

# Workshop on TinyML for Sustainable Development

## Advancements of TinyML in Santander, Colombia driven by the TinyML community and EdgeImpulse

Silvia Alejandra Sotelo López  
July 25<sup>th</sup> , 2024

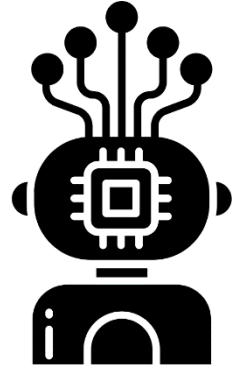


Acreditación Institucional  
ALTA CALIDAD • MULTICAMPUS  
Res. MEN No. 17228 del 24 de octubre de 2018 • 6 años • Vigilada Mineducación





Universidad  
Pontificia  
Bolivariana

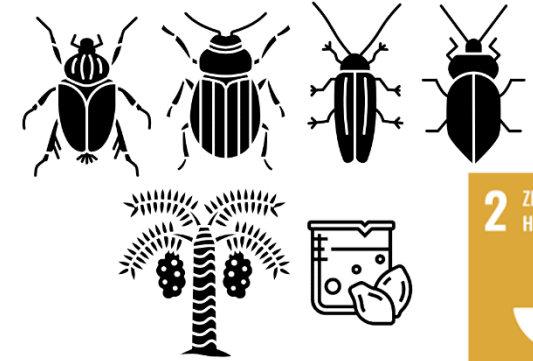


TinyML in  
Santander, CO

Graduation  
and  
postgraduate  
courses

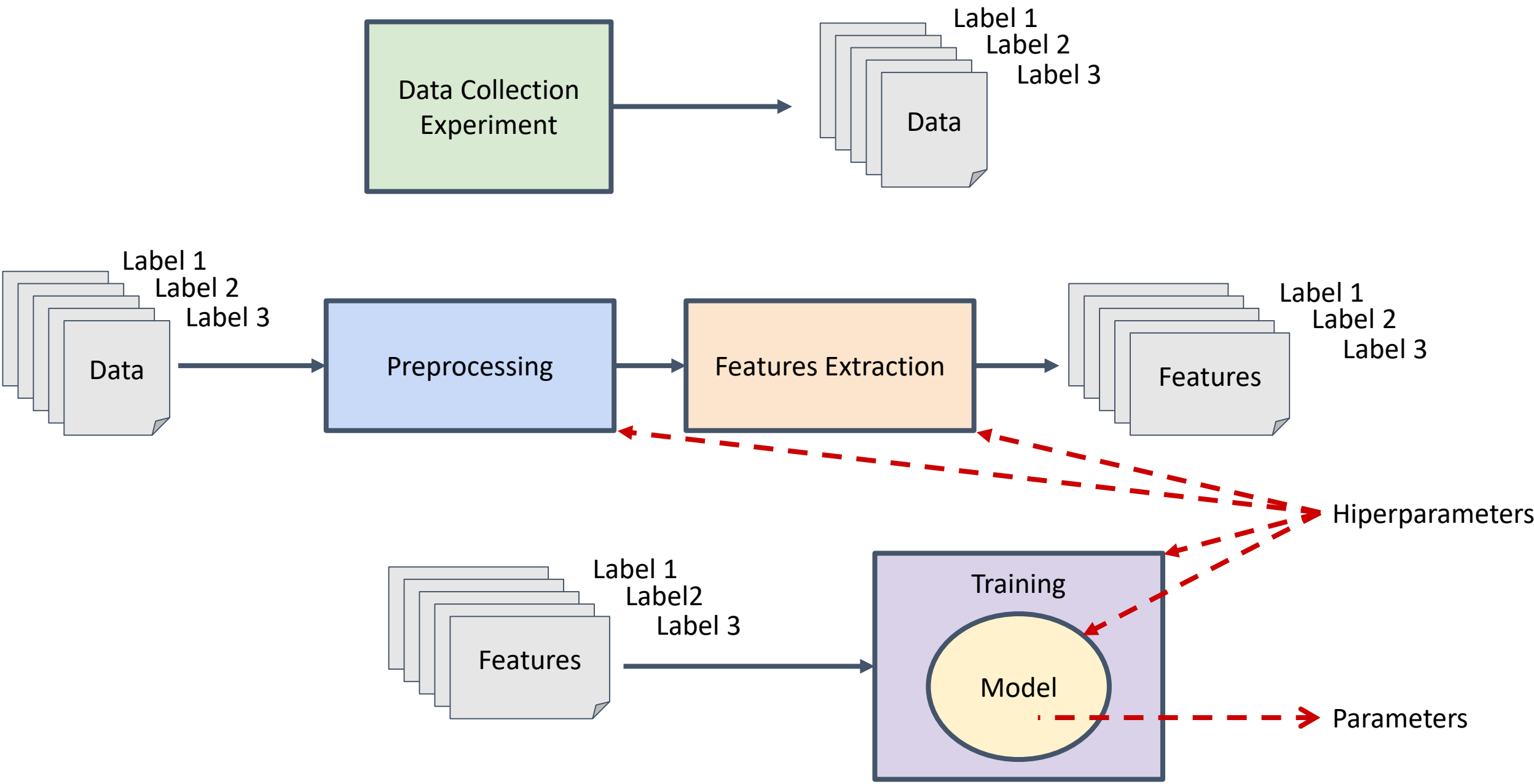
Student  
Research Hub

TinyML for Agroindustry

