## Modelo S com data augmentation e optimizer RMSProp

```
from tensorflow.keras.metrics import Metric
from tensorflow.keras import backend as K
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
import os
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Activation, Dropout, Flatten,
Dense, Conv2D, MaxPooling2D, BatchNormalization
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping,
CSVLogger, ReduceLROnPlateau
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.preprocessing.image import ImageDataGenerator
os.environ['TF CPP MIN LOG LEVEL'] = '2'
# CONSTANTES
BATCH SIZE = 64
IMG SIZE = 32
NUM CLASSES = 10 # n^{\circ} classes para identificar
NUM EPOCHS = 100
LEARNING RATE = 0.001
DENSE LAYERS = [256, 512, 1024, 1024]
# Folders do dataset
train_dirs = ['./dataset/train/train1', './dataset/train/train2',
               './dataset/train/train3', './dataset/train/train5']
validation dir = './dataset/validation'
test dir = './dataset/test'
train datagen = ImageDataGenerator(
    rescale=1./255,
    rotation range=15,
    width_shift range=0.01,
    height shift range=0.01,
    horizontal flip=True,
    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 junstar os trainning generators
def combined generator(generators):
    while True:
        for generator in generators:
            yield next(generator)
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')
Found 10000 images belonging to 10 classes.
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):
        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()
```

```
model = Sequential([
    Conv2D(DENSE LAYERS[0], (3, 3), input shape=(IMG SIZE, IMG SIZE,
3)),
    BatchNormalization(),
    Activation('relu'),
    MaxPooling2D((2, 2)),
    Dropout (0.3),
    Conv2D(DENSE LAYERS[1], (3, 3)),
    BatchNormalization(),
    Activation('relu'),
    MaxPooling2D((2, 2)),
    Dropout (0.5),
    Conv2D(DENSE LAYERS[2], (3, 3)),
    BatchNormalization(),
    Activation('relu'),
    MaxPooling2D((2, 2)),
    Dropout (0.5),
    Flatten(),
    Dense(DENSE LAYERS[3]),
    BatchNormalization(),
    Activation('relu'),
    Dropout (0.5),
    Dense(NUM CLASSES, activation='softmax')
1)
# Compilar o modelo
model.compile(optimizer=RMSprop(learning rate=LEARNING RATE),
              loss='categorical crossentropy',
              metrics=['accuracy', Precision(), Recall(), F1Score()])
model.summary()
Model: "sequential"
                              Output Shape
Layer (type)
                                                         Param #
 conv2d (Conv2D)
                              (None, 30, 30, 256)
                                                         7168
 batch normalization (BatchN (None, 30, 30, 256)
                                                         1024
 ormalization)
 activation (Activation)
                              (None, 30, 30, 256)
                                                         0
max_pooling2d (MaxPooling2D (None, 15, 15, 256)
                                                         0
```

dropout (Dropout)	(None, 15, 15, 256)	0
conv2d_1 (Conv2D)	(None, 13, 13, 512)	1180160
<pre>batch_normalization_1 (Batc hNormalization)</pre>	(None, 13, 13, 512)	2048
<pre>activation_1 (Activation)</pre>	(None, 13, 13, 512)	0
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 6, 6, 512)	0
dropout_1 (Dropout)	(None, 6, 6, 512)	0
conv2d_2 (Conv2D)	(None, 4, 4, 1024)	4719616
<pre>batch_normalization_2 (Batc hNormalization)</pre>	(None, 4, 4, 1024)	4096
activation_2 (Activation)	(None, 4, 4, 1024)	0
<pre>max_pooling2d_2 (MaxPooling 2D)</pre>	(None, 2, 2, 1024)	0
dropout_2 (Dropout)	(None, 2, 2, 1024)	0
flatten (Flatten)	(None, 4096)	0
dense (Dense)	(None, 1024)	4195328
<pre>batch_normalization_3 (Batc hNormalization)</pre>	(None, 1024)	4096
activation_3 (Activation)	(None, 1024)	0
dropout_3 (Dropout)	(None, 1024)	0
dense_1 (Dense)	(None, 10)	10250

Total params: 10,123,786 Trainable params: 10,118,154 Non-trainable params: 5,632

## # Definir os Callbacks

# Para salvar o melhor modelo com base na acurácia de validação checkpoint = ModelCheckpoint(

```
"models/modelo S com data augmentation.keras",
monitor='val accuracy', verbose=1, save best only=True, mode='max')
# Parar o treinamento se não houver melhoria na loss após x epochs
early stopping = EarlyStopping(
   monitor='val loss', patience=10, restore best weights=True)
# Salvar para csv
csv logger = CSVLogger(
    f'logs/modelo S com data augmentation.csv', append=True)
# Reduzir a learning rate se não houver melhoria na loss após x epochs
(lembrar de deixar este valor sempre menor que a patience no
early stopping!!)
