TRAFFIC SIGNS DETECTOR



Group 02

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E/16/022 E/16/025 E/16/222

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PROBLEM STATEMENT AND MOTIVATION

- Traffic signs convey, guide, restrict, warn, or instruct information
- Identification important to ensures road safety
- Insufficient illumination, partial occlusion and serious deformation are the factors that cause.
- Traffic signal detection is a challenging problem.
 - driving systems
 - automatic driving systems

SOLUTION

- Train and classify Traffic Signs using Convolutional neural networks.
- This will be done using TensorFlow, Keras and OPENCV in real time using a simple webcam.
- Vision based solution
 - Can use smartphone cameras or webcams mounted on vehicles to detect traffic signals
 - Minimum cost of technical devices is an advantage

Dataset used

- German Traffic Sign Recognition Benchmark (GTSRB)
 public archive with ID daaeac0d7ce1152aea9b61d9f1e19370
- 43 classes of signals ~ 34,000 images



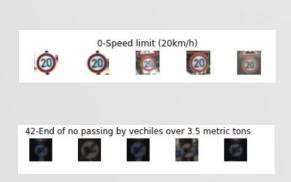
DATA **PREPROCESSING**

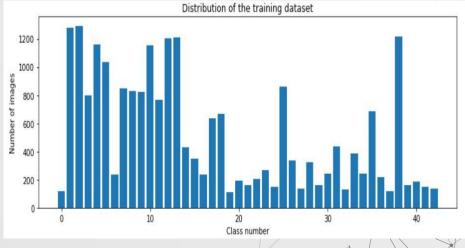
PROCEDURE

- Importing of images
- Splitting data into train, test and validation sets

```
testRatio = 0.2  # if 1000 images split will 200 for testing validationRatio = 0.2 # if 1000 images 20% of remaining 800 will be 160 for validation
```

Dataset consists of 43 classes





Sample images of classes

Distribution of training dataset

PROCEDURE

- Data preprocessing
 - Convert to grayscale
 - Standardize the lighting in an image
 - Normalize values between 0 and 1 instead of 0 and 255







After Preprocessing

- Image augmentation
 - To make images more generic
 - Shifting width and height
 - Magnitude of the shear angle and zoom level























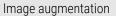




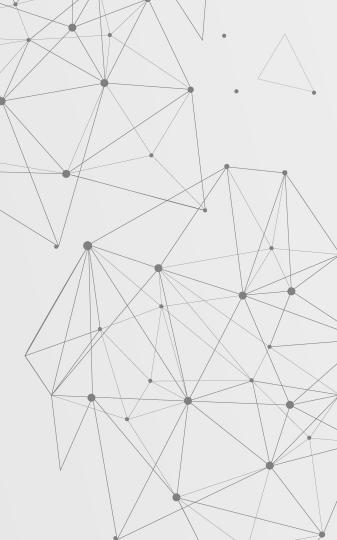








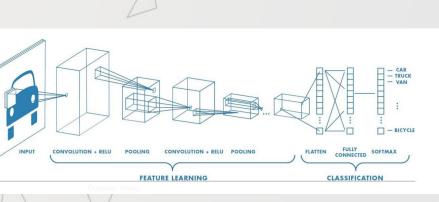
MODEL DESIGNING AND IMPLEMENTATION



PROCEDURE

- Developing convolution neural network model
 - Defining number of filters and size of a filter
 - Kernel size 5x5
 - Remove 2 pixels from each border (when 32x32 images used)
 - More convolution layers added
 - For increase accuracy
 - o In Pooling layers Scale down the feature map to generalize the solution
 - Dropout regularization
 - Better generalization
 - Reduce overfitting
 - 500 ,43 nodes in hidden layers in Dense Layer

Model Summary



Model:	"sequential"
wode1:	sequential

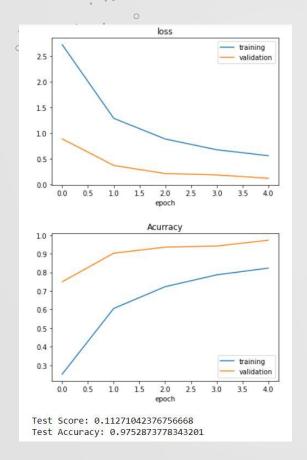
Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 28, 28, 60)	1560
conv2d_1 (Conv2D)	(None, 24, 24, 60)	90060
max_pooling2d (MaxPooling2D)	(None, 12, 12, 60)	0
conv2d_2 (Conv2D)	(None, 10, 10, 30)	16230
conv2d_3 (Conv2D)	(None, 8, 8, 30)	8130
max_pooling2d_1 (MaxPooling2	2 (None, 4, 4, 30)	0
dropout (Dropout)	(None, 4, 4, 30)	0
flatten (Flatten)	(None, 480)	0
dense (Dense)	(None, 500)	240500
dropout_1 (Dropout)	(None, 500)	0
dense_1 (Dense)	(None, 43)	21543