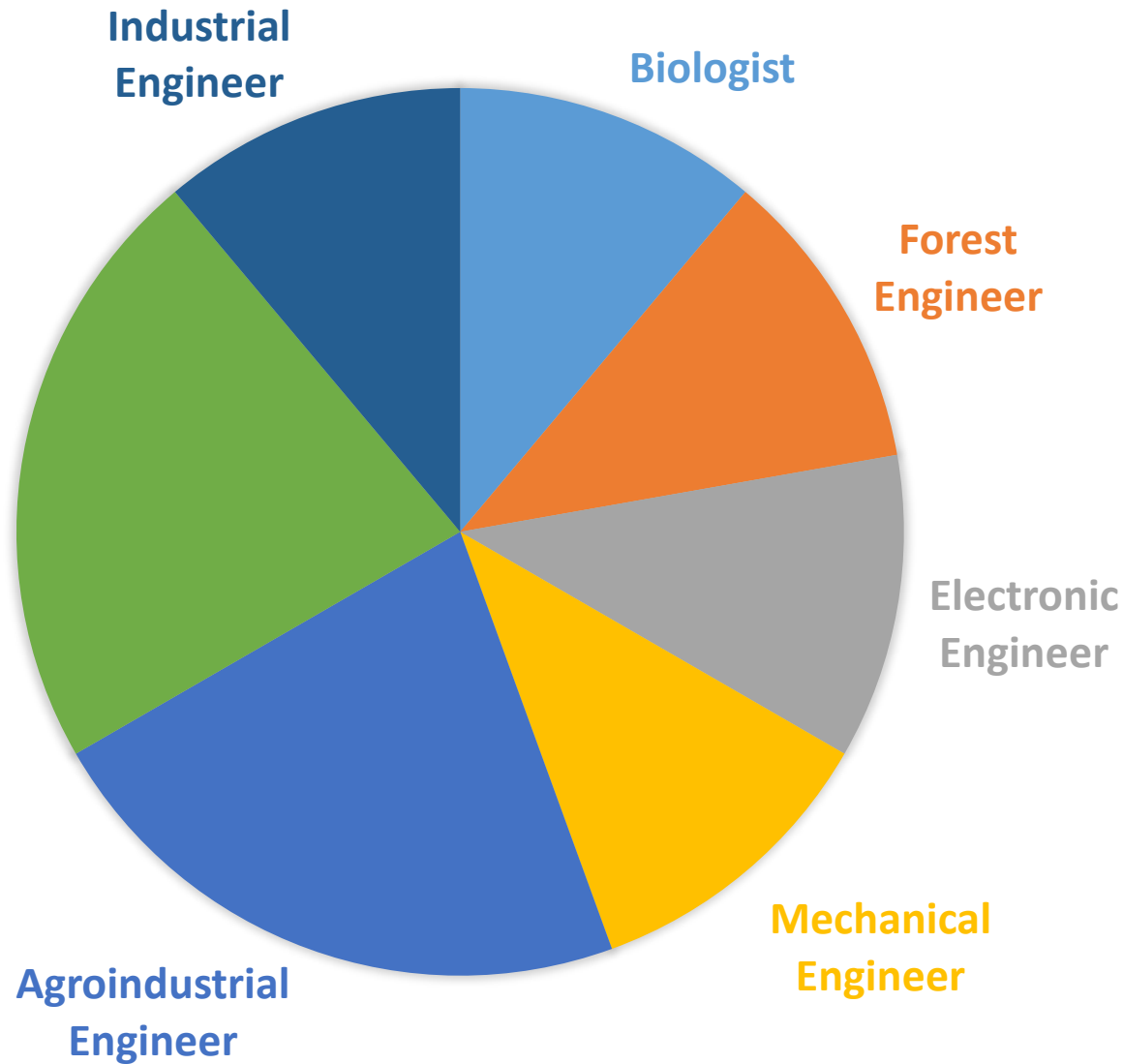


TinyML for Telemonitoring

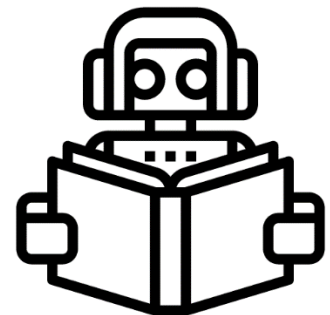




# Master's Course: MACHINE LEARNING APPLIED TO THE AGROINDUSTRY



**EDGE  
IMPULSE**



# AUTOMATED SYSTEM FOR THE CLASSIFICATION OF 4 CLASSES OF COLEOPTERS

Students: Johan Hernández, Hailyne Bohórquez, Fabián Osorio, Sandra López

Oil palm production contributes to Colombian economy, providing raw materials for various products such as palm oil, which is used in cooking, cosmetics, and industrial applications.



Red disease of Oil Palm

Significant economic losses

Due to the need to remove an infected trees.

