# TRABALHO DE INTELIGÊNCIA ARTIFICIAL

### **DEEP LEARNING**

## Rock-Paper-Scissors Images DATASET (link abaixo)

https://www.kaggle.com/datasets/drgfreeman/rockpaperscissors

#### **GUILHERME ZAGO CANESIN**

https://github.com/GuilhermeZCanesin/DeepLearning-2022.1.git

```
import keras
import tensorflow as tf
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D, AveragePooling2D
import numpy as np
import pandas as pd
from sklearn.metrics import classification_report, confusion_matrix
import matplotlib.pyplot as plt
import seaborn as sn
from tensorflow.keras.preprocessing.image import load_img, img_to_array, array_to_img
from numba import cuda
from keras.optimizers import Adam
```

# 2. DEFINIÇÃO DO MODELO PRÉ TREINADO

4. TAMBÉM CONGELANDO PARTE DO MODELO PARA TREINO DA ÚLTIMA CAMADA

MODELO:

INCEPTION V3

```
In [22]: base_model = tf.keras.applications.InceptionV3(weights='imagenet', include_top=False, input_shape=(200, 300, 3))
for layer in base_model.layers:
    layer.trainable = False
```

# 10. DEFINIÇÃO DE VARIÁVEIS DE MODELO

```
In [16]: batch_size = 32
# CLASSES (PEDRA - PAPEL - TESOURA)
num_classes = 3
dropout = 0.5
```

```
epochs = 500
lr = 0.001
```

# 5. ATRIBUINDO A FUNÇÃO DE ATIVAÇÃO DO MODELO

**FUNÇÃO** 

SOFTMAX

```
In [23]: x = Flatten()(base_model.output)
x = Dense(1024, activation='relu')(x)
x = Dropout(dropout)(x)
x = Dense(num_classes, activation='softmax')(x)
```

#### CRIANDO O OBJETO DO MODELO

```
In [24]: model = tf.keras.models.Model(base_model.input, x)
model.compile(Adam(learning_rate=lr),loss='categorical_crossentropy',metrics=['accuracy'])
```

#### 9. CRIANDO OS CONJUNTOS DE TESTE E TREINO

#### 1. BAIXANDO O CONJUNTO DE DADOS

```
In [28]: # INICIALIZANDO O OBJETO QUE RECUPERA AS AMOSTRAS DE TREINO COM A FUNÇÃO INCEPTIONV3
         train_data_gen = tf.keras.preprocessing.image.ImageDataGenerator(validation_split=0.2, preprocessing_function=tf.keras.applications.inception_v3.preprocess_input)
         # DEFININDO O CAMINHO DAS IMAGENS (PASTA INPUT), OS NOMES DAS PASTAS DEVEM TER OS MESMOS NOMES APRESENTADOS EM "classes" ABAIXO
         train_generator = train_data_gen.flow_from_directory('input/',
                                                           target size=(200, 300),
                                                           batch size=batch size,
                                                           class_mode="categorical",
                                                           shuffle=False,
                                                           classes=['paper', 'rock', 'scissors'],
                                                           subset='training')
         # DEFININDO O CAMINHO DAS IMAGENS (PASTA INPUT), OS NOMES DAS PASTAS DEVEM TER OS MESMOS NOMES APRESENTADOS EM "classes" ABAIXO
         test_generator = train_data_gen.flow_from_directory('input/',
                                                           target_size=(200, 300),
                                                           batch_size=16,
                                                           class_mode="categorical",
                                                           shuffle=False,
                                                           classes=['paper', 'rock', 'scissors'],
                                                           subset='validation')
```

Found 1751 images belonging to 3 classes. Found 437 images belonging to 3 classes.

