```
import datetime
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from tensorflow.keras import Model
from tensorflow.keras.models import Sequential
from\ tensorflow.keras.losses\ import\ categorical\_crossentropy
from tensorflow.keras.layers import Dense, Flatten, Conv2D, AveragePooling2D
from tensorflow.keras import datasets
from tensorflow.keras.utils import to_categorical
from __future__ import absolute_import, division, print_function, unicode_literals
(x_train, y_train), (x_test, y_test) = datasets.mnist.load_data()
print('x_train shape:', x_train.shape)
print(x_train.shape[0], 'train samples')
print(x_test.shape[0], 'test samples')
print(x_train[0].shape, 'image shape')
     x_train shape: (60000, 28, 28)
     60000 train samples
     10000 test samples
     (28, 28) image shape
# Agregar una dimension
x_train = x_train[:, :, :, np.newaxis]
x_test = x_test[:, :, :, np.newaxis]
print('x_train shape:', x_train.shape)
print(x_train.shape[0], 'train samples')
print(x_test.shape[0], 'test samples')
print(x_train[0].shape, 'image shape')
     x_train shape: (60000, 28, 28, 1)
     60000 train samples
     10000 test samples
     (28, 28, 1) image shape
# convertir clases a matrices
num_classes = 10
y_train = to_categorical(y_train, num_classes)
y_test = to_categorical(y_test, num_classes)
# Normalización
x_{train} = x_{train.astype('float32')}
x_test = x_test.astype('float32')
x_train /= 255
x_test /= 255
                                                                                   avg pool
                                         avg pool
                                                               5 × 5
                  5 \times 5
                                         f = 2
                                         s = 2
                                                                                                                                   Softmax
                  s = 1
                                                               s = 1
                           28 \times 28 \times 6
                                                  14 \times 14 \times 6
                                                                     10 \times 10 \times 16
                                                                                         5 x 5 x 16
     32 \times 32 \times 1
                                                                                                                                      10
                                                                                           400
                                                                                                            120
                                                                                                                        84
# LeNet-5 model
class LeNet(Sequential):
  def __init__(self, input_shape, nb_classes):
    super().__init__()
    self.add(Conv2D(6, kernel_size=(5, 5), strides=(1, 1), activation='tanh', input_shape=input_shape, padding="same"))
    self.add(AveragePooling2D(pool_size=(2, 2), strides=(2, 2), padding='valid'))
    \verb|self.add(Conv2D(16, kernel_size=(5, 5), strides=(1, 1), activation='tanh', padding='valid')|| \\
    self.add(AveragePooling2D(pool_size=(2, 2), strides=(2, 2), padding='valid'))
    self.add(Flatten())
    self.add(Dense(120, activation='tanh'))
```

self.add(Dense(84, activation='tanh'))

self.add(Dense(nb\_classes, activation='softmax'))

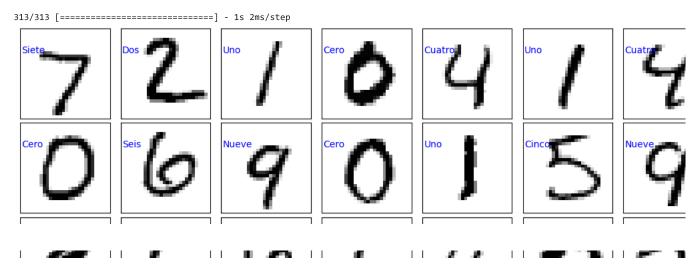
```
self.compile(optimizer='adam',
               loss=categorical_crossentropy,
                metrics=['accuracy'])
model = LeNet(x_train[0].shape, num_classes)
model.summary()
```

Model: "le\_net\_3"

Layer (type)	Output Shape	Param #
conv2d_5 (Conv2D)	(None, 28, 28, 6)	156
average_pooling2d_4 (Averag ePooling2D)	(None, 14, 14, 6)	0
conv2d_6 (Conv2D)	(None, 10, 10, 16)	2416
<pre>average_pooling2d_5 (Averag ePooling2D)</pre>	(None, 5, 5, 16)	0
flatten_2 (Flatten)	(None, 400)	0
dense_6 (Dense)	(None, 120)	48120
dense_7 (Dense)	(None, 84)	10164
dense_8 (Dense)	(None, 10)	850
Total params: 61,706		

Trainable params: 61,706 Non-trainable params: 0

```
model.fit(x_train, y=y_train,
          epochs=5,
          validation_data=(x_test, y_test),
          verbose=0)
     <keras.callbacks.History at 0x7f3f562f6260>
class_names = ['Cero', 'Uno', 'Dos', 'Tres', 'Cuatro',
                'Cinco', 'Seis', 'Siete', 'Ocho', 'Nueve']
y_pred = model.predict(x_test)
prediction_values=np.argmax(y_pred,axis=1)
fig = plt.figure(figsize=(15, 7))
fig.subplots_adjust(left=0, right=1, bottom=0, top=1, hspace=0.05, wspace=0.05)
for i in range(50):
 ax = fig.add_subplot(5, 10, i + 1, xticks=[], yticks=[])
 ax.imshow(x_test[i,:].reshape((28,28)),cmap=plt.cm.gray_r, interpolation='nearest')
 if prediction_values[i] == np.argmax(y_test[i]):
   ax.text(0, 7, class_names[prediction_values[i]], color='blue')
 else:
    ax.text(0, 7, class_names[prediction_values[i]], color='red')
```



Actividad: usar el modelo LeNet previo para resolver el problema de classificación de prendas mostrado a continuación.

```
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(x_train_, y_train_), (x_test_, y_test_) = datasets.fashion_mnist.load_data()
x_train_ = x_train_[:, :, :, np.newaxis]
x_{test} = x_{test}[:, :, :, np.newaxis]
y train = to categorical(y train , num classes)
y_test_ = to_categorical(y_test_, num_classes)
x_{train} = x_{train}.astype('float32')
x_{test} = x_{test}.astype('float32')
x_train_ /= 255
x_test_ /= 255
model_ = LeNet(x_train_[0].shape, num_classes)
model_.fit(x_train_, y = y_train_,
         epochs=5.
         validation_data=(x_test_, y_test_),
class_names = ['Camiseta/top', 'Pantalones', 'Jersey', 'Vestido', 'Abrigo',
               'Sandalia', 'Camisa', 'Sneaker', 'Bolso', 'Botines']
y_pred_ = model_.predict(x_test_)
prediction_values_ = np.argmax(y_pred_,axis=1)
fig = plt.figure(figsize=(15, 7))
fig.subplots_adjust(left=0, right=1, bottom=0, top=1, hspace=0.05, wspace=0.05)
for i in range(40):
 ax = fig.add_subplot(5, 8, i + 1, xticks=[], yticks=[])
 ax.imshow(x_test_[i,:].reshape((28,28)), cmap=plt.cm.gray_r, interpolation='nearest')
  if prediction_values_[i] == np.argmax(y_test_[i]):
   ax.text(0, 7, class_names[prediction_values_[i]], color='blue')
 else:
    ax.text(0, 7, class_names[prediction_values_[i]], color='red')
```

313/313 [=========] - 1s 2ms/step



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✓ 45 s completado a las 22:41