

# NHAN DIEN NGUOI

May 14, 2022

```
[ ]: # Sử dụng các thư viện sau để train mô hình
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import tensorflow as tf
from keras.utils import np_utils
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import SGD, RMSprop, Adam
from keras.callbacks import EarlyStopping
from sklearn.utils import validation
from sklearn import preprocessing
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import load_model
from tensorflow.keras.utils import load_img, img_to_array
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
[ ]: # Chuyển đến file Google Drive chứa bộ ảnh đã tải lên
%cd"/content/drive/MyDrive/Colab Notebooks/BT_AI/NHAN DIEN NGUOI/"
```

/content/drive/MyDrive/Colab Notebooks/BT\_AI/NHAN DIEN NGUOI

```
[ ]: # Tạo ra class ImageDataGenerator để chỉnh sửa ảnh
train_datagen=ImageDataGenerator(rescale=1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=True)
```

```
[ ]: # Tải bộ dữ liệu training set
training_set=train_datagen.flow_from_directory("/content/drive/MyDrive/Colab Notebooks/BT_AI/NHAN DIEN NGUOI/Training_set",target_size=(150,150),batch_size=32, class_mode='categorical')
```

Found 120 images belonging to 3 classes.

```
[ ]: # Tải bộ dữ liệu validation
```

```
validation=train_datagen.flow_from_directory("/content/drive/MyDrive/Colab_
↳Notebooks/BT_AI/NHAN DIEN NGUOI/Validation",target_size=(150,150),
↳batch_size=32, class_mode='categorical')
```

Found 18 images belonging to 3 classes.

```
[ ]: # Xem các nhãn có trong bộ dữ liệu training set
training_set.class_indices
```

```
[ ]: {'DUOC': 0, 'GIANG': 1, 'TRIEN': 2}
```

```
[ ]: # Xem các nhãn có trong bộ dữ liệu validation
validation.class_indices
```

```
[ ]: {'DUOC': 0, 'GIANG': 1, 'TRIEN': 2}
```

```
[ ]: # Tạo mô hình
model=Sequential()
# Tích chập 32 lần với mỗi lần là 3 hàng 3 cột
model.add(Conv2D(32,(3,3), activation='relu', kernel_initializer='he_uniform',
↳padding='same',input_shape=(150,150,3)))
model.add(Conv2D(32,(3,3), activation='relu', kernel_initializer='he_uniform',
↳padding='same'))
# Lấy phần tử lớn nhất ở trong 2 hàng và 2 cột
model.add(MaxPooling2D(2,2))
# Tích chập 64 lần với mỗi lần là 3 hàng 3 cột
model.add(Conv2D(64,(3,3), activation='relu', kernel_initializer='he_uniform',
↳padding='same'))
model.add(Conv2D(64,(3,3), activation='relu', kernel_initializer='he_uniform',
↳padding='same'))
# Lấy phần tử lớn nhất ở trong 2 hàng và 2 cột
model.add(MaxPooling2D(2,2))
# Tích chập 128 lần với mỗi lần là 3 hàng 3 cột
model.add(Conv2D(128,(3,3), activation='relu', kernel_initializer='he_uniform',
↳padding='same'))
model.add(Conv2D(128,(3,3), activation='relu', kernel_initializer='he_uniform',
↳padding='same'))
# Lấy phần tử lớn nhất ở trong 2 hàng và 2 cột
model.add(MaxPooling2D(2,2))
# Duỗi thẳng dữ liệu
model.add(Flatten())
# Tạo lớp ẩn thứ nhất với 128 tín hiệu ra
model.add(Dense(128,activation='relu',kernel_initializer='he_uniform'))
model.add(Dropout(0.2))
# Tạo lớp ẩn thứ hai với 3 tín hiệu ra
model.add(Dense(3,activation='softmax'))
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 150, 150, 32)	896
conv2d_1 (Conv2D)	(None, 150, 150, 32)	9248
max_pooling2d (MaxPooling2D)	(None, 75, 75, 32)	0
conv2d_2 (Conv2D)	(None, 75, 75, 64)	18496
conv2d_3 (Conv2D)	(None, 75, 75, 64)	36928
max_pooling2d_1 (MaxPooling2D)	(None, 37, 37, 64)	0
conv2d_4 (Conv2D)	(None, 37, 37, 128)	73856
conv2d_5 (Conv2D)	(None, 37, 37, 128)	147584
max_pooling2d_2 (MaxPooling2D)	(None, 18, 18, 128)	0
flatten (Flatten)	(None, 41472)	0
dense (Dense)	(None, 128)	5308544
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 3)	387

Total params: 5,595,939  
Trainable params: 5,595,939  
Non-trainable params: 0

