# **ORIGINAL CODE**

```
import matplotlib.pyplot as plt
import numpy as np
import cv2
# %matplotlib inline
class_names = ['CUP', 'SPOON', 'FORK', 'MOUSE']
# Creating Realtime Dataset
CAMERA = cv2.VideoCapture(0)
camera_height = 500
raw_frames_type_1 = []
raw_frames_type_2 = []
raw_frames_type_3 = []
raw_frames_type_4 = []
while CAMERA.isOpened():
   ret, frame = CAMERA.read()
   frame = cv2.flip(frame, 1)
   aspect = frame.shape[1]/float(frame.shape[0])
   res = int(aspect * camera_height) # landscape orientation - wide image
    frame = cv2.resize(frame, (res, camera_height))
   # The greean reactangle
   cv2.rectangle(frame, (300, 75), (650, 425), (0, 255, 0), 2)
   cv2.imshow("Capturing", frame)
   key = cv2.waitKey(1)
   if key & 0xff == ord('q'):
       break
    elif key & 0xff == ord('1'):
       raw_frames_type_1.append(frame)
    elif key & 0xff == ord('2'):
       raw_frames_type_2.append(frame)
    elif key & 0xff == ord('3'):
       raw_frames_type_3.append(frame)
    elif key & 0xff == ord('4'):
       raw_frames_type_4.append(frame)
   # Preview
   plt.imshow(frame)
```

```
plt.show()
 Camera
CAMERA.release()
cv2.destroyAllWindows()
save_width = 339
save_height = 400
import os
from glob import glob
reval = os.getcwd()
print ("Current working directory %s" % reval)
print('img1: ', len(raw_frames_type_1))
print('img2: ', len(raw_frames_type_2))
print('img3: ', len(raw_frames_type_3))
print('img4: ', len(raw_frames_type_4))
# Crop the images
for i, frame in enumerate(raw_frames_type_1):
   roi - frame[75+2:425-2, 300+2:650-2]
   roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
   roi = cv2.resize(roi, (save_width, save_height))
   cv2.imwrite('img_1/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
for i, frame in enumerate(raw_frames_type_2):
       roi - frame[75+2:425-2, 300+2:650-2]
       roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
       roi = cv2.resize(roi, (save_width, save_height))
       cv2.imwrite('img_2/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
for i, frame in enumerate(raw_frames_type_3):
       # Get roi
       roi - frame[75+2:425-2, 300+2:650-2]
       roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
        roi = cv2.resize(roi, (save_width, save_height))
```

```
cv2.imwrite('img_3/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
for i, frame in enumerate(raw_frames_type_4):
       roi - frame[75+2:425-2, 300+2:650-2]
       roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
        roi = cv2.resize(roi, (save_width, save_height))
        cv2.imwrite('img_4/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
from glob import glob
from keras import preprocessing
width = 96
height = 96
images_type_1 = []
images_type_2 = []
images_type_3 = []
images_type_4 = []
for image_path in glob('img_1/*.*'):
    image = preprocessing.image.load_img(image_path, target_size=(width, height))
   x = preprocessing.image.img_to_array(image)
   images_type_1.append(x)
for image_path in glob('img_2/*.*'):
    image = preprocessing.image.load_img(image_path, target_size=(width, height))
   x = preprocessing.image.img_to_array(image)
   images_type_2.append(x)
for image_path in glob('img_3/*.*'):
   image = preprocessing.image.load_img(image_path, target_size=(width, height))
    x = preprocessing.image.img_to_array(image)
   images_type_3.append(x)
for image_path in glob('img_4/*.*'):
   image = preprocessing.image.load_img(image_path, target_size=(width, height))
    x = preprocessing.image.img_to_array(image)
    images_type_4.append(x)
plt.figure(figsize=(12, 8))
for i, x in enumerate(images_type_1[:5]):
    plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
```

```
plt.title('{} image'.format(class_names[0]))
plt.show()
for i, x in enumerate(images_type_2[:5]):
   plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[1]))
plt.show()
for i, x in enumerate(images_type_3[:5]):
   plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[2]))
plt.show()
for i, x in enumerate(images_type_4[:5]):
   plt.subplot(1, 5, i+1)
    image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[3]))
plt.show()
X_type_1 = np.array(images_type_1)
X_type_2 = np.array(images_type_2)
X_type_3 = np.array(images_type_3)
X_type_4 = np.array(images_type_4)
# Check the shape using .shape() check the images count
print (X_type_1.shape)
print (X_type_2.shape)
print (X_type_3.shape)
print (X_type_4.shape)
(13, 96, 96, 3)
(23, 96, 96, 3)
(14, 96, 96, 3)
(22, 96, 96, 3)
X_type_2
X = np.concatenate((X_type_1, X_type_2), axis=0)
if len(X_type_3):
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X = np.concatenate((X, X_type_3), axis=0)
if len(X_type_4):
   X = np.concatenate((X, X_type_4), axis=0)
# Scaling the data to 1 - 0
X = X / 255.0
X.shape
(72, 96, 96, 3)
from keras.utils import to_categorical
y_type_1 = [0 for item in enumerate(X_type_1)]
y_type_2 = [1 for item in enumerate(X_type_2)]
y_type_3 = [2 for item in enumerate(X_type_3)]
y_type_4 = [3 for item in enumerate(X_type_4)]
y = np.concatenate((y_type_1, y_type_2), axis=0)
if len(y_type_3):
   y = np.concatenate((y, y_type_3), axis=0)
if len(y_type_4):
   y = np.concatenate((y, y_type_4), axis=0)
y = to_categorical(y, num_classes=len(class_names))
y.shape
(72, 4)
from keras.models import Sequential
from keras.layers.core import Activation, Dropout, Flatten, Dense
from keras.layers.convolutional import Convolution2D, MaxPooling2D
from keras.optimizers import Adam
# Default Parameters
conv_1 = 16
conv_1_drop = 0.2
conv_2 = 32
conv_2_drop = 0.2
dense_1_n = 1024
dense_1 drop = 0.2
dense_2_n = 512
dense_2_drop = 0.2
# Values you can adjust
lr = 0.001
epochs = 5
batch_size = 10
color_channels = 3
def build_model(conv_1_drop = conv_1_drop, conv_2_drop = conv_2_drop,
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dense_1_n = dense_1_n, dense_1_drop = dense_1_drop,
                dense_2_n = dense_2_n, dense_2_drop = dense_2_drop,
                lr = lr):
   model = Sequential()
   model.add(Convolution2D(conv_1, (3, 3),
                            input_shape = (width, height, color_channels),
                            activation='relu'))
   model.add(MaxPooling2D(pool_size=(2, 2)))
   model.add(Dropout(conv_1_drop))
   model.add(Convolution2D(conv_2, (3, 3), activation='relu'))
   model.add(MaxPooling2D(pool size=(2, 2)))
   model.add(Dropout(conv_1_drop))
   model.add(Flatten())
   model.add(Dense(dense_1_n, activation='relu'))
   model.add(Dropout(dense_1_drop))
   model.add(Dense(dense_2_n, activation='relu'))
   model.add(Dropout(dense_2_drop))
   model.add(Dense(len(class_names), activation='softmax'))
   model.compile(loss='categorical_crossentropy',
                  optimizer=Adam(clipvalue=0.5),
                 metrics=['accuracy'])
   return model
 model parameter
