

TRƯỜNG ĐẠI HỌC SƯ PHẠM KỸ THUẬT THÀNH PHỐ HỒ CHÍ MINH

KHOA CƠ KHÍ CHẾ TẠO MÁY

BỘ MÔN CƠ ĐIỆN TỬ



**HCMUTE**

**BÀI TẬP**

# **TRÍ TUỆ NHÂN TẠO**

*Họ và Tên: Nguyễn Trọng Nhân*

*Mã số Sinh viên: 19146065*

*Lớp Trí Tuệ Nhân Tạo Sáng Thứ 7*

*Thành phố Hồ Chí Minh, tháng 5 năm 2022*

# I. SỬ DỤNG CIFAR100 NHẬN DIỆN TỪ DATASET CÓ SẴN

## 1. Code Tổng Quan

```
#Import libraries
import pandas as pd
from tensorflow.keras import datasets, layers, models
import matplotlib.pyplot as plt
from keras.models import Sequential
import numpy as np
from keras.utils import np_utils
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import SGD, Adam, RMSprop
from keras.callbacks import EarlyStopping

#Import dataset
(train_images, train_labels), (test_images, test_labels) = datasets.cifar100.load_data()

#Set data type
train_images = train_images.astype('float32')
test_images = test_images.astype('float32')
train_images /= 255
test_images /= 255
train_labels = np_utils.to_categorical(train_labels, 100)
test_labels = np_utils.to_categorical(test_labels, 100)

#Create model for train
model = Sequential()
model.add(Conv2D(input_shape=(32, 32, 3), kernel_initializer='he_uniform', kernel_size=(2, 2), padding='same', strides=(2, 2), filters=32))
model.add(MaxPooling2D(pool_size=(2, 2), strides=(1, 1), padding='same'))
model.add(Conv2D(kernel_size=(2, 2), padding='same', strides=(2, 2), filters=64))
model.add(MaxPooling2D(pool_size=(2, 2), strides=(1, 1), padding='same'))
model.add(Flatten())
model.add(Dense(512, activation='relu'))
model.add(Dense(256, activation='relu'))
model.add(Dense(100, activation='softmax'))
model.summary()

#Compile and Training
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['accuracy'])
history = model.fit(train_images, train_labels, batch_size=32, epochs=100, verbose=1, validation_data=(test_images, test_labels), callbacks=[EarlyStopping(monitor='val_loss', patience=10)])

#Draw plot and make evaluate
plt.plot(history.history['accuracy'])
plt.xlabel('epoch')
```

```

plt.legend(['accuracy'])
plt.show()

score = model.evaluate(test_images, test_labels, verbose=1)
print('Test error:', score[0])
print('Test accuracy: ', score[1])

#Make test and predict
n=int(input("Index ? "))
plt.imshow(test_images[n].reshape(32,32,3))
y_predict = model.predict(test_images[n].reshape(1,32,32,3))
print('Predicted value: ', np.argmax(y_predict))
print('Correct value: ', np.argmax(test_labels[n]))

```

## 2. Code trên Colab và kết quả của từng lệnh

+ Code
+ Text

RAM
Disk

Editing

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3s

```

#Import libraries
import pandas as pd
from tensorflow.keras import datasets, layers, models
import matplotlib.pyplot as plt
from keras.models import Sequential
import numpy as np
from keras.utils import np_utils
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import SGD, Adam, RMSprop
from keras.callbacks import EarlyStopping

```

[2]
8s

```

#Import dataset
(train_images, train_labels), (test_images, test_labels) = datasets.cifar100.load_data()

```

Download data from <https://www.cs.toronto.edu/~kriz/cifar-100-python.tar.gz>  
169009152/169001437 [=====] - 2s 0us/step  
169017344/169001437 [=====] - 2s 0us/step

[3]
7s

```

#Set data type
train_images=train_images.astype('float32')
test_images=test_images.astype('float32')
train_images/=255
test_images/=255
train_labels=np_utils.to_categorical(train_labels,100)
test_labels=np_utils.to_categorical(test_labels,100)

```

```

▶ #Create model for train
model = Sequential()
model.add(Conv2D(input_shape=(32, 32, 3), kernel_initializer='he_uniform', kernel_size=(2,2), padding='same',
model.add(MaxPooling2D(pool_size=(2,2), strides=(1,1), padding='same'))
model.add(Conv2D(kernel_size=(2,2), padding='same', strides=(2,2), filters=64))
model.add(MaxPooling2D(pool_size=(2, 2), strides=(1,1), padding='same'))
model.add(Flatten())
model.add(Dense(512, activation='relu'))
model.add(Dense(256, activation='relu'))
model.add(Dense(100, activation='softmax'))
model.summary()

```

Model: "sequential"

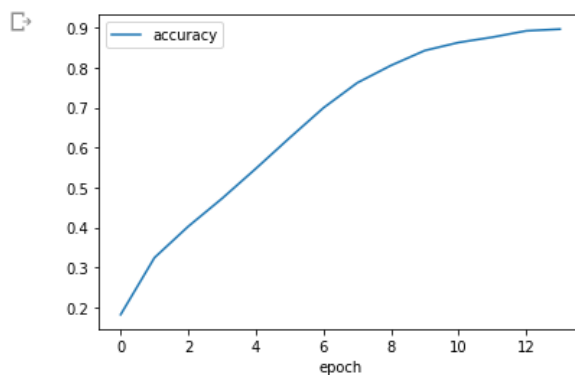
Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 16, 16, 32)	416
max_pooling2d (MaxPooling2D)	(None, 16, 16, 32)	0
conv2d_1 (Conv2D)	(None, 8, 8, 64)	8256
max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 64)	0
flatten (Flatten)	(None, 4096)	0
dense (Dense)	(None, 512)	2097664
dense_1 (Dense)	(None, 256)	131328
dense_2 (Dense)	(None, 100)	25700
=====		
Total params: 2,263,364		
Trainable params: 2,263,364		
Non-trainable params: 0		

```
#Compile and Training
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['accuracy'])
history=model.fit(train_images,train_labels,batch_size=32,epochs=100,verbose=1,validation_data=(test_images,
```

```
Epoch 1/100
1563/1563 [=====] - 19s 5ms/step - loss: 3.4659 - accuracy: 0.1819 - val_loss: 2.981
Epoch 2/100
1563/1563 [=====] - 7s 5ms/step - loss: 2.6691 - accuracy: 0.3242 - val_loss: 2.6056
Epoch 3/100
1563/1563 [=====] - 7s 4ms/step - loss: 2.2848 - accuracy: 0.4025 - val_loss: 2.5361
Epoch 4/100
1563/1563 [=====] - 7s 4ms/step - loss: 1.9549 - accuracy: 0.4724 - val_loss: 2.4883
Epoch 5/100
1563/1563 [=====] - 7s 5ms/step - loss: 1.6301 - accuracy: 0.5473 - val_loss: 2.6236
Epoch 6/100
1563/1563 [=====] - 7s 5ms/step - loss: 1.3161 - accuracy: 0.6243 - val_loss: 2.8521
Epoch 7/100
1563/1563 [=====] - 7s 4ms/step - loss: 1.0212 - accuracy: 0.6992 - val_loss: 3.1297
Epoch 8/100
1563/1563 [=====] - 7s 4ms/step - loss: 0.7883 - accuracy: 0.7619 - val_loss: 3.4431
Epoch 9/100
1563/1563 [=====] - 7s 5ms/step - loss: 0.6256 - accuracy: 0.8055 - val_loss: 3.9524
Epoch 10/100
1563/1563 [=====] - 7s 4ms/step - loss: 0.4997 - accuracy: 0.8428 - val_loss: 4.2755
Epoch 11/100
1563/1563 [=====] - 7s 4ms/step - loss: 0.4384 - accuracy: 0.8629 - val_loss: 4.7422
Epoch 12/100
1563/1563 [=====] - 7s 4ms/step - loss: 0.3974 - accuracy: 0.8761 - val_loss: 5.0034
Epoch 13/100
1563/1563 [=====] - 7s 5ms/step - loss: 0.3429 - accuracy: 0.8920 - val_loss: 5.4976
Epoch 14/100
1563/1563 [=====] - 7s 5ms/step - loss: 0.3291 - accuracy: 0.8963 - val_loss: 5.6835
```

```
#Draw plot and make evaluate
plt.plot(history.history['accuracy'])
plt.xlabel('epoch')
plt.legend(['accuracy'])
plt.show()

score = model.evaluate(test_images,test_labels, verbose=1)
print('Test error:',score[0])
print('Test accuracy: ',score[1])
```



```
313/313 [=====] - 1s 5ms/step - loss: 5.6835 - accuracy: 0.3614
Test error: 5.683497905731201
Test accuracy: 0.3614000082015991
```

```
#Make test and predict
n=int(input("Index: "))
plt.imshow(test_images[n].reshape(32,32,3))
y_predict = model.predict(test_images[n].reshape(1,32,32,3))
print('Predicted value: ', np.argmax(y_predict))
print('Correct value: ', np.argmax(test_labels[n]))
```