reduce lr = ReduceLROnPlateau(
   monitor='val_loss', factor=0.5, patience=3, verbose=1)
# calcular passos por epoch
steps per epoch = sum([gen.samples // BATCH SIZE for gen in
train generators])
# 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 generator.samples // BATCH SIZE,
    callbacks=[checkpoint, early stopping, csv logger, reduce lr]
)
# 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 1/100
624/624 [============== ] - ETA: 0s - loss: 1.6637 -
accuracy: 0.4272 - precision: 0.5460 - recall: 0.2850 - f1 score:
0.3745
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 0.51332, saving model to
models\modelo S com data augmentation.keras
1.6637 - accuracy: 0.4272 - precision: 0.5460 - recall: 0.2850 -
f1 score: 0.3745 - val loss: 1.3844 - val accuracy: 0.5133 -
val_precision: 0.7241 - val_recall: 0.3057 - val_f1_score: 0.4299 -
lr: 0.0010
Epoch 2/100
accuracy: 0.5618 - precision: 0.6893 - recall: 0.4292 - f1 score:
0.5290
Epoch 2: val accuracy improved from 0.51332 to 0.58664, saving model
to models\modelo S com data augmentation.keras
1.2339 - accuracy: 0.5618 - precision: 0.6893 - recall: 0.4292 -
f1 score: 0.5290 - val loss: 1.1930 - val accuracy: 0.5866 -
val precision: 0.6948 - val recall: 0.4849 - val f1 score: 0.5712 -
lr: 0.0010
Epoch 3/100
accuracy: 0.6115 - precision: 0.7350 - recall: 0.4941 - f1 score:
0.5909
Epoch 3: val accuracy did not improve from 0.58664
1.0889 - accuracy: 0.6115 - precision: 0.7350 - recall: 0.4941 -
f1 score: 0.5909 - val loss: 1.6433 - val accuracy: 0.4916 -
val precision: 0.5664 - val recall: 0.4251 - val f1 score: 0.4857 -
lr: 0.0010
Epoch 4/100
accuracy: 0.6451 - precision: 0.7558 - recall: 0.5382 - f1 score:
0.6287
Epoch 4: val accuracy did not improve from 0.58664
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1.0073 - accuracy: 0.6451 - precision: 0.7560 - recall: 0.5381 -
f1 score: 0.6287 - val loss: 1.2640 - val accuracy: 0.5733 -
val precision: 0.6641 - val recall: 0.4999 - val f1 score: 0.5704 -
lr: 0.0010
Epoch 5/100
accuracy: 0.6707 - precision: 0.7703 - recall: 0.5698 - f1 score:
0.6551
Epoch 5: val accuracy improved from 0.58664 to 0.68830, saving model
to models\modelo S com data augmentation.keras
624/624 [============= ] - 33s 53ms/step - loss:
0.9421 - accuracy: 0.6707 - precision: 0.7703 - recall: 0.5698 -
fl_score: 0.6551 - val_loss: 0.9065 - val_accuracy: 0.6883 -
val precision: 0.8154 - val recall: 0.5613 - val f1 score: 0.6649 -
lr: 0.0010
Epoch 6/100
accuracy: 0.6825 - precision: 0.7783 - recall: 0.5902 - f1_score:
Epoch 6: val accuracy did not improve from 0.68830