## **EXECUTANDO O TREINO DO MODELO**

6. OTIMIZADOR: ADAM

## 7. MÉTRICA DE AVALIAÇÃO: ACURÁCIA

### 8. MÉTRICA DE ERRO: CATEGORICAL CROSS ENTROPY

```
Epoch 1/500
Epoch 2/500
Epoch 3/500
Epoch 4/500
Epoch 5/500
Epoch 6/500
54/54 [============== ] - 249s 5s/step - loss: 0.0270 - accuracy: 0.9913 - val loss: 0.0450 - val accuracy: 0.9884
Epoch 7/500
Epoch 9/500
Epoch 10/500
Epoch 11/500
54/54 [============= - 283s 5s/step - loss: 0.0091 - accuracy: 0.9965 - val_loss: 0.0299 - val_accuracy: 0.9884
Epoch 12/500
Epoch 13/500
Epoch 14/500
Epoch 15/500
54/54 [============= - 230s 4s/step - loss: 0.0087 - accuracy: 0.9959 - val_loss: 0.0194 - val_accuracy: 0.9931
Epoch 16/500
54/54 [============== ] - 220s 4s/step - loss: 0.0028 - accuracy: 0.9994 - val loss: 0.0369 - val accuracy: 0.9907
Epoch 17/500
Epoch 18/500
Epoch 19/500
Epoch 20/500
Epoch 21/500
54/54 [============= - 219s 4s/step - loss: 0.0085 - accuracy: 0.9983 - val_loss: 0.0255 - val_accuracy: 0.9931
Epoch 22/500
Epoch 23/500
Epoch 24/500
54/54 [============== - 222s 4s/step - loss: 0.0117 - accuracy: 0.9959 - val loss: 0.0386 - val accuracy: 0.9931
Epoch 25/500
54/54 [============= - 221s 4s/step - loss: 0.0094 - accuracy: 0.9959 - val_loss: 0.0297 - val_accuracy: 0.9931
Epoch 26/500
Epoch 28/500
```

```
Epoch 29/500
54/54 [============== ] - 220s 4s/step - loss: 0.0056 - accuracy: 0.9959 - val loss: 0.0238 - val accuracy: 0.9977
Epoch 30/500
54/54 [============= - 222s 4s/step - loss: 0.0067 - accuracy: 0.9971 - val_loss: 0.0331 - val_accuracy: 0.9954
Epoch 31/500
Epoch 32/500
Epoch 33/500
Epoch 34/500
54/54 [============= - 222s 4s/step - loss: 0.0097 - accuracy: 0.9942 - val_loss: 0.0276 - val_accuracy: 0.9954
Epoch 35/500
54/54 [============== ] - 225s 4s/step - loss: 0.0061 - accuracy: 0.9983 - val loss: 0.0156 - val accuracy: 0.9977
Epoch 36/500
Epoch 37/500
Epoch 38/500
Epoch 39/500
Epoch 40/500
Epoch 41/500
Epoch 42/500
Epoch 43/500
Epoch 44/500
Epoch 45/500
54/54 [============== ] - 219s 4s/step - loss: 0.0590 - accuracy: 0.9820 - val loss: 0.0049 - val accuracy: 1.0000
Epoch 46/500
54/54 [============ - 219s 4s/step - loss: 0.0434 - accuracy: 0.9884 - val_loss: 0.0070 - val_accuracy: 0.9954
Epoch 47/500
Epoch 48/500
Epoch 49/500
Epoch 50/500
54/54 [============== ] - 220s 4s/step - loss: 0.0774 - accuracy: 0.9740 - val loss: 0.2646 - val accuracy: 0.9583
Epoch 51/500
54/54 [============= ] - 221s 4s/step - loss: 0.0618 - accuracy: 0.9773 - val loss: 0.0124 - val accuracy: 0.9954
Epoch 52/500
54/54 [============ ] - 223s 4s/step - loss: 0.0329 - accuracy: 0.9860 - val_loss: 0.0203 - val_accuracy: 0.9954
Epoch 53/500
54/54 [===========] - 219s 4s/step - loss: 0.0161 - accuracy: 0.9930 - val loss: 6.7486e-04 - val accuracy: 1.0000
Epoch 54/500
Epoch 55/500
```

```
Epoch 56/500
Epoch 57/500
Epoch 58/500
54/54 [===========] - 219s 4s/step - loss: 0.0159 - accuracy: 0.9953 - val loss: 2.5907e-04 - val accuracy: 1.0000
Epoch 59/500
Epoch 60/500
Epoch 61/500
54/54 [============== ] - 219s 4s/step - loss: 0.0318 - accuracy: 0.9907 - val loss: 0.0248 - val accuracy: 0.9977
Epoch 62/500
Epoch 64/500
54/54 [============== - 221s 4s/step - loss: 0.0237 - accuracy: 0.9930 - val_loss: 0.0233 - val_accuracy: 0.9977
Epoch 65/500
Epoch 66/500
54/54 [============ - 221s 4s/step - loss: 0.0189 - accuracy: 0.9942 - val_loss: 0.0425 - val_accuracy: 0.9954
Epoch 67/500
Epoch 68/500
Epoch 69/500
54/54 [============== - 220s 4s/step - loss: 0.0153 - accuracy: 0.9948 - val loss: 0.0089 - val accuracy: 0.9954
Epoch 70/500
54/54 [============= - 219s 4s/step - loss: 0.0281 - accuracy: 0.9942 - val_loss: 0.0063 - val_accuracy: 0.9977
Epoch 71/500
Epoch 72/500
54/54 [============== ] - 223s 4s/step - loss: 0.0130 - accuracy: 0.9948 - val loss: 0.0125 - val accuracy: 0.9977
Epoch 73/500
Epoch 74/500
54/54 [============= - 220s 4s/step - loss: 0.0308 - accuracy: 0.9930 - val loss: 0.0136 - val accuracy: 0.9954
Epoch 75/500
Epoch 76/500
Epoch 77/500
54/54 [============= - 219s 4s/step - loss: 0.0077 - accuracy: 0.9965 - val_loss: 0.0077 - val_accuracy: 0.9977
Epoch 78/500
Epoch 79/500
Epoch 80/500
54/54 [============= - 219s 4s/step - loss: 0.0226 - accuracy: 0.9936 - val_loss: 0.0082 - val_accuracy: 0.9954
Epoch 81/500
54/54 [============ - 219s 4s/step - loss: 0.0092 - accuracy: 0.9965 - val_loss: 0.0086 - val_accuracy: 0.9954
Epoch 83/500
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Epoch 84/500
54/54 [============== ] - 220s 4s/step - loss: 0.0173 - accuracy: 0.9924 - val loss: 0.0126 - val accuracy: 0.9954
Epoch 85/500
54/54 [============== - 221s 4s/step - loss: 0.0247 - accuracy: 0.9907 - val_loss: 0.0164 - val_accuracy: 0.9977
Epoch 86/500
Epoch 87/500
Epoch 88/500
Epoch 89/500
54/54 [============= - 222s 4s/step - loss: 0.1062 - accuracy: 0.9653 - val_loss: 0.0055 - val_accuracy: 1.0000
Epoch 90/500
Epoch 91/500
54/54 [============= - 220s 4s/step - loss: 0.0501 - accuracy: 0.9821 - val_loss: 0.0210 - val_accuracy: 0.9954
Epoch 92/500
Epoch 93/500
Epoch 94/500
Epoch 95/500
54/54 [============== ] - 232s 4s/step - loss: 0.0475 - accuracy: 0.9849 - val loss: 0.0071 - val accuracy: 0.9977
Epoch 96/500
Epoch 97/500
Epoch 98/500
Epoch 99/500
54/54 [===========] - 219s 4s/step - loss: 0.0196 - accuracy: 0.9924 - val loss: 7.7546e-04 - val accuracy: 1.0000
Epoch 100/500
54/54 [============== ] - 222s 4s/step - loss: 0.0215 - accuracy: 0.9913 - val loss: 0.0107 - val accuracy: 0.9954
Epoch 101/500
Epoch 102/500
54/54 [============= - 219s 4s/step - loss: 0.1536 - accuracy: 0.9738 - val_loss: 0.5487 - val_accuracy: 0.9306
Epoch 103/500
Epoch 104/500
Epoch 105/500
Epoch 106/500
Epoch 107/500
54/54 [============== ] - 220s 4s/step - loss: 0.0276 - accuracy: 0.9878 - val loss: 0.0068 - val accuracy: 0.9977
Epoch 108/500
Epoch 109/500
Epoch 110/500
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Epoch 111/500
Epoch 112/500
Epoch 113/500
Epoch 114/500
Epoch 115/500
54/54 [============== ] - 219s 4s/step - loss: 0.0357 - accuracy: 0.9831 - val loss: 0.0069 - val accuracy: 0.9977
Epoch 116/500
54/54 [============== ] - 221s 4s/step - loss: 0.0634 - accuracy: 0.9808 - val loss: 0.0320 - val accuracy: 0.9931
Epoch 117/500
Epoch 119/500
54/54 [============== - 219s 4s/step - loss: 0.0190 - accuracy: 0.9907 - val_loss: 0.0064 - val_accuracy: 0.9977
Epoch 120/500
Epoch 121/500
Epoch 122/500
Epoch 123/500
Epoch 124/500
Epoch 125/500
Epoch 126/500
54/54 [============== ] - 221s 4s/step - loss: 0.0228 - accuracy: 0.9924 - val loss: 0.0016 - val accuracy: 1.0000
Epoch 127/500
Epoch 128/500
Epoch 129/500
Epoch 130/500
Epoch 131/500
54/54 [============= - 220s 4s/step - loss: 0.0745 - accuracy: 0.9744 - val_loss: 0.0027 - val_accuracy: 1.0000
Epoch 132/500
Epoch 133/500
Epoch 134/500
54/54 [============== - 219s 4s/step - loss: 0.0213 - accuracy: 0.9936 - val loss: 0.0097 - val accuracy: 0.9977
Epoch 135/500
54/54 [============== - 220s 4s/step - loss: 0.0255 - accuracy: 0.9913 - val_loss: 0.0062 - val_accuracy: 0.9977
Epoch 136/500
54/54 [============== ] - 220s 4s/step - loss: 0.0206 - accuracy: 0.9930 - val loss: 0.0982 - val accuracy: 0.9907
Epoch 138/500
```