Index ? 21  
Predicted value: 48  
Correct value: 49



## II. SỬ DỤNG MNIST NHẬN DIỆN TỪ DATASET CÓ SẴN

### 1. Code Tổng Quan

```
#Import libraries
from keras.datasets import mnist
from sklearn import utils
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import RMSprop, Adam
import numpy as np
from keras.utils import np_utils
import pandas as pd
from keras.callbacks import EarlyStopping
import matplotlib.pyplot as plt

#Import dataset
(x_train, y_train), (x_test, y_test) = mnist.load_data()

#Set data type
x_train = x_train.astype('float32')
x_test = x_test.astype('float32')

x_train = x_train.reshape(60000, 28, 28, 1)
x_test = x_test.reshape(10000, 28, 28, 1)

x_train /= 255
x_test /= 255
```

```

y_train=np_utils.to_categorical(y_train,10)
y_test=np_utils.to_categorical(y_test,10)

#Create model for train
model = Sequential()
model.add(Conv2D(64, (3, 3), activation='relu', input_shape=(28, 28, 1)))
model.add(MaxPooling2D(2, 2))
model.add(Conv2D(64, (3, 3), input_shape=(28, 28), activation='relu'))
model.add(MaxPooling2D(2, 2))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(Dense(10, activation='softmax'))
model.summary()
#Compile and Training
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['a
ccuracy'])
history=model.fit(x_train,y_train,batch_size=128,epochs=500,verbose=1,valida
tion_data=(x_test,y_test),callbacks=[EarlyStopping(monitor='val_loss',patien
ce=20)])
#Draw plot and make evaluate
plt.plot(history.history['accuracy'])
plt.xlabel('epoch')
plt.legend(['accuracy'])
plt.show()

score = model.evaluate(x_test, y_test, verbose=0)
print('Test error:',score[0])
print('Test accuracy: ',score[1])

```

## 2. Code trên Colab và kết quả của từng lệnh

```
[1] #Import libraries
from keras.datasets import mnist
from sklearn import utils
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import RMSprop, Adam
import numpy as np
from keras.utils import np_utils
import pandas as pd
from keras.callbacks import EarlyStopping
import matplotlib.pyplot as plt
```

```
[2] #Import dataset
(x_train,y_train),(x_test,y_test)=mnist.load_data()
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz>  
11493376/11490434 [=====] - 0s 0us/step  
11501568/11490434 [=====] - 0s 0us/step

```
[3] #Set data type
x_train=x_train.astype('float32')
x_test=x_test.astype('float32')

x_train=x_train.reshape(60000,28,28,1)
x_test=x_test.reshape(10000,28,28,1)

x_train/=255
x_test/=255

y_train=np_utils.to_categorical(y_train,10)
y_test=np_utils.to_categorical(y_test,10)
```



```

#Create model for train
model = Sequential()
model.add(Conv2D(64,(3,3),activation='relu',input_shape=(28,28,1)))
model.add(MaxPooling2D(2,2))
model.add(Conv2D(64,(3,3),input_shape=(28,28),activation='relu'))
model.add(MaxPooling2D(2,2))
model.add(Flatten())
model.add(Dense(128,activation='relu'))
model.add(Dense(10,activation='softmax'))
model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 26, 26, 64)	640
max_pooling2d (MaxPooling2D)	(None, 13, 13, 64)	0
conv2d_1 (Conv2D)	(None, 11, 11, 64)	36928
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	0
flatten (Flatten)	(None, 1600)	0
dense (Dense)	(None, 128)	204928
dense_1 (Dense)	(None, 10)	1290
Total params: 243,786		
Trainable params: 243,786		
Non-trainable params: 0		

```

#Compile and Training
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['accuracy'])
history=model.fit(x_train,y_train,batch_size=128,epochs=500,verbose=1,validation_data=(x_test,y_test),callba

```

Epoch 1/500  
 469/469 [=====] - 85s 180ms/step - loss: 0.2003 - accuracy: 0.9428 - val\_loss: 0.071  
 Epoch 2/500  
 469/469 [=====] - 81s 172ms/step - loss: 0.0530 - accuracy: 0.9843 - val\_loss: 0.051  
 Epoch 3/500  
 469/469 [=====] - 88s 189ms/step - loss: 0.0373 - accuracy: 0.9881 - val\_loss: 0.031

```

469/469 [=====] - 81s 174ms/step - loss: 0.0039 - accuracy: 0.9992 - val_loss: 0.0436
Epoch 17/500
469/469 [=====] - 81s 174ms/step - loss: 0.0028 - accuracy: 0.9992 - val_loss: 0.0436
Epoch 18/500
469/469 [=====] - 82s 176ms/step - loss: 0.0057 - accuracy: 0.9980 - val_loss: 0.0364
Epoch 19/500
469/469 [=====] - 80s 171ms/step - loss: 0.0026 - accuracy: 0.9993 - val_loss: 0.0364
Epoch 20/500
469/469 [=====] - 81s 172ms/step - loss: 0.0029 - accuracy: 0.9990 - val_loss: 0.0436
Epoch 21/500
469/469 [=====] - 81s 174ms/step - loss: 0.0032 - accuracy: 0.9989 - val_loss: 0.0436
Epoch 22/500
469/469 [=====] - 80s 171ms/step - loss: 0.0011 - accuracy: 0.9996 - val_loss: 0.0364
Epoch 23/500
469/469 [=====] - 81s 172ms/step - loss: 0.0057 - accuracy: 0.9981 - val_loss: 0.0364
Epoch 24/500
469/469 [=====] - 81s 172ms/step - loss: 0.0018 - accuracy: 0.9994 - val_loss: 0.0364
Epoch 25/500
469/469 [=====] - 81s 172ms/step - loss: 0.0018 - accuracy: 0.9995 - val_loss: 0.0364

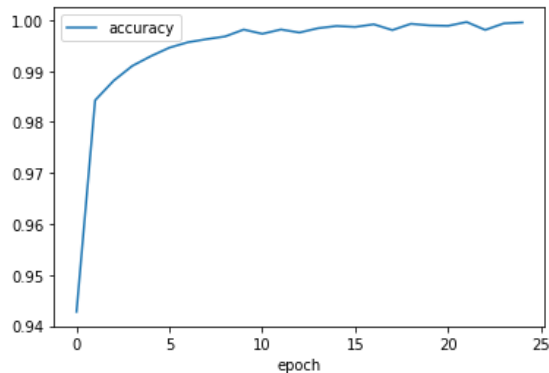
```

```

[6] #Draw plot and make evaluate
plt.plot(history.history['accuracy'])
plt.xlabel('epoch')
plt.legend(['accuracy'])
plt.show()

score = model.evaluate(x_test, y_test, verbose=0)
print('Test error:',score[0])
print('Test accuracy: ',score[1])