Preventative strategies involve regular monitoring of plantations

**Preventive actions: Monitoring in search of the black palm weevil (*Rhynchophorus palmarum*), its main vector**



Black Weevil



Red Weevil



Bearded Weevil

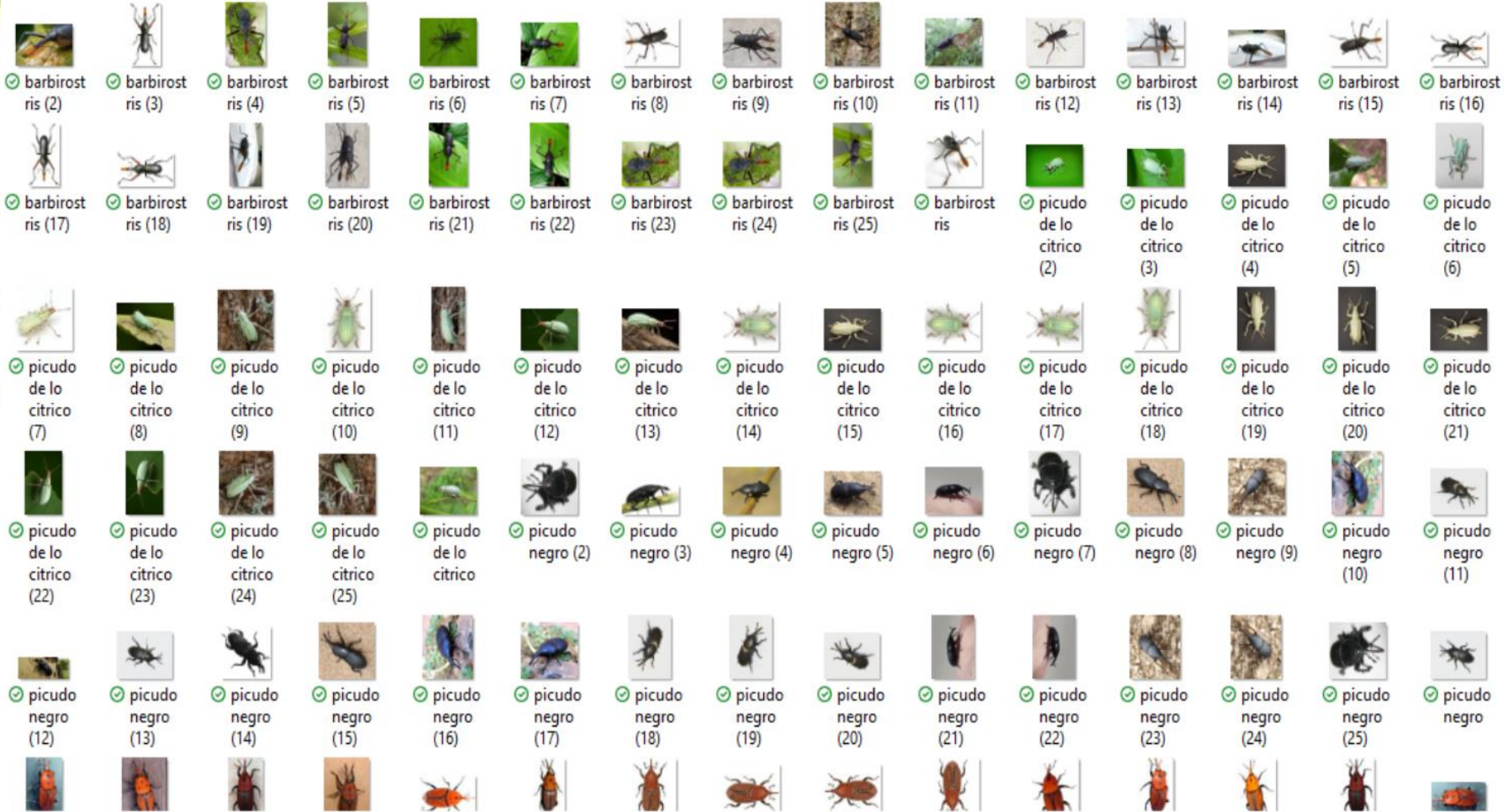


Citrus Root Weevil

# Dataset (25 images per class)




Red disease of Oil Palm



# Model I: Dense Neural Network

DATOS RECOLECTADOS  
100 artículos 

DIVISIÓN DE ENTREN...  
80 % / 20 % 

### Datos de imagen

Ejes de entrada  
imagen

Ancho de la ima...    Altura de imagen  
   

Modo de cambio de tamaño  
 

 Para obtener una precisión óptima con la transferencia de bloques de aprendizaje, utilice un tamaño de imagen de 96x96 o 160x160.

### Imagen

Nombre

Ejes de entrada (1)  
 imagen

### Clasificación

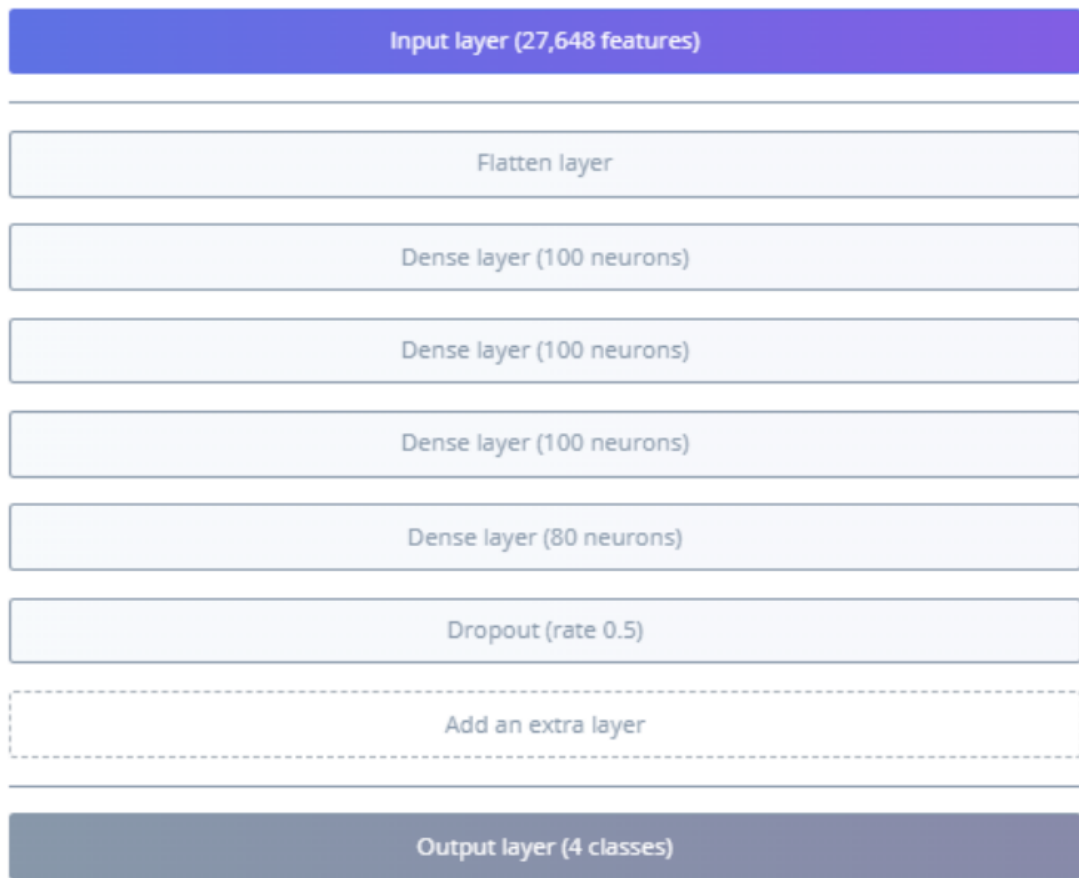
Nombre

Funciones de entrada  
 Imagen

Funciones de salida  
4 (Gorgojo Barbudo, PICUDO DE LO CITRICO, PICUDO NEGRO, PICUDO ROJO)

