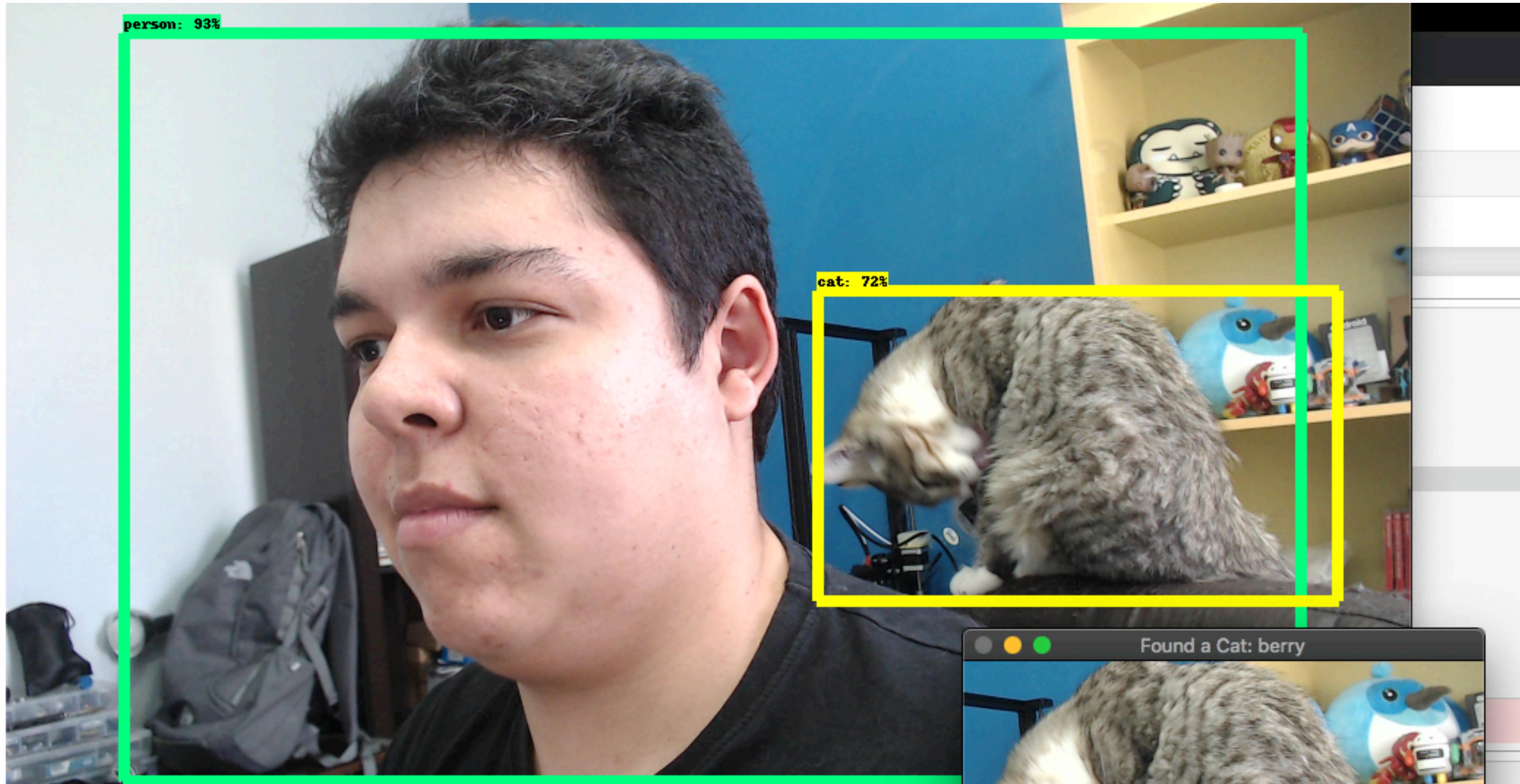
popcorn

raspberry

# Melhor dos dois mundos

Usar um modelo pré treinado para detecção de objetos primeiro e depois o modelo customizado





```
def prepare_image(file):
    img_path = ''
    img = image.load_img(img_path)
    img_array = image.img_to_array(img)
    img_array_expanded_dims = np.expand_dims(img_array, axis=0)
    return keras.applications.mobilenet_v2.preprocess_input(img_array_expanded_dims)
```



WARNING:tensorflow:From /usr/local/lib/python3.7/site-packages/tensorflow/py  
ocate\_with (from tensorflow.python.framework.ops) is deprecated and will be

# IoT Edge - Local Monitor

indoor-camera-8455

6/1/2019, 12:14:29 PM



Nothing found.