```



```

Test error: 0.03459727764129639
Test accuracy: 0.9936000108718872

```

```

✓ 10s [7] #Make test and predict
n=int(input("Index ? "))
print(x_test.shape)

y_predict = model.predict(x_test[n].reshape(1,28,28,1))
print('Predicted value: ', np.argmax(y_predict))
print('Correct value: ',np.argmax(y_test[n]))

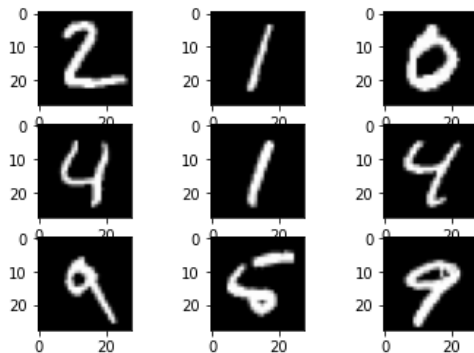
(x_tr,y_tr),(x_t,y_t)=mnist.load_data()
for i in range(1,10):
    plt.subplot(330+i)
    plt.imshow(x_t[i],cmap='gray')

```

```

Index ? 35
(10000, 28, 28, 1)
Predicted value: 2
Correct value: 2

```



### III. SỬ DỤNG MNIST NHẬN DIỆN FASHION TỪ DATASET CÓ SẴN

#### 1. Code Tổng Quan

```

#Import libraries
from keras.datasets import fashion_mnist
from sklearn import utils
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import RMSprop, Adam
import numpy as np
from keras.utils import np_utils
import pandas as pd
from keras.callbacks import EarlyStopping
import matplotlib.pyplot as plt

#Import dataset
(x_train,y_train),(x_test,y_test)=fashion_mnist.load_data()

#Set data type
x_train=x_train.astype('float32')

```

```

x_test=x_test.astype('float32')

x_train=x_train.reshape(60000,28,28,1)
x_test=x_test.reshape(10000,28,28,1)

x_train/=255
x_test/=255

y_train=np_utils.to_categorical(y_train,10)
y_test=np_utils.to_categorical(y_test,10)

#Create model for train
model = Sequential()
model.add(Conv2D(64, (3,3),activation='relu',input_shape=(28,28,1)))
model.add(MaxPooling2D(2,2))
model.add(Conv2D(64, (3,3),input_shape=(28,28),activation='relu'))
model.add(MaxPooling2D(2,2))
model.add(Flatten())
model.add(Dense(128,activation='relu'))
model.add(Dense(10, activation='softmax'))
model.summary()
#Compile and Training
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['accuracy'])
history=model.fit(x_train,y_train,batch_size=128,epochs=500,verbose=1,validation_data=(x_test,y_test),callbacks=[EarlyStopping(monitor='val_loss',patience=20)])
#Draw plot and make evaluate

plt.plot(history.history['accuracy'])
plt.xlabel('epoch')
plt.legend(['accuracy'])
plt.show()

score = model.evaluate(x_test, y_test, verbose=0)
print('Test error:',score[0])
print('Test accuracy: ',score[1])
#Make test and predict
n=int(input("Index ? "))
print(x_test.shape)

y_predict = model.predict(x_test[n].reshape(1,28,28,1))
print('Predicted value: ', np.argmax(y_predict))
print('Correct value: ',np.argmax(y_test[n]))



(x_tr,y_tr),(x_t,y_t)=fashion_mnist.load_data()
for i in range(1,10):

```

```
plt.subplot(330+i)
plt.imshow(x_t[i], cmap='gray')
```

## 2. Code trên Colab và kết quả của từng lệnh

+ Code + Text

RAM  Disk 

Editing

```
# Import libraries
from keras.datasets import fashion_mnist
from sklearn import utils
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Conv2D, MaxPooling2D, Flatten
from tensorflow.keras.optimizers import RMSprop, Adam
import numpy as np
from keras.utils import np_utils
import pandas as pd
from keras.callbacks import EarlyStopping
import matplotlib.pyplot as plt
```

```
[2] # Import dataset
(x_train, y_train), (x_test, y_test) = fashion_mnist.load_data()
```

Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-labels-idx1-ubyte.gz>  
32768/29515 [=====] - 0s 0us/step  
40960/29515 [=====] - 0s 0us/step  
Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/train-images-idx3-ubyte.gz>  
26427392/26421880 [=====] - 0s 0us/step  
26435584/26421880 [=====] - 0s 0us/step  
Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-labels-idx1-ubyte.gz>  
16384/5148 [=====]  
Downloading data from <https://storage.googleapis.com/tensorflow/tf-keras-datasets/t10k-images-idx3-ubyte.gz>  
4423680/4422102 [=====] - 0s 0us/step  
4431872/4422102 [=====] - 0s 0us/step

```
[3] # Set data type
x_train = x_train.astype('float32')
x_test = x_test.astype('float32')

x_train = x_train.reshape(60000, 28, 28, 1)
x_test = x_test.reshape(10000, 28, 28, 1)

x_train /= 255
x_test /= 255

y_train = np_utils.to_categorical(y_train, 10)
y_test = np_utils.to_categorical(y_test, 10)
```

```

▶ #Create model for train
model = Sequential()
model.add(Conv2D(64,(3,3),activation='relu',input_shape=(28,28,1)))
model.add(MaxPooling2D(2,2))
model.add(Conv2D(64,(3,3),input_shape=(28,28),activation='relu'))
model.add(MaxPooling2D(2,2))
model.add(Flatten())
model.add(Dense(128,activation='relu'))
model.add(Dense(10, activation='softmax'))
model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 26, 26, 64)	640
max_pooling2d (MaxPooling2D)	(None, 13, 13, 64)	0
conv2d_1 (Conv2D)	(None, 11, 11, 64)	36928
max_pooling2d_1 (MaxPooling2D)	(None, 5, 5, 64)	0
flatten (Flatten)	(None, 1600)	0
dense (Dense)	(None, 128)	204928
dense_1 (Dense)	(None, 10)	1290
=====		
Total params: 243,786		
Trainable params: 243,786		
Non-trainable params: 0		



### #Compile and Training

