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import numpy as np
import pandas as pd
import pydicom as dicom
import os
import matplotlib.pyplot as plt
import cv2
import math
import tensorflow._api.v2.compat.v1 as tf
tf.disable_v2_behavior()
import pandas as pd
import tflearn
from tflearn.layers.conv import conv_3d, max_pool_3d
from tflearn.layers.core import input_data, dropout, fully_connected
from tflearn.layers.estimator import regression
import numpy as np
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix
from tkinter import *
from tkinter import messagebox,ttk
import tkinter as tk
from PIL import Image,ImageTk
class LCD_CNN:
def __init__(self,root):
self.root=root
#window size
self.root.geometry("1006x500+0+0")
self.root.resizable(False, False)
self.root.title("Lung Cancer Detection")
img4=Image.open(r"Images\Lung-Cancer-Detection.jpg")
img4=img4.resize((1006,500),Image.ANTIALIAS)
#Antialiasing is a technique used in digital imaging to reduce the visual defects that occur when high-r
self.photoimg4=ImageTk.PhotoImage(img4)
bg_img=Label(self.root,image=self.photoimg4)
bg_img.place(x=0,y=50,width=1006,height=500)
# title Label
title_lbl=Label(text="Lung Cancer Detection",font=("Bradley Hand ITC",30,"bold"),bg="black",fg="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="white-lbl=Label(text="Lung Cancer Detection",fg="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label(text="white-lbl=Label
title_lbl.place(x=0,y=0,width=1006,height=50)
#button 1
self.b1=Button(text="Import Data",cursor="hand2",command=self.import_data,font=("Times New Ror
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from unittest import result

from xml.dom.expatbuilder import parseString

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self.b1.place(x=80,y=130,width=180,height=30)
#button 2
self.b2=Button(text="Pre-Process Data",cursor="hand2",command=self.preprocess_data,font=("Times
self.b2.place(x=80,y=180,width=180,height=30)
self.b2["state"] = "disabled"
self.b2.config(cursor="arrow")
#button 3
self.b3=Button(text="Train Data",cursor="hand2",command=self.train_data,font=("Times New Roman
self.b3.place(x=80,y=230,width=180,height=30)
self.b3["state"] = "disabled"
self.b3.config(cursor="arrow")
#Data Import lets you upload data from external sources and combine it with data you collect via Anal
def import_data(self):
##Data directory
self.dataDirectory = 'sample_images/'
self.lungPatients = os.listdir(self.dataDirectory)
##Read labels csv
self.labels = pd.read_csv('stage1_labels.csv', index_col=0)
##Setting x*y size to 50
##Setting x*y size to 10
self.size = 10
## Setting z-dimension (number of slices to 20)
## Setting z-dimension (number of slices to 5)
self.NoSlices = 5
messagebox.showinfo("Import Data", "Data Imported Successfully!")
self.b1["state"] = "disabled"
self.b1.config(cursor="arrow")
self.b2["state"] = "normal"
self.b2.config(cursor="hand2")
# Data preprocessing is the process of transforming raw data into an understandable format.
def preprocess_data(self):
def chunks(l, n):
count = 0
for i in range(0, len(l), n):
if (count < self.NoSlices):</pre>
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yield l[i:i + n]
count = count + 1
def mean(l):
return sum(l) / len(l)
#Average
def dataProcessing(patient, labels_df, size=10, noslices=5, visualize=False):
label = labels_df._get_value(patient, 'cancer')
path = self.dataDirectory + patient
slices = [dicom.read_file(path + '/' + s) for s in os.listdir(path)]
slices.sort(key=lambda x: int(x.ImagePositionPatient[2]))
new_slices = []
slices = [cv2.resize(np.array(each_slice.pixel_array), (size, size)) for each_slice in slices]
chunk_sizes = math.floor(len(slices) / noslices)
for slice_chunk in chunks(slices, chunk_sizes):
slice_chunk = list(map(mean, zip(*slice_chunk)))
new_slices.append(slice_chunk)
if label == 1: #Cancer Patient
label = np.array([0, 1])
elif label == 0: #Non Cancerous Patient
label = np.array([1, 0])
return np.array(new_slices), label
imageData = []
#Check if Data Labels is available in CSV or not
for num, patient in enumerate(self.lungPatients):
if num \% 100 == 0:
if num % 50 == 0:
print('Saved -', num)
try:
img_data, label = dataProcessing(patient, self.labels, size=self.size, noslices=self.NoSlices)
imageData.append([img_data, label,patient])
except KeyError as e:
print('Data is unlabeled')
##Results= Image Data and lable.