```
[ ]: # Biên dịch mô hình
model.
    ↪ compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
# Train mô hình với 100 lần học (epochs), mỗi lần học 50 dữ liệu (batch_size),
    ↪ khi sai số không thay đổi trong 70 lần học thì sẽ dừng việc train
history=model.fit(training_set, epochs=500, batch_size=50, verbose=1,
    ↪ validation_data=validation, callbacks=[EarlyStopping(monitor='val_loss',
    ↪ patience=70)])
```

Epoch 1/500  
4/4 [=====] - 49s 10s/step - loss: 9.5841 - accuracy:  
0.3417 - val\_loss: 1.0349 - val\_accuracy: 0.3333  
Epoch 2/500  
4/4 [=====] - 3s 824ms/step - loss: 1.0482 - accuracy:  
0.4000 - val\_loss: 0.8978 - val\_accuracy: 0.5556  
Epoch 3/500  
4/4 [=====] - 3s 821ms/step - loss: 0.8057 - accuracy:  
0.6250 - val\_loss: 0.6096 - val\_accuracy: 0.7778  
Epoch 4/500  
4/4 [=====] - 3s 819ms/step - loss: 0.5387 - accuracy:  
0.7583 - val\_loss: 0.6315 - val\_accuracy: 0.7778  
Epoch 5/500  
4/4 [=====] - 3s 817ms/step - loss: 0.2423 - accuracy:  
0.9333 - val\_loss: 0.3073 - val\_accuracy: 0.9444  
Epoch 6/500  
4/4 [=====] - 3s 810ms/step - loss: 0.1199 - accuracy:  
0.9667 - val\_loss: 0.5202 - val\_accuracy: 0.8889  
Epoch 7/500  
4/4 [=====] - 3s 817ms/step - loss: 0.1303 - accuracy:  
0.9583 - val\_loss: 0.2891 - val\_accuracy: 0.8889  
Epoch 8/500  
4/4 [=====] - 3s 812ms/step - loss: 0.0763 - accuracy:  
0.9833 - val\_loss: 0.2223 - val\_accuracy: 0.9444  
Epoch 9/500  
4/4 [=====] - 3s 814ms/step - loss: 0.0474 - accuracy:  
0.9917 - val\_loss: 0.2098 - val\_accuracy: 0.9444  
Epoch 10/500  
4/4 [=====] - 3s 898ms/step - loss: 0.0418 - accuracy:  
0.9833 - val\_loss: 0.1330 - val\_accuracy: 0.9444  
Epoch 11/500  
4/4 [=====] - 3s 827ms/step - loss: 0.0288 - accuracy:  
0.9833 - val\_loss: 0.0753 - val\_accuracy: 0.9444  
Epoch 12/500  
4/4 [=====] - 3s 869ms/step - loss: 0.0454 - accuracy:  
0.9750 - val\_loss: 0.1005 - val\_accuracy: 0.9444  
Epoch 13/500  
4/4 [=====] - 3s 883ms/step - loss: 0.0267 - accuracy:  
0.9917 - val\_loss: 0.2709 - val\_accuracy: 0.8889  
Epoch 14/500  
4/4 [=====] - 3s 818ms/step - loss: 0.0104 - accuracy:  
1.0000 - val\_loss: 0.1374 - val\_accuracy: 0.9444  
Epoch 15/500  
4/4 [=====] - 3s 872ms/step - loss: 0.0063 - accuracy:  
1.0000 - val\_loss: 0.0984 - val\_accuracy: 0.9444  
Epoch 16/500  
4/4 [=====] - 3s 889ms/step - loss: 0.0207 - accuracy:  
0.9917 - val\_loss: 0.0695 - val\_accuracy: 0.9444

Epoch 17/500  
4/4 [=====] - 3s 807ms/step - loss: 0.0081 - accuracy: 1.0000 - val\_loss: 0.1124 - val\_accuracy: 0.9444  
Epoch 18/500  
4/4 [=====] - 3s 816ms/step - loss: 0.0033 - accuracy: 1.0000 - val\_loss: 0.0746 - val\_accuracy: 0.9444  
Epoch 19/500  
4/4 [=====] - 3s 811ms/step - loss: 2.8018e-04 - accuracy: 1.0000 - val\_loss: 0.0170 - val\_accuracy: 1.0000  
Epoch 20/500  
4/4 [=====] - 3s 813ms/step - loss: 4.0767e-04 - accuracy: 1.0000 - val\_loss: 0.0144 - val\_accuracy: 1.0000  
Epoch 21/500  
4/4 [=====] - 3s 791ms/step - loss: 0.0438 - accuracy: 0.9833 - val\_loss: 0.6166 - val\_accuracy: 0.8889  
Epoch 22/500  
4/4 [=====] - 3s 798ms/step - loss: 0.0165 - accuracy: 0.9917 - val\_loss: 0.1135 - val\_accuracy: 0.9444  
Epoch 23/500  
4/4 [=====] - 3s 878ms/step - loss: 0.1535 - accuracy: 0.9750 - val\_loss: 0.2592 - val\_accuracy: 0.8889  
Epoch 24/500  
4/4 [=====] - 3s 811ms/step - loss: 0.0576 - accuracy: 0.9833 - val\_loss: 0.2158 - val\_accuracy: 0.8889  
Epoch 25/500  
4/4 [=====] - 3s 823ms/step - loss: 0.0070 - accuracy: 1.0000 - val\_loss: 0.3083 - val\_accuracy: 0.8889  
Epoch 26/500  
4/4 [=====] - 3s 866ms/step - loss: 0.0139 - accuracy: 0.9917 - val\_loss: 0.0377 - val\_accuracy: 1.0000  
Epoch 27/500  
4/4 [=====] - 3s 879ms/step - loss: 0.0040 - accuracy: 1.0000 - val\_loss: 0.2242 - val\_accuracy: 0.9444  
Epoch 28/500  