### Funciones de salida

4 (Gorgojo Barbudo, PICUDO DE LO CITRICO, PICUDO NEGRO, PICUDO ROJO)



## Model

Model version: [?](#)

Quantized (int8) ▾

### Last training performance (validation set)



ACCURACY  
100.0%



LOSS  
0,12

### Confusion matrix (validation set)

	GORGOJO BARBUDO	PICUDO DE LO CITR	PICUDO NEGRO	PICUDO ROJO
GORGOJO BARBUDO	100%	0%	0%	0%
PICUDO DE LO CITR	0%	100%	0%	0%
PICUDO NEGRO	0%	0%	100%	0%
PICUDO ROJO	0%	0%	0%	100%
F1 SCORE	1.00	1.00	1.00	1.00

### Resultados de las pruebas del modelo



EXACTITUD  
100.00%

	GORGOJO BARBUDO	PICUDO DE LO CÍTRICO	PICUDO NEGRO	PICUDO ROJO	INCIERTO
GORGOJO BARBUDO	100%	0%	0%	0%	0%
PICUDO DE LO CÍTRICO	0%	100%	0%	0%	0%
PICUDO NEGRO	0%	0%	100%	0%	0%
PICUDO ROJO	0%	0%	0%	100%	0%
PUNTUACIÓN F1	1.00	1.00	1.00	1.00	





Inferencing...

### Gorgojo Barbudo

Time per inference: 8 ms.

GORGJO B...PICUDO NE... PICUDO RO... PICUD...

5...	1.00	0.00	0.00	0.00
5...	1.00	0.00	0.00	0.00
5...	1.00	0.00	0.00	0.00
5...	1.00	0.00	0.00	0.00
5...	1.00	0.00	0.00	0.00



Inferencing...

### Picudo de lo Críticos

Time per inference: 5 ms.

GORGJO B...PICUDO NE... PICUDO RO... PICUD...

4...	0.00	0.00	0.00	1.00
4...	0.00	0.00	0.00	1.00
4...	0.00	0.00	0.00	1.00
4...	0.00	0.00	0.00	1.00
4...	0.00	0.00	0.00	1.00



Inferencing...

### Picudo Rojo

Time per inference: 5 ms.

GORGJO B...PICUDO NE... PICUDO RO... PICUD...

3...	0.00	0.00	1.00	0.00
3...	0.00	0.00	1.00	0.00
3...	0.00	0.00	1.00	0.00
3...	0.00	0.00	1.00	0.00
3...	0.00	0.00	1.00	0.00



Inferencing...

### Picudo Negro

Time per inference: 4 ms.

GORGJO B...PICUDO NE... PICUDO RO... PICUD...

1...	0.00	1.00	0.00	0.00
1...	0.00	1.00	0.00	0.00
1...	0.00	1.00	0.00	0.00
1...	0.00	1.00	0.00	0.00
1...	0.00	1.00	0.00	0.00

tphone.edgeimpulse.com

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## STUDENT RESEARCH HUB UPB-BUCARAMANGA



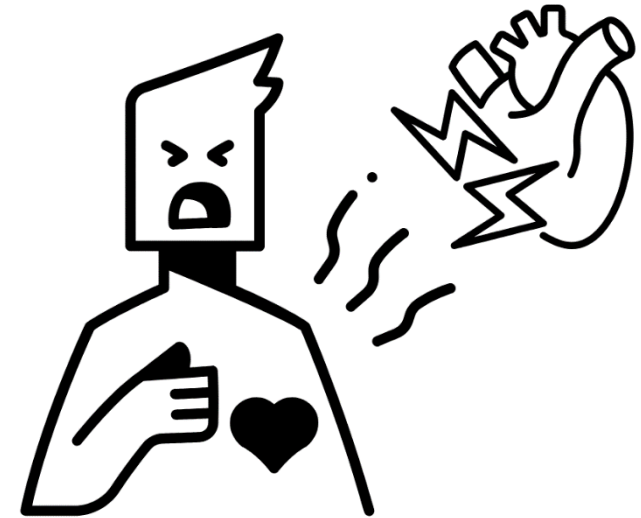
Emmanuel Angarita  
Brayan Arenas  
Alejandra Muñoz  
Santiago Pérez  
Julian Zabaleta

# AUTOMATIC WEIGHT ESTIMATION USING KINEMATIC SIGNALS OF THE GAIT CYCLE THROUGH NEURAL NETWORKS FOR TELEMONITORING OF SUBJECTS WITH HEART FAILURE

**Heart failure** is a chronic condition where the heart muscle is unable to pump enough blood to meet the body's needs.

Weight gain of ~ 3 kg or more in 2 days is a **WARNING SIGN** that something may be wrong!!!

