```
Edit View Run Kernel Tabs Settings Help
          ModelTraining.ipynb
           + % □ □ ▶ ■ C → Code
                                                                                                                                                                                                                            Python 3
       8
0
              [1]: import numpy as np import pandas as pd
B
                      import tensorflow as tf
                      from keras.preprocessing.image import ImageDataGenerator, load_img
                      from keras.layers import Conv2D, Dense, BatchNormalization, Activation, Dropout, MaxPool2D, Flatten
°¢
                     from keras.optimizers import Adam, RMSprop, SGD
                     from keras import regularizers
                     from keras.callbacks import ModelCheckpoint
from keras.callbacks import ModelCheckpoint, CSVLogger, TensorBoard, EarlyStopping, ReduceLROnPlateau
                     {\color{red}\mathsf{import}}\ {\color{red}\mathsf{datetime}}
                      import matplotlib.pyplot as plt
                     from keras.utils import plot_model
              [2]: train_dir = './Dataset/train/
test_dir = './Dataset/test/'
                     row, col = 48, 48
                     input_size = row, col
               [3]: plt.figure(figsize=(14,22))
                      for expression in os.listdir(train_dir):
                          img = load_img((train_dir + expression +'/'+ os.listdir(train_dir + expression)[1]))
                          plt.imshow(img)
                          plt.title(expression)
                     i += 1
plt.show()
                                                                    fear
                                                                                                           neutral
                                                                                                                                 sad
                           angry
                                              disgust
                                                                                       happy
                                                                                                                                                   surprise
               [4]: train_datagen = ImageDataGenerator(rescale=1./255,
                                                                 zoom_range=0.3,
horizontal_flip=True)
                     training_set = train_datagen.flow_from_directory(train_dir,
                                                                                  batch size=64,
                                                                                  target_size=(48,48),
                                                                                 shuffle=True,
color_mode='grayscale',
class_mode='categorical')
                     test_datagen = ImageDataGenerator(rescale=1./255)
test_set = test_datagen.flow_from_directory(test_dir,
                                                                                  target_size=(48,48),
                                                                                 shuffle=True,
color_mode='grayscale',
class_mode='categorical')
                     Found 28709 images belonging to 7 classes. Found 7178 images belonging to 7 classes.
               [5]: training_set.class_indices
              [5]: {'angry': 0,
'disgust': 1,
'fear': 2,
'happy': 3,
'neutral': 4,
                       'sad': 5,
'surprise': 6}
               [6]: model= tf.keras.models.Sequential()
                     model.add(Conv2D(32, kernel_size=(3, 3), padding='same', activation='relu', input_shape=(48, 48,1)))
model.add(Conv2D(64,(3,3), padding='same', activation='relu'))
model.add(BatchNormalization())
                     model.add(MaxPool2D(pool_size=(2, 2)))
                     model.add(Dropout(0.25))
                     model.add(Conv2D(128,(5,5), padding='same', activation='relu'))
model.add(BatchNormalization())
                     model.add(MaxPool2D(pool_size=(2, 2)))
                     model.add(Dropout(0.25))
                     model. add (Conv2D (512, (3, 3), padding='same', activation='relu', kernel\_regularizer=regularizers. 12 (0.01)))
                     model.add(BatchNormalization())
model.add(MaxPool2D(pool_size=(2, 2)))
                     model.add(Dropout(0.25))
                     model.add(Conv2D(512,(3,3), padding='same', activation='relu', kernel_regularizer=regularizers.12(0.01)))
model.add(BatchNormalization())
                     model.add(MaxPool2D(pool_size=(2, 2)))
                     model.add(Dropout(0.25))
                     model.add(Flatten())
                     model.add(Dense(256,activation = 'relu'))
model.add(BatchNormalization())
                      model.add(Dropout(0.25))
```

```
model.add(BatchNormalization())
     model.add(Dropout(0.25))
    model.add(Dense(7, activation='softmax'))
    model.compile(
       optimizer = Adam(lr=0.0001),
loss='categorical_crossentropy',
metrics=['accuracy']
[11]: checkpoint = ModelCheckpoint("test1.h5",
                           monitor="val loss".
