Trabalho_2_Deep_Learning

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1 Projeto 2 - Deep Learning

- Carlos Vinicius R Gonçalves Ra:1914723
- Gabriel Sana Ferreira da Silva Ra:1914766

2 Bibliotecas

Para a realização dessa atividade foi utilizada as seguintes biblioteca: * Pandas - Biblioteca de software criada para a linguagem Python para manipulação e análise de dados. * Numpy - Biblioteca Python que é usada principalmente para realizar cálculos em Arrays Multidimensionais. * Keras - Biblioteca open-source que dispõe uma interface em Python para redes neurais artificiais. O keras age como uma interface para a biblioteca TensorFlow, do Google. * Matplotlib - Biblioteca para a visualização de dados em Python. Ele apresenta uma API orientada a objetos que permite a criação de gráficos em 2D de uma forma simples e com poucos comandos.

```
import pandas as pd
import numpy as np
import os
import keras
import matplotlib.pyplot as plt

from sklearn.metrics import classification_report, confusion_matrix
from keras.layers import Dense,GlobalAveragePooling2D
from keras.applications import MobileNet
from keras.preprocessing import image
from keras.applications.mobilenet import preprocess_input
from keras.preprocessing.image import ImageDataGenerator
from keras.models import Model
from keras.optimizers import Adam
```

3 Importação do modelo e criação das camadas de neurônios

- Para o modelo pré-treinado da literatura, dentre os disponíveis no link (https://keras.io/api/applications/) foi utilizado o MobileNet, que por sua vez utiliza os pesos pré treinados da ImageNet.
- Foi utilizado o 'GlobalAveragePooling2D' para a extração das características na primeira camada.

- Foi utilizado o 'relu' (Rectified Linear Unit) no cálculo de função de ativação das camadas intermediárias.
- Foi utilizado a ativação 'softmax' na última camada, com suas devidas 3 classes para a predição das classes do dataset.

```
[3]: base_model=MobileNet(weights='imagenet',include_top=False) #importa o modelo de_
    →mobilenet e descarta a última camada de 1000 neurônios.

x=base_model.output
x=GlobalAveragePooling2D()(x)
x=Dense(1024,activation='relu')(x) #adicionamos camadas densas para que o_
    →modelo possa aprender funções mais complexas e classificar para melhores_
    →resultados.
x=Dense(1024,activation='relu')(x) #dense layer 2
x=Dense(512,activation='relu')(x) #dense layer 3
preds=Dense(3,activation='softmax')(x) #final layer com softmax activation
```

WARNING:tensorflow:`input_shape` is undefined or non-square, or `rows` is not in [128, 160, 192, 224]. Weights for input shape (224, 224) will be loaded as the default.

```
[4]: model=Model(inputs=base_model.input,outputs=preds)
#especificar as entradas
#especifique as saídas
#agora um modelo foi criado com base em nossa arquitetura
```

```
[5]: for layer in model.layers[:20]:
    layer.trainable=False
    for layer in model.layers[20:]:
        layer.trainable=True
```

4 Base de Dados

Para a realização deste projeto foi necessário escolher um dataset do site kaggle para o treinamento do modelo de deep learning. O dataset selecionado pelo grupo foi o "Rock-Paper-Scissors-Images", ou "Jo-Ken-Po", que representa o jogo pedra papel e tesoura.

Esse dataset contém 2189 imagens dos gestos usados para representar uma pedra ou papel ou tesoura no jogo. Essas imagens estão separadas em pastas representando cada uma das três classes e o código a seguir é responsável por fazer a divisão entre treino (80%) e teste (20%) dessas imagens.

```
[6]: train_data_gen = ___ 

→ ImageDataGenerator(preprocessing_function=preprocess_input,validation_split=0.

→8, rescale=1./255, shear_range = 0.2, zoom_range = 0.2) #included in our__ 

→ dependencies 
#Geração dos dados de treino (80%)

train_generator = train_data_gen.flow_from_directory('dataset', target_size=(64, 64), batch_size=32,
```

Found 1749 images belonging to 3 classes. Found 439 images belonging to 3 classes.