Found a **cat**



✓ Found Jam



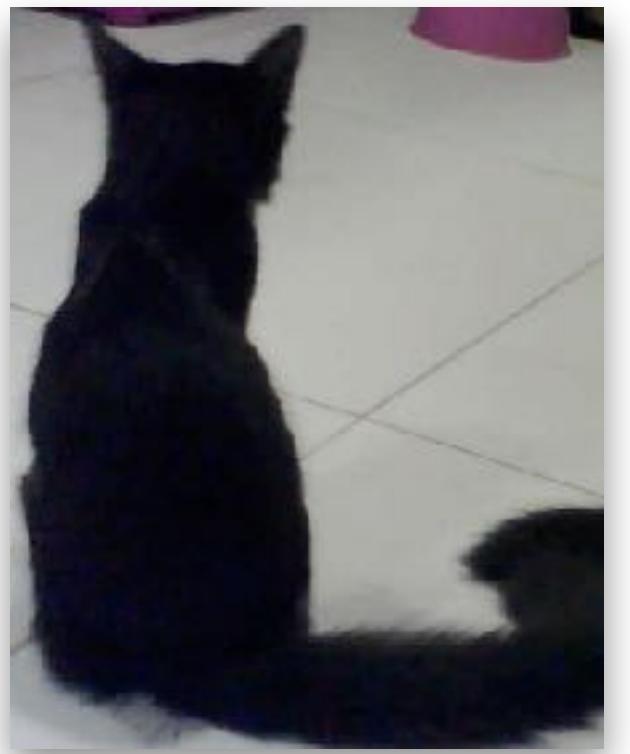
Found a **cat**



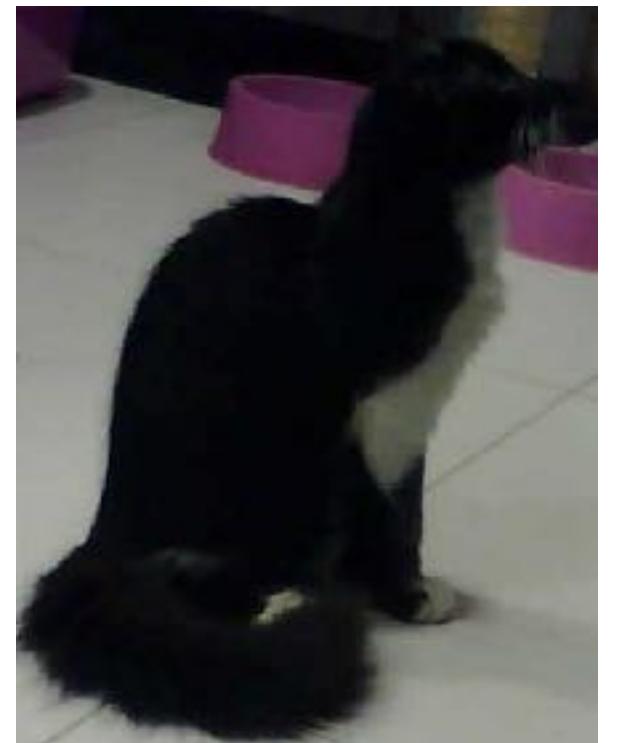
✓ Found **Berry**



✗ Found **Jam**



✓ Found **Muffin**



Found two **cats**



✓ Found **Berry**

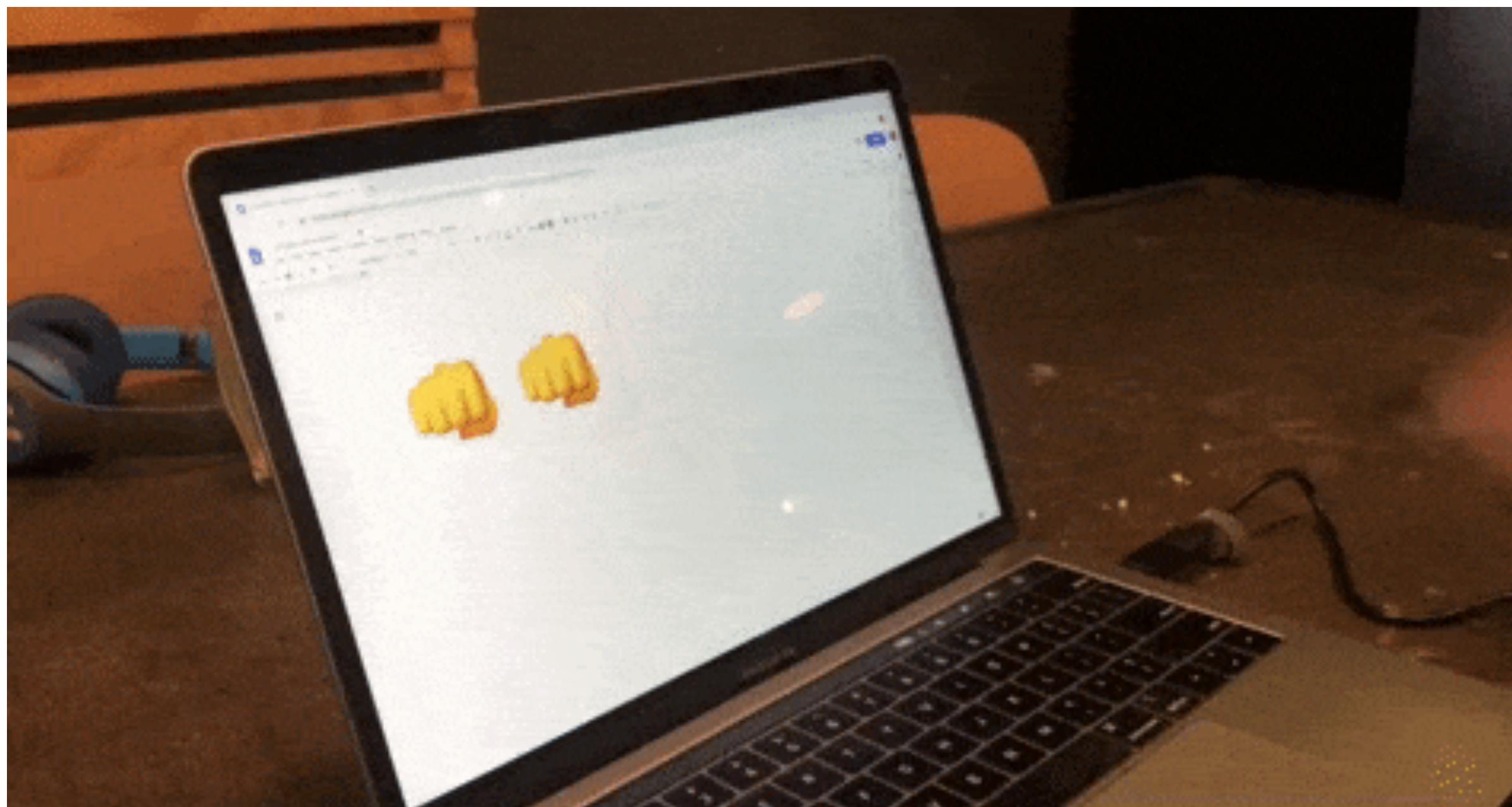


# Raquete de Tênis Inteligente



# Inspirações

Outros projetos de TinyML com Tensorflow



[medium.com/@devdevcharlie/play-street-fighter-with-body-movements-using-arduino-and-tensorflow-js-6b0e4734e118](https://medium.com/@devdevcharlie/play-street-fighter-with-body-movements-using-arduino-and-tensorflow-js-6b0e4734e118)

