> 1) Configuración del entorno

[] → 2 celdas ocultas

2) Cargar y explorar los datos

```
ruta_train = '/content/drive/MyDrive/Ev01/data/train'
ruta_test = '/content/drive/MyDrive/Ev01/data/test'
(x_train, y_train), (x_test, y_test) = cifar10.load_data()
x_train = x_train.astype('float32') / 255.0
x_{test} = x_{test.astype('float32')} / 255.0
y_train_cat = to_categorical(y_train, 10)
y_test_cat = to_categorical(y_test, 10)
print(f"x\_train \ shape: \{x\_train.shape\}")
print(f"y_train shape: {y_train_cat.shape}")
x_train shape: (50000, 32, 32, 3)
     y_train shape: (50000, 10)
model = models.Sequential([
   layers.Conv2D(32, (3, 3), activation='relu', input_shape=(32, 32, 3)),
    layers.MaxPooling2D(pool_size=(2, 2)),
   layers.BatchNormalization(),
   layers.Dropout(0.3),
    layers.Conv2D(64, (3, 3), activation='relu'),
   layers.MaxPooling2D(pool_size=(2, 2)),
    layers.BatchNormalization(),
    layers.Dropout(0.3),
   layers.Conv2D(128, (3, 3), activation='relu'),
    layers.BatchNormalization(),
    layers.Dropout(0.3),
    layers.Conv2D(256, (3, 3), activation='relu'),
    layers.BatchNormalization(),
    layers.Dropout(0.3),
    layers.Flatten(),
    layers.Dense(512, activation='relu'),
    layers.Dropout(0.5),
    layers.Dense(10, activation='softmax')
])
model.summary()
```

/usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/super().__init__(activity_regularizer=activity_regularizer, **kwargs)

Model: "sequential_5"

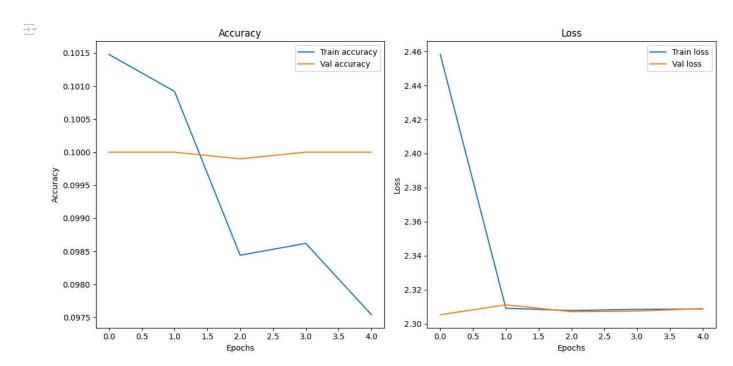
Layer (type)	Output Shape	Param #
conv2d_17 (Conv2D)	(None, 30, 30, 32)	896
max_pooling2d_15 (MaxPooling2D)	(None, 15, 15, 32)	0
batch_normalization_11 (BatchNormalization)	(None, 15, 15, 32)	128
dropout_14 (Dropout)	(None, 15, 15, 32)	0
conv2d_18 (Conv2D)	(None, 13, 13, 64)	18,496
max_pooling2d_16 (MaxPooling2D)	(None, 6, 6, 64)	0
batch_normalization_12 (BatchNormalization)	(None, 6, 6, 64)	256
dropout_15 (Dropout)	(None, 6, 6, 64)	0
conv2d_19 (Conv2D)	(None, 4, 4, 128)	73,856
batch_normalization_13 (BatchNormalization)	(None, 4, 4, 128)	512
dropout_16 (Dropout)	(None, 4, 4, 128)	0
conv2d_20 (Conv2D)	(None, 2, 2, 256)	295,168
batch_normalization_14 (BatchNormalization)	(None, 2, 2, 256)	1,024
dropout_17 (Dropout)	(None, 2, 2, 256)	0
flatten_5 (Flatten)	(None, 1024)	0
dense_10 (Dense)	(None, 512)	524,800
dropout_18 (Dropout)	(None, 512)	0
dense_11 (Dense)	(None, 10)	5,130

