

Mini Project Report

American Sign Language Detection

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Statement of Purpose:

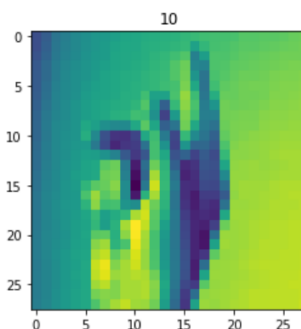
Develop a Machine Learning model for detection of American Sign Language Detection and Conversion. MNIST dataset is used for training the model using Keras and Tensor flow. By applying neural network algorithms we develop a model which can identify sign language with an accuracy of 70%. We use a classification algorithm to classify the image into 26 classes.

Results on sample data:

```
In [0]: img = test_X[6]
test_img = img.reshape((1,784))
img_class = model.predict_classes(test_img)
prediction = img_class[0]
classname = img_class[0]
print("Class: ",classname)
```

Class: 10

```
In [0]: img = img.reshape((28,28))
plt.imshow(img)
plt.title(classname)
plt.show()
```



When a test image is passed to the model it classifies it and gives the class (label) of that image which is passed.

Dataset used:

MNIST DATASET

<https://www.kaggle.com/datamunge/sign-language-mnist>

Literature Referenced:

International Conference on Advances in Computing, Communication Control and Networking (ICACCCN2018)

Dynamic Tool for American Sign Language Finger Spelling Interpreter

ISBN:978-1-5386-4119-4/18

Technologies Used:

Keras

Python

Tensorflow

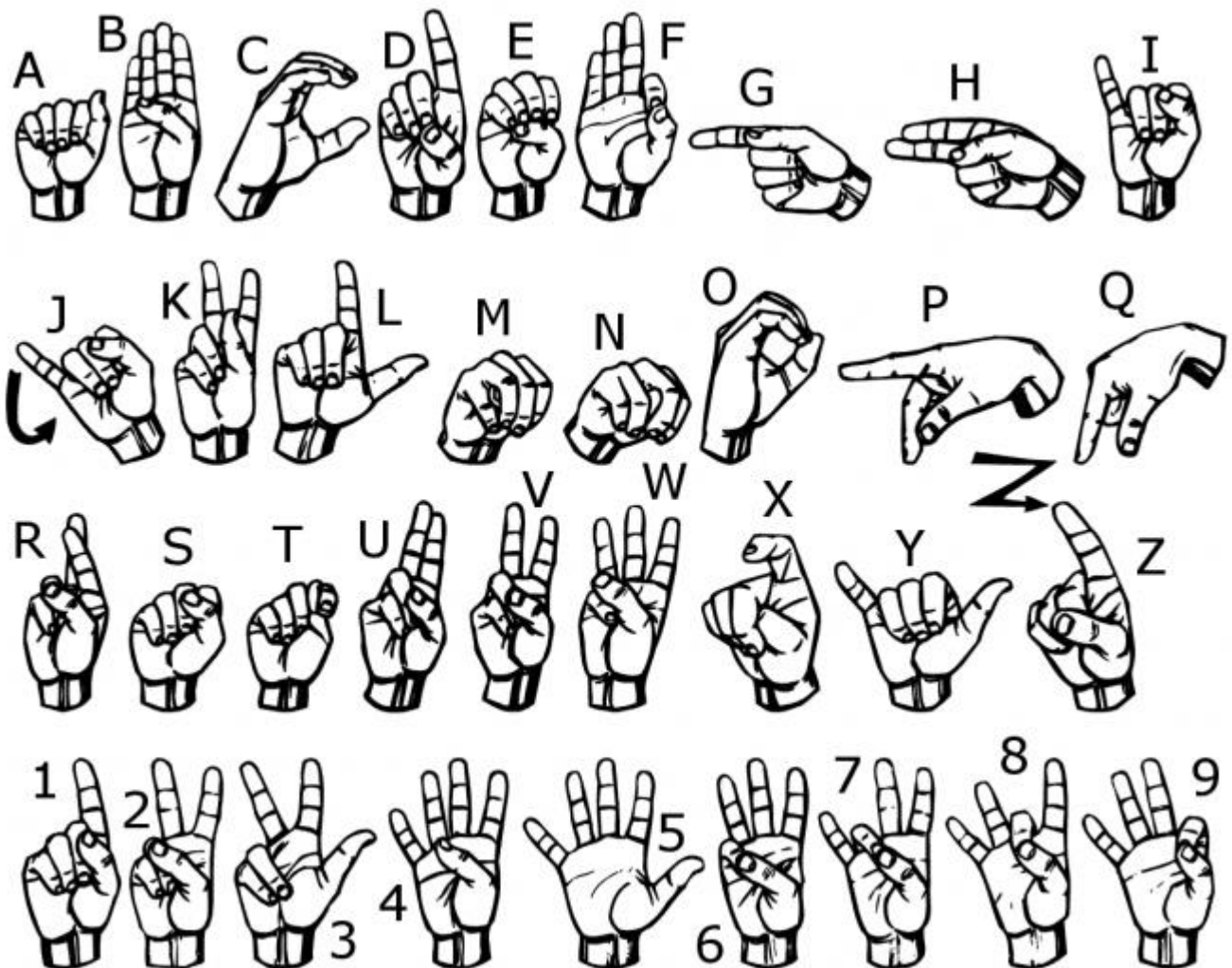
Google Colab

Machine Learning Schema

1) Dataset

We will use MNIST (Modified [National Institute of Standards and Technology](#)) dataset.

Basically, our dataset consists many images of 24 (except J and Z) American Sign Language alphabets. Each image has size 28x28 pixel which means total 784 pixels per image.



- Loading the dataset to Colab

Our dataset is in CSV (Comma-separated values) format. train_X and test_X contain the values of each pixel. train_Y and test_Y contains the label of image. You can use the following code to see the dataset:

- **Pre-processing**

train_X and test_X consists an array of all the pixel pixel values. We have to create an image from these values. Our image size is 28x28 hence we have to divide the array into 28x28 pixel groups. To do that we will use the following code:

2) Build and Train the Model

We will use CNN (Convolutional Neural Network) to recognise the alphabets. We are going to use keras.

As you can observe, like any other CNN our model consists couple of Conv2D and MaxPooling layers followed by some fully connected layers (Dense).

The first Conv2D (Convolutional) layer takes input image of shape (28, 28, and 1). The last fully connected layer gives us output for 26 alphabets.

We are using a Dropout after 2nd Conv2D layer to regularise our training.

We are using softmax activation function in the final layer. Which will give us probability for each alphabet as an output.

Layer (type)	Output Shape	Param #
=====		
conv2d_1 (Conv2D)	(None, 28, 28, 8)	80
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 8)	0
conv2d_2 (Conv2D)	(None, 14, 14, 16)	1168
dropout_1 (Dropout)	(None, 14, 14, 16)	0
max_pooling2d_2 (MaxPooling2D)	(None, 3, 3, 16)	0
dense_1 (Dense)	(None, 3, 3, 128)	2176
flatten_1 (Flatten)	(None, 1152)	0
dense_2 (Dense)	(None, 26)	29978
=====		
Total params: 33,402		
Trainable params: 33,402		
Non-trainable params: 0		

We are using [SGD](#) optimiser to compile our model.