4/4 [=====] - 3s 818ms/step - loss: 0.0082 - accuracy: 0.9917 - val\_loss: 0.0541 - val\_accuracy: 0.9444  
Epoch 29/500  
4/4 [=====] - 3s 867ms/step - loss: 0.0667 - accuracy: 0.9750 - val\_loss: 0.0038 - val\_accuracy: 1.0000  
Epoch 30/500  
4/4 [=====] - 3s 816ms/step - loss: 0.0490 - accuracy: 0.9750 - val\_loss: 0.3983 - val\_accuracy: 0.8889  
Epoch 31/500  
4/4 [=====] - 3s 799ms/step - loss: 0.0224 - accuracy: 0.9917 - val\_loss: 0.4612 - val\_accuracy: 0.8889  
Epoch 32/500  
4/4 [=====] - 3s 872ms/step - loss: 0.0127 - accuracy: 0.9917 - val\_loss: 0.0543 - val\_accuracy: 1.0000

Epoch 33/500  
4/4 [=====] - 3s 806ms/step - loss: 0.0020 - accuracy: 1.0000 - val\_loss: 0.0028 - val\_accuracy: 1.0000

Epoch 34/500  
4/4 [=====] - 3s 804ms/step - loss: 0.0061 - accuracy: 1.0000 - val\_loss: 0.0187 - val\_accuracy: 1.0000

Epoch 35/500  
4/4 [=====] - 3s 805ms/step - loss: 0.0019 - accuracy: 1.0000 - val\_loss: 0.0090 - val\_accuracy: 1.0000

Epoch 36/500  
4/4 [=====] - 3s 796ms/step - loss: 0.0026 - accuracy: 1.0000 - val\_loss: 0.1771 - val\_accuracy: 0.9444

Epoch 37/500  
4/4 [=====] - 3s 798ms/step - loss: 0.0024 - accuracy: 1.0000 - val\_loss: 0.0525 - val\_accuracy: 0.9444

Epoch 38/500  
4/4 [=====] - 3s 796ms/step - loss: 2.0562e-04 - accuracy: 1.0000 - val\_loss: 0.1917 - val\_accuracy: 0.9444

Epoch 39/500  
4/4 [=====] - 3s 821ms/step - loss: 1.9179e-04 - accuracy: 1.0000 - val\_loss: 0.0393 - val\_accuracy: 0.9444

Epoch 40/500  
4/4 [=====] - 3s 800ms/step - loss: 1.1046e-04 - accuracy: 1.0000 - val\_loss: 1.6600e-04 - val\_accuracy: 1.0000

Epoch 41/500  
4/4 [=====] - 3s 799ms/step - loss: 1.1412e-04 - accuracy: 1.0000 - val\_loss: 5.6217e-04 - val\_accuracy: 1.0000

Epoch 42/500  
4/4 [=====] - 3s 817ms/step - loss: 1.6413e-05 - accuracy: 1.0000 - val\_loss: 2.2651e-04 - val\_accuracy: 1.0000

Epoch 43/500  
4/4 [=====] - 3s 811ms/step - loss: 2.5798e-05 - accuracy: 1.0000 - val\_loss: 0.0287 - val\_accuracy: 1.0000

Epoch 44/500  
4/4 [=====] - 3s 805ms/step - loss: 5.1364e-06 - accuracy: 1.0000 - val\_loss: 2.6498e-04 - val\_accuracy: 1.0000

Epoch 45/500  
4/4 [=====] - 3s 808ms/step - loss: 3.5781e-05 - accuracy: 1.0000 - val\_loss: 0.1520 - val\_accuracy: 0.9444

Epoch 46/500  
4/4 [=====] - 3s 807ms/step - loss: 3.3779e-05 - accuracy: 1.0000 - val\_loss: 2.2514e-04 - val\_accuracy: 1.0000

Epoch 47/500  
4/4 [=====] - 3s 902ms/step - loss: 1.6809e-05 - accuracy: 1.0000 - val\_loss: 1.9621e-04 - val\_accuracy: 1.0000

Epoch 48/500  
4/4 [=====] - 3s 884ms/step - loss: 2.3096e-05 - accuracy: 1.0000 - val\_loss: 0.0015 - val\_accuracy: 1.0000

Epoch 49/500  
4/4 [=====] - 3s 820ms/step - loss: 0.0092 - accuracy: 0.9917 - val\_loss: 0.0014 - val\_accuracy: 1.0000

Epoch 50/500  
4/4 [=====] - 3s 816ms/step - loss: 0.0028 - accuracy: 1.0000 - val\_loss: 0.5138 - val\_accuracy: 0.9444

Epoch 51/500  
4/4 [=====] - 3s 812ms/step - loss: 0.0360 - accuracy: 0.9667 - val\_loss: 0.0084 - val\_accuracy: 1.0000

Epoch 52/500  
4/4 [=====] - 3s 818ms/step - loss: 0.0033 - accuracy: 1.0000 - val\_loss: 0.1161 - val\_accuracy: 0.9444

Epoch 53/500  
4/4 [=====] - 3s 812ms/step - loss: 0.0280 - accuracy: 0.9833 - val\_loss: 0.0990 - val\_accuracy: 0.9444

Epoch 54/500  
4/4 [=====] - 3s 875ms/step - loss: 0.0048 - accuracy: 1.0000 - val\_loss: 0.0283 - val\_accuracy: 1.0000

Epoch 55/500  
4/4 [=====] - 3s 870ms/step - loss: 9.9329e-04 - accuracy: 1.0000 - val\_loss: 0.0054 - val\_accuracy: 1.0000

Epoch 56/500  
4/4 [=====] - 3s 815ms/step - loss: 0.0022 - accuracy: 1.0000 - val\_loss: 0.0672 - val\_accuracy: 0.9444

Epoch 57/500  
4/4 [=====] - 3s 869ms/step - loss: 0.0054 - accuracy: 1.0000 - val\_loss: 0.0149 - val\_accuracy: 1.0000

Epoch 58/500  
4/4 [=====] - 3s 811ms/step - loss: 0.0019 - accuracy: 1.0000 - val\_loss: 0.1433 - val\_accuracy: 0.9444

Epoch 59/500  