0.9013 - accuracy: 0.6825 - precision: 0.7783 - recall: 0.5902 -
f1 score: 0.6713 - val loss: 0.9239 - val accuracy: 0.6822 -
val precision: 0.7608 - val recall: 0.6066 - val f1 score: 0.6750 -
lr: 0.0010
Epoch 7/100
accuracy: 0.7023 - precision: 0.7922 - recall: 0.6177 - f1 score:
0.6941
Epoch 7: val accuracy improved from 0.68830 to 0.70473, saving model
to models\modelo S com data augmentation.keras
0.8488 - accuracy: 0.7023 - precision: 0.7922 - recall: 0.6177 -
f1 score: 0.6941 - val loss: 0.8522 - val accuracy: 0.7047 -
val precision: 0.7846 - val recall: 0.6320 - val f1 score: 0.7001 -
lr: 0.0010
Epoch 8/100
accuracy: 0.7180 - precision: 0.8038 - recall: 0.6343 - f1_score:
0.7091
Epoch 8: val accuracy did not improve from 0.70473
0.8124 - accuracy: 0.7181 - precision: 0.8039 - recall: 0.6344 -
f1 score: 0.7091 - val loss: 1.0332 - val accuracy: 0.6625 -
val_precision: 0.7245 - val_recall: 0.6025 - val_f1_score: 0.6579 -
lr: 0.0010
Epoch 9/100
accuracy: 0.7259 - precision: 0.8093 - recall: 0.6476 - f1 score:
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0.7195
Epoch 9: val accuracy did not improve from 0.70473
0.7815 - accuracy: 0.7259 - precision: 0.8093 - recall: 0.6476 -
f1 score: 0.7195 - val loss: 1.0233 - val accuracy: 0.6643 -
val_precision: 0.7378 - val_recall: 0.5998 - val_f1_score: 0.6617 -
lr: 0.0010
Epoch 10/100
accuracy: 0.7350 - precision: 0.8153 - recall: 0.6592 - f1 score:
0.7290
Epoch 10: val accuracy improved from 0.70473 to 0.70873, saving model
to models\modelo S com data augmentation.keras
0.7596 - accuracy: 0.7351 - precision: 0.8153 - recall: 0.6593 -
f1 score: 0.7291 - val loss: 0.8323 - val accuracy: 0.7087 -
val precision: 0.7951 - val recall: 0.6423 - val f1 score: 0.7106 -
lr: 0.0010
Epoch 11/100
accuracy: 0.7440 - precision: 0.8179 - recall: 0.6731 - f1 score:
0.7385
Epoch 11: val accuracy did not improve from 0.70873
0.7322 - accuracy: 0.7437 - precision: 0.8177 - recall: 0.6730 -
f1 score: 0.7383 - val loss: 0.9851 - val accuracy: 0.6774 -
val_precision: 0.7392 - val_recall: 0.6284 - val_f1_score: 0.6793 -
lr: 0.0010
Epoch 12/100
accuracy: 0.7523 - precision: 0.8231 - recall: 0.6834 - f1 score:
Epoch 12: val_accuracy did not improve from 0.70873
0.7132 - accuracy: 0.7524 - precision: 0.8231 - recall: 0.6835 -
f1 score: 0.7469 - val loss: 1.0373 - val accuracy: 0.6635 -
val precision: 0.7242 - val recall: 0.6150 - val f1 score: 0.6651 -
lr: 0.0010
Epoch 13/100
accuracy: 0.7599 - precision: 0.8302 - recall: 0.6942 - f1_score:
0.7561
Epoch 13: val_accuracy did not improve from 0.70873
Epoch 13: ReduceLROnPlateau reducing learning rate to
0.0005000000237487257.