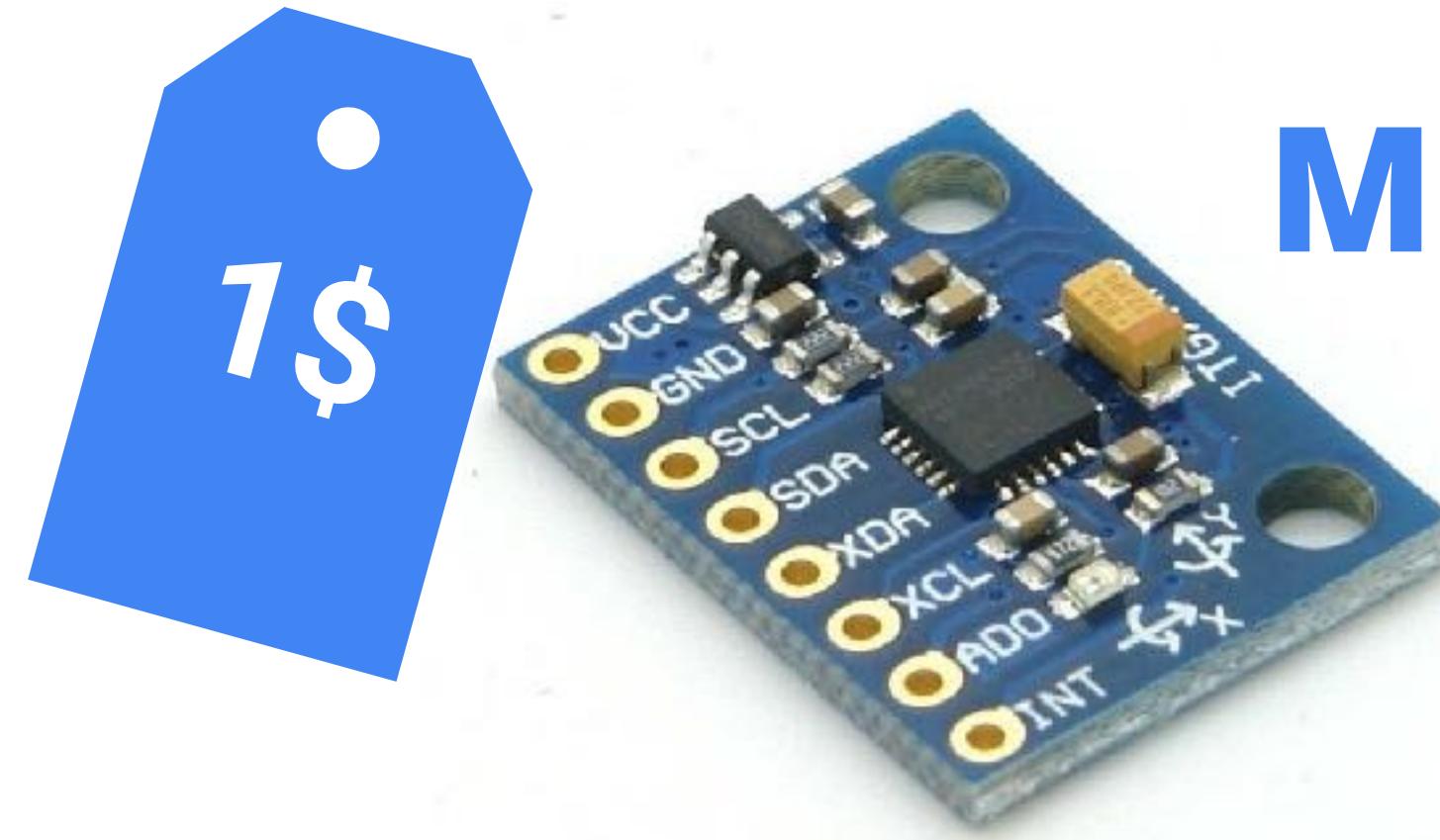
[blog.arduino.cc/2019/10/15/get-started-with-machine-learning-on-arduino/](https://blog.arduino.cc/2019/10/15/get-started-with-machine-learning-on-arduino/)

# Raquete Inteligente

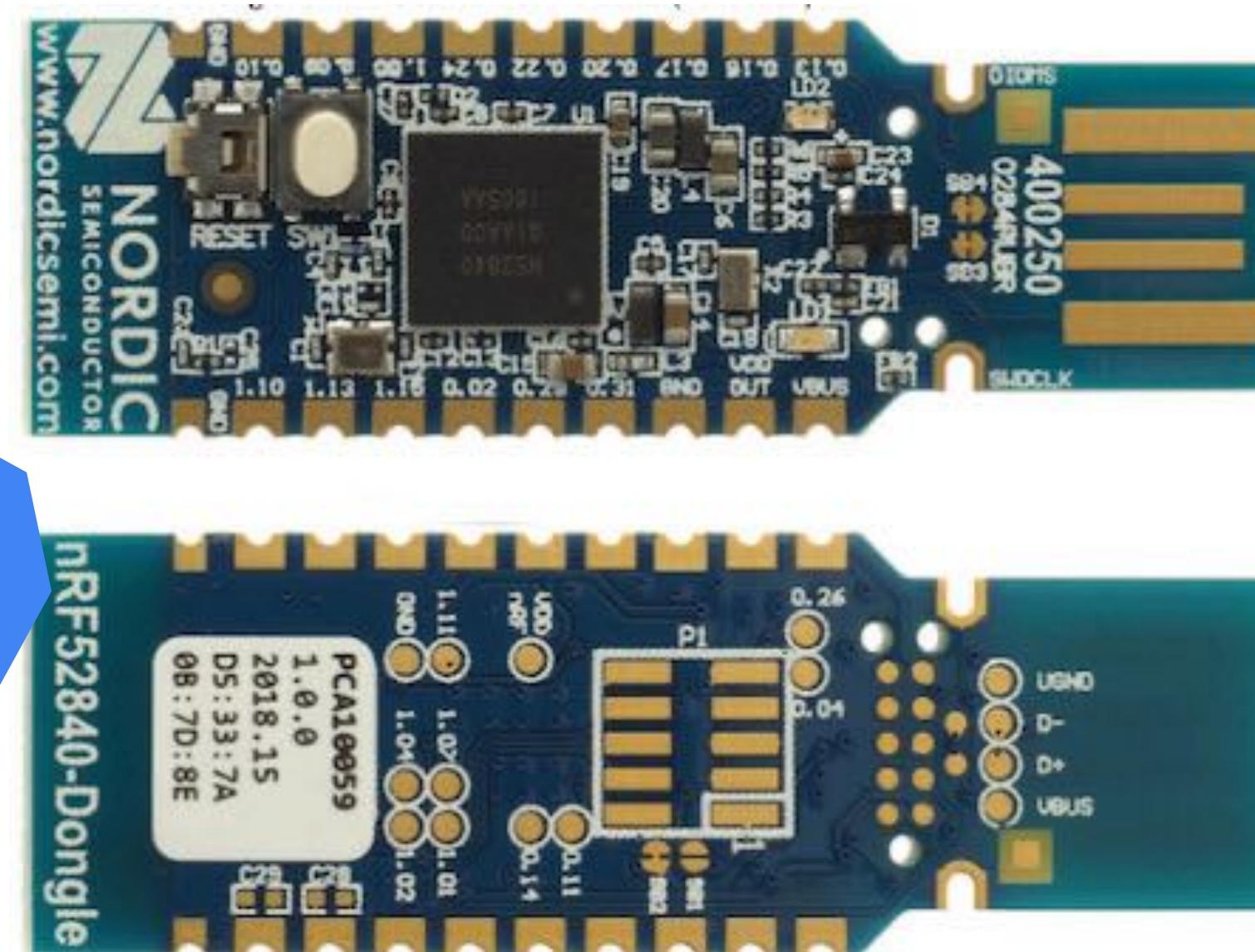
Classificar os diferentes movimentos do tênis - Backhand, forehand e serve