4/4 [=====] - 3s 889ms/step - loss: 7.5553e-04 - accuracy: 1.0000 - val\_loss: 0.0563 - val\_accuracy: 0.9444

Epoch 60/500  
4/4 [=====] - 3s 818ms/step - loss: 4.5218e-05 - accuracy: 1.0000 - val\_loss: 0.0052 - val\_accuracy: 1.0000

Epoch 61/500  
4/4 [=====] - 3s 813ms/step - loss: 4.1851e-04 - accuracy: 1.0000 - val\_loss: 0.1324 - val\_accuracy: 0.9444

Epoch 62/500  
4/4 [=====] - 3s 818ms/step - loss: 5.2588e-04 - accuracy: 1.0000 - val\_loss: 0.0341 - val\_accuracy: 1.0000

Epoch 63/500  
4/4 [=====] - 3s 876ms/step - loss: 0.0010 - accuracy: 1.0000 - val\_loss: 0.0037 - val\_accuracy: 1.0000

Epoch 64/500  
4/4 [=====] - 3s 812ms/step - loss: 1.5767e-04 - accuracy: 1.0000 - val\_loss: 0.1316 - val\_accuracy: 0.9444

Epoch 65/500  
4/4 [=====] - 3s 825ms/step - loss: 1.0414e-04 - accuracy: 1.0000 - val\_loss: 6.6374e-04 - val\_accuracy: 1.0000

Epoch 66/500  
4/4 [=====] - 3s 802ms/step - loss: 4.7250e-04 - accuracy: 1.0000 - val\_loss: 0.0147 - val\_accuracy: 1.0000

Epoch 67/500  
4/4 [=====] - 3s 817ms/step - loss: 7.5292e-05 - accuracy: 1.0000 - val\_loss: 6.1791e-04 - val\_accuracy: 1.0000

Epoch 68/500  
4/4 [=====] - 3s 811ms/step - loss: 3.5288e-05 - accuracy: 1.0000 - val\_loss: 1.9992e-04 - val\_accuracy: 1.0000

Epoch 69/500  
4/4 [=====] - 3s 800ms/step - loss: 6.0052e-05 - accuracy: 1.0000 - val\_loss: 0.0450 - val\_accuracy: 0.9444

Epoch 70/500  
4/4 [=====] - 3s 877ms/step - loss: 5.7562e-06 - accuracy: 1.0000 - val\_loss: 1.4304e-04 - val\_accuracy: 1.0000

Epoch 71/500  
4/4 [=====] - 3s 807ms/step - loss: 4.7825e-05 - accuracy: 1.0000 - val\_loss: 0.0022 - val\_accuracy: 1.0000

Epoch 72/500  
4/4 [=====] - 3s 853ms/step - loss: 0.0014 - accuracy: 1.0000 - val\_loss: 0.0725 - val\_accuracy: 1.0000

Epoch 73/500  
4/4 [=====] - 3s 804ms/step - loss: 8.2541e-04 - accuracy: 1.0000 - val\_loss: 0.0938 - val\_accuracy: 0.9444

Epoch 74/500  
4/4 [=====] - 3s 870ms/step - loss: 0.0020 - accuracy: 1.0000 - val\_loss: 4.9517e-04 - val\_accuracy: 1.0000

Epoch 75/500  
4/4 [=====] - 3s 810ms/step - loss: 3.3351e-05 - accuracy: 1.0000 - val\_loss: 0.2909 - val\_accuracy: 0.9444

Epoch 76/500  
4/4 [=====] - 3s 807ms/step - loss: 9.8627e-06 - accuracy: 1.0000 - val\_loss: 0.2040 - val\_accuracy: 0.9444

Epoch 77/500  
4/4 [=====] - 3s 820ms/step - loss: 3.3212e-04 - accuracy: 1.0000 - val\_loss: 0.0113 - val\_accuracy: 1.0000

Epoch 78/500  
4/4 [=====] - 3s 874ms/step - loss: 6.8308e-05 - accuracy: 1.0000 - val\_loss: 0.0944 - val\_accuracy: 0.9444

Epoch 79/500  
4/4 [=====] - 3s 868ms/step - loss: 0.0013 - accuracy: 1.0000 - val\_loss: 1.4702e-06 - val\_accuracy: 1.0000

Epoch 80/500  
4/4 [=====] - 3s 805ms/step - loss: 9.6757e-07 - accuracy: 1.0000 - val\_loss: 0.2314 - val\_accuracy: 0.9444



Epoch 81/500  
4/4 [=====] - 3s 878ms/step - loss: 0.0191 - accuracy: 0.9917 - val\_loss: 4.8001e-05 - val\_accuracy: 1.0000  
Epoch 82/500  
4/4 [=====] - 3s 801ms/step - loss: 6.3079e-04 - accuracy: 1.0000 - val\_loss: 0.4359 - val\_accuracy: 0.8889  
Epoch 83/500  
4/4 [=====] - 3s 816ms/step - loss: 0.2974 - accuracy: 0.9667 - val\_loss: 0.1011 - val\_accuracy: 0.9444  
Epoch 84/500  
4/4 [=====] - 3s 832ms/step - loss: 2.1551 - accuracy: 0.7750 - val\_loss: 0.0013 - val\_accuracy: 1.0000  
Epoch 85/500  
4/4 [=====] - 3s 806ms/step - loss: 0.3329 - accuracy: 0.9500 - val\_loss: 0.1940 - val\_accuracy: 0.8889  
Epoch 86/500  
4/4 [=====] - 3s 817ms/step - loss: 0.1488 - accuracy: 0.9667 - val\_loss: 0.0570 - val\_accuracy: 1.0000  
Epoch 87/500  
4/4 [=====] - 3s 830ms/step - loss: 0.0516 - accuracy: 0.9750 - val\_loss: 0.1202 - val\_accuracy: 0.9444  
Epoch 88/500  
4/4 [=====] - 3s 813ms/step - loss: 0.0335 - accuracy: 0.9833 - val\_loss: 0.1781 - val\_accuracy: 0.8889  
Epoch 89/500  
4/4 [=====] - 3s 824ms/step - loss: 0.0419 - accuracy: 0.9917 - val\_loss: 0.0850 - val\_accuracy: 0.9444  