np.save('imageDataNew-{}-{}-npy'.format(self.size, self.size, self.NoSlices), imageData)
messagebox.showinfo("Pre-Process Data", "Data Pre-Processing Done Successfully!")
self.b2["state"] = "disabled"
self.b2.config(cursor="arrow")
self.b3["state"] = "normal"
self.b3.config(cursor="hand2")
# Data training is the process of training the model based on the dataset and then predict on new data.
def train_data(self):
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imageData = np.load('imageDataNew-10-10-5.npy',allow_pickle=True)
trainingData = imageData[0:45]
validationData = imageData[45:50]
training_data=Label(text="Total Training Data: " + str(len(trainingData)),font=("Times New Roman",1
training_data.place(x=750,y=150,width=200,height=18)
validation_data=Label(text="Total Validation Data: " + str(len(validationData)),font=("Times New Ron
validation_data.place(x=750,y=190,width=200,height=18)
x = tf.placeholder('float')
y = tf.placeholder('float')
size = 10
keep_rate = 0.8
NoSlices = 5
def convolution3d(x, W):
return tf.nn.conv3d(x, W, strides=[1, 1, 1, 1, 1], padding='SAME')
def maxpooling3d(x):
return tf.nn.max_pool3d(x, ksize=[1, 2, 2, 2, 1], strides=[1, 2, 2, 2, 1], padding='SAME')
def cnn(x):
x = tf.reshape(x, shape=[-1, size, size, NoSlices, 1])
convolution1 = tf.nn.relu(
convolution3d(x, tf.Variable(tf.random_normal([3, 3, 3, 1, 32]))) + tf.Variable(tf.random_normal([32]))
convolution1 = maxpooling3d(convolution1)
convolution2 = tf.nn.relu(
convolution3d(convolution1, tf.Variable(tf.random_normal([3, 3, 3, 32, 64]))) + tf.Variable(
tf.random_normal([64])))
convolution2 = maxpooling3d(convolution2)
convolution3 = tf.nn.relu(
convolution3d(convolution2, tf.Variable(tf.random_normal([3, 3, 3, 64, 128]))) + tf.Variable(
tf.random_normal([128])))
convolution3 = maxpooling3d(convolution3)
convolution4 = tf.nn.relu(
convolution3d(convolution3, tf.Variable(tf.random_normal([3, 3, 3, 128, 256]))) + tf.Variable(
tf.random_normal([256])))
convolution4 = maxpooling3d(convolution4)
convolution5 = tf.nn.relu(
convolution3d(convolution4, tf.Variable(tf.random_normal([3, 3, 3, 256, 512]))) + tf.Variable(
tf.random_normal([512])))
convolution5 = maxpooling3d(convolution4)
fullyconnected = tf.reshape(convolution5, [-1, 256])
fullyconnected = tf.nn.relu(
tf.matmul(fullyconnected, tf.Variable(tf.random_normal([256, 256]))) + tf.Variable(tf.random_normal(
fullyconnected = tf.nn.dropout(fullyconnected, keep_rate)
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output = tf.matmul(fullyconnected, tf.Variable(tf.random_normal([256, 2]))) + tf.Variable(tf.random_n
return output
def network(x):
prediction = cnn(x)
cost = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits(logits=prediction, labels=y))
optimizer = tf.train.AdamOptimizer(learning_rate=1e-3).minimize(cost)
epochs = 100
with tf.Session() as session:
session.run(tf.global_variables_initializer())
for epoch in range(epochs):
epoch_{loss} = 0
for data in trainingData:
try:
X = data[0]
Y = data[1]
_, c = session.run([optimizer, cost], feed_dict={x: X, y: Y})
epoch_loss += c
except Exception as e:
pass
correct = tf.equal(tf.argmax(prediction, 1), tf.argmax(y, 1))
# if tf.argmax(prediction, 1) == 0:
accuracy = tf.reduce_mean(tf.cast(correct, 'float'))
print('Epoch', epoch + 1, 'completed out of', epochs, 'loss:', epoch_loss)
# print('Correct:',correct.eval({x:[i[0] for i in validationData], y:[i[1] for i in validationData]}))
print('Accuracy:', accuracy.eval({x: [i[0] for i in validationData], y: [i[1] for i in validationData]}))
#print('Final Accuracy:', accuracy.eval({x: [i[0] for i in validationData], y: [i[1] for i in validationData]})
