Modelo S com Data Augmentation

```
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 Adam
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'
```

Data Augmentation

Testámos vários tipos de data augmentation, com mais e menos intensidade. Concluímos que os valores escolhidos (rotações de 15 graus, deslocamentos horizontais e verticais de 0,01 e horizontal_flip) proporcionavam os melhores resultados, mesmo que essa melhora seja tão pouco significativa que nos deixe a pensar se efetivamente compensa de todo este tempo extra de treino.

Chegamos à conclusão que dado que as imagens do dataset têm baixa resolução, adicionar demasiada data augmentation prejudicava os resultados. Isto ocorre porque transformações excessivas em imagens de baixa qualidade podem distorcer os dados de forma significativa, dificultando a capacidade do modelo de aprender características relevantes. Portanto, um equilíbrio cuidadoso foi necessário para melhorar a robustez do modelo sem comprometer a integridade dos dados visuais.

```
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.
```

Teste com outras intensidades de data augmentation:

train_datagen = ImageDataGenerator(rescale=1./255, rotation_range=15, width_shift_range=0.01, height_shift_range=0.01, shear_range=0.01, zoom_range=0.01, horizontal_flip=True, fill_mode='nearest')

Validation Accuracy Obtida: 0.8462539911270142

train_datagen = ImageDataGenerator(rescale=1./255, rotation_range=5, height_shift_range=0.01, horizontal_flip=True, fill_mode='nearest')

Validation Accuracy Obtida: 0.8409455418586731

train_datagen = ImageDataGenerator(rescale=1./255, rotation_range=2, width_shift_range=0.01, horizontal_flip=True, fill_mode='nearest')

Validation Accuracy Obtida: 0.8501634340286255

```
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')
])
# Compilar o modelo
model.compile(optimizer=Adam(learning rate=LEARNING RATE),
              loss='categorical crossentropy',
              metrics=['accuracy', Precision(), Recall(), F1Score()])
model.summary()
Model: "sequential"
Layer (type)
                              Output Shape
                                                        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
                              (None, 15, 15, 256)
 dropout (Dropout)
                                                        0
 conv2d 1 (Conv2D)
                              (None, 13, 13, 512)
                                                        1180160
 batch normalization 1 (Batc
                               (None, 13, 13, 512)
                                                        2048
 hNormalization)
 activation 1 (Activation)
                              (None, 13, 13, 512)
                                                        0
max pooling2d 1 (MaxPooling (None, 6, 6, 512)
                                                        0
 2D)
```

(None, 6, 6, 512)

0

dropout 1 (Dropout)

```
conv2d 2 (Conv2D)
                             (None, 4, 4, 1024)
                                                        4719616
 batch normalization 2 (Batc (None, 4, 4, 1024)
                                                        4096
 hNormalization)
 activation 2 (Activation)
                             (None, 4, 4, 1024)
                                                        0
max pooling2d 2 (MaxPooling (None, 2, 2, 1024)
                                                        0
 2D)
 dropout 2 (Dropout)
                             (None, 2, 2, 1024)
                                                        0
 flatten (Flatten)
                             (None, 4096)
                                                        0
 dense (Dense)
                             (None, 1024)
                                                        4195328
 batch normalization 3 (Batc (None, 1024)
                                                        4096
 hNormalization)
 activation_3 (Activation)
                                                        0
                             (None, 1024)
dropout 3 (Dropout)
                             (None, 1024)
 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 adam.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 adam.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
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 adam.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 adam.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
1.0073 - accuracy: 0.6451 - precision: 0.7560 - recall: 0.5381 -
fl_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 adam.keras
0.9421 - accuracy: 0.6707 - precision: 0.7703 - recall: 0.5698 -
f1 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
624/624 [============= ] - 33s 53ms/step - loss:
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:
Epoch 7: val accuracy improved from 0.68830 to 0.70473, saving model
to models\modelo S com data augmentation adam.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:
0.7195
Epoch 9: val accuracy did not improve from 0.70473
624/624 [============= ] - 32s 52ms/step - loss:
0.7815 - accuracy: 0.7259 - precision: 0.8093 - recall: 0.6476 -
fl_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:
Epoch 10: val accuracy improved from 0.70473 to 0.70873, saving model
to models\modelo S com data augmentation adam.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:
0.7468
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.
0.6828 - accuracy: 0.7599 - precision: 0.8302 - recall: 0.6942 -
f1 score: 0.7561 - val loss: 1.0965 - val accuracy: 0.6664 -
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 adam.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:
0.7919
Epoch 16: val accuracy improved from 0.75220 to 0.78335, saving model
to models\modelo_S_com_data_augmentation_adam.keras
624/624 [============= ] - 33s 52ms/step - loss:
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 adam.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:
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 -
fl_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 adam.keras
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
624/624 [============= ] - 33s 53ms/step - loss:
0.4904 - accuracy: 0.8275 - precision: 0.8746 - recall: 0.7838 -
fl_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
624/624 [============= ] - 33s 53ms/step - loss:
0.4848 - accuracy: 0.8288 - precision: 0.8758 - recall: 0.7859 -
fl_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.
```

```
624/624 [============ ] - 33s 53ms/step - loss:
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:
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:
Epoch 26: val accuracy improved from 0.81941 to 0.83894, saving model
to models\modelo S com data augmentation adam.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 - val_recall
lr: 1.2500e-04
Epoch 27/100
accuracy: 0.8425 - precision: 0.8850 - recall: 0.8033 - f1_score:
0.8422
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
accuracy: 0.8439 - precision: 0.8860 - recall: 0.8064 - f1_score:
Epoch 28: val_accuracy improved from 0.83894 to 0.84095, saving model
to models\modelo S com data augmentation adam.keras
624/624 [============= ] - 33s 53ms/step - loss:
0.4444 - accuracy: 0.8439 - precision: 0.8860 - recall: 0.8064 -
f1 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 -
fl_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 -
lr: 6.2500e-05
```

```
Epoch 34/100
accuracy: 0.8563 - precision: 0.8935 - recall: 0.8212 - f1 score:
Epoch 34: val accuracy did not improve from 0.84095
Epoch 34: ReduceLROnPlateau reducing learning rate to
3.125000148429535e-05.
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:
Epoch 35: val accuracy improved from 0.84095 to 0.84565, saving model
to models\modelo S com data augmentation adam.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 -
f1_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:
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.
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:
0.8571
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:
0.8583
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
```

Podemos ver acima que existe melhora mesmo que pouco significativa quando comparado à versão sem data augmentation.

```
# 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 adam.png')
plt.tight layout()
```