Epoch 90/500  
4/4 [=====] - 3s 809ms/step - loss: 0.0280 - accuracy: 0.9833 - val\_loss: 0.0637 - val\_accuracy: 0.9444  
Epoch 91/500  
4/4 [=====] - 3s 863ms/step - loss: 0.0027 - accuracy: 1.0000 - val\_loss: 0.0606 - val\_accuracy: 0.9444  
Epoch 92/500  
4/4 [=====] - 3s 814ms/step - loss: 0.0018 - accuracy: 1.0000 - val\_loss: 0.0091 - val\_accuracy: 1.0000  
Epoch 93/500  
4/4 [=====] - 3s 813ms/step - loss: 0.0070 - accuracy: 1.0000 - val\_loss: 0.0052 - val\_accuracy: 1.0000  
Epoch 94/500  
4/4 [=====] - 3s 803ms/step - loss: 0.0094 - accuracy: 1.0000 - val\_loss: 0.0262 - val\_accuracy: 1.0000  
Epoch 95/500  
4/4 [=====] - 3s 816ms/step - loss: 0.0306 - accuracy: 0.9833 - val\_loss: 0.0028 - val\_accuracy: 1.0000  
Epoch 96/500  
4/4 [=====] - 3s 802ms/step - loss: 0.0147 - accuracy: 0.9917 - val\_loss: 0.2648 - val\_accuracy: 0.9444

Epoch 97/500  
4/4 [=====] - 3s 809ms/step - loss: 0.0051 - accuracy: 1.0000 - val\_loss: 0.8357 - val\_accuracy: 0.8889

Epoch 98/500  
4/4 [=====] - 3s 868ms/step - loss: 0.0218 - accuracy: 0.9917 - val\_loss: 0.1141 - val\_accuracy: 0.9444

Epoch 99/500  
4/4 [=====] - 3s 805ms/step - loss: 0.0978 - accuracy: 0.9917 - val\_loss: 0.0796 - val\_accuracy: 0.9444

Epoch 100/500  
4/4 [=====] - 3s 809ms/step - loss: 0.0159 - accuracy: 0.9917 - val\_loss: 0.2835 - val\_accuracy: 0.9444

Epoch 101/500  
4/4 [=====] - 3s 803ms/step - loss: 0.0035 - accuracy: 1.0000 - val\_loss: 0.6152 - val\_accuracy: 0.8889

Epoch 102/500  
4/4 [=====] - 3s 813ms/step - loss: 0.0035 - accuracy: 1.0000 - val\_loss: 0.2964 - val\_accuracy: 0.8889

Epoch 103/500  
4/4 [=====] - 3s 807ms/step - loss: 0.0050 - accuracy: 1.0000 - val\_loss: 0.2632 - val\_accuracy: 0.9444

Epoch 104/500  
4/4 [=====] - 3s 859ms/step - loss: 8.2262e-04 - accuracy: 1.0000 - val\_loss: 0.0175 - val\_accuracy: 1.0000

Epoch 105/500  
4/4 [=====] - 3s 800ms/step - loss: 0.0269 - accuracy: 0.9917 - val\_loss: 0.4828 - val\_accuracy: 0.9444

Epoch 106/500  
4/4 [=====] - 3s 885ms/step - loss: 0.1626 - accuracy: 0.9917 - val\_loss: 0.0077 - val\_accuracy: 1.0000

Epoch 107/500  
4/4 [=====] - 3s 803ms/step - loss: 0.0252 - accuracy: 0.9833 - val\_loss: 0.0059 - val\_accuracy: 1.0000

Epoch 108/500  
4/4 [=====] - 3s 797ms/step - loss: 0.0194 - accuracy: 0.9917 - val\_loss: 0.0361 - val\_accuracy: 1.0000

Epoch 109/500  
4/4 [=====] - 3s 808ms/step - loss: 0.0061 - accuracy: 1.0000 - val\_loss: 0.0668 - val\_accuracy: 0.9444

Epoch 110/500  
4/4 [=====] - 3s 799ms/step - loss: 0.0118 - accuracy: 1.0000 - val\_loss: 0.0197 - val\_accuracy: 1.0000

Epoch 111/500  
4/4 [=====] - 3s 803ms/step - loss: 0.0011 - accuracy: 1.0000 - val\_loss: 0.0827 - val\_accuracy: 0.9444

Epoch 112/500  
4/4 [=====] - 3s 805ms/step - loss: 8.9603e-04 - accuracy: 1.0000 - val\_loss: 8.5003e-04 - val\_accuracy: 1.0000

Epoch 113/500  
4/4 [=====] - 3s 862ms/step - loss: 5.8740e-05 - accuracy: 1.0000 - val\_loss: 1.0255e-04 - val\_accuracy: 1.0000  
Epoch 114/500  
4/4 [=====] - 3s 801ms/step - loss: 8.3505e-04 - accuracy: 1.0000 - val\_loss: 2.4915e-04 - val\_accuracy: 1.0000  
Epoch 115/500  
4/4 [=====] - 3s 811ms/step - loss: 0.0127 - accuracy: 0.9917 - val\_loss: 0.0147 - val\_accuracy: 1.0000  
Epoch 116/500  
4/4 [=====] - 3s 819ms/step - loss: 6.8556e-04 - accuracy: 1.0000 - val\_loss: 0.0017 - val\_accuracy: 1.0000  
Epoch 117/500  
4/4 [=====] - 3s 797ms/step - loss: 3.7433e-04 - accuracy: 1.0000 - val\_loss: 0.0048 - val\_accuracy: 1.0000  
Epoch 118/500  
4/4 [=====] - 3s 876ms/step - loss: 3.3961e-04 - accuracy: 1.0000 - val\_loss: 0.0030 - val\_accuracy: 1.0000  
Epoch 119/500  
4/4 [=====] - 3s 807ms/step - loss: 3.9029e-04 - accuracy: 1.0000 - val\_loss: 0.0012 - val\_accuracy: 1.0000  
Epoch 120/500  
4/4 [=====] - 3s 807ms/step - loss: 3.6425e-04 - accuracy: 1.0000 - val\_loss: 0.0021 - val\_accuracy: 1.0000  
