

## APPENDIX

### CODING

#### GABOR FILTER FUNCTION

```
1. def gaborFilter(img):
2.     filter = []
3.     num_filters = 16
4.     ksize = 35
5.     sigma = 10.0
6.     lmbda = 30.0
7.     gamma = 0.5
8.     psi = 0
9.     for theta in np.arange(0, np.pi, np.pi / num_filters):
10.        krn = cv2.getGaborKernel((ksize, ksize), sigma, theta, lmbda,
11.                                gamma, psi, ktype=cv2.CV_64F)
12.        krn /= 1.0 * krn.sum()
13.        filter.append(krn)
14.    filteredImage = np.zeros_like(img)
15.    depth = -1
16.    for krn in filter:
17.        filtering = cv2.filter2D(img, depth, krn)
18.        np.maximum(filteredImage, filtering, filteredImage)
19.    return filteredImage
```

#### CHAN-VESE SEGMENTATION FUNCTION

```
1. def chanveseFilter(img):
2.     cv = chan_vese(img, mu=0.25, lambda1=1, lambda2=1, tol=1e-3,
3.                   max_num_iter=200, dt=0.5,
4.                   init_level_set="checkerboard", extended_output=True)
5.     return cv
```

#### COMBINED FEATURE / BLEND IMAGE FUNCTION

```
1. def blendImage(imgGabor, imgCV):
2.     imgCV = 255 - imgCV
3.     blendImg = cv2.subtract(imgCV, imgGabor, dtype=cv2.CV_8U)
4.     return blendImg
```

#### IMAGE PREPROCESSING FUNCTION

```
1. def imgPreProcess(img):
2.     gbImg = gaborFilter(img)
3.     cvImg = chanveseFilter(img)
4.     resultImg = blendImage(np.asarray(cvImg[1]), gbImg)
5.     resultImg = 2.2 * resultImg + 75
6.     return resultImg
```

## ONE-HOT ENCODING CONVERSION FUNCTION

```
1. def convert_to_one_hot(Y, C):
2.     Y = np.eye(C)[Y.reshape(-1)].T
3.     return Y
```

## DATA LOAD FUNCTION

```
1. def loadData(DATA_DIR):
2.     DATA_Images = []
3.     progress = 1
4.     for imgFile in os.listdir(DATA_DIR):
5.         print('Image File Name : ',imgFile)
6.         tempHolder1 = cv2.imread(os.path.join(DATA_DIR,imgFile))
7.         tempHolder1 = cv2.cvtColor(tempHolder1, cv2.COLOR_BGR2YCR_CB)
8.         tempHolder2 = cv2.resize(tempHolder1[:, :, 1
9.         (200,200),interpolation=cv2.INTER_AREA)
10.        print(" Success Reading Image ...")
11.        imgProcess = imgPreProcess(tempHolder2)
12.        print(" Success Processing Image ...(",progress,")")
13.        progress += 1
14.        if 'au' in imgFile.lower():
15.            DATA_Images.append([imgProcess,1])
16.        elif 'tp' in imgFile.lower():
17.            DATA_Images.append([imgProcess,0])
18.        TRAIN_img, TEST_img = train_test_split(DATA_Images, test_size=0.2,
19.        random_state=25)
20.        train_set_x = []
21.        train_set_y = []
22.        test_set_x = []
23.        test_set_y = []
24.        for xTr,yTr in TRAIN_img:
25.            train_set_x.append(xTr)
26.            train_set_y.append(yTr)
27.        for xTe,yTe in TEST_img:
28.            test_set_x.append(xTe)
29.            test_set_y.append(yTe)
30.        return train_set_x, train_set_y, test_set_x, test_set_y
```

## LOAD AND PREPARE DATASET

```
1. ROWS = 200
2. COLS = 200
3. CHANNELS = 1
4. CLASSES = 2
5.
6. DATA_DIRECT = "C:\\Users\\...\\ProjectDocument\\TrainValData\\"
7. train_set_x, train_set_y, test_set_x, test_set_y = loadData(DATA_DIRECT)
8. train_set_x = np.array(train_set_x,dtype="uint8")
9. train_set_y = np.array(train_set_y,dtype="uint8")
10. test_set_x = np.array(test_set_x,dtype="uint8")
11. test_set_y = np.array(test_set_y,dtype="uint8")
12. X_train = train_set_x/255
13. X_test = test_set_x/255
```

```

14.Y_train = convert_to_one_hot(train_set_y, CLASSES).T
15.Y_test = convert_to_one_hot(test_set_y, CLASSES).T

```

### IDENTITY BLOCK FUNCTION

```

1. def identity_block(X, f, filters, stage, block):
2.     conv_name_base = 'res' + str(stage) + block + '_branch'
3.     bn_name_base = 'bn' + str(stage) + block + '_branch'
4.
5.     F1, F2, F3 = filters
6.
7.     X_shortcut = X
8.
9.     X = Conv2D(filters = F1, kernel_size = (1, 1), strides = (1,1),
padding = 'valid', name = conv_name_base + '2a', kernel_initializer
= glorot_uniform(seed=0))(X)
10.    X = BatchNormalization(axis = 3, name = bn_name_base + '2a')(X)
11.    X = Activation('relu')(X)
12.
13.    X = Conv2D(filters = F2, kernel_size = (f, f), strides = (1,1),
padding = 'same', name = conv_name_base + '2b', kernel_initializer
= glorot_uniform(seed=0))(X)
14.    X = BatchNormalization(axis = 3, name = bn_name_base + '2b')(X)
15.    X = Activation('relu')(X)
16.
17.    X = Conv2D(filters = F3, kernel_size = (1, 1), strides = (1,1),
padding = 'valid', name = conv_name_base + '2c', kernel_initializer
= glorot_uniform(seed=0))(X)
18.    X = BatchNormalization(axis = 3, name = bn_name_base + '2c')(X)
19.
20.    X = Add()([X, X_shortcut])
21.    X = Activation('relu')(X)
22.    return X

