# In [1]:

import csv, json, random
import numpy as np
import matplotlib.pyplot as plt
import tensorflow as tf

### In [4]:

```
samples = 35393 # 2~35394
image_height = 48
image width = 48
emotions count = 8
images = []
emotions = []
emotion_labels = []
with open('./dataset.csv') as file:
   reader = csv.reader(file)
   for line in reader:
        image_pixels = line[-1].split()
        if len(image_pixels) == 1:
            emotion_labels = line[2:2+emotions_count]
            assert(emotion_labels == ['neutral', 'happiness', 'surprise', 'sadness', 'anger
            continue
        image = []
        for i in range(image_height):
            row = []
            for j in range(image_width):
                row.append(image_pixels[image_width*i+j])
            row = list(map(int, row))
            image.append(row)
        images.append(image)
        emotion = []
        for i in range(2, 2+emotions_count):
            emotion.append(line[i])
        emotion = list(map(float, emotion))
        emotions.append(emotion)
images = np.array(images).reshape(samples, image height, image width, 1)
emotions = np.array(emotions)
print("images shape:", images.shape)
print("emotions shape:", emotions.shape)
print("emotion_labels:", emotion_labels)
# # check the last 9 images
# plt.figure(figsize=(10, 10))
# for i in range(35384,35393):
     ax = plt.subplot(3, 3, i-35383)
#
#
     plt.imshow(images[i].astype("uint8"))
     plt.title(emotion labels[np.argmax(emotions[i])])
     plt.axis("off")
images shape: (35393, 48, 48, 1)
emotions shape: (35393, 8)
emotion_labels: ['neutral', 'happiness', 'surprise', 'sadness', 'anger', 'di
sgust', 'fear', 'contempt']
```

## In [5]:

```
emotions
```

# Out[5]:

```
, 0.
              , 0.
                                    , ..., 0.2
array([[0.4
                                                    , 0.
                                                               , 0.
                                                                         ],
       [0.75
               , 0.
                          , 0.125
                                                              , 0.
                                    , ..., 0.
                                                    , 0.
                                                                         ],
       [0.555556, 0.
                          , 0.
                                                    , 0.
                                                               , 0.
                                     , ..., 0.
                                                                         ],
       ...,
       [0.
                , 0.
                          , 0.
                                                               , 0.2
                                     , ..., 0.1
                                                    , 0.
                          , 0.
                                                   , 0.
                , 1.
                                    , ..., 0.
                                                              , 0.
       [0.
       [0.222222, 0.
                          , 0.
                                    , ..., 0.111111, 0.
                                                               , 0.
                                                                         11)
```

# In [6]:

```
for i in range(emotions.shape[0]):
    max_indices = []
    maximum = np.amax(emotions[i]);
    for j in range(emotions.shape[1]):
        if emotions[i][j] == maximum:
            max_indices.append(j)
    no_of_maxs = len(max_indices)
    for j in range(emotions.shape[1]):
        if j in max_indices:
            emotions[i][j] = 1/no_of_maxs
        else:
            emotions[i][j] = 0
emotions
```

## Out[6]:

```
In [7]:
```

```
for i in range(10,30):
   print(emotions[i])
[0. 0. 0. 0. 1. 0. 0. 0.]
[1. 0. 0. 0. 0. 0. 0. 0.]
[1. 0. 0. 0. 0. 0. 0. 0.]
[1. 0. 0. 0. 0. 0. 0. 0.]
[0. 1. 0. 0. 0. 0. 0. 0.]
[0. 0. 1. 0. 0. 0. 0. 0.]
[0. 1. 0. 0. 0. 0. 0. 0.]
[0. 0. 0. 1. 0. 0. 0. 0.]
[1. 0. 0. 0. 0. 0. 0. 0.]
[1. 0. 0. 0. 0. 0. 0. 0.]
[0. 0. 0. 0. 0. 0. 1. 0.]
[0. 0. 0. 0. 1. 0. 0. 0.]
[0. 1. 0. 0. 0. 0. 0. 0.]
[0. 1. 0. 0. 0. 0. 0. 0.]
[0.5 0. 0.5 0. 0. 0. 0.
                             0. ]
[0. 0. 0. 0. 1. 0. 0. 0.]
[0. 1. 0. 0. 0. 0. 0. 0.]
[0. 0. 1. 0. 0. 0. 0. 0.]
[0. 1. 0. 0. 0. 0. 0. 0.]
[1. 0. 0. 0. 0. 0. 0. 0.]
In [8]:
1/2
Out[8]:
0.5
In [9]:
images = tf.convert_to_tensor(images)
emotions = tf.convert_to_tensor(emotions)
images = tf.image.grayscale_to_rgb(images)
print("images shape:", images.shape)
print("emotions shape:", emotions.shape)
images shape: (35393, 48, 48, 3)
emotions shape: (35393, 8)
In [10]:
# images = tf.image.resize(images, [224,224])
# print("images shape:", images.shape)
In [11]:
from tensorflow.python.keras.applications.vgg16 import preprocess_input
from tensorflow.python.keras import layers
# choose one method:
images = layers.Rescaling(1./127.5, offset= -1)(images)
#images = preprocess input(images)
```