```
model.compile(optimizer='Adam', loss='categorical_crossentropy', metrics=['accuracy'])  
history=model.fit(x_train,y_train,batch_size=128,epochs=500,verbose=1,validation_data=(x_test,y_test),call
```



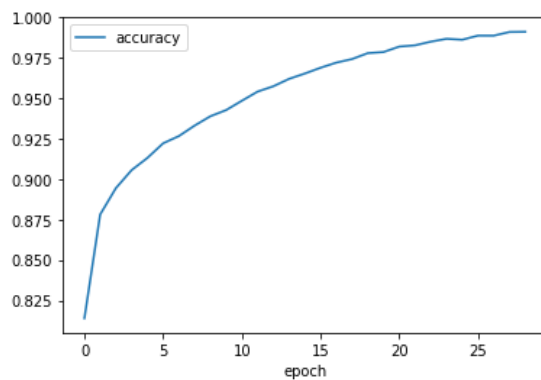
```
Epoch 1/500  
469/469 [=====] - 90s 189ms/step - loss: 0.5222 - accuracy: 0.8140 - val_loss: 0.1  
Epoch 2/500  
469/469 [=====] - 82s 176ms/step - loss: 0.3391 - accuracy: 0.8782 - val_loss: 0.1  
Epoch 3/500  
469/469 [=====] - 82s 175ms/step - loss: 0.2921 - accuracy: 0.8945 - val_loss: 0.1  
Epoch 4/500  
469/469 [=====] - 83s 178ms/step - loss: 0.2597 - accuracy: 0.9057 - val_loss: 0.1  
Epoch 5/500  
469/469 [=====] - 83s 176ms/step - loss: 0.2385 - accuracy: 0.9132 - val_loss: 0.1  
Epoch 6/500  
469/469 [=====] - 83s 177ms/step - loss: 0.2153 - accuracy: 0.9222 - val_loss: 0.1  
Epoch 7/500  
469/469 [=====] - 82s 174ms/step - loss: 0.1982 - accuracy: 0.9267 - val_loss: 0.1  
Epoch 8/500  
469/469 [=====] - 81s 172ms/step - loss: 0.1811 - accuracy: 0.9332 - val_loss: 0.1  
Epoch 9/500  
469/469 [=====] - 81s 173ms/step - loss: 0.1661 - accuracy: 0.9390 - val_loss: 0.1  
Epoch 10/500  
469/469 [=====] - 81s 172ms/step - loss: 0.1529 - accuracy: 0.9428 - val_loss: 0.1  
Epoch 11/500  
469/469 [=====] - 80s 171ms/step - loss: 0.1390 - accuracy: 0.9485 - val_loss: 0.1  
Epoch 12/500  
469/469 [=====] - 81s 173ms/step - loss: 0.1252 - accuracy: 0.9542 - val_loss: 0.1  
Epoch 13/500  
469/469 [=====] - 81s 173ms/step - loss: 0.1149 - accuracy: 0.9575 - val_loss: 0.1  
Epoch 14/500  
469/469 [=====] - 80s 171ms/step - loss: 0.1039 - accuracy: 0.9621 - val_loss: 0.1  
Epoch 15/500  
469/469 [=====] - 79s 169ms/step - loss: 0.0928 - accuracy: 0.9654 - val_loss: 0.1
```

```
✓ [5] Epoch 27/500
9m 469/469 [=====] - 80s 171ms/step - loss: 0.0298 - accuracy: 0.9888 - val_loss: 0.4
Epoch 28/500
469/469 [=====] - 80s 170ms/step - loss: 0.0256 - accuracy: 0.9911 - val_loss: 0.4
Epoch 29/500
469/469 [=====] - 80s 170ms/step - loss: 0.0245 - accuracy: 0.9912 - val_loss: 0.4
```

✓ 4s **#Draw plot and make evaluate**

```
plt.plot(history.history['accuracy'])
plt.xlabel('epoch')
plt.legend(['accuracy'])
plt.show()

score = model.evaluate(x_test, y_test, verbose=0)
print('Test error:',score[0])
print('Test accuracy: ',score[1])
```



```
Test error: 0.4804755747318268
Test accuracy: 0.9136999845504761
```



```
#Make test and predict
n=int(input("Index ? "))
print(x_test.shape)

y_predict = model.predict(x_test[n].reshape(1,28,28,1))
print('Predicted value: ', np.argmax(y_predict))
print('Correct value: ',np.argmax(y_test[n]))

(x_tr,y_tr),(x_t,y_t)=fashion_mnist.load_data()
for i in range(1,10):
    plt.subplot(330+i)
    plt.imshow(x_t[i],cmap='gray')
```

Index ? 52  
(10000, 28, 28, 1)  
Predicted value: 5  
Correct value: 5



## IV. NHẬN DIỆN TRÁI CÂY DÙNG CNN

### 1. Code Tổng Quan

```
#Import libraries
import glob
import cv2
import pandas as pd
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from random import randint
from keras.utils import np_utils
from keras.models import Sequential, load_model
from keras.layers import Dense, Flatten, Dropout, Conv2D, MaxPooling2D
from tensorflow.keras.optimizers import RMSprop, SGD, Adam
from tensorflow.keras.preprocessing.image import load_img, img_to_array
from tensorflow.keras.preprocessing.image import ImageDataGenerator
#Connect with my drive
from google.colab import drive
drive.mount('/content/drive')
#Get dataset from drive
```

```

train=ImageDataGenerator(rescale=1/255)
validation=ImageDataGenerator(rescale=1/255)
training_set=train.flow_from_directory('/content/drive/MyDrive/AI/FRUIT/train/train',target_size=(150,150), batch_size=40, class_mode='categorical')
validation_set=validation.flow_from_directory('/content/drive/MyDrive/AI/FRUIT/train/val',target_size=(150,150), batch_size=40, class_mode='categorical')
)
#Check label of dataset
training_set.class_indices
#Create Model for train
model = Sequential()
model.add(Conv2D(16, (3,3),padding='same',kernel_initializer='he_normal',input_shape=(150,150,3)))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(32, (3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(64, (3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())
model.add(Dense(512,activation='relu',kernel_initializer='he_normal'))
model.add(Dense(10,activation='softmax'))
model.summary()
#Training
opt = SGD(lr=0.01, momentum=0.9)
model.compile(optimizer=RMSprop(),loss='categorical_crossentropy', metrics=['accuracy'])
from keras.callbacks import EarlyStopping
history = model.fit(training_set, epochs = 10, validation_data = validation_set, verbose=1, callbacks=[EarlyStopping(monitor='val_loss', patience=15)])
#Save a file after train
model.save('CNN_FRUIT.h5')
#Draw plot and evaluate
score = model.evaluate(validation_set,verbose=0)
print('Test error: ',score[0])
print('Test accuracy: ',score[1])
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.ylabel('accuracy')
plt.xlabel('epochs')
plt.legend(['train','Validation'])
plt.show()
#Load trained file
CNN_FRUIT =load_model('CNN_FRUIT.h5')

```

```

#Check detect
img_path = '/content/drive/MyDrive/AI/FRUIT/test/52.jpg'
img=load_img(img_path,target_size=(150,150))
plt.imshow(img)
img=img_to_array(img)
img=img.reshape(1,150,150,3)
img=img.astype('float32')
img=img/255
Fruit=np.argmax(CNN_FRUIT.predict(img),axis=1)
pred = model.predict(img)
classes = ['passionfruit','peaches','pears','pineapples','plums','pomegranates','raspberries','strawberries','tomatoes','watermelons']
print(np.argmax(pred))
if Fruit==0: print(classes[0])
elif Fruit==1: print(classes[1])
elif Fruit==2: print(classes[2])
elif Fruit==3: print(classes[3])
elif Fruit==4: print(classes[4])
elif Fruit==5: print(classes[5])
elif Fruit==6: print(classes[6])
elif Fruit==7: print(classes[7])
elif Fruit==8: print(classes[8])
elif Fruit==9: print(classes[9])

```

## 2. Code trên Colab và kết quả của từng lệnh

```
# Import libraries
import glob
import cv2
import pandas as pd
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from random import randint
from keras.utils import np_utils
from keras.models import Sequential, load_model
from keras.layers import Dense, Flatten, Dropout, Conv2D, MaxPooling2D
from tensorflow.keras.optimizers import RMSprop, SGD, Adam
from tensorflow.keras.preprocessing.image import load_img, img_to_array
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
[ ] #Connect with my drive
    from google.colab import drive
    drive.mount('/content/drive')
```

Mounted at /content/drive

```
[ ]
```

```
drive/MyDrive/AI/FRUIT/train/train',target_size=(150,150), batch_size=40, class_mode='categorical')
content/drive/MyDrive/AI/FRUIT/train/val',target_size=(150,150), batch_size=40, class_mode='categorical')
```

```
Found 338 images belonging to 10 classes.
Found 338 images belonging to 10 classes.
```