```

### CONVOLUTIONAL BLOCK FUNCTION

```

1. def convolutional_block(X, f, filters, stage, block, s = 2):
2.     conv_name_base = 'res' + str(stage) + block + '_branch'
3.     bn_name_base = 'bn' + str(stage) + block + '_branch'
4.
5.     F1, F2, F3 = filters
6.
7.     X_shortcut = X
8.
9.
10.    X = Conv2D(F1, (1, 1), strides = (s,s), name = conv_name_base +
'2a', kernel_initializer = glorot_uniform(seed=0))(X)
11.    X = BatchNormalization(axis = 3, name = bn_name_base + '2a')(X)
12.    X = Activation('relu')(X)
13.
14.    X = Conv2D(filters=F2, kernel_size=(f, f), strides=(1, 1),
padding='same', name=conv_name_base+'2b', kernel_initializer=glorot_un
iform(seed=0))(X)

```

```

15. X = BatchNormalization(axis=3, name=bn_name_base + '2b')(X)
16. X = Activation('relu')(X)
17.
18. X = Conv2D(filters=F3, kernel_size=(1, 1), strides=(1, 1),
padding='valid',name=conv_name_base+'2c',kernel_initializer=glorot_u
niform(seed=0))(X)
19. X = BatchNormalization(axis=3, name=bn_name_base + '2c')(X)
20.
21. X_shortcut = Conv2D(F3, (1, 1), strides = (s,s), name =
conv_name_base + '1', kernel_initializer = glorot_uniform(seed=0))
(X_shortcut)
22. X_shortcut = BatchNormalization(axis = 3, name = bn_name_base + '1')
(X_shortcut)
23.
24. X = Add()([X, X_shortcut])
25. X = Activation('relu')(X)
26.
27. return X

```

### RESNET50 FUNCTION

```

1. def ResNet50(input_shape = (200, 200, 1), classes = 2):
2.     X_input = Input(input_shape)
3.
4.     X = ZeroPadding2D((3, 3))(X_input)
5.
6.     X = Conv2D(64, (7, 7), strides = (2, 2), name = 'conv1',
kernel_initializer = glorot_uniform(seed=0))(X)
7.     X = BatchNormalization(axis = 3, name = 'bn_conv1')(X)
8.     X = Activation('relu')(X)
9.     X = MaxPooling2D((3, 3), strides=(2, 2))(X)
10.
11.     X = convolutional_block(X, f = 3, filters = [64, 64, 256], stage =
2, block='a', s = 1)
12.     X = identity_block(X, 3, [64, 64, 256], stage=2, block='b')
13.     X = identity_block(X, 3, [64, 64, 256], stage=2, block='c')
14.
15.     X = convolutional_block(X, f = 3, filters = [128, 128, 512], stage =
3, block='a', s = 2)
16.     X = identity_block(X, 3, [128, 128, 512], stage=3, block='b')
17.     X = identity_block(X, 3, [128, 128, 512], stage=3, block='c')
18.     X = identity_block(X, 3, [128, 128, 512], stage=3, block='d')
19.
20.     X = convolutional_block(X, f = 3, filters = [256, 256, 1024], stage
= 4, block='a', s = 2)
21.     X = identity_block(X, 3, [256, 256, 1024], stage=4, block='b')
22.     X = identity_block(X, 3, [256, 256, 1024], stage=4, block='c')
23.     X = identity_block(X, 3, [256, 256, 1024], stage=4, block='d')
24.     X = identity_block(X, 3, [256, 256, 1024], stage=4, block='e')
25.     X = identity_block(X, 3, [256, 256, 1024], stage=4, block='f')
26.
27.     X = convolutional_block(X, f = 3, filters = [512, 512, 2048], stage
= 5, block='a', s = 2)
28.     X = identity_block(X, 3, [512, 512, 2048], stage=5, block='b')

```

```

29. X = identity_block(X, 3, [512, 512, 2048], stage=5, block='c')
30.
31. X = AveragePooling2D((2, 2), name='avg_pool')(X)
32.
33. X = Flatten()(X)
34. X = Dense(classes, activation='softmax', name='fc' + str(classes),
kernel_initializer = glorot_uniform(seed=0))(X)
35.
36. model = Model(inputs = X_input, outputs = X, name='ResNet50')
37.
38. return model

```

### BUILD RESNET50 MODEL

```

1. model = ResNet50(input_shape = (ROWS, COLS, CHANNELS), classes =
CLASSES)
2.
3. model.compile(optimizer='adam', loss='categorical_crossentropy',
metrics=['accuracy'])
4.
5. K.set_value(model.optimizer.learning_rate, 0.0000001)
6.
7. model.fit(X_train, Y_train, epochs = 50, batch_size = 16)
8.
9. preds = model.evaluate(X_test, Y_test)
10.print ("Loss = " + str(preds[0]))
11.print ("Test Accuracy = " + str(preds[1]))
12.
13.model.summary()
14.
15.model.save('ResNet50.h5')

```

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