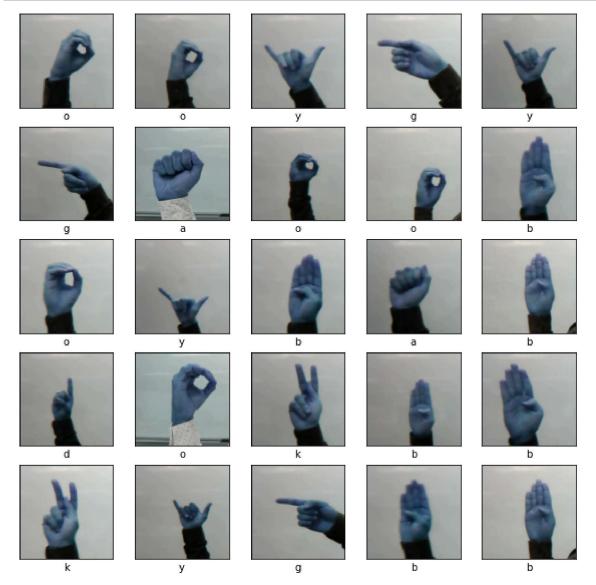
```
In [12]:
      import os
      import random
      from imutils import paths
      import matplotlib.pyplot as plt
      import argparse
      import cv2
      import tensorflow as tf
      from tensorflow.keras.models import load model
      from sklearn.preprocessing import LabelBinarizer
      from sklearn.model_selection import train_test_split
      from sklearn.metrics import classification report
      from tensorflow.keras.models import Sequential
      from tensorflow.keras.layers import Dense
      from tensorflow.keras.optimizers import SGD
      import numpy as np
      from tensorflow import keras
In [13]: | data = []
      labels = []
      imagePaths = list(paths.list_images("gesture"))
      for imagePath in imagePaths:
        image = cv2.imread(imagePath)
        data.append(image)
        # extract the label from the image path and update the labels list
        label = imagePath.split(os.path.sep)[-1][0]
        labels.append(int(label))
In [3]: |data = np.array(data, dtype="float") / 255.0
      labels = np.array(labels)
In [4]: labels
2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 3, 3, 3, 3, 3, 3, 3, 3,
          3, 3, 3, 3, 3, 3, 3, 3, 3, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4,
          7, 7, 7, 7, 7, 7, 7, 7, 7, 7])
In [5]: (trainX, testX, trainY, testY) = train_test_split(data,
        labels, test_size=0.15, random_state=22)
```

```
In [48]: plt.figure(figsize=(10,10))
    for i in range(25):
        plt.subplot(5,5,i+1)
        plt.xticks([])
        plt.yticks([])
        plt.grid(False)
        plt.imshow(trainX[i], cmap=plt.cm.gray)
        plt.xlabel(trainY[i])
    plt.show()
```



```
In [7]: | model = keras.Sequential([
          keras.layers.InputLayer(input_shape=(400, 400, 3)),
          keras.layers.Reshape(target shape=(400, 400, 3)),
          keras.layers.Conv2D(filters=32, kernel size=(3, 3), activation=tf.nn.relu),
          keras.layers.Conv2D(filters=64, kernel_size=(3, 3), activation=tf.nn.relu),
          keras.layers.MaxPooling2D(pool_size=(2, 2)),
          keras.layers.Dropout(0.25),
          keras.layers.Flatten(),
          keras.layers.Dense(10)
        ])
        # Define how to train the model
        model.compile(optimizer='adam',
                      loss=tf.keras.losses.SparseCategoricalCrossentropy(from logits=True
                     metrics=['accuracy'])
        # Train the digit classification model
        model.fit(trainX, trainY, epochs=5)
        Epoch 1/5
        8/8 [============ ] - 140s 18s/step - loss: 31.4515 - accurac
        y: 0.1030
        Epoch 2/5
        8/8 [============== ] - 121s 15s/step - loss: 2.8156 - accuracy:
        0.3305
```

8/8 [============] - 117s 15s/step - loss: 0.9184 - accuracy:

8/8 [==============] - 117s 15s/step - loss: 0.4348 - accuracy:

8/8 [===========] - 117s 15s/step - loss: 0.1656 - accuracy:

Out[7]: <tensorflow.python.keras.callbacks.History at 0x28568e57908>

Epoch 3/5

0.6953 Epoch 4/5

0.8412 Epoch 5/5

0.9700

In [8]: model.summary()

Model: "sequential_1"

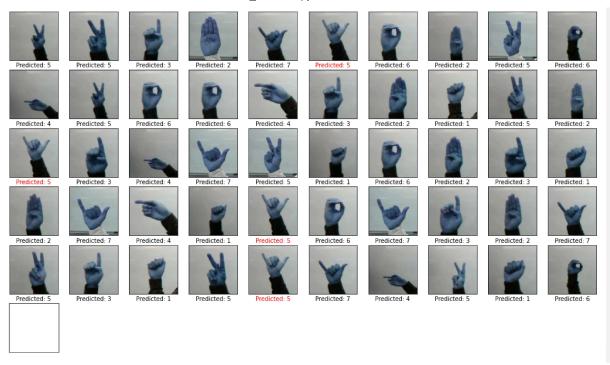
| Layer (type) | Output | Shape | Param # |
|------------------------------|--------|---------------|----------|
| reshape_1 (Reshape) | (None, | 400, 400, 3) | 0 |
| conv2d_2 (Conv2D) | (None, | 398, 398, 32) | 896 |
| conv2d_3 (Conv2D) | (None, | 396, 396, 64) | 18496 |
| max_pooling2d_1 (MaxPooling2 | (None, | 198, 198, 64) | 0 |
| dropout_1 (Dropout) | (None, | 198, 198, 64) | 0 |
| flatten_1 (Flatten) | (None, | 2509056) | 0 |
| dense_1 (Dense) | (None, | 10) | 25090570 |

Total params: 25,109,962 Trainable params: 25,109,962 Non-trainable params: 0

Test accuracy: 0.9047619104385376

```
In [15]: def get label color(val1, val2):
           if val1 == val2:
             return 'black'
           else:
             return 'red'
         # Predict the labels of digit images in our test dataset.
         predictions = model.predict(testX)
         prediction_digits = np.argmax(predictions, axis=1)
         # dataset, we will highlight it in red color.
         plt.figure(figsize=(18, 18))
         for i in range(100):
           ax = plt.subplot(10, 10, i+1)
           plt.xticks([])
           plt.yticks([])
           plt.grid(False)
           image_index = random.randint(0, len(prediction_digits))
           plt.imshow(testX[image_index], cmap=plt.cm.gray)
           ax.xaxis.label.set color(get label color(prediction digits[image index],\
                                                     testY[image index]))
           plt.xlabel('Predicted: %d' % prediction_digits[image_index])
         plt.show()
```

IndexError: index 42 is out of bounds for axis 0 with size 42



```
In [17]: # Convert Keras model to TF Lite format.
    converter = tf.lite.TFLiteConverter.from_keras_model(model)
    tflite_float_model = converter.convert()

# Show model size in KBs.
    float_model_size = len(tflite_float_model) / 1024
    print('Float model size = %dKBs.' % float_model_size)
```

Float model size = 98087KBs.

Quantized model size = 24526KBs, which is about 25% of the float model size.

```
In [27]: def evaluate tflite model(tflite model):
           # Initialize TFLite interpreter using the model.
           interpreter = tf.lite.Interpreter(model content=tflite model)
           interpreter.allocate tensors()
           input tensor_index = interpreter.get_input_details()[0]["index"]
           output = interpreter.tensor(interpreter.get_output_details()[0]["index"])
           # Run predictions on every image in the "test" dataset.
           prediction digits = []
           for test_image in trainX:
             # Pre-processing: add batch dimension and convert to float32 to match with
             # the model's input data format.
             test_image = np.expand_dims(test_image, axis=0).astype(np.float32)
             interpreter.set_tensor(input_tensor_index, test_image)
             # Run inference.
             interpreter.invoke()
             # Post-processing: remove batch dimension and find the digit with highest
             # probability.
             digit = np.argmax(output()[0])
             prediction_digits.append(digit)
           # Compare prediction results with ground truth labels to calculate accuracy.
           accurate count = 0
           for index in range(len(prediction digits)):
             if prediction digits[index] == trainY[index]:
               accurate count += 1
           accuracy = accurate count * 1.0 / len(prediction digits)
           return accuracy
         # Evaluate the TF Lite float model. You'll find that its accurary is identical
         # to the original TF (Keras) model because they are essentially the same model
         # stored in different format.
         float accuracy = evaluate tflite model(tflite float model)
         print('Float model accuracy = %.4f' % float_accuracy)
         # Evalualte the TF Lite quantized model.
         # Don't be surprised if you see quantized model accuracy is higher than
         # the original float model. It happens sometimes :)
         quantized accuracy = evaluate tflite model(tflite quantized model)
         print('Quantized model accuracy = %.4f' % quantized_accuracy)
         print('Accuracy drop = %.4f' % (float_accuracy - quantized_accuracy))
         Float model accuracy = 0.9957
         Quantized model accuracy = 0.9957
         Accuracy drop = 0.0000
In [28]: | f = open('gesture.tflite', "wb")
         f.write(tflite quantized model)
         f.close()
```

In []: