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Федеральное государственное бюджетное образовательное  
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НИЖЕГОРОДСКИЙ ГОСУДАРСТВЕННЫЙ ТЕХНИЧЕСКИЙ УНИВЕРСИТЕТ  
ИМ. Р.Е. АЛЕКСЕЕВА  
ИНСТИТУТ РАДИОЭЛЕКТРОНИКИ И ИНФОРМАЦИОННЫХ ТЕХНОЛОГИЙ

Курс “Аппаратное и программное обеспечение роботизированных систем”

Отчет по лабораторной работе №3

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Нижний Новгород 2021

## Тема работы:

Классификация изображений с использованием свёрточных нейронных сетей.

## Задание:

Выполнить анализ статьи, разобрать структуру сети, реализовать сеть в Keras, оценить точность работы сети.

Вариант данных: MNIST

Вариант модели сети: Xception

## Листинг программы:

```
[ ] import keras
import tensorflow as tf

from keras.applications.xception import Xception, preprocess_input
from keras.models import Sequential, Model
from keras.layers import Lambda, Input
from keras.backend import tf as ktf
from keras.layers.core import Flatten, Dense, Dropout
from keras.utils import np_utils
from keras.optimizers import SGD, Adam
from scipy import misc
import numpy as np

mnist = tf.keras.datasets.mnist

[ ] (x_train, y_train), (x_test, y_test) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11493376/11490434 [=====] - 0s 0us/step

[ ] # Размер изображений
img_width, img_height = 28, 28
# Размер мини-выборки
batch_size = 32
# Кол-во изображений для обучения
nb_train_samples = 60000
# Кол-во изображений для теста
nb_test_samples = 10000
```

```
[ ] # Нормалтизация изображений к значениям -1, 1
x_train = tf.keras.utils.normalize(x_train, axis=1)
x_test = tf.keras.utils.normalize(x_test, axis=1)

x_train = x_train.astype('float32')
x_test = x_test.astype('float32')
x_train = x_train / 255.0
x_test = x_test / 255.0
x_train = np.stack((x_train,)*3, axis=-1)
x_test = np.stack((x_test,)*3, axis=-1)

y_train = np_utils.to_categorical(y_train)
y_test = np_utils.to_categorical(y_test)
num_classes = y_test.shape[1]
```

```
# Создание экземпляра модели сети
xception = tf.keras.applications.Xception(
    include_top=False,
    weights="imagenet",
    input_shape= (128,128,3)
)
# input_shape : Необязательный кортеж формы
# include_top : логическое значение, следует ли включать полностью
#подключенный уровень в верхнюю часть сети.
# weights : один из None(случайная инициализация), «imagenet»
#(предварительное обучение в ImageNet) или путь к загружаемому файлу весов.
# Сверточная часть сети
xception.trainable = False
trainable = False
for layer in xception.layers:
    if layer.name == 'block5_sepconv1':
        trainable = True
        layer.trainable = trainable
# # Сверточную часть сети обучать не надо

xception.summary()
```

Создадим экземпляр модели:

Model: "xception"

Layer (type)	Output Shape	Param #	Connected to
input_13 (InputLayer)	[(None, 128, 128, 3)]	0	
block1_conv1 (Conv2D)	(None, 63, 63, 32)	864	input_13[0][0]
block1_conv1_bn (BatchNormaliza	(None, 63, 63, 32)	128	block1_conv1[0][0]
block1_conv1_act (Activation)	(None, 63, 63, 32)	0	block1_conv1_bn[0][0]
block1_conv2 (Conv2D)	(None, 61, 61, 64)	18432	block1_conv1_act[0][0]
block1_conv2_bn (BatchNormaliza	(None, 61, 61, 64)	256	block1_conv2[0][0]
block1_conv2_act (Activation)	(None, 61, 61, 64)	0	block1_conv2_bn[0][0]
block2_sepconv1 (SeparableConv2	(None, 61, 61, 128)	8768	block1_conv2_act[0][0]
block2_sepconv1_bn (BatchNormal	(None, 61, 61, 128)	512	block2_sepconv1[0][0]
block2_sepconv2_act (Activation	(None, 61, 61, 128)	0	block2_sepconv1_bn[0][0]
block2_sepconv2 (SeparableConv2	(None, 61, 61, 128)	17536	block2_sepconv2_act[0][0]
block2_sepconv2_bn (BatchNormal	(None, 61, 61, 128)	512	block2_sepconv2[0][0]
conv2d_48 (Conv2D)	(None, 31, 31, 128)	8192	block1_conv2_act[0][0]
block2_pool (MaxPooling2D)	(None, 31, 31, 128)	0	block2_sepconv2_bn[0][0]
batch_normalization_48 (BatchNo	(None, 31, 31, 128)	512	conv2d_48[0][0]
add_144 (Add)	(None, 31, 31, 128)	0	block2_pool[0][0] batch_normalization_48[0][0]
block3_sepconv1_act (Activation	(None, 31, 31, 128)	0	add_144[0][0]

block3_sepconv1	(SeparableConv2 (None, 31, 31, 256)	33920	block3_sepconv1_act[0][0]
block3_sepconv1_bn	(BatchNormal (None, 31, 31, 256)	1024	block3_sepconv1[0][0]
block3_sepconv2_act	(Activation (None, 31, 31, 256)	0	block3_sepconv1_bn[0][0]
block3_sepconv2	(SeparableConv2 (None, 31, 31, 256)	67840	block3_sepconv2_act[0][0]
block3_sepconv2_bn	(BatchNormal (None, 31, 31, 256)	1024	block3_sepconv2[0][0]
conv2d_49	(Conv2D) (None, 16, 16, 256)	32768	add_144[0][0]
block3_pool	(MaxPooling2D) (None, 16, 16, 256)	0	block3_sepconv2_bn[0][0]
batch_normalization_49	(BatchNo (None, 16, 16, 256)	1024	conv2d_49[0][0]
add_145	(Add) (None, 16, 16, 256)	0	block3_pool[0][0] batch_normalization_49[0][0]
block4_sepconv1_act	(Activation (None, 16, 16, 256)	0	add_145[0][0]
block4_sepconv1	(SeparableConv2 (None, 16, 16, 728)	188672	block4_sepconv1_act[0][0]
block4_sepconv1_bn	(BatchNormal (None, 16, 16, 728)	2912	block4_sepconv1[0][0]
block4_sepconv2_act	(Activation (None, 16, 16, 728)	0	block4_sepconv1_bn[0][0]
block4_sepconv2	(SeparableConv2 (None, 16, 16, 728)	536536	block4_sepconv2_act[0][0]
block4_sepconv2_bn	(BatchNormal (None, 16, 16, 728)	2912	block4_sepconv2[0][0]
conv2d_50	(Conv2D) (None, 8, 8, 728)	186368	add_145[0][0]
block4_pool	(MaxPooling2D) (None, 8, 8, 728)	0	block4_sepconv2_bn[0][0]
batch_normalization_50	(BatchNo (None, 8, 8, 728)	2912	conv2d_50[0][0]
add_146	(Add) (None, 8, 8, 728)	0	block4_pool[0][0] batch normalization 50[0][0]

block5_sepconv1_act	(Activation (None, 8, 8, 728))	0	add_146[0][0]
block5_sepconv1	(SeparableConv2 (None, 8, 8, 728))	536536	block5_sepconv1_act[0][0]
block5_sepconv1_bn	(BatchNormal (None, 8, 8, 728))	2912	block5_sepconv1[0][0]
block5_sepconv2_act	(Activation (None, 8, 8, 728))	0	block5_sepconv1_bn[0][0]
block5_sepconv2	(SeparableConv2 (None, 8, 8, 728))	536536	block5_sepconv2_act[0][0]
block5_sepconv2_bn	(BatchNormal (None, 8, 8, 728))	2912	block5_sepconv2[0][0]
block5_sepconv3_act	(Activation (None, 8, 8, 728))	0	block5_sepconv2_bn[0][0]
block5_sepconv3	(SeparableConv2 (None, 8, 8, 728))	536536	block5_sepconv3_act[0][0]
block5_sepconv3_bn	(BatchNormal (None, 8, 8, 728))	2912	block5_sepconv3[0][0]
add_147 (Add)	(None, 8, 8, 728)	0	block5_sepconv3_bn[0][0] add_146[0][0]
block6_sepconv1_act	(Activation (None, 8, 8, 728))	0	add_147[0][0]
block6_sepconv1	(SeparableConv2 (None, 8, 8, 728))	536536	block6_sepconv1_act[0][0]
block6_sepconv1_bn	(BatchNormal (None, 8, 8, 728))	2912	block6_sepconv1[0][0]
block6_sepconv2_act	(Activation (None, 8, 8, 728))	0	block6_sepconv1_bn[0][0]
block6_sepconv2	(SeparableConv2 (None, 8, 8, 728))	536536	block6_sepconv2_act[0][0]
block6_sepconv2_bn	(BatchNormal (None, 8, 8, 728))	2912	block6_sepconv2[0][0]
block6_sepconv3_act	(Activation (None, 8, 8, 728))	0	block6_sepconv2_bn[0][0]
block6_sepconv3	(SeparableConv2 (None, 8, 8, 728))	536536	block6_sepconv3_act[0][0]
▶			
block13_sepconv1	(SeparableConv (None, 8, 8, 728))	536536	block13_sepconv1_act[0][0]
block13_sepconv1_bn	(BatchNorma (None, 8, 8, 728))	2912	block13_sepconv1[0][0]
block13_sepconv2_act	(Activatio (None, 8, 8, 728))	0	block13_sepconv1_bn[0][0]
block13_sepconv2	(SeparableConv (None, 8, 8, 1024))	752024	block13_sepconv2_act[0][0]
block13_sepconv2_bn	(BatchNorma (None, 8, 8, 1024))	4096	block13_sepconv2[0][0]
conv2d_51 (Conv2D)	(None, 4, 4, 1024)	745472	add_154[0][0]
block13_pool	(MaxPooling2D) (None, 4, 4, 1024)	0	block13_sepconv2_bn[0][0]
batch_normalization_51	(BatchNo (None, 4, 4, 1024))	4096	conv2d_51[0][0]
add_155 (Add)	(None, 4, 4, 1024)	0	block13_pool[0][0] batch_normalization_51[0][0]
block14_sepconv1	(SeparableConv (None, 4, 4, 1536))	1582080	add_155[0][0]
block14_sepconv1_bn	(BatchNorma (None, 4, 4, 1536))	6144	block14_sepconv1[0][0]
block14_sepconv1_act	(Activatio (None, 4, 4, 1536))	0	block14_sepconv1_bn[0][0]
block14_sepconv2	(SeparableConv (None, 4, 4, 2048))	3159552	block14_sepconv1_act[0][0]
block14_sepconv2_bn	(BatchNorma (None, 4, 4, 2048))	8192	block14_sepconv2[0][0]
block14_sepconv2_act	(Activatio (None, 4, 4, 2048))	0	block14_sepconv2_bn[0][0]
=====			
Total params: 20,861,480			
Trainable params: 0			
Non-trainable params: 20,861,480			
(None, 128, 128, 3)			

```

▶ # Преобразуем изображение
inp = Input(shape=(None, None, 3))
out = Lambda(lambda image: ktf.image.resize(image, (128, 128)))(inp)

inputLayer = Model(inputs=inp, outputs=out, name="resizer")
inputLayer.summary()