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Epoch 139/500
54/54 [============== ] - 220s 4s/step - loss: 0.0157 - accuracy: 0.9953 - val loss: 0.0459 - val accuracy: 0.9884
Epoch 140/500
54/54 [============= - 219s 4s/step - loss: 0.0267 - accuracy: 0.9884 - val_loss: 0.0013 - val_accuracy: 1.0000
Epoch 141/500
Epoch 142/500
54/54 [============= - 221s 4s/step - loss: 0.0282 - accuracy: 0.9878 - val loss: 0.0033 - val accuracy: 1.0000
Epoch 143/500
Epoch 144/500
54/54 [==========] - 219s 4s/step - loss: 0.0188 - accuracy: 0.9942 - val_loss: 3.3289e-04 - val_accuracy: 1.0000
Epoch 145/500
Epoch 146/500
54/54 [============= - 222s 4s/step - loss: 0.0098 - accuracy: 0.9988 - val_loss: 0.0078 - val_accuracy: 0.9977
Epoch 147/500
Epoch 148/500
Epoch 149/500
Epoch 150/500
54/54 [============== ] - 222s 4s/step - loss: 0.0179 - accuracy: 0.9959 - val loss: 0.0040 - val accuracy: 0.9977
Epoch 151/500
Epoch 152/500
Epoch 153/500
Epoch 154/500
54/54 [============= - 221s 4s/step - loss: 0.0216 - accuracy: 0.9930 - val loss: 0.0067 - val accuracy: 0.9977
Epoch 155/500
Epoch 156/500
54/54 [===========] - 219s 4s/step - loss: 0.0125 - accuracy: 0.9959 - val_loss: 9.7528e-04 - val_accuracy: 1.0000
Epoch 157/500
54/54 [============= - 220s 4s/step - loss: 0.0232 - accuracy: 0.9936 - val_loss: 0.0435 - val_accuracy: 0.9931
Epoch 158/500
Epoch 159/500
54/54 [============== - 219s 4s/step - loss: 0.0638 - accuracy: 0.9919 - val loss: 0.0042 - val accuracy: 0.9977
Epoch 160/500
54/54 [============== ] - 221s 4s/step - loss: 0.0409 - accuracy: 0.9907 - val loss: 0.0104 - val accuracy: 0.9977
Epoch 161/500
Epoch 162/500
Epoch 163/500
Epoch 164/500
Epoch 165/500
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```
Epoch 166/500
Epoch 167/500
54/54 [===========] - 223s 4s/step - loss: 0.0690 - accuracy: 0.9825 - val loss: 1.9901e-04 - val accuracy: 1.0000
Epoch 168/500
54/54 [============= - 229s 4s/step - loss: 0.0378 - accuracy: 0.9843 - val loss: 0.0030 - val accuracy: 1.0000
Epoch 169/500
Epoch 170/500
54/54 [============== ] - 224s 4s/step - loss: 0.0371 - accuracy: 0.9860 - val loss: 0.0043 - val accuracy: 1.0000
Epoch 171/500
Epoch 172/500
54/54 [============== - 226s 4s/step - loss: 0.0298 - accuracy: 0.9889 - val loss: 0.0151 - val accuracy: 0.9954
Epoch 174/500
54/54 [============ - 223s 4s/step - loss: 0.0315 - accuracy: 0.9895 - val_loss: 0.0109 - val_accuracy: 0.9954
Epoch 175/500
Epoch 176/500
54/54 [============ - 219s 4s/step - loss: 0.0140 - accuracy: 0.9936 - val_loss: 0.0111 - val_accuracy: 0.9977
Epoch 177/500
Epoch 178/500
Epoch 179/500
54/54 [===========] - 220s 4s/step - loss: 0.0233 - accuracy: 0.9895 - val loss: 3.5600e-04 - val accuracy: 1.0000
Epoch 180/500
Epoch 181/500
Epoch 182/500
Epoch 183/500
Epoch 184/500
Epoch 185/500
Epoch 186/500
54/54 [============== - 221s 4s/step - loss: 0.0222 - accuracy: 0.9942 - val_loss: 0.0040 - val_accuracy: 0.9977
Epoch 187/500
Epoch 188/500
Epoch 189/500
Epoch 190/500
54/54 [============= - 219s 4s/step - loss: 0.0335 - accuracy: 0.9907 - val_loss: 0.0100 - val_accuracy: 0.9931
Epoch 191/500
54/54 [============== ] - 221s 4s/step - loss: 0.0271 - accuracy: 0.9913 - val loss: 0.0113 - val accuracy: 0.9954
54/54 [============= - 219s 4s/step - loss: 0.0271 - accuracy: 0.9913 - val_loss: 0.0166 - val_accuracy: 0.9954
Epoch 193/500
```

