New Section

```
from google.colab import files
uploaded = files.upload()
for fn in uploaded.keys():
  print('User uploaded file "{name}" with length {length} bytes'.format(
      name=fn, length=len(uploaded[fn])))
     Choose Files | archive (1).zip

    archive (1).zip(application/x-zip-compressed) - 63252113 bytes, last modified: 7/30/2023 - 100% done

     Saving archive (1).zip to archive (1) (1).zip
     User uploaded file "archive (1) (1).zip" with length 63252113 bytes
from zipfile import ZipFile
file name = "archive (1).zip"
with ZipFile(file_name, 'r') as zip:
  zip.extractall()
  print("Done")
     Done
from google.colab import drive
drive.mount('/content/drive')
import numpy as np
import cv2
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D
from keras.optimizers import Adam
from keras.layers import MaxPooling2D
from keras.preprocessing.image import ImageDataGenerator
from keras.preprocessing import image
train dir = 'train'
val dir = 'test'
train_datagen = ImageDataGenerator(rescale=1./255)
val_datagen = ImageDataGenerator(rescale=1./255)
train generator = train datagen.flow from directory(
        train dir,
        target_size=(48,48),
        batch_size=64,
        color mode="grayscale",
        class mode='categorical')
validation_generator = val_datagen.flow_from_directory(
        val_dir,
        target size=(48,48),
        batch_size=64,
```

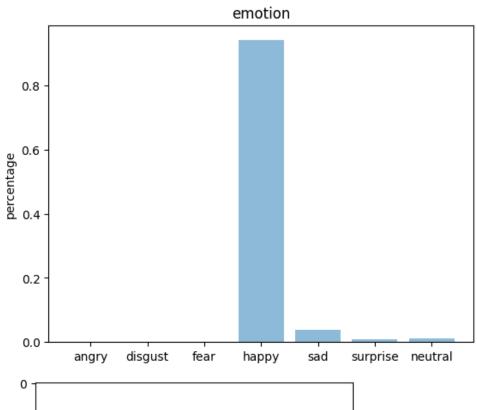
color_mode="grayscale",
class mode='categorical')