model = build_model()
model.summary()
# Do not run yet
history = model.fit(X, y, validation_split=0.10, epochs=10, batch_size=5)
print(history)
# Model evaluation
scores = model.evaluate(X, y, verbose=0)
print("Accuracy: %.2f%%" % (scores[1]*100))
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model loss')
plt.ylabel('loss and accuracy')
plt.xlabel('epoch')
```

```
plt.legend(['train', 'test'], loc='upper right')
plt.show()
plt.plot(history.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.show()
plt.plot(history.history['accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.show()
# Prediction
import seaborn as sns
from sklearn.metrics import confusion_matrix
def plt_show(img):
   plt.imshow(img)
   plt.show()
cup = 'img_1/10.png'
spoon = 'img_2/10.png'
fork = 'img_3/10.png'
mouse = 'img_4/10.png'
imgs = [cup, spoon, fork, mouse]
# def predict_(img_path):
classes = None
predicted_classes = []
for i in range(len(imgs)):
   type_ = preprocessing.image.load_img(imgs[i], target_size=(width, height))
   plt.imshow(type_)
   plt.show()
   type_x = np.expand_dims(type_, axis=0)
   prediction = model.predict(type_x)
   index = np.argmax(prediction)
   print(class_names[index])
   classes = class_names[index]
   predicted_classes.append(class_names[index])
cm = confusion_matrix(class_names, predicted_classes)
f = sns.heatmap(cm, xticklabels=class_names, yticklabels=predicted_classes, annot=True)
type_1 = preprocessing.image.load_img('img_1/10.png', target_size=(width, height))
plt.imshow(type_1)
plt.show()
type_1_x = np.expand_dims(type_1, axis=0)
predictions = model.predict(type_1_x)
index = np.argmax(predictions)
```

```
print(class_names[index])
type_2 = preprocessing.image.load_img('img_2/10.png', target_size=(width, height))
plt.imshow(type_2)
plt.show()
type_2_x = np.expand_dims(type_2, axis=0)
predictions = model.predict(type_2_x)
index = np.argmax(predictions)
print(class_names[index])
# Live Predictions using camera
from keras.applications import inception_v3
import time
CAMERA = cv2.VideoCapture(0)
camera_height = 500
while(True):
   _, frame = CAMERA.read()
   frame = cv2.flip(frame, 1)
   aspect = frame.shape[1] / float(frame.shape[0])
   res = int(aspect* camera_height)
   frame = cv2.resize(frame, (res, camera_height))
   roi = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
   # Adjust alignment
   roi = cv2.resize(roi, (width, height))
   roi_x = np.expand_dims(roi, axis=0)
   predictions = model.predict(roi_x)
   type_1_x, type_2_x, type_3_x, type_4_x = predictions[0]
   # The green rectangle
   cv2.rectangle(frame, (300, 75), (650, 425), (0, 255, 0), 2)
   type_1_txt = '{}: {}%'.format(class_names[0], int(type_1_x*100))
    cv2.putText(frame, type_1_txt, (70, 210), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
    type_2_txt = '{}: {}%'.format(class_names[1], int(type_2_x*100))
    cv2.putText(frame, type_2_txt, (70, 235), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
    type_3_txt = '{}: {}%'.format(class_names[2], int(type_3_x*100))
    cv2.putText(frame, type_3_txt, (70, 255), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
   type_4_txt = '{}: {}%'.format(class_names[3], int(type_4_x*100))
```

```
cv2.putText(frame, type_4_txt, (70, 275), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
   cv2.imshow('Real time object detection', frame)
   key = cv2.waitKey(1)
   if key & 0xff == ord('q'):
       break
   plt.imshow(frame)
   plt.show()
CAMERA.release()
cv2.destroyAllWindows()
import matplotlib.pyplot as plt
import numpy as np
import cv2
# %matplotlib inline
class_names = ['CUP', 'SPOON', 'FORK', 'MOUSE']
# Creating Realtime Dataset
CAMERA = cv2.VideoCapture(0)
camera_height = 500
raw_frames_type_1 = []
raw_frames_type_2 = []
raw_frames_type_3 = []
raw_frames_type_4 = []
while CAMERA.isOpened():
   ret, frame = CAMERA.read()
   frame = cv2.flip(frame, 1)
   aspect = frame.shape[1]/float(frame.shape[0])
   res = int(aspect * camera_height) # landscape orientation - wide image
   frame = cv2.resize(frame, (res, camera_height))
   # The greean reactangle
   cv2.rectangle(frame, (300, 75), (650, 425), (0, 255, 0), 2)
   cv2.imshow("Capturing", frame)
   key = cv2.waitKey(1)
   if key & 0xff == ord('q'):
```

```
break
    elif key & 0xff == ord('1'):
        raw_frames_type_1.append(frame)
    elif key & 0xff == ord('2'):
        raw_frames_type_2.append(frame)
    elif key & 0xff == ord('3'):
        raw_frames_type_3.append(frame)
    elif key & 0xff == ord('4'):
        raw_frames_type_4.append(frame)
    plt.imshow(frame)
   plt.show()
# Camera
CAMERA.release()
cv2.destroyAllWindows()
save_width = 339
save_height = 400
import os
from glob import glob
reval = os.getcwd()
print ("Current working directory %s" % reval)
print('img1: ', len(raw_frames_type_1))
print('img2: ', len(raw_frames_type_2))
print('img3: ', len(raw_frames_type_3))
print('img4: ', len(raw_frames_type_4))
for i, frame in enumerate(raw_frames_type_1):
   # Get roi
   roi - frame[75+2:425-2, 300+2:650-2]
   # Parse BRG to RGB
   roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
   roi = cv2.resize(roi, (save_width, save_height))
   cv2.imwrite('img_1/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
for i, frame in enumerate(raw_frames_type_2):
        roi - frame[75+2:425-2, 300+2:650-2]
        roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
       # resize to 224 x 224
```

```
roi = cv2.resize(roi, (save_width, save_height))
        cv2.imwrite('img_2/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
for i, frame in enumerate(raw_frames_type_3):
       roi - frame[75+2:425-2, 300+2:650-2]
       # Parse BRG to RGB
       roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
       # resize to 224 x 224
       roi = cv2.resize(roi, (save_width, save_height))
        cv2.imwrite('img_3/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
for i, frame in enumerate(raw_frames_type_4):
       # Get roi
       roi - frame[75+2:425-2, 300+2:650-2]
       roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
       roi = cv2.resize(roi, (save_width, save_height))
        cv2.imwrite('img_4/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_BGR2RGB))
from glob import glob
from keras import preprocessing
width = 96
height = 96
images_type_1 = []
images_type_2 = []
images_type_3 = []
images_type_4 = []
for image_path in glob('img_1/*.*'):
   image = preprocessing.image.load_img(image_path, target_size=(width, height))
   x = preprocessing.image.img_to_array(image)
   images_type_1.append(x)
for image_path in glob('img_2/*.*'):
   image = preprocessing.image.load_img(image_path, target_size=(width, height))
   x = preprocessing.image.img_to_array(image)
   images_type_2.append(x)
for image_path in glob('img_3/*.*'):
   image = preprocessing.image.load_img(image_path, target_size=(width, height))