```
54/54 [============== - 219s 4s/step - loss: 0.0268 - accuracy: 0.9924 - val loss: 0.0043 - val accuracy: 1.0000
Epoch 194/500
54/54 [============== ] - 220s 4s/step - loss: 0.0164 - accuracy: 0.9959 - val loss: 0.0028 - val accuracy: 1.0000
Epoch 195/500
54/54 [============= - 221s 4s/step - loss: 0.0219 - accuracy: 0.9924 - val_loss: 0.0027 - val_accuracy: 0.9977
Epoch 196/500
Epoch 197/500
Epoch 198/500
54/54 [===========] - 219s 4s/step - loss: 0.0185 - accuracy: 0.9953 - val loss: 8.9049e-04 - val accuracy: 1.0000
Epoch 199/500
54/54 [============= - 221s 4s/step - loss: 0.0136 - accuracy: 0.9965 - val_loss: 0.0026 - val_accuracy: 0.9977
Epoch 200/500
54/54 [============== ] - 220s 4s/step - loss: 0.0174 - accuracy: 0.9953 - val loss: 0.0026 - val accuracy: 0.9977
Epoch 201/500
54/54 [============= - 219s 4s/step - loss: 0.0139 - accuracy: 0.9965 - val_loss: 0.0027 - val_accuracy: 0.9977
Epoch 202/500
Epoch 203/500
Epoch 204/500
Epoch 205/500
54/54 [============== ] - 220s 4s/step - loss: 0.0339 - accuracy: 0.9930 - val loss: 0.0073 - val accuracy: 0.9954
Epoch 206/500
Epoch 207/500
Epoch 208/500
Epoch 209/500
Epoch 210/500
54/54 [============== ] - 219s 4s/step - loss: 0.0200 - accuracy: 0.9953 - val loss: 0.0025 - val accuracy: 0.9977
Epoch 211/500
54/54 [===========] - 221s 4s/step - loss: 0.0108 - accuracy: 0.9965 - val_loss: 5.5992e-04 - val_accuracy: 1.0000
Epoch 212/500
54/54 [============= - 219s 4s/step - loss: 0.0290 - accuracy: 0.9907 - val_loss: 0.0070 - val_accuracy: 0.9977
Epoch 213/500
Epoch 214/500
Epoch 215/500
Epoch 216/500
Epoch 217/500
Epoch 218/500
Epoch 219/500
Epoch 220/500
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Epoch 221/500
Epoch 222/500
Epoch 223/500
Epoch 224/500
Epoch 225/500
54/54 [============== ] - 221s 4s/step - loss: 0.0267 - accuracy: 0.9930 - val loss: 0.0193 - val accuracy: 0.9931
Epoch 226/500
54/54 [============== ] - 221s 4s/step - loss: 0.0279 - accuracy: 0.9919 - val loss: 0.0049 - val accuracy: 0.9977
Epoch 227/500
Epoch 229/500
54/54 [============== - 221s 4s/step - loss: 0.0357 - accuracy: 0.9924 - val_loss: 0.0041 - val_accuracy: 0.9977
Epoch 230/500
Epoch 231/500
54/54 [============== - 231s 4s/step - loss: 0.0386 - accuracy: 0.9901 - val_loss: 0.0042 - val_accuracy: 0.9977
Epoch 232/500
Epoch 233/500
Epoch 234/500
Epoch 235/500
Epoch 236/500
Epoch 237/500
54/54 [============== - 220s 4s/step - loss: 0.0234 - accuracy: 0.9936 - val loss: 0.0041 - val accuracy: 0.9977
Epoch 238/500
Epoch 239/500
54/54 [===========] - 224s 4s/step - loss: 0.0201 - accuracy: 0.9919 - val loss: 6.9230e-04 - val accuracy: 1.0000
Epoch 240/500
Epoch 241/500
54/54 [============== - 219s 4s/step - loss: 0.0187 - accuracy: 0.9936 - val_loss: 0.0113 - val_accuracy: 0.9977
Epoch 242/500
54/54 [==========] - 220s 4s/step - loss: 0.0132 - accuracy: 0.9948 - val_loss: 6.4353e-04 - val_accuracy: 1.0000
Epoch 243/500
54/54 [===========] - 220s 4s/step - loss: 0.0114 - accuracy: 0.9953 - val loss: 6.2864e-04 - val accuracy: 1.0000
Epoch 244/500
Epoch 245/500
54/54 [============== - 219s 4s/step - loss: 0.0192 - accuracy: 0.9924 - val_loss: 0.0045 - val_accuracy: 0.9977
Epoch 246/500
54/54 [============== ] - 219s 4s/step - loss: 0.0128 - accuracy: 0.9942 - val loss: 0.0029 - val accuracy: 1.0000
54/54 [============= - 219s 4s/step - loss: 0.0169 - accuracy: 0.9948 - val_loss: 0.0010 - val_accuracy: 1.0000
Epoch 248/500
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Epoch 249/500
Epoch 250/500
54/54 [===========] - 219s 4s/step - loss: 0.0209 - accuracy: 0.9913 - val_loss: 9.3031e-04 - val_accuracy: 1.0000
Epoch 251/500
Epoch 252/500
Epoch 253/500
Epoch 254/500
54/54 [============= - 219s 4s/step - loss: 0.0525 - accuracy: 0.9866 - val_loss: 0.0901 - val_accuracy: 0.9815
Epoch 255/500
Epoch 256/500
54/54 [============== - 221s 4s/step - loss: 0.0255 - accuracy: 0.9884 - val_loss: 0.0174 - val_accuracy: 0.9954
Epoch 257/500
Epoch 258/500
Epoch 259/500
Epoch 260/500
Epoch 261/500
Epoch 262/500
Epoch 263/500
Epoch 264/500
54/54 [===========] - 221s 4s/step - loss: 0.0179 - accuracy: 0.9919 - val loss: 5.2017e-05 - val accuracy: 1.0000
Epoch 265/500
Epoch 266/500
54/54 [============= - 218s 4s/step - loss: 0.0660 - accuracy: 0.9942 - val_loss: 0.0046 - val_accuracy: 0.9977
Epoch 267/500
54/54 [============== ] - 219s 4s/step - loss: 0.0160 - accuracy: 0.9953 - val loss: 0.0386 - val accuracy: 0.9954
Epoch 268/500
Epoch 269/500
Epoch 270/500