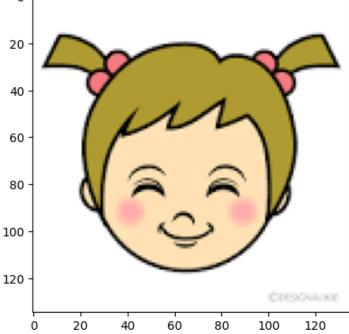
```
Found 28709 images belonging to 7 classes.
    Found 7178 images belonging to 7 classes.
emotion model = Sequential()
emotion model.add(Conv2D(32, kernel size=(3, 3), activation='relu', input shape=(48,48,1)))
emotion model.add(Conv2D(64, kernel_size=(3, 3), activation='relu'))
emotion_model.add(MaxPooling2D(pool_size=(2, 2)))
emotion model.add(Dropout(0.25))
emotion model.add(Conv2D(128, kernel size=(3, 3), activation='relu'))
emotion model.add(MaxPooling2D(pool size=(2, 2)))
emotion_model.add(Conv2D(128, kernel_size=(3, 3), activation='relu'))
emotion model.add(MaxPooling2D(pool size=(2, 2)))
emotion_model.add(Dropout(0.25))
emotion_model.add(Flatten())
emotion model.add(Dense(1024, activation='relu'))
emotion_model.add(Dropout(0.5))
emotion_model.add(Dense(7, activation='softmax'))
emotion_model.compile(loss='categorical_crossentropy',optimizer=Adam(lr=0.0001, decay=1e-6),metrics=['accurac
emotion_model_info = emotion_model.fit_generator(
      train_generator,
      steps_per_epoch=28709 // 64,
      epochs=60,
      validation_data=validation_generator,
      validation_steps=7178 // 64)
    <ipython-input-7-e3e2c1d7010d>:2: UserWarning: `Model.fit generator` is deprecated and will be remov \(^\)
     emotion model info = emotion model.fit generator(
    Epoch 1/60
    448/448 [============ ] - 17s 34ms/step - loss: 1.6249 - accuracy: 0.3670 - val los
    Epoch 2/60
    448/448 [============ ] - 14s 30ms/step - loss: 1.5327 - accuracy: 0.4125 - val los
    Epoch 3/60
    448/448 [============ ] - 14s 32ms/step - loss: 1.4633 - accuracy: 0.4383 - val los
    Epoch 4/60
    Epoch 5/60
    448/448 [============ ] - 14s 32ms/step - loss: 1.3584 - accuracy: 0.4829 - val los
    Epoch 6/60
    448/448 [============= ] - 14s 31ms/step - loss: 1.3113 - accuracy: 0.5049 - val los
    Epoch 7/60
    448/448 [============ ] - 15s 32ms/step - loss: 1.2683 - accuracy: 0.5204 - val los
    Epoch 8/60
    448/448 [============ ] - 14s 32ms/step - loss: 1.2377 - accuracy: 0.5329 - val los
    Epoch 9/60
    448/448 [============ ] - 14s 31ms/step - loss: 1.2003 - accuracy: 0.5484 - val los
    Epoch 10/60
    448/448 [=============== ] - 14s 31ms/step - loss: 1.1723 - accuracy: 0.5611 - val_los
    Epoch 11/60
    448/448 [============ ] - 14s 32ms/step - loss: 1.1466 - accuracy: 0.5712 - val los
    Epoch 12/60
    Epoch 13/60
    448/448 [=============== ] - 14s 31ms/step - loss: 1.0913 - accuracy: 0.5939 - val_los
    Epoch 14/60
    Epoch 15/60
```

```
448/448 [============ ] - 14s 32ms/step - loss: 1.0448 - accuracy: 0.6099 - val los
    Epoch 16/60
    448/448 [=========== ] - 14s 32ms/step - loss: 1.0221 - accuracy: 0.6237 - val los
    Epoch 17/60
    448/448 [============ ] - 14s 32ms/step - loss: 0.9951 - accuracy: 0.6317 - val los
    Epoch 18/60
    Epoch 19/60
    448/448 [============ ] - 14s 30ms/step - loss: 0.9486 - accuracy: 0.6515 - val los
    Epoch 20/60
    Epoch 21/60
    448/448 [============ ] - 14s 31ms/step - loss: 0.9065 - accuracy: 0.6655 - val los
    Epoch 22/60
    448/448 [=============== ] - 14s 30ms/step - loss: 0.8824 - accuracy: 0.6772 - val_los
    Epoch 23/60
    448/448 [============== ] - 13s 30ms/step - loss: 0.8615 - accuracy: 0.6811 - val_los
    Epoch 24/60
    448/448 [============ ] - 14s 32ms/step - loss: 0.8379 - accuracy: 0.6961 - val los
    Epoch 25/60
    448/448 [============== ] - 17s 38ms/step - loss: 0.8156 - accuracy: 0.7010 - val_los
    Epoch 26/60
    448/448 [=========== ] - 14s 32ms/step - loss: 0.7936 - accuracy: 0.7110 - val los
    Epoch 27/60
    448/448 [============== ] - 14s 31ms/step - loss: 0.7760 - accuracy: 0.7191 - val_los
emotion model.save('model.h5')
from keras.models import load model
emotion model = load model('model.h5')
def emotion analysis(emotions):
   objects = ('angry', 'disgust', 'fear', 'happy', 'sad', 'surprise', 'neutral')
   y_pos = np.arange(len(objects))
   plt.bar(y pos, emotions, align='center', alpha=0.5)
   plt.xticks(y_pos, objects)
   plt.ylabel('percentage')
   plt.title('emotion')
   plt.show()
from IPython.display import display, Javascript
from google.colab.output import eval js
from base64 import b64decode
def take photo(filename='photo.jpg', quality=0.8):
 js = Javascript('''
   async function takePhoto(quality) {
     const div = document.createElement('div');
     const capture = document.createElement('button');
     capture.textContent = 'Capture';
    div.appendChild(capture);
     const video = document.createElement('video');
    video.style.display = 'block';
     const stream = await navigator.mediaDevices.getUserMedia({video: true});
    document.body.appendChild(div);
```

```
div.appendChild(video);
     video.srcObject = stream;
      await video.play();
      // Resize the output to fit the video element.
      google.colab.output.setIframeHeight(document.documentElement.scrollHeight, true);
      // Wait for Capture to be clicked.
      await new Promise((resolve) => capture.onclick = resolve);
     const canvas = document.createElement('canvas');
      canvas.width = video.videoWidth;
      canvas.height = video.videoHeight;
     canvas.getContext('2d').drawImage(video, 0, 0);
      stream.getVideoTracks()[0].stop();
     div.remove();
     return canvas.toDataURL('image/jpeg', quality);
    }
    ''')
  display(js)
  data = eval_js('takePhoto({})'.format(quality))
 binary = b64decode(data.split(',')[1])
 with open(filename, 'wb') as f:
    f.write(binary)
  return filename
take photo()
     'photo.jpg'
from tensorflow.keras.utils import load img
from keras.models import load model
import matplotlib.pyplot as plt
import cv2
from tensorflow.keras.utils import load img
from keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.utils import img_to_array
import numpy as np
import matplotlib.pyplot as plt
file = '/content/images (1).png'
true_image =load_img(file)
img =load img(file, color mode="grayscale", target size=(48, 48))
x =img to array(img)
x = np.expand_dims(x, axis = 0)
x /= 255
custom = emotion model.predict(x)
emotion_analysis(custom[0])
x = np.array(x, 'float32')
x = x.reshape([48, 48]);
plt.imshow(true_image)
plt.show()
```

1/1 [=======] - 0s 21ms/step





✓ 1s completed at 6:20 PM

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