12F FMA To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags. VOC-NOTICE: GPU memory for this assignment is capped at 1024MiB 2023-10-05 15:08:06.591857: E tensorflow/compiler/xla/stream_executor/cuda/cuda_driver.cc:267] failed call to cuInit: CUDA_ERROR_NO_DEVICE: no CUDA-capable device is detected In [2]: dataflow = ImageDataGenerator(rescale = 1.0 / 255.0) In [3]: # creating a train variable to store all the training data into a variable train = dataflow.flow_from_directory('./data/train/', class_mode = 'binary') Found 40 images belonging to 2 classes. In [4]: # creating a test variable to store all the testing data into a variable test = dataflow.flow_from_directory('./data/test', class_mode = 'binary') Found 20 images belonging to 2 classes. In [5]: # intializing the model with a variable name model model = models.Sequential() #Convolutional layer 1 with 32 filters of kernel size[5,5] model.add(layers.Conv2D(32, (5,5), activation = 'relu', padding = 'same', input_shape = (256, 256, 3))) # Pooling layer 1 with pool size[2,2] and stride 2 model.add(layers.MaxPooling2D(2,2)) # Convolutional layer 2 with 64 filters of kernel size[5,5] model.add(layers.Conv2D(64, (5,5), activation = 'relu')) # Pooling layer 2 with pool size[2,2] and stride 2 model.add(layers.MaxPooling2D(2,2)) # Dense layer whose output size is fixed in the hyper parameter: fc_size=32 model.add(layers.Flatten()) model.add(layers.Dense(32, activation='relu')) #Dropout layer with dropout probability 0.4 model.add(layers.Dropout(0.4)) model.add(layers.Dense(1, activation = 'sigmoid')) In [6]: # displaying the summary of the model model.summary() Model: "sequential" Layer (type) Output Shape Param # ______ 2432 conv2d (Conv2D) (None, 256, 256, 32) max_pooling2d (MaxPooling2D (None, 128, 128, 32) conv2d_1 (Conv2D) (None, 124, 124, 64) 51264 max_pooling2d_1 (MaxPooling (None, 62, 62, 64) 0 2D) flatten (Flatten) (None, 246016) 0 dense (Dense) (None, 32) 7872544 0 dropout (Dropout) (None, 32) dense_1 (Dense) (None, 1) 33 ______ Total params: 7,926,273 Trainable params: 7,926,273 Non-trainable params: 0 In [7]: sgd_opt = tf.keras.optimizers.SGD(learning_rate = 0.01) In [8]: model.compile(optimizer = sgd_opt, loss = 'binary_crossentropy', metrics = ['accuracy']) In [9]: model.fit(train, validation_data= test, epochs = 100) Epoch 2/100 Epoch 3/100 Epoch 4/100 Epoch 5/100 Epoch 6/100 Epoch 7/100 Epoch 8/100 Epoch 9/100 Epoch 10/100 Epoch 11/100 Epoch 12/100 Epoch 13/100 Epoch 14/100 Epoch 15/100 Epoch 16/100 Epoch 17/100 Epoch 18/100 Epoch 19/100 Epoch 20/100 Epoch 21/100 Epoch 22/100 Epoch 23/100 Epoch 24/100 Epoch 25/100 Epoch 26/100 Epoch 27/100 Epoch 28/100 Epoch 29/100 Epoch 30/100 Epoch 31/100 Epoch 32/100 Epoch 33/100 Epoch 34/100 Epoch 35/100 Epoch 36/100 Epoch 37/100 Epoch 38/100 Epoch 39/100 Epoch 40/100 Epoch 42/100 Epoch 43/100 Epoch 44/100 Epoch 45/100 Epoch 46/100 Epoch 47/100 Epoch 48/100 Epoch 50/100 Epoch 51/100 Epoch 52/100 Epoch 53/100 Epoch 54/100 Epoch 55/100 Epoch 56/100 Epoch 57/100 Epoch 58/100 Epoch 59/100 Epoch 60/100 Epoch 61/100 Epoch 62/100 Epoch 63/100 Epoch 64/100 Epoch 65/100 Epoch 66/100 Epoch 67/100 Epoch 68/100 Epoch 69/100 Epoch 70/100 Epoch 71/100 Epoch 72/100 Epoch 73/100

2023-10-05 15:08:03.491300: I tensorflow/core/platform/cpu_feature_guard.cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX2 AVX5

Epoch 74/100 Epoch 75/100 Epoch 76/100 Epoch 77/100 Epoch 78/100 Epoch 79/100 Epoch 80/100 Epoch 81/100 Epoch 82/100 Epoch 83/100

Epoch 84/100

Epoch 85/100

Epoch 86/100

Epoch 87/100

Epoch 88/100

Epoch 89/100

Epoch 90/100

Epoch 91/100

Epoch 92/100

Epoch 93/100

Epoch 94/100

Epoch 95/100

Epoch 96/100

Epoch 97/100

Epoch 98/100

Epoch 99/100

Out[9]: <keras.callbacks.History at 0x7fc428bf38e0>

from matplotlib import pyplot as plt

plt.title('ACCURACY') plt.ylabel('Accuracy') plt.xlabel('Epochs')

plt.show()

0.9

0.8

plt.plot(model.history.history['accuracy']) plt.plot(model.history.history['val_accuracy'])

plt.legend(['train', 'test'], loc = 'upper right')

ACCURACY

Epochs

100

In [10]: # comparing the training data and testing data reults through a line chart

In [1]: # importing the required libraries import tensorflow as tf from tensorflow import keras

from keras.preprocessing.image import ImageDataGenerator

from tensorflow.keras import models,layers