## Modelo T com data augmentation e optimizer Adam

```
import json
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 Adam
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.applications import ResNet50
from tensorflow.keras.regularizers import l2
from tensorflow.keras.mixed_precision import set_global_policy
```

## Definição de constantes

```
# 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  # nº classes para identificar
NUM_EPOCHS = 60
LEARNING_RATE = 0.0001
DENSE_LAYERS = [1024, 512, 256, 128]

INFO:tensorflow:Mixed precision compatibility check (mixed_float16):
OK
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
```

#### Data set

```
validation_dir = './dataset/validation'
test_dir = './dataset/test'
```

### Data augmentation

Aqui decidimos aplicar um data augmentation mais robusto, até para testar novos parametros o que se mostrou efetivo, (mesmo que a diferença nao seja tão nítida) é possivel constatar uma evolução

```
# Data Augmentation
train datagen = ImageDataGenerator(
    rescale=1./255,
    rotation range=20, # Increase rotation range
    width shift range=0.1,
    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 do modelo ResNet50 - deixar include top=False
base model = ResNet50(weights='imagenet', include top=False,
                      input_shape=(IMG SIZE, IMG SIZE, 3))
Found 10000 images belonging to 10 classes.
```

# UnFreeze Layers - 100 layers

Aqui descongelamos as ultimas 100 camadas antes das 50 descongeladas no modelo T sem data augmentaion para comparar os resultados, e percebemos que a melhoria não foi assim tão significativa, dado que o tempo de treino é bastante superior quanto mais camadas descongeladas.

```
# Descongelar camadas (nao meter valores demasiado altos)
for layer in base_model.layers[-100:]:
    layer.trainable = True
```

## Funções para obter as métricas

```
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)
```

```
v true = K.cast(v 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):
        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 o modelo

```
# Definir as layers do modelo com parametros ajustados para reduzir o
overfittina
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')
1)
# Compilar o modelo
model.compile(optimizer=Adam(learning rate=LEARNING RATE),
              loss='categorical crossentropy',
              metrics=['accuracy', Precision(), Recall(), F1Score()])
model.summary()
Model: "sequential"
                             Output Shape
Layer (type)
                                                        Param #
 resnet50 (Functional)
                              (None, 5, 5, 2048)
                                                        23587712
 batch normalization (BatchN (None, 5, 5, 2048)
                                                        8192
 ormalization)
 global average pooling2d (G (None, 2048)
                                                        0
```

lobal Average Dealing 2D		
lobalAveragePooling2D)		
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

# Callbacks

```
verbose=1) # More aggressive schedule
csv_logger = CSVLogger(
    f'logs/modelo_T_com_data_augmentation_adam.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
```

### Resultados

```
# 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 adam adam.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:
0.4052
Epoch 2: val accuracy improved from 0.14233 to 0.72516, saving model
to models\modelo T com data augmentation adam adam.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:
Epoch 3: val accuracy improved from 0.72516 to 0.78746, saving model
to models\modelo T com data augmentation adam adam.keras
8.1400 - accuracy: 0.6660 - precision: 0.7825 - recall: 0.5227 -
f1_score: 0.6267 - val_loss: 4.6251 - val accuracy: 0.7875 -
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_adam_adam.keras
624/624 [==============] - 166s 267ms/step - loss:
3.3969 - accuracy: 0.7371 - precision: 0.8242 - recall: 0.6336 -
fl_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:
Epoch 5: val_accuracy did not improve from 0.83313
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
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 -
fl_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
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
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 adam adam.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
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:
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_adam_adam.keras
0.4505 - accuracy: 0.9230 - precision: 0.9405 - recall: 0.9064 -
fl_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 adam adam.keras
0.3569 - accuracy: 0.9377 - precision: 0.9508 - recall: 0.9249 -
f1 score: 0.9377 - val loss: 0.5075 - val accuracy: 0.8875 -
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:
0.9436
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
624/624 [=============== ] - ETA: 0s - loss: 0.2715 -
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 -
f1 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
Epoch 22: val accuracy improved from 0.88752 to 0.89042, saving model
to models\modelo T com data augmentation adam adam.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 -
f1 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 adam adam.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: 0.8872 - val f1 score: 0.8933 -
lr: 4.0000e-06
Epoch 25/60
accuracy: 0.9665 - precision: 0.9732 - recall: 0.9605 - f1 score:
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 -
f1 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
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 adam adam.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:
```

```
0.9698
Epoch 28: val accuracy did not improve from 0.89243
0.1852 - accuracy: 0.9702 - precision: 0.9747 - recall: 0.9650 -
f1 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 adam adam.keras
624/624 [============== ] - 167s 267ms/step - loss:
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 -
fl_score: 0.9736 - val_loss: 0.4801 - val accuracy: 0.8919 -
```

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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:
0.9735
Epoch 33: val accuracy improved from 0.89253 to 0.89263, saving model
to models\modelo T com data augmentation adam adam.keras
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:
Epoch 34: val_accuracy did not improve from 0.89263
624/624 [============== ] - 166s 266ms/step - loss:
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 - f1_score:
Epoch 35: val accuracy improved from 0.89263 to 0.89343, saving model
to models\modelo T com data augmentation adam adam.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 adam adam.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
624/624 [=============== ] - ETA: 0s - loss: 0.1607 -
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 adam adam.keras
624/624 [============= ] - 170s 272ms/step - loss:
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 -
fl_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:
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
624/624 [============== ] - ETA: 0s - loss: 0.1570 -
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
624/624 [============= ] - ETA: 0s - loss: 0.1577 -
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
0.1542 - accuracy: 0.9758 - precision: 0.9797 - recall: 0.9717 -
fl_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
624/624 [=============== ] - ETA: 0s - loss: 0.1580 -
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
624/624 [============== ] - ETA: 0s - loss: 0.1557 -
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
accuracy: 0.9748 - precision: 0.9789 - recall: 0.9709 - f1 score:
0.9749
Epoch 54: val accuracy did not improve from 0.89413
0.1584 - accuracy: 0.9748 - precision: 0.9789 - recall: 0.9709 -
fl_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:
0.9750
Epoch 56: val accuracy improved from 0.89413 to 0.89433, saving model
```

```
to models\modelo T com data augmentation adam adam.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
624/624 [============== ] - ETA: 0s - loss: 0.1577 -
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:
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 adam.png')
plt.tight layout()
#plt.show()
```