624/624 [============= ] - 33s 52ms/step - loss:
0.6828 - accuracy: 0.7599 - precision: 0.8302 - recall: 0.6942 -
fl_score: 0.7561 - val_loss: 1.0965 - val_accuracy: 0.6664 -
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val precision: 0.7220 - val recall: 0.6169 - val f1 score: 0.6653 -
lr: 0.0010
Epoch 14/100
accuracy: 0.7854 - precision: 0.8491 - recall: 0.7262 - f1 score:
0.7829
Epoch 14: val accuracy improved from 0.70873 to 0.75220, saving model
to models\modelo S com data augmentation.keras
624/624 [============= ] - 33s 52ms/step - loss:
0.6160 - accuracy: 0.7854 - precision: 0.8492 - recall: 0.7263 -
f1 score: 0.7830 - val loss: 0.7720 - val accuracy: 0.7522 -
val_precision: 0.8182 - val_recall: 0.7037 - val_f1_score: 0.7567 -
lr: 5.0000e-04
Epoch 15/100
accuracy: 0.7926 - precision: 0.8514 - recall: 0.7357 - f1 score:
0.7893
Epoch 15: val_accuracy did not improve from 0.75220
0.5974 - accuracy: 0.7924 - precision: 0.8513 - recall: 0.7356 -
f1 score: 0.7892 - val loss: 0.7612 - val accuracy: 0.7490 -
val precision: 0.7991 - val recall: 0.7057 - val f1 score: 0.7495 -
lr: 5.0000e-04
Epoch 16/100
accuracy: 0.7931 - precision: 0.8525 - recall: 0.7394 - f1 score:
Epoch 16: val accuracy improved from 0.75220 to 0.78335, saving model
to models\modelo S com data augmentation.keras
0.5868 - accuracy: 0.7932 - precision: 0.8526 - recall: 0.7395 -
fl_score: 0.7920 - val_loss: 0.6476 - val_accuracy: 0.7834 -
val precision: 0.8406 - val recall: 0.7339 - val f1 score: 0.7836 -
lr: 5.0000e-04
Epoch 17/100
accuracy: 0.8002 - precision: 0.8562 - recall: 0.7487 - f1 score:
0.7989
Epoch 17: val accuracy improved from 0.78335 to 0.78566, saving model
to models\modelo_S_com_data_augmentation.keras
0.5644 - accuracy: 0.8002 - precision: 0.8562 - recall: 0.7487 -
f1 score: 0.7989 - val loss: 0.6252 - val_accuracy: 0.7857 -
val precision: 0.8405 - val recall: 0.7421 - val f1 score: 0.7882 -
lr: 5.0000e-04
Epoch 18/100
accuracy: 0.8050 - precision: 0.8607 - recall: 0.7552 - f1 score:
0.8045
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Epoch 18: val accuracy did not improve from 0.78566
0.5551 - accuracy: 0.8050 - precision: 0.8607 - recall: 0.7552 -
f1 score: 0.8045 - val loss: 0.8144 - val accuracy: 0.7459 -
val precision: 0.7968 - val recall: 0.7088 - val f1 score: 0.7502 -
lr: 5.0000e-04
Epoch 19/100
accuracy: 0.8074 - precision: 0.8620 - recall: 0.7607 - f1_score:
0.8082
Epoch 19: val accuracy did not improve from 0.78566
0.5477 - accuracy: 0.8075 - precision: 0.8620 - recall: 0.7607 -
f1 score: 0.8082 - val loss: 0.8062 - val accuracy: 0.7353 -
val_precision: 0.7881 - val_recall: 0.6938 - val_f1_score: 0.7379 -
lr: 5.0000e-04
Epoch 20/100
accuracy: 0.8137 - precision: 0.8636 - recall: 0.7656 - f1 score:
0.8117
Epoch 20: val accuracy did not improve from 0.78566
Epoch 20: ReduceLROnPlateau reducing learning rate to
0.0002500000118743628.
624/624 [============= ] - 32s 52ms/step - loss:
0.5324 - accuracy: 0.8137 - precision: 0.8636 - recall: 0.7656 -
f1 score: 0.8117 - val loss: 0.6503 - val accuracy: 0.7834 -
val precision: 0.8299 - val recall: 0.7440 - val f1 score: 0.7846 -
lr: 5.0000e-04
Epoch 21/100
accuracy: 0.8213 - precision: 0.8692 - recall: 0.7770 - f1 score:
0.8205
Epoch 21: val accuracy improved from 0.78566 to 0.81941, saving model
to models\modelo S com data augmentation.keras
624/624 [============ ] - 34s 54ms/step - loss:
0.5101 - accuracy: 0.8213 - precision: 0.8692 - recall: 0.7770 -
f1 score: 0.8205 - val_loss: 0.5298 - val_accuracy: 0.8194 -
val precision: 0.8635 - val recall: 0.7851 - val f1 score: 0.8224 -
lr: 2.5000e-04
Epoch 22/100
accuracy: 0.8275 - precision: 0.8746 - recall: 0.7838 - f1 score:
0.8267
Epoch 22: val accuracy did not improve from 0.81941
0.4904 - accuracy: 0.8275 - precision: 0.8746 - recall: 0.7838 -
f1 score: 0.8267 - val loss: 0.5830 - val accuracy: 0.8066 -
val_precision: 0.8527 - val_recall: 0.7686 - val_f1_score: 0.8085 -
```

```
lr: 2.5000e-04
Epoch 23/100
accuracy: 0.8288 - precision: 0.8758 - recall: 0.7859 - f1 score:
0.8284
Epoch 23: val accuracy did not improve from 0.81941
0.4848 - accuracy: 0.8288 - precision: 0.8758 - recall: 0.7859 -
f1 score: 0.8284 - val loss: 0.6124 - val accuracy: 0.7964 -
val precision: 0.8386 - val recall: 0.7634 - val f1 score: 0.7992 -
lr: 2.5000e-04
Epoch 24/100
accuracy: 0.8348 - precision: 0.8783 - recall: 0.7940 - f1 score:
0.8340
Epoch 24: val accuracy did not improve from 0.81941
Epoch 24: ReduceLROnPlateau reducing learning rate to
0.0001250000059371814.
