

Entrenamiento de una red neuronal con

✓ arquitectura AlexNet usando Tensorflow / Keras

Comienza a programar o [generar](#) con IA.

```
import tensorflow as tf
from tensorflow.keras.datasets import cifar10
from tensorflow.keras.utils import to_categorical
import os

# Cargar dataset CIFAR-10
(x_train, y_train), (x_test, y_test) = cifar10.load_data()

# Convertir etiquetas a one-hot encoding
y_train = to_categorical(y_train, 10)
y_test = to_categorical(y_test, 10)

# Parámetros
BATCH_SIZE = 32
IMAGE_SIZE = (227, 227) # AlexNet espera 227x227

# Función de preprocesamiento (redimensiona imágenes y normaliza)
def preprocess_image(image, label):
    image = tf.image.resize(image, IMAGE_SIZE) / 255.0 # Normalización
    return image, label

# Crear datasets con `tf.data.Dataset`
train_dataset = (
    tf.data.Dataset.from_tensor_slices((x_train, y_train))
    .map(preprocess_image, num_parallel_calls=tf.data.experimental.AUTOTUNE)
    .batch(BATCH_SIZE)
    .prefetch(tf.data.experimental.AUTOTUNE)
)

test_dataset = (
    tf.data.Dataset.from_tensor_slices((x_test, y_test))
    .map(preprocess_image, num_parallel_calls=tf.data.experimental.AUTOTUNE)
    .batch(BATCH_SIZE)
    .prefetch(tf.data.experimental.AUTOTUNE)
)

from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense

def create_alexnet():
    model = Sequential([
        # Capa 1: Conv + ReLU + Pool
        Conv2D(96, (11, 11), strides=4, padding="same", input_shape=(227, 227, 3)),
        ReLU(),
        MaxPooling2D(pool_size=(3, 3), strides=2),

        # Capa 2: Conv + ReLU + Pool
        Conv2D(256, (5, 5), padding="same"),
        ReLU(),
        MaxPooling2D(pool_size=(3, 3), strides=2),
```

NotFoundError ✕

...



Explica el error:

NotFoundError: Graph execution error:

Detected at node StatefulPartitionedCall

File "<frozen runpy>", line 198, in _run

File "<frozen runpy>", line 88, in _run

File "/usr/local/lib/python3.11/dist-pa

File "/usr/local/lib/python3.11/dist-pa

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File "/usr/local/lib/python3.11/dist-pa

File "/usr/lib/python3.11/asyncio/base_

File "/usr/lib/python3.11/asyncio/base_

File "/usr/lib/python3.11/asyncio/event

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File "/usr/local/lib/python3.11/dist-pa

File "<ipython-input-4-fc452574286e>",

File "/usr/local/lib/python3.11/dist-pa

```

# Capas 3, 4, 5: Conv + ReLU
Conv2D(384, (3, 3), padding="same"),
ReLU(),
Conv2D(384, (3, 3), padding="same"),
ReLU(),
Conv2D(256, (3, 3), padding="same"),
ReLU(),
MaxPooling2D(pool_size=(3, 3), strides=2),

# Aplanar
Flatten(),

# Capas densas
Dense(4096),
ReLU(),
Dropout(0.5),
Dense(4096),
ReLU(),
Dropout(0.5),

# Capa de salida con 10 clases (CIFAR-10)
Dense(10, activation="softmax")
])
return model

# Ruta del archivo de pesos
WEIGHTS_PATH = "alexnet_weights.h5"
WEIGHTS_PATH = "alexnet_weights.weights.h5"

# Crear modelo
model = create_alexnet()

# Verificar si existe un modelo preentrenado
if os.path.exists(WEIGHTS_PATH):
    print(" ♦ Cargando pesos preentrenados...")
    model.load_weights(WEIGHTS_PATH)
else:
    print(" ♦ Entrenando desde cero...")

/usr/local/lib/python3.11/dist-packages/keras/src/layers/convoluti
super().__init__(activity_regularizer=activity_regularizer, **kw
♦ Cargando pesos preentrenados...

# Compilar modelo
model.compile(optimizer="adam", loss="categorical_crossentropy", metri

# Entrenar solo si no hay pesos preentrenados
if not os.path.exists(WEIGHTS_PATH):
    model.fit(train_dataset, epochs=15, validation_data=test_dataset)

# Guardar pesos entrenados
model.save_weights(WEIGHTS_PATH)
print(f"✅ Pesos guardados en {WEIGHTS_PATH}")

import numpy as np
np.argmax(model.predict(test_dataset)[0])

