

# Model Building

## Importing libraries

```
# This library helps add support for large, multi-dimensional arrays and matrices import
numpy as np
#open source used for both ML and DL for computation
import tensorflow as tf #it is a plain stack of layers
from tensorflow.keras.models import Sequential
#Dense Layer is the regular deeply connected neural network layer from
tensorflow.keras.layers import Dense, Flatten, Dropout
#Flatten-used for flattening the input or change the dimension, MaxPooling2D-for downsampling
from tensorflow.keras.layers import Convolution2D, MaxPooling2D
#Its used for different augmentation of the image
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

## Augmenting the data

```
#setting parameter for Image Data augmentation to the training data
train_datagen = ImageDataGenerator(rescale=1./255, shear_range=0.2,
zoom_range=0.2, horizontal_flip=True)
#Image Data augmentation to the testing data
test_datagen=ImageDataGenerator(rescale=1./255)
```

## Loading our data and performing data augmentation

```
#performing data agumentation to train data
x_train = train_datagen.flow_from_directory(r'C:\\Users\\Midhun\\OneDrive\\Desktop\\Gestu
                                             target_size=(64, 64), batch_size=3,
                                             color_mode='grayscale',
                                             class_mode='categorical')

#performing data agumentation to test data
x_test = test_datagen.flow_from_directory(r'C:\\Users\\Midhun\\OneDrive\\Desktop\\Gesture
                                           target_size=(64, 64), batch_size=3,
                                           color_mode='grayscale',
                                           class_mode='categorical')
```

Found 594 images belonging to 6 classes. Found  
30 images belonging to 6 classes.

```
print(x_train.class_indices)#checking the number of classes
{'0': 0, '1': 1, '2': 2, '3': 3, '4': 4, '5': 5}
```

## Model Creation

```
# Initializing the CNN model = Sequential()
```

```
# First convolution layer and pooling
model.add(Convolution2D(32, (3, 3), input_shape=(64, 64, 1), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
```

```
# Second convolution layer and pooling
model.add(Convolution2D(32, (3, 3), activation='relu'))
# input_shape is going to be the pooled feature maps from the previous convolution layer
model.add(MaxPooling2D(pool_size=(2, 2)))
```

```
# Flattening the layers i.e. input layer model.add(Flatten())
```

```
# Adding a fully connected layer, i.e. Hidden Layer model.add(Dense(units=512
, activation='relu'))
```

```
# softmax for categorical analysis, Output Layer model.add(Dense(units=6,
activation='softmax'))
```

```
model.summary()#summary of our model
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 32)	320
max_pooling2d (MaxPooling2D)	(None, 31, 31, 32)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	9248
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 32)	0
flatten (Flatten)	(None, 6272)	0
dense (Dense)	(None, 512)	3211776
dense_1 (Dense)	(None, 6)	3078
Total params: 3,224,422		
Trainable params: 3,224,422		
Non-trainable params: 0		

## Model Compilation

```
# Compiling the CNN
# categorical_crossentropy for more than 2
model.compile(optimizer='adam', loss='categorical_crossentropy',
metrics=['accuracy'])
```

