Facial Expression Recognition with Keras

Task 1: Import Libraries

```
In [2]: import numpy as np
        import seaborn as sns
        import matplotlib.pyplot as plt
        import utils
        import os
        %matplotlib inline
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        from tensorflow.keras.layers import Dense, Input, Dropout, Flatten, Conv
        from tensorflow.keras.layers import BatchNormalization, Activation, Max
        Pooling2D
        from tensorflow.keras.models import Model, Sequential
        from tensorflow.keras.optimizers import Adam
        from tensorflow.keras.callbacks import ModelCheckpoint, ReduceLROnPlate
        from tensorflow.keras.utils import plot model
        from IPython.display import SVG, Image
        from livelossplot import PlotLossesTensorFlowKeras
        import tensorflow as tf
        print("Tensorflow version:", tf. version )
```

Tensorflow version: 2.1.0

Task 2: Plot Sample Image





```
datagen train = ImageDataGenerator(horizontal flip=True)
train generator = datagen train.flow from directory("train/", target si
ze=(img size,img size),
                                                     color mode='graysca
le',
                                                     batch size=batch si
ze,
                                                     class mode='categor
ical',
                                                     shuffle=True)
datagen validation = ImageDataGenerator(horizontal flip=True)
validation generator = datagen train.flow from directory("test/", targe
t size=(img size,img size),
                                                     color mode='graysca
le',
                                                     batch size=batch si
ze,
                                                     class mode='categor
ical',
                                                     shuffle=True)
# print(type(train generator))
# print(train generator)
```

Found 28709 images belonging to 7 classes. Found 7178 images belonging to 7 classes.

Task 4: Create CNN Model

Inspired by Goodfellow, I.J., et.al. (2013). Challenged in representation learning: A report of three machine learning contests. *Neural Networks*, 64, 59-63. doi:10.1016/j.neunet.2014.09.005

```
In [9]: model = Sequential()
```

```
# 1 - conv layer
# 64 filters, 3 by 3
model.add(Conv2D(64, (3,3), padding='same', input shape=(48,48,1)))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Dropout(0.25))
# 2 - conv layer
model.add(Conv2D(128, (5,5), padding='same'))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Dropout(0.25))
# 3 - conv layer
model.add(Conv2D(512, (3,3), padding='same'))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Dropout(0.25))
# 4 - conv layer
model.add(Conv2D(512, (3,3), padding='same'))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Dropout(0.25))
model.add(Flatten())
model.add(Dense(256))
model.add(BatchNormalization())
model.add(Activation('relu'))
model.add(Dropout(0.25))
model.add(Dense(512))
model.add(BatchNormalization())
model.add(Activation('relu'))
```

```
model.add(Dropout(0.25))
model.add(Dense(7, activation='softmax'))

opt = Adam(lr=0.0005)
model.compile(optimizer=opt, loss='categorical_crossentropy', metrics=[
'accuracy'])
model.summary()
```

Model: "sequential"

Layer (type)	Output	Shape	Param #
conv2d (Conv2D)	(None,	48, 48, 64)	640
batch_normalization (BatchNo	(None,	48, 48, 64)	256
activation (Activation)	(None,	48, 48, 64)	0
<pre>max_pooling2d (MaxPooling2D)</pre>	(None,	24, 24, 64)	0
dropout (Dropout)	(None,	24, 24, 64)	0
conv2d_1 (Conv2D)	(None,	24, 24, 128)	204928
<pre>batch_normalization_1 (Batch</pre>	(None,	24, 24, 128)	512
activation_1 (Activation)	(None,	24, 24, 128)	0
<pre>max_pooling2d_1 (MaxPooling2</pre>	(None,	12, 12, 128)	0
dropout_1 (Dropout)	(None,	12, 12, 128)	0
conv2d_2 (Conv2D)	(None,	12, 12, 512)	590336
<pre>batch_normalization_2 (Batch</pre>	(None,	12, 12, 512)	2048
activation_2 (Activation)	(None,	12, 12, 512)	0
<pre>max_pooling2d_2 (MaxPooling2</pre>	(None,	6, 6, 512)	0

dropout_2 (Dropout)	(None, 6	, 6, 512)	0
conv2d_3 (Conv2D)	(None, 6	, 6, 512)	2359808
batch_normalization_3 (Batch	(None, 6	, 6, 512)	2048
activation_3 (Activation)	(None, 6	, 6, 512)	0
max_pooling2d_3 (MaxPooling2	(None, 3	, 3, 512)	0
dropout_3 (Dropout)	(None, 3	, 3, 512)	0
flatten (Flatten)	(None, 4	608)	0
dense (Dense)	(None, 2	56)	1179904
batch_normalization_4 (Batch	(None, 2	56)	1024
activation_4 (Activation)	(None, 2	56)	0
dropout_4 (Dropout)	(None, 2	56)	0
dense_1 (Dense)	(None, 5	12)	131584
batch_normalization_5 (Batch	(None, 5	12)	2048
activation_5 (Activation)	(None, 5	12)	0
dropout_5 (Dropout)	(None, 5	12)	0
dense_2 (Dense)	(None, 7)	3591
Total params: 4,478,727 Trainable params: 4,474,759 Non-trainable params: 3,968			

In []:

Task 6: Train and Evaluate Model

```
In [10]:
          epochs = 15
          steps per epoch = train generator.n//train generator.batch size
          validation steps = validation generator.n//validation generator.batch s
          ize
          checkpoint = ModelCheckpoint("model weights.h5", monitor='val accuracy'
          , save weights only=True,
                                         mode='max', verbose=1)
          reduce lr = ReduceLROnPlateau(monitor='val loss', factor=0.1, patience=
          2, min lr=0.00001,
                                          model='auto')
          callbacks = [PlotLossesTensorFlowKeras(), checkpoint, reduce_lr]
          history = model.fit(x=train generator, steps per epoch=steps per epoch,
                               epochs=epochs,
                               validation data=validation generator,
                               validation steps=validation steps,
                               callbacks=callbacks
                        Log-loss (cost function)
                                                                   accuracy
          1.35
                                                 0.64
          1.30
                                                 0.62
          1.25
                                                 0.60
          1.20
                                                 0.58
          1.15
                                                 0.56
          1.10
                                                 0.54
          1.05
                                                 0.52
          1.00
                                                 0.50
          0.95
          Log-loss (cost function):
```

```
training
                0.949, max:
                           1.798, cur:
                                        0.949)
         (min:
validation (min:
                0.998, max:
                            1.705, cur:
                                        0.998)
accuracy:
training
         (min:
                0.309, max:
                            0.643, cur:
                                        0.643)
validation (min:
                0.355, max:
                            0.630, cur:
                                        0.625)
Epoch 00015: saving model to model weights.h5
87 - accuracy: 0.6431 - val loss: 0.9983 - val accuracy: 0.6253
```

Task 7: Represent Model as JSON String