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Model Building
Predictions
Date: 11 November 2022
Team ID: PNT2022TMID50335
Project Name: Emerging Methods for early Detection Of forest fire
'''import model building libraries'''
#to define linear initialisation import sequential
from keras.models import Sequential
#to add layers import Dense
from keras.layers import Dense
#to create Convolution kernel import Covolution2D
from keras.layers import Convolution2D
#import Maxpooling Layer
from keras.layers import Dense
from keras.layers import Convolution2D
#import Flatten Layer
from keras.layers import Flatten
#import maxpooling layer
from keras.layers import MaxPooling2D
import warnings
warnings.filterwarnings('ignore')
#initializing model
model=Sequential()
Adding CNN Layer
Task 1:
#add cnn layer
model.add(Conv2D(filters=32,kernel size=2,padding="same",activation
="relu"))
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model.add(MaxPooling2D(pool size=2))
model.add(Conv2D(filters=64,kernel size=2,padding="same",activation="r
elu"))
model.add(MaxPooling2D(pool size=2))
Task 2:
#flattening layer
model.add(Flatten())
Task 3:
#Adding PoolingLayer
model.add(MaxPooling2D(pool size=(2,2)))
Adding Dense Layer
#adding DenseLayer
model.add(Dense(500,activation="relu"))
model.add(Dense(2,activation="softmax"))
print("Adding dense layer on top")
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(10))
print("Complete architecture of the model")
model.summary()
Configuring the Learning Process
#Configuring the learning process
model.compile(loss='binary crossentrophy',optimizer="adam",
metrices=["accuracy"])
Test The Model
#importing the required libraries
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D
from tensorflow.keras.layers import MaxPool2D
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dropout
from tensorflow.keras.layers import Dense
#loading data
(X train,y train) , (X test,y test)=mnist.load data()
#reshaping data
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X train = X train.reshape((X train.shape[0], X train.shape[1],
X train.shape[2], 1))
X test =
X test.reshape((X test.shape[0], X test.shape[1], X test.shape[2],1))
#checking the shape after reshaping
print(X train.shape)
print(X test.shape)
#normalizing the pixel values
X train=X train/255
X test=X test/255
Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/mnist.npz
(60000, 28, 28, 1)
(10000, 28, 28, 1)
#defining model
model=Sequential()
#adding convolution layer
model.add(Conv2D(32,(3,3),activation='relu',input shape=(28,28,1)))
#adding pooling layer
model.add(MaxPool2D(2,2))
#adding fully connected layer
model.add(Flatten())
model.add(Dense(100,activation='relu'))
#adding output layer
model.add(Dense(10,activation='softmax'))
#compiling the model
model.compile(loss='sparse categorical crossentropy',optimizer='adam',
metrics=['accuracy'])
#fitting the model
model.fit(X train,y train,epochs=10)
Epoch 1/10
0.1474 - accuracy: 0.9566
Epoch 2/10
0.0534 - accuracy: 0.9837
Epoch 3/10
0.0338 - accuracy: 0.9893
Epoch 4/10
0.0218 - accuracy: 0.9931
Epoch 5/10
0.0149 - accuracy: 0.9952
Epoch 6/10
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0.0101 - accuracy: 0.9966
Epoch 7/10
0.0084 - accuracy: 0.9973
Epoch 8/10
0.0069 - accuracy: 0.9977
Epoch 9/10
0.0051 - accuracy: 0.9985
Epoch 10/10
0.0045 - accuracy: 0.9985
<keras.callbacks.History at 0x7f3bbd6264d0>
Save The Model
import tensorflow as tf
from tensorflow.keras import datasets, layers, models
import matplotlib.pyplot as plt
plt.plot(history.history['accuracy'], label='accuracy')
plt.plot(history.history['val accuracy'], label = 'val accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.ylim([0.5, 1])
plt.legend(loc='lower right')
test loss, test acc = model.evaluate(test images, test labels,
verbose=2)
Predictions
#import the models
from keras.models import load model
#import image classs from keras
from keras.preprocessing import image
#import numpy
import numpy as np
#import cv2
import cv2
#give any random image path
img=image.load img('/content/with fire (18).jpg')
x=image.img to array(img)
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#expand the image shape
x=np.expand_dims(x,axis=0)
pred=model.predict_classes(x)
pred.
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