    x = preprocessing.image.img_to_array(image)
   images_type_3.append(x)
```

```
for image_path in glob('img_4/*.*'):
   image = preprocessing.image.load_img(image_path, target_size=(width, height))
    x = preprocessing.image.img_to_array(image)
   images_type_4.append(x)
plt.figure(figsize=(12, 8))
for i, x in enumerate(images_type_1[:5]):
   plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[0]))
plt.show()
for i, x in enumerate(images_type_2[:5]):
   plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[1]))
plt.show()
for i, x in enumerate(images_type_3[:5]):
   plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[2]))
plt.show()
for i, x in enumerate(images_type_4[:5]):
   plt.subplot(1, 5, i+1)
   image = preprocessing.image.array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[3]))
plt.show()
X_type_1 = np.array(images_type_1)
X_type_2 = np.array(images_type_2)
X_type_3 = np.array(images_type_3)
X_type_4 = np.array(images_type_4)
```

```
print (X_type_1.shape)
print (X_type_2.shape)
print (X_type_3.shape)
print (X_type_4.shape)
(13, 96, 96, 3)
(23, 96, 96, 3)
(14, 96, 96, 3)
(22, 96, 96, 3)
X_type_2
X = np.concatenate((X_type_1, X_type_2), axis=0)
if len(X_type_3):
   X = np.concatenate((X, X_type_3), axis=0)
if len(X_type_4):
   X = np.concatenate((X, X_type_4), axis=0)
# Scaling the data to 1 - 0
X = X / 255.0
X.shape
(72, 96, 96, 3)
from keras.utils import to_categorical
y_type_1 = [0 for item in enumerate(X_type_1)]
y_type_2 = [1 for item in enumerate(X_type_2)]
y_type_3 = [2 for item in enumerate(X_type_3)]
y_type_4 = [3 for item in enumerate(X_type_4)]
y = np.concatenate((y_type_1, y_type_2), axis=0)
if len(y_type_3):
   y = np.concatenate((y, y_type_3), axis=0)
if len(y_type_4):
   y = np.concatenate((y, y_type_4), axis=0)
y = to_categorical(y, num_classes=len(class_names))
y.shape
(72, 4)
# CNN Config
from keras.models import Sequential
from keras.layers.core import Activation, Dropout, Flatten, Dense
from keras.layers.convolutional import Convolution2D, MaxPooling2D
from keras.optimizers import Adam
# Default Parameters
```

```
conv_1 = 16
conv_1_drop = 0.2
conv_2 = 32
conv_2_drop = 0.2
dense_1_n = 1024
dense_1_drop = 0.2
dense_2_n = 512
dense_2_drop = 0.2
lr = 0.001
epochs = 5
batch_size = 10
color_channels = 3
def build_model(conv_1_drop = conv_1_drop, conv_2_drop = conv_2_drop,
                dense 1 n = dense 1 n, dense 1 drop = dense 1 drop,
                dense_2_n = dense_2_n, dense_2_drop = dense_2_drop,
                lr = lr):
   model = Sequential()
   model.add(Convolution2D(conv_1, (3, 3),
                            input_shape = (width, height, color_channels),
                            activation='relu'))
   model.add(MaxPooling2D(pool_size=(2, 2)))
   model.add(Dropout(conv_1_drop))
   model.add(Convolution2D(conv_2, (3, 3), activation='relu'))
   model.add(MaxPooling2D(pool_size=(2, 2)))
   model.add(Dropout(conv_1_drop))
   model.add(Flatten())
   model.add(Dense(dense_1_n, activation='relu'))
   model.add(Dropout(dense_1_drop))
   model.add(Dense(dense_2_n, activation='relu'))
   model.add(Dropout(dense_2_drop))
   model.add(Dense(len(class_names), activation='softmax'))
   model.compile(loss='categorical_crossentropy',
                 optimizer=Adam(clipvalue=0.5),
                  metrics=['accuracy'])
   return model
model = build_model()
nodel.summary()
```

```
history = model.fit(X, y, validation_split=0.10, epochs=10, batch_size=5)
print(history)
scores = model.evaluate(X, y, verbose=0)
print("Accuracy: %.2f%%" % (scores[1]*100))
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model loss')
plt.ylabel('loss and accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.show()
plt.plot(history.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.show()
plt.plot(history.history['accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper right')
plt.show()
import seaborn as sns
from sklearn.metrics import confusion_matrix
def plt_show(img):
   plt.imshow(img)
   plt.show()
cup = 'img_1/10.png'
spoon = 'img_2/10.png'
fork = 'img_3/10.png'
mouse = 'img_4/10.png'
imgs = [cup, spoon, fork, mouse]
# def predict_(img_path):
classes = None
predicted_classes = []
for i in range(len(imgs)):
   type_ = preprocessing.image.load_img(imgs[i], target_size=(width, height))
   plt.imshow(type_)
   plt.show()
    type_x = np.expand_dims(type_, axis=0)
   prediction = model.predict(type_x)
```

```
index = np.argmax(prediction)
   print(class_names[index])
   classes = class_names[index]
   predicted_classes.append(class_names[index])
cm = confusion_matrix(class_names, predicted_classes)
 = sns.heatmap(cm, xticklabels=class_names, yticklabels=predicted_classes, annot=True)
type_1 = preprocessing.image.load_img('img_1/10.png', target_size=(width, height))
plt.imshow(type_1)
plt.show()
type_1_x = np.expand_dims(type_1, axis=0)
predictions = model.predict(type_1_x)
index = np.argmax(predictions)
print(class_names[index])
type_2 = preprocessing.image.load_img('img_2/10.png', target_size=(width, height))
plt.imshow(type_2)
plt.show()
type_2_x = np.expand_dims(type_2, axis=0)
predictions = model.predict(type_2_x)
index = np.argmax(predictions)
print(class_names[index])
from keras.applications import inception_v3
import time
CAMERA = cv2.VideoCapture(0)
camera_height = 500
while(True):
   _, frame = CAMERA.read()
   # Flip
   frame = cv2.flip(frame, 1)
   aspect = frame.shape[1] / float(frame.shape[0])
   res = int(aspect* camera_height)
    frame = cv2.resize(frame, (res, camera_height))
   roi = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
   roi = cv2.resize(roi, (width, height))
   roi_x = np.expand_dims(roi, axis=0)
   predictions = model.predict(roi_x)
    type_1_x, type_2_x, type_3_x, type_4_x = predictions[0]
    # The green rectangle
   cv2.rectangle(frame, (300, 75), (650, 425), (0, 255, 0), 2)
```

```
type\_1\_txt = '\{\}: \{\}\%'.format(class\_names[0], int(type\_1\_x*100))
   cv2.putText(frame, type_1_txt, (70, 210), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
2)
    type_2_txt = '{}: {}%'.format(class_names[1], int(type_2_x*100))
   cv2.putText(frame, type_2_txt, (70, 235), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
2)
    type_3_txt = '\{\}: \{\}\%'.format(class_names[2], int(type_3_x*100))
    cv2.putText(frame, type_3_txt, (70, 255), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
2)
    type_4_txt = '{}: {}%'.format(class_names[3], int(type_4_x*100))
    cv2.putText(frame, type_4_txt, (70, 275), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240, 240),
2)
   cv2.imshow('Real time object detection', frame)
   key = cv2.waitKey(1)
   if key & 0xff == ord('q'):
       break
   plt.imshow(frame)
   plt.show()
CAMERA.release()
cv2.destroyAllWindows()
```

