IF702 - REDES NEURAIS

Grupo:

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Reconhecimento de Dígitos com CNN

Importação de datasets e tratamento dos dados

```
!pip install keras-tuner --quiet
!pip install optuna --quiet
!pip install ipython-autotime --quiet
from keras_tuner import RandomSearch
import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
import keras
import optuna
(train images, train labels), (test images, test labels)=tf.keras.datasets.mnist.load data()
train_images=train_images.reshape((60000,28,28,1))
test_images=test_images.reshape((10000,28,28,1))
train images, test images=train images/255.0 ,test images/255.0
```

```
Construindo a CNN
  #Building a CNN Model
 #Building a CNM Model
model=tf.keras.models.Sequential()
model.add(tf.keras.layers.Conv2D(32, (3,3),activation='relu',input_shape=(28,28,1)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.Conv2D(64, (3,3),activation='relu'))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.FlaxPooling2D((2,2)))
model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Dense(64, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='softmax'))
 model.summary()
model.compile(optimizer='adam',loss='sparse_categorical_crossentropy',metrics='accuracy')
Model: "sequential"
                                                     Output Shape
Layer (type)
conv2d (Conv2D)
                                                    (None, 26, 26, 32)
max_pooling2d (MaxPooling2D) (None, 13, 13, 32)
conv2d_1 (Conv2D)
                                                    (None, 11, 11, 64)
                                                                                                       18496
max_pooling2d_1 (MaxPooling2 (None, 5, 5, 64)
conv2d 2 (Conv2D)
                                                    (None, 3, 3, 64)
                                                                                                       36928
flatten (Flatten)
                                                    (None, 576)
                                                                                                       0
dense (Dense)
                                                    (None, 64)
                                                                                                       36928
dense_1 (Dense)
                                                      (None, 10)
                                                                                                       650
Total params: 93,322
Trainable params: 93,322
Non-trainable params: 0
 model.fit(train_images,train_labels,epochs=1,batch_size=10)
scores = model.evaluate(test_images,test_labels, verbose=0)
6000/6000 [====
                                                                ======] - 16s 2ms/step - loss: 0.1172 - accuracy: 0.9637
def build_model(hp):
        build_model(hp):
    # create model object
model=tf.keras.models.Sequential()
model=add(tf.keras.layers.Conv2D(32, (3,3),activation='relu',input_shape=(28,28,1)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.NaxPooling2D((2,2)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.NaxPooling2D((2,2)))
model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Platten())
model.add(tf.keras.layers.Dense(64, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='softmax'))
         model.summary()
model.compile(optimizer='adam',loss='sparse_categorical_crossentropy',metrics='accuracy')
return model
 tuner.search(train_images,train_labels,epochs=3,validation_data=(train_images,train_labels))
INFO: tensor flow: Reloading \ Oracle \ from \ existing \ project \ ./untitled\_project/trial/untitled\_project/oracle.json \ Model: "sequential"
```

Layer (type)	Output	Shape	Param #
conv2d (Conv2D)	(None,	26, 26, 32)	320
max_pooling2d (MaxPooling2D)	(None,	13, 13, 32)	Θ
conv2d_1 (Conv2D)	(None,	11, 11, 64)	18496
max_pooling2d_1 (MaxPooling2	(None,	5, 5, 64)	Θ

```
conv2d 2 (Conv2D)
                                               (None, 3, 3, 64)
                                                                                          36928
flatten (Flatten)
                                               (None, 576)
                                                                                          0
dense (Dense)
                                                                                          36928
                                               (None, 64)
dense 1 (Dense)
                                                (None, 10)
                                                                                          650
Total params: 93,322
Trainable params: 93,322
Non-trainable params: 0
\overline{\text{INFO:}} tensorflow:Reloading \ Tuner \ from \ ./untitled\_project/trial/untitled\_project/tuner0.json \ INFO:\\ tensorflow:Oracle \ triggered \ exit
 best hps=tuner.get best hyperparameters(num trials=1)[0]
# Sugestão de Adriano (128 layers na 3a CNN)
def build_model2(hp):
       # create model object
model=tf.keras.models.Sequential()
      model=tf.keras.models.Sequential()
model.add(tf.keras.layers.Conv2D(32, (3,3),activation='relu',input_shape=(28,28,1)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.Conv2D(64, (3,3),activation='relu'))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.Conv2D(128,(3,3),activation='relu'))
model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Dense(128, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='relu'))
       model.summary()
model.compile(optimizer='adam',loss='sparse_categorical_crossentropy',metrics='accuracy')
return model
 tuner2 = RandomSearch(build_model2,
                                 objective='val_accuracy',
max_trials = 5,directory='./untitled_project/trial2/')
 tuner2.search(train_images,train_labels,epochs=3,validation_data=(train_images,train_labels))
INFO: tensor flow: Reloading \ Oracle \ from \ existing \ project \ ./untitled\_project/trial2/untitled\_project/oracle.json \ Model: "sequential"
Layer (type)
                                               Output Shape
                                                                                          Param #
conv2d (Conv2D)
                                                                                          320
                                               (None, 26, 26, 32)
max pooling2d (MaxPooling2D) (None, 13, 13, 32)
                                                                                          0
                                               (None, 11, 11, 64)
                                                                                          18496
conv2d 1 (Conv2D)
max_pooling2d_1 (MaxPooling2 (None, 5, 5, 64)
conv2d_2 (Conv2D)
                                               (None, 3, 3, 128)
                                                                                          73856
flatten (Flatten)
                                               (None, 1152)
                                               (None, 128)
                                                                                          147584
dense (Dense)
dense_1 (Dense)
                                               (None, 10)
                                                                                          1290
Total params: 241,546
Trainable params: 241,546
Non-trainable params: 0
INFO:tensorflow:Reloading Tuner from ./untitled_project/trial2/untitled_project/tuner0.json INFO:tensorflow:Oracle triggered exit
# Sugestão de Adriano (128 layers na 3a CNN com dilation_rate menor=2) def build_model3(hp):
      build_model3(hp):
# create model object
model=tf.keras.models.Sequential()
model.add(tf.keras.layers.Conv2D(32, (2,2),activation='relu',input_shape=(28,28,1)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.MaxPooling2D((2,2)))
model.add(tf.keras.layers.Sequential (2,2))
model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Platten())
model.add(tf.keras.layers.Dense(128, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='rolu'))
model.add(tf.keras.layers.Dense(10, activation='softmax'))
       model.summary()
model.compile(optimizer='adam',loss='sparse_categorical_crossentropy',metrics='accuracy')
return model
 tuner3 = RandomSearch(build model3,
                                 objective='val_accuracy',
max_trials = 5, directory='./untitled_project/trial3/')
 tuner 3. search (train\_images, train\_labels, epochs=3, validation\_data=(train\_images, train\_labels))
INFO: tensor flow: Reloading \ Oracle \ from \ existing \ project \ ./untitled\_project/trial3/untitled\_project/oracle.json \ Model: "sequential"
Layer (type)
                                               Output Shape
                                                                                          Param #
conv2d (Conv2D)
                                               (None, 27, 27, 32)
                                                                                          160
max_pooling2d (MaxPooling2D) (None, 13, 13, 32)
conv2d 1 (Conv2D)
                                                                                          8256
                                               (None, 12, 12, 64)
max_pooling2d_1 (MaxPooling2 (None, 6, 6, 64)
conv2d_2 (Conv2D)
                                               (None, 5, 5, 128)
                                                                                          32896
                                               (None, 3200)
flatten (Flatten)
                                                                                          0
dense (Dense)
                                               (None, 128)
                                                                                          409728
dense 1 (Dense)
                                               (None, 10)
                                                                                          1290
Total params: 452,330
Trainable params: 452,330
Non-trainable params: 0
INFO: tensorflow: Reloading \ Tuner \ from \ ./untitled\_project/trial3/untitled\_project/tuner\theta. json \ INFO: tensorflow: Oracle triggered exit
 from keras.layers import LSTM
from sklearn.metrics import accuracy_score, plot_confusion_matrix, confusion_matrix, classification_report
```