#### In [12]:

```
training_samples = 28317  # 2~28318 (Training)
validation_samples = 3541  # 28319~31859 (PublicTest)
test_samples = 3535  # 31860~35394 (PrivateTest)

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: (31858, 48, 48, 3)
```

In [13]:

training\_emotions shape: (31858, 8) test\_images shape: (3535, 48, 48, 3)

test emotions shape: (3535, 8)

```
# # check the last 9 images
# plt.figure(figsize=(10, 10))
# for i in range(6):
# ax = plt.subplot(3, 3, i+1)
# plt.imshow(special_images[i].astype("uint8"))
# plt.axis("off")
```

# In [14]:

```
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
def loss_func(y_true, y_pred):
    size = y_true.shape[0]
    loss = 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():
            loss += 1
    return loss/size
```

## In [15]:

```
from tensorflow.python.keras.applications.vgg16 import VGG16
from tensorflow.python.keras.layers import Dense, GlobalAveragePooling2D
from tensorflow.python.keras.models import Model
from tensorflow.python.keras import layers, Sequential

vgg_model = VGG16(include_top=False, weights="imagenet", input_shape=(48,48,3))
vgg_model.trainable=False
model = Sequential([
    vgg_model,
    layers.GlobalAveragePooling2D(),
    layers.Dense(1024, activation='relu'),
    layers.Dense(emotions_count, activation='softmax'),
])
#model.summary()
```

## In [16]:

```
from tensorflow.python.keras import losses, metrics

#m = top_k_categorical_accuracy(y_true, y_pred, k=3)
#model.compile(optimizer='sgd', loss=losses.CategoricalCrossentropy(), metrics = [metrics.C model.compile(optimizer='adam', loss=losses.CategoricalCrossentropy(), metrics = [model_accuracy(), met
```

## In [17]:

```
model.fit(x=training images,
        y=training_emotions,
        batch_size=32,
        epochs=10,
        validation_data=(test_images, test_emotions))
```

C:\Users\Dark1\AppData\Roaming\Python\Python38\site-packages\tensorflow\pyth on\data\ops\dataset\_ops.py:4211: UserWarning: Even though the `tf.config.exp erimental\_run\_functions\_eagerly` option is set, this option does not apply t o tf.data functions. To force eager execution of tf.data functions, please u se `tf.data.experimental.enable\_debug\_mode()`.

```
warnings.warn(
```

```
Epoch 1/10
996/996 [========= ] - 865s 869ms/step - loss: 1.3427 -
model_acc: 0.5156 - val_loss: 1.2979 - val_model_acc: 0.5364
996/996 [============= ] - 860s 863ms/step - loss: 1.2241 -
model_acc: 0.5599 - val_loss: 1.2503 - val_model_acc: 0.5464
Epoch 3/10
996/996 [============== ] - 860s 864ms/step - loss: 1.1559 -
model_acc: 0.5877 - val_loss: 1.2163 - val_model_acc: 0.5628
Epoch 4/10
996/996 [=============== ] - 862s 866ms/step - loss: 1.0921 -
model acc: 0.6180 - val loss: 1.2049 - val model acc: 0.5757
Epoch 5/10
996/996 [=========== ] - 873s 876ms/step - loss: 1.0272 -
model_acc: 0.6416 - val_loss: 1.1879 - val_model_acc: 0.5745
996/996 [================ ] - 866s 870ms/step - loss: 0.9626 -
model_acc: 0.6666 - val_loss: 1.2157 - val_model_acc: 0.5802
Epoch 7/10
996/996 [============== ] - 865s 869ms/step - loss: 0.9017 -
model_acc: 0.6892 - val_loss: 1.2101 - val_model_acc: 0.5855
Epoch 8/10
996/996 [========== ] - 865s 869ms/step - loss: 0.8356 -
model_acc: 0.7151 - val_loss: 1.2166 - val_model_acc: 0.5942
996/996 [================ ] - 866s 869ms/step - loss: 0.7759 -
model_acc: 0.7394 - val_loss: 1.2362 - val_model_acc: 0.5880
Epoch 10/10
996/996 [============ ] - 866s 869ms/step - loss: 0.7180 -
model acc: 0.7620 - val loss: 1.3056 - val model acc: 0.5801
Out[17]:
```

<tensorflow.python.keras.callbacks.History at 0x23326616190>

## In [18]:

```
m = metrics.CategoricalAccuracy()
m.update_state([[0, 1, 1], [0, 1, 0]], [[0.1, 0.9, 0.98], [0.05, 0.95, 0]])
m.result().numpy()
```

## Out[18]:

0.5

In [ ]:		
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