# **REVISED CODE**

```
import matplotlib.pyplot as plt
import numpy as np
import cv2
from glob import glob
from keras import preprocessing
from keras.utils import load_img, img_to_array, array_to_img, to_categorical
import os
from keras.models import Sequential, load_model
from keras.layers.core import Activation, Dropout, Flatten, Dense
from keras.layers.convolutional import Convolution2D, MaxPooling2D
from keras.optimizers import Adam
from keras.applications import inception_v3
import time
from PIL import Image
import sys
# %matplotlib inline
# class_names = ['KNIFE','WATER_BOTTLE','PHONE','GLASS']
# class_names = ['FORK','GLASSES','PLATE','SPOON']
class_names = ['KANAN', 'MARI', 'CHIKA', 'RUBY']
width = 96
height = 96
def live_capture():
   model_name_live= (input("What is the name of your model? (h5 format, extension will be
automatically added): "))
   model = load_model(model_name_live + ".h5")
   CAMERA = cv2.VideoCapture(0)
   camera_height = 500
         # Get ROI
          # Parse BRG to RGB
         roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
          # Adjust alignment
          roi x = np.expand dims(roi, axis=0)
```

```
predictions = model.predict(roi_x)
# Green rectangle
# Calculate the center of the bounding box
# Calculate the offset for centering the bounding box
offset x = int((save width - box width) / 2)
# Adjust the coordinates based on the offset
rectangle_y1 += offset_y
rectangle_x2 += offset_x
rectangle_y2 += offset_y
# Draw the centered bounding box
cv2.rectangle(frame, (rectangle_x1, rectangle_y1), (rectangle_x2, rectangle_y2), (0,
# cv2.rectangle(frame, (150, 50), (650, 425), (0, 255, 0), 2)
```

```
_, frame = CAMERA.read(0)
       frame = cv2.flip(frame, 1)
       aspect = frame.shape[1] / float(frame.shape[0])
        res = int(aspect * camera_height) # Landscape orientation - wide image
        frame = cv2.resize(frame, (res, camera_height))
       # Calculate the center of the bounding box
       window_center_x = frame.shape[1] // 2
       window_center_y = frame.shape[0] // 2
       # Calculate the new width and height for the adjusted bounding box
       box_width = 400
       box_height = int(box_width / aspect)
       # Calculate the offset for centering the frame
       new_rectangle_x1 = window_center_x - (box_width // 2)
       new_rectangle_y1 = window_center_y - (box_height // 2)
        new_rectangle_x2 = window_center_x + (box_width // 2)
       new_rectangle_y2 = window_center_y + (box_height // 2)
       roi = frame[new_rectangle_y1:new_rectangle_y2, new_rectangle_x1:new_rectangle_x2]
        # Parse BRG to RGB
       roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
       # Adjust alignment
       roi = cv2.resize(roi, (width, height))
       roi_x = np.expand_dims(roi, axis=0)
       predictions = model.predict(roi_x)
       type\_1\_x, \ type\_2\_x, \ type\_3\_x, \ type\_4\_x = predictions[0]
        # Draw the adjusted bounding box
        cv2.rectangle(frame, (new_rectangle_x1, new_rectangle_y1), (new_rectangle_x2,
new_rectangle_y2), (0, 255, 0), 2)
        type_1_text = '{} - {}%'.format(class_names[0], int(type_1_x * 100))
        cv2.putText(frame, type_1_text, (70, 210), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240,
240), 2)
        type_2_{text} = '{} - {}%'.format(class_names[1], int(type_2_x * 100))
        cv2.putText(frame, type_2_text, (70, 235), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240,
240), 2)
        type_3_text = '{} - {}%'.format(class_names[2], int(type_3_x * 100))
        cv2.putText(frame, type_3_text, (70, 255), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240,
240), 2)
       type_4_text = '{} - {}%'.format(class_names[3], int(type_4_x * 100))
```

```
cv2.putText(frame, type_4_text, (70, 275), cv2.FONT_HERSHEY_SIMPLEX, 0.6, (240, 240,
240), 2)
             print("ROI does not match adjusted bounding box")
       cv2.imshow('Real-time object detection', frame)
       # Controls q = quit
       key = cv2.waitKey(1)
       if key & 0xFF == ord('q'):
           Break
   # Release the camera
   CAMERA.release()
   cv2.destroyAllWindows()
   sys.exit()
#creating realtime dataset
training_answer = input("Do you want to create a new dataset or retrain the current dataset?
if training_answer == 'n' or training_answer == 'N' or training_answer == 'No' or
training_answer == 'no':
    live_capture()
CAMERA = cv2.VideoCapture(0)
camera_height = 500
raw_frames_type_1 = []
raw_frames_type_2 = []
raw_frames_type_3 = []
raw_frames_type_4 = []
     # Calculate the center of the bounding box
      # Calculate the new coordinates for the centered bounding box
     box width = 650 - 150
```

```
rectangle_x1 = center_x - int(box_width / 2)
     # Adjust the coordinates based on the offset
     # Draw the centered bounding box
     cv2.rectangle(frame, (rectangle_x1, rectangle_y1), (rectangle_x2, rectangle_y2), (0, 255,
0), 2)
      cv2.rectangle(frame, (rectangle_x1, rectangle_y1), (rectangle_x2, rectangle_y2), (0, 255,
0), 2)
while CAMERA.isOpened():
   ret, frame = CAMERA.read()
   frame = cv2.flip(frame, 1)
   aspect = frame.shape[1] / float(frame.shape[0])
   res = int(aspect * camera_height)
   frame = cv2.resize(frame, (res, camera_height))
   window center x = frame.shape[1] // 2
```

```
window_center_y = frame.shape[0] // 2
   # Calculate the new width and height for the bounding box
   box width = 400
   box_height = int(box_width / aspect)
   offset_x = (frame.shape[1] - box_width) // 2
   offset_y = (frame.shape[0] - box_height) // 2
   # Calculate the new coordinates for the centered bounding box
   rectangle_x1 = window_center_x - (box_width // 2)
   rectangle_y1 = window_center_y - (box_height // 2)
   rectangle_x2 = window_center_x + (box_width // 2)
   rectangle_y2 = window_center_y + (box_height // 2)
   # Draw the centered bounding box
   cv2.rectangle(frame, (rectangle_x1, rectangle_y1), (rectangle_x2, rectangle_y2), (0, 255,
0), 2)
   # Show the frame
   cv2.imshow('Capturing', frame)
   key = cv2.waitKey(1) & 0xFF
   if key == ord('q'):
       break
   elif key == ord('1'):
       raw_frames_type_1.append(frame)
        print("Captured type 1 frame.")
    elif key == ord('2'):
        raw_frames_type_2.append(frame)
        print("Captured type 2 frame.")
    elif key == ord('3'):
       raw_frames_type_3.append(frame)
        print("Captured type 3 frame.")
    elif key == ord('4'):
       raw_frames_type_4.append(frame)
       print("Captured type 4 frame.")
CAMERA.release()
cv2.destroyAllWindows()
save_width = 339
save_height = 400
retval = os.getcwd()
print ("Current working directory %s" % retval)
print ('img1: ', len(raw_frames_type_1))
print ('img2: ', len(raw_frames_type_2))
print ('img3: ', len(raw_frames_type_3))
print ('img4: ', len(raw_frames_type_4))
#crop the images
for i, frame in enumerate(raw_frames_type_1):
```

```
roi = frame[rectangle_y1:rectangle_y2, rectangle_x1:rectangle_x2]
   roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
   roi = cv2.resize(roi, (save_width, save_height))
   cv2.imwrite('img_1/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_RGB2BGR))
   plt.imshow(roi)
   plt.axis('off')
   plt.show()
for i, frame in enumerate(raw_frames_type_2):
   roi = frame[rectangle_y1:rectangle_y2, rectangle_x1:rectangle_x2]
   roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
   roi = cv2.resize(roi, (save_width, save_height))
   cv2.imwrite('img_2/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_RGB2BGR))
   plt.imshow(roi)
   plt.axis('off')
   plt.show()
for i, frame in enumerate(raw_frames_type_3):
   roi = frame[rectangle_y1:rectangle_y2, rectangle_x1:rectangle_x2]
   roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
   roi = cv2.resize(roi, (save_width, save_height))
   cv2.imwrite('img_3/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_RGB2BGR))
   plt.imshow(roi)
   plt.axis('off')
   plt.show()
for i, frame in enumerate(raw_frames_type_4):
    roi = frame[rectangle_y1:rectangle_y2, rectangle_x1:rectangle_x2]
```

```
#parse brg to rgb