5 Definição do otimizador, steps e realização do treinamento e teste

- O otimizador utilizado pelo grupo foi o 'Adam', que apresenta um melhor resultado.
- A métrica de avaliação do treinamento foi a acurácia.
- A métrica de erro utilizado foi a 'categorical_crossentropy', já que o dataset utilizado apresenta mais do que duas classes, impossibilitando o uso do 'binary_crossentropy'.
- O treinamento foi realizado em 500 epochs.

```
accuracy: 0.9632 - val_loss: 2.6271 - val_accuracy: 0.3317
Epoch 4/500
accuracy: 0.9750 - val_loss: 1.0373 - val_accuracy: 0.6322
Epoch 5/500
54/54 [============= ] - 16s 303ms/step - loss: 0.0971 -
accuracy: 0.9725 - val_loss: 4.5178 - val_accuracy: 0.6274
Epoch 6/500
accuracy: 0.9764 - val_loss: 9.5787 - val_accuracy: 0.3365
Epoch 7/500
accuracy: 0.9745 - val_loss: 1.6418 - val_accuracy: 0.6202
Epoch 8/500
accuracy: 0.9745 - val_loss: 12.1697 - val_accuracy: 0.3317
Epoch 9/500
54/54 [============= ] - 17s 307ms/step - loss: 0.0987 -
accuracy: 0.9696 - val_loss: 4.3325 - val_accuracy: 0.3341
Epoch 10/500
accuracy: 0.9818 - val_loss: 1.8286 - val_accuracy: 0.6538
Epoch 11/500
accuracy: 0.9895 - val_loss: 12.6313 - val_accuracy: 0.3462
Epoch 12/500
accuracy: 0.9760 - val_loss: 0.4219 - val_accuracy: 0.9663
Epoch 13/500
accuracy: 0.9677 - val_loss: 0.9848 - val_accuracy: 0.6562
Epoch 14/500
accuracy: 0.9638 - val loss: 1.9498 - val accuracy: 0.3510
Epoch 15/500
accuracy: 0.9824 - val_loss: 14.5133 - val_accuracy: 0.3221
Epoch 16/500
accuracy: 0.9684 - val_loss: 19.2740 - val_accuracy: 0.3365
Epoch 17/500
accuracy: 0.9635 - val_loss: 26.4419 - val_accuracy: 0.3389
Epoch 18/500
accuracy: 0.9867 - val_loss: 8.7733 - val_accuracy: 0.3341
Epoch 19/500
```

```
accuracy: 0.9750 - val_loss: 22.7262 - val_accuracy: 0.3365
Epoch 20/500
accuracy: 0.9914 - val_loss: 8.0728 - val_accuracy: 0.5457
Epoch 21/500
accuracy: 0.9778 - val_loss: 13.9867 - val_accuracy: 0.3317
Epoch 22/500
54/54 [============= ] - 18s 339ms/step - loss: 0.0390 -
accuracy: 0.9886 - val_loss: 3.8246 - val_accuracy: 0.4423
Epoch 23/500
accuracy: 0.9830 - val_loss: 2.3932 - val_accuracy: 0.6659
Epoch 24/500
accuracy: 0.9896 - val_loss: 24.6374 - val_accuracy: 0.3365
Epoch 25/500
54/54 [============= ] - 16s 300ms/step - loss: 0.0245 -
accuracy: 0.9901 - val_loss: 24.8931 - val_accuracy: 0.3221
Epoch 26/500
accuracy: 0.9952 - val_loss: 15.5425 - val_accuracy: 0.3341
Epoch 27/500
accuracy: 0.9904 - val_loss: 9.2503 - val_accuracy: 0.5505
Epoch 28/500
accuracy: 0.9804 - val_loss: 28.3348 - val_accuracy: 0.3365
Epoch 29/500
54/54 [============ ] - 17s 320ms/step - loss: 0.0247 -
accuracy: 0.9958 - val_loss: 0.4850 - val_accuracy: 0.7043
Epoch 30/500
accuracy: 0.9920 - val_loss: 8.9016 - val_accuracy: 0.3317
Epoch 31/500
54/54 [============= ] - 17s 311ms/step - loss: 0.0061 -
accuracy: 0.9986 - val_loss: 2.2905 - val_accuracy: 0.6779
Epoch 32/500
accuracy: 0.9933 - val_loss: 2.7070 - val_accuracy: 0.3269
Epoch 33/500
accuracy: 0.9886 - val_loss: 13.4739 - val_accuracy: 0.3221
Epoch 34/500
accuracy: 0.9792 - val_loss: 8.2792 - val_accuracy: 0.3245
Epoch 35/500
```

```
accuracy: 0.9875 - val_loss: 8.7024 - val_accuracy: 0.3317
Epoch 36/500
accuracy: 0.9722 - val_loss: 19.4724 - val_accuracy: 0.3341
Epoch 37/500
accuracy: 0.9950 - val_loss: 4.3731 - val_accuracy: 0.4447
Epoch 38/500
accuracy: 0.9926 - val_loss: 4.1820 - val_accuracy: 0.6082
Epoch 39/500
accuracy: 0.9863 - val_loss: 17.0178 - val_accuracy: 0.3413
Epoch 40/500
accuracy: 0.9921 - val_loss: 21.8703 - val_accuracy: 0.3341
Epoch 41/500
accuracy: 0.9952 - val_loss: 22.4195 - val_accuracy: 0.3558
Epoch 42/500
accuracy: 0.9908 - val_loss: 32.3237 - val_accuracy: 0.3317
Epoch 43/500
accuracy: 0.9941 - val_loss: 33.7887 - val_accuracy: 0.3341
Epoch 44/500
accuracy: 0.9924 - val_loss: 40.0359 - val_accuracy: 0.3293
Epoch 45/500
accuracy: 0.9932 - val_loss: 46.1657 - val_accuracy: 0.3413
Epoch 46/500
accuracy: 0.9993 - val loss: 59.9077 - val accuracy: 0.3365
Epoch 47/500
accuracy: 0.9881 - val_loss: 36.3791 - val_accuracy: 0.3293
Epoch 48/500
accuracy: 0.9975 - val_loss: 27.3736 - val_accuracy: 0.3293
Epoch 49/500
accuracy: 0.9856 - val_loss: 21.9677 - val_accuracy: 0.3341
Epoch 50/500
accuracy: 0.9887 - val_loss: 29.7851 - val_accuracy: 0.3269
Epoch 51/500
```

```
accuracy: 0.9943 - val_loss: 16.8720 - val_accuracy: 0.3389
Epoch 52/500
accuracy: 0.9764 - val loss: 29.8850 - val accuracy: 0.3413
Epoch 53/500
accuracy: 0.9829 - val_loss: 25.3516 - val_accuracy: 0.3245
Epoch 54/500
accuracy: 0.9698 - val_loss: 12.2010 - val_accuracy: 0.5192
Epoch 55/500
accuracy: 0.9825 - val_loss: 36.4433 - val_accuracy: 0.3221
Epoch 56/500
accuracy: 0.9770 - val_loss: 19.5120 - val_accuracy: 0.3798
Epoch 57/500
accuracy: 0.9933 - val_loss: 54.9786 - val_accuracy: 0.3341
Epoch 58/500
accuracy: 0.9957 - val_loss: 71.8453 - val_accuracy: 0.3413
Epoch 59/500
accuracy: 0.9987 - val_loss: 59.1999 - val_accuracy: 0.3365
Epoch 60/500
accuracy: 0.9977 - val_loss: 17.6527 - val_accuracy: 0.3197
Epoch 61/500
accuracy: 0.9767 - val_loss: 1.3973 - val_accuracy: 0.4255
Epoch 62/500
accuracy: 0.9723 - val_loss: 27.4339 - val_accuracy: 0.3365
Epoch 63/500
54/54 [============= ] - 17s 323ms/step - loss: 0.0869 -
accuracy: 0.9825 - val_loss: 53.9124 - val_accuracy: 0.3293
Epoch 64/500
accuracy: 0.9807 - val_loss: 66.0329 - val_accuracy: 0.3221
Epoch 65/500
accuracy: 0.9925 - val_loss: 23.2149 - val_accuracy: 0.3534
Epoch 66/500
54/54 [============= ] - 18s 337ms/step - loss: 0.0319 -
accuracy: 0.9916 - val_loss: 20.0540 - val_accuracy: 0.3389
Epoch 67/500
```

```
accuracy: 0.9988 - val_loss: 13.4462 - val_accuracy: 0.4928
Epoch 68/500
accuracy: 0.9969 - val loss: 25.8290 - val accuracy: 0.3293
Epoch 69/500
54/54 [============ ] - 16s 298ms/step - loss: 0.0283 -
accuracy: 0.9921 - val_loss: 24.0323 - val_accuracy: 0.3438
Epoch 70/500
accuracy: 0.9919 - val_loss: 32.7899 - val_accuracy: 0.3341
Epoch 71/500
accuracy: 1.0000 - val_loss: 39.1873 - val_accuracy: 0.3341
Epoch 72/500
accuracy: 0.9970 - val_loss: 17.9413 - val_accuracy: 0.3413
Epoch 73/500
accuracy: 0.9959 - val_loss: 0.1151 - val_accuracy: 0.9591
Epoch 74/500
accuracy: 0.9909 - val_loss: 1.9834 - val_accuracy: 0.6130
Epoch 75/500
accuracy: 0.9987 - val_loss: 34.3543 - val_accuracy: 0.3317
Epoch 76/500
accuracy: 0.9942 - val_loss: 2.7662 - val_accuracy: 0.4663
Epoch 77/500
54/54 [============= ] - 17s 313ms/step - loss: 0.0068 -
accuracy: 0.9988 - val_loss: 7.8735 - val_accuracy: 0.4135
Epoch 78/500
accuracy: 0.9940 - val_loss: 32.8413 - val_accuracy: 0.3365
Epoch 79/500
54/54 [============= ] - 17s 314ms/step - loss: 0.0128 -
accuracy: 0.9930 - val_loss: 1.5313 - val_accuracy: 0.6851
Epoch 80/500
accuracy: 0.9983 - val_loss: 1.9092 - val_accuracy: 0.6827
Epoch 81/500
accuracy: 0.9934 - val_loss: 35.8909 - val_accuracy: 0.3197
Epoch 82/500
accuracy: 0.9950 - val_loss: 9.7760 - val_accuracy: 0.5673
Epoch 83/500
```

```
accuracy: 0.9950 - val_loss: 3.9986 - val_accuracy: 0.6466
Epoch 84/500
accuracy: 0.9966 - val_loss: 24.8852 - val_accuracy: 0.3534
Epoch 85/500
accuracy: 0.9987 - val_loss: 16.9621 - val_accuracy: 0.3341
Epoch 86/500
54/54 [============= ] - 19s 345ms/step - loss: 0.0078 -
accuracy: 0.9960 - val_loss: 39.5791 - val_accuracy: 0.5409
Epoch 87/500
accuracy: 0.9986 - val_loss: 9.0084 - val_accuracy: 0.4183
Epoch 88/500
accuracy: 0.9993 - val_loss: 13.2334 - val_accuracy: 0.6683
Epoch 89/500
accuracy: 0.9974 - val_loss: 43.1830 - val_accuracy: 0.3389
Epoch 90/500
accuracy: 0.9912 - val_loss: 106.8340 - val_accuracy: 0.3317
Epoch 91/500
accuracy: 0.9948 - val_loss: 42.2665 - val_accuracy: 0.3197
Epoch 92/500
accuracy: 0.9975 - val_loss: 0.9375 - val_accuracy: 0.6394
Epoch 93/500
accuracy: 1.0000 - val_loss: 55.8504 - val_accuracy: 0.3486
Epoch 94/500
accuracy: 0.9895 - val loss: 14.0523 - val accuracy: 0.5168
Epoch 95/500
accuracy: 0.9762 - val_loss: 14.0737 - val_accuracy: 0.3221
Epoch 96/500
accuracy: 0.9787 - val_loss: 4.0765 - val_accuracy: 0.4784
Epoch 97/500
accuracy: 0.9856 - val_loss: 46.4800 - val_accuracy: 0.3365
Epoch 98/500
accuracy: 0.9901 - val_loss: 46.4905 - val_accuracy: 0.3462
Epoch 99/500
```

```
accuracy: 0.9956 - val_loss: 37.1175 - val_accuracy: 0.3510
Epoch 100/500
accuracy: 0.9967 - val_loss: 2.8773 - val_accuracy: 0.6611
Epoch 101/500
54/54 [============= ] - 19s 345ms/step - loss: 0.0114 -
accuracy: 0.9955 - val_loss: 7.3593 - val_accuracy: 0.5865
Epoch 102/500
54/54 [============= ] - 19s 353ms/step - loss: 0.0091 -
accuracy: 0.9972 - val_loss: 18.8954 - val_accuracy: 0.4375
Epoch 103/500
accuracy: 0.9963 - val_loss: 10.5248 - val_accuracy: 0.4976
Epoch 104/500
accuracy: 0.9982 - val_loss: 4.9357 - val_accuracy: 0.5793
Epoch 105/500
54/54 [============= ] - 19s 344ms/step - loss: 0.0019 -
accuracy: 0.9998 - val_loss: 9.5911 - val_accuracy: 0.3462
Epoch 106/500
accuracy: 0.9910 - val_loss: 10.7497 - val_accuracy: 0.4255
Epoch 107/500
accuracy: 0.9944 - val_loss: 33.2379 - val_accuracy: 0.3221
Epoch 108/500
accuracy: 0.9975 - val_loss: 48.3126 - val_accuracy: 0.3269
Epoch 109/500
accuracy: 0.9992 - val_loss: 1.5244 - val_accuracy: 0.7236
Epoch 110/500
accuracy: 0.9772 - val loss: 48.6761 - val accuracy: 0.3365
Epoch 111/500
accuracy: 0.9980 - val_loss: 28.7520 - val_accuracy: 0.4543
Epoch 112/500
accuracy: 0.9946 - val_loss: 13.1986 - val_accuracy: 0.3870
Epoch 113/500
accuracy: 0.9848 - val_loss: 57.4512 - val_accuracy: 0.3341
Epoch 114/500
accuracy: 0.9963 - val_loss: 70.0984 - val_accuracy: 0.3365
Epoch 115/500
```

```
accuracy: 0.9980 - val_loss: 2.5083 - val_accuracy: 0.4543
Epoch 116/500