```
[ ] #Check label of dataset
training_set.class_indices
```

```
{'passionfruit': 0,
 'peaches': 1,
 'pears': 2,
 'pineapples': 3,
 'plums': 4,
 'pomegranates': 5,
 'raspberries': 6,
 'strawberries': 7,
 'tomatoes': 8,
 'watermelons': 9}
```

```
[ ] #Create Model for train
model = Sequential()
model.add(Conv2D(16,(3,3),padding='same',kernel_initializer='he_normal',input_shape=(150,150,3)))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(32,(3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(64,(3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())
model.add(Dense(512,activation='relu',kernel_initializer='he_normal'))
model.add(Dense(10,activation='softmax'))
model.summary()
```

🔗 Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 150, 150, 16)	448
max_pooling2d (MaxPooling2D)	(None, 75, 75, 16)	0
conv2d_1 (Conv2D)	(None, 75, 75, 32)	4640
max_pooling2d_1 (MaxPooling2D)	(None, 37, 37, 32)	0

↑ ↓ 🔗 💬 ⚙️ 📄 🗑️ ⋮

▶

```
#Training
opt = SGD(lr=0.01, momentum=0.9)
model.compile(optimizer=RMSprop(),loss='categorical_crossentropy', metrics=['accuracy'])
from keras.callbacks import EarlyStopping
history = model.fit(training_set, epochs = 10, validation_data = validation_set, verbose=1, callbacks=[Early
```

🔗

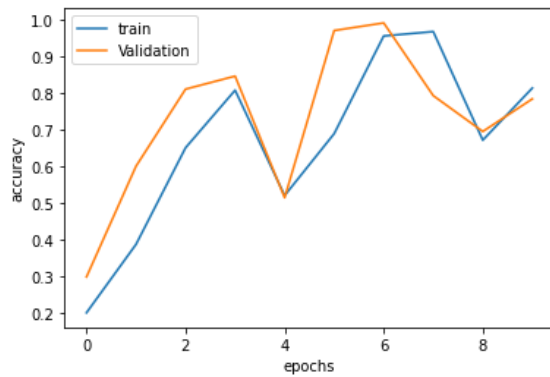
```
/usr/local/lib/python3.7/dist-packages/keras/optimizer_v2/gradient_descent.py:102: UserWarning: The `lr` argu
super(SGD, self).__init__(name, **kwargs)
Epoch 1/10
9/9 [=====] - 262s 29s/step - loss: 57.5984 - accuracy: 0.2012 - val_loss: 21.3964 - va
Epoch 2/10
9/9 [=====] - 21s 2s/step - loss: 12.2867 - accuracy: 0.3876 - val_loss: 2.3318 - va
Epoch 3/10
9/9 [=====] - 21s 2s/step - loss: 1.8595 - accuracy: 0.6509 - val_loss: 0.6878 - val
Epoch 4/10
9/9 [=====] - 21s 2s/step - loss: 0.8379 - accuracy: 0.8077 - val_loss: 0.4958 - val
Epoch 5/10
9/9 [=====] - 21s 2s/step - loss: 6.3797 - accuracy: 0.5207 - val_loss: 3.3969 - val
Epoch 6/10
9/9 [=====] - 21s 2s/step - loss: 2.2402 - accuracy: 0.6893 - val_loss: 0.1163 - val
Epoch 7/10
9/9 [=====] - 21s 2s/step - loss: 0.1391 - accuracy: 0.9556 - val_loss: 0.0451 - val
Epoch 8/10
9/9 [=====] - 21s 2s/step - loss: 0.0961 - accuracy: 0.9675 - val_loss: 1.0558 - val
Epoch 9/10
9/9 [=====] - 21s 2s/step - loss: 2.1547 - accuracy: 0.6716 - val_loss: 2.4518 - val
Epoch 10/10
9/9 [=====] - 20s 2s/step - loss: 1.0251 - accuracy: 0.8136 - val_loss: 0.8940 - val
```

[ ]

```
#Save a file after train
model.save('CNN_FRUIT.h5')
```

```
#Draw plot and evaluate
score = model.evaluate(validation_set,verbose=0)
print('Test error: ',score[0])
print('Test accuracy: ',score[1])
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.ylabel('accuracy')
plt.xlabel('epochs')
plt.legend(['train','Validation'])
plt.show()
```

Test error: 0.8940480351448059  
Test accuracy: 0.784023642539978

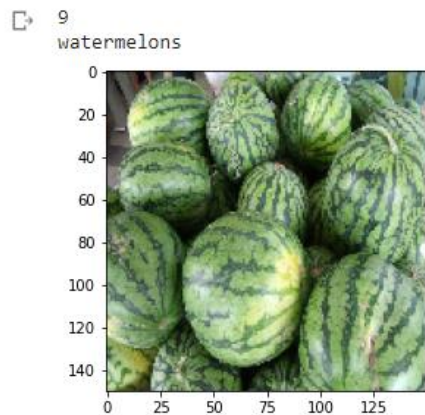


```
[ ] #Load trained file
CNN_FRUIT =load_model('CNN_FRUIT.h5')
```

```

#Check detect
img_path = './content/drive/MyDrive/AI/FRUIT/test/52.jpg'
img=load_img(img_path,target_size=(150,150))
plt.imshow(img)
img=img_to_array(img)
img=img.reshape(1,150,150,3)
img=img.astype('float32')
img=img/255
Fruit=np.argmax(CNN_FRUIT.predict(img),axis=1)
pred = model.predict(img)
classes = ['passionfruit', 'peaches', 'pears', 'pineapples', 'plums', 'pomegranates', 'raspberries', 'strawberries']
print(np.argmax(pred))
if Fruit==0: print(classes[0])
elif Fruit==1: print(classes[1])
elif Fruit==2: print(classes[2])
elif Fruit==3: print(classes[3])
elif Fruit==4: print(classes[4])
elif Fruit==5: print(classes[5])
elif Fruit==6: print(classes[6])
elif Fruit==7: print(classes[7])
elif Fruit==8: print(classes[8])
elif Fruit==9: print(classes[9])

```



## V. NHẬN DIỆN TIỀN DÙNG CNN

### 1. Code Tổng Quan

```

#Import libraries
import glob
import cv2
import pandas as pd
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from random import randint
from keras.utils import np_utils
from keras.models import Sequential, load_model
from keras.layers import Dense, Flatten, Dropout, Conv2D, MaxPooling2D
from tensorflow.keras.optimizers import RMSprop, SGD, Adam
from tensorflow.keras.preprocessing.image import load_img, img_to_array

```



```

from tensorflow.keras.preprocessing.image import ImageDataGenerator
#Connect with my drive
from google.colab import drive
drive.mount('/content/drive')
#Get dataset from drive
train=ImageDataGenerator(rescale=1/255)
validation=ImageDataGenerator(rescale=1/255)
training_set=train.flow_from_directory('/content/drive/MyDrive/AI/Tien/train
/train',target_size=(150,150), batch_size=40, class_mode='categorical')
validation_set=validation.flow_from_directory('/content/drive/MyDrive/AI/Tie
n/train/validation',target_size=(150,150), batch_size=40, class_mode='catego
rical')
#Check label of dataset
training_set.class_indices
#Create Model for train
model = Sequential()
model.add(Conv2D(16, (3,3),padding='same',kernel_initializer='he_normal',input_shape=(150,150,3)))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(32, (3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(64, (3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())
model.add(Dense(512,activation='relu',kernel_initializer='he_normal'))
model.add(Dense(6,activation='softmax'))
model.summary()
#Training
opt = SGD(lr=0.01, momentum=0.9)
model.compile(optimizer=RMSprop(),loss='categorical_crossentropy', metrics=
['accuracy'])
from keras.callbacks import EarlyStopping
history = model.fit(training_set, epochs = 10, validation_data = validation_
set, verbose=1, callbacks=[EarlyStopping(monitor='val_loss', patience=15)])
#Save a file after train
model.save('CNN_MONEY.h5')
#Draw plot and evaluate
score = model.evaluate(validation_set,verbose=0)
print('Test error: ',score[0])
print('Test accuracy: ',score[1])
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.ylabel('accuracy')

```