## ICare



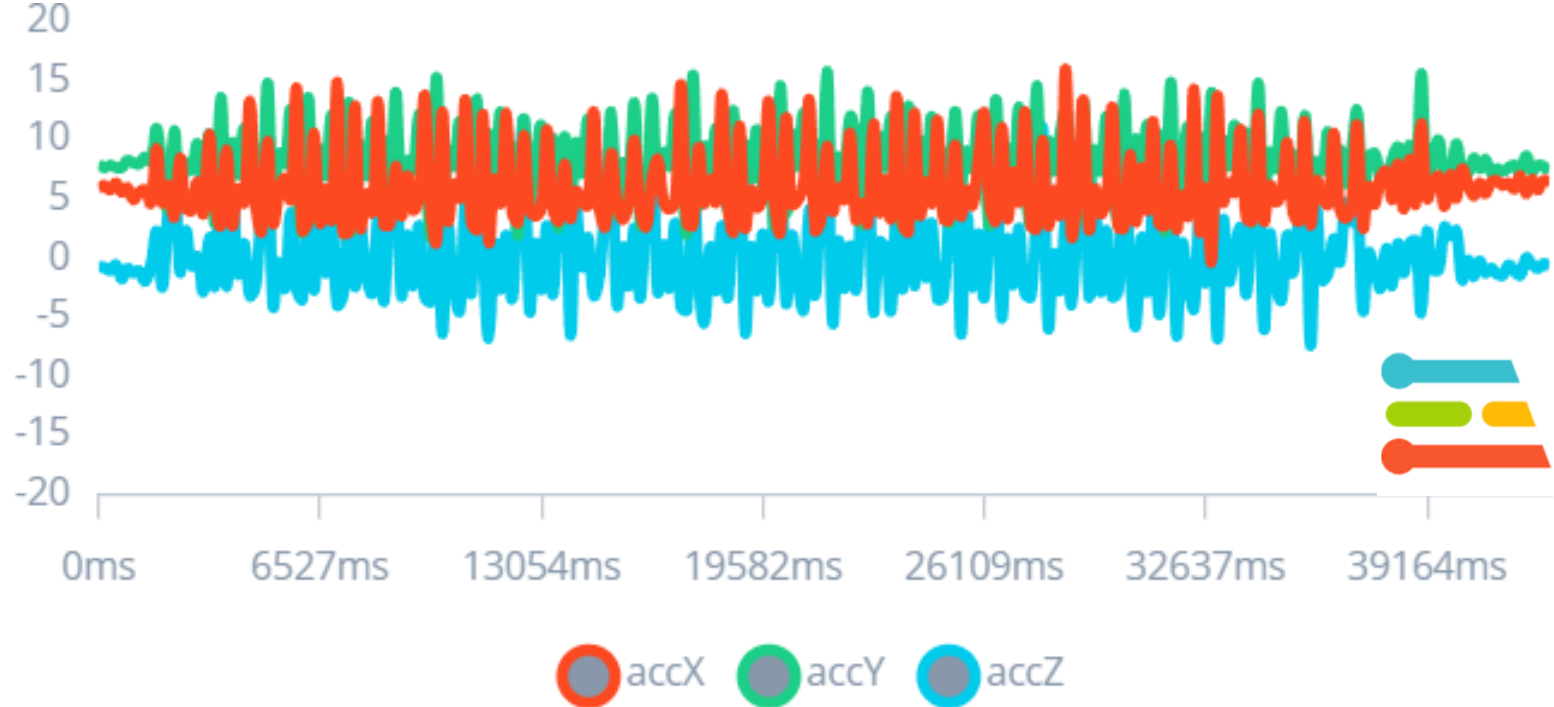
High probability to Re-hospitalization

Impact on patient outcomes and healthcare costs.

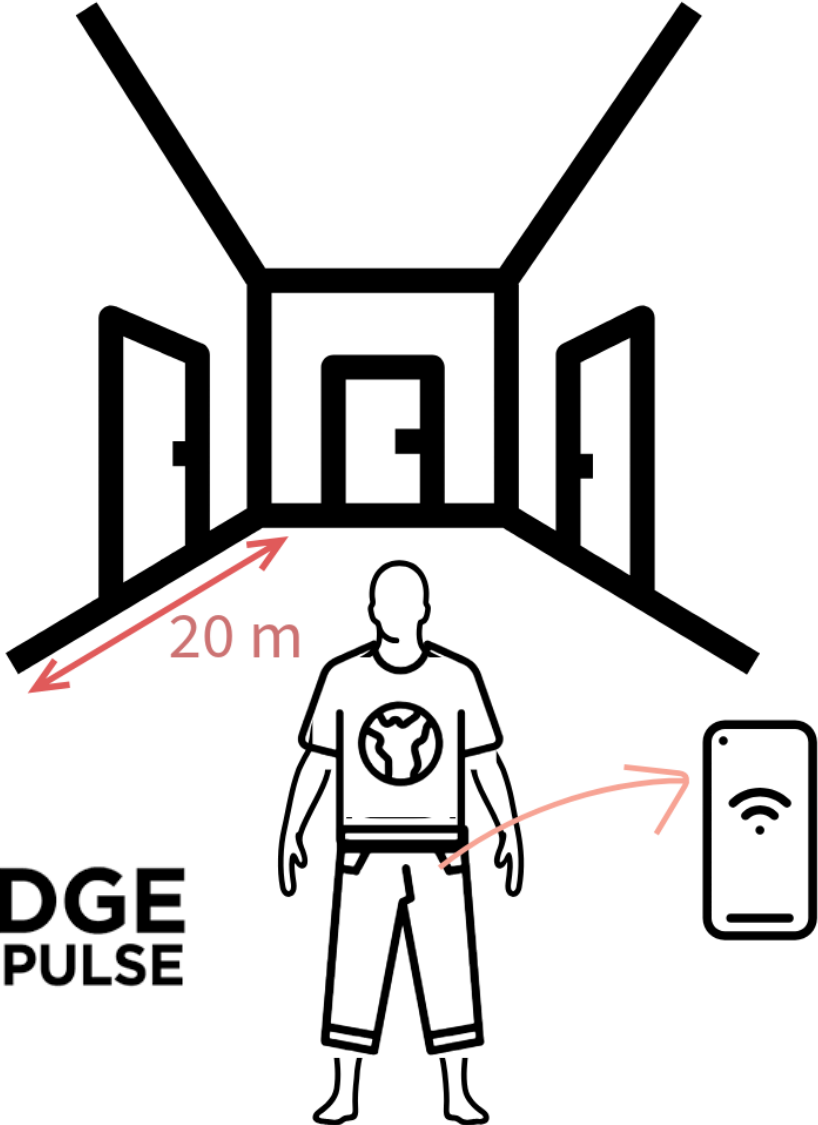
# Experiment Overview

With the mobile device in the front pocket of pant, the subject walked in a straight line for 3 minutes along a 20 m hallway.