                           mode="min",
                           save_best_only = True,
                           verbose=1)
    callbacks = [checkpoint]
[12]: epochs = 60
    batch_size = 64
[13]: %%time
    history = model.fit(x = training_set,
                    epochs = epoch
                    callbacks = callbacks,
                    validation data = test set)
    Epoch 00002
449/449 [==
     Epoch 3/60
449/449 [==
           [=============] - ETA: 0s - loss: 6.6892 - accuracy: 0.2783
1003: val_loss improved from 7.17838 to 5.93084, saving model to test1.h5
[=========] - 32s 72ms/step - loss: 6.6892 - accuracy: 0.2783 - val_loss: 5.9308 - val_accuracy: 0.3396
     Epoch 5/60
448/449 [==
     - accuracy: 0.3359 - val_loss: 4.1160 - val_accuracy: 0.3897
     Epoch 6/60
     Epoch 0/00
448/449 [================] - ETA: 0s - loss: 2.8926 - accuracy: 0.4079
Epoch 00008: val_loss improved from 3.11482 to 2.53663, saving model to test1.h5
449/449 [================] - 23s 52ms/step - loss: 2.8924 - accuracy: 0.4077 - val_loss: 2.5366 - val_accuracy: 0.4755
     449/449 [===
Epoch 9/60
448/449 [===
            448/449 [=
     Epoch 0001
449/449 [=
     Epoch 12/60
449/449 [===
    Epoch 00013:
    Epoch 14/60
448/449 [===
     Epoch 16/60
448/449 [========>] - ETA: 0s - loss: 1.6628 - accuracy: 0.5371
Epoch 00016: val_loss did not improve from 1.66457
449/449 [=========] - 23s 51ms/step - loss: 1.6630 - accuracy: 0.5371 - val_loss: 1.6901 - val_accuracy: 0.5124
Epoch 17/60
449/449 [=========] - ETA: 0s - loss: 1.6245 - accuracy: 0.5445
Epoch 00017: val_loss improved from 1.66457 to 1.56306, saving model to test1.h5
449/449 [==========] - 23s 52ms/step - loss: 1.6245 - accuracy: 0.5445 - val_loss: 1.5631 - val_accuracy: 0.5653
     449/449 [===
Epoch 18/60
     Epoch 18/66
448/449 [==
     448/449 [===================]. - ETA: 0s - loss: 1.6004 - accuracy: 0.5502

Epoch 00018: val_loss improved from 1.56306 to 1.51144, saving model to test1.h5

449/449 [===================] - 23s 52ms/step - loss: 1.6005 - accuracy: 0.5502 - val_loss: 1.5114 - val_accuracy: 0.5834
     448/449 [==
Epoch 00020
449/449 [==
            448/449 [=======>] - ETA: 0s - loss: 1.5213 - accuracy: 0.5721

Epoch 00021: val_loss improved from 1.45896 to 1.43982, saving model to test1.h5

449/449 [=========] - 23s 52ms/step - loss: 1.5212 - accuracy: 0.5722 - val_loss: 1.4398 - val_accuracy: 0.6025

Epoch 22/60
     Epoch 21/60
448/449 [===
```

```
448/449 [============]- - EIA: 0S - 10SS: 1.5029 - accuracy: 0.5/6/
Epoch 00022: val_loss did not improve from 1.43982
449/449 [============]- 23s 51ms/step - loss: 1.5030 - accuracy: 0.5766 - val_loss: 1.4997 - val_accuracy: 0.5800
Epoch 23/60
449/449 [=============] - ETA: 0s - loss: 1.4952 - accuracy: 0.5770
                              ======] - ETA: 0s - loss: 1.4952 - accuracy: 0.5770
prove from 1.43982
======] - 23s 51ms/step - loss: 1.4952 - accuracy: 0.5770 - val_loss: 1.4672 - val_accuracy: 0.5935
Epoch 00023
449/449 [==
       023: val_loss did not improve fr
Epoch 24/60
449/449 [===
449/449 [===
Epoch 25/60
449/449 [===
        Epoch 00025
449/449 [==
[==================] - ETA: 0s - loss: 1.4558 - accuracy: 0.5965
1027: val_loss did not improve from 1.41754
[===============] - 23s 52ms/step - loss: 1.4558 - accuracy: 0.5965 - val_loss: 1.4276 - val_accuracy: 0.6085
449/449 [==
Epoch 00027
Epoch 28/60
448/449 [===
449/449 [===
Epoch 30/60
449/449 [===
                                   =] - ETA: 0s - loss: 1.4353 - accuracy: 0.6051
       [======================] - [17. 03 - 1.4959 uccaracy. 0.0092
|930: val loss did not improve from 1.41690
[================] - 25s 56ms/step - loss: 1.4353 - accuracy: 0.6051 - val_loss: 1.4171 - val_accuracy: 0.6112
Epoch 00032
449/449 [==
Epoch 33/60
448/449 [===
448/449 [==============] - ETA: 0s - loss: 1.4184 - accuracy: 0.6102