```

Model: "resizer"

Layer (type)	Output Shape	Param #
input_12 (InputLayer)	[(None, None, None, 3)]	0
lambda_11 (Lambda)	(None, 128, 128, 3)	0
Total params: 0		
Trainable params: 0		
Non-trainable params: 0		

```

▶ # Создание модели составной сети
model = Sequential()
# Добавляем слой для преобразования размера изображения
model.add(inputLayer)
# Добавляем сверточные слои
model.add(xception)
# Преобразуем двумерный массив Xception в одномерный
model.add(Flatten())
# Полносвязный слой
model.add(Dense(256, activation='relu'))
# Слой регуляризации (для предотвращения переобучения)
model.add(Dropout(0.5))
# Кол-во классов
model.add(Dense(num_classes, activation='softmax'))

# Собираем модель
model.build(input_shape)
model.summary()

```

Model: "sequential\_12"

Layer (type)	Output Shape	Param #
resizer (Functional)	(None, 128, 128, 3)	0
module_wrapper_12 (ModuleWra	(None, 4, 4, 2048)	20861480
flatten_11 (Flatten)	(None, 32768)	0
dense_22 (Dense)	(None, 256)	8388864
dropout_11 (Dropout)	(None, 256)	0
dense_23 (Dense)	(None, 10)	2570
Total params: 29,252,914		
Trainable params: 8,391,434		
Non-trainable params: 20,861,480		

```

[ ] # Компилируем составную сеть
epochs = 5
model.compile(loss='categorical_crossentropy',
              optimizer=Adam(lr=1e-5),
              metrics=['accuracy'])

```

```
[ ] model.fit(x_train, y_train, validation_data=(x_test, y_test), epochs = 6, batch_size=batch_size)
```

```
Epoch 1/6
1875/1875 [=====] - 106s 55ms/step - loss: 1.6371 - accuracy: 0.4794 - val_loss: 0.7300 - val_accuracy: 0.8429
Epoch 2/6
1875/1875 [=====] - 108s 58ms/step - loss: 0.7638 - accuracy: 0.7883 - val_loss: 0.5194 - val_accuracy: 0.8768
Epoch 3/6
1875/1875 [=====] - 101s 54ms/step - loss: 0.6025 - accuracy: 0.8299 - val_loss: 0.4305 - val_accuracy: 0.8892
Epoch 4/6
1875/1875 [=====] - 101s 54ms/step - loss: 0.5142 - accuracy: 0.8502 - val_loss: 0.3788 - val_accuracy: 0.9047
Epoch 5/6
1875/1875 [=====] - 101s 54ms/step - loss: 0.4621 - accuracy: 0.8651 - val_loss: 0.3533 - val_accuracy: 0.9049
Epoch 6/6
1875/1875 [=====] - 101s 54ms/step - loss: 0.4144 - accuracy: 0.8773 - val_loss: 0.3252 - val_accuracy: 0.9125
(60000, 28, 28, 3)
```

```
# Final evaluation of the model
scores = model.evaluate(x_test, y_test, verbose=0)
print("Accuracy: %.2f%%" % (scores[1]*100))
```

```
Accuracy: 91.25%
```