- Capturar dados de movimento
  - Acelerômetro e Giroscópio
  - Captura via Bluetooth
  - Janelas de 1 segundo de captura
    - $60 * 6$  pontos por segundo ( 16ms de intervalo ) = 360 de entrada
- Treinar modelo para classificar os movimentos
  - Análise espectral dos dados de movimento
  - Classificador com Redes Neurais
- Exportar modelo TFLite pra ser usado com TF Lite Micro

# MPU6050



1\$



10\$

# nrf52840 Dongle



20\$

# Particle Xenon

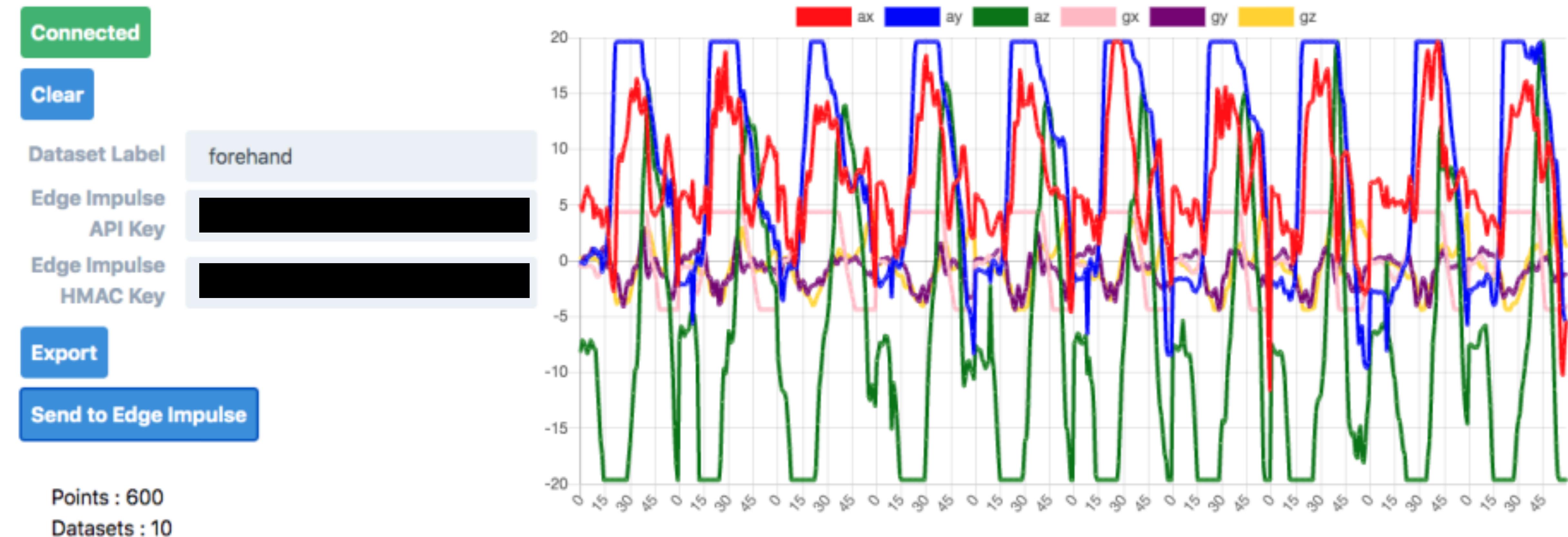




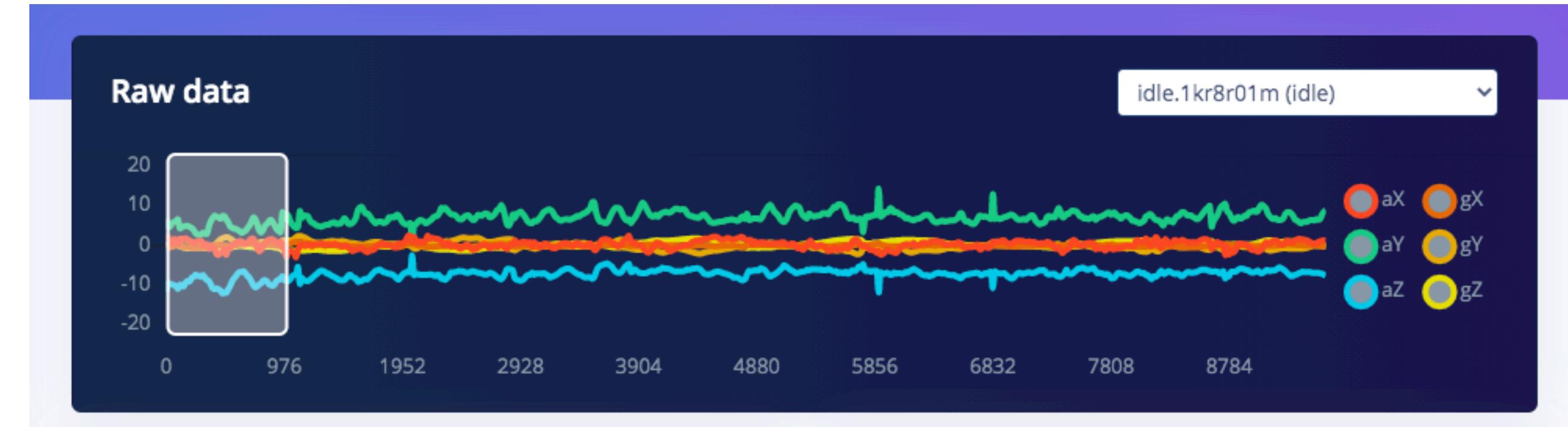
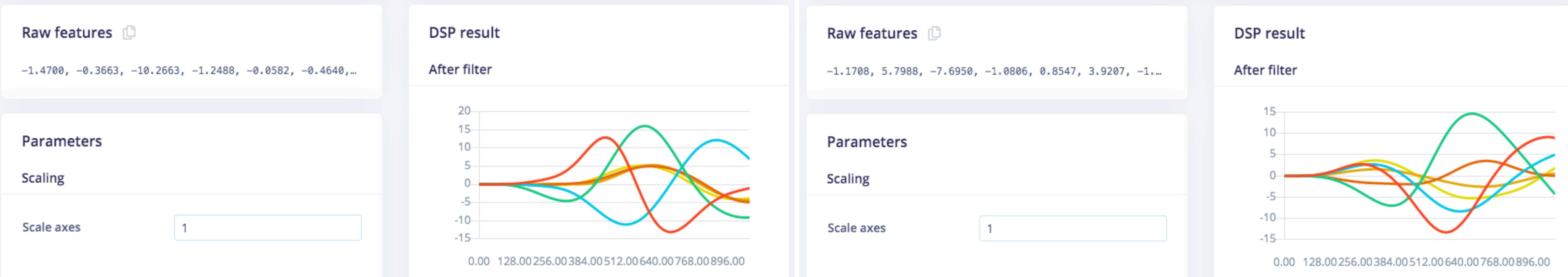
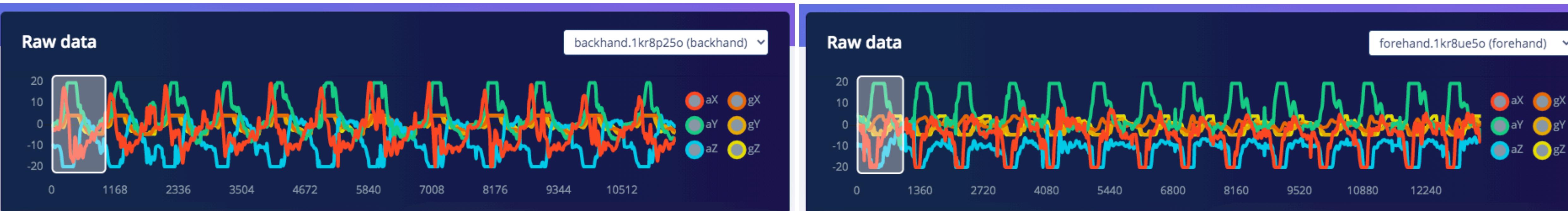
# Capturas dos dados