54/54 [============== ] - 219s 4s/step - loss: 0.0162 - accuracy: 0.9930 - val loss: 0.0029 - val accuracy: 0.9977
Epoch 271/500
54/54 [============== ] - 219s 4s/step - loss: 0.0166 - accuracy: 0.9924 - val loss: 0.0036 - val accuracy: 0.9977
Epoch 272/500
Epoch 273/500
Epoch 274/500
54/54 [===========] - 219s 4s/step - loss: 0.0202 - accuracy: 0.9930 - val loss: 2.0696e-08 - val accuracy: 1.0000
Epoch 275/500
54/54 [===========] - 219s 4s/step - loss: 0.0281 - accuracy: 0.9930 - val loss: 2.5497e-04 - val accuracy: 1.0000
```

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Epoch 276/500
Epoch 277/500
54/54 [===========] - 219s 4s/step - loss: 0.0243 - accuracy: 0.9919 - val loss: 1.3592e-06 - val accuracy: 1.0000
Epoch 278/500
Epoch 279/500
Epoch 280/500
Epoch 281/500
Epoch 282/500
54/54 [===========] - 219s 4s/step - loss: 0.0155 - accuracy: 0.9936 - val loss: 3.1260e-06 - val accuracy: 1.0000
Epoch 284/500
54/54 [===========] - 219s 4s/step - loss: 0.0167 - accuracy: 0.9930 - val_loss: 2.8025e-06 - val_accuracy: 1.0000
Epoch 285/500
Epoch 286/500
Epoch 287/500
Epoch 288/500
Epoch 289/500
54/54 [============== ] - 220s 4s/step - loss: 0.0150 - accuracy: 0.9930 - val loss: 0.0432 - val accuracy: 0.9977
Epoch 290/500
54/54 [============= - 219s 4s/step - loss: 0.0286 - accuracy: 0.9913 - val_loss: 0.0169 - val_accuracy: 0.9954
Epoch 291/500
54/54 [============== ] - 219s 4s/step - loss: 0.0128 - accuracy: 0.9965 - val loss: 0.0274 - val accuracy: 0.9954
Epoch 292/500
Epoch 293/500
Epoch 294/500
Epoch 295/500
Epoch 296/500
54/54 [============== - 221s 4s/step - loss: 0.0199 - accuracy: 0.9942 - val_loss: 0.0026 - val_accuracy: 1.0000
Epoch 297/500
Epoch 298/500
Epoch 299/500
Epoch 300/500
54/54 [============= - 219s 4s/step - loss: 0.0087 - accuracy: 0.9959 - val_loss: 0.0011 - val_accuracy: 1.0000
Epoch 301/500
Epoch 303/500
```

```
Epoch 304/500
Epoch 305/500
54/54 [============== - 222s 4s/step - loss: 0.0138 - accuracy: 0.9924 - val_loss: 0.0043 - val_accuracy: 0.9977
Epoch 306/500
Epoch 307/500
Epoch 308/500
Epoch 309/500
54/54 [============ - 221s 4s/step - loss: 0.0136 - accuracy: 0.9953 - val_loss: 0.0458 - val_accuracy: 0.9954
Epoch 310/500
54/54 [============== ] - 219s 4s/step - loss: 0.0386 - accuracy: 0.9930 - val loss: 0.0066 - val accuracy: 0.9977
Epoch 311/500
54/54 [===========] - 219s 4s/step - loss: 0.0461 - accuracy: 0.9913 - val_loss: 3.8561e-04 - val_accuracy: 1.0000
Epoch 312/500
54/54 [===========] - 220s 4s/step - loss: 0.0471 - accuracy: 0.9907 - val loss: 4.2198e-05 - val accuracy: 1.0000
Epoch 313/500
54/54 [============== ] - 237s 4s/step - loss: 0.1223 - accuracy: 0.9889 - val loss: 0.0317 - val accuracy: 0.9977
Epoch 314/500
Epoch 315/500
54/54 [============== ] - 220s 4s/step - loss: 0.0323 - accuracy: 0.9889 - val loss: 0.0286 - val accuracy: 0.9954
Epoch 316/500
Epoch 317/500
Epoch 318/500
Epoch 319/500
Epoch 320/500
54/54 [============= ] - 235s 4s/step - loss: 0.0161 - accuracy: 0.9913 - val loss: 0.0133 - val accuracy: 0.9954
Epoch 321/500
Epoch 322/500
Epoch 323/500
Epoch 324/500
Epoch 325/500
Epoch 326/500
Epoch 327/500
Epoch 328/500
Epoch 329/500
Epoch 330/500
```

```
Epoch 331/500
Epoch 332/500
Epoch 333/500
Epoch 334/500
Epoch 335/500
Epoch 336/500
Epoch 337/500
54/54 [============= - 300s 6s/step - loss: 0.0116 - accuracy: 0.9954 - val loss: 0.0058 - val accuracy: 0.9954
Epoch 339/500
54/54 [============ - 270s 5s/step - loss: 0.0117 - accuracy: 0.9953 - val_loss: 0.0056 - val_accuracy: 0.9954
Epoch 340/500
Epoch 341/500
Epoch 342/500
Epoch 343/500
Epoch 344/500
Epoch 345/500
Epoch 346/500
54/54 [============== ] - 249s 5s/step - loss: 0.0047 - accuracy: 0.9983 - val loss: 0.0022 - val accuracy: 0.9977
Epoch 347/500
Epoch 348/500
54/54 [===========] - 231s 4s/step - loss: 0.0103 - accuracy: 0.9965 - val loss: 3.9813e-04 - val accuracy: 1.0000
Epoch 349/500
54/54 [===========] - 227s 4s/step - loss: 0.0059 - accuracy: 0.9977 - val loss: 3.9850e-04 - val accuracy: 1.0000
Epoch 350/500
Epoch 351/500
54/54 [============] - 225s 4s/step - loss: 0.0104 - accuracy: 0.9977 - val_loss: 4.0436e-04 - val_accuracy: 1.0000
Epoch 352/500
Epoch 353/500
Epoch 354/500
Epoch 355/500
54/54 [============] - 225s 4s/step - loss: 0.0434 - accuracy: 0.9936 - val_loss: 1.6133e-04 - val_accuracy: 1.0000
Epoch 356/500
54/54 [============] - 227s 4s/step - loss: 0.0208 - accuracy: 0.9965 - val_loss: 2.4197e-04 - val_accuracy: 1.0000
Epoch 358/500
```

```
Epoch 359/500
Epoch 360/500
54/54 [============= - 229s 4s/step - loss: 0.0096 - accuracy: 0.9959 - val_loss: 0.0377 - val_accuracy: 0.9977
Epoch 361/500
54/54 [===========] - 227s 4s/step - loss: 0.0132 - accuracy: 0.9942 - val_loss: 1.8673e-05 - val_accuracy: 1.0000
Epoch 362/500
Epoch 363/500
54/54 [===========] - 225s 4s/step - loss: 0.0667 - accuracy: 0.9936 - val loss: 9.3022e-06 - val accuracy: 1.0000
Epoch 364/500
54/54 [============= - 228s 4s/step - loss: 0.0341 - accuracy: 0.9930 - val_loss: 0.0128 - val_accuracy: 0.9954
Epoch 365/500
Epoch 366/500
Epoch 367/500
Epoch 368/500
Epoch 369/500
54/54 [============= - 227s 4s/step - loss: 0.0247 - accuracy: 0.9936 - val loss: 0.0033 - val accuracy: 0.9977
Epoch 370/500
54/54 [============== ] - 227s 4s/step - loss: 0.0119 - accuracy: 0.9936 - val loss: 0.0033 - val accuracy: 0.9977
Epoch 371/500
Epoch 372/500
54/54 [============= - 233s 4s/step - loss: 0.0139 - accuracy: 0.9948 - val_loss: 0.0033 - val_accuracy: 0.9977
Epoch 373/500
Epoch 374/500
Epoch 375/500
Epoch 376/500
54/54 [============= - 229s 4s/step - loss: 0.0270 - accuracy: 0.9959 - val_loss: 0.0057 - val_accuracy: 0.9977
Epoch 377/500
Epoch 378/500
Epoch 379/500
Epoch 380/500
54/54 [===========] - 231s 4s/step - loss: 0.0051 - accuracy: 0.9988 - val_loss: 5.7408e-05 - val_accuracy: 1.0000
Epoch 381/500
Epoch 382/500
54/54 [===========] - 227s 4s/step - loss: 0.0058 - accuracy: 0.9983 - val_loss: 8.2727e-05 - val_accuracy: 1.0000
Epoch 383/500
54/54 [===========] - 227s 4s/step - loss: 0.0118 - accuracy: 0.9959 - val loss: 8.7093e-05 - val accuracy: 1.0000
Epoch 384/500
54/54 [===========] - 229s 4s/step - loss: 0.0073 - accuracy: 0.9983 - val loss: 9.7997e-05 - val accuracy: 1.0000
Epoch 385/500
54/54 [===========] - 227s 4s/step - loss: 0.0216 - accuracy: 0.9965 - val loss: 6.3936e-04 - val accuracy: 1.0000
```

```
Epoch 386/500
54/54 [===========] - 226s 4s/step - loss: 0.0118 - accuracy: 0.9942 - val loss: 6.9829e-04 - val accuracy: 1.0000
Epoch 387/500
Epoch 388/500
Epoch 389/500
Epoch 390/500
Epoch 391/500
Epoch 392/500
Epoch 394/500
54/54 [============== - 226s 4s/step - loss: 0.0356 - accuracy: 0.9977 - val_loss: 0.0031 - val_accuracy: 0.9977
Epoch 395/500
Epoch 396/500
54/54 [============= - 274s 5s/step - loss: 0.0106 - accuracy: 0.9959 - val_loss: 0.0024 - val_accuracy: 0.9977
Epoch 397/500
Epoch 398/500
Epoch 399/500
54/54 [============== - 229s 4s/step - loss: 0.1153 - accuracy: 0.9930 - val loss: 0.0044 - val accuracy: 0.9977
Epoch 400/500
54/54 [============] - 227s 4s/step - loss: 0.0188 - accuracy: 0.9977 - val_loss: 6.2322e-04 - val_accuracy: 1.0000
Epoch 401/500
Epoch 402/500
Epoch 403/500
54/54 [===========] - 228s 4s/step - loss: 0.0077 - accuracy: 0.9977 - val loss: 3.8210e-05 - val accuracy: 1.0000
Epoch 404/500
54/54 [===========] - 227s 4s/step - loss: 0.0160 - accuracy: 0.9959 - val loss: 4.8841e-04 - val accuracy: 1.0000
Epoch 405/500
54/54 [===========] - 226s 4s/step - loss: 0.0081 - accuracy: 0.9965 - val loss: 5.9236e-04 - val accuracy: 1.0000
Epoch 406/500
54/54 [============ - 226s 4s/step - loss: 0.0097 - accuracy: 0.9965 - val_loss: 0.0058 - val_accuracy: 0.9954
Epoch 407/500
Epoch 408/500
54/54 [===========] - 248s 5s/step - loss: 0.0072 - accuracy: 0.9965 - val loss: 3.6061e-04 - val accuracy: 1.0000
Epoch 409/500
Epoch 410/500
54/54 [============= - 258s 5s/step - loss: 0.0129 - accuracy: 0.9953 - val_loss: 0.0159 - val_accuracy: 0.9954
Epoch 411/500
54/54 [============== ] - 323s 6s/step - loss: 0.0126 - accuracy: 0.9953 - val loss: 0.0105 - val accuracy: 0.9931
Epoch 413/500
```

```
Epoch 414/500
Epoch 415/500
Epoch 416/500
Epoch 417/500
Epoch 418/500
Epoch 419/500
54/54 [==========] - 227s 4s/step - loss: 0.0098 - accuracy: 0.9959 - val_loss: 1.0191e-06 - val_accuracy: 1.0000
Epoch 420/500
Epoch 421/500
54/54 [============== ] - 229s 4s/step - loss: 0.0102 - accuracy: 0.9959 - val_loss: 7.6652e-07 - val_accuracy: 1.0000
Epoch 422/500
54/54 [===========] - 226s 4s/step - loss: 0.0037 - accuracy: 0.9994 - val loss: 3.3304e-07 - val accuracy: 1.0000
Epoch 423/500
Epoch 424/500
Epoch 425/500
54/54 [============== ] - 229s 4s/step - loss: 0.2075 - accuracy: 0.9913 - val loss: 0.1020 - val accuracy: 0.9977
Epoch 426/500
Epoch 427/500
Epoch 428/500
Epoch 429/500
Epoch 430/500
54/54 [============== ] - 272s 5s/step - loss: 0.1515 - accuracy: 0.9930 - val loss: 0.0224 - val accuracy: 0.9977
Epoch 431/500
54/54 [===========] - 272s 5s/step - loss: 0.0337 - accuracy: 0.9965 - val_loss: 1.3004e-04 - val_accuracy: 1.0000
Epoch 432/500
Epoch 433/500
Epoch 434/500
Epoch 435/500
Epoch 436/500