Epoch 121/500  
4/4 [=====] - 3s 814ms/step - loss: 8.9153e-04 - accuracy: 1.0000 - val\_loss: 7.9630e-04 - val\_accuracy: 1.0000  
Epoch 122/500  
4/4 [=====] - 3s 810ms/step - loss: 2.6549e-04 - accuracy: 1.0000 - val\_loss: 0.0114 - val\_accuracy: 1.0000  
Epoch 123/500  
4/4 [=====] - 3s 798ms/step - loss: 3.5392e-04 - accuracy: 1.0000 - val\_loss: 0.0011 - val\_accuracy: 1.0000  
Epoch 124/500  
4/4 [=====] - 3s 806ms/step - loss: 1.5260e-04 - accuracy: 1.0000 - val\_loss: 4.8266e-04 - val\_accuracy: 1.0000  
Epoch 125/500  
4/4 [=====] - 3s 868ms/step - loss: 5.9919e-04 - accuracy: 1.0000 - val\_loss: 8.2175e-05 - val\_accuracy: 1.0000  
Epoch 126/500  
4/4 [=====] - 3s 785ms/step - loss: 1.8349e-04 - accuracy: 1.0000 - val\_loss: 0.0138 - val\_accuracy: 1.0000  
Epoch 127/500  
4/4 [=====] - 3s 861ms/step - loss: 1.1010e-04 - accuracy: 1.0000 - val\_loss: 0.0021 - val\_accuracy: 1.0000  
Epoch 128/500  
4/4 [=====] - 3s 804ms/step - loss: 2.0689e-05 - accuracy: 1.0000 - val\_loss: 1.2806e-04 - val\_accuracy: 1.0000

Epoch 129/500  
4/4 [=====] - 3s 806ms/step - loss: 1.3858e-04 - accuracy: 1.0000 - val\_loss: 0.0014 - val\_accuracy: 1.0000  
Epoch 130/500  
4/4 [=====] - 3s 866ms/step - loss: 5.5066e-04 - accuracy: 1.0000 - val\_loss: 0.0028 - val\_accuracy: 1.0000  
Epoch 131/500  
4/4 [=====] - 3s 860ms/step - loss: 4.2275e-04 - accuracy: 1.0000 - val\_loss: 1.2819e-04 - val\_accuracy: 1.0000  
Epoch 132/500  
4/4 [=====] - 3s 809ms/step - loss: 2.3354e-05 - accuracy: 1.0000 - val\_loss: 0.0027 - val\_accuracy: 1.0000  
Epoch 133/500  
4/4 [=====] - 3s 868ms/step - loss: 1.1476e-04 - accuracy: 1.0000 - val\_loss: 8.5050e-04 - val\_accuracy: 1.0000  
Epoch 134/500  
4/4 [=====] - 3s 797ms/step - loss: 6.5698e-05 - accuracy: 1.0000 - val\_loss: 0.1241 - val\_accuracy: 0.9444  
Epoch 135/500  
4/4 [=====] - 3s 800ms/step - loss: 1.7796e-04 - accuracy: 1.0000 - val\_loss: 5.7137e-04 - val\_accuracy: 1.0000  
Epoch 136/500  
4/4 [=====] - 3s 811ms/step - loss: 3.8592e-05 - accuracy: 1.0000 - val\_loss: 3.5235e-04 - val\_accuracy: 1.0000  
Epoch 137/500  
4/4 [=====] - 3s 853ms/step - loss: 8.4676e-04 - accuracy: 1.0000 - val\_loss: 0.0303 - val\_accuracy: 1.0000  
Epoch 138/500  
4/4 [=====] - 3s 806ms/step - loss: 6.7666e-05 - accuracy: 1.0000 - val\_loss: 0.0083 - val\_accuracy: 1.0000  
Epoch 139/500  
4/4 [=====] - 3s 823ms/step - loss: 2.6106e-05 - accuracy: 1.0000 - val\_loss: 7.3015e-05 - val\_accuracy: 1.0000  
Epoch 140/500  
4/4 [=====] - 3s 794ms/step - loss: 1.1433e-04 - accuracy: 1.0000 - val\_loss: 0.0037 - val\_accuracy: 1.0000  
Epoch 141/500  
4/4 [=====] - 3s 821ms/step - loss: 9.3830e-05 - accuracy: 1.0000 - val\_loss: 4.4389e-05 - val\_accuracy: 1.0000  
Epoch 142/500  
4/4 [=====] - 3s 805ms/step - loss: 1.1898e-05 - accuracy: 1.0000 - val\_loss: 1.4831e-04 - val\_accuracy: 1.0000  
Epoch 143/500  
4/4 [=====] - 3s 809ms/step - loss: 6.7843e-06 - accuracy: 1.0000 - val\_loss: 0.0049 - val\_accuracy: 1.0000  
Epoch 144/500  
4/4 [=====] - 3s 876ms/step - loss: 1.5747e-05 - accuracy: 1.0000 - val\_loss: 4.7308e-04 - val\_accuracy: 1.0000

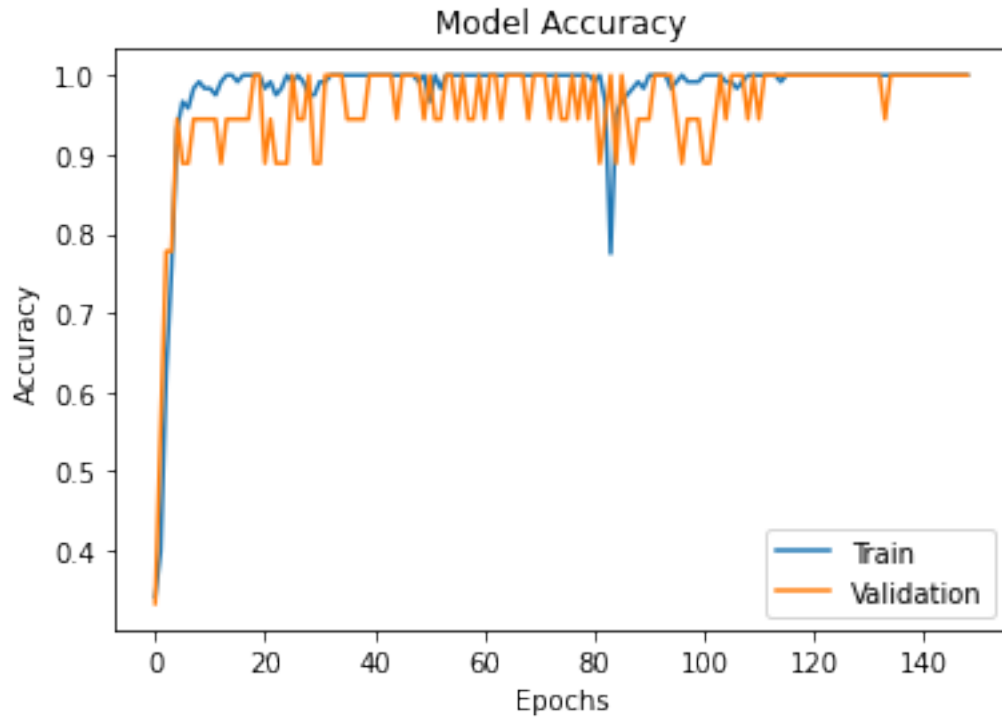
```
Epoch 145/500
4/4 [=====] - 3s 804ms/step - loss: 3.1685e-05 -
accuracy: 1.0000 - val_loss: 0.0041 - val_accuracy: 1.0000
Epoch 146/500
4/4 [=====] - 3s 799ms/step - loss: 1.2386e-05 -
accuracy: 1.0000 - val_loss: 1.0165e-04 - val_accuracy: 1.0000
Epoch 147/500
4/4 [=====] - 3s 871ms/step - loss: 1.2327e-05 -
accuracy: 1.0000 - val_loss: 1.1872e-04 - val_accuracy: 1.0000
Epoch 148/500
4/4 [=====] - 3s 863ms/step - loss: 8.8018e-06 -
accuracy: 1.0000 - val_loss: 9.8862e-04 - val_accuracy: 1.0000
Epoch 149/500
4/4 [=====] - 3s 798ms/step - loss: 9.5811e-06 -
accuracy: 1.0000 - val_loss: 3.9504e-04 - val_accuracy: 1.0000
```

