CSEE5590 - Python and Deep Learning for Engineering & Science

Project Increment – 3

Team number: 5

Project Title

Sign Language for Numbers

Team Members

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The Story and its details:

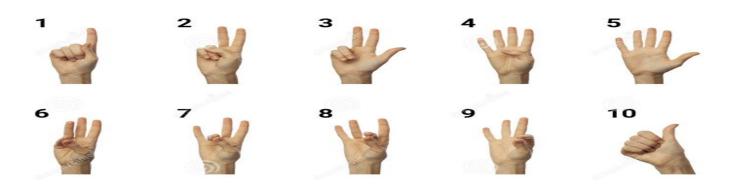
There have been several milestone reached in innovation and a great deal of technology advancement and research has been done to help the individuals who are physically challenged. Supporting the reason, Deep learning and computer vision can be utilized in an extensive way for community of people who can't speak or listen.

The Sign language is very important for people who have hearing and speaking deficiency generally called Deaf and Mute. It is the only mode of communication for such people to convey their messages and it becomes very important for people to understand their language.

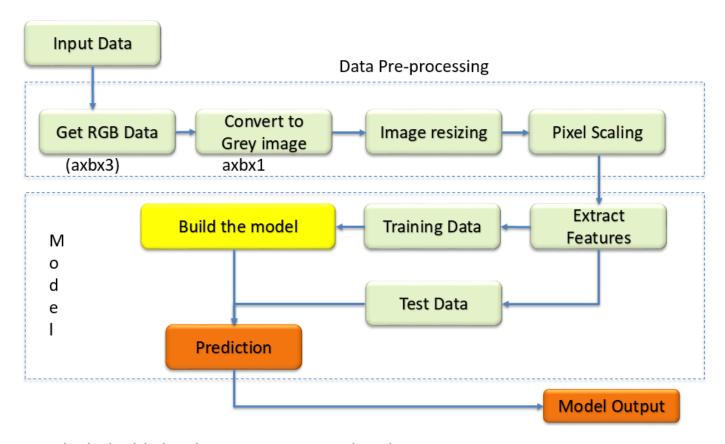
This project proposes the method or algorithm for an application which would help to recognize the different signs using palm/finger gestures. This will ease and make a better communication channel to rest of the world.

The Data and its details:

- Sign language datasets for hand gesture recognition for numbers
- Source of data Kaggle Link in reference section
- Sample hand signs are in below picture
- Total number of training images 1500 images for each sign = 16,500
- Covering hand images 0-9 and unknown (Total classes -11)
- Input data type RGB Images(3 channels)
- Assumption All input images are front facing

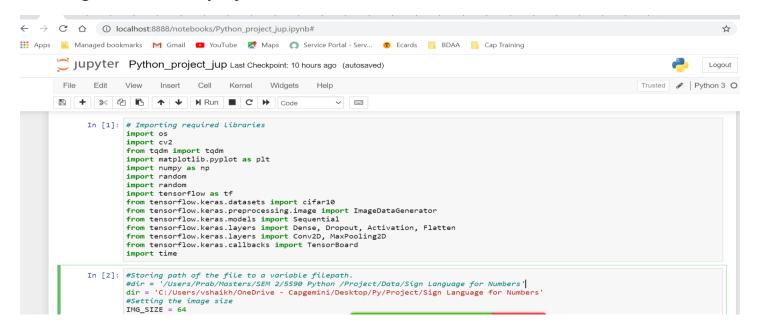


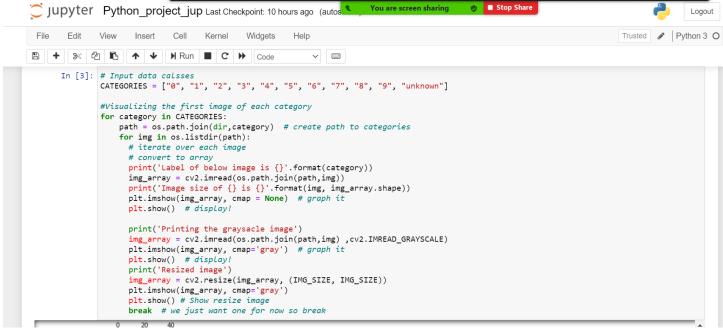
The Model Building Blocks:



- 1. Blocks highlighted in green are completed
- 2. Build the model Model 1 (CNN)implementation is complete and evaluation is in progress
- 3. Prediction and ensemble modelling Next steps