```

plt.xlabel('epochs')
plt.legend(['train', 'Validation'])
plt.show()
#Load trained file
CNN_money =load_model('CNN_MONEY.h5')
#Check detect
img_path = '/content/drive/MyDrive/AI/Tien/test/20k/20k (9).jpg'
img=load_img(img_path,target_size=(150,150))
plt.imshow(img)
img=img_to_array(img)
img=img.reshape(1,150,150,3)
img=img.astype('float32')
img=img/255
money=np.argmax(CNN_money.predict(img),axis=1)
pred = model.predict(img)
classes = ['01k', '02k', '05k', '10k', '20k', '50k']

print(np.argmax(pred))
if money==0: print(classes[0])
elif money==1: print(classes[1])
elif money==2: print(classes[2])
elif money==3: print(classes[3])
elif money==4: print(classes[4])

```

## 2. Code trên Colab và kết quả của từng lệnh

```
1s ▶ #Import libraries
import glob
import cv2
import pandas as pd
import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from random import randint
from keras.utils import np_utils
from keras.models import Sequential, load_model
from keras.layers import Dense, Flatten, Dropout, Conv2D, MaxPooling2D
from tensorflow.keras.optimizers import RMSprop, SGD, Adam
from tensorflow.keras.preprocessing.image import load_img, img_to_array
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
7s [2] #Connect with my drive
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
1s [3]
MyDrive/AI/Tien/train/train',target_size=(150,150), batch_size=40, class_mode='categorical')
/drive/MyDrive/AI/Tien/train/validation',target_size=(150,150), batch_size=40, class_mode='categorical')
```

Found 120 images belonging to 6 classes.  
Found 120 images belonging to 6 classes.

```
1s [4] #Check label of dataset
training_set.class_indices
```

```
{'01k': 0, '02k': 1, '05k': 2, '10k': 3, '20k': 4, '50k': 5}
```

Double-click (or enter) to edit

```

#Create Model for train
model = Sequential()
model.add(Conv2D(16,(3,3),padding='same',kernel_initializer='he_normal',input_shape=(150,150,3)))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(32,(3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Conv2D(64,(3,3),padding='same',kernel_initializer='he_normal'))
model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())
model.add(Dense(512,activation='relu',kernel_initializer='he_normal'))
model.add(Dense(6,activation='softmax'))
model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 150, 150, 16)	448
max_pooling2d (MaxPooling2D)	(None, 75, 75, 16)	0
conv2d_1 (Conv2D)	(None, 75, 75, 32)	4640
max_pooling2d_1 (MaxPooling2D)	(None, 37, 37, 32)	0
conv2d_2 (Conv2D)	(None, 37, 37, 64)	18496
max_pooling2d_2 (MaxPooling2D)	(None, 18, 18, 64)	0
flatten (Flatten)	(None, 20736)	0
dense (Dense)	(None, 512)	10617344
dense_1 (Dense)	(None, 6)	3078
=====		
Total params: 10,644,006		
Trainable params: 10,644,006		
Non-trainable params: 0		

```

▶ #Training
opt = SGD(lr=0.01, momentum=0.9)
model.compile(optimizer=RMSprop(), loss='categorical_crossentropy', metrics=['accuracy'])
from keras.callbacks import EarlyStopping
history = model.fit(training_set, epochs = 10, validation_data = validation_set, verbose=1, callbacks=[Early

```

↳ /usr/local/lib/python3.7/dist-packages/keras/optimizer\_v2/gradient\_descent.py:102: UserWarning: The `lr` argu  
super(SGD, self).\_\_init\_\_(name, \*\*kwargs)  
Epoch 1/10  
3/3 [=====] - 63s 24s/step - loss: 87.0198 - accuracy: 0.1250 - val\_loss: 225.6956 -  
Epoch 2/10  
3/3 [=====] - 5s 2s/step - loss: 139.7541 - accuracy: 0.2500 - val\_loss: 80.0438 - \
Epoch 3/10  
3/3 [=====] - 5s 2s/step - loss: 55.6856 - accuracy: 0.1417 - val\_loss: 19.5561 - va  
Epoch 4/10  
3/3 [=====] - 5s 2s/step - loss: 17.7833 - accuracy: 0.2333 - val\_loss: 4.8543 - val  
Epoch 5/10  
3/3 [=====] - 5s 2s/step - loss: 2.5973 - accuracy: 0.5833 - val\_loss: 2.1795 - val\_  
Epoch 6/10  
3/3 [=====] - 5s 2s/step - loss: 0.7957 - accuracy: 0.7917 - val\_loss: 1.2054 - val\_  
Epoch 7/10  
3/3 [=====] - 5s 2s/step - loss: 0.3386 - accuracy: 0.8917 - val\_loss: 0.9113 - val\_  
Epoch 8/10  
3/3 [=====] - 5s 2s/step - loss: 0.1427 - accuracy: 0.9750 - val\_loss: 0.9072 - val\_  
Epoch 9/10  
3/3 [=====] - 5s 2s/step - loss: 0.1201 - accuracy: 0.9833 - val\_loss: 0.7979 - val\_  
Epoch 10/10  
3/3 [=====] - 5s 2s/step - loss: 0.0503 - accuracy: 1.0000 - val\_loss: 0.7753 - val\_

```

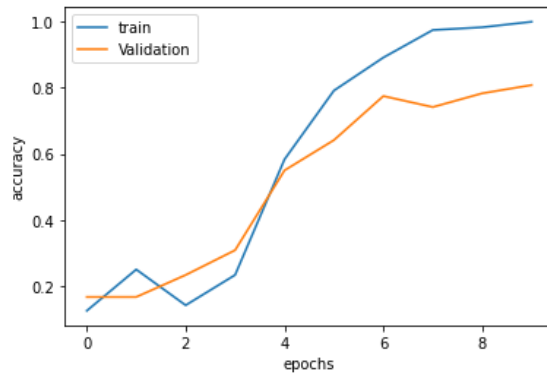
[7] #Save a file after train
model.save('CNN_MONEY.h5')

```

✓  
2s

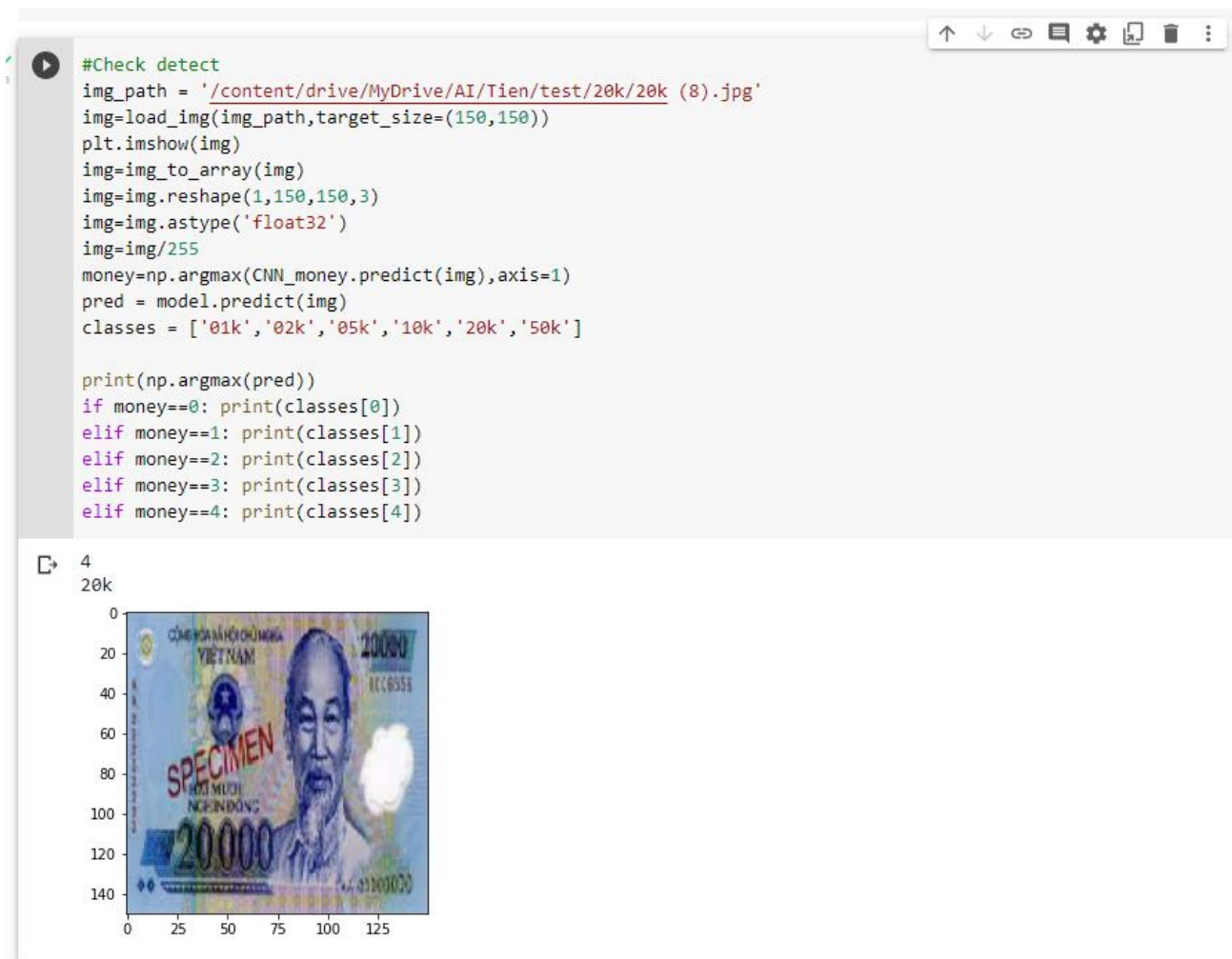
```
#Draw plot and evaluate
score = model.evaluate(validation_set,verbose=0)
print('Test error: ',score[0])
print('Test accuracy: ',score[1])
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.ylabel('accuracy')
plt.xlabel('epochs')
plt.legend(['train','Validation'])
plt.show()
```

Test error: 0.77530837059021  
Test accuracy: 0.8083333373069763



✓  
0s

```
[9] #Load trained file
CNN_money =load_model('CNN_MONEY.h5')
```



## VI. NHẬN DIỆN THỨC ẪN DÙNG CNN

### 1. Code Tổng Quan

```
#Import libraries
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Flatten, Conv2D, MaxPoolin
g2D
from keras.datasets import mnist, fashion_mnist, cifar10, cifar100
from tensorflow.keras.optimizers import Adam, RMSprop, SGD
from keras.utils import np_utils
import cv2
import matplotlib.pyplot as plt
import numpy as np
import tensorflow.compat.v2 as tf
from tensorflow import keras
import os
import pickle
```

```

import pandas as pd
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img_to_array
from keras.models import load_model
from keras.preprocessing import image
#Connect with my drive
from google.colab import drive
drive.mount('/content/drive')
#Get dataset from drive
train_path = '/content/drive/MyDrive/AI/MonAn'
path_img = []
labels = ['BanhMi', 'Banhcuon', 'Banhtrangnuong', 'BunBo', 'Bunthitnuong', 'Comtam', 'Goicuon', 'HuTieu', 'Miquang', 'Pho']
x_train = []
y_train = []
x_test = []
y_test = []
# Create dataset
for i in labels:
    path = os.path.join(train_path, i)
    index_label = labels.index(i)
    for j in os.listdir(path):
        img_path = os.path.join(path, j)
        img = image.load_img(img_path, target_size=(300,300)) # độ phân giải t
        ùy ae chọn nhé!
        img = img_to_array(img)
        img = img.reshape(300,300,3) #hàm reshape phải có cùng độ phân giải vớ
        i target_size của nhé
        img = img.astype('float32')
        img = img/255
        x_train.append(img)
        y_train.append(index_label)

# Process data
x_train = np.array(x_train)
y_train = np.array(y_train)
y_train = np_utils.to_categorical(y_train)
#Create Model for train
model = Sequential()
model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_unifo
rm', padding='same', input_shape=(300, 300, 3)))
model.add(MaxPooling2D(2,2))
model.add(Flatten())
model.add(Dense(256, activation='relu', kernel_initializer='he_uniform'))
model.add(Dense(10, activation='softmax'))
model.summary()
#Compile and Training

```



```

opt=SGD(learning_rate=0.01,momentum=0.9)
model.compile(loss='binary_crossentropy',optimizer=opt, metrics=['accuracy'])
)
history = model.fit(x_train,y_train,epochs = 20)
#Draw plot and make evaluate
plt.subplot(2,1,1)
plt.plot(history.history['loss'])
plt.title('Model Loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['train','validation'])
plt.show()

plt.subplot(2,1,2)
plt.plot(history.history['accuracy'])
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['train','validation'])
plt.show()
#Check detect
img = image.load_img('/content/Banhtrangnuong/Banhtrangnuong_8.jpg', target_size=(300,300))
plt.imshow(img)
img = img_to_array(img)
img = img.reshape(1,300,300,3)
img = img.astype('float32')
img = img/255
print('This is: '+ labels[np.argmax(model.predict(img))])

img1 = image.load_img('/content/HuTieu/HuTieu_7.jpg', target_size=(300,300))

plt.imshow(img1)
img1 = img_to_array(img1)
img1 = img1.reshape(1,300,300,3)
img1 = img1.astype('float32')
img1 = img1/255
print('Day la '+ labels[np.argmax(model.predict(img1))])

img2 = image.load_img('/content/BunBo/BunBo_9.jpg', target_size=(300,300))
plt.imshow(img2)
img2 = img_to_array(img2)
img2 = img2.reshape(1,300,300,3)
img2 = img2.astype('float32')
img2 = img2/255
print('Day la '+ labels[np.argmax(model.predict(img2))])

```

## 2. Code trên Colab và kết quả của từng lệnh

+ Code + Text

✓ RAM  
Disk

Editing



```
[1] #Import libraries
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout, Flatten, Conv2D, MaxPooling2D
from keras.datasets import mnist, fashion_mnist, cifar10, cifar100
from tensorflow.keras.optimizers import Adam, RMSprop, SGD
from keras.utils import np_utils
import cv2
import matplotlib.pyplot as plt
import numpy as np
import tensorflow.compat.v2 as tf
from tensorflow import keras
import os
import pickle
import pandas as pd
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img_to_array
from keras.models import load_model
from keras.preprocessing import image
```

```
#Connect with my drive
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
[3] #Get dataset from drive
train_path = '/content/drive/MyDrive/AI/MonAn'
path_img = []
labels = ['BanhMi', 'Banhcuon', 'Banhtrangnuong', 'BunBo', 'Bunhitnuong', 'Comtam', 'Goicuon', 'HuTieu', 'Miquang',
x_train = []
y_train = []
x_test = []
y_test =[]
# Create dataset
for i in labels:
    path = os.path.join(train_path,i)
    index_label = labels.index(i)
    for j in os.listdir(path):
        img_path = os.path.join(path, j)
        img = image.load_img(img_path, target_size=(300,300)) # độ phân giải tùy ae chọn nhé!
        img = img_to_array(img)
        img = img.reshape(300,300,3) #hàm reshape phải có cùng độ phân giải với target_size của nhé
        img = img.astype('float32')
        img = img/255
        x_train.append(img)
        y_train.append(index_label)

# Process data
x_train = np.array(x_train)
y_train = np.array(y_train)
y_train = np_utils.to_categorical(y_train)
```

```
[4] #Create Model for train
model = Sequential()
model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform', padding='same', input_shape=
model.add(MaxPooling2D(2,2))
model.add(Flatten())
model.add(Dense(256,activation='relu',kernel_initializer='he_uniform'))
model.add(Dense(10,activation='softmax'))
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 300, 300, 32)	896
max_pooling2d (MaxPooling2D)	(None, 150, 150, 32)	0



### #Compile and Training