Epoch 437/500
Epoch 438/500
54/54 [===========] - 256s 5s/step - loss: 0.0235 - accuracy: 0.9965 - val loss: 9.8287e-05 - val accuracy: 1.0000
Epoch 439/500
54/54 [===========] - 270s 5s/step - loss: 0.0157 - accuracy: 0.9936 - val loss: 5.1326e-08 - val accuracy: 1.0000
Epoch 440/500
54/54 [===========] - 299s 6s/step - loss: 0.0106 - accuracy: 0.9965 - val loss: 1.4625e-08 - val accuracy: 1.0000
```

```
Epoch 441/500
54/54 [===========] - 265s 5s/step - loss: 0.0072 - accuracy: 0.9977 - val loss: 1.4349e-08 - val accuracy: 1.0000
Epoch 442/500
54/54 [===========] - 266s 5s/step - loss: 0.0076 - accuracy: 0.9983 - val loss: 1.4625e-08 - val accuracy: 1.0000
Epoch 443/500
54/54 [===========] - 258s 5s/step - loss: 0.0164 - accuracy: 0.9924 - val loss: 8.8303e-09 - val accuracy: 1.0000
Epoch 444/500
54/54 [============== ] - 284s 5s/step - loss: 0.0220 - accuracy: 0.9959 - val loss: 0.0037 - val accuracy: 0.9977
Epoch 445/500
Epoch 446/500
Epoch 447/500
54/54 [===========] - 322s 6s/step - loss: 0.0199 - accuracy: 0.9965 - val loss: 7.1153e-06 - val accuracy: 1.0000
Epoch 449/500
54/54 [============] - 270s 5s/step - loss: 0.0106 - accuracy: 0.9959 - val_loss: 1.3910e-05 - val_accuracy: 1.0000
Epoch 450/500
54/54 [===========] - 273s 5s/step - loss: 0.0089 - accuracy: 0.9971 - val_loss: 1.3883e-05 - val_accuracy: 1.0000
Epoch 451/500
54/54 [============] - 259s 5s/step - loss: 0.0109 - accuracy: 0.9959 - val_loss: 1.3667e-05 - val_accuracy: 1.0000
Epoch 452/500
54/54 [===========] - 288s 5s/step - loss: 0.0088 - accuracy: 0.9965 - val loss: 1.3654e-05 - val accuracy: 1.0000
Epoch 453/500
54/54 [============ ] - 341s 6s/step - loss: 0.0074 - accuracy: 0.9971 - val loss: 1.0883e-05 - val accuracy: 1.0000
Epoch 454/500
Epoch 455/500
54/54 [============] - 287s 5s/step - loss: 0.0080 - accuracy: 0.9965 - val_loss: 2.8418e-06 - val_accuracy: 1.0000
Epoch 456/500
Epoch 457/500
Epoch 458/500
54/54 [===========] - 236s 4s/step - loss: 0.0094 - accuracy: 0.9959 - val loss: 1.8507e-06 - val accuracy: 1.0000
Epoch 459/500
54/54 [===========] - 236s 4s/step - loss: 0.0057 - accuracy: 0.9983 - val loss: 1.8305e-06 - val accuracy: 1.0000
Epoch 460/500
54/54 [============== ] - 235s 4s/step - loss: 0.0074 - accuracy: 0.9977 - val loss: 0.0606 - val accuracy: 0.9977
Epoch 461/500
54/54 [============] - 234s 4s/step - loss: 0.0053 - accuracy: 0.9988 - val_loss: 2.7348e-04 - val_accuracy: 1.0000
Epoch 462/500
Epoch 463/500
Epoch 464/500
Epoch 465/500
54/54 [============== ] - 382s 7s/step - loss: 0.0202 - accuracy: 0.9983 - val_loss: 1.1700e-07 - val_accuracy: 1.0000
Epoch 466/500
54/54 [============= ] - 410s 8s/step - loss: 0.0072 - accuracy: 0.9977 - val_loss: 1.9316e-09 - val_accuracy: 1.0000
Epoch 468/500
```

```
Epoch 469/500
Epoch 470/500
54/54 [============== ] - 391s 7s/step - loss: 0.0065 - accuracy: 0.9983 - val_loss: 9.1265e-07 - val_accuracy: 1.0000
Epoch 471/500
54/54 [==========] - 372s 7s/step - loss: 0.0139 - accuracy: 0.9971 - val_loss: 8.9130e-08 - val_accuracy: 1.0000
Epoch 472/500
54/54 [===========] - 372s 7s/step - loss: 0.0082 - accuracy: 0.9971 - val loss: 5.4637e-08 - val accuracy: 1.0000
Epoch 473/500
54/54 [===========] - 360s 7s/step - loss: 0.0174 - accuracy: 0.9965 - val loss: 3.7253e-08 - val accuracy: 1.0000
Epoch 474/500
Epoch 475/500
Epoch 476/500
54/54 [===========] - 233s 4s/step - loss: 0.0095 - accuracy: 0.9965 - val_loss: 3.2899e-04 - val_accuracy: 1.0000
Epoch 477/500
54/54 [===========] - 231s 4s/step - loss: 0.0064 - accuracy: 0.9971 - val loss: 4.2621e-04 - val accuracy: 1.0000
Epoch 478/500
Epoch 479/500
Epoch 480/500
54/54 [============== ] - 229s 4s/step - loss: 0.0383 - accuracy: 0.9948 - val loss: 0.0025 - val accuracy: 0.9977
Epoch 481/500
Epoch 482/500
54/54 [===========] - 232s 4s/step - loss: 0.0032 - accuracy: 1.0000 - val_loss: 4.5444e-07 - val_accuracy: 1.0000
Epoch 483/500
54/54 [============= ] - 340s 6s/step - loss: 0.0079 - accuracy: 0.9977 - val loss: 1.3245e-07 - val accuracy: 1.0000
Epoch 484/500
54/54 [===========] - 368s 7s/step - loss: 0.0055 - accuracy: 0.9983 - val loss: 1.0707e-07 - val accuracy: 1.0000
Epoch 485/500
Epoch 486/500
54/54 [============= ] - 424s 8s/step - loss: 0.0030 - accuracy: 0.9994 - val_loss: 9.6028e-08 - val_accuracy: 1.0000
Epoch 487/500
Epoch 488/500
Epoch 489/500
Epoch 490/500
Epoch 491/500
Epoch 492/500
54/54 [===========] - 395s 7s/step - loss: 0.0057 - accuracy: 0.9983 - val_loss: 0.0000e+00 - val_accuracy: 1.0000
Epoch 493/500
Epoch 494/500
Epoch 495/500
```

