Team Id:PNT2022TMID46656 Model Building

Adding The Convolution Layer In []: import numpy as np import matplotlib.pyplot as plt In []: from tensorflow.keras.preprocessing.image import ImageDataGenerator In []: # Training Datagen train datagen ImageDataGenerator(rescale=1/255, zoom range=0.2, horizontal flip=True, vertica 1 flip=False) # Testing Datagen test datagen = ImageDataGenerator(rescale=1/255) In []: # Training Dataset x train=train datagen.flow from directory(r'/content/drive/MyDrive/Dataset/t raining set', target size=(64,64), class mode='categorical', batch size=900) # Testing Dataset x test=test datagen.flow from directory(r'/content/drive/MyDrive/Dataset/tes t set',target size=(64,64), class mode='categorical',batch size=900) Found 15760 images belonging to 9 classes. Found 2250 images belonging to 9 classes. In []: # let img1 be an image with no features img1 = np.array([np.array([200, 200]), np.array([200, 200])]) img2 =np.array([np.array([200, 200]), np.array([0, 0])]) img3 =np.array([np.array([200, 0]), np.array([200, 0])]) kernel horizontal = np.array([np.array([2, 2]), np.array([-2, -2])])print(kernel horizontal, 'is a kernel for detecting horizontal edges') kernel vertical = np.array([np.array([2, -2]), np.array([2, -2])]) print(kernel vertical, 'is a kernel for detecting vertical edges') In []: # We will apply the kernels on the images by # elementwise multiplication followed by summation def apply kernel(img, kernel): return np.sum(np.multiply(img, kernel)) # Visualizing img1 plt.imshow(img1) plt.axis('off') plt.title('img1') plt.show()

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# Checking for horizontal and vertical features in image1 print('Horizontal
edge confidence score:', apply kernel(img1,
kernel horizontal)) print('Vertical edge confidence score:',
apply kernel (img1,
kernel vertical))
In []:
# Visualizing img2
plt.imshow(img2)
plt.axis('off')
plt.title('img2') plt.show()
# Checking for horizontal and vertical features in image2 print('Horizontal
edge confidence score:', apply kernel(img2,
kernel horizontal)) print('Vertical edge confidence score:',
apply kernel(img2,
kernel vertical))
In []:
# Visualizing img3
plt.imshow(img3)
plt.axis('off')
plt.title('img3') plt.show()
# Checking for horizontal and vertical features in image3 print('Horizontal
edge confidence score:', apply kernel(img3,
kernel horizontal)) print('Vertical edge confidence score:',
apply kernel(img3,
kernel vertical))
In []:
print("Len x-train : ", len(x train)) print("Len
x-test : ", len(x test))
Len x-train : 18 Len
x-test: 3
In []:
# The Class Indices in Training Dataset
x train.class indices
Out[]:
{'A': 0, 'B': 1, 'C': 2, 'D': 3, 'E': 4, 'F': 5, 'G': 6, 'H': 7, 'I': 8}
Model Creation
In []:
# Importing Libraries
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D, MaxPooling2D, Flatten, Dense
In []:
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# Creating Model model=Sequential()
In[]:
# Adding Layers
model.add(Convolution2D(32,(3,3))
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