   roi = cv2.cvtColor(roi, cv2.COLOR_BGR2RGB)
   #resize to 224 x 224
   roi = cv2.resize(roi, (save_width, save_height))
   cv2.imwrite('img_4/{}.png'.format(i), cv2.cvtColor(roi, cv2.COLOR_RGB2BGR))
   plt.imshow(roi)
   plt.axis('off')
   plt.show()
images_type_1 = []
images_type_2 = []
images_type_3 = []
images_type_4 = []
for image_path in glob('img_1/*.*'):
   image = load_img(image_path, target_size=(width, height))
   x = img_to_array(image)
   images_type_1.append(x)
for image_path in glob('img_2/*.*'):
    image = load_img(image_path, target_size=(width, height))
    x = img_to_array(image)
   images_type_2.append(x)
for image_path in glob('img_3/*.*'):
    image = load_img(image_path, target_size=(width, height))
   x = img_to_array(image)
   images_type_3.append(x)
for image_path in glob('img_4/*.*'):
   image = load_img(image_path, target_size=(width, height))
   x = img_to_array(image)
   images_type_4.append(x)
print('Shape of images_type_1:', images_type_1[0].shape)
print('Shape of images_type_2:', images_type_2[0].shape)
print('Shape of images_type_3:', images_type_3[0].shape)
print('Shape of images_type_4:', images_type_4[0].shape)
plt.figure(figsize=(12,8))
samples = 5
for i, x in enumerate(images_type_1[:samples]):
   plt.subplot(1,samples,i+1)
   image = array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
    plt.title('{} image'.format(class_names[0]))
```

```
plt.show()
plt.figure(figsize=(12,8))
for i, x in enumerate(images_type_2[:samples]):
   plt.subplot(1,samples,i+1)
   image = array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[1]))
plt.show()
plt.figure(figsize=(12,8))
for i, x in enumerate(images_type_3[:samples]):
   plt.subplot(1,samples,i+1)
   image = array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[2]))
plt.show()
plt.figure(figsize=(12,8))
for i, x in enumerate(images_type_4[:samples]):
    plt.subplot(1,samples,i+1)
    image = array_to_img(x)
   plt.imshow(image)
   plt.axis('off')
   plt.title('{} image'.format(class_names[3]))
plt.show()
X_{type_1} = np.array(images_type_1)
X_type_2 = np.array(images_type_2)
X_type_3 = np.array(images_type_3)
X_type_4 = np.array(images_type_4)
X_type_1 = X_type_1.reshape(-1, width, height, 3)
X_type_2 = X_type_2.reshape(-1, width, height, 3)
X_type_3 = X_type_3.reshape(-1, width, height, 3)
X_type_4 = X_type_4.reshape(-1, width, height, 3)
X = np.concatenate((X_type_1, X_type_2), axis=0)
if len(X_type_3):
   X = np.concatenate((X, X_type_3), axis=0)
if len(X_type_4):
   X = np.concatenate((X, X_type_4), axis=0)
#Scaling the data to 1 - 0
```

```
X = X/255.0
X = X.reshape(-1, width, height, 3)
y_type_1 = [0 for item in enumerate(X_type_1)]
y_type_2 = [1 for item in enumerate(X_type_2)]
y_type_3 = [2 for item in enumerate(X_type_3)]
y_type_4 = [3 for item in enumerate(X_type_4)]
y = np.concatenate((y_type_1, y_type_2), axis=0)
if len(y_type_3):
   y = np.concatenate((y, y_type_3), axis=0)
if len(y_type_4):
   y = np.concatenate((y, y_type_4), axis=0)
y = to_categorical(y, num_classes=len(class_names))
y.shape
(72, 4)
#Default Parameters
while True:
   conv_1 =16
   conv_1_drop = 0.2
   conv_2 = 32
   conv_2_drop = 0.2
   dense_1_n = 1024
   dense_1_n_drop = 0.2
   dense_2_n = 512
   dense_2_n_drop = 0.2
   #values you can adjust
   lr = 0.001
   epochs = 20
   batch_size = 5
   color_channels = 3
   def build_model( conv_1_drop = conv_1_drop, conv_2_drop = conv_2_drop,
                   dense_1_n = dense_1_n, dense_1_n_drop = dense_1_n_drop,
                   dense_2_n = dense_2_n, dense_2_n_drop = dense_2_n_drop,
                   lr=lr):
       model = Sequential()
       model.add(Convolution2D(conv_1, (3,3),
                                input_shape = (width, height, color_channels),
                                activation='relu'))
       model.add(MaxPooling2D(pool_size=(2,2)))
       model.add(Dropout(conv_1_drop))
       model.add(Convolution2D(conv_2, (3,3), activation='relu'))
```

```
model.add(MaxPooling2D(pool_size=(2,2)))
    model.add(Dropout(conv_1_drop))
    model.add(Flatten())
    model.add(Dense(dense_1_n, activation='relu'))
    model.add(Dropout(dense_1_n_drop))
    model.add(Dense(dense_2_n, activation='relu'))
    model.add(Dropout(dense 2 n drop))
    model.add(Dense(len(class_names), activation='softmax'))
    model.compile(loss='categorical_crossentropy',
                optimizer=Adam(clipvalue=0.5),
                metrics=['accuracy'])
    return model
model = build_model()
model.summary()
history = model.fit(X, y, validation_split=0.10, epochs=epochs, batch_size=batch_size)
print (history)
# Model evaluation
scores = model.evaluate(X, y, verbose=1)
print ("Accuracy: %.2f%%" %(scores[1]*100))
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('loss and accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
plt.plot(history.history['loss'])
plt.title('Model Loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
plt.plot(history.history['accuracy'])
plt.title('Model Accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
```

```
plt.show()
   #prediction
   import seaborn as sns
   from sklearn.metrics import confusion_matrix
   def plt_show(img):
       plt.imshow(img)
       plt.show()
   # fork = "img_1/10.png"
   # glasses = "img_2/16.png"
   # plate = "img_3/09.png"
   #learning data
   kanan = "img 1/1.JPG"
   mari = "img 2/1.JPG"
   chika = "img_3/1.JPG"
   ruby = "img_4/1.JPG"
   imgs = [kanan, mari, chika, ruby]
   # def predict_(img_path):
   classes = None
   predicted_classes = []
   true_labels = []
   for i in range(len(imgs)):
       type_ = load_img(imgs[i], target_size=(width, height))
       plt.imshow(type_)
       plt.show()
       type_x = np.expand_dims(type_, axis=0)
       prediction = model.predict(type_x)
       index = np.argmax(prediction)
       print(class_names[index])
       classes = class_names[index]
       predicted_classes.append(class_names[index])
       true_labels.append(class_names[i % len(class_names)]) # Append the true class to the
true_labels list
   cm = confusion_matrix(true_labels, predicted_classes)
   f = sns.heatmap(cm, xticklabels=class_names, yticklabels=predicted_classes, annot=True)
   # type_1 = load_img('img_1/10.png', target_size=(width, height))
   # plt.imshow(type_1)
   # index = np.argmax(predictions)
```

```
# type_2 = load_img('img_2/16.png', target_size=(width, height))
# index = np.argmax(predictions)
# type_3 = load_img('img_3/09.png', target_size=(width, height))
# predictions = model.predict(type 3 x)
# index = np.argmax(predictions)
# print(class_names[index])
# type_4 = load_img('img_4/10.png', target_size=(width, height))
# type_4_x = np.expand_dims(type_4, axis=0)
# predictions = model.predict(type_4_x)
# index = np.argmax(predictions)
type_1 = load_img('img_1/1.JPG', target_size=(width, height))
plt.imshow(type_1)
plt.show()
type_1_x = np.expand_dims(type_1, axis=0)
predictions = model.predict(type_1_x)
index = np.argmax(predictions)
print(class_names[index])
type_2 = load_img('img_2/1.JPG', target_size=(width, height))
plt.imshow(type_2)
plt.show()
type_2_x = np.expand_dims(type_2, axis=0)
predictions = model.predict(type_2_x)
index = np.argmax(predictions)
print(class_names[index])
type_3 = load_img('img_3/1.JPG', target_size=(width, height))
plt.imshow(type_3)
plt.show()
type 3 x = np.expand dims(type 3, axis=0)
```

```
predictions = model.predict(type_3_x)
   index = np.argmax(predictions)
   print(class_names[index])
   type_4 = load_img('img_4/1.jpg', target_size=(width, height))
   plt.imshow(type_4)
   plt.show()
   type_4_x = np.expand_dims(type_4, axis=0)
   predictions = model.predict(type_4_x)
   index = np.argmax(predictions)
   print(class_names[index])
   answer = input("Do you want to train and evaluate the model again? (y/n): ")
   if answer.lower() == "n" or answer.lower() == "no" or answer.lower() == "N" or
answer.lower() == "no":
       Break
print ("What do you want to name the model file? (e.g. model): ")
model_name = input()
model.save(model_name + '.h5') # Keras model
live_capture()
```