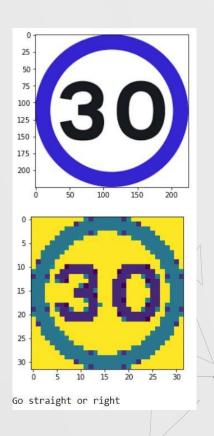
Total params: 378,023 Trainable params: 378,023 Non-trainable params: 0

```
Epoch 1/10
446/446 [=========== ] - 74s 165ms/step - loss: 2.6546 - accuracy: 0.2611 - val loss: 1.0425 - val accuracy:
0.6649
Epoch 2/10
446/446 [=========== ] - 74s 165ms/step - loss: 1.3224 - accuracy: 0.5922 - val loss: 0.3956 - val accuracy:
0.8664
Epoch 3/10
0.9339
Epoch 4/10
446/446 [========== ] - 69s 154ms/step - loss: 0.6974 - accuracy: 0.7817 - val loss: 0.1787 - val accuracy:
0.9456
Epoch 5/10
446/446 [============ ] - 66s 147ms/step - loss: 0.5742 - accuracy: 0.8184 - val loss: 0.1277 - val accuracy:
0.9652
Epoch 6/10
446/446 [=========== ] - 67s 150ms/step - loss: 0.4951 - accuracy: 0.8413 - val loss: 0.0863 - val accuracy:
0.9777
Epoch 7/10
446/446 [=========== ] - 65s 145ms/step - loss: 0.4433 - accuracy: 0.8600 - val loss: 0.0854 - val accuracy:
0.9772
Epoch 8/10
446/446 [=========== ] - 70s 158ms/step - loss: 0.4038 - accuracy: 0.8727 - val loss: 0.0680 - val accuracy:
0.9826
Epoch 9/10
0.9813
Epoch 10/10
446/446 [=========== ] - 70s 157ms/step - loss: 0.3520 - accuracy: 0.8890 - val loss: 0.0488 - val accuracy:
0.9878
```

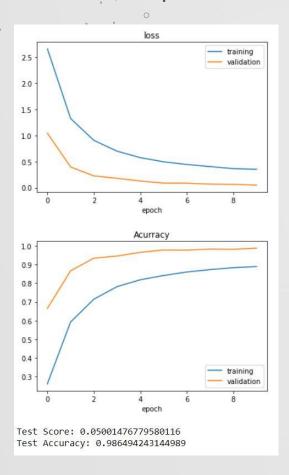
MODEL TESTING AND EVALUATION

• For 5 epochs

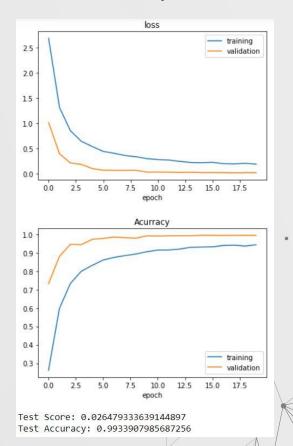




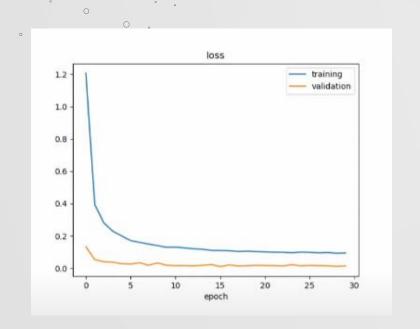
• For 10 epochs •

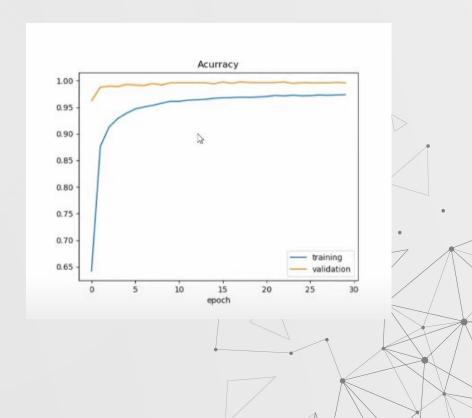


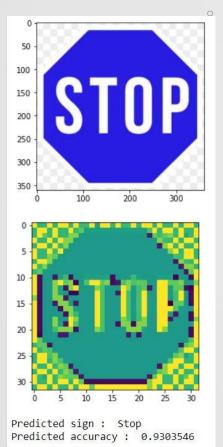
For 20 epochs

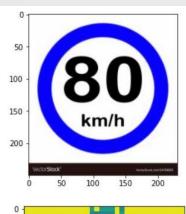


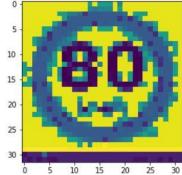
• For 30 epochs











Speed limit (80km/h)
Predicted sign : Speed limit (80km/h)
Predicted accuracy : 0.9247637