Working screens from project:







```
□ | T | 8 | 4 | 1 | T | T | T | T | T | Code
                                                                       ~
       In [4]: # Defining a funcation to read image data, resizing and convering to required format
                 def get_images(directory):
                     output = []
                      print("Loading {}".format(directory))
                      # Iterate through each folder corresponding to a category
                      for category in CATEGORIES:
                          path = os.path.join(directory,category) # create path to categories
class_num = CATEGORIES.index(category) # get the classification
                           print('Processing folder {}'.format(category))
                           # Iterate through each image in folder
                           for file in tqdm(os.listdir(path)):
                             # Get the path name of the image
                             img_path = os.path.join(path, file)
                             # Open and read the image in grayscale
                             image_array = cv2.imread(img_path,cv2.IMREAD_GRAYSCALE)
                             # Resize the image
                             image_resize = cv2.resize(image_array, (IMG_SIZE, IMG_SIZE))
                             #print('type of Image', type(Images))
#print('type of image', type(image))
# Append the image and its corresponding label to the output
                             output.append([image_resize, class_num])
                          #Images = np.array(Images, dtype = 'float32')
#Labels = np.array(Labels, dtype = 'int32')
                      return output
```

```
In [5]: # Loading input image data in list
              image_with_label = get_images(dir)
                                                                                                    | 35/1500 [00:00<00:04, 349.14it/s]
              \label{loading C:Users/vshaikh/OneDrive - Capgemini/Desktop/Py/Project/Sign \ Language \ for \ Numbers
              Processing folder 0
                         | 1500/1500 [00:03<00:00, 458.06it/s]
                                                                                                    | 33/1500 [00:00<00:04, 324.32it/s]
              Processing folder 1
              100%|
3%|
                                                                                                   1500/1500 [00:03<00:00, 458.99it/s]
                                                                                                    | 47/1500 [00:00<00:03, 466.59it/s]
              Processing folder 2
                                                                                                   1500/1500 [00:02<00:00, 511.77it/s]
                                                                                                    | 53/1500 [00:00<00:02, 526.15it/s]
              Processing folder 3
              100%
                                                                                                   1500/1500 [00:03<00:00, 493.34it/s]
                6%|
                                                                                                   | 86/1500 [00:00<00:03, 409.93it/s]
               Processing folder 4
              100%|
                                                                                                   1500/1500 [00:02<00:00, 502.75it/s]
                7%
                                                                                                    | 98/1500 [00:00<00:03, 451.44it/s]
              Processing folder 5
                                                                                                    1500/1500 [00:06<00:00, 245.13it/s]
                100%
                                                                                                     | 25/1500 [00:00<00:06, 241.04it/s]
                  2% İ
                Processing folder 9
                100%
                                                                                                    1500/1500 [00:03<00:00, 385.66it/s]
                  3%|
                                                                                                     | 43/1500 [00:00<00:03, 422.40it/s]
                Processing folder unknown
                                                                                                 | 1500/1500 [00:02<00:00, 562.62it/s]
                100%
        In [6]: # Checking input data length
                len(image_with_label)
        Out[6]: 16500
        In [7]: # Plotting image from input data
                plt.imshow(image_with_label[1500][0], cmap='gray') # graph it
        Out[7]: <matplotlib.image.AxesImage at 0x1fe07f80970>
                  0
                 10
                 20
                 30
                 40
~
      In [8]: # Input data lebel length
len(image_with_label[0][0])
      Out[8]: 64
      In [9]: # Printing image data at position 0
print(image_with_label[0])
              ...,
[117, 119, 114, ..., 55, 80, 98],
[111, 116, 121, ..., 57, 74, 82],
[108, 117, 133, ..., 65, 74, 64]], dtype=uint8), 0]
     In [10]: # Print image data at poisiton 15000
print(image_with_label[15000])
              ...,
[223, 220, 219, ..., 212, 210, 212],
[218, 216, 216, ..., 213, 205, 171],
[219, 216, 216, ..., 199, 136, 114]], dtype=uint8), 10]
```