- **1.** Commented out the import statement from 'keras.preprocessing import image' because it is not used.
- **2.** Removed the line '%matplotlib inline' as it is not necessary.
- **3.** Removed the line 'print ("Current working directory %s" % retval) as it is not used.
- **4.** Replaced the values `50:425` and `150:650` with `rectangle\_y1:rectangle\_y2` and `rectangle\_x1:rectangle\_x2` respectively in the code for getting the ROI (region of interest) in the `for` loops.
- **5.** Added a check for the lengths of `X\_type\_3` and `X\_type\_4` before concatenating them to `X` in the code block:

```
if len(X_type_3):
    X = np.concatenate((X, X_type_3), axis=0)

if len(X_type_4):
    X = np.concatenate((X, X_type_4), axis=0)
```

**6.** Added a check for the lengths of 'y\_type\_3` and 'y\_type\_4` before concatenating them to 'y` in the code block:

```
if len(y_type_3):
    y = np.concatenate((y, y_type_3), axis=0)

if len(y_type_4):
    y = np.concatenate((y, y_type_4), axis=0)
```

- 7. Added missing parentheses in the line 'y.shape' to print the shape of 'y'.
- **8.** Added missing indentation to the line '(72, 4)' to align it with the previous line.
- **9.** Removed the unused import statement 'from keras.applications import inception\_v3`.
- **10.** Added missing parentheses in the line 'y = to\_categorical(y, num\_classes=len(class\_names))'.
- **11.** Indented the entire code block under the 'if \_\_name\_\_ == '\_\_main\_\_': ` condition.
- **12.** Added missing 'import statement import seaborn as sns' for generating a confusion matrix.
- **13.** Indented the 'plt show(img)' function.

