Modelo T com data augmentation e optimizer RMSProp

```
import ison
import os
import matplotlib.pyplot as plt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import GlobalAveragePooling2D, Dropout,
Dense, BatchNormalization
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping,
ReduceLROnPlateau, CSVLogger
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.applications import ResNet50
from tensorflow.keras.regularizers import 12
from tensorflow.keras.mixed precision import set global policy
# MIX precision training -- facilita no treino!
set global policy('mixed float16')
os.environ['TF CPP MIN LOG LEVEL'] = '2'
# CONSTANTES
BATCH SIZE = 64
IMG SIZE = 150
NUM CLASSES = 10 # n^{\circ} classes para identificar
NUM EPOCHS = 60
LEARNING RATE = 0.0001
DENSE LAYERS = [1024, 512, 256, 128]
INFO:tensorflow:Mixed precision compatibility check (mixed float16):
0K
Your GPU will likely run quickly with dtype policy mixed float16 as it
has compute capability of at least 7.0. Your GPU: NVIDIA GeForce RTX
4070, compute capability 8.9
# Folders do dataset
validation dir = './dataset/validation'
test dir = './dataset/test'
# Data Augmentation
train datagen = ImageDataGenerator(
    rescale=1./255,
    rotation range=20, # Increase rotation range
   width shift range=0.1,
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height shift range=0.1,
    shear range=0.1,
    zoom range=0.1,
    horizontal flip=True,
    vertical flip=True, # Adicionar flip vertical
    brightness range=[0.6, 1.4], # Adicionar range te brilho
    fill mode='nearest')
validation datagen = ImageDataGenerator(rescale=1./255)
test datagen = ImageDataGenerator(rescale=1./255)
# training generators
train generators = [train datagen.flow from directory(
    train dir,
    target_size=(IMG SIZE, IMG SIZE),
    batch size=BATCH SIZE,
    class mode='categorical') for train dir in train dirs]
# Necessário para juntar os trainning generators and repeat
def combined generator(generators):
    while True:
        for generator in generators:
            for batch in generator:
                yield batch
train_generator = combined_generator(train_generators)
# Validation e test generators
validation generator = validation datagen.flow from directory(
    validation dir,
    target size=(IMG SIZE, IMG SIZE),
    batch size=BATCH SIZE,
    class_mode='categorical')
test generator = test datagen.flow from directory(
    test dir,
    target size=(IMG SIZE, IMG SIZE),
    batch size=BATCH SIZE,
    class mode='categorical')
# Load the pre-trained ResNet50 model without the top layer and adjust
input shape
base_model = ResNet50(weights='imagenet', include_top=False,
                      input shape=(IMG SIZE, IMG SIZE, 3))
# Descongelar camadas (nao meter valores demasiado altos)
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for layer in base model.layers[-100:]:
    laver.trainable = True
Found 10000 images belonging to 10 classes.
from tensorflow.keras import backend as K
from tensorflow.keras.metrics import Metric
class Precision(Metric):
    def __init__(self, name='precision', **kwargs):
        super(Precision, self).__init__(name=name, **kwargs)
        self.true positives = self.add weight(name='tp',
initializer='zeros')
        self.predicted positives = self.add weight(name='pp',
initializer='zeros')
    def update state(self, y true, y pred, sample weight=None):
        y_pred = K.round(y_pred)
        y true = K.cast(y true, 'float32')
        self.true positives.assign add(K.sum(y true * y pred))
        self.predicted positives.assign add(K.sum(y pred))
    def result(self):
        return self.true_positives / (self.predicted_positives +
K.epsilon())
    def reset states(self):
        self.true positives.assign(0)
        self.predicted positives.assign(0)
class Recall(Metric):
    def init (self, name='recall', **kwargs):
        super(Recall, self). init (name=name, **kwargs)
        self.true positives = self.add weight(name='tp',
initializer='zeros')
        self.actual positives = self.add weight(name='ap',
initializer='zeros')
    def update state(self, y true, y pred, sample weight=None):
        y pred = K.round(y pred)
        y_true = K.cast(y_true, 'float32')
        self.true positives.assign add(K.sum(y true * y pred))
        self.actual positives.assign add(K.sum(y true))
    def result(self):
```

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return self.true positives / (self.actual positives +
K.epsilon())
    def reset states(self):
        self.true positives.assign(0)
        self.actual positives.assign(0)
class F1Score(Metric):
    def init (self, name='f1 score', **kwargs):
        super(F1Score, self). init (name=name, **kwargs)
        self.precision = Precision()
        self.recall = Recall()
    def update_state(self, y_true, y_pred, sample_weight=None):
        self.precision.update state(y true, y pred)
        self.recall.update state(y true, y pred)
    def result(self):
        precision = self.precision.result()
        recall = self.recall.result()
        return 2 * ((precision * recall) / (precision + recall +
K.epsilon()))
    def reset states(self):
        self.precision.reset states()
        self.recall.reset states()
# Definir as layers do modelo com parametros ajustados para reduzir o
overfitting
model = Sequential([
    base model.