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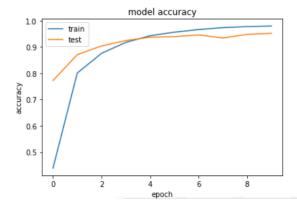
```
In [11]: # performing randon suffle
random.shuffle(image_with_label)
            # Checking sample data
for sample in image with_label[:10]:
    print(sample[1])
            8
            10
            5
             6
             10
 In [12]: # Splitting input data images into X (features)and Y (label)
            X = []
Y = []
            for features,label in image_with_label:
    X.append(features)
    Y.append(label)
            print(X[0].reshape(-1, IMG_SIZE, IMG_SIZE, 1))
            X = np.array(X).reshape(-1, IMG_SIZE, IMG_SIZE, 1)
             [[[135]
In [12]: # Splitting input data images into X (features)and Y (label)
           X = []
Y = []
            for features,label in image_with_label:
                X.append(features)
                 Y.append(label)
            print(X[0].reshape(-1, IMG_SIZE, IMG_SIZE, 1))
           X = np.array(X).reshape(-1, IMG_SIZE, IMG_SIZE, 1)
            [[[[135]
                [137]
                [140]
                [131]
                [129]
               [127]]
              [[136]
                [139]
                [142]
               [133]
                [131]
               [129]]
              [[138]
                  الالادمتا
   In [13]: Y
               9,
7,
3,
9,
9,
5,
8,
4,
7,
2,
5,
2,
5,
0,
               7,
2,
               1,
```

```
In [15]: # Using pickle to serialize the data
import pickle
pickle_out = open("C:/Users/vshaikh/OneDrive - Capgemini/Desktop/Py/Project/Sign Language for Numbers/X.pickle","wb")
pickle_dump(X, pickle_out)
pickle_out.close()

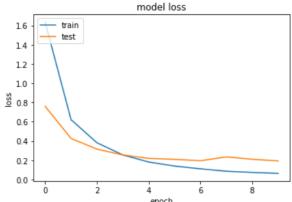
pickle_out = open("C:/Users/vshaikh/OneDrive - Capgemini/Desktop/Py/Project/Sign Language for Numbers/Y.pickle","wb")
pickle_dump(Y, pickle_out)
pickle_out.close()
```

```
In [*]: # Loading pickel data
        pickle_in = open("C:/Users/vshaikh/OneDrive - Capgemini/Desktop/Py/Project/Sign Language for Numbers/X.pickle","rb")
        X = pickle.load(pickle_in)
        X = np.array(X).reshape(-1, IMG_SIZE, IMG_SIZE, 1)
        pickle in = open("C:/Users/vshaikh/OneDrive - Capgemini/Desktop/Py/Project/Sign Language for Numbers/Y.pickle","rb")
          = pickle.load(pickle_in)
        Y = np.array(Y)
        # Normalizing the data
        X = X/255.0
        # Defining sequential model (CNN) -- Model1
        model = Sequential()
        model.add(Conv2D(16, (2,2), input_shape=X.shape[1:], activation='relu'))
        model.add(MaxPoolingZD(pool_size=(2, 2), strides=(2, 2), padding='same'))
model.add(Conv2D(32, (3,3), activation='relu'))
        model.add(MaxPooling2D(pool_size=(3, 3), strides=(3, 3), padding='same'))
        model.add(Conv2D(64, (5,5), activation='relu'))
        model.add(MaxPooling2D(pool_size=(5, 5), strides=(5, 5), padding='same'))
        model.add(Flatten())
        model.add(Dense(128, activation='relu'))
        model.add(Dropout(0.2))
        model.add(Dense(11, activation='softmax')) # size must be equal to number of classes i.e. 11
        # Compiling the model
        model.compile(loss='sparse_categorical_crossentropy',
                       optimizer='adam',
                       metrics=['accuracy'])
        # Fitting model on training data
        model.fit(X, Y, batch_size=32, epochs=10, validation_split=0.2)
```

```
In [17]: # Plotting Accuracy and Loss using history object
import matplotlib.pyplot as plt
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```



```
In [18]: plt.plot(history.history['loss'])
    plt.plot(history.history['val_loss'])
    plt.title('model loss')
    plt.ylabel('loss')
    plt.xlabel('epoch')
    plt.legend(['train', 'test'], loc='upper left')
    plt.show()
```



Work sharing/Module sharing between teammates:

Prabhajan Trivedi -

- 1. Worked on prepressing of data
- 2. Now working on building model 2

Harshita Patil -

- 1. Working on writing preprocessed data to CSV file so that it can be shared across multiple models that we are planning to implement.
- 2. Worked on building first model 1

Vasim Shaikh -

- 1. Used pre-processed data and worked on building first model and training the model 1
- 2. Worked on documentation and collaborating with team members

Shireesha Maddi -

- 1. Worked on data preprocessing along with Prabhanjan T
- 2. Now working on model 3 and visualization

Any issues, blockages with the project:

1. Converting image data in to CSV so that it can used as an input for multiple models

```
import pandas as pd
       import matplotlib.pyplot as plt
       import numpy as np
[10] dfi = pd.read_csv('/content/drive/MyDrive/img.csv')
[11] dfi.shape
       (55, 3)
[12] X = []
       X = dfi[1]
    X = dfi[1]
                                           Traceback (most recent call last)
    /usr/local/lib/python3.6/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
                          return self._engine.get_loc(casted_key)
     -> 2895
       2896
                       except KeyError as err:
    pandas/_libs/index.pyx in pandas._libs.index.IndexEngine.get_loc()
    pandas/_libs/index.pyx in pandas._libs.index.IndexEngine.get_loc()
    pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
    pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
    KeyError: 1
    The above exception was the direct cause of the following exception:
                                           Traceback (most recent call last)
                                     2 frames
    /usr/local/lib/python3.6/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
       2895
                          return self._engine.get_loc(casted_key)
       2896
                       except KeyError as err:
     -> 2897
                          raise KeyError(key) from err
       2898
       2899
                   if tolerance is not None:
    KeyError: 1
```

2. Unable to upload data to Github due to size issue

Next Steps –

- 1. Complete remaining model building and training
- 2. Implementation of ensemble model
- 3. Model evaluation

Github Link - https://github.com/vasimshaikh39/python2020

References:

- 1. https://www.kaggle.com/muhammadkhalid/sign-language-for-numbers
- 2. https://heartbeat.fritz.ai/introduction-to-machine-learning-model-evaluation-fa859e1b2d7f
- 3. https://data-flair.training/blogs/sign-language-recognition-python-ml-opency/
- 4. https://towardsdatascience.com/metrics-to-evaluate-your-machine-learning-algorithm-f10ba6e38234
- 5. https://www.kaggle.com/joshbeau/numerical-sign-language-recognition-tensorflow