accuracy: 0.9967 - val_loss: 42.8185 - val_accuracy: 0.3245
Epoch 117/500
54/54 [============= ] - 19s 347ms/step - loss: 0.0039 -
accuracy: 1.0000 - val_loss: 9.5193 - val_accuracy: 0.5913
Epoch 118/500
54/54 [============= ] - 19s 348ms/step - loss: 0.0029 -
accuracy: 0.9984 - val_loss: 65.9858 - val_accuracy: 0.3534
Epoch 119/500
accuracy: 0.9962 - val_loss: 71.7023 - val_accuracy: 0.3389
Epoch 120/500
accuracy: 0.9974 - val_loss: 27.9667 - val_accuracy: 0.3365
Epoch 121/500
accuracy: 0.9890 - val_loss: 27.1689 - val_accuracy: 0.3269
Epoch 122/500
accuracy: 0.9951 - val_loss: 11.9578 - val_accuracy: 0.5938
Epoch 123/500
accuracy: 0.9979 - val_loss: 43.8122 - val_accuracy: 0.3413
Epoch 124/500
accuracy: 0.9644 - val_loss: 4.4448 - val_accuracy: 0.4375
Epoch 125/500
accuracy: 0.9805 - val_loss: 74.3904 - val_accuracy: 0.3293
Epoch 126/500
accuracy: 0.9956 - val loss: 94.6304 - val accuracy: 0.3438
Epoch 127/500
accuracy: 0.9994 - val_loss: 70.4924 - val_accuracy: 0.3293
Epoch 128/500
accuracy: 0.9958 - val_loss: 109.1095 - val_accuracy: 0.3269
Epoch 129/500
accuracy: 0.9919 - val_loss: 3.3741 - val_accuracy: 0.6707
Epoch 130/500
accuracy: 0.9985 - val_loss: 141.9347 - val_accuracy: 0.3365
Epoch 131/500
```

```
accuracy: 0.9937 - val_loss: 69.6763 - val_accuracy: 0.3293
Epoch 132/500
accuracy: 1.0000 - val_loss: 78.8748 - val_accuracy: 0.3942
Epoch 133/500
accuracy: 1.0000 - val_loss: 64.7945 - val_accuracy: 0.5721
Epoch 134/500
54/54 [============= ] - 19s 344ms/step - loss: 0.0045 -
accuracy: 0.9978 - val_loss: 169.6505 - val_accuracy: 0.3245
Epoch 135/500
accuracy: 0.9999 - val_loss: 69.8450 - val_accuracy: 0.4447
Epoch 136/500
accuracy: 0.9997 - val_loss: 5.8184 - val_accuracy: 0.6707
Epoch 137/500
accuracy: 0.9939 - val_loss: 116.1733 - val_accuracy: 0.3365
Epoch 138/500
accuracy: 0.9972 - val_loss: 122.7325 - val_accuracy: 0.3293
Epoch 139/500
accuracy: 0.9763 - val_loss: 102.0887 - val_accuracy: 0.3966
Epoch 140/500
accuracy: 0.9881 - val_loss: 29.8844 - val_accuracy: 0.3389
Epoch 141/500
54/54 [============= ] - 19s 350ms/step - loss: 0.0260 -
accuracy: 0.9899 - val_loss: 9.1920 - val_accuracy: 0.6202
Epoch 142/500
accuracy: 0.9791 - val_loss: 3.7597 - val_accuracy: 0.6683
Epoch 143/500
54/54 [============= ] - 18s 338ms/step - loss: 0.0045 -
accuracy: 0.9982 - val_loss: 6.9729 - val_accuracy: 0.6346
Epoch 144/500
accuracy: 0.9989 - val_loss: 0.9452 - val_accuracy: 0.8966
Epoch 145/500
accuracy: 0.9974 - val_loss: 2.0514 - val_accuracy: 0.6707
Epoch 146/500
accuracy: 0.9976 - val_loss: 3.5278 - val_accuracy: 0.3942
Epoch 147/500
```

```
accuracy: 0.9991 - val_loss: 0.0643 - val_accuracy: 0.9856
Epoch 148/500
54/54 [============= ] - 19s 346ms/step - loss: 4.7868e-05 -
accuracy: 1.0000 - val_loss: 0.0736 - val_accuracy: 0.9880
Epoch 149/500
accuracy: 1.0000 - val_loss: 0.2402 - val_accuracy: 0.9712
Epoch 150/500
accuracy: 0.9999 - val_loss: 0.1989 - val_accuracy: 0.9567
Epoch 151/500
accuracy: 0.9871 - val_loss: 14.6794 - val_accuracy: 0.6562
Epoch 152/500
accuracy: 0.9974 - val_loss: 18.5263 - val_accuracy: 0.5144
Epoch 153/500
accuracy: 0.9977 - val_loss: 1.7204 - val_accuracy: 0.6490
Epoch 154/500
accuracy: 1.0000 - val_loss: 1.0072 - val_accuracy: 0.7163
Epoch 155/500
accuracy: 1.0000 - val_loss: 0.0429 - val_accuracy: 0.9880
Epoch 156/500
accuracy: 0.9915 - val_loss: 50.3642 - val_accuracy: 0.3389
Epoch 157/500
accuracy: 0.9750 - val_loss: 3.0523 - val_accuracy: 0.6370
Epoch 158/500
accuracy: 0.9934 - val loss: 59.6182 - val accuracy: 0.3293
Epoch 159/500
accuracy: 0.9899 - val_loss: 66.6053 - val_accuracy: 0.3317
Epoch 160/500
accuracy: 0.9953 - val_loss: 33.1099 - val_accuracy: 0.3510
Epoch 161/500
accuracy: 0.9975 - val_loss: 53.4686 - val_accuracy: 0.3269
Epoch 162/500
accuracy: 0.9978 - val_loss: 38.0449 - val_accuracy: 0.3317
Epoch 163/500
```

```
accuracy: 1.0000 - val_loss: 13.6990 - val_accuracy: 0.5192
Epoch 164/500
accuracy: 0.9986 - val_loss: 124.4887 - val_accuracy: 0.3317
Epoch 165/500
accuracy: 0.9998 - val_loss: 119.7394 - val_accuracy: 0.3341
Epoch 166/500
accuracy: 0.9999 - val_loss: 120.6628 - val_accuracy: 0.3221
Epoch 167/500
54/54 [============ ] - 19s 344ms/step - loss: 0.0169 -
accuracy: 0.9974 - val_loss: 104.6451 - val_accuracy: 0.3269
Epoch 168/500
accuracy: 0.9975 - val_loss: 10.7130 - val_accuracy: 0.5529
Epoch 169/500
accuracy: 0.9890 - val_loss: 107.4608 - val_accuracy: 0.3293
Epoch 170/500
accuracy: 0.9994 - val_loss: 114.7476 - val_accuracy: 0.3365
Epoch 171/500
accuracy: 1.0000 - val_loss: 112.0127 - val_accuracy: 0.3293
Epoch 172/500
accuracy: 1.0000 - val_loss: 40.4314 - val_accuracy: 0.4928
Epoch 173/500
accuracy: 1.0000 - val_loss: 7.3669 - val_accuracy: 0.6635
Epoch 174/500
accuracy: 1.0000 - val_loss: 1.7648 - val_accuracy: 0.7740
Epoch 175/500
54/54 [============= ] - 17s 318ms/step - loss: 1.9974e-05 -
accuracy: 1.0000 - val_loss: 0.1623 - val_accuracy: 0.9760
Epoch 176/500
accuracy: 1.0000 - val_loss: 0.0846 - val_accuracy: 0.9880
Epoch 177/500
accuracy: 0.9998 - val_loss: 89.7190 - val_accuracy: 0.3389
Epoch 178/500
accuracy: 0.9902 - val_loss: 1.3296 - val_accuracy: 0.6466
Epoch 179/500
```

```
accuracy: 0.9941 - val_loss: 67.8263 - val_accuracy: 0.3317
Epoch 180/500
accuracy: 0.9992 - val_loss: 91.8273 - val_accuracy: 0.3389
Epoch 181/500
accuracy: 0.9984 - val_loss: 54.3687 - val_accuracy: 0.3317
Epoch 182/500
accuracy: 0.9954 - val_loss: 68.0787 - val_accuracy: 0.3245
Epoch 183/500
accuracy: 0.9979 - val_loss: 73.2711 - val_accuracy: 0.3317
Epoch 184/500
54/54 [============= - - 16s 293ms/step - loss: 0.0035 -
accuracy: 0.9987 - val_loss: 4.3280 - val_accuracy: 0.4062
Epoch 185/500
accuracy: 0.9987 - val_loss: 0.9301 - val_accuracy: 0.6611
Epoch 186/500
accuracy: 0.9979 - val_loss: 78.3781 - val_accuracy: 0.3269
Epoch 187/500
accuracy: 0.9980 - val_loss: 62.6589 - val_accuracy: 0.3606
Epoch 188/500
accuracy: 0.9954 - val_loss: 26.6324 - val_accuracy: 0.6611
Epoch 189/500
54/54 [=========== ] - 17s 309ms/step - loss: 0.0019 -
accuracy: 0.9988 - val_loss: 22.0281 - val_accuracy: 0.6707
Epoch 190/500
54/54 [============= ] - 17s 305ms/step - loss: 0.0025 -
accuracy: 0.9997 - val loss: 1.1803 - val accuracy: 0.7332
Epoch 191/500
accuracy: 0.9952 - val_loss: 2.9656 - val_accuracy: 0.6731
Epoch 192/500
54/54 [============== ] - 17s 310ms/step - loss: 0.0108 -
accuracy: 0.9962 - val_loss: 57.7008 - val_accuracy: 0.3245
Epoch 193/500
accuracy: 0.9998 - val_loss: 84.0329 - val_accuracy: 0.3269
Epoch 194/500
accuracy: 0.9995 - val_loss: 81.6532 - val_accuracy: 0.3221
Epoch 195/500
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accuracy: 0.9985 - val_loss: 98.9346 - val_accuracy: 0.3245
Epoch 196/500
accuracy: 0.9998 - val_loss: 48.7523 - val_accuracy: 0.3413
Epoch 197/500
accuracy: 0.9986 - val_loss: 50.3682 - val_accuracy: 0.3413
Epoch 198/500
54/54 [============= ] - 17s 308ms/step - loss: 0.0032 -
accuracy: 0.9975 - val_loss: 27.6761 - val_accuracy: 0.3582
Epoch 199/500
accuracy: 1.0000 - val_loss: 6.1784 - val_accuracy: 0.5409
Epoch 200/500
54/54 [============ - - 17s 314ms/step - loss: 0.0020 -
accuracy: 0.9989 - val_loss: 1.0828 - val_accuracy: 0.7668
Epoch 201/500
54/54 [============ ] - 17s 310ms/step - loss: 2.1327e-04 -
accuracy: 1.0000 - val_loss: 0.0938 - val_accuracy: 0.9832
Epoch 202/500
accuracy: 1.0000 - val_loss: 0.0368 - val_accuracy: 0.9952
Epoch 203/500
accuracy: 1.0000 - val_loss: 0.0493 - val_accuracy: 0.9952
Epoch 204/500
accuracy: 1.0000 - val_loss: 0.0469 - val_accuracy: 0.9952
Epoch 205/500
accuracy: 1.0000 - val_loss: 0.0313 - val_accuracy: 0.9952
Epoch 206/500
accuracy: 1.0000 - val loss: 0.3871 - val accuracy: 0.9327
Epoch 207/500
54/54 [============= ] - 17s 309ms/step - loss: 0.0266 -
accuracy: 0.9969 - val_loss: 36.7802 - val_accuracy: 0.3269
Epoch 208/500
accuracy: 0.9882 - val_loss: 19.2613 - val_accuracy: 0.3534
Epoch 209/500
54/54 [============= ] - 17s 313ms/step - loss: 0.0219 -
accuracy: 0.9933 - val_loss: 35.2791 - val_accuracy: 0.3221
Epoch 210/500
accuracy: 0.9969 - val_loss: 97.3211 - val_accuracy: 0.3269
Epoch 211/500
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accuracy: 0.9960 - val_loss: 103.3467 - val_accuracy: 0.3269
Epoch 212/500
accuracy: 0.9971 - val loss: 125.3893 - val accuracy: 0.3365
Epoch 213/500
accuracy: 0.9997 - val_loss: 116.7638 - val_accuracy: 0.3365
Epoch 214/500
accuracy: 0.9929 - val_loss: 126.9912 - val_accuracy: 0.3365
Epoch 215/500
accuracy: 0.9844 - val_loss: 83.8732 - val_accuracy: 0.3365
Epoch 216/500
accuracy: 0.9961 - val_loss: 72.9643 - val_accuracy: 0.3365
Epoch 217/500
54/54 [============= ] - 17s 308ms/step - loss: 0.0118 -
accuracy: 0.9968 - val_loss: 0.4971 - val_accuracy: 0.8582
Epoch 218/500
accuracy: 0.9906 - val_loss: 0.2487 - val_accuracy: 0.9519
Epoch 219/500
accuracy: 0.9909 - val_loss: 109.0853 - val_accuracy: 0.3269
Epoch 220/500
accuracy: 0.9965 - val_loss: 8.9789 - val_accuracy: 0.4014
Epoch 221/500
54/54 [============= ] - 17s 315ms/step - loss: 0.0054 -
accuracy: 0.9955 - val_loss: 3.0936 - val_accuracy: 0.6178
Epoch 222/500
54/54 [============= ] - 17s 311ms/step - loss: 1.1319e-04 -
accuracy: 1.0000 - val loss: 0.4795 - val accuracy: 0.9351
Epoch 223/500
54/54 [============= ] - 17s 311ms/step - loss: 3.9691e-04 -
accuracy: 1.0000 - val_loss: 0.0668 - val_accuracy: 0.9880
Epoch 224/500
accuracy: 1.0000 - val_loss: 3.1132 - val_accuracy: 0.5913
Epoch 225/500
54/54 [============= ] - 17s 308ms/step - loss: 1.5566e-04 -
accuracy: 1.0000 - val_loss: 0.0916 - val_accuracy: 0.9784
Epoch 226/500
accuracy: 0.9998 - val_loss: 1.1883 - val_accuracy: 0.7909
Epoch 227/500
```