Captura via Web Bluetooth e envia para a plataforma Edge Impulse

TinyML Tennis Capture



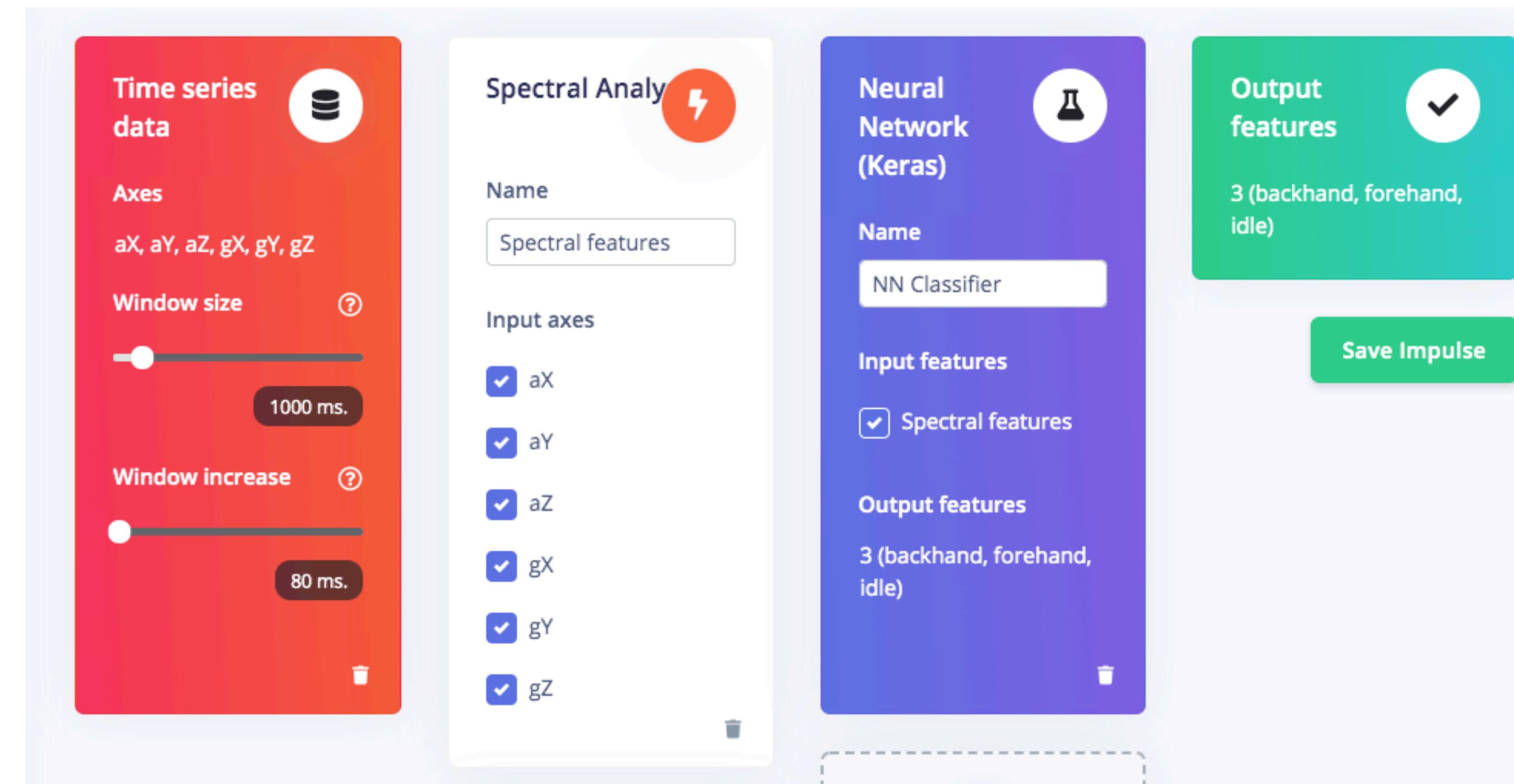
# EDGE IMPULSE



# Exportando o modelo

SDK da Edge Impulse usa Tensorflow Micro + DSP

- Analise espectral usa DSP
- Classificador usa TensorFlow Micro
- Estatísticas final do Modelo
  - 4kb de RAM e 28k de Flash 😍