## Model fitting

```
# It will generate packets of train and test data for training
model.fit_generator(x_train, steps_per_epoch = 594/3 , epochs =
25,
                    validation_data = x_test, validation_steps =
30/3 )
```

```
C:\Users\Midhun\anaconda3\lib\site-packages\tensorflow\python\keras\engine\training.py:194
0: UserWarning: `Model.fit_generator` is deprecated and will be removed in a future versi
on. Please use `Model.fit`, which supports generators.
  warnings.warn("`Model.fit_generator` is deprecated and
Epoch 1/25
198/198 [=====] - 7s 34ms/step - loss: 1.3144 - accuracy: 0.4798
- val_loss: 0.7614 - val_accuracy: 0.7000
Epoch 2/25
198/198 [=====] - 7s 34ms/step - loss: 0.6828 - accuracy: 0.7155
- val_loss: 0.5644 - val_accuracy: 0.8000
Epoch 3/25
198/198 [=====] - 7s 33ms/step - loss: 0.4049 - accuracy: 0.8552
- val_loss: 0.7858 - val_accuracy: 0.7667
Epoch 4/25
198/198 [=====] - 7s 34ms/step - loss: 0.3155 - accuracy: 0.8721 -
val_loss: 0.2433 - val_accuracy: 0.9667
Epoch 5/25
198/198 [=====] - 7s 34ms/step - loss: 0.1929 - accuracy: 0.9327
- val_loss: 0.3210 - val_accuracy: 0.9667
Epoch 6/25
```

```
198/198 [=====] - 7s 34ms/step - loss: 0.1761 - accuracy: 0.9495
- val_loss: 0.5928 - val_accuracy: 0.9000
Epoch 7/25
198/198 [=====] - 7s 34ms/step - loss: 0.1257 - accuracy: 0.9613
- val_loss: 0.3547 - val_accuracy: 0.9333
Epoch 8/25
198/198 [=====] - 7s 34ms/step - loss: 0.0959 - accuracy: 0.9630 -
val_loss: 0.4215 - val_accuracy: 0.9667
Epoch 9/25
198/198 [=====] - 7s 35ms/step - loss: 0.1353 - accuracy: 0.9444
- val_loss: 0.3127 - val_accuracy: 0.9667
Epoch 10/25
198/198 [=====] - 7s 34ms/step - loss: 0.0985 - accuracy: 0.9697
- val_loss: 0.3157 - val_accuracy: 0.9667
Epoch 11/25
198/198 [=====] - 7s 34ms/step - loss: 0.0824 - accuracy: 0.9747
- val_loss: 0.3259 - val_accuracy: 0.9667
Epoch 12/25
198/198 [=====] - 7s 34ms/step - loss: 0.0474 - accuracy: 0.9865 -
val_loss: 0.4769 - val_accuracy: 0.9333
Epoch 13/25
198/198 [=====] - 7s 35ms/step - loss: 0.0570 - accuracy: 0.9848
- val_loss: 0.1794 - val_accuracy: 0.9667
Epoch 14/25
198/198 [=====] - 7s 34ms/step - loss: 0.0704 - accuracy: 0.9714
- val_loss: 0.4142 - val_accuracy: 0.9333
Epoch 15/25
198/198 [=====] - 7s 34ms/step - loss: 0.0749 - accuracy: 0.9697
- val_loss: 0.4670 - val_accuracy: 0.9667
Epoch 16/25
198/198 [=====] - 7s 34ms/step - loss: 0.0529 - accuracy: 0.9832
- val_loss: 0.3779 - val_accuracy: 0.9667
Epoch 17/25
198/198 [=====] - 7s 34ms/step - loss: 0.0412 - accuracy: 0.9899
- val_loss: 0.4824 - val_accuracy: 0.9000
Epoch 18/25
198/198 [=====] - 7s 35ms/step - loss: 0.0394 - accuracy: 0.9798
- val_loss: 0.2046 - val_accuracy: 0.9667
Epoch 19/25
198/198 [=====] - 7s 35ms/step - loss: 0.0870 - accuracy: 0.9747
- val_loss: 0.3575 - val_accuracy: 0.9667
Epoch 20/25
198/198 [=====] - 7s 34ms/step - loss: 0.0308 - accuracy: 0.9899 -
val_loss: 0.3367 - val_accuracy: 0.9667
Epoch 21/25
198/198 [=====] - 7s 35ms/step - loss: 0.0381 - accuracy: 0.9865
- val_loss: 0.2932 - val_accuracy: 0.9333 Epoch 22/25
198/198 [=====] - 7s 35ms/step - loss: 0.0422 - accuracy: 0.9865
- val_loss: 0.5818 - val_accuracy: 0.8667
Epoch 23/25
198/198 [=====] - 7s 34ms/step - loss: 0.0319 - accuracy: 0.9865
- val_loss: 0.3459 - val_accuracy: 0.9667
Epoch 24/25
198/198 [=====] - 7s 35ms/step - loss: 0.0385 - accuracy: 0.9815 -
val_loss: 0.3301 - val_accuracy: 0.9667
Epoch 25/25
198/198 [=====] - ETA: 0s - loss: 0.0138 - accuracy: 0.99 - 7s 3
<tensorflow.python.keras.callbacks.History at 0x1b89d20da60>
```

## Saving model

```
# Save the model model.save('gesture.h5')
```

```
model_json = model.to_json() with open("model-  
bw.json", "w") as json_file:  
    json_file.write(model_json)
```