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accuracy: 0.9955 - val_loss: 16.0098 - val_accuracy: 0.3269
Epoch 228/500
accuracy: 0.9853 - val_loss: 11.7751 - val_accuracy: 0.3365
Epoch 229/500
54/54 [============= ] - 17s 313ms/step - loss: 0.0831 -
accuracy: 0.9929 - val_loss: 12.8509 - val_accuracy: 0.3702
Epoch 230/500
54/54 [============= ] - 17s 309ms/step - loss: 0.0433 -
accuracy: 0.9898 - val_loss: 0.0908 - val_accuracy: 0.9736
Epoch 231/500
accuracy: 0.9878 - val_loss: 13.5961 - val_accuracy: 0.5457
Epoch 232/500
accuracy: 0.9959 - val_loss: 115.1288 - val_accuracy: 0.4231
Epoch 233/500
54/54 [============= ] - 17s 308ms/step - loss: 0.0346 -
accuracy: 0.9874 - val_loss: 70.7416 - val_accuracy: 0.3438
Epoch 234/500
accuracy: 0.9987 - val_loss: 0.9138 - val_accuracy: 0.8894
Epoch 235/500
accuracy: 0.9934 - val_loss: 2.9870 - val_accuracy: 0.7284
Epoch 236/500
accuracy: 0.9992 - val_loss: 9.5957 - val_accuracy: 0.6490
Epoch 237/500
accuracy: 0.9995 - val_loss: 1.9733 - val_accuracy: 0.7933
Epoch 238/500
54/54 [============= ] - 17s 307ms/step - loss: 7.8776e-04 -
accuracy: 1.0000 - val loss: 1.5001 - val accuracy: 0.8582
Epoch 239/500
54/54 [============= ] - 17s 308ms/step - loss: 0.0027 -
accuracy: 0.9996 - val_loss: 0.2352 - val_accuracy: 0.9832
Epoch 240/500
54/54 [============= ] - 17s 311ms/step - loss: 7.6048e-04 -
accuracy: 0.9999 - val_loss: 0.0998 - val_accuracy: 0.9928
Epoch 241/500
accuracy: 1.0000 - val_loss: 0.1304 - val_accuracy: 0.9880
Epoch 242/500
accuracy: 1.0000 - val_loss: 0.0818 - val_accuracy: 0.9928
Epoch 243/500
```

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accuracy: 1.0000 - val_loss: 0.0838 - val_accuracy: 0.9904
Epoch 244/500
accuracy: 1.0000 - val loss: 0.1188 - val accuracy: 0.9880
Epoch 245/500
54/54 [============= ] - 17s 310ms/step - loss: 2.1338e-05 -
accuracy: 1.0000 - val_loss: 0.1272 - val_accuracy: 0.9928
Epoch 246/500
54/54 [============= ] - 17s 311ms/step - loss: 7.1030e-06 -
accuracy: 1.0000 - val_loss: 0.0381 - val_accuracy: 0.9952
Epoch 247/500
accuracy: 1.0000 - val_loss: 0.0645 - val_accuracy: 0.9928
Epoch 248/500
accuracy: 1.0000 - val_loss: 0.0579 - val_accuracy: 0.9928
Epoch 249/500
54/54 [============ ] - 17s 307ms/step - loss: 6.7506e-06 -
accuracy: 1.0000 - val_loss: 0.0655 - val_accuracy: 0.9904
Epoch 250/500
accuracy: 1.0000 - val_loss: 0.2088 - val_accuracy: 0.9688
Epoch 251/500
accuracy: 0.9979 - val_loss: 213.0587 - val_accuracy: 0.3317
Epoch 252/500
accuracy: 0.9996 - val_loss: 98.8510 - val_accuracy: 0.5577
Epoch 253/500
accuracy: 0.9937 - val_loss: 219.8500 - val_accuracy: 0.3389
Epoch 254/500
accuracy: 0.9992 - val loss: 166.5837 - val accuracy: 0.3966
Epoch 255/500
accuracy: 1.0000 - val_loss: 55.3658 - val_accuracy: 0.6082
Epoch 256/500
accuracy: 1.0000 - val_loss: 3.1707 - val_accuracy: 0.7548
Epoch 257/500
54/54 [============= ] - 17s 314ms/step - loss: 1.7971e-04 -
accuracy: 1.0000 - val_loss: 0.5948 - val_accuracy: 0.9399
Epoch 258/500
accuracy: 0.9981 - val_loss: 45.5208 - val_accuracy: 0.3173
Epoch 259/500
```