```
[ ]: # Đánh giá độ chính xác của mô hình
Score=model.evaluate(training_set,verbose=0)
print('Train Loss', Score[0])
print('Train Accuracy', Score[1])
```

```
Train Loss 2.385093239354319e-06
Train Accuracy 1.0
```

```
[ ]: # Vẽ đồ thị giữa số lần học (Epochs) và độ chính xác (Accuracy)
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epochs')
plt.legend(['Train', 'Validation'])
plt.show
```

```
[ ]: <function matplotlib.pyplot.show>
```



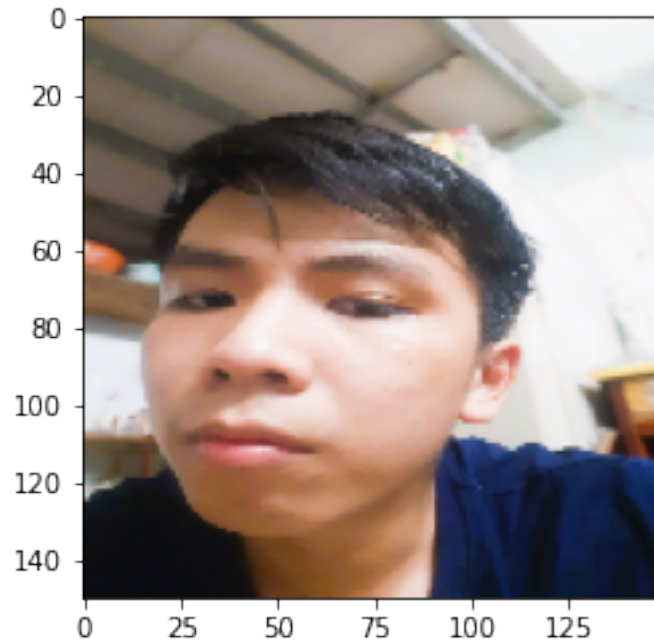
```
[ ]: # Lưu lại
model.save("NHAN DIEN NGUOI.h5")

[ ]: # Tải mô hình
model_CNN=load_model('NHAN DIEN NGUOI.h5')