**Backhand  
Led green**



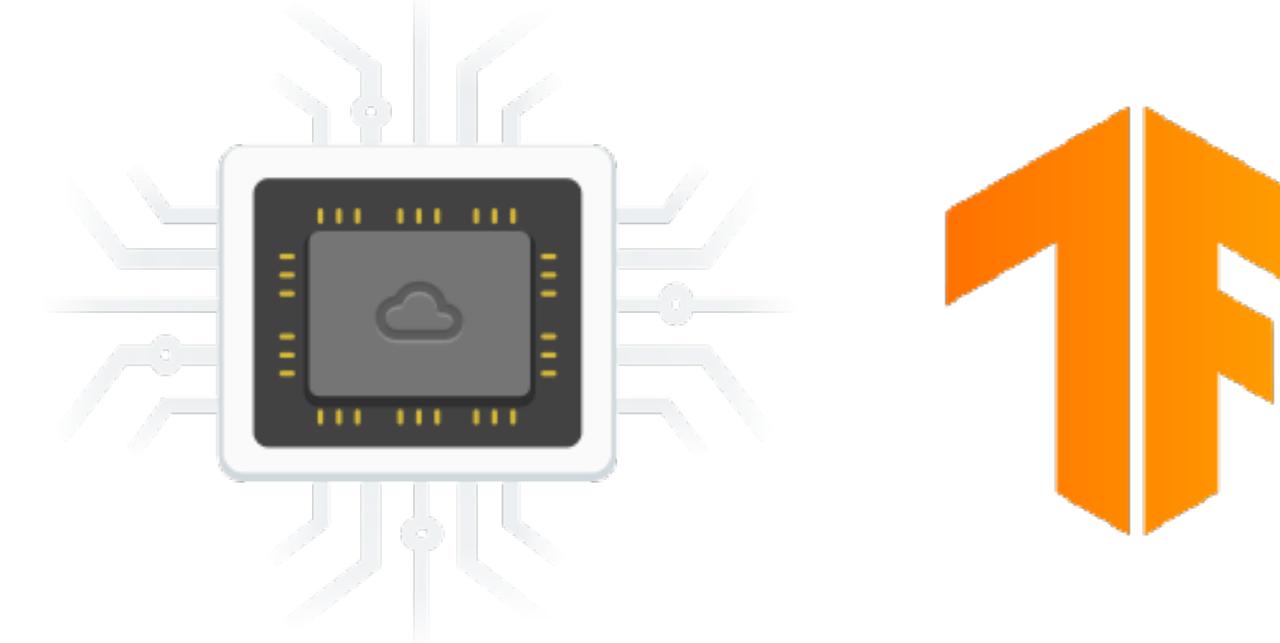
# Próximos passos

Melhorar o modelo e compartilhar os dados

- Capturar mais dados
  - Serve/Saques
  - Diferentes pessoas, alturas, estilos de jogo
  - Capturar dados em jogo/treino
- Disponibilizar o dataset
- Treinar modelo apenas com TensorFlow
  - O quanto o processamento DSP faz diferença ? ( provavelmente muita )
  - Usar com ESP32 - Mais acessível no Brasil

# Links e Referências

- Projetos
  - [github.com/alvarowolfx/gcloud-iot-edge-tensorflow](https://github.com/alvarowolfx/gcloud-iot-edge-tensorflow)
  - [github.com/alvarowolfx/tinyml-smart-tennis-sensor](https://github.com/alvarowolfx/tinyml-smart-tennis-sensor)
- Links
  - [edgeimpulse.com](https://edgeimpulse.com)
  - [www.tensorflow.org/lite/microcontrollers](https://www.tensorflow.org/lite/microcontrollers)
  - [blog.arduino.cc/2019/10/15/get-started-with-machine-learning-on-arduino/](https://blog.arduino.cc/2019/10/15/get-started-with-machine-learning-on-arduino/)
  - [medium.com/@devdevcharlie/play-street-fighter-with-body-movements-using-arduino-and-tensorflow-js-6b0e4734e118](https://medium.com/@devdevcharlie/play-street-fighter-with-body-movements-using-arduino-and-tensorflow-js-6b0e4734e118)



# OBRIGADO!

[AVIEBRANTZ.COM](http://AVIEBRANTZ.COM)



**Alvaro Viebrantz**

Google Developer Expert for IoT and Product Engineer at Leverege

[aviebrantz.com](http://aviebrantz.com)

 [@alvaroviebrantz](https://twitter.com/alvaroviebrantz)