Total params: 920,266 (3.51 MB)
Trainable params: 919,306 (3.51 MB)

```
initial_learning_rate = 0.001
{\tt lr\_schedule = tf.keras.optimizers.schedules.ExponentialDecay(}
    initial_learning_rate, decay_steps=100000, decay_rate=0.96, staircase=True)
\verb|model.compile(optimizer=tf.keras.optimizers.Adam(learning\_rate=lr\_schedule)|,\\
              loss='categorical_crossentropy',
              metrics=['accuracy'])
datagen = tf.keras.preprocessing.image.ImageDataGenerator(
    rotation_range=20,
    width_shift_range=0.2,
    height shift range=0.2,
    horizontal_flip=True,
    zoom_range=0.2,
    shear_range=0.2,
datagen.fit(x_train)
history = model.fit(datagen.flow(x_train, y_train_cat, batch_size=128),
      epochs=30,
      validation_data=(x_test, y_test_cat),
      \verb|callbacks=[tf.keras.callbacks.EarlyStopping(monitor='val_loss', patience=3, restore\_best\_weights=True)||
→ Epoch 1/30
     /usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` cl
       self._warn_if_super_not_called()
                                 - 45s 91ms/step - accuracy: 0.2356 - loss: 2.5210 - val_accuracy: 0.1187 - val_loss: 4.4226
     391/391 -
     Epoch 2/30
     391/391 -
                                 - 30s 76ms/step - accuracy: 0.3890 - loss: 1.6817 - val_accuracy: 0.4318 - val_loss: 1.5459
     Epoch 3/30
     391/391 -
                                 – 31s 78ms/step - accuracy: 0.4427 - loss: 1.5474 - val_accuracy: 0.5032 - val_loss: 1.3937
     Epoch 4/30
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.4721 - loss: 1.4687 - val_accuracy: 0.5611 - val_loss: 1.2249
     Epoch 5/30
```

```
391/391 •
                                 - 29s 74ms/step - accuracy: 0.4922 - loss: 1.4159 - val_accuracy: 0.5968 - val_loss: 1.1100
     Epoch 6/30
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.5094 - loss: 1.3672 - val_accuracy: 0.5716 - val_loss: 1.1927
     Epoch 7/30
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.5268 - loss: 1.3232 - val_accuracy: 0.6281 - val_loss: 1.0484
     Epoch 8/30
                                 - 29s 75ms/step - accuracy: 0.5466 - loss: 1.2859 - val accuracy: 0.5410 - val loss: 1.3163
     391/391
     Epoch 9/30
     391/391 -
                                 - 42s 78ms/step - accuracy: 0.5472 - loss: 1.2662 - val accuracy: 0.5213 - val loss: 1.4209
     Epoch 10/30
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.5615 - loss: 1.2324 - val_accuracy: 0.6186 - val_loss: 1.0827
test_loss, test_acc = model.evaluate(x_test, y_test_cat)
print(f"Test accuracy: {test_acc * 100:.2f}%")
y_pred = model.predict(x_test)
y_pred_labels = np.argmax(y_pred, axis=1)
print(classification_report(np.argmax(y_test_cat, axis=1), y_pred_labels))
    313/313 -
                                 - 1s 3ms/step - accuracy: 0.6258 - loss: 1.0420
     ₹emi
     313/313 -
                                 - 1s 3ms/step
                   precision
                                recall f1-score
                                                    support
                0
                        0.74
                                  0.60
                                             0.66
                                                       1000
                1
                        0.78
                                  0.70
                                            0.74
                                                       1000
                2
                        0.65
                                  0.39
                                             0.49
                                                       1000
                                  0.29
                        0.57
                                             0.38
                                                       1000
                3
                                  0.50
                                             0.55
                4
                        0.61
                                                       1000
                        0.58
                                  0.56
                                             0.57
                                                       1000
                6
                        0.54
                                  0.86
                                             0.66
                                                       1000
                        0.59
                                  0.76
                                             0.67
                                                       1000
                        0.75
                                  0.76
                                            0.75
                                                       1000
                8
                                  0.86
                                                       1000
                9
                        0.58
                                            0.69
         accuracy
                                             0.63
                                                      10000
                        0.64
                                  0.63
                                             0.62
                                                      10000
        macro avg
     weighted avg
                        0.64
                                  0.63
                                             0.62
                                                      10000
for lr in [0.001, 0.01, 0.1]:
    model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=lr),
                  loss='categorical_crossentropy',
                  metrics=['accuracy'])
    history = model.fit(datagen.flow(x_train, y_train_cat, batch_size=128),
                        epochs=5,
                        validation data=(x test, y test cat))
⇒ Epoch 1/5
     /usr/local/lib/python3.11/dist-packages/keras/src/trainers/data adapters/py dataset adapter.py:121: UserWarning: Your `PyDataset` cl
       self._warn_if_super_not_called()
     391/391 -
                                 - 47s 93ms/step - accuracy: 0.5277 - loss: 1.3270 - val accuracy: 0.5777 - val loss: 1.3052
     Epoch 2/5
     391/391 -
                                 - 30s 75ms/step - accuracy: 0.5426 - loss: 1.2840 - val accuracy: 0.6098 - val loss: 1.1100
     Epoch 3/5
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.5579 - loss: 1.2543 - val_accuracy: 0.6140 - val_loss: 1.0764
     Epoch 4/5
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.5643 - loss: 1.2301 - val_accuracy: 0.6486 - val_loss: 1.0277
     Epoch 5/5
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.5766 - loss: 1.2000 - val_accuracy: 0.6034 - val_loss: 1.2162
     Epoch 1/5
     391/391 -
                                 - 44s 90ms/step - accuracy: 0.2949 - loss: 1.9502 - val accuracy: 0.3269 - val loss: 2.1056
     Epoch 2/5
     391/391 -
                                 - 30s 76ms/step - accuracy: 0.3940 - loss: 1.6976 - val_accuracy: 0.3215 - val_loss: 2.3465
     Epoch 3/5
     391/391 -
                                 - 30s 76ms/step - accuracy: 0.4273 - loss: 1.6100 - val accuracy: 0.4566 - val loss: 1.8348
     Epoch 4/5
     391/391 -
                                 - 30s 76ms/step - accuracy: 0.4504 - loss: 1.5744 - val_accuracy: 0.4240 - val_loss: 1.6816
     Epoch 5/5
     391/391 -
                                 - 30s 76ms/step - accuracy: 0.4587 - loss: 1.5341 - val_accuracy: 0.5682 - val_loss: 1.3093
     Epoch 1/5
     391/391 -
                                  • 46s 92ms/step - accuracy: 0.1081 - loss: 3.0286 - val_accuracy: 0.1000 - val_loss: 2.3053
     Epoch 2/5
     391/391 -
                                 - 28s 73ms/step - accuracy: 0.1026 - loss: 2.3082 - val accuracy: 0.1000 - val loss: 2.3111
     Epoch 3/5
     391/391 -
                                 - 29s 73ms/step - accuracy: 0.0961 - loss: 2.3084 - val accuracy: 0.0999 - val loss: 2.3070
     Epoch 4/5
     391/391 -
                                 - 29s 74ms/step - accuracy: 0.1007 - loss: 2.3085 - val_accuracy: 0.1000 - val_loss: 2.3074
     Epoch 5/5
     391/391 -
                                 - 41s 73ms/step - accuracy: 0.0993 - loss: 2.3085 - val_accuracy: 0.1000 - val_loss: 2.3089
```

```
plt.figure(figsize=(12, 6))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Train accuracy')
plt.plot(history.history['val_accuracy'], label='Val accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Train loss')
plt.plot(history.history['val_loss'], label='Val loss')
plt.title('Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.tight_layout()
plt.show()
```

