```
In [1]:
         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']
         #data augmentation: mirror version
         samples = 67251 # 2~67252
         training samples = 28317*2 # 2~56635 (Training)
         validation samples = 3541*2 # 56636~63717 (PublicTest)
         test samples = 3535  # 63718~67252 (PrivateTest)
         image path = "./dataset/images.npy"
         emotion multi path = "./dataset/emotions multi.npy"
         emotion single path = "./dataset/emotions single.npy"
In [2]:
         images = np.load(image path)
         emotions multi = np.load(emotion multi path)
         emotions single = np.load(emotion single path)
         print(images.shape)
         print(emotions multi.shape)
         print(emotions single.shape)
         (67251, 48, 48, 1)
         (67251, 8)
        (67251, 8)
In [3]:
         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():
                     acc += 1
             return acc/size
In [4]:
         #emotions = emotions single
         emotions = emotions multi
         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: (63716, 48, 48, 3)
        training emotions shape: (63716, 8)
        test images shape: (3535, 48, 48, 3)
        test emotions shape: (3535, 8)
In [5]:
         from tensorflow.python.keras.applications import vgg16, resnet v2
         from tensorflow.python.keras import optimizers
         from tensorflow.python.keras.optimizer v2 import adam
In [6]:
         base_model = vgg16.VGG16(include_top=False,
                                  weights="imagenet",
                                  input_shape=(48,48,3))
```

```
base model.trainable=True
model = Sequential([
   base model,
  lavers.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=losses.CategoricalCrossentropy(),
        metrics = [tf.keras.metrics.TopKCategoricalAccuracy(k=2)])
model.fit(x=training images,
      v=training emotions,
      batch size=32,
      epochs=40,
      validation data=(test images, test emotions))
C:\Users\Darkl\anaconda3\lib\site-packages\tensorflow\python\data\ops\dataset ops.py:3703: UserWarning: Even though the `tf.c
onfig.experimental run functions eagerly` option is set, this option does not apply to tf.data functions. To force eager exec
ution of tf.data functions, please use `tf.data.experimental.enable.debug mode()`.
 warnings.warn(
Epoch 1/40
9159 - val top k categorical accuracy: 0.9191
Epoch 2/40
8717 - val top k categorical accuracy: 0.9364
Epoch 3/40
8667 - val top k categorical accuracy: 0.9355
Epoch 4/40
8439 - val top k categorical accuracy: 0.9454
Epoch 5/40
8314 - val top k categorical accuracy: 0.9508
Epoch 6/40
8394 - val top k categorical accuracy: 0.9508
Epoch 7/40
8406 - val top k categorical accuracy: 0.9482
```

```
Epoch 8/40
8443 - val top k categorical accuracy: 0.9530
Epoch 9/40
8674 - val top k categorical accuracy: 0.9485
Epoch 10/40
8920 - val top k categorical accuracy: 0.9454
Epoch 11/40
9148 - val top k categorical accuracy: 0.9511
Epoch 12/40
9158 - val top k categorical accuracy: 0.9468
Epoch 13/40
9558 - val top k categorical accuracy: 0.9485
Epoch 14/40
9516 - val top k categorical accuracy: 0.9499
Epoch 15/40
9720 - val top k categorical accuracy: 0.9477
Epoch 16/40
9877 - val top k categorical accuracy: 0.9502
Epoch 17/40
9913 - val top k categorical accuracy: 0.9502
Epoch 18/40
9986 - val top k categorical accuracy: 0.9454
Epoch 19/40
0079 - val top k categorical accuracy: 0.9468
Epoch 20/40
0360 - val top k categorical accuracy: 0.9522
Epoch 21/40
0244 - val top k categorical accuracy: 0.9513
Epoch 22/40
```

```
0301 - val top k categorical accuracy: 0.9511
Epoch 23/40
0506 - val top k categorical accuracy: 0.9479
Epoch 24/40
0201 - val top k categorical accuracy: 0.9528
Epoch 25/40
0365 - val top k categorical accuracy: 0.9533
Epoch 26/40
0446 - val top k categorical accuracy: 0.9499
Epoch 27/40
0319 - val top k categorical accuracy: 0.9511
Epoch 28/40
0376 - val top k categorical accuracy: 0.9522
Epoch 29/40
0628 - val top k categorical accuracy: 0.9519
Epoch 30/40
0493 - val top k categorical accuracy: 0.9539
Epoch 31/40
0768 - val top k categorical accuracy: 0.9519
Epoch 32/40
0414 - val top k categorical accuracy: 0.9522
Epoch 33/40
0758 - val top k categorical accuracy: 0.9494
Epoch 34/40
0628 - val top k categorical accuracy: 0.9530
Epoch 35/40
0594 - val top k categorical_accuracy: 0.9516
Epoch 36/40
0806 - val top k categorical accuracy: 0.9454
Epoch 37/40
```

```
0939 - val top k categorical accuracy: 0.9494
    Epoch 38/40
    1154 - val top k categorical accuracy: 0.9513
    Epoch 39/40
    0534 - val top k categorical accuracy: 0.9522
    Epoch 40/40
    0866 - val top k categorical accuracy: 0.9471
    <tensorflow.python.keras.callbacks.History at 0x22bf8ccc790>
Out[6]:
In [ ]:
    base model = resnet v2.ResNet50V2(include top=False,
                 weights="imagenet",
                 input shape=(48,48,3))
    base model.trainable=True
    model = Sequential([
      base model,
      layers.GlobalAveragePooling2D(),
      layers.Dense(2048, activation='relu'),
      layers.Dense(2048, activation='relu'),
      layers.Dense(emotions count, activation='softmax'),
    1)
    model.compile(optimizer=adam.Adam(learning rate=1e-4),
            loss=losses.CategoricalCrossentropy(),
            metrics = [model accl)
    model.fit(x=training images,
          y=training emotions,
          batch size=32,
          epochs=40,
          validation data=(test images, test emotions))
    Epoch 1/40
    1 acc: 0.7386
    Epoch 2/40
    l acc: 0.7710
    Epoch 3/40
```

```
l acc: 0.7831
 Epoch 4/40
 1 acc: 0.7881
 Epoch 5/40
 1 acc: 0.7998
 Epoch 6/40
 1 acc: 0.7972
 Epoch 7/40
 1 acc: 0.8084
 Epoch 8/40
 1 acc: 0.8067
 Epoch 9/40
 1 acc: 0.7955
 Epoch 10/40
 l acc: 0.8112
 Epoch 11/40
 l acc: 0.8051
 Epoch 12/40
 1 acc: 0.8030
 Epoch 13/40
 l acc: 0.8138
 Epoch 14/40
 l acc: 0.8126
 Epoch 15/40
 l acc: 0.8188
 Epoch 16/40
  633/1992 [======>...... - ETA: 3:26 - loss: 0.5951 - model acc: 0.9481
In [ ]:
```