x1 = accuracy.eval({x: [i[0] for i in validationData], y: [i[1] for i in validationData]})
final_accuracy=Label(text="Final Accuracy: " + str(x1),font=("Times New Roman",13,"bold"),bg="black
final_accuracy.place(x=750,y=230,width=200,height=18)
patients = []
actual = []
predicted = []
finalprediction = tf.argmax(prediction, 1)
actualprediction = tf.argmax(y, 1)
for i in range(len(validationData)):
patients.append(validationData[i][2])
for i in finalprediction.eval(\{x: [i[0] \text{ for i in validationData}], y: [i[1] \text{ for i in validationData}]\}\}:
if(i==1):
predicted.append("Cancer")
else:
predicted.append("No Cancer")
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for i in actualprediction.eval({x: [i[0] for i in validationData], y: [i[1] for i in validationData]}):
if(i==1):
actual.append("Cancer")
actual.append("No Cancer")
for i in range(len(patients)):
print("-----")
print("Patient: ",patients[i])
print("Actual: ", actual[i])
print("Predicted: ", predicted[i])
print("-----")
# messagebox.showinfo("Result", "Patient: " + ' '.join(map(str,patients)) + "\nActual: " + str(actual) + "
y_actual = pd.Series(
(actualprediction.eval({x: [i[0] for i in validationData], y: [i[1] for i in validationData]})),
name='Actual')
y_predicted = pd.Series(
(finalprediction.eval({x: [i[0] for i in validationData], y: [i[1] for i in validationData]})),
name='Predicted')
df_confusion = pd.crosstab(y_actual, y_predicted).reindex(columns=[0,1],index=[0,1], fill_value=0)
print('Confusion Matrix:\n')
print(df_confusion)
prediction_label=Label(text=">>>> P R E D I C T I O N <<<<",font=("Times New Roman",14,"bold"),bg=
prediction_label.place(x=0,y=458,width=1006,height=20)
result1 = []
for i in range(len(validationData)):
result1.append(patients[i])
if(y_actual[i] == 1):
result1.append("Cancer")
else:
result1.append("No Cancer")
if(y_predicted[i] == 1):
result1.append("Cancer")
else:
result1.append("No Cancer")
# print(result1)
total_rows = int(len(patients))
total_columns = int(len(result1)/len(patients))
heading = ["Patient: ", "Actual: ", "Predicted: "]
self.root.geometry("1006x"+str(500+(len(patients)*20)-20)+"+0+0")
self.root.resizable(False, False)
for i in range(total_rows):
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for j in range(total_columns):
self.e = Entry(root, width=42, fg='black', font=('Times New Roman',12,'bold'))
self.e.grid(row=i, column=j)
self.e.place(x=(j*335),y=(478+i*20))
self.e.insert(END, heading[j] + result1[j + i*3])
self.e["state"] = "disabled"
self.e.config(cursor="arrow")
self.b3["state"] = "disabled"
self.b3.config(cursor="arrow")
messagebox.showinfo("Train Data", "Model Trained Successfully!")
## Function to plot confusion matrix
def plot_confusion_matrix(df_confusion, title='Confusion matrix', cmap=plt.cm.gray_r):\
plt.matshow(df_confusion, cmap=cmap) # imshow
# plt.title(title)
plt.colorbar()
tick_marks = np.arange(len(df_confusion.columns))
plt.title(title)
plt.xticks(tick_marks, df_confusion.columns, rotation=45)
plt.yticks(tick_marks, df_confusion.index)
# plt.tight_layout()
plt.ylabel(df_confusion.index.name)
plt.xlabel(df_confusion.columns.name)
plt.show()
plot_confusion_matrix(df_confusion)
# print(y_true,y_pred)
# print(confusion_matrix(y_true, y_pred))
# print(actualprediction.eval({x:[i[0] for i in validationData], y:[i[1] for i in validationData]}))
# print(finalprediction.eval({x:[i[0] for i in validationData], y:[i[1] for i in validationData]}))
network(x)
# For GUI
if __name__ == "__main__":
root=Tk()
obj=LCD_CNN(root)
root.mainloop()
root.mainloop()
```