```

40/40 ————— 7s 164ms/step
7

File ["/usr/local/lib/python3.11/dist-pa](#)

File ["/usr/local/lib/python3.11/dist-pa](#)

File ["/usr/local/lib/python3.11/dist-pa](#)

could not find registered transfer manage
[[[{{node StatefulPartitionedCall}}]]

✦ Hypothesis

The error message "NotFoundError: could not find registered transfer manager for platform Host -- check target linkage" usually indicates an issue with TensorFlow's internal communication between devices, specifically between the host (CPU) and the device where the computation is supposed to happen (GPU or TPU).

In this case, it's likely that:

1. The pre-trained weights file (alexnet_weights.weights.h5) was saved in an environment with different hardware (CPU, GPU, TPU) or TensorFlow version than the current environment. This can cause compatibility issues.
2. The path to the weights file is incorrect, or the file is corrupted.

Suggested Changes

Ruta del archivo de pesos
WEIGHTS_PATH = "alexnet_weights.h5" # (
#WEIGHTS_PATH = "alexnet_weights.weight:

Crear modelo
model = create_alexnet()

Verificar si existe un modelo preentren
if os.path.exists(WEIGHTS_PATH):
 print(" ♦ Cargando pesos preentrena
 model.load_weights(WEIGHTS_PATH) # I
else:
 print(" ♦ Entrenando desde cero..."

[Use code with caution](#)

Explanation of changes:

```
import pandas as pd
```

```
pd.values(y_test)
```



AttributeError

Traceback (most recent

call last)

<ipython-input-8-a789fe06a185> in <cell line: 0>()

```
1 import pandas as pd
```

```
2
```

```
----> 3 pd.values(y_test)
```



Próximos pasos: [Explicar error](#)

```
y_test
```



```
array([[0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 1., 0.],
       [0., 0., 0., ..., 0., 1., 0.],
       ...,
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 1., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 1., 0., 0.]])
```

```
# Ruta del archivo de pesos
```

```
WEIGHTS_PATH = "alexnet_weights.h5" # Changed file extension to .h5
```

```
#WEIGHTS_PATH = "alexnet_weights.weights.h5"
```

```
# Crear modelo
```

```
model = create_alexnet()
```

```
# Verificar si existe un modelo preentrenado
```

```
if os.path.exists(WEIGHTS_PATH):
```

```
    print(" ♦ Cargando pesos preentrenados...")
```

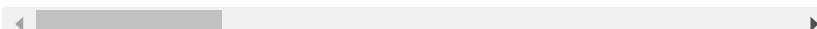
```
    model.load_weights(WEIGHTS_PATH) # Loads weights from the file if
```

```
else:
```

```
    print(" ♦ Entrenando desde cero...")
```



```
/usr/local/lib/python3.11/dist-packages/keras/src/layers/convoluti
super().__init__(activity_regularizer=activity_regularizer, **kw
 ♦ Entrenando desde cero...
```



```
# Cargar dataset CIFAR-10
```

```
(x_train, y_train), (x_test, y_test) = cifar10.load_data()
```

```
x_test[0]
```



```
ndarray (32, 32, 3) show data
```



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```
categories = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog',
```

1. **Change the file extension:** The original code attempts to load weights from a file named "alexnet_weights.weights.h5". This is an unusual naming convention for Keras/TensorFlow weight files. Typically, the file extension is simply ".h5". By changing WEIGHTS_PATH to "alexnet_weights.h5", we can ensure the code looks for the correct file extension which is commonly used for saving model weights in Keras.
2. **Ensure the weights file exists:** Double-check that the weights file ("alexnet_weights.h5") actually exists in the specified location. If you are using pre-trained weights, make sure you have downloaded the file and placed it in the correct directory.
3. **Save with the correct extension:** When saving the weights, make sure to use the ".h5" extension:



