Traffic Sign Classifier

Dataset Exploration

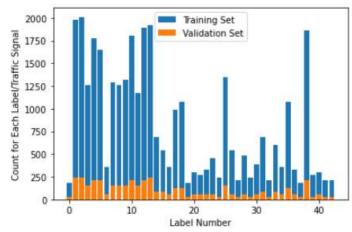
Dataset Summary

Number of training examples: 34799
Number of validation examples: 4410
Number of testing examples: 12630

Input image shape: (32, 32, 3) Number of output classes: 43

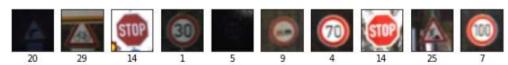
Exploration Visualization

The below histogram shows the number of examples for each output label both from the testing set and the validation set to visualize the relative difference in each set:



The below picture shows a random set of 10 images along with the assigned label number from the training set to visualize the differences in images:

[25972, 10781, 29727, 2395, 13477, 11815, 7250, 29866, 34296, 24137]



Design and Test Model Architecture

Preprocessing

Preprocessing is a very important step in developing a convolutional neural network as we want the inputs to be on the same scale

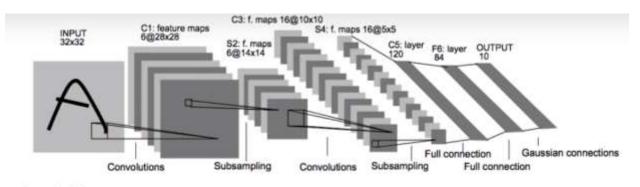
Step 1: Mean for all the channels was calculated to ensure that light intensity did not play major role in the learning process and in general grayscale images provide better results

Step 2: The images were normalized to ensure that the data has a mean of zero and equal variance and hence ensure the images are on a similar scale

Step 3: The images were shuffled to ensure that the model did not learn the order of images

Model Architecture

The starting point for the model was the LeNet5 and the model immediately gave an accuracy of 89% on the validation test. To increase the accuracy a dropout layer was added after the second fully connected layer and initially a keep probability was kept at 0.5. this layer helped in reducing the chances of overfitting. The picture below is from the LeNet5 developed by Yan LeCun. With respect to this, the model for this training has another layer after the second fully connected layer which uses dropout to reduce overfitting. Also, the number of output labels in our model is 43 instead of 10.



Source: Yan LeCun

The table below give the input and output sizes for all the layers, along with the activation function and type of padding used

S.No.	Layer	Input Size	Output Size
1	5x5 Convolutional Layer	Image 32x32x1	28x28x6
	Stride: [1, 1]		
	Padding: Valid		
2	Activation Function:	28x28x6	28x28x6
	RELU		
3	Max Pooling	28x28x6	14x14x6
	Kernel size: [2, 2]		
	Stride = [2, 2]		
4	5x5 Convolutional Layer	14x14x6	10x10x16
	Stride: [1, 1]		
	Padding: Valid		
5	Activation Function:	10x10x16	10x10x16
	RELU		
6	Max Pooling	10x10x16	5x5x16
	Kernel size: [2, 2]		
	Stride: [2, 2]		
7	Flatten Layer	5x5x16	400
8	Fully Connected	400	120
9	Activation Function:	120	120
	RELU		
10	Fully Connected	120	84
11	Activation Function:	84	84
	RELU		
12	Dropout Layer	84	84
13	Fully Connected Output Layer	84	43

Model Training

Adams Optimizer was used it is one of the most computationally efficient optimizers and has little memory requirements. It is a stochastic gradient descent optimizer Batch size was kept at 128. Epochs were increased to 100 initially to see if the model was overfitting and slowly reduced to 50 as the model was converging and achieving an accuracy of more than 93%

Other hyperparameters are mentioned in the table below:

Hyperparameter	Variable Name	Value			
Learning Rate	rate	0.001			
Epochs	epochs	50			
Batch size	batch	128			
Keep probability	keep_prob	0.7			

Mean of weights	mu	0
Sigma of weights	sigma	0.1

Solution Approach

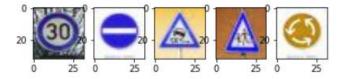
The above hyperparameters were drastically increased and decreased to understand how they impacted the result while keeping the learning rate low and keeping the epochs close to 100. Once the architecture and the other hyperparameters gave satisfactory results, the learning rate was slowly increased, and the epochs were reduced

The model gave validation accuracy of 94.4% and test set accuracy of 91.9%

Test a Model on New Images

Acquiring New Images

The following images were downloaded from the internet and converted to 32x32 using MS Paint. As these images were converted to smaller pixels, some of the details maybe lost and hence it could be difficult for the model to classify them correctly



Performance on New Images

The model achieved an accuracy of 0.8 on the downloaded images as compared to an accuracy of 0.944 on the test set provided. But this accuracy could vary if the images were not clear enough

Model Certainties – Softmax Probabilities

```
INFO:tensorflow:Restoring parameters from ./lenet
TopKV2(values=array([[9.99999881e-01, 1.02738014e-07, 3.99582700e-10, 5.2660955
5e-11,
        2.55827443e-11],
       [9.99999881e-01, 1.16982790e-07, 2.37826148e-10, 4.25931568e-14,
        2.40105154e-17],
       [9.40245688e-01, 2.95954701e-02, 2.87118424e-02, 9.84704238e-04,
       4.03189857e-04],
       [1.00000000e+00, 9.56782875e-10, 1.27988592e-11, 8.57621092e-12,
       3.15923405e-15],
       [9.99999642e-01, 3.70805225e-07, 2.22822607e-12, 8.57511897e-14,
        2.80224909e-14]], dtype=float32), indices=array([[ 1, 38, 2, 3, 5],
       [17, 9, 34, 12, 40],
       [23, 28, 19, 29, 20],
       [28, 24, 20, 29, 30],
       [40, 9, 12, 41, 17]]))
```