11/18/21, 8:39 AM

```
In [1]:
         # data augmentation test: rotate different degree (pay attention to adjustable filename etc.)
         import numpy as np
         import matplotlib.pyplot as plt
         import tensorflow as tf
         from tensorflow.python.keras.layers import Dense, GlobalAveragePooling2D
         from tensorflow.python.keras.models import Model
         from tensorflow.python.keras import layers, Sequential, losses, metrics
         image\ height = 48
         image width = 48
         emotions count = 8
         emotion_labels = ['neutral', 'happiness', 'surprise', 'sadness',
                            'anger', 'disgust', 'fear', 'contempt']
         # !!! change sample size
         samples = 99109 # 2~99110
         training samples = 28317*3 # (Training)
         validation samples = 3541*3 # (PublicTest)
         test samples = 3535
                                 # (PrivateTest)
         #!!! change npy folder name
         image path = "./dataset r40/images.npy"
         emotion multi path = "./dataset r40/emotions multi.npy"
         emotion single path = "./dataset r40/emotions single.npy"
         images = np.load(image path)
         emotions multi = np.load(emotion multi path)
         emotions single = np.load(emotion single path)
         # !!! change s/m dataset
         #emotions = emotions single
         emotions = emotions multi
         print(images.shape)
         print(emotions multi.shape)
         print(emotions_single.shape)
        (99109, 48, 48, 1)
        (99109, 8)
        (99109, 8)
In [2]:
         cce = losses.CategoricalCrossentropy()
         mse = losses.MeanSquaredError()
```

```
tf.config.run functions eagerly(True)
         def model acc(y true, y pred):
             size = y true.shape[0]
             acc = 0
             for i in range(size):
                 true = y true[i]
                 pred = y pred[i]
                 index max = tf.argmax(pred).numpy()
                 if true[index max].numpy()==tf.reduce max(true).numpy():
             return acc/size
In [3]:
         images = tf.convert to tensor(images)
         #images = tf.image.grayscale to rgb(images)
         emotions = tf.convert to tensor(emotions)
         #images = tf.image.resize(images, [224,224])
         images = layers.Rescaling(1./127.5, offset= -1)(images)
         training size = training samples + validation samples
         test size = test samples
         training images = images[:training size]
         test images = images[training size:]
         training emotions = emotions[:training size]
         test emotions = emotions[training size:]
         print("training images shape:", training images.shape)
         print("training emotions shape:", training emotions.shape)
         print("test images shape:", test images.shape)
         print("test emotions shape:", test emotions.shape)
        training images shape: (95574, 48, 48, 1)
        training emotions shape: (95574, 8)
        test images shape: (3535, 48, 48, 1)
        test emotions shape: (3535, 8)
In [4]:
         from tensorflow.python.keras.applications import vgg16, resnet v2
         from tensorflow.python.keras import optimizers
         from tensorflow.python.keras.optimizer v2 import adam
In [5]:
         base model = vgg16.VGG16(include top=False,
                                  weights="imagenet",
```

11/18/21, 8:39 AM model_rotate

```
input shape=(48,48,3))
base model.trainable=True
model = Sequential([
  base model,
  layers.GlobalAveragePooling2D(),
  layers.Dense(4096, activation='relu'),
  layers.Dense(4096, activation='relu'),
  layers.Dense(emotions count, activation='softmax'),
1)
model.compile(optimizer=adam.Adam(learning rate=1e-4),
        loss=mse,
        metrics = [model acc])
model.fit(x=tf.image.grayscale to rgb(training images),
      y=training emotions,
      batch size=32,
      epochs=40,
      validation data=(tf.image.grayscale to rgb(test images), test emotions))
C:\Users\Darkl\anaconda3\lib\site-packages\tensorflow\python\data\ops\dataset ops.py:3703: UserWarning: Even though
the `tf.config.experimental run functions eagerly` option is set, this option does not apply to tf.data functions. T
o force eager execution of tf.data functions, please use `tf.data.experimental.enable.debug mode()`.
 warnings.warn(
Epoch 1/40
val model acc: 0.7489
Epoch 2/40
val model acc: 0.7864
Epoch 3/40
val model acc: 0.8031
Epoch 4/40
val model acc: 0.8245
Epoch 5/40
val model acc: 0.8284
Epoch 6/40
val model acc: 0.8236
Epoch 7/40
val model acc: 0.8326
```

```
Epoch 8/40
val model acc: 0.8318
Epoch 9/40
val model acc: 0.8313
Epoch 10/40
val model acc: 0.8354
Epoch 11/40
val model acc: 0.8315
Epoch 12/40
val model acc: 0.8351
Epoch 13/40
val model acc: 0.8366
Epoch 14/40
val model acc: 0.8332
Epoch 15/40
val model acc: 0.8428
Epoch 16/40
val model acc: 0.8428
Epoch 17/40
val model acc: 0.8442
Epoch 18/40
val model acc: 0.8447
Epoch 19/40
val model acc: 0.8427
Epoch 20/40
val model acc: 0.8422
Epoch 21/40
val model acc: 0.8399
Epoch 22/40
val model acc: 0.8397
Epoch 23/40
```

11/18/21, 8:39 AM model_rotate

```
val model acc: 0.8387
Epoch 24/40
val model acc: 0.8438
Epoch 25/40
val model acc: 0.8325
Epoch 26/40
val model acc: 0.8385
Epoch 27/40
val model acc: 0.8312acc: 0
Epoch 28/40
val model acc: 0.8438
Epoch 29/40
val model acc: 0.8416
Epoch 30/40
val model acc: 0.8438
Epoch 31/40
val model acc: 0.8419
Epoch 32/40
val model acc: 0.8410
Epoch 33/40
val model acc: 0.8368
Epoch 34/40
val model acc: 0.8481
Epoch 35/40
8 - val model acc: 0.8484
Epoch 36/40
1 - val model acc: 0.8396
Epoch 37/40
val model acc: 0.8405
Epoch 38/40
```

```
8 - val model acc: 0.8430
       Epoch 39/40
       1 - val model_acc: 0.8391
       Epoch 40/40
       9 - val model acc: 0.8405
       <tensorflow.python.keras.callbacks.History at 0x1f700a74d00>
Out[5]:
In [6]:
        image path = "./dataset rran/images.npy"
        emotion multi path = "./dataset rran/emotions multi.npy"
        emotion single path = "./dataset rran/emotions single.npy"
        images = np.load(image path)
        emotions multi = np.load(emotion multi path)
        emotions single = np.load(emotion single path)
        # !!! change s/m dataset
        #emotions = emotions single
        emotions = emotions multi
        print(images.shape)
        print(emotions multi.shape)
        print(emotions single.shape)
        images = tf.convert_to_tensor(images)
        #images = tf.image.grayscale to rgb(images)
        emotions = tf.convert to tensor(emotions)
        #images = tf.image.resize(images, [224,224])
        images = layers.Rescaling(1./127.5, offset= -1)(images)
        training size = training samples + validation samples
        test size = test samples
        training images = images[:training size]
        test images = images[training size:]
        training emotions = emotions[:training size]
        test emotions = emotions[training size:]
        print("training images shape:", training images.shape)
        print("training emotions shape:", training emotions.shape)
        print("test_images shape:", test_images.shape)
        print("test_emotions shape:", test_emotions.shape)
        base model = vgg16.VGG16(include top=False,
                              weights="imagenet",
                              input shape=(48,48,3))
        base model.trainable=True
```