Final Project

- Hallym Univ. 인공지능개론 최종 과제용 파일입니다
- 제출/발표: 2018/6/22 금 12:30 발표
- 조원 1 박상범/ 조원 2 전찬혁
- Data set 개수 : 총 3007개
- 촬영 장소 및 일시
 - 。 산악 협력관
 - 2018/6/21

```
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import tensorflow as tf
from tensorflow.python.framework import ops
from dd_nnutil_hallym1 import *
import random
import time
```

▼ Dataset 업로드 확인

['0', '1', '2', '3', '4', '5']

위와 다르게 출력이 된다면 애초에 상위 폴더이름이 dataset1 인지 확인해보세요.

▼ Data 검토해보기

Data의 종류. 현재는 6가지의 data를 사용하므로 6을 사용합니다.

```
nclasses = 6
```

그려볼 이미지 index. 아래의 숫자를 바꾸어 가며 아래 cell에 그림이 제대로 표시가 되는지 확인 해보세요. 최종적으로는 직접 생성한 이미지가 display되도록 idx를 설정하세요

```
idx= 152

fig, ax = plt.subplots(figsize=(20, 10), dpi=80)
for i in range(nclasses):
```

```
img1, ntot = load_image_test(folder=folder1, img_class=i, idx=idx)
   print('class', i, '--', ntot)
   img1c = centered_crop(img1, output_side_length=128)
   plt.subplot(1,nclasses,i+1)
   plt.title('Class {} - #{}'.format(i, idx))
   plt.imshow(img1c)
('class', 0, '--', 500)
     ('class', 1, '--', 500)
     ('class', 2, '--', 500)
     ('class', 3, '--', 500)
     ('class', 4, '--', 500)
     ('class', 5, '--', 500)
                 Class 0 - #152
                                                   Class 1 - #152
                                                                                     Class 2 - #152
        0
                                         0
                                                                           0
       20
                                         20
                                                                          20
       40
                                         40
                                                                          40
       60
                                         60
                                                                          60
       80
                                         80
                                                                          80
      100
                                        100
                                                                         100
      120
                                        120
                                                                         120
```

100

Hyper-parameters

learning_rate=0.01 num_epochs=50 minibatch_size=32

▼ 본격적으로 시작 - Data 로드 하기

```
X_train, Y_train_orig, X_test, Y_test_orig = \(\frac{1}{2}\) load_dataset(folder=folder1, nclasses=nclasses)

\(\frac{1}{2}\) ('./dataset2/0/*.JPG', '-->', (500,)) ('./dataset2/1/*.JPG', '-->', (500,)) ('./dataset2/2/*.JPG', '-->', (500,)) ('./dataset2/3/*.JPG', '-->', (500,)) ('./dataset2/4/*.JPG', '-->', (500,)) ('./dataset2/5/*.JPG', '-->', (500,)) ('./dataset2/5/*.JPG', '-->', (500,))

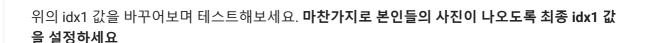
일반 숫자를 one-hot encoding으로!
```

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```
Y_train = convert_to_one_hot(Y_train_orig, nclasses).T
Y_test = convert_to_one_hot(Y_test_orig, nclasses).T
Data shape 살펴보기. Dimension을 살펴보세요
```

```
print ("number of training examples = " + str(X_train.shape[0]))
print ("number of test examples = " + str(X_test.shape[0]))
print ("X_train shape: " + str(X_train.shape))
```

```
print ("Y_train shape: " + str(Y_train.shape))
print ("X_test shape: " + str(X_test.shape))
print ("Y_test shape: " + str(Y_test.shape))
    number of training examples = 2100
      number of test examples = 900
      X_train shape: (2100, 128, 128, 3)
      Y_train shape: (2100, 6)
      X_test shape: (900, 128, 128, 3)
      Y_test shape: (900, 6)
# display images
idx1 = 250
x1 = X_{train}[idx1]
y1 = Y_{train[idx1]}
plt.figure()
plt.imshow(x1)
plt.title('idx #{}-- class #{}'.format(idx1. v1))
     Text(0.5,1,u'idx #250-- class #[1. 0. 0. 0. 0. 0.]')
               idx #250-- class #[1. 0. 0. 0. 0. 0.]
         0
        20
        40
```



120

▼ Placeholders 만들기

60

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100

120

0

20

40

60

80

100

```
def create_placeholders(n_H0, n_W0, n_C0, n_y):
    X = tf.placeholder(tf.float32, [None, n_H0, n_W0, n_C0])
    Y = tf.placeholder(tf.float32, [None, n_y])
    return X, Y
```

▼ Parameter 초기화 하기

- W1, W2의 크기, 갯수를 원하는 대로 변경하세요.
- W3, W4.. 등이 필요한 경우 자유롭게 넣어보세요

return parameters

▼ Forward propagation

• 아래는 다음과 같은 ConvNet을 구현한 것입니다. 자유롭게 stride, pooling의 ksize 변경해 보세요.