(train_images,train_labels),(test_images,test_labels)=tf.keras.datasets.mnist.load_data()

```
#Reshape data
train_images=train_images.reshape((60000,28,28,1))
test_images=test_images.reshape((10000,28,28,1))
train images, test images=train images/255.0 ,test images/255.0
```

Redes utilizadas

Camadas finais de adensamento com função de ativação 'relu' e de categorização com função de ativação 'softmax'.

CNN1:

- Neurônios: 32 à 128
- · Dilatação: 1 ou 2
- · Camadas CNN: 2

CNN2:

- Neurônios: 32 à 128
- Dilatação: 1 ou 2
- Camadas CNN: 3

CNN3:

1875/1875 [=

Neurônios: 32 à 128

```
    Dilatação: 1

  • Camadas CNN: 5
    (x_train, y_train), (x_test, y_test) = (train_images,train_labels),(test_images,test_labels)
    activation = trial.suggest_categorical('activation', ['sigmoid', 'softmax', 'tanh', 'relu','softplus','softsign','selu','elu','exponential'])
    n_neurons = trial.suggest_int('n_neurons', 16, 64)
dilation = trial.suggest_int('dilation', 1, 2)
loss = trial.suggest_categorical('loss', ['sparse_categorical_crossentropy'])
optimizer = trial.suggest_categorical('optimizer', ['sgd', 'adam','rmsprop','adamax'])
    \verb|model.add(tf.keras.layers.Conv2D(n neurons, (dilation, dilation), activation=activation, input\_shape=(28,28,1)))|
    model.add(tf.keras.layers.Conv2U(n_neurons, (dllation,dilation),activation=activation,inpl
model.add(tf.keras.layers.MaxPooling2D((dilation,dilation)))
model.add(tf.keras.layers.Conv2D(n_neurons*2, (dilation,dilation),activation=activation))
model.add(tf.keras.layers.Platten())
model.add(tf.keras.layers.Dense(n_neurons*2, activation='relu'))
model.add(tf.keras.layers.Dense(10, activation='softmax'))
model.compile(loss=loss,optimizer=optimizer,metrics=["accuracy"])
    \label{eq:history} \begin{array}{ll} \mbox{history} = \mbox{model.fit}(x\_train, \ y\_train, \ \mbox{epochs=8, verbose=1}) \\ \mbox{y\_pred} = \mbox{np.argmax}(\mbox{model.predict}(x\_test), \ \mbox{axis=-1}) \\ \mbox{y\_true} = \mbox{y\_test} \end{array}
     return accuracy_score(y_true, y_pred)
 study_1 = optuna.create_study(direction='maximize')
study_1.optimize(cnn, n_trials=16)
 \hbox{[I 2021-11-05 13:13:11,446] A new study created in memory with name: no-name-0212a617-dc7d-4797-b20d-8c84dd82265d ] } \\
Número de neurônios: 32
Otimizador: sgd
Epoch 1/8
 1875/1875 [=
                                               ========1 - 5s 3ms/step - loss: 0.3251 - accuracy: 0.9087
Epoch 2/8
                                 1875/1875 [=
Epoch 5/8
 1875/1875 [=
                                    1875/1875 [=
                                              [I 2021-11-05 13:13:49,505] Trial 0 finished with value: 0.9729 and parameters: {'activation': 'selu', 'n_neurons': 32, 'dilation': 1, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'sgd'}. Best is trial 0 with value: 0.9729. Função de ativação: softplus Número de neurônios: 24
Otimizador: rmsprop
| December 2007 | Control 
Epoch 3/8
 Epoch 6/8
 [I 2021-11-05 13:14:32,475] Trial 1 finished with value: 0.1135 and parameters: {'activation': 'softplus', 'n_neurons': 24, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 0 with value: 0.9729.
Função de ativação: selu
Número de neurônios: 20
Epoch 2/8
 Fnoch 3/8
 Epoch 4/8
1875/1875 [=
                                Epoch 5/8
```

```
Epoch 7/8
1875/1875 [=
                     Fnoch 8/8
[I 2021-11-05 13:15:04,483] Trial 2 finished with value: 0.9794 and parameters: {'activation': 'selu', 'n_neurons': 20, 'dilation': 2, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'adamax'}. Best is trial 2 with value: 0.9794. Função de ativação: softplus Número de neurônios: 61
Otimizador: adamax
Fnoch 1/8
1875/1875 [===
                         Torsy | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1075 | 1
Epoch 4/8
1875/1875 [=
                        -----] - 9s 5ms/step - loss: 2.3012 - accuracy: 0.1124
Epoch 6/8
Fnoch 7/8
1875/1875 [=
Epoch 8/8
1875/1875 [=
                          10/2/10/5 [=========] - 95 5ms/step - loss: 2.3012 - accuracy: 0.1124
[I 2021-11-05 13:16:19,574] Trial 3 finished with value: 0.1135 and parameters: {'activation': 'softplus', 'n_neurons': 61, 'dilation': 1, 'loss': 'sparse_categorica l_crossentropy', 'optimizer': 'adamax'}. Best is trial 2 with value: 0.9794.
Função de ativação: softsign
Número de neurônios: 39
Otimizador: sgd
Epoch 1/8
1875/1875 [===========] - 4s 2ms/step - loss: 0.5266 - accuracy: 0.8632
Fnoch 2/8
                                                ====] - 9s 5ms/step - loss: 2.3012 - accuracy: 0.1124
========] - 4s 2ms/step - loss: 0.1489 - accuracy: 0.9560
Epoch 5/8
Epoch 6/8
1875/1875 [============] - 4s 2ms/step - loss: 0.1136 - accuracy: 0.9665
Epoch 7/8
1875/1875 [============] - 4s 2ms/step - loss: 0.1025 - accuracy: 0.9700