0.4737 - accuracy: 0.8348 - precision: 0.8783 - recall: 0.7940 -
f1 score: 0.8340 - val loss: 0.5317 - val accuracy: 0.8186 -
val precision: 0.8635 - val recall: 0.7862 - val f1 score: 0.8230 -
lr: 2.5000e-04
Epoch 25/100
accuracy: 0.8390 - precision: 0.8811 - recall: 0.7992 - f1 score:
0.8381
Epoch 25: val accuracy did not improve from 0.81941
0.4588 - accuracy: 0.8391 - precision: 0.8812 - recall: 0.7993 -
f1 score: 0.8382 - val loss: 0.5647 - val accuracy: 0.8150 -
val precision: 0.8550 - val recall: 0.7817 - val f1 score: 0.8167 -
lr: 1.2500e-04
Epoch 26/100
accuracy: 0.8404 - precision: 0.8841 - recall: 0.8028 - f1_score:
0.8415
Epoch 26: val accuracy improved from 0.81941 to 0.83894, saving model
to models\modelo_S_com_data_augmentation.keras
0.4485 - accuracy: 0.8404 - precision: 0.8841 - recall: 0.8028 -
f1 score: 0.8415 - val loss: 0.4765 - val accuracy: 0.8389 -
val_precision: 0.8798 - val_recall: 0.8079 - val f1 score: 0.8423 -
lr: 1.2500e-04
Epoch 27/100
accuracy: 0.8425 - precision: 0.8850 - recall: 0.8033 - f1 score:
0.8422
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Epoch 27: val accuracy did not improve from 0.83894
0.4502 - accuracy: 0.8425 - precision: 0.8849 - recall: 0.8032 -
f1 score: 0.8421 - val loss: 0.5164 - val accuracy: 0.8256 -
val precision: 0.8675 - val recall: 0.7941 - val f1 score: 0.8292 -
lr: 1.2500e-04
Epoch 28/100
624/624 [============== ] - ETA: 0s - loss: 0.4444 -
accuracy: 0.8439 - precision: 0.8860 - recall: 0.8064 - f1 score:
0.8444
Epoch 28: val accuracy improved from 0.83894 to 0.84095, saving model
to models\modelo_S_com_data_augmentation.keras
0.4444 - accuracy: 0.8439 - precision: 0.8860 - recall: 0.8064 -
fl_score: 0.8444 - val_loss: 0.4683 - val_accuracy: 0.8409 -
val precision: 0.8782 - val recall: 0.8080 - val f1 score: 0.8416 -
lr: 1.2500e-04
Epoch 29/100
accuracy: 0.8468 - precision: 0.8871 - recall: 0.8101 - f1 score:
0.8468
Epoch 29: val accuracy did not improve from 0.84095
0.4370 - accuracy: 0.8468 - precision: 0.8870 - recall: 0.8101 -
f1 score: 0.8468 - val loss: 0.5098 - val accuracy: 0.8290 -
val precision: 0.8667 - val recall: 0.8007 - val f1 score: 0.8324 -
lr: 1.2500e-04
Epoch 30/100
accuracy: 0.8485 - precision: 0.8885 - recall: 0.8136 - f1 score:
0.8494
Epoch 30: val_accuracy did not improve from 0.84095
0.4317 - accuracy: 0.8485 - precision: 0.8885 - recall: 0.8136 -
f1 score: 0.8494 - val loss: 0.4960 - val accuracy: 0.8333 -
val precision: 0.8694 - val recall: 0.8052 - val f1 score: 0.8360 -
lr: 1.2500e-04
Epoch 31/100
accuracy: 0.8513 - precision: 0.8891 - recall: 0.8157 - f1 score:
0.8508
Epoch 31: val accuracy did not improve from 0.84095
Epoch 31: ReduceLROnPlateau reducing learning rate to
6.25000029685907e-05.
0.4264 - accuracy: 0.8513 - precision: 0.8891 - recall: 0.8157 -
f1 score: 0.8508 - val loss: 0.5034 - val accuracy: 0.8332 -
val_precision: 0.8702 - val_recall: 0.8038 - val_f1_score: 0.8357 -
```

```
lr: 1.2500e-04
Epoch 32/100
accuracy: 0.8514 - precision: 0.8903 - recall: 0.8158 - f1 score:
0.8514
Epoch 32: val accuracy did not improve from 0.84095
0.4222 - accuracy: 0.8515 - precision: 0.8904 - recall: 0.8158 -
f1 score: 0.8515 - val loss: 0.4868 - val accuracy: 0.8371 -
val precision: 0.8739 - val recall: 0.8071 - val f1 score: 0.8392 -
lr: 6.2500e-05
Epoch 33/100
accuracy: 0.8561 - precision: 0.8947 - recall: 0.8217 - f1 score:
0.8566
Epoch 33: val accuracy did not improve from 0.84095
0.4143 - accuracy: 0.8561 - precision: 0.8947 - recall: 0.8217 -
f1 score: 0.8566 - val loss: 0.4835 - val accuracy: 0.8397 -
val_precision: 0.8741 - val_recall: 0.8100 - val_f1 score: 0.8408 - val_f1
lr: 6.2500e-05
Epoch 34/100
accuracy: 0.8563 - precision: 0.8935 - recall: 0.8212 - f1 score:
0.8558
Epoch 34: val accuracy did not improve from 0.84095
Epoch 34: ReduceLROnPlateau reducing learning rate to
3.125000148429535e-05.
624/624 [============ ] - 33s 52ms/step - loss:
0.4125 - accuracy: 0.8561 - precision: 0.8934 - recall: 0.8210 -
f1 score: 0.8557 - val loss: 0.4861 - val accuracy: 0.8378 -
val precision: 0.8720 - val recall: 0.8075 - val f1 score: 0.8385 -
lr: 6.2500e-05
Epoch 35/100
accuracy: 0.8580 - precision: 0.8927 - recall: 0.8232 - f1_score:
0.8566
Epoch 35: val accuracy improved from 0.84095 to 0.84565, saving model
to models\modelo_S_com_data_augmentation.keras
0.4080 - accuracy: 0.8580 - precision: 0.8927 - recall: 0.8232 -
f1 score: 0.8566 - val loss: 0.4550 - val accuracy: 0.8457 -
val precision: 0.8807 - val recall: 0.8161 - val f1 score: 0.8472 -
lr: 3.1250e-05
Epoch 36/100
accuracy: 0.8566 - precision: 0.8932 - recall: 0.8243 - f1 score:
0.8574
```

```
Epoch 36: val accuracy did not improve from 0.84565
0.4093 - accuracy: 0.8566 - precision: 0.8932 - recall: 0.8243 -
f1 score: 0.8574 - val loss: 0.4811 - val accuracy: 0.8397 -
val precision: 0.8737 - val recall: 0.8095 - val f1 score: 0.8404 -
lr: 3.1250e-05
Epoch 37/100
accuracy: 0.8562 - precision: 0.8931 - recall: 0.8220 - f1_score:
0.8561
Epoch 37: val accuracy did not improve from 0.84565
0.4075 - accuracy: 0.8562 - precision: 0.8931 - recall: 0.8220 -
f1 score: 0.8561 - val loss: 0.4680 - val accuracy: 0.8415 -
val_precision: 0.8764 - val_recall: 0.8128 - val_f1_score: 0.8434 -
lr: 3.1250e-05
Epoch 38/100
accuracy: 0.8569 - precision: 0.8941 - recall: 0.8252 - f1 score:
0.8582
Epoch 38: val accuracy did not improve from 0.84565
Epoch 38: ReduceLROnPlateau reducing learning rate to
1.5625000742147677e-05.