20 UPB Students  
Mean: 68.5 kg  
Standard Deviation: 19.6 kg  
Minimum: 46.9 kg  
Maximum: 130 kg  
Sample frequency: 62.5 Hz



Students: Alejandra Muñoz, Daniel Russo, Santiago Pérez, Julián Zabaleta



**EDGE  
IMPULSE**

Accelerometer: TDK-Invensense icm4x6xx  
Moto G 30 y Moto G 22.  
Sample frequency: 62.5 Hz

**Time series data**

Input axes (3)  
accX, accY, accZ

Window size  
10.000 ms.

Window increase  
500 ms.

Frequency (Hz)  
62.5

Zero-pad data

**Spectral Analysis**

Name  
Spectral features

Input axes (3)  
 accX  
 accY  
 accZ

**Regression (Keras)**

Name  
Regression

Input features  
 Spectral features

Output features  
1 (Scalar value 46.9...130)

**Output features**

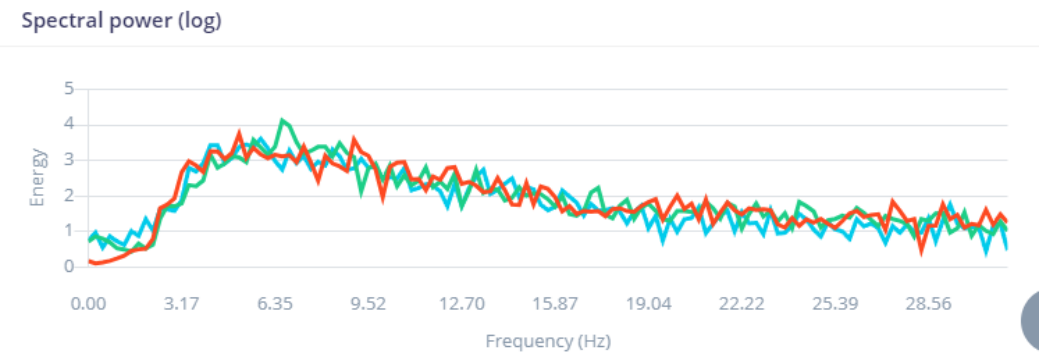
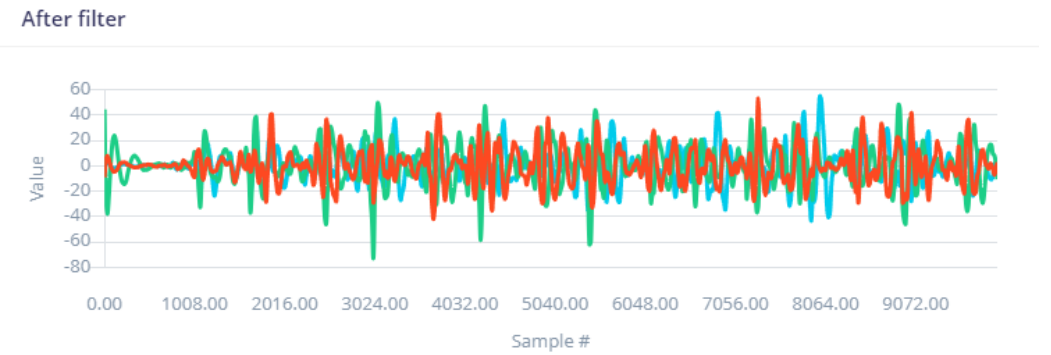
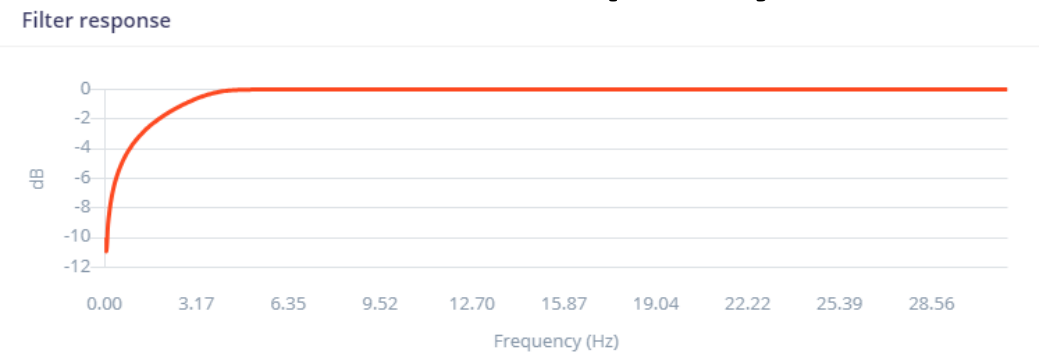
1 (Scalar value 46.9...130)

Save Impulse

# FEATURES

## Time domain

## Frequency domain



### Parameters

Autotune parameters

### Filter

Scale axes ②

Type ②

Cut-off frequency ②

Order ②

### Analysis

FFT length ②

Take log of spectrum? ②

Overlap FFT frames? ②

## Advanced training settings

Validation set size [?](#)

%

Split train/validation set on metadata key [?](#)

Profile int8 model [?](#)