Epoch 8/8
[I 2021-11-05 13:16:51,268] Trial 4 finished with value: 0.9689 and parameters: {'activation': 'softsign', 'n_neurons': 39, 'dilation': 2, 'loss': 'sparse_categorica l_crossentropy', 'optimizer': 'softsign'
| Número de neurônios: 39 | Número de neurôni
                  1875/1875 [=
Otimizador: adam
Epoch 1/8
1875/1875 [====
                          Epoch 2/8
1875/1875 [====
                        Epoch 3/8
Fnoch 4/8
1875/1875 [=====
                        Fnoch 7/8
Epoch 2/8
                           1875/1875 [=
Epoch 3/8
Epoch 5/8
Epoch 6/8
EPOCH 0/6
1875/1875 [============] - 4s 2ms/step - loss: 0.0258 - accuracy: 0.9925
Epoch 7/8
1875/1875 [=====================] - 4s 2ms/step - loss: 0.0224 - accuracy: 0.9938
Epoch 8/8
[I 2021-11-05 13:18:00,618] Trial 6 finished with value: 0.987 and parameters: ('activation': 'relu', 'n_neurons': 21, 'dilation': 2, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 6 with value: 0.987.
ssentropy', 'optimizer': 'rms
Função de ativação: softsign
Número de neurônios: 42
Otimizador: rmsprop
Epoch 2/8
1875/1875 [========] - 8s 4ms/step - loss: 0.1269 - accuracy: 0.9637
Epoch 5/8
1875/1875 [======
Epoch 6/8
                        Epoch 7/8
[I 2021-11-05 13:19:04,305] Trial 7 finished with value: 0.9704 and parameters: {'activation': 'softsign', 'n_neurons': 42, 'dilation': 1, 'loss': 'sparse_categorica l_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 6 with value: 0.987. Função de ativação: selu Número de neurônios: 17
Otimizador: adamax
Epoch 3/8
1875/1875 [----
Epoch 4/8
1875/1875 [------] - 4s 2ms/step - loss: 0.1279 - accuracy: 0.9622
Epoch 5/8
1875/1875 [=
                         -----] - 4s 2ms/step - loss: 0.1066 - accuracy: 0.9677
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Epoch 7/8
1875/1875 [=
                Fnoch 8/8
[I 2021-11-05 13:19:39,487] Trial 8 finished with value: 0.9591 and parameters: {'activation': 'selu', 'n_neurons': 17, 'dilation': 1, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'adamax'}. Best is trial 6 with value: 0.987. Função de ativação: exponential Número de neurônios: 34
Otimizador: rmsprop
Fnoch 1/8
1875/1875 [====
                  Epoch 4/8
1875/1875 [=
                 Epoch 6/8
Fnoch 7/8
1875/1875 [=
Epoch 8/8
1875/1875 [=
                  ==] - 4s 2ms/step - loss: 0.0601 - accuracy: 0.9825
Epoch 3/8
1875/1875 [============] - 4s 2ms/step - loss: 0.0242 - accuracy: 0.9921
Epoch 4/8
1875/1875 [============] - 4s 2ms/step - loss: 0.0168 - accuracy: 0.9944
                   Epoch 5/8
                1875/1875 [=====
Epoch 8/8
1875/1875 [=
            [I 2021-11-05 13:20:51,066] Trial 10 finished with value: 0.9882 and parameters: {'activation': 'relu', 'n_neurons': 51, 'dilation': 2, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'adam'}. Best is trial 10 with value: 0.9882.
Função de ativação: relu
Número de neurônios: 53
Otimizador: adam
Epoch 1/8
1875/1875 [===
                  Epoch 2/8
1875/1875 [==
                 Epoch 3/8
Fnoch 4/8
1875/1875 [====
                Fnoch 7/8
[I 2021-11-05 13:21:26,741] Trial 11 finished with value: 0.9886 and parameters: {'activation': 'relu', 'n_neurons': 53, 'dilation': 2, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'adam'}. Best is trial 11 with value: 0.9886. Função de ativação: relu Número de neurônios: 56
Otimizador: adam
Epoch 1/8
1875/1875 [======
                 Epoch 2/8
                   1875/1875 [=
Epoch 3/8
Epoch 5/8
1875/1875 [=====
                Epoch 6/8
EPOCH 0/6
1875/1875 [============] - 4s 2ms/step - loss: 0.0088 - accuracy: 0.9972
Epoch 7/8
1875/1875 [=====================] - 4s 2ms/step - loss: 0.0081 - accuracy: 0.9972
Epoch 8/8
[I 2021-11-05 13:22:03,102] Trial 12 finished with value: 0.9893 and parameters: {'activation': 'relu', 'n_neurons': 56, 'dilation': 2, 'loss': 'sparse_categorical_c rossentropy', 'optmizer': 'adam'}. Best is trial 12 with value: 0.9893.
Função de ativação: relu
Número de neurônios: 62
Otimizador: adam
Epoch 2/8
1875/1875 [========] - 4s 2ms/step - loss: 0.0387 - accuracy: 0.9877
Epoch 4/8
Epoch 5/8
1875/1875 [====
                 Epoch 7/8
[I 2021-11-05 13:22:39,955] Trial 13 finished with value: 0.9862 and parameters: {'activation': 'relu', 'n_neurons': 62, 'dilation': 2, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'adam'}. Best is trial 12 with value: 0.9893. Função de ativação: elu Número de neurônios: 52
Otimizador: adam
Temperature to the second seco
Epoch 2/8
1875/1875 [====
                 Epoch 3/8
1875/1875 [----
Epoch 4/8
1875/1875 [------] - 4s 2ms/step - loss: 0.0361 - accuracy: 0.9883
Epoch 5/8
1875/1875 [
                  -----] - 4s 2ms/step - loss: 0.0227 - accuracy: 0.9926
```

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1875/1875 [
                  Fnoch 8/8
[I 2021-11-05 13:23:14,891] Trial 14 finished with value: 0.9822 and parameters: {'activation': 'elu', 'n_neurons': 52, 'dilation': 2, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'adam'}. Best is trial 12 with value: 0.9893. Função de ativação: tanh Número de neurônios: 53
Otimizador: adam
Fnoch 1/8
1875/1875 [=
                  =========] - 5s 2ms/step - loss: 0.1474 - accuracy: 0.9549
Epoch 4/8
1875/1875 [=
                  ========] - 4s 2ms/step - loss: 0.0287 - accuracy: 0.9908
Epoch 6/8
Fnoch 7/8
1875/1875 [==
Epoch 8/8
1875/1875 [==
                       =======] - 4s 2ms/step - loss: 0.0205 - accuracy: 0.9934
                                  ==] - 4s 2ms/step - loss: 0.0166 - accuracy: 0.9949
[I 2021-11-05 13:23:50,908] Trial 15 finished with value: 0.9828 and parameters: {'activation': 'tanh', 'n_neurons': 53, 'dilation': 2, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'adam'}. Best is trial 12 with value: 0.9893.
def cnn2(trial):
   (x_train, y_train), (x_test, y_test) = (train_images,train_labels),(test_images,test_labels)
  activation = trial.suggest_categorical('activation', ['sigmoid', 'softmax', 'tanh', 'relu','softplus','softsign','selu','elu','exponential'])
  n_neurons = trial.suggest_int('n_neurons', 16, 64)
dilation = trial.suggest_int('dilation', 1, 2)
loss = trial.suggest_categorical('loss', ['sparse_categorical_crossentropy'])
optimizer = trial.suggest_categorical('optimizer', ['sgd', 'adam','rmsprop','adamax'])
  print("Função de ativação: {}\nNúmero de neurônios: {}\nOtimizador: {}\".format(activation,n neurons.optimizer))
  model = keras.models.Sequential()
  \label{local_model_add} $$ model_add(tf.keras.layers.Conv2D(n_neurons, (dilation,dilation),activation=activation,input\_shape=(28,28,1))) $$ model_add(tf.keras.layers.MaxPooling2D((dilation,dilation))) $$ $$
                   ADICIONAMOS MAIS UMA CAMADA CNN -
  \label{eq:model_add(tf.keras.layers.Conv2D(n_neurons*2, (dilation,dilation),activation=activation))} \\ model.add(tf.keras.layers.Flatten()) \\ model.add(tf.keras.layers.Dense(n_neurons*2, activation='relu')) \\ model.add(tf.keras.layers.Dense(10, activation='softmax')) \\ model.add(tf.keras.layers.Dense(10, activation='softmax')) \\ model.compile(loss=loss,optimizer=optimizer,metrics=["accuracy"]) \\ \\
  history = model.fit(x_train, y_train, epochs=8, verbose=1)
y_pred = np.argmax(model.predict(x_test), axis=-1)
y_true = y_test
return accuracy_score(y_true, y_pred)
study_2 = optuna.create_study(direction='maximize')
study_2.optimize(cnn2, n_trials=16)
[I 2021-11-05 13:23:50,941] A new study created in memory with name: no-name-fd4b0ef8-84d3-496c-9361-d9lac0a9eblf Função de ativação: softmax Número de neurônios: 20
Otimizador: rmsprop
Epoch 3/8
1875/1875 [=
               Epoch 4/8
1875/1875 [=
                 -----] - 7s 4ms/step - loss: 2.3013 - accuracy: 0.1124
Fnoch 6/8
1875/1875 [=
                   ========] - 7s 4ms/step - loss: 2.3013 - accuracy: 0.1124
Epoch 7/8
1875/1875 [====
                Epoch 8/8
In 2021-11-05 13:24:45,081] Trial 0 finished with value: 0.1135 and parameters: {'activation': 'softmax', 'n_neurons': 20, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 0 with value: 0.1135.
Função de ativação: selu
Número de neurônios: 36
Otimizador: adamax
Espech 1/8
Epoch 1/8
1875/1875 [==
                        Epoch 3/8
1875/1875 [=
                 Epoch 4/8
                  -----] - 8s 4ms/step - loss: 0.0822 - accuracy: 0.9750
1875/1875 [=
Epoch 5/8
Epoch 6/8
1875/1875 [==
                  Epoch 7/8
[I 2021-11-05 13:25:46,116] Trial 1 finished with value: 0.9689 and parameters: {'activation': 'selu', 'n_neurons': 36, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'adamax'}. Best is trial 1 with value: 0.9689. Função de ativação: softplus Número de neurônios: 45
Otimizador: rmsprop
Epoch 1/8
1875/1875 [===
                   Epoch 3/8
1875/1875 [========] - 11s 6ms/step - loss: 0.3430 - accuracy: 0.8993
Epoch 4/8
1875/1875 [======
Epoch 5/8
1875/1875 [======
```

Epoch 7/8

```
Epoch 8/8
[I 2021-11-05 13:27:13,336] Trial 2 finished with value: 0.9213 and parameters: {'activation': 'softplus', 'n_neurons': 45, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 1 with value: 0.9689.
Função de ativação: exponential
Número de neurônios: 16
Otimizador: adamax
Fnoch 2/8
Epoch 3/8
1875/1875 [=======] - 4s 2ms/step - loss: 0.2351 - accuracy: 0.9291
Epoch 5/8
1875/1875 [==
                 Epoch 7/8
Fnoch 8/8
1875/1875 [=
                   [I 2021-11-05 13:27:49,889] Trial 3 finished with value: 0.9751 and parameters: {'activation': 'exponential', 'n_neurons': 16, 'dilation': 2, 'loss': 'sparse_categor ical_crossentropy', 'optimizer': 'adamax'}. Best is trial 3 with value: 0.9751. Função de ativação: selu Número de neurônios: 41
Otimizador: adamax
Epoch 5/8

Epoch 5/8

Epoch 5/8

Epoch 5/8

Epoch 5/8

Epoch 5/8
Epoch 5/8
1875/1875 [==
                  Epoch 6/8
1875/1875 [=
                          ======== 1 - 5s 2ms/step - loss: 0.0111 - accuracy: 0.9966
Epoch 2/8
Epoch 3/8
1875/1875 [==
                 Epoch 4/8
Fnoch 5/8
Fnoch 8/8
1875/1875 [=
            [I 2021-11-05 13:29:16,820] Trial 5 finished with value: 0.9695 and parameters: {'activation': 'tanh', 'n_neurons': 20, 'dilation': 1, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'rmsprop'}. Best is trial 4 with value: 0.9893.
Função de ativação: softplus
Número de neurônios: 53
Otimizador: adamax
Epoch 3/8
                   1875/1875 [=
Epoch 4/8
Epoch 6/8
Epoch 7/8
EPOCH 7/6
B1875/1875 [==========] - 5s 3ms/step - loss: 0.0356 - accuracy: 0.9888
Epoch 8/8
1875/1875 [============] - 5s 3ms/step - loss: 0.0302 - accuracy: 0.9899
[I 2021-11-05 13:29:58,553] Trial 6 finished with value: 0.9879 and parameters: {'activation': 'softplus', 'n_neurons': 53, 'dilation': 2, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'adamax'}. Best is trial 4 with value: 0.9893. Função de ativação: sigmoid Número de neurônios: 63
Epoch 3/8
1875/1875 [====
                Tors | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 | 103 |
Epoch 5/8
Epoch 6/8
1875/1875 [====
Epoch 7/8
                 [I 2021-11-05 13:31:18,595] Trial 7 finished with value: 0.135 and parameters: {'activation': 'sigmoid', 'n_neurons': 63, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'sgd'}. Best is trial 4 with value: 0.9893.

Número de neurônios: 63
Otimizador: adam
Epoch 1/8
Epoch 3/8
1875/1875 [====
                 =======] - 11s 6ms/step - loss: 0.2185 - accuracy: 0.9388
Epoch 4/8
1875/1875 [---
Epoch 5/8
1875/1875 [------] - 11s 6ms/step - loss: 0.1528 - accuracy: 0.9570
Epoch 6/8
1875/1875 [=
                -----] - 11s 6ms/step - loss: 0.1188 - accuracy: 0.9653
```

```
Epoch 8/8
[I 2021-11-05 13:32:47,390] Trial 8 finished with value: 0.961 and parameters: {'activation': 'exponential', 'n_neurons': 63, 'dilation': 1, 'loss': 'sparse_categori cal_crossentropy', 'optimizer': 'adam'}. Best is trial 4 with value: 0.9893. Função de ativação: elu Número de neurônios: 19
Otimizador: adamax
Fnoch 2/8
Epoch 3/8
1875/1875 [=======] - 4s 2ms/step - loss: 0.0705 - accuracy: 0.9783
Epoch 5/8
1875/1875 [==
                 ========] - 4s 2ms/step - loss: 0.0403 - accuracy: 0.9878
Epoch 7/8
Fnoch 8/8
1875/1875 [=
                   [I 2021-11-05 13:33:23,811] Trial 9 finished with value: 0.9878 and parameters: {'activation': 'elu', 'n_neurons': 19, 'dilation': 2, 'loss': 'sparse_categorical_crosentropy', 'optimizer': 'adamax'}. Best is trial 4 with value: 0.9893. Função de ativação: selu Número de neurônios: 35
Otimizador: sqd
| Default 2006 | Squip | Squip
Epoch 5/8
1875/1875 [==
                  Epoch 6/8
1875/1875 [=
                           [I 2021-11-05 13:33:57,578] Trial 10 finished with value: 0.9843 and parameters: {'activation': 'selu', 'n_neurons': 35, 'dilation': 2, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'sgd'}. Best is trial 4 with value: 0.9893.
Epoch 2/8
Epoch 3/8
1875/1875 [==
                 Epoch 4/8
Fnoch 5/8
Fnoch 8/8
1875/1875 [=
            =======] - 5s 3ms/step - loss: 2.3012 - accuracy: 0.1124
[I 2021-11-05 13:34:38,332] Trial I1 finished with value: 0.1135 and parameters: {'activation': 'softplus', 'n_neurons': 49, 'dilation': 2, 'loss': 'sparse_categoric al_crossentropy', 'optimizer': 'adamax'}. Best is trial 4 with value: 0.9893.
Função de ativação: softsign
Número de neurônios: 53
Otimizador: adamax
Epoch 3/8
                   1875/1875 [==
Epoch 4/8
Epoch 6/8
Epoch 7/8
EPOCH 7/8
1875/1875 [==========] - 5s 3ms/step - loss: 0.0154 - accuracy: 0.9952
Epoch 8/8
1875/1875 [===========] - 5s 3ms/step - loss: 0.0127 - accuracy: 0.9961
[I 2021-11-05 13:35:20,639] Trial 12 finished with value: 0.9917 and parameters: {'activation': 'softsign', 'n_neurons': 53, 'dilation': 2, 'loss': 'sparse_categoric al_crossentropy', 'optimizer': 'adamax'}. Best is trial 12 with value: 0.9917. Função de ativação: softsign
Número de neurônios: 55
Otimizador: adam
Fanch 1/8
Epoch 3/8
1875/1875 [=====
                Epoch 5/8
Epoch 6/8
1875/1875 [====
Epoch 7/8
                 [1 2021-11-05 13:36:01,848] Trial 13 finished with value: 0.9813 and parameters: {'activation': 'softsign', 'n_neurons': 55, 'dilation': 2, 'loss': 'sparse_categoric al_crossentropy', 'optimizer': 'adam'}. Best is trial 12 with value: 0.9917. Função de ativação: softsign Número de neurônios: 31
Otimizador: adamax
Epoch 1/8
Epoch 3/8
1875/1875 [====
                 =======] - 4s 2ms/step - loss: 0.0575 - accuracy: 0.9829
Epoch 4/8
1875/1875 [----
Epoch 5/8
1875/1875 [------] - 4s 2ms/step - loss: 0.0370 - accuracy: 0.9889
Epoch 6/8
1875/1875 [=
                -----] - 4s 2ms/step - loss: 0.0274 - accuracy: 0.9917
```

```
[I 2021-11-05 13:36:43,751] Trial 14 finished with value: 0.99 and parameters: {'activation': 'softsign', 'n_neurons': 31, 'dilation': 2, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'adamax'}. Best is trial 12 with value: 0.9917. Função de ativação: softsign Número de neurônios: 29
Otimizador: adamax
Fnoch 2/8
Epoch 5/8
1875/1875 [=
             -----] - 5s 3ms/step - loss: 0.0388 - accuracy: 0.9882
             Epoch 7/8
1875/1875 [=
                         ===] - 5s 2ms/step - loss: 0.0246 - accuracy: 0.9924
[I 2021-11-05 13:37:22,545] Trial 15 finished with value: 0.9886 and parameters: {'activation': 'softsign', 'n_neurons': 29, 'dilation': 2, 'loss': 'sparse_categoric al_crossentropy', 'optimizer': 'adamax'}. Best is trial 12 with value: 0.9917.
  (x\_train, \ y\_train), \ (x\_test, \ y\_test) = (train\_images, train\_labels), (test\_images, test\_labels)
 activation = trial.suggest_categorical('activation', ['sigmoid', 'softmax', 'tanh', 'relu','softplus','softsign','selu','elu','exponential'])
  n neurons = trial.suggest int('n neurons', 16, 64)
 dilation = trial.suggest_int('dilation', 1, 1)
loss = trial.suggest_categorical('loss', ['sparse_categorical_crossentropy'])
optimizer = trial.suggest_categorical('optimizer', ['sgd', 'adam','rmsprop','adamax'])
 \label{local_model_add(tf.keras.layers.Conv2D(n_neurons, (dilation, dilation), activation=activation, input\_shape=(28,28,1))) \\ model.add(tf.keras.layers.MaxPooling2D((dilation, dilation)))
              ADICIONAMOS MAIS TRÊS CAMADA CNN
 \label{eq:model_add(tf.keras.layers.Conv2D(n_neurons*2, (dilation,dilation),activation=activation))} \\ model.add(tf.keras.layers.Flatten()) \\ model.add(tf.keras.layers.Dense(n_neurons*2, activation='relu')) \\ model.add(tf.keras.layers.Dense(10, activation='softmax')) \\ model.compile(loss=loss,optimizer=optimizer,metrics=["accuracy"]) \\ \end{aligned}
  history = model.fit(x_train, y_train, epochs=8, verbose=1)
y_pred = np.argmax(model.predict(x_test), axis=-1)
y_true = y_test
  return accuracy_score(y_true, y_pred)
study\_3 = optuna.create\_study(direction='maximize')\\ study\_3.optimize(cnn3, n\_trials=16)
[I 2021-11-05 13:37:22,583] A new study created in memory with name: no-name-bb39e995-7f55-4052-b3f5-188b7096ee06 Função de ativação: sigmoid Número de neurônios: 43 Otimizador: adamax
Epoch 5/8
1875/1875 [==
Epoch 6/8
            -----] - 13s 7ms/step - loss: 2.3012 - accuracy: 0.1124
[I 2021-11-05 13:39:08,714] Trial 0 finished with value: 0.1135 and parameters: {'activation': 'sigmoid', 'n_neurons': 43, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'adamax'}. Best is trial 0 with value: 0.1135. Função de ativação: softsign Número de neurônios: 32
Otimizador: rmsprop
Epoch 2/8
1875/1875 [=
            =======] - 10s 6ms/step - loss: 0.1273 - accuracy: 0.9620
Epoch 3/8
Epoch 5/8
1875/1875 [=
            Epoch 6/8
B05/1875 [============] - 10s 5ms/step - loss: 0.0646 - accuracy: 0.9809
            1875/1875 [=
Epoch 2/8
1875/1875 [=
          Fnoch 5/8
1875/1875 [=
```

Epoch 8/8

```
Epoch 7/8
1875/1875 [=
       Fnoch 8/8
[I 2021-11-05 13:42:10,612] Trial 2 finished with value: 0.9713 and parameters: {'activation': 'tanh', 'n_neurons': 37, 'dilation': 1, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'rmsprop'}. Best is trial 1 with value: 0.9725. Função de ativação: softmax Número de neurônios: 42
Otimizador: rmsprop
Fnoch 1/8
1875/1875 [===:
          1875/1875 [=
         Epoch 6/8
Fnoch 7/8
1875/1875 [=
Epoch 8/8
1875/1875 [=
          ==] - 17s 9ms/step - loss: 2.3014 - accuracy: 0.1124
[I 2021-11-05 13:44:34,662] Trial 3 finished with value: 0.1135 and parameters: {'activation': 'softmax', 'n_neurons': 42, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 1 with value: 0.9725.
======== ] - 12s 6ms/step - loss: 0.2901 - accuracy: 0.9162
Epoch 4/8
1875/1875 [=
          Epoch 5/8
         1875/1875 [=====
[I 2021-11-05 13:46:57,749] Trial 4 finished with value: 0.9704 and parameters: {'activation': 'elu', 'n_neurons': 41, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'sgd'}. Best is trial 1 with value: 0.9725.
Função de ativação: elu
Número de neurônios: 60
Otimizador: responso
Otimizador: rmsprop
Epoch 1/8
1875/1875 [====
          Enoch 2/8
1875/1875 [=
         Epoch 3/8
Fnoch 4/8
1875/1875 [====
         Epoch 5/8 1875/1875 [=============] - 17s 9ms/step - loss: 0.0876 - accuracy: 0.9774
Fnoch 7/8
[I 2021-11-05 13:49:17,789] Trial 5 finished with value: 0.9647 and parameters: {'activation': 'elu', 'n_neurons': 60, 'dilation': 1, 'loss': 'sparse_categorical_cro ssentropy', 'optimizer': 'rmsprop'}. Best is trial 1 with value: 0.9725.
Função de ativação: relu
Epoch 2/8
          1875/1875 [=
Epoch 3/8
Epoch 5/8
Epoch 6/8
EPOCH 0/8
1875/1875 [============] - 10s 5ms/step - loss: 0.0554 - accuracy: 0.9827
Epoch 7/8
1875/1875 [===========] - 10s 5ms/step - loss: 0.0487 - accuracy: 0.9848
Epoch 8/8
[I 2021-11-05 13:55:39,762] Trial 6 finished with value: 0.9744 and parameters: {'activation': 'relu', 'n_neurons': 38, 'dilation': 1, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'rsgd'}. Best is trial 6 with value: 0.9744.

Número de neurônios: 64
Otimizador: rmsprop
Epoch 4/8
Epoch 5/8
1875/1875 [===
         Epoch
[I 2021-11-05 13:53:32,748] Trial 7 finished with value: 0.1135 and parameters: {'activation': 'softmax', 'n_neurons': 64, 'dilation': 1, 'loss': 'sparse_categorical_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 6 with value: 0.9744.
Função de ativação: exponential
Número de neurônios: 19
Otimizador: rmsprop
Epoch 2/8
1875/1875 [=======] - 8s 4ms/step - loss: nan - accuracy: 0.0987
Epoch 3/8
Epoch 5/8
1875/1875 [=
         1875/1875 [=
```

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Epoch 7/8
1875/1875 [=
          Fnoch 8/8
[I 2021-11-05 13:54:38,064] Trial 8 finished with value: 0.098 and parameters: {'activation': 'exponential', 'n_neurons': 19, 'dilation': 1, 'loss': 'sparse_categori cal_crossentropy', 'optimizer': 'rmsprop'}. Best is trial 6 with value: 0.9744. Função de ativação: selu Número de neurônios: 41
Otimizador: sqd
Fnoch 1/8
Epoch 4/8
1875/1875 [=
          Epoch 6/8
Fnoch 7/8
1875/1875 [=
Epoch 8/8
1875/1875 [=
           ======== ] - 12s 6ms/step - loss: 0.0320 - accuracy: 0.9900
                     ==] - 12s 6ms/step - loss: 0.0254 - accuracy: 0.9921
[I 2021-11-05 13:56:14,523] Trial 9 finished with value: 0.9758 and parameters: {'activation': 'selu', 'n_neurons': 41, 'dilation': 1, 'loss': 'sparse_categorical_cr ossentropy', 'optimizer': 'sgd'}. Best is trial 9 with value: 0.9758.
Função de ativação: selu
Número de neurônios: 52
Otimizador: adam
Epoch 1/8
1875/1875 [========
        Epoch 5/8
          =======] - 13s 7ms/step - loss: 0.0822 - accuracy: 0.9748
1875/1875 [=====
Epoch 8/8
1875/1875 [=
       [I 2021-11-05 13:55:03,140] Trial 10 finished with value: 0.9691 and parameters: {'activation': 'selu', 'n_neurons': 52, 'dilation': 1, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'adam'}. Best is trial 9 with value: 0.9758. Função de ativação: relu Número de neurônios: 27
Otimizador: sqd
Epoch 1/8
1875/1875 [===
           Epoch 2/8
1875/1875 [=
          Epoch 3/8
Fnoch 4/8
1875/1875 [====
          Fnoch 7/8
-----] - 8s 4ms/step - loss: 0.0528 - accuracy: 0.9833
[I 2021-11-05 13:59:04,956] Trial 11 finished with value: 0.9747 and parameters: {'activation': 'relu', 'n_neurons': 27, 'dilation': 1, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'sgd'}. Best is trial 9 with value: 0.9758.
Função de ativação: relu
Epoch 2/8
           1875/1875 [=
Epoch 3/8
Epoch 5/8
Epoch 6/8
EPOCH 0/6

1875/1875 [============] - 7s 4ms/step - loss: 0.0809 - accuracy: 0.9750

Epoch 7/8

1875/1875 [==================] - 7s 4ms/step - loss: 0.0732 - accuracy: 0.9773
Epoch 8/8
[I 2021-11-05 14:00:03,935] Trial 12 finished with value: 0.90508 and parameters: {'activation': 'relu', 'n_neurons': 23, 'dilation': 1, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'sgd'}. Best is trial 9 with value: 0.9758. Função de ativação: softplus
Número de neurônios: 27
Otimizador: sgd
Epoch 4/8
Epoch 5/8
1875/1875 [====
Epoch 6/8
          Epoch 7/8
[I 2021-11-05 14:01:07,859] Trial 13 finished with value: 0.1135 and parameters: {'activation': 'softplus', 'n_neurons': 27, 'dilation': 1, 'loss': 'sparse_categoric al_crossentropy', 'optimizer': 'sgd'}. Best is trial 9 with value: 0.9758. Função de ativação: selu Número de neurônios: 16
Otimizador: sqd
Epoch 2/8
1875/1875 [====
          Epoch 3/8
1875/1875 [----
Epoch 4/8
1875/1875 [------] - 6s 3ms/step - loss: 0.1042 - accuracy: 0.9685
Epoch 5/8
1875/1875 [=
          -----] - 6s 3ms/step - loss: 0.0842 - accuracy: 0.9745
1875/1875 [=
```

```
Epoch 7/8
1875/1875 [=
                                                                    Fnoch 8/8
  [I 2021-11-05 14:01:55,763] Trial 14 finished with value: 0.9657 and parameters: {'activation': 'selu', 'n_neurons': 16, 'dilation': 1, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'sgd'}. Best is trial 9 with value: 0.9758. Função de ativação: selu Número de neurônios: 49
 Otimizador: adam
 | Stanfold 
 Epoch 3/8 [===============] - 13s 7ms/step - loss: 0.1156 - accuracy: 0.9656
 Epoch 4/8
1875/1875 [=
Epoch 5/8
1875/1875 [=
                                                             ========] - 13s 7ms/step - loss: 0.0980 - accuracy: 0.9706
                                                                    [I 2021-11-05 14:03:41,345] Trial 15 finished with value: 0.9717 and parameters: {'activation': 'selu', 'n_neurons': 49, 'dilation': 1, 'loss': 'sparse_categorical_c rossentropy', 'optimizer': 'adam'}. Best is trial 9 with value: 0.9758.
```