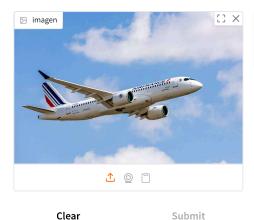


```
model.save('/content/drive/MyDrive/Deep/modelo.h5')
    WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is a
model = tf.keras.models.load_model('/content/drive/MyDrive/Deep/modelo.h5')
    WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until y
def cargar_imagen(imagen):
    img = image.load_img(imagen, target_size=(32, 32))
    img_array = image.img_to_array(img)
    img_array = np.expand_dims(img_array, axis=0)
    img_array = img_array.astype('float32') / 255.0
   return img_array
def predecir_imagen(imagen):
    img_preprocesada = cargar_imagen(imagen)
    predicciones = model.predict(img_preprocesada)
    clase_predicha = np.argmax(predicciones, axis=1)
    clases = ['avión', 'automóvil', 'pájaro', 'gato', 'ciervo', 'perro', 'rana', 'caballo', 'barco', 'camión']
    return f"Predicción: {clases[clase_predicha[0]]} - Probabilidad: {np.max(predicciones)}"
interface = gr.Interface(fn=predecir_imagen, inputs=gr.Image(type="filepath"), outputs="text")
interface.launch()
```

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Colab notebook detected. To show errors in colab notebook, set debug=True in launch() * Running on public URL: https://f663f64c6b7122b29c.gradio.live

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output Predicción: avión - Probabilidad: N 11239449679851532

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