```
model.save_weights(WEIGHTS_PATH)
```

[Use code with caution](#)

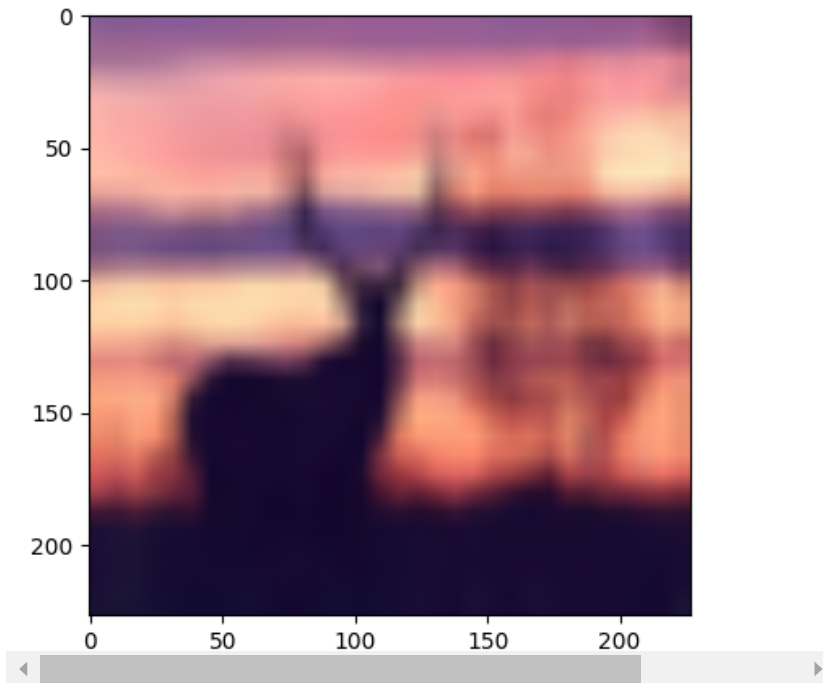
By making these changes, we can address the potential issues related to file paths and extensions. After making the changes above and executing the code, if the error still shows up it is due to a hardware conflict of your environment with the hardware where the weights were originally saved. You should train the model from scratch in this case.

Califica esta respuesta

```
im=tf.image.resize(x_test[211], IMAGE_SIZE) / 255.0
```

```
import matplotlib.pyplot as plt
plt.imshow(im)
```

```
<matplotlib.image.AxesImage at 0x7a0aeadf8350>
```



```
np.argmax(model.predict(np.expand_dims(im, axis=0)))
```

```
1/1 ————— 0s 32ms/step
9
```

```
x_test[3]
```

```
ndarray (32, 32, 3) show data
```



```
x_test[11]
```

```
ndarray (32, 32, 3) show data
```

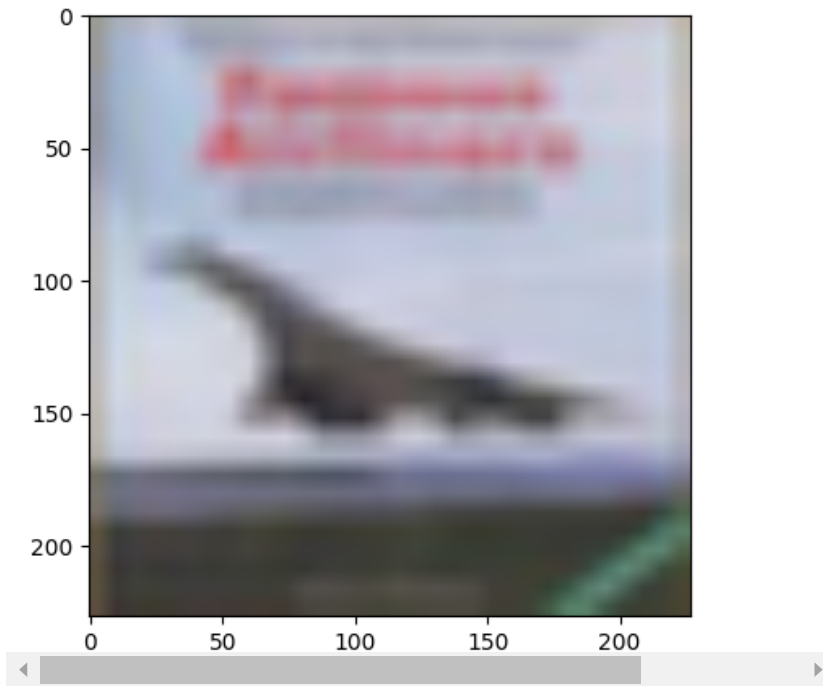


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```
im=tf.image.resize(x_test[3], IMAGE_SIZE) / 255.0
```

```
import matplotlib.pyplot as plt
plt.imshow(im)
```