```
CONV2D -> RELU -> MAXPOOL -> CONV2D -> RELU -> MAXPOOL -> FLATTEN -> FC
```

• CONV2D -> RELU -> MAXPOOL 을 하나의 덩어리로 생각하면 좋습니다

```
def forward_propagation(X, parameters):
    W1 = parameters['W1']
    W2 = parameters['W2']
    # CONV2D: stride of 1, padding 'SAME'
    Z1 = tf.nn.conv2d(X, W1, strides=[1, 1, 1, 1], padding='SAME')
    # RELU
    A1 = tf.nn.relu(Z1)
    # MAXPOOL: window 8x8, sride 8, padding 'SAME'
    P1 = tf.nn.max_pool(A1, ksize=[1, 8, 8, 1], strides=[1, 8, 8, 1], padding='SAME')
    # CONV2D: filters W2, stride 1, padding 'SAME'
Z2 = tf.nn.conv2d(P1, W2, strides=[1, 1, 1, 1], padding='SAME')
    # RELU
    A2 = tf.nn.relu(Z2)
    # MAXPOOL: window 4x4, stride 4, padding 'SAME'
    P2 = tf.nn.max_pool(A2, ksize=[1, 4, 4, 1], strides=[1, 4, 4, 1], padding='SAME')
    # FLATTEN
    P2 = tf.contrib.layers.flatten(P2)
    # FULLY-CONNECTED without non-linear activation function (not not call softmax).
    # 6 neurons in output layer. Hint: one of the arguments should be "activation_fn=None"
    Z3 = tf.contrib.layers.fully_connected(P2, 6, activation_fn=None)
    Y_hat = Z3
    return Y_hat
```

▼ Cost 계산하기

```
def compute_cost(Y_hat, Y):
    # cost = tf.nn.softmax_cross_entropy_with_logits(logits = Y_hat, labels = Y)
    cost = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits_v2(logits=Y_hat, labels=Y_return_cost
```

▼ Model

지금까지 함수로 만들었던 기능들을 불러 만들어 네트워크를 구성합니다

```
(m, n_H0, n_W0, n_C0) = X_train.shape
n_y = Y_train.shape[1]