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accuracy: 0.9958 - val_loss: 82.4774 - val_accuracy: 0.3317
Epoch 260/500
accuracy: 0.9961 - val_loss: 60.1401 - val_accuracy: 0.3197
Epoch 261/500
54/54 [============ ] - 17s 306ms/step - loss: 0.0033 -
accuracy: 0.9997 - val_loss: 6.5259 - val_accuracy: 0.4111
Epoch 262/500
accuracy: 0.9915 - val_loss: 1.8188 - val_accuracy: 0.6659
Epoch 263/500
accuracy: 1.0000 - val_loss: 0.0595 - val_accuracy: 0.9832
Epoch 264/500
accuracy: 1.0000 - val_loss: 0.0369 - val_accuracy: 0.9928
Epoch 265/500
54/54 [============ ] - 17s 310ms/step - loss: 1.6088e-04 -
accuracy: 1.0000 - val_loss: 0.0281 - val_accuracy: 0.9952
Epoch 266/500
accuracy: 1.0000 - val_loss: 1.2125 - val_accuracy: 0.7861
Epoch 267/500
accuracy: 1.0000 - val_loss: 0.0675 - val_accuracy: 0.9880
Epoch 268/500
accuracy: 1.0000 - val_loss: 0.0571 - val_accuracy: 0.9880
Epoch 269/500
54/54 [============ ] - 17s 311ms/step - loss: 0.0104 -
accuracy: 0.9969 - val_loss: 4.9695 - val_accuracy: 0.3486
Epoch 270/500
accuracy: 0.9951 - val_loss: 8.2384 - val_accuracy: 0.5986
Epoch 271/500
54/54 [============= ] - 17s 310ms/step - loss: 0.0045 -
accuracy: 0.9993 - val_loss: 20.6670 - val_accuracy: 0.4038
Epoch 272/500
accuracy: 0.9933 - val_loss: 121.5924 - val_accuracy: 0.3413
Epoch 273/500
54/54 [============ ] - 17s 311ms/step - loss: 0.0013 -
accuracy: 0.9997 - val_loss: 175.7887 - val_accuracy: 0.3341
Epoch 274/500
accuracy: 0.9991 - val_loss: 22.3897 - val_accuracy: 0.3269
Epoch 275/500
```

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accuracy: 0.9992 - val_loss: 13.2568 - val_accuracy: 0.3438
Epoch 276/500
accuracy: 1.0000 - val loss: 2.0511 - val accuracy: 0.6418
Epoch 277/500
54/54 [============ ] - 17s 307ms/step - loss: 1.1806e-04 -
accuracy: 1.0000 - val_loss: 0.3532 - val_accuracy: 0.9231
Epoch 278/500
54/54 [============ ] - 17s 311ms/step - loss: 1.0192e-05 -
accuracy: 1.0000 - val_loss: 0.0976 - val_accuracy: 0.9856
Epoch 279/500
accuracy: 1.0000 - val_loss: 1.0599 - val_accuracy: 0.8630
Epoch 280/500
accuracy: 0.9966 - val_loss: 60.9927 - val_accuracy: 0.3221
Epoch 281/500
accuracy: 0.9949 - val_loss: 27.1026 - val_accuracy: 0.3269
Epoch 282/500
accuracy: 0.9919 - val_loss: 38.0822 - val_accuracy: 0.3245
Epoch 283/500
accuracy: 1.0000 - val_loss: 6.0436 - val_accuracy: 0.4976
Epoch 284/500
accuracy: 1.0000 - val_loss: 0.4862 - val_accuracy: 0.9207
Epoch 285/500
accuracy: 0.9969 - val_loss: 19.3160 - val_accuracy: 0.6442
Epoch 286/500
accuracy: 0.9970 - val loss: 4.7806 - val accuracy: 0.6611
Epoch 287/500
54/54 [============= ] - 17s 311ms/step - loss: 0.0048 -
accuracy: 0.9986 - val_loss: 3.2098 - val_accuracy: 0.5889
Epoch 288/500
accuracy: 0.9953 - val_loss: 201.1188 - val_accuracy: 0.3317
Epoch 289/500
accuracy: 0.9992 - val_loss: 5.4820 - val_accuracy: 0.4231
Epoch 290/500
accuracy: 1.0000 - val_loss: 1.8663 - val_accuracy: 0.6106
Epoch 291/500
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accuracy: 1.0000 - val_loss: 0.2614 - val_accuracy: 0.9255
Epoch 292/500
accuracy: 0.9989 - val_loss: 0.0917 - val_accuracy: 0.9712
Epoch 293/500
54/54 [============= ] - 17s 311ms/step - loss: 0.0046 -
accuracy: 0.9992 - val_loss: 3.5203 - val_accuracy: 0.6490
Epoch 294/500
54/54 [============= ] - 17s 307ms/step - loss: 0.0055 -
accuracy: 1.0000 - val_loss: 1.8396 - val_accuracy: 0.6827
Epoch 295/500
accuracy: 0.9998 - val_loss: 0.6158 - val_accuracy: 0.8438
Epoch 296/500
accuracy: 0.9935 - val_loss: 46.2791 - val_accuracy: 0.3149
Epoch 297/500
accuracy: 1.0000 - val_loss: 10.0443 - val_accuracy: 0.3365
Epoch 298/500
accuracy: 0.9994 - val_loss: 102.4854 - val_accuracy: 0.3197
Epoch 299/500
accuracy: 0.9985 - val_loss: 186.1451 - val_accuracy: 0.3269
Epoch 300/500
accuracy: 0.9829 - val_loss: 113.1238 - val_accuracy: 0.3317
Epoch 301/500
54/54 [============= ] - 17s 312ms/step - loss: 0.0082 -
accuracy: 0.9981 - val_loss: 132.8485 - val_accuracy: 0.3245
Epoch 302/500
accuracy: 0.9938 - val loss: 38.0350 - val accuracy: 0.3245
Epoch 303/500
accuracy: 0.9998 - val_loss: 127.2534 - val_accuracy: 0.3221
Epoch 304/500
accuracy: 0.9968 - val_loss: 14.7443 - val_accuracy: 0.3197
Epoch 305/500
accuracy: 0.9959 - val_loss: 15.6937 - val_accuracy: 0.3630
Epoch 306/500
accuracy: 0.9968 - val_loss: 7.3809 - val_accuracy: 0.6202
Epoch 307/500
```