## **Data Preprocessing:**

- The original code loads images using the load\_img() function from keras.preprocessing.image. In the new code, it directly imports the load\_img() function from keras.utils.
- The loaded images are converted to arrays using img\_to\_array() function from keras.preprocessing.image in the original code. In the new code, it directly imports the img\_to\_array() function from keras.utils.

## **Image Cropping and Saving:**

- The original code manually defined the region of interest (ROI) for cropping images using specific coordinates (50:425, 150:650). In the new code, the ROI is calculated dynamically based on the center coordinates and box dimensions.

- The cropping and saving of images are performed for each type (type\_1, type\_2, type\_3, type\_4) separately using individual loops and saving images to different directories ('img\_1/', 'img\_2/', 'img\_3/', 'img\_4/').
- -After cropping and resizing the ROI, the images are saved using the cv2.imwrite() function.

## **Model Training and Evaluation:**

- The original code defined the model architecture using the Sequential model from keras.models. The new code keeps this architecture definition unchanged.
- The model is compiled with the loss function 'categorical\_crossentropy' and the optimizer 'Adam' in both the original and new code.
- In the new code, a variable history is used to store the training history of the model, which is returned by the model.fit() function.
- The model's training progress and evaluation are visualized using the matplotlib.pyplot library. The new code plots the accuracy and loss curves during training.

### **Live Predictions using Camera:**

- The new code introduces a section for live predictions using a camera. It captures frames from the camera, performs real-time object detection on the captured frames, and overlays the predicted labels on the frames.
- The camera frames are processed similarly to the cropped images, where the ROI is dynamically calculated based on the center coordinates and box dimensions.
- The predictions for the types are obtained using the trained model and displayed as text on the frames using cv2.putText() function.
- The processed frames with predicted labels are displayed in a real-time video feed using cv2.imshow() function.

#### **Import Statements:**

- The original code imports the glob module directly, while the new code imports glob from glob module.
- The original code imports preprocessing from keras, while the new code imports preprocessing from keras.utils.

#### Variables:

- The original code assigns the variable class\_names with different sets of class names based on the commented lines. In the new code, it assigns class\_names to the list ['FORK', 'GLASSES', 'PLATE', 'SPOON'] directly.

## **Directory Creation:**

- The original code does not include code for creating directories to save the cropped images (img\_1/, img\_2/, img\_3/, img\_4/). The new code assumes that these directories already exist and saves the images accordingly.

### **Image Visualization:**

- In the new code, the image visualization using matplotlib.pyplot is modified to show only a subset of images. For example, it shows only the first 5 images for images\_type\_1 and the first 10 images for images\_type\_2, images\_type\_3, and images\_type\_4.
- The subplot titles in the new code are updated to display the corresponding class names dynamically using class\_names list.

### **Live Predictions using Camera:**

- In the new code, the real-time object detection using a camera is implemented at the end. It includes capturing frames from the camera, processing them, performing predictions, and displaying the frames with predicted labels in real-time.
- **14.** Indented the code block under the 'for i in range(len(imgs)): 'loop, which includes the code for displaying images and predictions.
- **15.** Indented the code block under the 'if \_\_name\_\_ == '\_\_main\_\_': condition, which includes the code for live predictions using the camera.

```
# Calculate the center of the window
  window_center_x = frame.shape[1] // 2
  window_center_y = frame.shape[0] // 2

# Calculate the new width and height for the bounding box
  box_width = 400
  box_height = int(box_width / aspect)

# Calculate the offset for centering the bounding box
  offset_x = (frame.shape[1] - box_width) // 2
  offset_y = (frame.shape[0] - box_height) // 2

# Calculate the new coordinates for the centered bounding box
  rectangle_x1 = window_center_x - (box_width // 2)
  rectangle_y1 = window_center_y - (box_height // 2)
  rectangle_x2 = window_center_x + (box_width // 2)
```

```
rectangle_y2 = window_center_y + (box_height // 2)

# Draw the centered bounding box
    cv2.rectangle(frame, (rectangle_x1, rectangle_y1), (rectangle_x2, rectangle_y2), (0, 255,
0), 2)
```

- **16.** The center of the window or frame is calculated by dividing the width and height of the frame by 2 using the // operator. These values are stored in window\_center\_x and window\_center\_y, respectively.
- **17.** The code specifies the desired width of the bounding box as box\_width. To maintain the aspect ratio, the corresponding height is calculated as box height = int(box width / aspect).
- **18.** An offset is computed to center the bounding box within the frame. The offset\_x is determined by subtracting the box width from the frame width and dividing by 2 ((frame.shape[1] box\_width) // 2). Similarly, offset\_y is calculated by subtracting the box height from the frame height and dividing by 2 ((frame.shape[0] box\_height) // 2).
- **19.** The new coordinates for the centered bounding box are determined based on the window center and the calculated box dimensions. The top-left corner of the box is specified as (rectangle\_x1, rectangle\_y1), which is obtained by subtracting half the box width from the window center (window\_center\_x (box\_width // 2)) for the x-coordinate and subtracting half the box height from the window center (window\_center\_y (box\_height // 2)) for the y-coordinate. Similarly, the bottom-right corner of the box is given by (rectangle\_x2, rectangle\_y2), which is obtained by adding half the box width to the window center (window\_center\_x + (box\_width // 2)) for the x-coordinate and adding half the box height to the window center (window center y + (box\_height // 2)) for the y-coordinate.
- **20.** Finally, a bounding box is drawn on the frame using cv2.rectangle(). The function takes the frame as the first argument, the top-left corner coordinates (rectangle\_x1, rectangle\_y1), the bottom-right corner coordinates (rectangle\_x2, rectangle\_y2), the color of the rectangle (in this case, green with (0, 255, 0)), and the thickness of the rectangle border (2 pixels).

```
for i, frame in enumerate(raw_frames_type_1):
    #get roi
    roi = frame[rectangle_y1:rectangle_y2, rectangle_x1:rectangle_x2]
    # roi = frame[50:425, 150:650]
```

- **21.** The region of interest (ROI) is calculated using array slicing in Python. The ROI represents a specific rectangular region within the frame image.
- **22.** The ROI is extracted from the frame image using array slicing. The syntax [y1:y2, x1:x2] is used to specify the region of interest.

- **23.** The rectangle\_y1 and rectangle\_y2 variables represent the vertical (y-axis) coordinates of the top-left corner and the bottom-right corner of the bounding box, respectively. These coordinates define the range of rows in the frame that will be included in the ROI.
- **24.** Similarly, the rectangle\_x1 and rectangle\_x2 variables represent the horizontal (x-axis) coordinates of the top-left corner and the bottom-right corner of the bounding box, respectively. These coordinates define the range of columns in the frame that will be included in the ROI.
- **25.** By specifying frame[rectangle\_y1:rectangle\_y2, rectangle\_x1:rectangle\_x2], the code slices the frame array to extract the region of interest defined by the specified row and column ranges.
- **26.** The resulting roi variable contains the extracted portion of the frame image that corresponds to the bounding box defined by the coordinates (rectangle x1, rectangle y1) and (rectangle x2, rectangle y2).

```
def live_capture():
    model_name_live= (input("What is the name of your model? (h5 format, extension will be
automatically added): "))
    model = load_model(model_name_live + ".h5")
```

- **27.** Added a function "live\_capture()" to be used as direct path if the user does not want to train any further and just want to proceed to live detection.
- **28.** Given the option to choose the name of the model as well, given that it is the current directory that the terminal is on.
- **29.** Automatically adds the ".h5" extension to the model name specified by the user.

```
answer = input("Do you want to train and evaluate the model again? (y/n): ")

if answer.lower() == "n" or answer.lower() == "no" or answer.lower() == "N" or
answer.lower() == "no":
    break

# Options for saving the model
print ("What do you want to name the model file? (e.g. model): ")
model_name = input()
model.save(model_name + '.h5') # Keras model

live_capture()
```

- **30.** Converted the whole model building and evaluation into a while statement that will repeat if the user does not choose the key words/letters: (N, n, no, No).
- **31.** Gave an option to save the model in a custom file name that automatically adds the extension as well.
- **32.** After the saving, the code will proceed to live detection.

-----END-----