[ ]: # Kiểm tra hình của từng bạn với bộ dữ liệu test_set

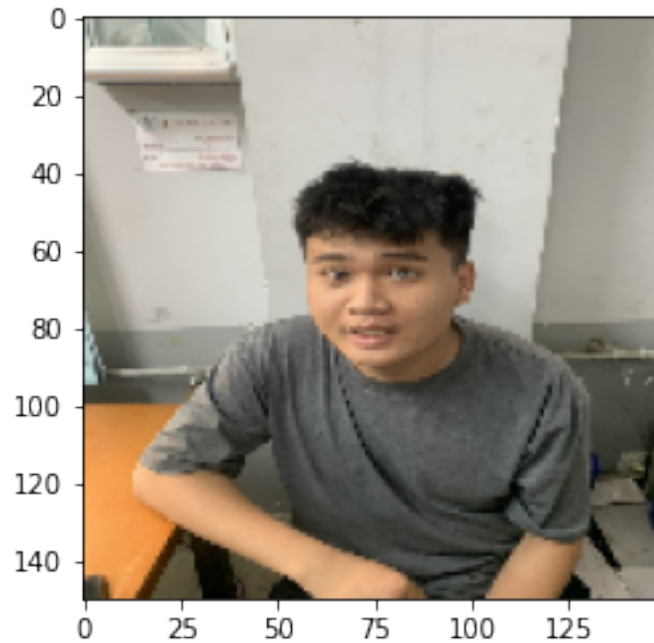
[ ]: img=load_img("/content/drive/MyDrive/Colab Notebooks/BT_AI/NHAN DIEN NGUOI/
↳Test_set/DuocT2.jpg",target_size=(150,150))
plt.imshow(img)
img=img_to_array(img)
img=img.astype('float32')
img=img/255
img=np.expand_dims(img,axis=0)
result=model_CNN.predict(img)
if round(result[0][0])==1:
    prediction='DUOC'
if round(result[0][1])==1:
    prediction='GIANG'
if round(result[0][2])==1:
    prediction='TRIEN'
print(prediction)
```

DUOC



```
[ ]: img=load_img("/content/drive/MyDrive/Colab Notebooks/BT_AI/NHAN DIEN NGUOI/
↳Test_set/GiangT1.jpg",target_size=(150,150))
plt.imshow(img)
img=img_to_array(img)
img=img.astype('float32')
img=img/255
img=np.expand_dims(img,axis=0)
result=model_CNN.predict(img)
if round(result[0][0])==1:
    prediction='DUOC'
if round(result[0][1])==1:
    prediction='GIANG'
if round(result[0][2])==1:
    prediction='TRIEU'
print(prediction)
```

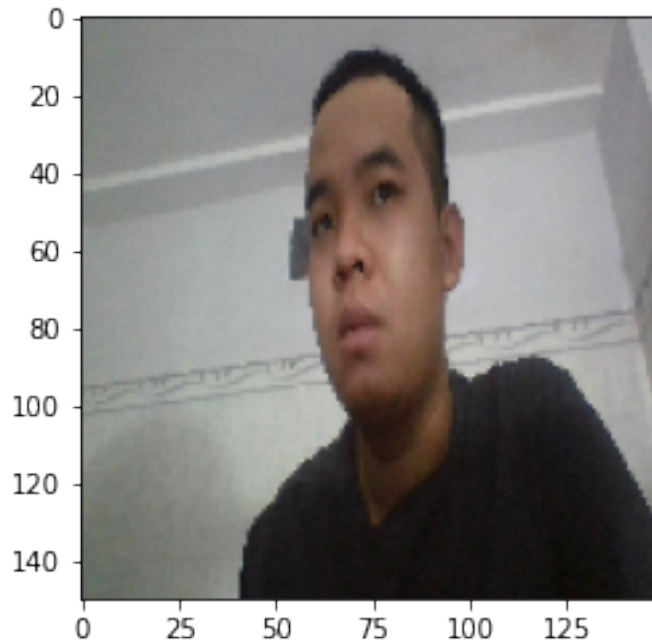
GIANG



```
[ ]: img=load_img("/content/drive/MyDrive/Colab Notebooks/BT_AI/NHAN DIEN NGUOI/
↪Test_set/TrienT3.jpg",target_size=(150,150))
plt.imshow(img)
img=img_to_array(img)
img=img.astype('float32')
img=img/255
img=np.expand_dims(img,axis=0)
result=model_CNN.predict(img)
if round(result[0][0])==1:
    prediction='DUOC'
if round(result[0][1])==1:
    prediction='GIANG'
if round(result[0][2])==1:
    prediction='TRIEN'
print(prediction)
```

TRIEN





```
[ ]: # Chuyển sang pdf
from google.colab import drive
drive.mount('/content/drive')
!sudo apt-get install texlive-xetex texlive-fonts-recommended
↳ texlive-generic-recommended
!wget -nc https://raw.githubusercontent.com/brpy/colab-pdf/master/colab_pdf.py
from colab_pdf import colab_pdf
colab_pdf('NHAN DIEN NGUOI.ipynb')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call `drive.mount("/content/drive", force_remount=True)`.

Reading package lists... Done

Building dependency tree

Reading state information... Done

The following packages were automatically installed and are no longer required:

libnvidia-common-460 nsight-compute-2020.2.0

Use 'sudo apt autoremove' to remove them.

The following additional packages will be installed:

fonts-droid-fallback fonts-lato fonts-lmodern fonts-noto-mono fonts-texgyre  
 javascript-common libcupsfilters1 libcupsimage2 libgs9 libgs9-common  
 libijs-0.35 libjbig2dec0 libjs-jquery libkpathsea6 libpotrace0 libptexenc1  
 libruby2.5 libsynchronet1 libtexlua52 libtexluajit2 libzip-0-13 lmodern  
 poppler-data preview-latex-style rake ruby ruby-did-you-mean ruby-minitest  
 ruby-net-telnet ruby-power-assert ruby-test-unit ruby2.5  
 rubygems-integration t1utils tex-common tex-gyre texlive-base