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accuracy: 0.9976 - val_loss: 59.1746 - val_accuracy: 0.3197
Epoch 308/500
accuracy: 0.9959 - val_loss: 8.4888 - val_accuracy: 0.3606
Epoch 309/500
accuracy: 0.9996 - val_loss: 2.3146 - val_accuracy: 0.7380
Epoch 310/500
accuracy: 1.0000 - val_loss: 40.3080 - val_accuracy: 0.6394
Epoch 311/500
accuracy: 0.9999 - val_loss: 136.1943 - val_accuracy: 0.3534
Epoch 312/500
accuracy: 0.9994 - val_loss: 77.3868 - val_accuracy: 0.4399
Epoch 313/500
54/54 [============= ] - 17s 306ms/step - loss: 2.1050e-04 -
accuracy: 1.0000 - val_loss: 40.4712 - val_accuracy: 0.6130
Epoch 314/500
accuracy: 0.9999 - val_loss: 7.0279 - val_accuracy: 0.6851
Epoch 315/500
accuracy: 0.9997 - val_loss: 17.9665 - val_accuracy: 0.5409
Epoch 316/500
accuracy: 1.0000 - val_loss: 0.2254 - val_accuracy: 0.9784
Epoch 317/500
accuracy: 1.0000 - val_loss: 0.0663 - val_accuracy: 0.9904
Epoch 318/500
accuracy: 1.0000 - val loss: 0.0484 - val accuracy: 0.9928
Epoch 319/500
accuracy: 1.0000 - val_loss: 0.0468 - val_accuracy: 0.9952
Epoch 320/500
accuracy: 0.9973 - val_loss: 154.8115 - val_accuracy: 0.3245
Epoch 321/500
accuracy: 1.0000 - val_loss: 163.0662 - val_accuracy: 0.3173
Epoch 322/500
accuracy: 1.0000 - val_loss: 150.2540 - val_accuracy: 0.3317
Epoch 323/500
```

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accuracy: 0.9999 - val_loss: 32.2344 - val_accuracy: 0.3245
Epoch 324/500
accuracy: 0.9999 - val loss: 0.0548 - val accuracy: 0.9928
Epoch 325/500
54/54 [============= ] - 17s 310ms/step - loss: 0.0029 -
accuracy: 0.9973 - val_loss: 0.7241 - val_accuracy: 0.8438
Epoch 326/500
54/54 [============== ] - 17s 311ms/step - loss: 0.0286 -
accuracy: 0.9956 - val_loss: 120.4672 - val_accuracy: 0.3389
Epoch 327/500
accuracy: 0.9969 - val_loss: 8.3452 - val_accuracy: 0.4327
Epoch 328/500
54/54 [============ ] - 17s 311ms/step - loss: 6.7081e-04 -
accuracy: 1.0000 - val_loss: 1.9350 - val_accuracy: 0.6683
Epoch 329/500
54/54 [============ ] - 17s 318ms/step - loss: 2.0056e-04 -
accuracy: 1.0000 - val_loss: 0.5134 - val_accuracy: 0.8462
Epoch 330/500
accuracy: 1.0000 - val_loss: 0.1644 - val_accuracy: 0.9688
Epoch 331/500
accuracy: 1.0000 - val_loss: 0.2595 - val_accuracy: 0.9760
Epoch 332/500
54/54 [============= ] - 17s 312ms/step - loss: 4.4680e-06 -
accuracy: 1.0000 - val_loss: 0.1188 - val_accuracy: 0.9856
Epoch 333/500
54/54 [============ ] - 17s 313ms/step - loss: 2.1494e-05 -
accuracy: 1.0000 - val_loss: 0.0571 - val_accuracy: 0.9880
Epoch 334/500
54/54 [============= ] - 17s 309ms/step - loss: 3.6969e-06 -
accuracy: 1.0000 - val loss: 0.0731 - val accuracy: 0.9928
Epoch 335/500
54/54 [============= ] - 17s 311ms/step - loss: 2.7960e-06 -
accuracy: 1.0000 - val_loss: 0.0775 - val_accuracy: 0.9904
Epoch 336/500
accuracy: 1.0000 - val_loss: 0.0564 - val_accuracy: 0.9904
Epoch 337/500
54/54 [============= ] - 17s 307ms/step - loss: 2.6077e-06 -
accuracy: 1.0000 - val_loss: 0.1205 - val_accuracy: 0.9928
Epoch 338/500
accuracy: 1.0000 - val_loss: 0.0676 - val_accuracy: 0.9952
Epoch 339/500
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accuracy: 1.0000 - val_loss: 0.0709 - val_accuracy: 0.9952
Epoch 340/500
accuracy: 1.0000 - val loss: 0.0406 - val accuracy: 0.9952
Epoch 341/500
54/54 [============= ] - 17s 309ms/step - loss: 9.8416e-06 -
accuracy: 1.0000 - val_loss: 0.0480 - val_accuracy: 0.9952
Epoch 342/500
accuracy: 1.0000 - val_loss: 0.0475 - val_accuracy: 0.9928
Epoch 343/500
accuracy: 1.0000 - val_loss: 0.1340 - val_accuracy: 0.9952
Epoch 344/500
accuracy: 1.0000 - val_loss: 0.0864 - val_accuracy: 0.9952
Epoch 345/500
54/54 [============ ] - 17s 307ms/step - loss: 3.5314e-07 -
accuracy: 1.0000 - val_loss: 0.0555 - val_accuracy: 0.9952
Epoch 346/500
accuracy: 1.0000 - val_loss: 0.0599 - val_accuracy: 0.9904
Epoch 347/500
accuracy: 1.0000 - val_loss: 0.1232 - val_accuracy: 0.9928
Epoch 348/500
accuracy: 0.9960 - val_loss: 44.0763 - val_accuracy: 0.3317
Epoch 349/500
accuracy: 0.9904 - val_loss: 68.9829 - val_accuracy: 0.3317
Epoch 350/500
accuracy: 0.9930 - val_loss: 41.2128 - val_accuracy: 0.3894
Epoch 351/500
54/54 [============= ] - 17s 310ms/step - loss: 0.0032 -
accuracy: 0.9993 - val_loss: 13.1568 - val_accuracy: 0.5072
Epoch 352/500
accuracy: 0.9891 - val_loss: 1.1000 - val_accuracy: 0.7596
Epoch 353/500
accuracy: 0.9965 - val_loss: 2.2588 - val_accuracy: 0.5601
Epoch 354/500
accuracy: 0.9991 - val_loss: 99.9162 - val_accuracy: 0.3269
Epoch 355/500
```