624/624 [============= ] - 32s 52ms/step - loss:
0.4074 - accuracy: 0.8568 - precision: 0.8941 - recall: 0.8251 -
f1 score: 0.8582 - val loss: 0.4960 - val accuracy: 0.8347 -
val precision: 0.8707 - val recall: 0.8051 - val f1 score: 0.8366 -
lr: 3.1250e-05
Epoch 39/100
accuracy: 0.8563 - precision: 0.8940 - recall: 0.8242 - f1 score:
0.8577
Epoch 39: val_accuracy did not improve from 0.84565
0.4023 - accuracy: 0.8563 - precision: 0.8940 - recall: 0.8242 -
fl_score: 0.8577 - val_loss: 0.4667 - val_accuracy: 0.8419 -
val precision: 0.8771 - val recall: 0.8125 - val_f1_score: 0.8436 -
lr: 1.5625e-05
Epoch 40/100
accuracy: 0.8591 - precision: 0.8949 - recall: 0.8250 - f1_score:
0.8585
Epoch 40: val accuracy did not improve from 0.84565
0.4052 - accuracy: 0.8592 - precision: 0.8949 - recall: 0.8250 -
f1 score: 0.8586 - val loss: 0.4813 - val accuracy: 0.8407 -
val precision: 0.8721 - val recall: 0.8095 - val f1 score: 0.8396 -
lr: 1.5625e-05
```

```
Epoch 41/100
accuracy: 0.8601 - precision: 0.8952 - recall: 0.8250 - f1 score:
Epoch 41: val accuracy did not improve from 0.84565
Epoch 41: ReduceLROnPlateau reducing learning rate to
7.812500371073838e-06.
624/624 [============== ] - 32s 52ms/step - loss:
0.3980 - accuracy: 0.8601 - precision: 0.8952 - recall: 0.8250 -
f1 score: 0.8586 - val loss: 0.4651 - val accuracy: 0.8431 -
val precision: 0.8772 - val recall: 0.8130 - val f1 score: 0.8439 -
lr: 1.5625e-05
Epoch 42/100
accuracy: 0.8565 - precision: 0.8932 - recall: 0.8237 - f1_score:
Epoch 42: val accuracy did not improve from 0.84565
624/624 [============= ] - 32s 52ms/step - loss:
0.4066 - accuracy: 0.8565 - precision: 0.8932 - recall: 0.8238 -
fl_score: 0.8571 - val_loss: 0.4732 - val_accuracy: 0.8419 -
val_precision: 0.8754 - val_recall: 0.8125 - val_f1_score: 0.8428 -
lr: 7.8125e-06
Epoch 43/100
accuracy: 0.8567 - precision: 0.8915 - recall: 0.8253 - f1 score:
Epoch 43: val accuracy did not improve from 0.84565
0.4079 - accuracy: 0.8567 - precision: 0.8915 - recall: 0.8253 -
f1 score: 0.8571 - val loss: 0.4705 - val accuracy: 0.8424 -
val precision: 0.8752 - val recall: 0.8127 - val f1 score: 0.8428 -
lr: 7.8125e-06
Epoch 44/100
accuracy: 0.8574 - precision: 0.8949 - recall: 0.8246 - f1 score:
Epoch 44: val accuracy did not improve from 0.84565
Epoch 44: ReduceLROnPlateau reducing learning rate to
3.906250185536919e-06.
0.4046 - accuracy: 0.8573 - precision: 0.8948 - recall: 0.8246 -
f1 score: 0.8583 - val loss: 0.4751 - val accuracy: 0.8412 -
val precision: 0.8749 - val recall: 0.8115 - val f1 score: 0.8420 -
lr: 7.8125e-06
Epoch 45/100
accuracy: 0.8594 - precision: 0.8966 - recall: 0.8275 - f1 score:
```

```
0.8606
Epoch 45: val accuracy did not improve from 0.84565
0.3970 - accuracy: 0.8594 - precision: 0.8966 - recall: 0.8275 -
f1 score: 0.8607 - val loss: 0.4668 - val accuracy: 0.8434 -
val precision: 0.8766 - val recall: 0.8140 - val f1 score: 0.8441 -
lr: 3.9063e-06
- accuracy: 0.8505 - precision: 0.8816 - recall: 0.8231 - f1 score:
0.8514
Test Loss: 0.4495638310909271
Test Accuracy: 0.8504999876022339
Test Precision: 0.8816409707069397
Test Recall: 0.8230999708175659
Test F1 Score: 0.851365327835083
# Plots do treino
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 S com data augmentation rmsprop.png')
plt.tight layout()
```