```
opt=SGD(learning_rate=0.01,momentum=0.9)
model.compile(loss='binary_crossentropy',optimizer=opt, metrics=['accuracy'])
history = model.fit(x_train,y_train,epochs = 20)
```

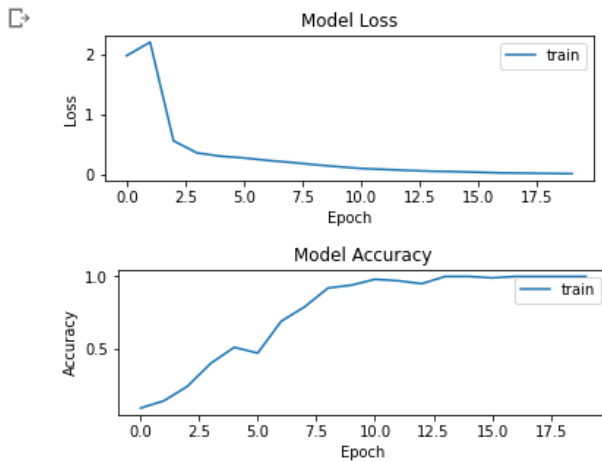


```
Epoch 1/20
4/4 [=====] - 12s 89ms/step - loss: 1.9712 - accuracy: 0.0900
Epoch 2/20
4/4 [=====] - 0s 67ms/step - loss: 2.1966 - accuracy: 0.1400
Epoch 3/20
4/4 [=====] - 0s 65ms/step - loss: 0.5580 - accuracy: 0.2400
Epoch 4/20
4/4 [=====] - 0s 66ms/step - loss: 0.3583 - accuracy: 0.4000
Epoch 5/20
4/4 [=====] - 0s 67ms/step - loss: 0.3034 - accuracy: 0.5100
Epoch 6/20
4/4 [=====] - 0s 66ms/step - loss: 0.2739 - accuracy: 0.4700
Epoch 7/20
4/4 [=====] - 0s 66ms/step - loss: 0.2339 - accuracy: 0.6900
Epoch 8/20
4/4 [=====] - 0s 67ms/step - loss: 0.2004 - accuracy: 0.7900
Epoch 9/20
4/4 [=====] - 0s 66ms/step - loss: 0.1629 - accuracy: 0.9200
Epoch 10/20
4/4 [=====] - 0s 67ms/step - loss: 0.1293 - accuracy: 0.9400
Epoch 11/20
4/4 [=====] - 0s 66ms/step - loss: 0.0997 - accuracy: 0.9800
Epoch 12/20
4/4 [=====] - 0s 66ms/step - loss: 0.0832 - accuracy: 0.9700
Epoch 13/20
4/4 [=====] - 0s 67ms/step - loss: 0.0672 - accuracy: 0.9500
Epoch 14/20
4/4 [=====] - 0s 66ms/step - loss: 0.0530 - accuracy: 1.0000
Epoch 15/20
4/4 [=====] - 0s 68ms/step - loss: 0.0474 - accuracy: 1.0000
Epoch 16/20
4/4 [=====] - 0s 66ms/step - loss: 0.0364 - accuracy: 0.9900
Epoch 17/20
4/4 [=====] - 0s 68ms/step - loss: 0.0251 - accuracy: 1.0000
Epoch 18/20
4/4 [=====] - 0s 69ms/step - loss: 0.0213 - accuracy: 1.0000
Epoch 19/20
4/4 [=====] - 0s 66ms/step - loss: 0.0164 - accuracy: 1.0000
Epoch 20/20
4/4 [=====] - 0s 64ms/step - loss: 0.0121 - accuracy: 1.0000
```

✓  
0s

```
#Draw plot and make evaluate
plt.subplot(2,1,1)
plt.plot(history.history['loss'])
plt.title('Model Loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['train','validation'])
plt.show()

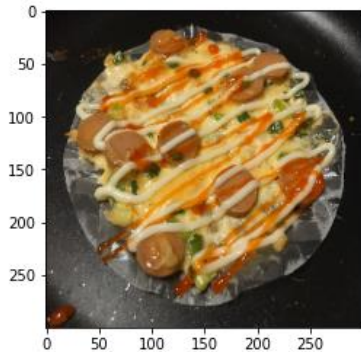
plt.subplot(2,1,2)
plt.plot(history.history['accuracy'])
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['train','validation'])
plt.show()
```



✓  
0s

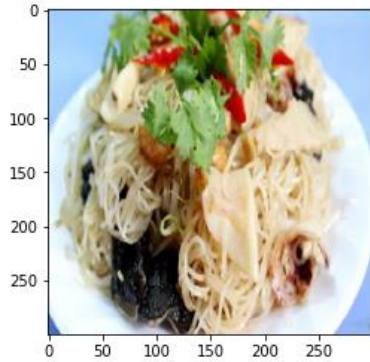
```
#Check detect
img = image.load_img('/content/drive/MyDrive/AI/MonAn/Banhtrangnuong/Banhtrangnuong_8.jpg', target_size=(300, 300))
plt.imshow(img)
img = img_to_array(img)
img = img.reshape(1,300,300,3)
img = img.astype('float32')
img = img/255
print('This is: ' + labels[np.argmax(model.predict(img))])
```

☞ This is: Banhtrangnuong



```
img1 = image.load_img('/content/drive/MyDrive/AI/MonAn/HuTieu/HuTieu_7.jpg', target_size=(300,300))
plt.imshow(img1)
img1 = img_to_array(img1)
img1 = img1.reshape(1,300,300,3)
img1 = img1.astype('float32')
img1 = img1/255
print('Day la ' + labels[np.argmax(model.predict(img1))])
```

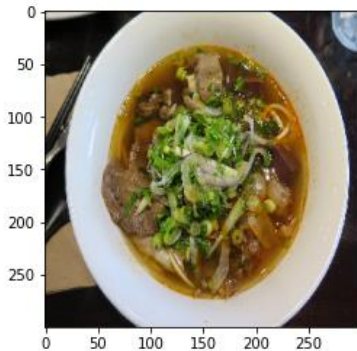
Day la HuTieu



✓  
1s

```
img2 = image.load_img('/content/drive/MyDrive/AI/MonAn/BunBo/BunBo_9.jpg', target_size=(300,300))
plt.imshow(img2)
img2 = img_to_array(img2)
img2 = img2.reshape(1,300,300,3)
img2 = img2.astype('float32')
img2 = img2/255
print('Day la ' + labels[np.argmax(model.predict(img2))])
```

Day la BunBo



## VII. GITHUB UPLOAD

[https://github.com/nhanguyene/HOMEWORK\\_AI\\_21\\_05\\_22](https://github.com/nhanguyene/HOMEWORK_AI_21_05_22)