Epoch 00033: val_loss did not improve from 1.39373
449/449 [===============] - 25s 55ms/step - loss: 1.4182 - accuracy: 0.6103 - val_loss: 1.4345 - val_accuracy: 0.6084
Epoch 00033: Val_loss did not improve from 1.39373
449/449 [==============] - 25s 55ms/step - loss: 1.4182 - accuracy: 0.6103 - Val_loss: 1.4345 - Val_accuracy: 0.6084
Epoch 34/60
448/449 [============] - ETA: 0s - loss: 1.4086 - accuracy: 0.6146
Epoch 00034: Val_loss did not improve from 1.39373
449/449 [=====================] - 25s 56ms/step - loss: 1.4086 - accuracy: 0.6145 - Val_loss: 1.4241 - Val_accuracy: 0.6172
Epoch 35/60
449/449 [===
Epoch 36/60
Epoch 37/6
449/449 [=
                                   =] - ETA: 0s - loss: 1.4036 - accuracy: 0.6213
449/449 [===
Epoch 39/60
449/449 [===
                                 ====] - 28s 63ms/step - loss: 1.3947 - accuracy: 0.6262 - val loss: 1.3934 - val accuracy: 0.6329
                                   =1 - ETA: 0s - loss: 1.4081 - accuracy: 0.6227
Epoch 00039: val_loss did not improve from 1.36851
449/449 [===================] - 34s 75ms/step - loss: 1.4081 - accuracy: 0.6227 - val_loss: 1.4057 - val_accuracy: 0.6204
Epoch 40/60
449/449 [===
        Epoch 00040:
449/449 [===
Epoch 41/60
Epoch 42/60
449/449 [===
Epoch 00043:
449/449 [===
Epoch 44/60
449/449 [===
                                 ====] - 28s 63ms/step - loss: 1.3842 - accuracy: 0.6316 - val loss: 1.3649 - val accuracy: 0.6478
                                   =| - ETA: 0s - loss: 1.3792 - accuracy: 0.6334
Epoch 00044: val_loss did not improve from 1.35783
449/449 [========] - 28s 63ms/step - loss: 1.3792 - accuracy: 0.6334 - val_loss: 1.3590 - val_accuracy: 0.6463
Epoch 45/60
Epoch 46/60
448/449 [===
Epoch 80047: Va_10ss did not improve From 1.35763
449/449 [=============] - 235 52ms/step - loss: 1.3756 - accuracy: 0.6359 - val_loss: 1.3670 - val_accuracy: 0.6452
Epoch 48/60
449/449 [==========] - ETA: 0s - loss: 1.3777 - accuracy: 0.6367
Epoch 80048: val_loss did not improve from 1.35783
449/449 [=====================] - 24s 53ms/step - loss: 1.3777 - accuracy: 0.6367 - val_loss: 1.3721 - val_accuracy: 0.636
                                 ====] - 23s 52ms/step - loss: 1.3756 - accuracy: 0.6359 - val_loss: 1.3670 - val_accuracy: 0.6452
448/449 [==================] - ETA: 0s - loss: 1.3678 - accuracy: 0.6386
Epoch 00050: val_loss did not improve from 1.35783
449/449 [====================] - 24s 52ms/step - loss: 1.3677 - accuracy: 0.6386 - val_loss: 1.3941 - val_accuracy: 0.6289
Epoch 51/60
449/449 [===
EPOCH 31/00
449/449 [========================] - ETA: 0s - loss: 1.3570 - accuracy: 0.6432
Epoch 00051: val_loss did not improve from 1.35783
449/449 [======================] - 26s 59ms/step - loss: 1.3570 - accuracy: 0.6432 - val_loss: 1.3826 - val_accuracy: 0.6485
```

```
25s 56ms/step - loss: 1.3743 - accuracy: 0.6458 - val_loss: 1.3867 - val_accuracy: 0.6432
                 Epoch 00053:
       Epoch 54/60
449/449 [===
       Epoch 00054:
449/449 [===
Epoch 55/60
449/449 [===
Epoch 00055:
449/449 [===
Epoch 56/60==
                                                 =] - 24s 53ms/step - loss: 1.3709 - accuracy: 0.6444 - val_loss: 1.3730 - val_accuracy: 0.6485
                [===================] - ETA: 0s - loss: 1.3648 - accuracy: 0.6473
055: val_loss did not improve from 1.35783
[===============] - 24s 53ms/step - loss: 1.3648 - accuracy: 0.6473 - val_loss: 1.4404 - val_accuracy: 0.6319
       Epoch 57/60
       :====] - 24s 53ms/step - loss: 1.3531 - accuracy: 0.6506 - val loss: 1.4153 - val accuracy: 0.6326
                                                 =] - ETA: 0s - loss: 1.3573 - accuracy: 0.6509 ETA: 1s - los
       Epoch 00058: val_loss did not improve from 1.35783
449/449 [====================] - 27s 60ms/step - loss: 1.3573 - accuracy: (
