

```
from google.colab import drive
drive.mount('/content/drive')
```

↗ Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
%cd "/content/drive/MyDrive/Semestre10/Computación/M2/datasetP"
ls
```

↗ /content/drive/MyDrive/Semestre10/Computación/M2/datasetP
augmented test train

```
#Imports
```

```
import matplotlib.pyplot as plt
import numpy as np
import os
import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
# Añadir los paths de las carpetas de 'train' y 'test'
train_dir = os.path.join('train')
test_dir = os.path.join('test')
```

```
# Empezamos con generación de datos en una variable llamada 'train_datagen'.
# Entrenamos el modelo al mismo tiempo que la generación de datos para no
# perder el tiempo y hacer uso óptimo de la RAM.
```

```
# Usamos la función ImageDataGenerator de TensorFlow
train_datagen = ImageDataGenerator(
    rescale = 1./255, # Reescalamos las imágenes
    rotation_range = 10, # Las rotamos levemente (10 grados)
    width_shift_range = 0.2, # Permitimos que se ensanche la imagen
    zoom_range = 0.3, # Hacemos zoom
    horizontal_flip = True) #Volteamos la imagen
```

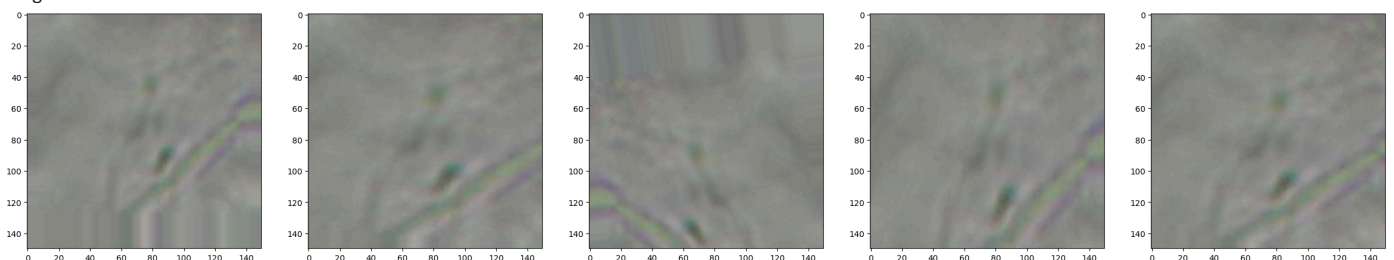
```
# Y BAM! Nuevos datos!
```

```
# Alimentamos los datos nuevos de 'train_datagen' a la función 'flow_from_directory'
# de TensorFlow
train_generator = train_datagen.flow_from_directory(
    train_dir, # le damos el path de entrenamiento
    target_size = (150, 150),
    # batch_size = 1 porque la RAM es un relajo XD
    batch_size = 1, # la cantidad de imágenes por conversión
    class_mode = 'categorical', # modo categórico porque tenemos
    # 5 clases datos
)
```

```
# Mostramos las nuevas fotos
plt.figure()
f, axarr = plt.subplots(1, 5, figsize=(30, 8))
```

```
for i in range(5) :
    axarr[i].imshow(train_generator[0][0][0])
```

↗ Found 7385 images belonging to 5 classes.
<Figure size 640x480 with 0 Axes>



```
train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size = (150, 150),
```

```

batch_size = 8,
class_mode = 'categorical',
)

```

```
images , labels = train_generator[0]
```

```

print(images.shape)
print(labels)

```

```

plt.figure()
#subplot(r,c) provide the no. of rows and columns
f, axarr = plt.subplots(1, images.shape[0], figsize=(30, 4))

```

```

for i in range(images.shape[0]) :
    axarr[i].imshow(images[i])

```

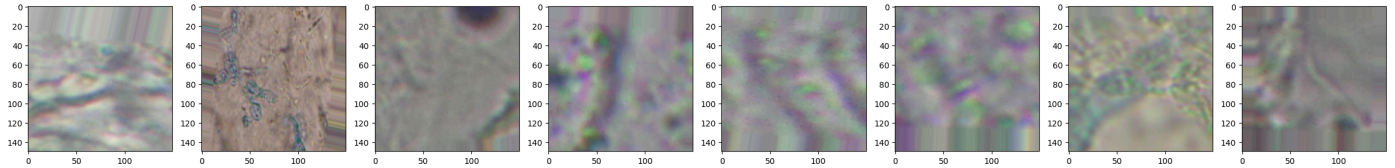
Found 7385 images belonging to 5 classes.

```

(8, 150, 150, 3)
[[0. 1. 0. 0. 0.]
 [0. 0. 0. 1. 0.]
 [0. 1. 0. 0. 0.]
 [1. 0. 0. 0. 0.]
 [1. 0. 0. 0. 0.]
 [0. 0. 1. 0. 0.]
 [0. 0. 0. 1. 0.]
 [0. 1. 0. 0. 0.]]

```

<Figure size 640x480 with 0 Axes>



```
path = "/content/drive/MyDrive/Semestre10/Computaci3n/M2/datasetP/"
```

```

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size = (150, 150),
    batch_size = 8,
    class_mode = 'binary',
    save_to_dir= path + '/augmented',
    save_prefix='aug',
    save_format='png'
)

```

Found 7385 images belonging to 5 classes.

```
# Inicio de la red neuronal convolutiva
```

```

from tensorflow.keras import optimizers
from tensorflow.keras import models
from tensorflow.keras import layers

```

```

model = models.Sequential()
model.add(layers.Conv2D(10, (3, 3), activation="relu", input_shape = (150,150,3)))
model.add(layers.Flatten())
model.add(layers.Dense(256,activation='relu'))
model.add(layers.Dense(1,activation='sigmoid'))


```

```
model.summary()
```

```

model.compile(loss='binary_crossentropy',
              optimizer=optimizers.RMSprop(learning_rate=2e-5),
              metrics=['acc'])

```

 Model: "sequential_6"

Layer (type)	Output Shape	Param #
conv2d_6 (Conv2D)	(None, 148, 148, 10)	280
flatten_6 (Flatten)	(None, 219040)	0
dense_12 (Dense)	(None, 256)	56,074,496
dense_13 (Dense)	(None, 1)	257

Total params: 56,075,033 (213.91 MB)

Trainable params: 56,075,033 (213.91 MB)

Non-trainable params: 0 (0.00 B)