```
In [1]: # ATTENTION: Please do not alter any of the provided code in the exercise. Only of # ATTENTION: Please do not add or remove any cells in the exercise. The grader with attention with attention with a training.

import csv import numpy as not import tensorflow as the from tensorflow.keras.preprocessing.image import ImageDataGenerator from os import getcwd
```

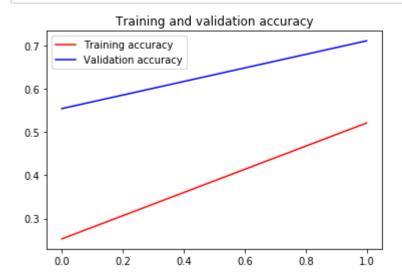
```
In [18]: | def get_data(filename):
           # You will need to write code that will read the file passed
           # into this function. The first line contains the column headers
           # so you should ignore it
           # Each successive line contians 785 comma separated values between 0 and 255
           # The first value is the label
           # The rest are the pixel values for that picture
           # The function will return 2 np.array types. One with all the labels
           # One with all the images
           # Tips:
           # If you read a full line (as 'row') then row[0] has the label
           # and row[1:785] has the 784 pixel values
           # Take a look at np.array split to turn the 784 pixels into 28x28
           # You are reading in strings, but need the values to be floats
           # Check out np.array().astype for a conversion
             with open(filename) as training file:
               # Your code starts here
                  lines_excluding_header = training_file.readlines()[1:]
                  size = len(lines excluding header)
                  labels = np.zeros(size)
                  images = np.zeros((size, 28, 28))
                  for index, line in enumerate(lines excluding header):
                      individual item = line.strip().split(",")
                      if line.strip().split(","):
                          #only first item is a label
                          labels[index] = int(individual item[0])
                          #everthing else is an image
                          image = np.asarray(individual item[1:], dtype=np.float32)
                          image = np.array split(image, 28)
                          images[index, :, :] = image
                # Your code ends here
              return images, labels
         path_sign_mnist_train = f"{getcwd()}/../tmp2/sign_mnist_train.csv"
         path_sign_mnist_test = f"{getcwd()}/../tmp2/sign_mnist_test.csv"
         training images, training labels = get data(path sign mnist train)
         testing_images, testing_labels = get_data(path_sign_mnist_test)
         # Keep these
         print(training_images.shape)
         print(training labels.shape)
         print(testing_images.shape)
         print(testing labels.shape)
         # Their output should be:
         # (27455, 28, 28)
         # (27455,)
         # (7172, 28, 28)
         # (7172,)
         (27455, 28, 28)
         (27455,)
         (7172, 28, 28)
         (7172,)
```

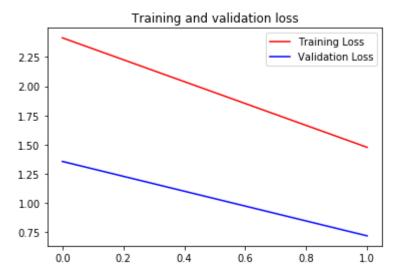
```
In [19]: # In this section you will have to add another dimension to the data
         # So, for example, if your array is (10000, 28, 28)
         # You will need to make it (10000, 28, 28, 1)
         # Hint: np.expand dims
         training_images = np.expand_dims(training_images, axis=3)
         testing images = np.expand dims(testing images, axis=3)
         # Create an ImageDataGenerator and do Image Augmentation
         train_datagen = ImageDataGenerator(
             rescale=1/255,
             rotation range=40,
             width_shift_range=0.2,
             height shift range=0.2,
             shear range=0.2,
             zoom_range=0.2,
             horizontal flip=True,
             fill mode='nearest'
         )
         validation datagen = ImageDataGenerator(
             rescale=1/255
         )
         # Keep These
         print(training images.shape)
         print(testing images.shape)
         # Their output should be:
         # (27455, 28, 28, 1)
         # (7172, 28, 28, 1)
         (27455, 28, 28, 1)
         (7172, 28, 28, 1)
```

```
In [32]: # Define the model
         # Use no more than 2 Conv2D and 2 MaxPooling2D
         from tensorflow.keras.optimizers import RMSprop
         model = tf.keras.models.Sequential([
             tf.keras.layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1
             tf.keras.layers.MaxPooling2D(2, 2),
             tf.keras.layers.Conv2D(64, (3, 3), activation='relu'),
             tf.keras.layers.MaxPooling2D(2, 2),
             tf.keras.layers.Flatten(),
             tf.keras.layers.Dense(512, activation='relu'),
             tf.keras.layers.Dropout(0.2),
             tf.keras.layers.Dense(25, activation='softmax')
         ])
         train_generator = train_datagen.flow(training_images, training_labels)
         validation_generator = validation_datagen.flow(testing_images, testing_labels)
         # Compile Model.
         model.compile(
             optimizer=RMSprop(lr=0.002),
             loss='sparse categorical crossentropy',
             metrics=['acc']
         )
         # Train the Model
         history = model.fit_generator(
             train generator,
             epochs=2,
             validation_data=validation_generator,
             verbose=2
         )
         model.evaluate(testing_images, testing_labels, verbose=0)
         Epoch 1/2
         858/858 - 76s - loss: 2.4145 - acc: 0.2528 - val_loss: 1.3569 - val_acc: 0.5540
         Epoch 2/2
         858/858 - 77s - loss: 1.4776 - acc: 0.5208 - val loss: 0.7201 - val acc: 0.7107
```

Out[32]: [190.11476199080408, 0.5064138]

```
In [34]:
         # Plot the chart for accuracy and loss on both training and validation
         %matplotlib inline
         import matplotlib.pyplot as plt
         acc = history.history['acc']
         val_acc = history.history['val_acc']
         loss = history.history['loss']
         val loss = history.history['val loss']
         epochs = range(len(acc))
         plt.plot(epochs, acc, 'r', label='Training accuracy')
         plt.plot(epochs, val_acc, 'b', label='Validation accuracy')
         plt.title('Training and validation accuracy')
         plt.legend()
         plt.figure()
         plt.plot(epochs, loss, 'r', label='Training Loss')
         plt.plot(epochs, val_loss, 'b', label='Validation Loss')
         plt.title('Training and validation loss')
         plt.legend()
         plt.show()
```





## **Submission Instructions**

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In [ ]: # Now click the 'Submit Assignment' button above.
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When you're done or would like to take a break, please run the two cells below to save your work and close the Notebook. This will free up resources for your fellow learners.