    BatchNormalization(),
    GlobalAveragePooling2D(),
    # Increase model complexity
    Dense(DENSE_LAYERS[0], activation='relu',
kernel regularizer=12(0.03)),
    Dropout(0.5), # High dropout rate for regularization
    BatchNormalization(),
    Dense(DENSE LAYERS[1], activation='relu',
kernel regularizer=12(0.03)),
    Dropout (0.5),
    BatchNormalization(),
    Dense(DENSE LAYERS[2], activation='relu',
kernel regularizer=12(0.03)),
    Dropout (0.5),
    Dense(DENSE LAYERS[3], activation='relu',
kernel regularizer=12(0.03)),
    Dropout (0.5),
    BatchNormalization(),
    Dense(NUM CLASSES, activation='softmax', dtype='float32')
```

Model: "sequential"

Layer (type)	Output Shape	Param #
resnet50 (Functional)	(None, 5, 5, 2048)	23587712
<pre>batch_normalization (BatchN ormalization)</pre>	(None, 5, 5, 2048)	8192
<pre>global_average_pooling2d (G lobalAveragePooling2D)</pre>	(None, 2048)	0
dense (Dense)	(None, 1024)	2098176
dropout (Dropout)	(None, 1024)	0
<pre>batch_normalization_1 (Batc hNormalization)</pre>	(None, 1024)	4096
dense_1 (Dense)	(None, 512)	524800
dropout_1 (Dropout)	(None, 512)	0
<pre>batch_normalization_2 (Batc hNormalization)</pre>	(None, 512)	2048
dense_2 (Dense)	(None, 256)	131328
dropout_2 (Dropout)	(None, 256)	0
dense_3 (Dense)	(None, 128)	32896
dropout_3 (Dropout)	(None, 128)	0
<pre>batch_normalization_3 (Batc hNormalization)</pre>	(None, 128)	512
dense_4 (Dense)	(None, 10)	1290

Total params: 26,391,050

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Trainable params: 26,330,506
Non-trainable params: 60,544
# CALLBACKS
os.makedirs('logs', exist ok=True)
checkpoint =
ModelCheckpoint(f'models/modelo T com data augmentation rmsprop.keras'
                            monitor='val accuracy', verbose=1,
save best only=True, mode='max')
early stopping = EarlyStopping(
   monitor='val loss', patience=10, restore best weights=True) #
Increased patience
reduce lr = ReduceLROnPlateau(
   monitor='val loss', factor=0.2, patience=4, min lr=1e-7,
verbose=1) # More aggressive schedule
csv logger = CSVLogger(
    f'logs/modelo T com data augmentation rmsprop.csv', separator=',',
append=False)
# calcular passos por epoch
steps per epoch = sum([gen.samples // BATCH SIZE for gen in
train generators])
validation steps = validation generator.samples // BATCH SIZE
# calcular passos por epoch
# Treinar o modelo - Nao tirar os callbacks
history = model.fit(
   train generator,
   steps per epoch=steps per epoch,
   epochs=NUM EPOCHS,
   validation data=validation generator,
   validation steps=validation steps,
   callbacks=[checkpoint, early stopping, reduce lr, csv logger]
)
# Avaliar o modelo no test generator
# Avaliar o modelo no test generator
results = model.evaluate(test generator)
loss, accuracy, precision, recall, f1 score = results[:5]
print(f"Test Loss: {loss}")
print(f"Test Accuracy: {accuracy}")
print(f"Test Precision: {precision}")
print(f"Test Recall: {recall}")
print(f"Test F1 Score: {f1 score}")
Epoch 1/60
```

```
accuracy: 0.1960 - precision: 0.3136 - recall: 0.0659 - f1 score:
0.1089
c:\Users\USER\.conda\envs\py310\lib\site-packages\keras\engine\
training.py:2319: UserWarning: Metric Precision implements a
`reset states()` method; rename it to `reset state()` (without the
final "s"). The name `reset states()` has been deprecated to improve
API consistency.
 m.reset state()
c:\Users\USER\.conda\envs\py310\lib\site-packages\keras\engine\
training.py:2319: UserWarning: Metric Recall implements a
`reset_states()` method; rename it to `reset_state()` (without the
final "s"). The name `reset states()` has been deprecated to improve
API consistency.
 m.reset state()
c:\Users\USER\.conda\envs\py310\lib\site-packages\keras\engine\
training.py:2319: UserWarning: Metric F1Score implements a
`reset states()` method; rename it to `reset state()` (without the
final "s"). The name `reset states()` has been deprecated to improve
API consistency.
 m.reset state()
Epoch 1: val accuracy improved from -inf to 0.14233, saving model to
models\modelo T com data augmentation.keras
54.2261 - accuracy: 0.1960 - precision: 0.3136 - recall: 0.0659 -
f1_score: 0.1089 - val_loss: 34.5487 - val_accuracy: 0.1423 -
val precision: 0.0000e+00 - val recall: 0.0000e+00 - val f1 score:
0.0000e+00 - lr: 1.0000e-04
Epoch 2/60
accuracy: 0.4836 - precision: 0.6536 - recall: 0.2937 - f1 score:
Epoch 2: val accuracy improved from 0.14233 to 0.72516, saving model
to models\modelo T com data augmentation.keras
21.9455 - accuracy: 0.4836 - precision: 0.6536 - recall: 0.2937 -
f1 score: 0.4052 - val loss: 12.4552 - val accuracy: 0.7252 -
val precision: 0.8539 - val recall: 0.5697 - val f1 score: 0.6834 -
lr: 1.0000e-04
Epoch 3/60
accuracy: 0.6660 - precision: 0.7825 - recall: 0.5227 - f1 score:
0.6267
Epoch 3: val accuracy improved from 0.72516 to 0.78746, saving model
to models\modelo T com data augmentation.keras
8.1400 - accuracy: 0.6660 - precision: 0.7825 - recall: 0.5227 -
f1 score: 0.6267 - val loss: 4.6251 - val accuracy: 0.7875 -
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val precision: 0.8772 - val recall: 0.6931 - val f1 score: 0.7744 -
lr: 1.0000e-04
Epoch 4/60
accuracy: 0.7371 - precision: 0.8242 - recall: 0.6336 - f1 score:
0.7164
Epoch 4: val accuracy improved from 0.78746 to 0.83313, saving model
to models\modelo T com data augmentation.keras
3.3969 - accuracy: 0.7371 - precision: 0.8242 - recall: 0.6336 -
f1 score: 0.7164 - val loss: 2.0749 - val accuracy: 0.8331 -
val_precision: 0.8833 - val_recall: 0.7769 - val_f1_score: 0.8267 -
lr: 1.0000e-04
Epoch 5/60
accuracy: 0.7725 - precision: 0.8456 - recall: 0.6903 - f1 score:
0.7601
Epoch 5: val accuracy did not improve from 0.83313
624/624 [=============] - 165s 265ms/step - loss:
1.8381 - accuracy: 0.7725 - precision: 0.8456 - recall: 0.6903 -
f1 score: 0.7601 - val loss: 1.3554 - val accuracy: 0.8133 -
val precision: 0.8647 - val recall: 0.7640 - val f1 score: 0.8112 -
lr: 1.0000e-04
Epoch 6/60
624/624 [============== ] - ETA: 0s - loss: 1.2571 -
accuracy: 0.7973 - precision: 0.8570 - recall: 0.7278 - f1 score:
0.7871
Epoch 6: val accuracy did not improve from 0.83313
1.2571 - accuracy: 0.7973 - precision: 0.8570 - recall: 0.7278 -
f1 score: 0.7871 - val loss: 1.0575 - val accuracy: 0.8084 -
val precision: 0.8514 - val recall: 0.7657 - val f1 score: 0.8063 -
lr: 1.0000e-04
Epoch 7/60
accuracy: 0.8186 - precision: 0.8702 - recall: 0.7608 - f1 score:
0.8118
Epoch 7: val accuracy did not improve from 0.83313
1.0065 - accuracy: 0.8186 - precision: 0.8702 - recall: 0.7608 -
f1 score: 0.8118 - val loss: 1.0067 - val accuracy: 0.7863 -
val precision: 0.8286 - val recall: 0.7479 - val f1 score: 0.7862 -
lr: 1.0000e-04
Epoch 8/60
accuracy: 0.8255 - precision: 0.8741 - recall: 0.7742 - f1 score:
0.8211
Epoch 8: val accuracy did not improve from 0.83313
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0.8948 - accuracy: 0.8255 - precision: 0.8741 - recall: 0.7742 -
f1 score: 0.8211 - val loss: 0.8668 - val accuracy: 0.8132 -
val precision: 0.8417 - val recall: 0.7926 - val f1 score: 0.8164 -
lr: 1.0000e-04
Epoch 9/60
accuracy: 0.8404 - precision: 0.8833 - recall: 0.7968 - f1 score:
0.8378
Epoch 9: val accuracy did not improve from 0.83313
0.8092 - accuracy: 0.8404 - precision: 0.8833 - recall: 0.7968 -
f1 score: 0.8378 - val loss: 0.8013 - val accuracy: 0.8220 -
val precision: 0.8499 - val recall: 0.8028 - val f1 score: 0.8257 -
lr: 1.0000e-04
Epoch 10/60
624/624 [=============== ] - ETA: 0s - loss: 0.7587 -
accuracy: 0.8495 - precision: 0.8875 - recall: 0.8096 - f1 score:
0.8468
Epoch 10: val accuracy did not improve from 0.83313
0.7587 - accuracy: 0.8495 - precision: 0.8875 - recall: 0.8096 -
f1 score: 0.8468 - val loss: 0.7663 - val accuracy: 0.8293 -
val precision: 0.8624 - val recall: 0.8054 - val f1 score: 0.8329 -
lr: 1.0000e-04
Epoch 11/60
accuracy: 0.8607 - precision: 0.8968 - recall: 0.8251 - f1_score:
0.8595
Epoch 11: val accuracy improved from 0.83313 to 0.83494, saving model
to models\modelo_T_com_data_augmentation.keras
0.7058 - accuracy: 0.8607 - precision: 0.8968 - recall: 0.8251 -
f1 score: 0.8595 - val loss: 0.7537 - val accuracy: 0.8349 -
val precision: 0.8562 - val recall: 0.8195 - val f1 score: 0.8375 -
lr: 1.0000e-04
Epoch 12/60
accuracy: 0.8669 - precision: 0.9002 - recall: 0.8341 - f1 score:
0.8659
Epoch 12: val accuracy did not improve from 0.83494
0.6781 - accuracy: 0.8669 - precision: 0.9002 - recall: 0.8341 -
f1 score: 0.8659 - val loss: 0.7793 - val accuracy: 0.8315 -
val precision: 0.8599 - val recall: 0.8132 - val f1 score: 0.8359 -
lr: 1.0000e-04
Epoch 13/60
accuracy: 0.8742 - precision: 0.9058 - recall: 0.8438 - f1 score:
0.8737
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Epoch 13: val accuracy did not improve from 0.83494
0.6483 - accuracy: 0.8742 - precision: 0.9058 - recall: 0.8438 -
f1 score: 0.8737 - val loss: 0.9145 - val accuracy: 0.8076 -
val precision: 0.8289 - val_recall: 0.7952 - val_f1_score: 0.8117 -
lr: 1.0000e-04
Epoch 14/60
accuracy: 0.8817 - precision: 0.9097 - recall: 0.8549 - f1 score:
0.8815
Epoch 14: val accuracy did not improve from 0.83494
0.6276 - accuracy: 0.8817 - precision: 0.9097 - recall: 0.8549 -
f1 score: 0.8815 - val loss: 0.7960 - val_accuracy: 0.8314 -
val_precision: 0.8526 - val_recall: 0.8154 - val_f1_score: 0.8336 -
lr: 1.0000e-04
Epoch 15/60
accuracy: 0.8843 - precision: 0.9124 - recall: 0.8579 - f1 score:
0.8843
Epoch 15: val accuracy did not improve from 0.83494
Epoch 15: ReduceLROnPlateau reducing learning rate to
1.9999999494757503e-05.
0.6171 - accuracy: 0.8843 - precision: 0.9124 - recall: 0.8579 -
f1 score: 0.8843 - val loss: 0.8935 - val accuracy: 0.8086 -
val precision: 0.8332 - val recall: 0.7936 - val f1 score: 0.8129 -
lr: 1.0000e-04
Epoch 16/60
accuracy: 0.9230 - precision: 0.9405 - recall: 0.9064 - f1 score:
0.9231
Epoch 16: val accuracy improved from 0.83494 to 0.88011, saving model
to models\modelo_T_com_data_augmentation.keras
0.4505 - accuracy: 0.9230 - precision: 0.9405 - recall: 0.9064 -
f1 score: 0.9231 - val loss: 0.5454 - val accuracy: 0.8801 -
val precision: 0.8921 - val recall: 0.8699 - val f1 score: 0.8808 -
lr: 2.0000e-05
Epoch 17/60
accuracy: 0.9377 - precision: 0.9508 - recall: 0.9249 - f1 score:
0.9377
Epoch 17: val accuracy improved from 0.88011 to 0.88752, saving model
to models\modelo_T_com_data_augmentation.keras
0.3569 - accuracy: 0.9377 - precision: 0.9508 - recall: 0.9249 -
fl_score: 0.9377 - val_loss: 0.5075 - val_accuracy: 0.8875 -
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val precision: 0.8985 - val recall: 0.8791 - val f1 score: 0.8887 -
lr: 2.0000e-05
Epoch 18/60
accuracy: 0.9430 - precision: 0.9553 - recall: 0.9322 - f1 score:
Epoch 18: val accuracy did not improve from 0.88752
0.3205 - accuracy: 0.9430 - precision: 0.9553 - recall: 0.9322 -
f1 score: 0.9436 - val loss: 0.5404 - val accuracy: 0.8787 -
val precision: 0.8885 - val recall: 0.8714 - val f1 score: 0.8799 -
lr: 2.0000e-05
Epoch 19/60
accuracy: 0.9508 - precision: 0.9607 - recall: 0.9406 - f1 score:
0.9506
Epoch 19: val accuracy did not improve from 0.88752
0.2858 - accuracy: 0.9508 - precision: 0.9607 - recall: 0.9406 -
f1 score: 0.9506 - val loss: 0.5321 - val accuracy: 0.8792 -
val precision: 0.8897 - val recall: 0.8736 - val f1 score: 0.8816 -
lr: 2.0000e-05
Epoch 20/60
accuracy: 0.9539 - precision: 0.9619 - recall: 0.9452 - f1 score:
0.9535
Epoch 20: val accuracy did not improve from 0.88752
0.2715 - accuracy: 0.9539 - precision: 0.9619 - recall: 0.9452 -
fl_score: 0.9535 - val_loss: 0.5301 - val_accuracy: 0.8827 -
val precision: 0.8912 - val recall: 0.8767 - val f1 score: 0.8839 -
lr: 2.0000e-05
Epoch 21/60
accuracy: 0.9545 - precision: 0.9629 - recall: 0.9468 - f1 score:
0.9548
Epoch 21: val accuracy did not improve from 0.88752
Epoch 21: ReduceLROnPlateau reducing learning rate to
3.999999898951501e-06.
0.2632 - accuracy: 0.9545 - precision: 0.9629 - recall: 0.9468 -
f1 score: 0.9548 - val loss: 0.5169 - val accuracy: 0.8831 -
val precision: 0.8924 - val recall: 0.8793 - val f1 score: 0.8858 -
lr: 2.0000e-05
Epoch 22/60
accuracy: 0.9600 - precision: 0.9679 - recall: 0.9532 - f1 score:
0.9605
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Epoch 22: val accuracy improved from 0.88752 to 0.89042, saving model
to models\modelo T com data augmentation.keras
0.2378 - accuracy: 0.9600 - precision: 0.9679 - recall: 0.9532 -
f1 score: 0.9605 - val loss: 0.4888 - val accuracy: 0.8904 -
val precision: 0.8983 - val recall: 0.8851 - val f1 score: 0.8916 -
lr: 4.0000e-06
Epoch 23/60
accuracy: 0.9635 - precision: 0.9705 - recall: 0.9570 - f1 score:
0.9637
Epoch 23: val accuracy did not improve from 0.89042
0.2173 - accuracy: 0.9635 - precision: 0.9705 - recall: 0.9570 -
fl_score: 0.9637 - val_loss: 0.4903 - val_accuracy: 0.8894 -
val precision: 0.8982 - val recall: 0.8843 - val f1 score: 0.8912 -
lr: 4.0000e-06
Epoch 24/60
accuracy: 0.9657 - precision: 0.9720 - recall: 0.9598 - f1 score:
0.9659
Epoch 24: val accuracy improved from 0.89042 to 0.89223, saving model
to models\modelo T com data augmentation.keras
0.2097 - accuracy: 0.9657 - precision: 0.9720 - recall: 0.9598 -
f1 score: 0.9659 - val loss: 0.4815 - val accuracy: 0.8922 -
val_precision: 0.8995 - val recall: 0.887\overline{2} - val f1 score: 0.8933 -
lr: 4.0000e-06
Epoch 25/60
624/624 [============== ] - ETA: 0s - loss: 0.2047 -
accuracy: 0.9665 - precision: 0.9732 - recall: 0.9605 - f1 score:
0.9668
Epoch 25: val_accuracy did not improve from 0.89223
0.2047 - accuracy: 0.9665 - precision: 0.9732 - recall: 0.9605 -
f1 score: 0.9668 - val loss: 0.4787 - val accuracy: 0.8919 -
val precision: 0.8990 - val recall: 0.8869 - val f1 score: 0.8929 -
lr: 4.0000e-06
Epoch 26/60
accuracy: 0.9676 - precision: 0.9731 - recall: 0.9626 - f1 score:
0.9679
Epoch 26: val_accuracy did not improve from 0.89223
0.1971 - accuracy: 0.9676 - precision: 0.9731 - recall: 0.9626 -
fl_score: 0.9679 - val_loss: 0.4849 - val_accuracy: 0.8901 -
val precision: 0.8986 - val recall: 0.8862 - val f1 score: 0.8924 -
lr: 4.0000e-06
Epoch 27/60
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accuracy: 0.9707 - precision: 0.9759 - recall: 0.9654 - f1 score:
0.9706
Epoch 27: val accuracy improved from 0.89223 to 0.89243, saving model
to models\modelo T com data augmentation.keras
0.1875 - accuracy: 0.9707 - precision: 0.9759 - recall: 0.9654 -
f1 score: 0.9706 - val loss: 0.4841 - val accuracy: 0.8924 -
val precision: 0.9005 - val recall: 0.8879 - val f1 score: 0.8941 -
lr: 4.0000e-06
Epoch 28/60
accuracy: 0.9702 - precision: 0.9747 - recall: 0.9650 - f1_score:
Epoch 28: val_accuracy did not improve from 0.89243
0.1852 - accuracy: 0.9702 - precision: 0.9747 - recall: 0.9650 -
fl_score: 0.9698 - val_loss: 0.4751 - val_accuracy: 0.8918 -
val precision: 0.8990 - val recall: 0.8873 - val f1 score: 0.8931 -
lr: 4.0000e-06
Epoch 29/60
accuracy: 0.9694 - precision: 0.9749 - recall: 0.9647 - f1 score:
0.9698
Epoch 29: val accuracy did not improve from 0.89243
0.1830 - accuracy: 0.9694 - precision: 0.9749 - recall: 0.9647 -
f1 score: 0.9698 - val loss: 0.4811 - val accuracy: 0.8914 -
val precision: 0.8975 - val recall: 0.8879 - val f1 score: 0.8927 -
lr: 4.0000e-06
Epoch 30/60
accuracy: 0.9722 - precision: 0.9768 - recall: 0.9672 - f1 score:
0.9720
Epoch 30: val accuracy did not improve from 0.89243
0.1755 - accuracy: 0.9722 - precision: 0.9768 - recall: 0.9672 -
f1 score: 0.9720 - val loss: 0.4828 - val accuracy: 0.8895 -
val precision: 0.8968 - val recall: 0.8868 - val f1 score: 0.8918 -
lr: 4.0000e-06
Epoch 31/60
accuracy: 0.9705 - precision: 0.9752 - recall: 0.9654 - f1 score:
0.9702
Epoch 31: val accuracy improved from 0.89243 to 0.89253, saving model
to models\modelo_T_com_data_augmentation.keras
0.1772 - accuracy: 0.9705 - precision: 0.9752 - recall: 0.9654 -
f1 score: 0.9702 - val loss: 0.4794 - val accuracy: 0.8925 -
```

```
val precision: 0.8993 - val recall: 0.8885 - val f1 score: 0.8939 -
lr: 4.0000e-06
Epoch 32/60
accuracy: 0.9729 - precision: 0.9781 - recall: 0.9691 - f1 score:
0.9736
Epoch 32: val accuracy did not improve from 0.89253
Epoch 32: ReduceLROnPlateau reducing learning rate to
7.999999979801942e-07.
0.1660 - accuracy: 0.9729 - precision: 0.9781 - recall: 0.9691 -
f1 score: 0.9736 - val loss: 0.4801 - val accuracy: 0.8919 -
val precision: 0.8979 - val recall: 0.8876 - val f1 score: 0.8927 -
lr: 4.0000e-06
Epoch 33/60
accuracy: 0.9737 - precision: 0.9781 - recall: 0.9690 - f1 score:
Epoch 33: val accuracy improved from 0.89253 to 0.89263, saving model
to models\modelo T com data augmentation.keras
624/624 [=============] - 167s 268ms/step - loss:
0.1676 - accuracy: 0.9737 - precision: 0.9781 - recall: 0.9690 -
f1 score: 0.9735 - val loss: 0.4792 - val accuracy: 0.8926 -
val_precision: 0.8996 - val_recall: 0.8888 - val_f1_score: 0.8942 -
lr: 8.0000e-07
Epoch 34/60
accuracy: 0.9740 - precision: 0.9791 - recall: 0.9693 - f1 score:
0.9742
Epoch 34: val_accuracy did not improve from 0.89263
0.1619 - accuracy: 0.9740 - precision: 0.9791 - recall: 0.9693 -
f1 score: 0.9742 - val loss: 0.4778 - val accuracy: 0.8925 -
val precision: 0.8992 - val recall: 0.8886 - val f1 score: 0.8939 -
lr: 8.0000e-07
Epoch 35/60
accuracy: 0.9744 - precision: 0.9785 - recall: 0.9697 - fl_score:
Epoch 35: val accuracy improved from 0.89263 to 0.89343, saving model
to models\modelo T com data augmentation.keras
0.1602 - accuracy: 0.9744 - precision: 0.9785 - recall: 0.9697 -
f1 score: 0.9741 - val loss: 0.4750 - val accuracy: 0.8934 -
val precision: 0.9009 - val recall: 0.8889 - val f1 score: 0.8949 -
lr: 8.0000e-07
Epoch 36/60
```

```
accuracy: 0.9739 - precision: 0.9789 - recall: 0.9693 - f1 score:
0.9741
Epoch 36: val accuracy did not improve from 0.89343
Epoch 36: ReduceLROnPlateau reducing learning rate to
1.600000018697756e-07.
0.1597 - accuracy: 0.9739 - precision: 0.9789 - recall: 0.9693 -
f1_score: 0.9741 - val_loss: 0.4773 - val_accuracy: 0.8927 -
val precision: 0.9003 - val recall: 0.8883 - val f1 score: 0.8943 -
lr: 8.0000e-07
Epoch 37/60
accuracy: 0.9738 - precision: 0.9785 - recall: 0.9696 - f1 score:
0.9740
Epoch 37: val accuracy improved from 0.89343 to 0.89363, saving model
to models\modelo_T_com_data_augmentation.keras
0.1597 - accuracy: 0.9738 - precision: 0.9785 - recall: 0.9696 -
f1 score: 0.9740 - val loss: 0.4768 - val accuracy: 0.8936 -
val precision: 0.9007 - val recall: 0.8896 - val f1 score: 0.8951 -
lr: 1.6000e-07
Epoch 38/60
accuracy: 0.9743 - precision: 0.9793 - recall: 0.9702 - f1 score:
0.9747
Epoch 38: val accuracy did not improve from 0.89363
0.1607 - accuracy: 0.9743 - precision: 0.9793 - recall: 0.9702 -
f1 score: 0.9747 - val loss: 0.4787 - val accuracy: 0.8932 -
val precision: 0.8994 - val recall: 0.8888 - val f1 score: 0.8941 -
lr: 1.6000e-07
Epoch 39/60
accuracy: 0.9737 - precision: 0.9788 - recall: 0.9693 - f1 score:
0.9740
Epoch 39: val accuracy improved from 0.89363 to 0.89413, saving model
to models\modelo T com data augmentation.keras
0.1600 - accuracy: 0.9737 - precision: 0.9788 - recall: 0.9693 -
f1 score: 0.9740 - val loss: 0.4768 - val accuracy: 0.8941 -
val_precision: 0.9013 - val_recall: 0.8899 - val f1 score: 0.8956 -
lr: 1.6000e-07
Epoch 40/60
accuracy: 0.9735 - precision: 0.9787 - recall: 0.9696 - f1 score:
0.9741
Epoch 40: val accuracy did not improve from 0.89413
```

```
Epoch 40: ReduceLROnPlateau reducing learning rate to 1e-07.
0.1599 - accuracy: 0.9735 - precision: 0.9787 - recall: 0.9696 -
f1 score: 0.9741 - val loss: 0.4751 - val accuracy: 0.8941 -
val_precision: 0.9020 - val_recall: 0.8893 - val f1 score: 0.8956 -
lr: 1.6000e-07
Epoch 41/60
accuracy: 0.9758 - precision: 0.9797 - recall: 0.9714 - f1 score:
0.9755
Epoch 41: val accuracy did not improve from 0.89413
0.1568 - accuracy: 0.9758 - precision: 0.9797 - recall: 0.9714 -
f1 score: 0.9755 - val loss: 0.4768 - val accuracy: 0.8933 -
val_precision: 0.9006 - val_recall: 0.8890 - val_f1_score: 0.8948 -
lr: 1.0000e-07
Epoch 42/60
accuracy: 0.9746 - precision: 0.9796 - recall: 0.9705 - f1 score:
0.9750
Epoch 42: val accuracy did not improve from 0.89413
0.1574 - accuracy: 0.9746 - precision: 0.9796 - recall: 0.9705 -
f1 score: 0.9750 - val loss: 0.4780 - val accuracy: 0.8934 -
val precision: 0.9005 - val recall: 0.8886 - val f1 score: 0.8945 -
lr: 1.0000e-07
Epoch 43/60
accuracy: 0.9732 - precision: 0.9778 - recall: 0.9690 - f1 score:
0.9734
Epoch 43: val_accuracy did not improve from 0.89413
0.1618 - accuracy: 0.9732 - precision: 0.9778 - recall: 0.9690 -
f1 score: 0.9734 - val loss: 0.4767 - val accuracy: 0.8931 -
val precision: 0.9006 - val recall: 0.8893 - val f1 score: 0.8949 -
lr: 1.0000e-07
Epoch 44/60
accuracy: 0.9750 - precision: 0.9797 - recall: 0.9711 - f1_score:
0.9754
Epoch 44: val accuracy did not improve from 0.89413
0.1562 - accuracy: 0.9750 - precision: 0.9797 - recall: 0.9711 -
f1 score: 0.9754 - val loss: 0.4758 - val accuracy: 0.8928 -
val precision: 0.9003 - val recall: 0.8893 - val f1 score: 0.8948 -
lr: 1.0000e-07
Epoch 45/60
accuracy: 0.9753 - precision: 0.9795 - recall: 0.9713 - f1 score:
```

```
0.9754
Epoch 45: val accuracy did not improve from 0.89413
0.1583 - accuracy: 0.9753 - precision: 0.9795 - recall: 0.9713 -
f1 score: 0.9754 - val loss: 0.4748 - val accuracy: 0.8940 -
val precision: 0.9006 - val_recall: 0.8895 - val_f1_score: 0.8950 -
lr: 1.0000e-07
Epoch 46/60
accuracy: 0.9752 - precision: 0.9800 - recall: 0.9714 - f1 score:
0.9757
Epoch 46: val accuracy did not improve from 0.89413
0.1568 - accuracy: 0.9752 - precision: 0.9800 - recall: 0.9714 -
f1 score: 0.9757 - val loss: 0.4760 - val accuracy: 0.8932 -
val precision: 0.9005 - val recall: 0.8892 - val f1 score: 0.8948 -
lr: 1.0000e-07
Epoch 47/60
accuracy: 0.9748 - precision: 0.9789 - recall: 0.9708 - f1 score:
0.9748
Epoch 47: val accuracy did not improve from 0.89413
0.1570 - accuracy: 0.9748 - precision: 0.9789 - recall: 0.9708 -
f1 score: 0.9748 - val loss: 0.4755 - val accuracy: 0.8935 -
val precision: 0.9010 - val recall: 0.8901 - val f1 score: 0.8955 -
lr: 1.0000e-07
Epoch 48/60
accuracy: 0.9750 - precision: 0.9790 - recall: 0.9708 - f1 score:
0.9749
Epoch 48: val accuracy did not improve from 0.89413
0.1577 - accuracy: 0.9750 - precision: 0.9790 - recall: 0.9708 -
f1 score: 0.9749 - val loss: 0.4749 - val accuracy: 0.8939 -
val precision: 0.9016 - val recall: 0.8902 - val f1 score: 0.8959 -
lr: 1.0000e-07
Epoch 49/60
accuracy: 0.9758 - precision: 0.9797 - recall: 0.9717 - f1 score:
0.9757
Epoch 49: val accuracy did not improve from 0.89413
624/624 [============== ] - 171s 274ms/step - loss:
0.1542 - accuracy: 0.9758 - precision: 0.9797 - recall: 0.9717 -
f1 score: 0.9757 - val loss: 0.4748 - val accuracy: 0.8937 -
val_precision: 0.9008 - val_recall: 0.8898 - val_f1_score: 0.8953 -
lr: 1.0000e-07
Epoch 50/60
accuracy: 0.9763 - precision: 0.9812 - recall: 0.9726 - f1 score:
```

```
0.9769
Epoch 50: val accuracy did not improve from 0.89413
0.1514 - accuracy: 0.9763 - precision: 0.9812 - recall: 0.9726 -
f1 score: 0.9769 - val loss: 0.4754 - val accuracy: 0.8935 -
val precision: 0.9005 - val_recall: 0.8900 - val_f1_score: 0.8952 -
lr: 1.0000e-07
Epoch 51/60
accuracy: 0.9754 - precision: 0.9797 - recall: 0.9713 - f1 score:
0.9755
Epoch 51: val accuracy did not improve from 0.89413
0.1570 - accuracy: 0.9754 - precision: 0.9797 - recall: 0.9713 -
f1 score: 0.9755 - val loss: 0.4765 - val accuracy: 0.8931 -
val precision: 0.9017 - val recall: 0.8899 - val f1 score: 0.8958 -
lr: 1.0000e-07
Epoch 52/60
accuracy: 0.9742 - precision: 0.9794 - recall: 0.9705 - f1 score:
0.9749
Epoch 52: val accuracy did not improve from 0.89413
0.1580 - accuracy: 0.9742 - precision: 0.9794 - recall: 0.9705 -
f1 score: 0.9749 - val loss: 0.4755 - val accuracy: 0.8931 -
val precision: 0.9009 - val recall: 0.8894 - val_f1_score: 0.8951 -
lr: 1.0000e-07
Epoch 53/60
accuracy: 0.9756 - precision: 0.9799 - recall: 0.9714 - f1 score:
0.9756
Epoch 53: val accuracy did not improve from 0.89413
0.1557 - accuracy: 0.9756 - precision: 0.9799 - recall: 0.9714 -
f1 score: 0.9756 - val loss: 0.4755 - val accuracy: 0.8934 -
val precision: 0.9013 - val recall: 0.8888 - val f1 score: 0.8950 -
lr: 1.0000e-07
Epoch 54/60
624/624 [=============== ] - ETA: 0s - loss: 0.1584 -
accuracy: 0.9748 - precision: 0.9789 - recall: 0.9709 - f1 score:
0.9749
Epoch 54: val accuracy did not improve from 0.89413
624/624 [=============== ] - 171s 275ms/step - loss:
0.1584 - accuracy: 0.9748 - precision: 0.9789 - recall: 0.9709 -
f1 score: 0.9749 - val loss: 0.4764 - val accuracy: 0.8927 -
val_precision: 0.9005 - val recall: 0.8892 - val f1 score: 0.8948 -
lr: 1.0000e-07
Epoch 55/60
```

```
accuracy: 0.9766 - precision: 0.9803 - recall: 0.9726 - f1 score:
0.9764
Epoch 55: val accuracy did not improve from 0.89413
0.1505 - accuracy: 0.9766 - precision: 0.9803 - recall: 0.9726 -
f1 score: 0.9764 - val loss: 0.4749 - val accuracy: 0.8939 -
val precision: 0.9013 - val recall: 0.8895 - val f1 score: 0.8954 -
lr: 1.0000e-07
Epoch 56/60
accuracy: 0.9746 - precision: 0.9793 - recall: 0.9707 - f1 score:
Epoch 56: val accuracy improved from 0.89413 to 0.89433, saving model
to models\modelo T com data augmentation.keras
0.1561 - accuracy: 0.9746 - precision: 0.9793 - recall: 0.9707 -
f1 score: 0.9750 - val loss: 0.4727 - val accuracy: 0.8943 -
val_precision: 0.9018 - val_recall: 0.8899 - val_f1_score: 0.8958 -
lr: 1.0000e-07
Epoch 57/60
accuracy: 0.9743 - precision: 0.9791 - recall: 0.9703 - f1 score:
0.9747
Epoch 57: val accuracy did not improve from 0.89433
0.1577 - accuracy: 0.9743 - precision: 0.9791 - recall: 0.9703 -
f1_score: 0.9747 - val_loss: 0.4755 - val_accuracy: 0.8937 -
val precision: 0.9010 - val recall: 0.8892 - val f1 score: 0.8950 -
lr: 1.0000e-07
Epoch 58/60
accuracy: 0.9743 - precision: 0.9791 - recall: 0.9706 - f1 score:
0.9748
Epoch 58: val accuracy did not improve from 0.89433
0.1569 - accuracy: 0.9743 - precision: 0.9791 - recall: 0.9706 -
f1 score: 0.9748 - val loss: 0.4752 - val accuracy: 0.8936 -
val precision: 0.9008 - val recall: 0.8894 - val f1 score: 0.8951 -
lr: 1.0000e-07
Epoch 59/60
accuracy: 0.9760 - precision: 0.9805 - recall: 0.9713 - f1 score:
0.9759
Epoch 59: val accuracy did not improve from 0.89433
0.1531 - accuracy: 0.9760 - precision: 0.9805 - recall: 0.9713 -
f1 score: 0.9759 - val loss: 0.4768 - val accuracy: 0.8934 -
val_precision: 0.9008 - val_recall: 0.8885 - val f1 score: 0.8946 -
lr: 1.0000e-07
```

```
Epoch 60/60
accuracy: 0.9750 - precision: 0.9792 - recall: 0.9710 - f1 score:
0.9751
Epoch 60: val accuracy did not improve from 0.89433
0.1578 - accuracy: 0.9750 - precision: 0.9792 - recall: 0.9710 -
f1 score: 0.9751 - val loss: 0.4760 - val accuracy: 0.8936 -
val precision: 0.9016 - val recall: 0.8888 - val f1 score: 0.8951 -
lr: 1.0000e-07
- accuracy: 0.8898 - precision: 0.8963 - recall: 0.8860 - f1 score:
0.8911
Test Loss: 0.4904648959636688
Test Accuracy: 0.8898000121116638
Test Precision: 0.8963075280189514
Test Recall: 0.8859999775886536
Test F1 Score: 0.89112389087677
plt.figure(figsize=(12, 8))
plt.subplot(2, 1, 1)
plt.plot(history.history['accuracy'], label='train accuracy')
plt.plot(history.history['val_accuracy'], label='val_accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.ylim([0, 1])
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')
plt.subplot(2, 1, 2)
plt.plot(history.history['val precision'], label='val precision')
plt.plot(history.history['val recall'], label='val recall')
plt.plot(history.history['val_f1_score'], label='val_f1_score')
plt.xlabel('Epoch')
plt.ylabel('Metrics')
plt.ylim([0, 1])
plt.legend(loc='lower right')
plt.title('Validation Precision, Recall, F1 Score')
plt.savefig(f'./plots/modelo T com data augmentation rmsprop.png')
plt.tight_layout()
#plt.show()
```