#### APRESENTANDO OS RESULTADOS

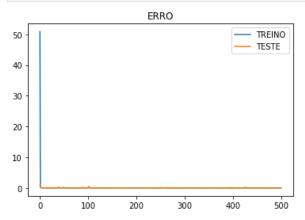
```
In [35]: results = model.evaluate_generator(test_generator, 437)
    print('Acc: %.3f, Loss: %.3f' % (results[1], results[0]))

    C:\Users\gui_r\AppData\Local\Temp\ipykernel_16068\2082521232.py:1: UserWarning: `Model.evaluate_generator` is deprecated and will be removed in a future version. Pleas
    e use `Model.evaluate`, which supports generators.
        results = model.evaluate_generator(test_generator, 437)
    WARNING:tensorflow:Your input ran out of data; interrupting training. Make sure that your dataset or generator can generate at least `steps_per_epoch * epochs` batches
    (in this case, 437 batches). You may need to use the repeat() function when building your dataset.
    Acc: 0.998, Loss: 0.029
```

### 11. APRESENTANDO OS GRÁFICOS DE LINHA

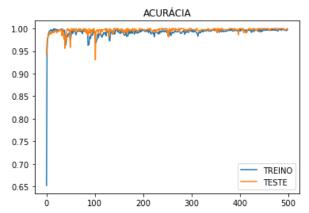
• ERRO DO TESTE E TREINO

```
In [36]: plt.title('ERRO')
  plt.plot(history.history['loss'], label='TREINO')
  plt.plot(history.history['val_loss'], label='TESTE')
  plt.legend()
  plt.show()
```

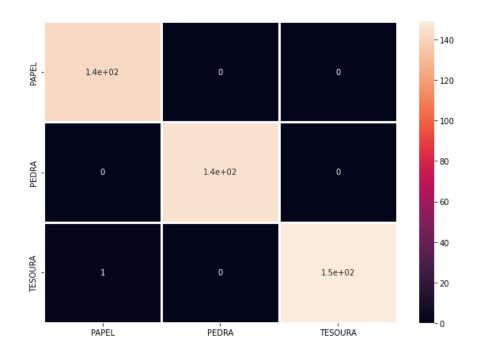


ACURÁCIA DO TESTE E TREINO

```
In [37]: plt.title('ACURÁCIA')
    plt.plot(history.history['accuracy'], label='TREINO')
    plt.plot(history.history['val_accuracy'], label='TESTE')
    plt.legend()
    plt.show()
```



### 12. APRESENTANDO A MATRIZ DE CONFUSÃO



## 13. APRESENTANDO OS DADOS DO CONJUNTO DE TESTES

# ACURÁCIA, PRECISÃO, REVOCAÇÃO E F1-SCORE

In [39]: print(classification\_report(test\_generator.classes, y\_pred)) recall f1-score support precision 142 0.99 1.00 1.00 1.00 1.00 1.00 145 1.00 0.99 1.00 150 437 accuracy 1.00 macro avg 1.00 1.00 1.00 437 weighted avg 1.00 1.00 1.00 437

In [ ]: