

DEEP LEARNING MICRO PROJECT

Real time object detection using pre-trained imagenet models

DEFINITION:

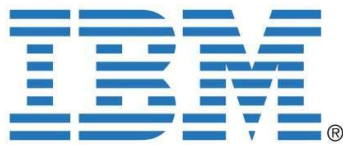
A convolutional neural network (CNN or convnet) is a subset of machine learning. It is one of the various types of artificial neural networks which are used for different applications and data types. A CNN is a kind of network architecture for deep learning algorithms and is specifically used for image recognition and tasks that involve the processing of pixel data.

There are other types of neural networks in deep learning, but for identifying and recognizing objects, CNNs are the network architecture of choice. This makes them highly suitable for computer vision (CV) tasks and for applications where object recognition is vital, such as self-driving cars and facial recognition.

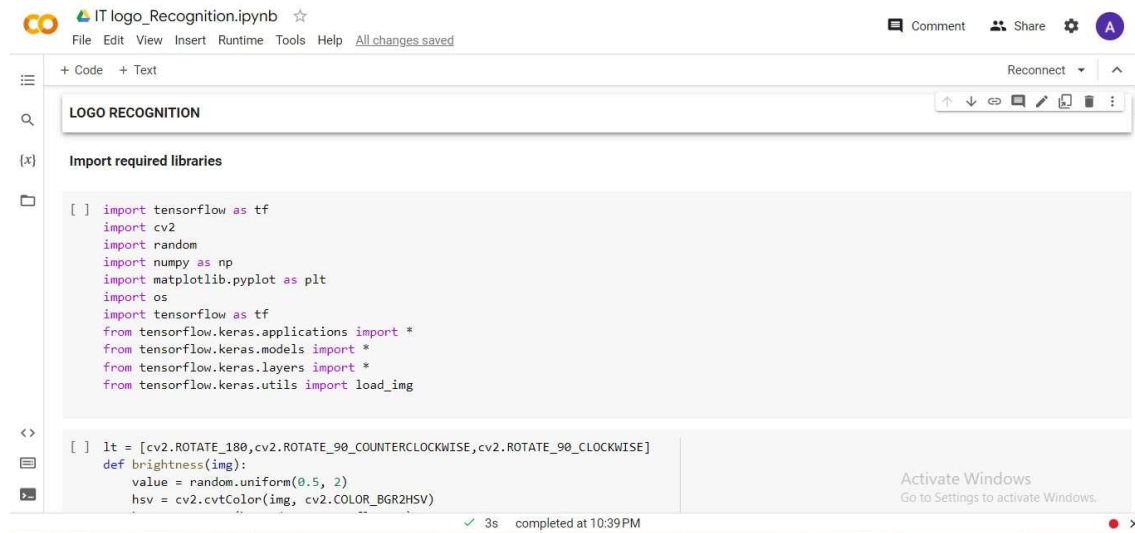
DATASET DESCRIPTION:

Data that we are using here is images. Images of Multinational company and IT company logos, which is in the format of jpg, jpeg, png.

For example:

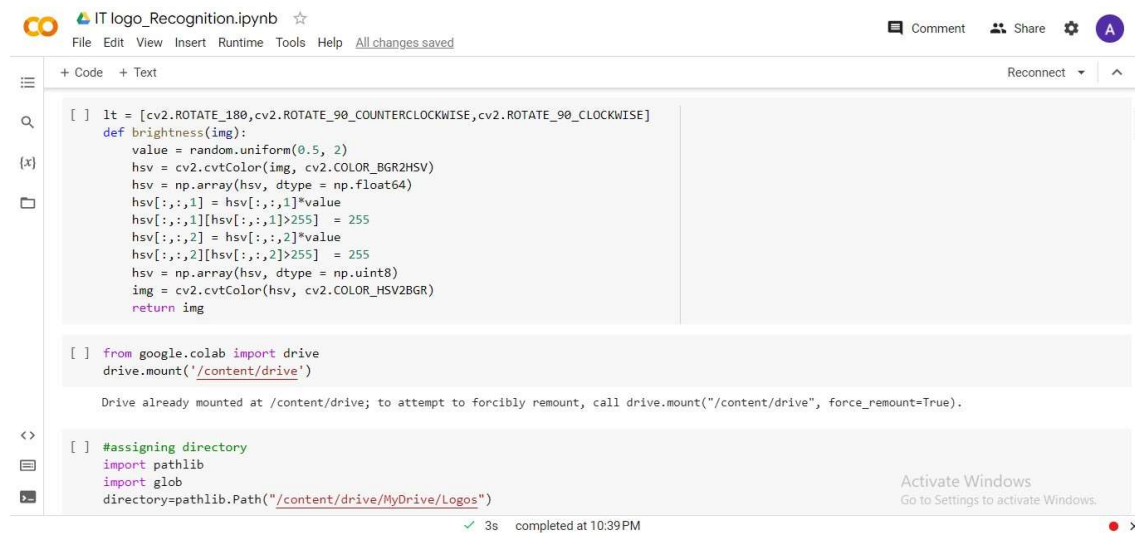


PROGRAM CODE:



```
[ ] import tensorflow as tf
import cv2
import random
import numpy as np
import matplotlib.pyplot as plt
import os
import tensorflow as tf
from tensorflow.keras.applications import *
from tensorflow.keras.models import *
from tensorflow.keras.layers import *
from tensorflow.keras.utils import load_img
```

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```
[ ] from google.colab import drive
drive.mount('/content/drive')

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

[ ] #assigning directory
import pathlib
import glob
directory=pathlib.Path("/content/drive/MyDrive/Logos")
```

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IT logo_Recognition.ipynb

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```
[ ] resultant="/content/augmentedimages"

[ ] items = os.listdir(directory)

classes=[]
count=0
images = []
labels=[]
for i in items:
    i1 = 0
    print(i)
    classes.append(i)
    path1 = f"{directory}/{i}"
    a = random.randint(5,10)
    img = cv2.imread(path1)
    img = cv2.resize(img,(224,224))
    k = i.split(".")[0]
    cv2.imwrite(f"{resultant}\\{k}{i1}.jpg",img)

    i1+=1
    while a!=0:
        img = cv2.rotate(img,lt[random.randint(0,2)])
        images.append(img)
```

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```
[ ] while a!=0:
    img = cv2.rotate(img,lt[random.randint(0,2)])
    images.append(img)
    cv2.imwrite(f"{resultant}\\{k}{i1}.jpg",img)
    i1+=1
    labels.append(count)
    if a%2==0:
        img = brightness(img)
        images.append(img)
        cv2.imwrite(f"{resultant}\\{k}{i1}.jpg",img)
        i1+=1
        labels.append(count)
    a-=1
    count+=1
images = np.array(images)
labels = np.array(labels)

Apple.png
Google.png
Dell.png
Accenture.png
IBM.jpg
Wipro.png
Microsoft.jpeg
Infosys.png
```

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```
[ ] images.shape

(77, 224, 224, 3)

[ ] # change the model here for alternatives

model = VGG16(weights="imagenet")
for i in model.layers:
    i.trainable = False

len(model.layers)

23

[ ] model.summary()

Model: "vgg16"
Layer (type) Output Shape Param #
-----
input_8 (InputLayer) [(None, 224, 224, 3)] 0
block1_conv1 (Conv2D) (None, 224, 224, 64) 1792
```

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```
[ ] block3_conv3 (Conv2D) (None, 56, 56, 256) 590080
block3_pool (MaxPooling2D) (None, 28, 28, 256) 0
block4_conv1 (Conv2D) (None, 28, 28, 512) 1180160
block4_conv2 (Conv2D) (None, 28, 28, 512) 2359808
block4_conv3 (Conv2D) (None, 28, 28, 512) 2359808
block4_pool (MaxPooling2D) (None, 14, 14, 512) 0
block5_conv1 (Conv2D) (None, 14, 14, 512) 2359808
block5_conv2 (Conv2D) (None, 14, 14, 512) 2359808
block5_conv3 (Conv2D) (None, 14, 14, 512) 2359808
flatten_7 (Flatten) (None, 100352) 0
dense_21 (Dense) (None, 512) 51380736
dense_22 (Dense) (None, 128) 65664
dense_23 (Dense) (None, 13) 1677

Total params: 66,162,765
Trainable params: 51,448,077
Non-trainable params: 14,714,688
```

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```
[ ] import tensorflow as tf
class myCallback(tf.keras.callbacks.Callback):
    def on_epoch_end(self, epoch, logs={}):
        print("call")
        if(logs.get('accuracy') > .99):
            print("\nReached %2.2f%% accuracy, so stopping training!!" %(99))
            self.model.stop_training = True
callbacks = myCallback()

[ ] # model1.summary()
model2.compile(optimizer="adam",loss="sparse_categorical_crossentropy",metrics=["accuracy"])
model2.fit(images,labels,epochs=10,callbacks=[callbacks])

Epoch 1/10
3/3 [=====] - ETA: 0s - loss: 7.1854 - accuracy: 0.4935 call
3/3 [=====] - 49s 13s/step - loss: 7.1854 - accuracy: 0.4935
Epoch 2/10
3/3 [=====] - ETA: 0s - loss: 1.3938 - accuracy: 0.9481 call
3/3 [=====] - 45s 13s/step - loss: 1.3938 - accuracy: 0.9481
Epoch 3/10
3/3 [=====] - ETA: 0s - loss: 1.0003e-04 - accuracy: 1.0000 call

Reached 99.00% accuracy, so stopping training!!
3/3 [=====] - 45s 13s/step - loss: 1.0003e-04 - accuracy: 1.0000
<keras.callbacks.History at 0x7f080abe0a30>

[ ] model2.evaluate(images,labels)
```

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```
[ ] model2.fit(images,labels,epochs=10,callbacks=[callbacks])

Epoch 1/10
3/3 [=====] - ETA: 0s - loss: 7.1854 - accuracy: 0.4935 call
3/3 [=====] - 49s 13s/step - loss: 7.1854 - accuracy: 0.4935
Epoch 2/10
3/3 [=====] - ETA: 0s - loss: 1.3938 - accuracy: 0.9481 call
3/3 [=====] - 45s 13s/step - loss: 1.3938 - accuracy: 0.9481
Epoch 3/10
3/3 [=====] - ETA: 0s - loss: 1.0003e-04 - accuracy: 1.0000 call

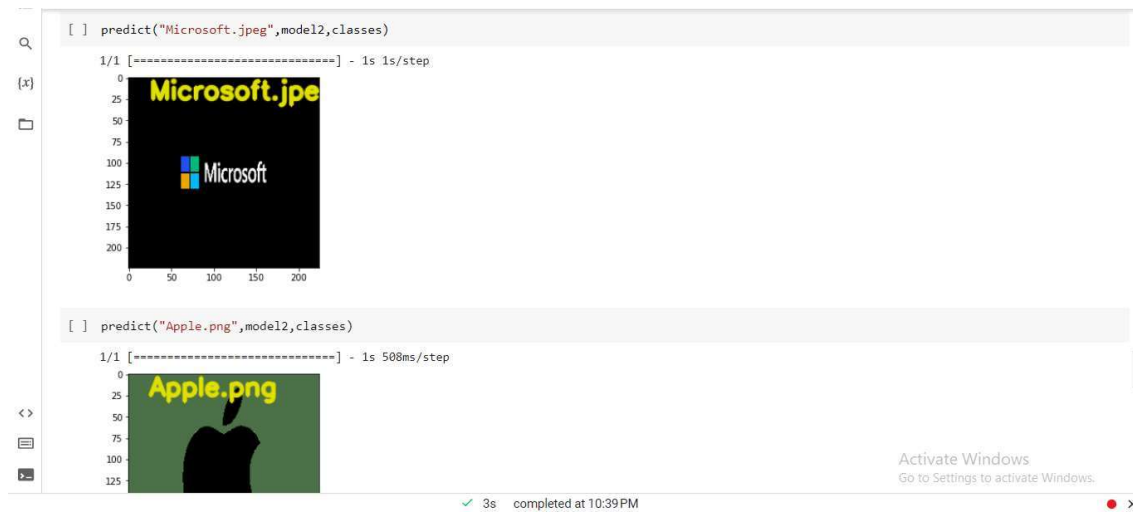
Reached 99.00% accuracy, so stopping training!!
3/3 [=====] - 45s 13s/step - loss: 1.0003e-04 - accuracy: 1.0000
<keras.callbacks.History at 0x7f080abe0a30>

[ ] model2.evaluate(images,labels)