Gráficos de resultados do processo de otimização

optuna.visualization.plot optimization history(study 1)

In []: optuna.visualization.plot_optimization_history(study_2)

	Como podemos ver acima, nossa primeira rede de CNN obteve melhor desempenho em sua otimização, dado que foi mais consistente (o primeiro trial do CNN2 e CNN3 foi um out	lier).
In []:	<pre>optuna.visualization.plot_slice(study_1)</pre>	
	4	
In []:	<pre>optuna.visualization.plot_slice(study_2)</pre>	
In []:	optuna.visualization.plot_slice(study_3)	

Resultados do plot de otimização de cada estudo

Resultado da comparação de parâmetros

- Funções de ativação: no geral, todas as funções de otimização apresentaram bons resultados (como visto no primeiro estudo). Nos outros dois estudos é visível que as funções "softmax", "softplus" e "sigmoid" foram as piores.
- Dilatação: no terceiro estudo, a rede permitiu apenas uso da dilatação com 1, portanto, vamos desconsiderá-lo. Os outros dois estudos apresentaram melhores resultados quando a dilatação foi 2.
- · A função de perda utilizada foi a mesma em todos.
- O número de neurônios foi um parâmetro bastante inconstante, não aparentou ser um diferencial em nenhum dos estudos.
- O otimizador que mais se destacou foi o "adam", e em segundo o "sgd", os outros demonstraram ser igualmente bons, com um pouco mais de outliers.

In []: optuna.visualization.plot_param_importances(study_1)

```
In [ ]: optuna.visualization.plot_param_importances(study_2)
```

```
In [ ]: optuna.visualization.plot_param_importances(study_3)
         ZeroDivisionError Traceback (most recent call last) <ipython-input-27-eba610d4746b> in <module>()
          ---> 1 optuna.visualization.plot_param_importances(study_3)
         /usr/local/lib/python3.7/dist-packages/optuna/visualization/_param_importances.py in plot_param_importances(study, evaluator, params, target, target_name)
         111
112
--> 113
114
                     importances = optuna.importance.get_param_importances(
    study, evaluator=evaluator, params=params, target=target
                     )
            115
         /usr/local/lib/python3.7/dist-packages/optuna/importance/__init__.py in get_param_importances(study, evaluator, params, target)
91 raise TypeError("Evaluator must be a subclass of BaseImportanceEvaluator.")
              92
         ---> 93
                     return evaluator.evaluate(study, params=params, target=target)
        128
         /usr/local/lib/python3.7/dist-packages/optuna/importance/_fanova/_panova.py in fit(self, X, y, search_spaces, column_to_encoded_columns)
22 self._forest.fit(X, y)
              72
73
                          self.\_trees = [\_FanovaTree(e.tree\_, search\_spaces) \ for e in self.\_forest.estimators\_] \\ self.\_column\_to\_encoded\_columns = column\_to\_encoded\_columns \\ self.\_variances = \{\}
         ---> 74
         73
                         self._trees = [_FanovaTree(e.tree_, search_spaces) for e in self._forest.estimators_]
```

```
75 self._column_to_encoded_columns = column_to_encoded_columns
76 self._variances = {}

/usr/local/lib/python3.7/dist-packages/optuna/importance/_fanova/_tree.py in __init__(self, tree, search_spaces)
21 self._search_spaces = search_spaces
22 statistics = self._precompute_statistics()
24 split_midpoints, split_sizes = self._precompute_split_midpoints_and_sizes()
25 subtree_active_features = self._precompute_subtree_active_features()

/usr/local/lib/python3.7/dist-packages/optuna/importance/_fanova/_tree.py in __precompute_statistics(self)
180 child_values.append(statistics(child_node_index, 0])
181 child_weights.append(statistics(child_node_index, 0])
182 value = numpy, average(child_values, weights=child_weights)
183 weight = float(numpy, sum(child_weights))
184 statistics[node_index] = [value, weight]

<_array_function__ internals> in average(*args, **kwargs)

/usr/local/lib/python3.7/dist-packages/numpy/lib/function_base.py in average(a, axis, weights, returned)
408 if np.any(scl == 0.0):
409 raise ZeroDivisionError(
--> 410 "Weights sum to zero, can't be normalized")
411
412 avg = np.multiply(a, wgt, dtype=result_dtype).sum(axis)/scl

ZeroDivisionError: Weights sum to zero, can't be normalized
```

Diferenciais

A função de ativação foi alta nos dois primeiros estudos, sendo um diferencial importante, os outros fatores pouco impactaram. No terceiro estudo houve algum problema no optuna que impossibilitou a análise.