## Neural network architecture

 Save

```
1 import tensorflow as tf
2 from tensorflow.keras.models import Sequential
3 from tensorflow.keras.layers import Dense, InputLayer, Dropout, Conv1D, Conv2D, Flatten, Reshape,
  MaxPooling1D, MaxPooling2D, AveragePooling2D, BatchNormalization, TimeDistributed, Permute, ReLU,
  Softmax
4 from tensorflow.keras.optimizers import Adam
5 EPOCHS = args.epochs or 500
6 LEARNING_RATE = args.learning_rate or 0.001
7 # this controls the batch size, or you can manipulate the tf.data.Dataset objects yourself
8 BATCH_SIZE = 32
9 train_dataset = train_dataset.batch(BATCH_SIZE, drop_remainder=False)
10 validation_dataset = validation_dataset.batch(BATCH_SIZE, drop_remainder=False)
11
12 # model architecture
13 model = Sequential()
14 model.add(Dense(100, activation='relu'))
15 model.add(Dense(80, activation='relu'))
16 model.add(Dense(50, activation='relu'))
17 model.add(Dense(classes, name='y_pred'))
18
19 # this controls the learning rate
20 opt = Adam(learning_rate=LEARNING_RATE, beta_1=0.9, beta_2=0.999)
21 callbacks.append(BatchLoggerCallback(BATCH_SIZE, train_sample_count, epochs=EPOCHS))
22
23 # train the neural network
24 model.compile(loss='mean_squared_error', optimizer=opt, metrics=None)
25 model.fit(train_dataset, epochs=EPOCHS, validation_data=validation_dataset, verbose=2, callbacks
  =callbacks)
26
27 # Use this flag to disable per-channel quantization for a model.
28 # This can reduce RAM usage for convolutional models, but may have
29 # an impact on accuracy.
30 disable_per_channel_quantization = False
```



LOSS

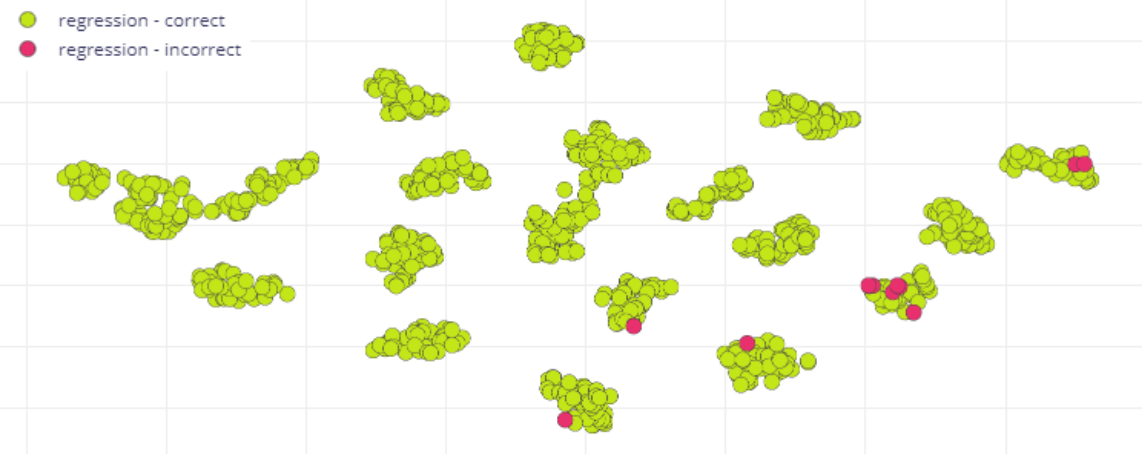
1,27

Metrics (validation set) 

METRIC	VALUE
Mean squared error <a href="#">?</a>	1.27
Mean absolute error <a href="#">?</a>	0.85
Explained variance score <a href="#">?</a>	1.00

Data explorer (full training set) [?](#)

Maximum absolute regression error is 3, [set thresholds](#).



On-device performance [?](#)

Engine: [?](#)

EON™ Compiler 



INFERRING TIME

16 ms.



PEAK RAM USAGE

1,8K



FLASH USAGE

57,3K



## Test data

Classify all



Set the 'expected outcome' for each sample to the desired outcome to automatically score the impulse. Maximum absolute regression error is 3, [set thresholds](#).

SAMPLE NAME	EXPECTED ...	LENGTH	ACCURACY	RESULT	ERROR ⓘ
caminata_g22_21.3g...	47.8	43s	30%	55.05 max, 51.28 ...	⋮
caminata_g30_20.3g...	60	45s	68%	66.49 max, 61.87 ...	⋮
80.caminata_g22_4....	80	45s	64%	83.13 max, 77.86 ...	⋮
78.caminata_g30_7....	78	53s	98%	80.89 max, 78.18 ...	⋮
63.caminata_g30_3....	63	55s	100%	64.96 max, 62.87 ...	⋮
79.5.caminata_g30_...	79.5	43s	97%	82.33 max, 80.24 ...	⋮
91.4.caminata_g22_...	91.4	46s	97%	94.04 max, 91.29 ...	⋮
130.caminata_g22_2...	130	46s	53%	137.8 max, 125.7...	⋮
63.4.caminata_g30_...	63.4	47s	89%	68.93 max, 64.86 ...	⋮



## Model testing output

(0)