ops.reset_default_graph()
X, Y = create_placeholders(n_H0, n_W0, n_C0, n_y)
parameters = initialize_parameters()
Y_hat = forward_propagation(X, parameters)
cost = compute_cost(Y_hat, Y)
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate).minimize(cost)
```

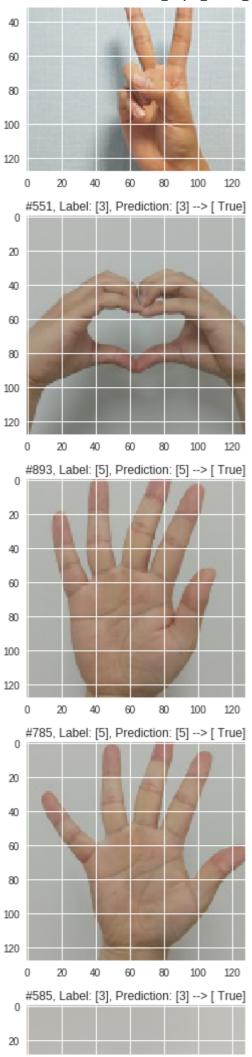
▼ Start the session to compute the tensorflow graph

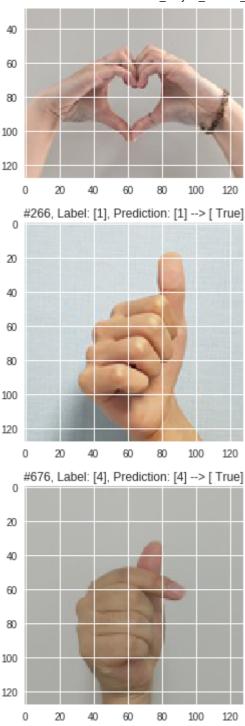
```
print_cost = True
vcosts = [] # cost를 저장할 빈 list
vtime = [] # 연산시간을 기록할 빈 list
# Reset the graph
#tf.reset_default_graph()
# Start interactive session
sess = tf.InteractiveSession()
seed = 0
sess.run(tf.global_variables_initializer())
# Do the training loop
for epoch in range(num_epochs):
   minibatch_cost = 0.
   num_minibatches = int(m / minibatch_size) # number of minibatches of size minibatch_si
   seed = seed + 1
   minibatches = random_mini_batches(X_train, Y_train, minibatch_size, seed=seed)
    for minibatch in minibatches:
        # Select a minibatch
        (minibatch_X, minibatch_Y) = minibatch
       # IMPORTANT: The line that runs the graph on a minibatch.
       # Run the session to execute the optimizer and the cost, the feedict should contain
       #_, temp_cost = sess.run([optimizer, cost], feed_dict={X: minibatch_X, Y: minibatch
       t0 = time.time()
       sess.run(optimizer, feed_dict={X: minibatch_X, Y: minibatch_Y})
       t_elapsed = time.time() - t0
       vtime.append(t_elapsed) # 시간을 측정하고 이를 list에 저장함 (append)
        temp_cost = sess.run(cost, feed_dict={X: minibatch_X, Y: minibatch_Y})
       minibatch_cost += temp_cost / num_minibatches
   # Print the cost every epoch
    if print_cost == True and epoch % 5 == 0:
       print("Cost after epoch %i: %f" % (epoch, minibatch_cost))
   vcosts.append(minibatch_cost)
Cost after epoch 0: 1.827218
     Cost after epoch 5: 0.173040
     Cost after epoch 10: 0.066017
     Cost after epoch 15: 0.063294
     Cost after epoch 20: 0.044596
     Cost after epoch 25: 0.022372
     Cost after epoch 30: 0.021404
     Cost after epoch 35: 0.023178
     Cost after epoch 40: 0.016705
     Cost after epoch 45: 0.003797
correct_prediction = tf.equal(tf.argmax(Y_hat, 1), tf.argmax(Y, 1))
accuracy = tf.reduce_mean(tf.cast(correct_prediction, "float"))
train_accuracy = accuracy.eval({X: X_train, Y: Y_train})
test_accuracy = accuracy.eval({X: X_test, Y: Y_test})
print("Train Accuracy:", train_accuracy)
```

```
Final Project 717005 2018 Spring Student v2.ipynb - Colaboratory
print("Test Accuracy:", test_accuracy)
print("Mean time to train for each batch: {:.3f} sec / batch size : {}".format(np.mean(vtim
•••
학습된 모델로 예측해보기 (correct predictions)
    • 10개 테스트 해보기 --> 자유롭게 원하는 대로 변경하여 테스트 해보세요
ntest = 10
for j in range(0,ntest):
    # Get one and predict
    r = random.randint(0, X_test.shape[0] - 1)
#print('Picked {} / {}'.format(r, X_test.shape[0]))
    v1 = sess.run(tf.argmax(Y_test[r:r+1], 1))
    t0 = time.time()
    v2 = sess.run(tf.argmax(Y_hat, 1), feed_dict={X: X_test[r:r+1]})
    t_elapsed = time.time() - t0
    bok = v1 == v2
    str1 = '#{}, Label: {}, Prediction: {}, {}, Time : {:.3f} sec'.format(r, v1, v2,bok, t_
    print(str1)
    str2 = '#{}, Label: {}, Prediction: {} --> {}'.format(r, v1, v2, bok)
    plt.figure()
    plt.imshow(X_test[r])
    plt.title(str2)
```

С→

sess.close()





▼ Cost 그려보기

```
plt.figure()
plt.plot(np.squeeze(vcosts))
plt.ylabel('cost')
plt.xlabel('iterations (per tens)')
plt.title("Learning rate =" + str(learning_rate))
```

 \Box

Text(0.5,1,u'Learning rate =0.01')



Cost 의 초반부 확대해서 그려보기

```
plt.figure()
plt.plot(np.squeeze(vcosts[:20]))
plt.ylabel('cost')
plt.xlabel('iterations (per tens)')
plt.title("Learning rate =" + str(learning_rate))
plt.show()
```