Epoch 59/60
448/449 [======================] - ETA: 0s - loss: 1.3590 - accuracy: 0.6511
Epoch 00059: val_loss did not improve from 1.35783
                 958: val_loss did not improve from 1.35783
[=================] - 27s 60ms/step - loss: 1.3573 - accuracy: 0.6509 - val_loss: 1.3693 - val_accuracy: 0.6449
       Wall time: 25min 49s
[14]: fig , ax = plt.subplots(1,2)
    train_acc = history.history['accuracy']
    train_loss = history.history['loss']
       fig.set_size_inches(12,4)
       ax[0].plot(history.history['accuracy'])
       ax[0].plot(history.history['val_accuracy'])
       ax[0].set_title('Training Accuracy vs Validation Accuracy')
ax[0].set_ylabel('Accuracy')
ax[0].set_xlabel('Epoch')
       ax[0].legend(['Train', 'Validation'], loc='upper left')
       ax[1].plot(history.history['loss'])
       ax[1].plot(history.history['val_loss'])
       ax[1].set_title('Training Loss vs Validation Loss')
ax[1].set_ylabel('Loss')
ax[1].set_xlabel('Epoch')
       ax[1].legend(['Train', 'Validation'], loc='upper left')
                  Training Accuracy vs Validation Accuracy
                                                                             Training Loss vs Validation Loss
                   Validation
                                                                          Validation
         0.5
                                                                 50 5
          0.4
          0.3
                                                                   3
                                                                   2
          0.2
                                                  50
                                                                                                          50
                     10
                                           40
                                                         60
                                                                             10
                                                                                    20
                                                                                                                 60
                             20
                                  Fnoch
                                                                                         Epoch
[16]: y_pred = model.predict(training_set)
       y_pred = np.argmax(y_pred, axis=1)
       class_labels = test_set.class_indices
class_labels = {v:k for k,v in class_labels.items()}
[16]: {0: 'angry',
1: 'disgust',
2: 'fear',
3: 'happy',
4: 'neutral',
        5: 'sad',
6: 'surprise'}
[17]: from sklearn.metrics import classification_report, confusion_matrix
       cm_train = confusion_matrix(training_set.classes, y_pred)
        print(cm_train)
       print('Classification Report')
target_names = list(class_labels.values())
       print(classification_report(training_set.classes, y_pred, target_names=target_names))
       [[ 550 65 275 1119 681 861 444]
[ 72 8 33 113 72 95 43]
[ 563 54 310 1086 704 898 482]
                99 519 1934 1253 1551
69 345 1365 850 1055
58 375 1330 829 1004
                                          815
575
         1044
          706
671
467
       [ 467 48
Classification
                     254 838
                                530
                                     674
                      precision
                                    recall f1-score
                                                         support
             angry
disgust
fear
                           0.14
                                                             3995
                           0.02
0.15
                                      0.02
0.08
                                                 0.02
0.10
                                                             436
4097
```