### Results

Model version: ⓘ

Unoptimized (float32) ▾

ACCURACY ⓘ  
**73.63%**

MEAN SQUARED ERROR  
**14.03**

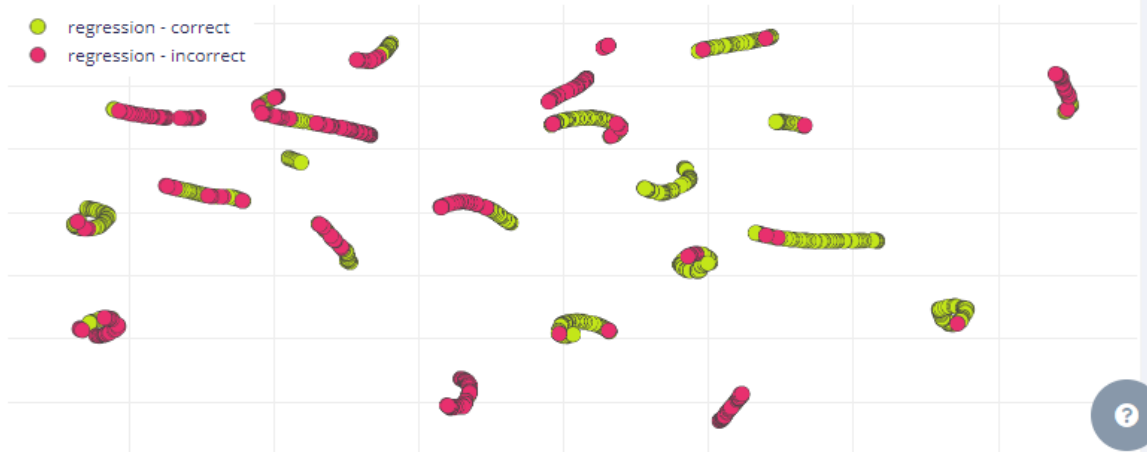
### Metrics for Regression

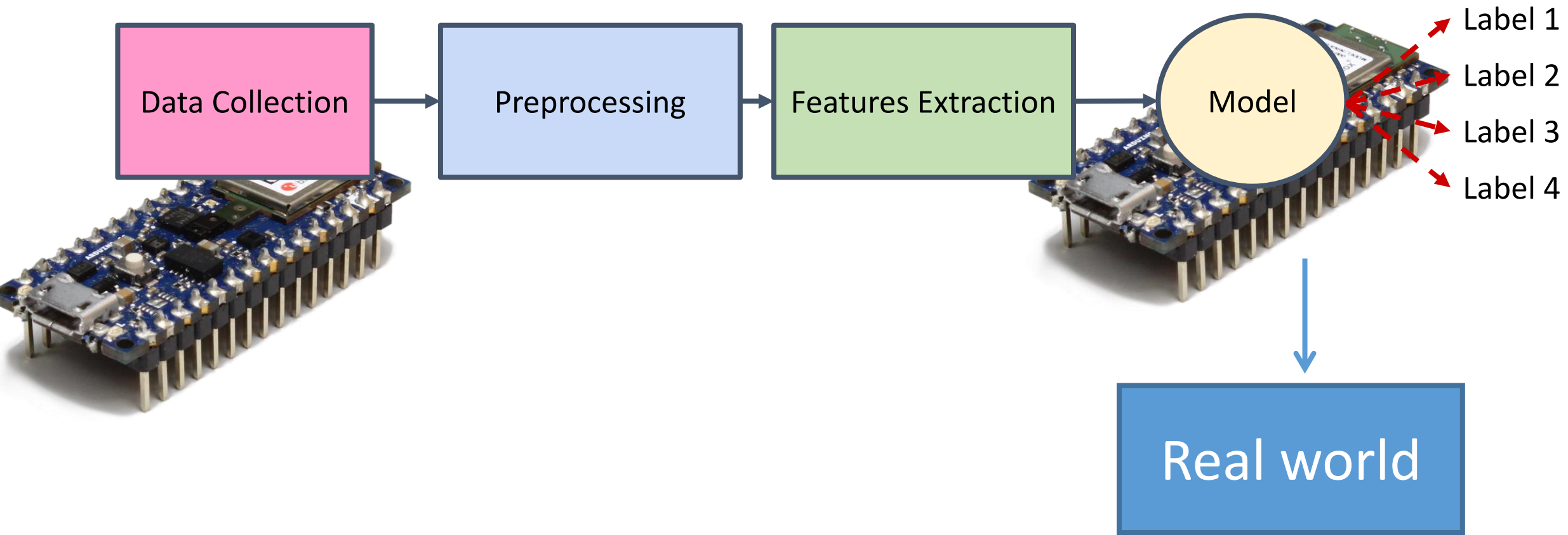


METRIC	VALUE
Mean squared error ⓘ	14.03
Mean absolute error ⓘ	2.37
Explained variance score ⓘ	0.96

### Feature explorer ⓘ

- regression - correct
- regression - incorrect







# SOLID WASTE CLASSIFIER IN THE ECOCAMPUS UPB-BUCARAMANGA USING TINYML



## ¡En la UPB Uso, pienso y clasifico!

Campaña Residuos  
Resolución 2384 de 2019.

**Blanco:** Residuos aprovechables  
(plástico, vidrio, papel, cartón, metales)

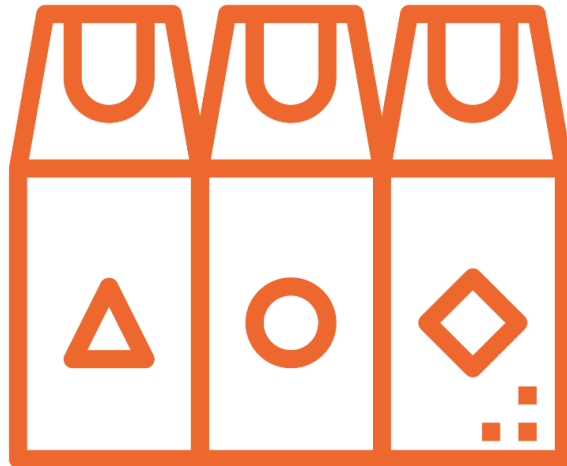
**Verde:** Residuos orgánicos aprovechables  
(cáscaras frutas y verduras, residuos de poda, resto de alimentos crudos)

**Negro:** Residuos NO aprovechables  
(Papel higiénico, servilletas, comida preparada)



#SinLímites

CAMPUS UPB BUCARAMANGA  
**GESTIÓN AMBIENTAL**  
www.upb.edu.co



# Muito Obrigada to the TinyML Community

For further information or collaboration opportunities,  
please contact me at [silvia.sotelo@upb.edu.co](mailto:silvia.sotelo@upb.edu.co)