 <matplotlib.image.AxesImage at 0x7dac3422e410>



```
np.argmax(model.predict(np.expand_dims(im, axis=0)))
```

 1/1 ————— 0s 39ms/step
0


```
import numpy as np
categories[np.argmax(model.predict(np.expand_dims(im, axis=0)))]
```

 1/1 ————— 0s 32ms/step

```
# Esto es para agregar un corchete. Importante, debe ser array o tensor.
import numpy as np
categories[np.argmax(model.predict(np.array([im])))]
```

 1/1 ————— 0s 31ms/step

```
im
```

 <tf.Tensor: shape=(227, 227, 3), dtype=float32, numpy=

```
array([[0.5294118, 0.37254903, 0.5921569 ],
       [0.5294118, 0.37254903, 0.5921569 ],
       [0.5294118, 0.37254903, 0.5921569 ],
       ...,
       [0.46666667, 0.2627451, 0.44313726],
       [0.46666667, 0.2627451, 0.44313726],
       [0.46666667, 0.2627451, 0.44313726]],

      [[0.5294118, 0.37254903, 0.5921569 ],
       [0.5294118, 0.37254903, 0.5921569 ],
       [0.5294118, 0.37254903, 0.5921569 ],
       ...,
       [0.46666667, 0.2627451, 0.44313726],
       [0.46666667, 0.2627451, 0.44313726],
       [0.46666667, 0.2627451, 0.44313726]],

      [[0.5294118, 0.37254903, 0.5921569 ],
```

```
[0.5294118 , 0.37254903, 0.5921569 ],
[0.5294118 , 0.37254903, 0.5921569 ],
...,
[0.46666667, 0.2627451 , 0.44313726],
[0.46666667, 0.2627451 , 0.44313726],
[0.46666667, 0.2627451 , 0.44313726]],

...,

[[0.09803922, 0.07058824, 0.20784314],
[0.09803922, 0.07058824, 0.20784314],
[0.09803922, 0.07058824, 0.20784314],
...,
[0.10980392, 0.07843138, 0.21568628],
[0.10980392, 0.07843138, 0.21568628],
[0.10980392, 0.07843138, 0.21568628]],

[[0.09803922, 0.07058824, 0.20784314],
[0.09803922, 0.07058824, 0.20784314],
[0.09803922, 0.07058824, 0.20784314],
...,
[0.10980392, 0.07843138, 0.21568628],
[0.10980392, 0.07843138, 0.21568628],
[0.10980392, 0.07843138, 0.21568628]],

[[0.09803922, 0.07058824, 0.20784314],
[0.09803922, 0.07058824, 0.20784314],
[0.09803922, 0.07058824, 0.20784314],
...,
[0.10980392, 0.07843138, 0.21568628],
[0.10980392, 0.07843138, 0.21568628],
[0.10980392, 0.07843138, 0.21568628]]], dtype=float32)>
```

```
imexp=np.expand_dims(im, axis=0)
```

```
imexp.shape
```

```
→ (1, 227, 227, 3)
```

Ingresa una instrucción aquí



0/2000

Es posible que las respuestas muestren información ofensiva o imprecisa que no represente las opiniones de Google. [Más información](#)

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