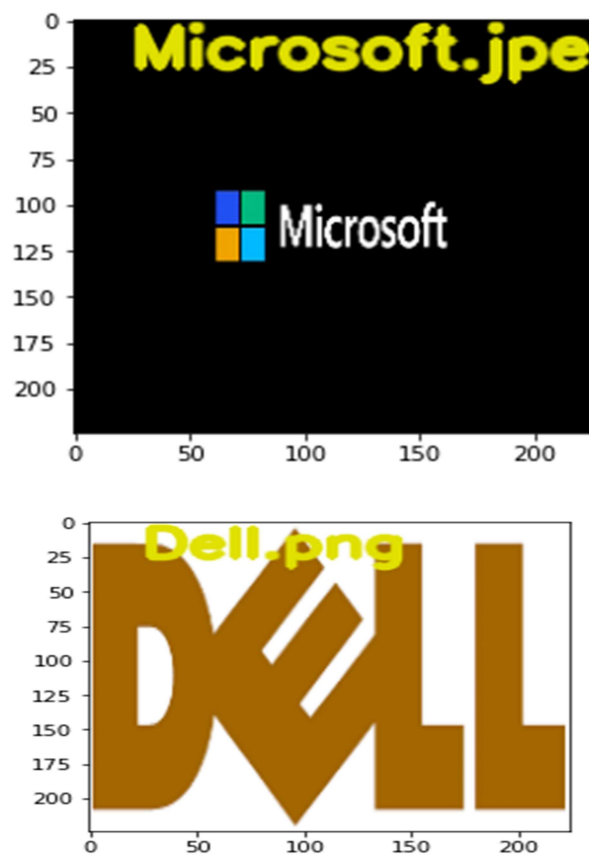
3/3 [=====] - 43s 13s/step - loss: 0.7537 - accuracy: 0.9610
[0.7536954879760742, 0.9610389471054077]

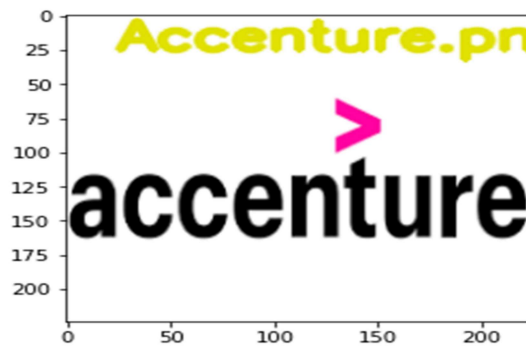
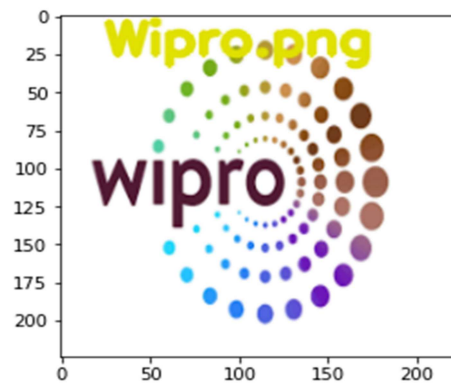
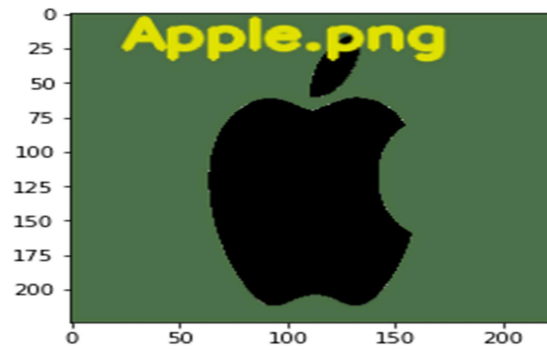
[ ] def predict(i,model,labels):
    path1 = f"{directory}/{i}"
    img = cv2.imread(path1)
    img = cv2.resize(img,(224,224))
    a = np.argmax(model.predict(np.array([img])))
    img = cv2.putText(img, labels[a], (25,25), cv2.FONT_HERSHEY_SIMPLEX,1, (225,225,0), 3, cv2.LINE_AA)
    plt.imshow(img)
```

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OUTPUT:





RESULTS:

Thus, the Images (LOGO) of the IT or MNCs are successfully recognized using convolution neural network.

Therefore, Accuracy = 0.96