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accuracy: 0.9998 - val_loss: 71.6443 - val_accuracy: 0.3341
Epoch 356/500
accuracy: 0.9909 - val_loss: 24.4450 - val_accuracy: 0.4062
Epoch 357/500
54/54 [============= ] - 17s 313ms/step - loss: 0.0023 -
accuracy: 0.9992 - val_loss: 30.2608 - val_accuracy: 0.5264
Epoch 358/500
54/54 [============= ] - 17s 321ms/step - loss: 1.9618e-04 -
accuracy: 1.0000 - val_loss: 6.1629 - val_accuracy: 0.6659
Epoch 359/500
accuracy: 0.9998 - val_loss: 0.0866 - val_accuracy: 0.9736
Epoch 360/500
accuracy: 0.9970 - val_loss: 0.1532 - val_accuracy: 0.9880
Epoch 361/500
accuracy: 0.9913 - val_loss: 0.5304 - val_accuracy: 0.8125
Epoch 362/500
accuracy: 0.9885 - val_loss: 211.8268 - val_accuracy: 0.3293
Epoch 363/500
accuracy: 0.9987 - val_loss: 290.0982 - val_accuracy: 0.3341
Epoch 364/500
accuracy: 1.0000 - val_loss: 217.3516 - val_accuracy: 0.3750
Epoch 365/500
accuracy: 0.9997 - val_loss: 82.2165 - val_accuracy: 0.6226
Epoch 366/500
accuracy: 1.0000 - val_loss: 140.3148 - val_accuracy: 0.6274
Epoch 367/500
accuracy: 0.9996 - val_loss: 0.1386 - val_accuracy: 0.9784
Epoch 368/500
accuracy: 0.9974 - val_loss: 109.3726 - val_accuracy: 0.0168
Epoch 369/500
accuracy: 1.0000 - val_loss: 40.7015 - val_accuracy: 0.2981
Epoch 370/500
accuracy: 1.0000 - val_loss: 2.4744 - val_accuracy: 0.6611
Epoch 371/500
```

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accuracy: 1.0000 - val_loss: 0.1195 - val_accuracy: 0.9832
Epoch 372/500
accuracy: 1.0000 - val loss: 0.0521 - val accuracy: 0.9928
Epoch 373/500
54/54 [============= ] - 19s 346ms/step - loss: 5.1356e-04 -
accuracy: 0.9995 - val_loss: 6.4172 - val_accuracy: 0.5096
Epoch 374/500
accuracy: 0.9996 - val_loss: 2.6064 - val_accuracy: 0.7139
Epoch 375/500
accuracy: 1.0000 - val_loss: 8.0773 - val_accuracy: 0.4952
Epoch 376/500
54/54 [============ ] - 19s 344ms/step - loss: 1.0185e-05 -
accuracy: 1.0000 - val_loss: 0.7807 - val_accuracy: 0.8990
Epoch 377/500
accuracy: 1.0000 - val_loss: 0.0219 - val_accuracy: 0.9952
Epoch 378/500
accuracy: 1.0000 - val_loss: 0.1891 - val_accuracy: 0.9976
Epoch 379/500
accuracy: 1.0000 - val_loss: 0.0659 - val_accuracy: 0.9928
Epoch 380/500
accuracy: 1.0000 - val_loss: 0.0457 - val_accuracy: 0.9952
Epoch 381/500
accuracy: 1.0000 - val_loss: 0.1370 - val_accuracy: 0.9976
Epoch 382/500
54/54 [============= ] - 19s 345ms/step - loss: 4.5496e-05 -
accuracy: 1.0000 - val loss: 0.0405 - val accuracy: 0.9952
Epoch 383/500
accuracy: 1.0000 - val_loss: 0.1401 - val_accuracy: 0.9928
Epoch 384/500
accuracy: 1.0000 - val_loss: 16.8042 - val_accuracy: 0.6851
Epoch 385/500
accuracy: 0.9981 - val_loss: 372.7028 - val_accuracy: 0.3413
Epoch 386/500
accuracy: 0.9998 - val_loss: 347.8006 - val_accuracy: 0.3293
Epoch 387/500
```

```
accuracy: 0.9998 - val_loss: 189.1129 - val_accuracy: 0.3654
Epoch 388/500
accuracy: 0.9992 - val_loss: 211.2339 - val_accuracy: 0.3341
Epoch 389/500
accuracy: 0.9974 - val_loss: 69.4624 - val_accuracy: 0.3341
Epoch 390/500
accuracy: 0.9956 - val_loss: 1.7959 - val_accuracy: 0.6899
Epoch 391/500
accuracy: 1.0000 - val_loss: 2.6708 - val_accuracy: 0.7837
Epoch 392/500
accuracy: 0.9994 - val_loss: 7.4775 - val_accuracy: 0.4135
Epoch 393/500
accuracy: 0.9999 - val_loss: 2.6664 - val_accuracy: 0.6370
Epoch 394/500
accuracy: 0.9990 - val_loss: 19.8276 - val_accuracy: 0.5481
Epoch 395/500
accuracy: 1.0000 - val_loss: 34.3325 - val_accuracy: 0.6538
Epoch 396/500
accuracy: 1.0000 - val_loss: 6.0294 - val_accuracy: 0.7019
Epoch 397/500
accuracy: 1.0000 - val_loss: 7.2987 - val_accuracy: 0.6947
Epoch 398/500
accuracy: 1.0000 - val loss: 143.3756 - val accuracy: 0.4712
Epoch 399/500
accuracy: 0.9977 - val_loss: 50.6277 - val_accuracy: 0.3389
Epoch 400/500
accuracy: 0.9940 - val_loss: 14.2041 - val_accuracy: 0.4736
Epoch 401/500
accuracy: 0.9989 - val_loss: 13.1102 - val_accuracy: 0.5889
Epoch 402/500
accuracy: 1.0000 - val_loss: 1.6152 - val_accuracy: 0.7572
Epoch 403/500
```

```
accuracy: 0.9926 - val_loss: 46.6000 - val_accuracy: 0.0529
Epoch 404/500
accuracy: 1.0000 - val_loss: 123.4665 - val_accuracy: 0.3269
Epoch 405/500
accuracy: 1.0000 - val_loss: 113.7809 - val_accuracy: 0.3293
Epoch 406/500
54/54 [============= ] - 18s 327ms/step - loss: 0.0175 -
accuracy: 0.9980 - val_loss: 4.7175 - val_accuracy: 0.5024
Epoch 407/500
accuracy: 0.9942 - val_loss: 81.9235 - val_accuracy: 0.3245
Epoch 408/500
54/54 [============= - - 18s 333ms/step - loss: 0.0129 -
accuracy: 0.9988 - val_loss: 1.2342 - val_accuracy: 0.7019
Epoch 409/500
54/54 [============ ] - 18s 338ms/step - loss: 4.2944e-04 -
accuracy: 1.0000 - val_loss: 0.0533 - val_accuracy: 0.9904
Epoch 410/500
accuracy: 1.0000 - val_loss: 0.0426 - val_accuracy: 0.9904
Epoch 411/500
accuracy: 0.9997 - val_loss: 9.9194 - val_accuracy: 0.4351
Epoch 412/500
accuracy: 1.0000 - val_loss: 0.1456 - val_accuracy: 0.9663
Epoch 413/500
accuracy: 1.0000 - val_loss: 0.0404 - val_accuracy: 0.9928
Epoch 414/500
accuracy: 0.9990 - val_loss: 58.7406 - val_accuracy: 0.3438
Epoch 415/500
accuracy: 0.9931 - val_loss: 54.6556 - val_accuracy: 0.3486
Epoch 416/500
accuracy: 0.9994 - val_loss: 87.3058 - val_accuracy: 0.3173
Epoch 417/500
accuracy: 0.9974 - val_loss: 71.8763 - val_accuracy: 0.3317
Epoch 418/500
accuracy: 0.9991 - val_loss: 5.5865 - val_accuracy: 0.5889
Epoch 419/500
```

```
accuracy: 0.9995 - val_loss: 3.0799 - val_accuracy: 0.7861
Epoch 420/500
accuracy: 1.0000 - val loss: 0.1304 - val accuracy: 0.9832
Epoch 421/500
accuracy: 1.0000 - val_loss: 0.0503 - val_accuracy: 0.9928
Epoch 422/500
54/54 [============= ] - 18s 336ms/step - loss: 1.7236e-04 -
accuracy: 1.0000 - val_loss: 0.0212 - val_accuracy: 0.9976
Epoch 423/500
accuracy: 0.9978 - val_loss: 2.4839 - val_accuracy: 0.6010
Epoch 424/500
accuracy: 1.0000 - val_loss: 0.1861 - val_accuracy: 0.9712
Epoch 425/500
54/54 [============ ] - 19s 342ms/step - loss: 1.0648e-04 -
accuracy: 1.0000 - val_loss: 0.0489 - val_accuracy: 0.9928
Epoch 426/500
accuracy: 1.0000 - val_loss: 0.1163 - val_accuracy: 0.9952
Epoch 427/500
accuracy: 0.9998 - val_loss: 0.0484 - val_accuracy: 0.9952
Epoch 428/500
accuracy: 1.0000 - val_loss: 0.0942 - val_accuracy: 0.9952
Epoch 429/500
54/54 [============ ] - 19s 343ms/step - loss: 2.7956e-04 -
accuracy: 1.0000 - val_loss: 0.2233 - val_accuracy: 0.9880
Epoch 430/500
accuracy: 1.0000 - val loss: 0.4814 - val accuracy: 0.9760
Epoch 431/500
accuracy: 1.0000 - val_loss: 0.1608 - val_accuracy: 0.9904
Epoch 432/500
accuracy: 1.0000 - val_loss: 0.3546 - val_accuracy: 0.9808
Epoch 433/500
accuracy: 1.0000 - val_loss: 0.1916 - val_accuracy: 0.9904
Epoch 434/500
accuracy: 1.0000 - val_loss: 0.0237 - val_accuracy: 0.9928
Epoch 435/500
```

```
accuracy: 1.0000 - val_loss: 0.0358 - val_accuracy: 0.9928
Epoch 436/500
54/54 [============= ] - 19s 347ms/step - loss: 2.9779e-06 -
accuracy: 1.0000 - val loss: 0.0283 - val accuracy: 0.9952
Epoch 437/500
accuracy: 1.0000 - val_loss: 80.7220 - val_accuracy: 0.4111
Epoch 438/500
accuracy: 0.9967 - val_loss: 3.4390 - val_accuracy: 0.3654
Epoch 439/500
accuracy: 0.9984 - val_loss: 0.6388 - val_accuracy: 0.7115
Epoch 440/500
accuracy: 0.9975 - val_loss: 90.4615 - val_accuracy: 0.2909
Epoch 441/500
accuracy: 0.9925 - val_loss: 3.4633 - val_accuracy: 0.6514
Epoch 442/500
accuracy: 0.9998 - val_loss: 128.8319 - val_accuracy: 0.3101
Epoch 443/500
accuracy: 1.0000 - val_loss: 136.9113 - val_accuracy: 0.3197
Epoch 444/500
accuracy: 1.0000 - val_loss: 0.7369 - val_accuracy: 0.9014
Epoch 445/500
54/54 [============= ] - 19s 347ms/step - loss: 0.0022 -
accuracy: 0.9987 - val_loss: 0.7469 - val_accuracy: 0.9135
Epoch 446/500
54/54 [============= ] - 19s 347ms/step - loss: 2.6374e-04 -
accuracy: 1.0000 - val_loss: 0.0244 - val_accuracy: 0.9976
Epoch 447/500
accuracy: 1.0000 - val_loss: 0.2677 - val_accuracy: 0.9928
Epoch 448/500
accuracy: 0.9993 - val_loss: 9.1079 - val_accuracy: 0.6827
Epoch 449/500
accuracy: 0.9986 - val_loss: 2.8010 - val_accuracy: 0.5264
Epoch 450/500
accuracy: 0.9952 - val_loss: 6.5534 - val_accuracy: 0.5312
Epoch 451/500
```

```
accuracy: 0.9997 - val_loss: 31.0955 - val_accuracy: 0.3654
Epoch 452/500
accuracy: 0.9997 - val_loss: 4.4112 - val_accuracy: 0.5793
Epoch 453/500
54/54 [============= ] - 18s 339ms/step - loss: 0.0107 -
accuracy: 0.9987 - val_loss: 2.6460 - val_accuracy: 0.7861
Epoch 454/500
54/54 [============= ] - 19s 343ms/step - loss: 1.2096e-05 -
accuracy: 1.0000 - val_loss: 0.7585 - val_accuracy: 0.9615
Epoch 455/500
accuracy: 1.0000 - val_loss: 0.2753 - val_accuracy: 0.9736
Epoch 456/500
accuracy: 1.0000 - val_loss: 0.0941 - val_accuracy: 0.9928
Epoch 457/500
54/54 [============= ] - 19s 344ms/step - loss: 3.7532e-06 -
accuracy: 1.0000 - val_loss: 0.0726 - val_accuracy: 0.9952
Epoch 458/500
accuracy: 1.0000 - val_loss: 0.1273 - val_accuracy: 0.9952
Epoch 459/500
accuracy: 1.0000 - val_loss: 0.1001 - val_accuracy: 0.9976
Epoch 460/500
accuracy: 1.0000 - val_loss: 0.0105 - val_accuracy: 0.9976
Epoch 461/500
accuracy: 1.0000 - val_loss: 0.0560 - val_accuracy: 0.9952
Epoch 462/500
accuracy: 1.0000 - val loss: 0.5174 - val accuracy: 0.9591
Epoch 463/500
accuracy: 1.0000 - val_loss: 0.2515 - val_accuracy: 0.9880
Epoch 464/500
accuracy: 1.0000 - val_loss: 0.2195 - val_accuracy: 0.9928
Epoch 465/500
accuracy: 1.0000 - val_loss: 0.0154 - val_accuracy: 0.9952
Epoch 466/500
accuracy: 1.0000 - val_loss: 0.0587 - val_accuracy: 0.9952
Epoch 467/500
```

```
accuracy: 1.0000 - val_loss: 0.0106 - val_accuracy: 0.9976
Epoch 468/500
accuracy: 1.0000 - val loss: 0.0098 - val accuracy: 0.9976
Epoch 469/500
accuracy: 1.0000 - val_loss: 0.0167 - val_accuracy: 0.9952
Epoch 470/500
54/54 [============= ] - 18s 338ms/step - loss: 4.7907e-06 -
accuracy: 1.0000 - val_loss: 0.0859 - val_accuracy: 0.9976
Epoch 471/500
accuracy: 1.0000 - val_loss: 7.5850e-05 - val_accuracy: 1.0000
Epoch 472/500
accuracy: 1.0000 - val_loss: 0.0111 - val_accuracy: 0.9976
Epoch 473/500
54/54 [============= ] - 18s 342ms/step - loss: 7.5358e-09 -
accuracy: 1.0000 - val_loss: 0.0134 - val_accuracy: 0.9976
Epoch 474/500
accuracy: 1.0000 - val_loss: 0.0992 - val_accuracy: 0.9976
Epoch 475/500
accuracy: 1.0000 - val_loss: 0.0195 - val_accuracy: 0.9976
Epoch 476/500
accuracy: 1.0000 - val_loss: 0.0143 - val_accuracy: 0.9976
Epoch 477/500
54/54 [============ ] - 19s 346ms/step - loss: 4.5837e-08 -
accuracy: 1.0000 - val_loss: 0.0767 - val_accuracy: 0.9952
Epoch 478/500
accuracy: 0.9966 - val_loss: 291.7560 - val_accuracy: 0.3341
Epoch 479/500
accuracy: 0.9849 - val_loss: 30.8915 - val_accuracy: 0.5264
Epoch 480/500
accuracy: 0.9999 - val_loss: 40.7386 - val_accuracy: 0.5409
Epoch 481/500
54/54 [============ ] - 18s 337ms/step - loss: 0.0273 -
accuracy: 0.9922 - val_loss: 202.3821 - val_accuracy: 0.3317
Epoch 482/500
accuracy: 0.9994 - val_loss: 173.0067 - val_accuracy: 0.3365
Epoch 483/500
```

```
accuracy: 0.9990 - val_loss: 54.3070 - val_accuracy: 0.5144
Epoch 484/500
accuracy: 0.9798 - val_loss: 9.7696 - val_accuracy: 0.3678
Epoch 485/500
accuracy: 0.9970 - val_loss: 108.9033 - val_accuracy: 0.3317
Epoch 486/500
accuracy: 0.9978 - val_loss: 0.1429 - val_accuracy: 0.9808
Epoch 487/500
accuracy: 0.9999 - val_loss: 0.3336 - val_accuracy: 0.9375
Epoch 488/500
accuracy: 0.9994 - val_loss: 0.4299 - val_accuracy: 0.8630
Epoch 489/500
54/54 [============ ] - 19s 346ms/step - loss: 7.9616e-05 -
accuracy: 1.0000 - val_loss: 0.0784 - val_accuracy: 0.9784
Epoch 490/500
accuracy: 1.0000 - val_loss: 0.0399 - val_accuracy: 0.9928
Epoch 491/500
accuracy: 0.9997 - val_loss: 0.0375 - val_accuracy: 0.9952
Epoch 492/500
accuracy: 0.9973 - val_loss: 86.2996 - val_accuracy: 0.3197
Epoch 493/500
accuracy: 0.9976 - val_loss: 0.1907 - val_accuracy: 0.9760
Epoch 494/500
accuracy: 1.0000 - val loss: 0.0684 - val accuracy: 0.9904
Epoch 495/500
accuracy: 1.0000 - val_loss: 0.0583 - val_accuracy: 0.9952
Epoch 496/500
accuracy: 1.0000 - val_loss: 0.0173 - val_accuracy: 0.9952
Epoch 497/500
accuracy: 1.0000 - val_loss: 0.0401 - val_accuracy: 0.9928
Epoch 498/500
accuracy: 1.0000 - val_loss: 0.1800 - val_accuracy: 0.9952
Epoch 499/500
```

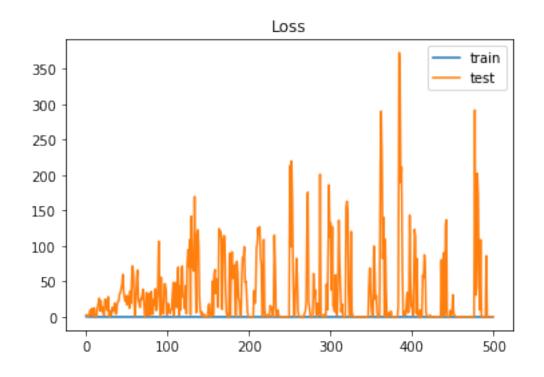
6 Acurácia

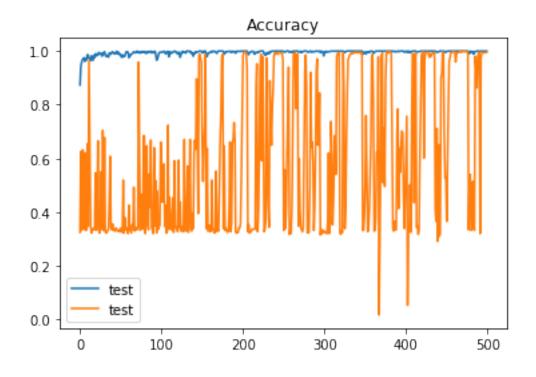
```
[8]: results = model.evaluate_generator(test_generator, 10) # Função para calcular au → acurácia
print('Acc: %.3f, Loss: %.3f' % (results[1], results[0]))

C:\ProgramData\Anaconda3\lib\site-
packages\tensorflow\python\keras\engine\training.py:1877: UserWarning:
`Model.evaluate_generator` is deprecated and will be removed in a future version. Please use `Model.evaluate`, which supports generators.
   warnings.warn('`Model.evaluate_generator` is deprecated and '
Acc: 0.994, Loss: 0.095
```

7 Curvas de Aprendizado do modelo

```
[9]: plt.title('Loss')
   plt.plot(history.history['loss'], label='train')
   plt.plot(history.history['val_loss'], label='test')
   plt.legend()
   plt.show()
   # Criando graficos para visualização dos resultados
   print()
   print()
   print()
   plt.title('Accuracy')
   plt.plot(history.history['accuracy'], label='test')
   plt.plot(history.history['val_accuracy'], label='test')
   plt.legend()
   plt.show()
```





8 Resultados Classificação e Matriz de Confusão

- Para obter os resultados do treinamento e usar o modelo treinado, precisa-se de fato realizar as previsões no dataset a fim de descobrir se o modelo está conseguindo prever de forma correta.
- Foi utilizada a função metrics.classification_report da biblioteca 'sklearn'.
- Dessa biblioteca, usamos os critérios de 'precision', 'recall' e 'f1-score' para obter as métricas de avaliação de resultados.
- Foram testadas 143 imagens da classe 'papel', 146 imagens da classe 'pedra' e 150 imagens da classe 'tesoura'
- Por fim, foi utilizado a biblioteca 'seaborn' para a apresentação da matriz de confusão, que é responsável por mostrar a quantidade de imagens previstas corretamente ou incorretamente.

```
[10]: | test_steps_per_epoch = np.math.ceil(test_generator.samples / test_generator.
       →batch_size) # Testes por época
      predictions = model.predict generator(test generator,
       →steps=test_steps_per_epoch) # Criação das predições a partir do, _
       →predict generator do modelo
      predicted_classes = np.argmax(predictions, axis=1)# classes de predição
      true_classes = test_generator.classes #Classes de predição
      class_labels = list(test_generator.class_indices.keys())
      report = classification_report(true_classes, predicted_classes,_
      →target_names=class_labels)
      print(report)
      matrix = confusion_matrix(true_classes,predicted_classes)
      df_cm = pd.DataFrame(matrix, index = [i for i in range(3)],
      columns = [i for i in range(3)])
      plt.figure(figsize = (10,7))
      sns.heatmap(df_cm, annot=True, linewidths=2.5)
```

C:\ProgramData\Anaconda3\lib\site-

packages\tensorflow\python\keras\engine\training.py:1905: UserWarning:

`Model.predict_generator` is deprecated and will be removed in a future version. Please use `Model.predict`, which supports generators.

warnings.warn('`Model.predict_generator` is deprecated and '

support
143
146
150
439
439
439

Matriz de Confusão:

